## QATAR UNIVERSIY

## COLLEGE OF ARTS AND SCIENCES

## ELEMENTAL COMPOSITION, SOURCE TRACKING, AND AIR QUALITY ASSESSMENT

## OF PM2.5 AND PM10 POLLUTION IN QATAR

BY

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of the Requirements

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## ABSTRACT

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Title: Elemental Composition, Source Tracking, and Air Quality Assessment of PM2.5 and PM10 Pollution in Qatar

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Particulate matter (PM) pollution is one of the major environmental pollution issues severely affecting human health and air quality all over the world. Based on the recent World Health Organization (WHO) report, PM levels were considered relatively high in Qatar. This might mainly be attributed to arid climate, but also due to rapid industrialization and urbanization as well as traffic. The literature on PM pollution and its source is limited in Qatar and the region. Therefore, this study was carried out to assess the air quality based on PM2.5 and PM10 levels at different locations in Qatar, determine the elemental composition of PM2.5 and PM10 to trace their sources, and create a map by using Geographical Information System (GIS) to show the air quality based on PM levels in select locations in Qatar.

A total of 100 samples (60 for PM2.5 and 40 for PM10) were collected using SKC Deployable Particulate Sampler (DPS) System for 24-hr during the months of September to December, 2016. The sampling was conducted at five different locations, namely, Qatar University (QU), Education City (EC), Aspire Zone (AZ), Whole Sale Market area (WM), and Al-Wakrah City (AW). The elemental composition of PM samples was determined using an inductively coupled plasma optical emission spectrometry (ICP-OES). The

relationship between the environmental conditions and PM levels were also established.

The health risks associated with different PM levels were calculated using the US EPA Air Quality Index (AQI) tool. The AQI values calculated based on the daily concentrations of PM2.5 and PM10 at each sampling location were computed on maps using GIS modeling system in combination with Google Earth.

The overall mean concentrations of 24-hr PM2.5 ranged from 50  $\mu$ g/m<sup>3</sup> to 64  $\mu$ g/m<sup>3</sup>, while PM10 levels were between 127  $\mu$ g/m<sup>3</sup> and 185  $\mu$ g/m<sup>3</sup>. The four months mean concentrations of PM2.5 were determined to be 50, 64, 55, 59, and 57  $\mu$ g/m<sup>3</sup> at QU, EC, AZ, WSM, AW, respectively. The average 24-hr PM10 levels were 138  $\mu$ g/m<sup>3</sup> at QU, 156  $\mu$ g/m<sup>3</sup> at EC, 127  $\mu$ g/m<sup>3</sup> at AZ, 185 $\mu$ g/m<sup>3</sup> at WM, and 160  $\mu$ g/m<sup>3</sup> at AW. The concentrations of PM2.5 detected at each station exceeded the WHO guideline (20  $\mu$ g/m<sup>3</sup>) by 2.5 to 3 fold during the study period.

The presence of high concentrations of Ca, Fe, Al, Fe, Sr, Mn, Na, and Mg indicated the major sources of PM to be soil/crustal. The identification of Ni, Co, Cr, Cd, Ba, Pb, V, and Zn were directly related to anthropogenic sources, specifically due to fossil fuel combustion and vehicular emission and these levels were reported at the highest levels at the wholesale market station. The AQI levels determined at all stations indicated that overall air quality at Qatar University and Aspire Zone area was considered to be "Moderate" for PM10 and "Unhealthy for sensitive group" for PM2.5 levels. While Education City, Whole sale Market, and Al-Wakrah city areas had "unhealthy" and "unhealthy for sensitive group" ratings for PM2.5 and PM10 levels, respectively.

The statistical analysis on determining the effect of sampling date and locations on the concentration of PM2.5 and PM10 showed that there is a significant relationship (p<0.01) between PM levels, sampling stations, and sampling date.

These findings highlight the need for more research on PM pollution 1) to determine seasonal levels since this study only covered four months (September-December), 2) to better understand the source of PM pollution (in addition to elements, the levels of Poly Aromatic Hydrocarbons should also be determined), and 3) to establish more effective control measures to protect public health and preserve the environment in Qatar.

# **DEDICATION**

I would like to dedicate my work to the residents of Qatar who need to know more about air

quality.

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# **Table of Contents**

DEDICATION vi
ACKNOWLEDGMENTS vii
List of Figures xi
List of Tables xii
CHAPTER I: INTRODUCTION 1
CHAPTER II: LITERATURE REVIEW 4
1.1. Particulate Matter 5
1.2. Size Distribution of Airborne Particulate Matter
1.3. PM Pollution Sources in Qatar10
1.4 Anthropogenic Sources
1.4.1 Industrial Activities
1.4.2 Traffic
1.6 Meteorology in Qatar 15
1.7 Chemical Composition of PM17
1.7.1 Elemental Composition of PM18
1.8 Health Effects of PMs 21
2.1 Justification
2.2 Objectives of the Study
CHAPTER III: MATERIALS AND METHODS

3.1 Study Site	
3.2 Sample Collection and on-site measurements of PM2.5 and PM10	
3.4 Calculation of PM Mass	
3.5 Identification of the elemental composition of PMs	32
3.6 Determination of PM sources using Enrichment factor (EF)	33
3.7 Determination of Air Quality	
3.8 Mapping of AQI values calculated based on PM concentrations and Land u	ıse 35
3.9 Statistical Analysis	
CHAPTER IV: RESULTS AND DISCUSSION	
4.1 PM Concentrations by Location and Month	
4.2 Mapping of AQI Values Based on PM2.5 and PM10 Concentrations	49
4.3 Particulate Matter Elemental Composition	52
4.4 Possible sources of PMs based on Enrichment Factor Analysis	62
CHAPTER V: CONCLUSION	65
REFERENCES	67
APPENDICES	78
Appendix A : Meteorological Data	78
Appendix B: PM Calculations	100
Appendix C: Statistical Analysis	101

# List of Figures

Figure 1. Size distribution of PM in ambient air (WHO, 2006) 10
Figure 2. TSP source emission by their size distribution (Vega et al., 2001)
Figure 3. Map of sampling locations
Figure 4. Deployable Particle Sampler
Figure 5. Exploded view of IMPACT Sampler
Figure 6. The comparison of the USEPA standards and the average mean concentrations
of PM2.5 and PM10 recorded at different locations during the study period
Figure 7. The percentage distribution of different Air Quality Index ratings during the
sampling months
Figure 8. Land use Illustration at Aspire Zone
Figure 9. Land Use Illustration at Qatar University
Figure 10. Land Use Illustration at Whole Sale Market
Figure 11. Land Use Illustration at Education City 48
Figure 12. PM2.5 based Air Quality Index Values at the sampling stations
Figure 13. PM10 based Air Quality Index Values at the sampling stations
Figure 14. Illustration of Land use at of Qatar University
Figure 15. Illustration of Land use at Education City
Figure 16. Average concentrations of heavy metals in PM2.5 collected from different
sampling stations
Figure 17. Average concentrations of heavy metals in PM10 samples collected from
different sampling stations

# List of Tables

Table 1 Ambient air pollution limits in Qatari Environment Protection Law 30 (2002)
and World Health Organization (WHO, 2005)7
Table 2 World Health Organization's report on PM levels in Qatar (WHO, 2016)
Table 3 Particulate matter (PM) standards established by USEPA, WHO, Qatar, and
several Middle East Countries (Tsiouri et al., 2015) 12
Table 4 Maximum and minimum temperatures measured at Hamad Airport during 2012
(MDPS, 2013)
Table 5 Maximum and minimum relative humidity (%) measured at Hamad Airport
during 2012 (MDPS, 2013) 17
Table 6 Elemental composition of PM10 samples collected in Iraq and Kuwait
(Naimabadi et al., 2016) 19
Table 7 Outdoor elemental composition of PM2.5 and PM10 (Saraga et al., 2017) 21
Table 8 Flow rate measurements of each PM2.5 and PM10 sampling device at the time of
calibration
Table 9 Tuning parameters of the ICP-OES33Table 10 Air Quality Index categories by Level of Health concern and Colors (AirNow,
2016)
Table 11 The mean concentrations of PM2.5 and PM10 samples collected from different
Stations and Meteorological data
Table 12 The Air Quality Index Values calculated based on the concentrations* of PM2.5
and PM10 during the study period
Table 13 Air Quality Index Values calculated based on the four month averages of PM2.5

and PM10 at different stations	
Table 14 Elemental composition concentration of PM2.5 samples collected	d from different
locations	55
Table 15 Elemental composition concentration of PM10 samples collected	from different
locations	56
Table 16 Enrichment Factor (EF) Values for elements determined in PM2.	5 and PM10
samples (µg/m <sup>3</sup> )	64

## **CHAPTER I: INTRODUCTION**

In the last few years, the State of Qatar has been going through many changes and developments in the economy, urban environments, and construction. The population of Qatar is rapidly growing, reaching 2,668,415 according to the latest statistics. This figure is expected to increase to three million by 2026 (MDPS, 2017). The fast population growth caused many environmental changes. Qatar's National Vision 2030 was developed to address these environmental changes by establishing a balance between economic growth, social development, and environmental protection. The vision emphasizes sustaining the environment for the future generations by balancing between developmental needs and the protection of the natural environment, land, sea, and air (Sillitoe, 2014).

As a result of this initiative, air quality research field has received much needed attention from researchers in Qatar as well as around the world. There is now a substantial body of epidemiological evidence that establishes a link between exposure to air pollution and increased mortality (especially premature death) and morbidity due to a wide range of adverse cardiovascular and respiratory problems (Lee et al., 2014). There are various pollutants in the air, such as nitrogen oxides (NOx), sulfur oxides (SOx), ozone (O<sub>3</sub>), particulate matter (PM), CO, CO<sub>2</sub>, hazardous air pollutants (e.g., aldehydes, PAHs, etc.).

Particulate matter are fine particles that are suspended in the air and originate from different sources (Laden et al., 2000), including natural and anthropogenic sources (EPA, 2015). Natural sources include wind-blown desert dust, sea spray aerosols, volcanoes, seismic activity, and wild fires (EPA, 2015; Putaud et al., 2004). Examples of anthropogenic sources include vehicle emissions from fuel combustion, domestic heating, incineration, construction and emissions from thermal power generation (EEA, 2015; Hassan et al., 2016). The major size distribution of PM is between 2.5 and 10  $\mu$ m (PM2.5 and PM10) (Khan et al., 2010). PM2.5 sources are mainly dust/soil, oil combustion, petrochemical industries, and traffic emissions (Brow et al., 2013). In contrast, PM10 mainly comes from natural sources and transportation (Lenschow et al., 2001).

Many studies have investigated the effects of different sizes of PM and their impact on human health (Davidson et al., 2005). Several health problems result from PM exposure, like respiratory and cardiovascular morbidity, asthma and other respiratory symptoms, and mortality related to lung cancer (Brook et al., 2010; Chen et al., 2016). The sources of PM can be identified based on the chemical and physical properties of these pollutants and their reactions with other chemicals suspended in the air (Ye et al., 2017). The health effects depend on the exposure time and doses (Shaughnessy et al., 2015). The effects also depend on the concentration of PM (PM2.5, PM10; unit mass/m<sup>3</sup>), the PM's complex compositions of trace metals and other elements. Therefore, it is important to identify the probable source of PM promptly.

The Middle East is considered one of the most polluted areas in terms of PM pollution (Elbayoumi et al., 2013). Rapid urbanization and construction in this region have created concerns about rising health problems related to PM pollution (Tsiouri et al., 2015). Factors that affect the dispersion and concentration of PM include temperature, humidity, the height of the mixing layer, and pressure (Khan et al., 2010; Marcazzan et al., 2001). Qatar is located in an arid region with desert features similar to other Middle

Eastern countries. The World Health Organization (WHO) has reported high levels of PM2.5 and PM10 in this area (WHO, 2016).

Since Qatar is a fast developing nation, air pollution issues related to PM pollution need urgent attention. Therefore, this study was carried out to monitor the concentrations of PM (PM2.5 and PM10) at different locations in Qatar by integrating emissions within a framework based on a geographic information system (GIS) using different modeling techniques. The study also aimed at investigating the air quality based on PM concentrations and elemental compositions of PM to identify the main source at these sampling sites. The results obtained in this study could be helpful to protect public health and the environment in Qatar.

## **CHAPTER II: LITERATURE REVIEW**

Air pollution problem has become an important public health issue as a result of massive population increases and industrial development (Loupa et al., 2016). Numerous air pollution studies offer indications relating diverse effects and diseases to the toxic substances present in air pollutants (Vallero, 2014). Air pollution in urban areas considered as significant problem, especially in developing countries (Mage et al., 1996). The World Health Organization (WHO) conducted a study on air pollution in 1958 which showed that scientists need to focus deeply on air pollution and its effects on humans, organisms, and earth systems (WHO, 2016). Specific studies carried out during the twentieth century have focused on acid rain in European counties and the United States (Patel et al., 1974; Schindler, 1988). Acid rain is a common type of pollution in many developed countries and can affect people' lives. However, acid rain is not a global issue, but there are other global air pollution problems that are occurring in tropical and desert countries. In the last part of the twentieth century, scientists started to understand the composition of air pollution, which opened up more opportunities to investigate air pollution in a wider view and to investigate more details by combining sources with other factors (Vallero, 2014).

Air pollution is caused by a mixture of complex components of solids and liquids that vary in size, composition, and origin (natural and anthropogenic) (Dockery et al., 1993; Samet et al., 2002; Brook, et al., 2004). The anthropogenic sources are made by people through automobiles, industry, construction, etc. In contrast, natural sources are mainly from dust, volcanoes, and forest fires, etc. (Kampa & Castanas, 2008). Current scientific evidence proves the fact that outdoor air pollution causes a variety of diseases, such as respiratory illnesses, cardiovascular illnesses, and death. One of the major air pollutants considered to be the most detrimental to human health is particulate matter (PM).

#### **1.1. Particulate Matter**

Particulate matter (PM) are a mixture of solid and liquid particles that are organic and inorganic chemicals (Jang et al., 1997; Laden et al., 2000). Anthropogenic sources that are most dominant in urban areas include industrial fuel combustion, domestic heating in houses, fuel burned by vehicles, road wear, and other sources (EEA, 2015). The elemental composition of PM depends on the source. For example, the elements commonly found in PM generated by power plants are nickel, zinc, sulfate, and mineral aerosol (sodium, magnesium, chloride). Sources including vehicles emit elements such as trace elements, and nitrate, elemental/organic carbon (Rodríguez et al., 2004). Toxic heavy metals like Cd, Pb, Cr, Zn, Ni, and As are usually emitted by the metal industry, and Al, Si, K, Ti, and Fe are distributed by coal combustion. Al and Fe are the main crustal elements that used to be compared with other elements to identify possibility of anthropogenic sources. PM with significant fractions of Si, Cl, and Fe mainly come from burning biomass (Rodríguez et al., 2004). Elements like Na, Cl, and Mg are mainly form sea spray and sea salts (Viana et al., 2008). There are six criteria air pollutants listed by the United States Environmental Protection Agency (USEPA) under the National Ambient Air Quality Standards (NAAQS) for Clean Air Requirement (USEPA, 2016). These air pollutants are ground-level ozone, PM, CO, lead, sulfur dioxide, and nitrogen dioxide. These pollutants were chosen based on their human health effects and the environmental damage they cause. The Ministry of Environment and Municipality in Qatar uses the same standards as listed by the USEPA with different pollution levels near the surrounding countries' levels (Tsiouri et al., 2015). The concentration limits of PM10 for 24 hours is 150  $\mu$ g/m<sup>3</sup> and 50  $\mu$ g/m<sup>3</sup> a year, which stays within the limit of other Middle East countries. For PM2.5 limits, the criteria set by the WHO is 50  $\mu$ g/m<sup>3</sup> for 24 hours and 20  $\mu$ g/m<sup>3</sup> annually, but in Qatar, there is no standard limit for PM2.5 (Table 1).

## Table 1

Ambient air pollution limits in Qatari Environment Protection Law 30 (2002) and World Health Organization (WHO, 2005)

		Concentration averaged over								
Dollutont	Unit	1 ho	our	8 ho	ours	24 h	ours	1 y	ear	
ronutant	Unit	Qatar WHO Qatar WHO Qatar WHO Qatar WHO								
Nitrogen dioxide		400	200			150		100		
(NO2)	µg/m <sup>3</sup>	400	200			150		100	40	
Particulate Matter	, 3					150	50	50	20	
<10 µm	µg/m <sup>3</sup>					150	50	50	20	
Particulate Matter	, 3									
<2.5 μm	µg/m³						25		10	
Carbon Monoxide	2									
(CO)	mg/m <sup>3</sup>	40	30	10	10					
Ground Level	. 2									
Ozone (O3)	µg/m³	235		120	100					
Sulfur dioxide	2									
(SO <sub>2</sub> )	µg/m³					365	20	80		

The most recent report on air pollution published by WHO (WHO, 2016) mentioned Qatar as one of the most polluted countries in terms of PM10 and PM2.5 levels (Table 2). The reported values were based on the official reported results, satellites data and modelling.

#### Table 2

Country	PM Urban	12.5 [μg/m <sup>2</sup> and rural	<sup>3</sup> ], areas	PM2.5 [μg/m <sup>3</sup> ], Urban areas		
Ootor	Median	Lower	Upper	Median	Lower	Upper
Qatai	103	67	160	105	69	159

World Health Organization's report on PM levels in Qatar (WHO, 2016)

#### 1.2. Size Distribution of Airborne Particulate Matter

Particulate matter are categorized into three sizes relying on their aerodynamic diameter (10, 2.5, and 1  $\mu$ m). PM10 which is also called inhalable coarse particles (<2.5-<10 $\mu$ m), can be found near roadways and dusty industries and have short lifetime in atmosphere compared to PM2.5. PM10 stays in the air for minutes to days and travel distance of less than one to hundreds Km (Joint & World Health, 2006). Common sources of PM10 are dust resuspension, mining, sea spray, construction, and demolition. PM2.5, which became the most studied pollution in the last ten years, was added to the air quality criteria because of its detrimental health effects on the respiratory system. It is called fine particles, with diameter less than 2.5 micrometer (<2.5). PM2.5 can travel long distances of up to 10<sup>6</sup> kilometers with ling lifetimes reaching days and weeks, and the main source

of PM2.5 is bringing fossil fuel (organic biomass) and combustion process (Joint & World Health, 2006). It can be found in haze or smoke and is emitted directly from the source. Its size lets PM2.5 be inhaled deeper in the lings.

Particulate matter that less than1 µm in diameter are called PM1. It can be travel beyond the lungs disrupts systemic vascular function and causing significant health effects (Rundell et al., 2007). Its atmospheric half-life is minuets to hours and it can travel less than tens of kilometers (Joint & World Health, 2006). Composition and mass of particulate matter could be divided to main two categories which are fine and coarse particles. Particulate matter that are fine and coarse fractions are demonstrated in Figure 1.

Understanding the land use around the sampling station considered important in air pollution studies. Geographical information system software support researchers to build graphical display of geographical information to be presented on maps. That visual tools are used to visualize data in term of numbers to build more clear decision support to the reality on geography map.



Figure 1. Size distribution of PM in ambient air (WHO, 2006).

#### **1.3. PM Pollution Sources in Qatar**

Increased urbanization, industrialization, and construction activities in Qatar have resulted in problems with air pollution (Tsiouri et al., 2015). Additionally, natural conditions such as dust storms often occur in the Arabian Gulf region are also contributing factors that directly impact air quality (Alam et al., 2014). Furthermore, the petroleum industry is the backbone of Arabian Gulf countries. The Qatar General Petroleum Corporation (QGPC) was established in 1974 (Bergendahl, 1985), and ranks as one of the largest petroleum and gas companies in the world. The recent industrial development in this region has led health organizations to establish a link between air pollution issues and petroleum industry activities (Chen et al., 2016). The total emissions of PM are mostly from natural sources, like ocean sprays, suspended dust from terrestrial areas, and burning fossil fuel. In a recent study by Hassan et al. (2016), it was found that the main source of particles size ranged between 0.25-32  $\mu$ m were from wind erosion of the loose soil, where the study took place near to construction area. The limits of PM in Qatar and the Middle East were reviewed by Tsiouri et al. (2015) who reported that most of the countries in this region have higher PM limits than non-dusty countries due to the concentration levels being naturally high in this region (Table 3).

## Table 3

Particulate matter (PM) standards established by USEPA, WHO, Qatar, and several Middle East Countries (Tsiouri et al., 2015)

Standards	PM type and averaging time							
	PM1	$10 (\mu g/m^3)$	PM2.5 (	ug/m <sup>3</sup> )				
-	24-h	Annual	24-h	Annual				
USEPA (EPA, 2010)	150	50	35	15				
EU (EC, 2010)	50	40	-	25				
WHO (WHO, 2006; WHO,	50	20	25	10				
2011)								
Jordan (Al-Zubi, 2011)	120	70	-	-				
Kuwait (IES, 2011)	150	90	35	15				
Lebanon (LEDO, 2001)	80	-	-	-				
Oman (HMR, 2010)	150	-	35	15				
Qatar (Abdel-Moati, 2008)	150 (1 h)	50 (3 month)	-	-				
Saudi Arabia (PME, 2012)	340	80	35	15				
Syria (ELARD, 2009)	100	-	-	-				
UAE (EAD, 2014)	150	-	-	-				

#### **1.4 Anthropogenic Sources**

#### **1.4.1 Industrial Activities**

Qatar has an area of 11,572.07 Km<sup>2</sup> and industrial cities are spread out in various locations for petroleum and chemical production. PM2.5 mainly originates from secondary pollutants (which react in the air), followed by primary pollutants (trace metals that could come from industry and transportation systems) which are mainly the sources of PM10 (Heal et al., 2012). Industrial and electrical power plants are recognized as the main stationary air pollution sources in urban cities. These sources are considered as point sources where the type of emission can be easily identified. In Qatar, there are two main power plants for electrical production with total production reaching 8000MW (QEWC, 2011). Elements like Ni, Cr, Cu, Sn, Zn, Mo, Sb, Pb, and Cd in PM could give indications industrial activities -including power plants- (Das et al., 2015).

Highlighting industrial sources depends on the results of sampling filters collected from sites. Based on official sources, there is no data available on PM sources from industrial activities in Qatar. Since high levels of PM2.5 come from anthropogenic sources such as industry and traffic, several stationary samplers are needed to have continuous readings for longer period near the sources and living areas.

#### 1.4.2 Traffic

Qatar is considered as one of the richest countries in the world when it comes to capital income. Transportation systems in Qatar are mainly based on private cars, and using public transportation is not common in this country. There is no metro system yet, and public bus transportation is not convenient for middle income people since using private cars is faster. Furthermore, trucks are the main method for transporting goods due to the absence of freight trains. Particles generated by vehicle activities come from several processes, such as the combustion of fossil fuel, resuspension of road/soil dust, tire friction, and brake linings (Laschober et al., 2004). Fine particulate matter is not only from natural sources, it is also come from vehicles (Hassan et al., 2016). PM emitted from or related to traffic varies during periods of high traffic density with poor air movement in urban areas. For example, human exposure to traffic pollutants is higher in street canyons (Vardoulakis et al., 2003) and can be detected if there is a continuous air monitoring during a 24 hour period. There are several elements associated with vehicle activities, such as Fe, Br, Cu, Zn, Ba, and Pb (Huang et al., 1994).

#### **1.4.3 Construction**

Qatar is considered one of the fastest developing countries with heavy construction activities supported by the strong economy. Since winning the World Cup 2020 bid, construction activities have been sped to build stadiums and infrastructures. A recent study conducted near construction sites in Qatar reported the increased concentrations of PM pollution mainly due to the presence of Calcisols (that resulted from the accumulation of secondary carbonate coming from the construction site) (Hassan et al., 2016). As Qatar characterized by its dry and arid environment (Gopalaswami et al., 2015), erosion of the soil and its suspension in the air column during construction activities will increase the particulate matter pollution.

#### **1.5 Natural Sources**

In Qatar, dust storms normally occur within specific months of the year. Qatar is a desert country where no fires could increase the pollutants significantly like forest and tropical forest regions. As a result, the elemental composition of PMs will be consisted of the minerals dust like Ca, Fe, Sr, Si, K, and Ti (Weckwerth, 2001) which are mostly are available in Qatar and outer regional countries.

#### **1.6 Meteorology in Qatar**

Qatar is one of the Gulf Cooperation Council (GCC) countries which is characterized by a desert biome. Qatar has characteristic of dry and arid region conditions (Gopalaswami et al., 2015). This leads to hot temperatures in summer and slightly colder temperatures in winter. However, the peninsula shape results in high relative humidity compared to other GCC countries. The most recent data published by the Ministry of Development Planning and Statistics (MDPS, 2013) on the temperature and relative humanity at the Hamad Airport and other stations is presented in (Table 4). According to the report, the maximum average daily temperature in summer reaches 43.3°C and at least 22.5°C in winter. In summer, the average maximum humidity is 66 to 88%, and minimum is 29 to 60%.

# Table 4

Maximum and minimum temperatures measured at Hamad Airport during 2012 (MDPS, 2013)

Month	Extr Tempe	eme erature		Number of days with				
(2012)	Absolute Max (°C)	Absolute Min (°C)	Max. Temp. and Min. Temp. ( $^{\circ}C$ )					
			>=25	>=30	>=35	>=40	>=45	<=10
January	28.0	8.6	5	0	0	0	0	2
February	29.8	11.5	4	0	0	0	0	0
March	37.8	12.8	19	5	2	0	0	0
April	39.8	19.5	30	25	7	0	0	0
May	46.8	27.2	31	31	31	4	1	0
June	47.7	27.6	30	31	30	28	6	0
July	47.6	30.0	31	30	31	31	8	0
August	46.1	29.6	31	31	31	23	2	0
September	43.5	25.5	30	30	30	15	0	0
October	40.5	25.8	31	31	19	2	0	0
November	35.0	18.4	8	17	1	0	0	0
December	31.3	14.3	16	1	0	0	0	0
Annual	47.7	8.6	266	232	182	103	17	2

## Table 5

Maximum and minimum relative humidity (%) measured at Hamad Airport during 2012 (MDPS, 2013)

	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
Min	60	59	49	42	34	29	34	43	43	47	55	62
Max	88	87	82	71	71	66	73	80	77	82	82	88

## **1.7 Chemical Composition of PM**

Particulate matter can differ in physical size and chemical composition ( Figure 2). Many different chemical compounds have been detected in PM, but they are mainly sulfates, nitrates, elements, polycyclic hydrocarbons, and other organic chemicals. Vega et al. (2001) indicated that the source of PM2.5 is crude oil combustion, followed by road and soil dust. PM10 originates from road, soil, and construction dust.



*Figure 2*. TSP source emission by their size distribution (Vega et al., 2001)

#### 1.7.1 Elemental Composition of PM

The elemental composition of PM depends on spatial and temporal factors. The composition reflects the source and activity around the area. Also, at a certain time of the year, there are higher PM concentrations and it contains a great variety of elements attached to the particles. The elemental composition of PM varies between areas and could depend on industrial, transportation, and meteorological factors. Naimabadi et al. (2016) reported on the PM10 concentration and its composition on normal days and during dust storms. It was concluded that there is no significant connection between their elemental compositions, but the study highlighted that elements in PM10 could lead to

cytotoxicity. The study area was similar to Qatar in terms of temperature and near Iraq and Kuwait. The elemental compositions of PM samples determined in this study are shown in Table 6 (Naimabadi et al., 2016).

## Table 6

Elemental composition of PM10 samples collected in Iraq and Kuwait (Naimabadi et al., 2016)

	Metal conten	ts of PM10 in	Metals contents of PM <sub>10</sub> in		
Elements	dust events	day (ng/m <sup>3</sup> )	normal da	ys (ng/m³)	
	Mean	SD	Mean	SD	
Al	35137.76	20751.96	8463.54	5462.89	
Fe	28387.17	26348.50	2309.10	1743.20	
Zn	34202.83	44192.53	16276	1534.50	
Pb	52.06	61.31	25.17	6.14	
Cr	72.62	81.36	8.24	9.20	
Cu	83.81	40.74	48.18	10.43	
Cd	19.50	15.49	18.66	7.98	
As	6.67	12.06	4.25	5.03	
V	82.96	93.32	6.27	0.30	
Ni	74.34	52.77	0.99	1.16	

Das et al. (2016) conducted a study in Baranagar, a crowded city in India with high PM emission from anthropogenic sources, between 2013-2014. It was highlighted in this study that the PM concentration surpassed the normal WHO organization limits and reached 783  $\mu$ g/m<sup>3</sup> (84–783  $\mu$ g/m<sup>3</sup>) for PM2.5 and 928  $\mu$ g/m<sup>3</sup> (167–928  $\mu$ g/m<sup>3</sup>) for PM10. Many toxic metals were detected in PM2.5, such as Cd, Cu, V, Cr, Ni, Zn, Mo, S, and Sb. Anthropogenic sources, mainly industry, have an enrichment factor for Ni, Cr, and Cu between 100 and 10 ng/m<sup>3</sup>, while Sn, Zn, Mo, Sb, Pb, and Cd had an enrichment factors between 1000 and 100 ng/m<sup>3</sup>. The elemental composition of PM2.5 in Kavala, Greece was investigated by Loupa et al. (2016). It was found that the highest concentrations of elemental components in PM2.5 were S (1321.0  $\mu$ g/m<sup>3</sup>), followed by Na (657.7  $\mu$ g/m<sup>3</sup>), K  $(374.68 \ \mu g/m^3)$ , Ca  $(448.00 \ \mu g/m^3)$ , Al  $(360.34 \ \mu m/m^3)$ , Si  $(325.50 \ \mu g/m^3)$ , Fe  $(147.30 \ \mu g/m^3)$  $\mu g/m^3$ ), Mg (126.32  $\mu g/m^3$ ), Zn (62.19  $\mu g/m^3$ ), and Ni (4.87  $\mu g/m^3$ ) (Loupa, Zarogianni, Karali, Kosmadakis, & Rapsomanikis, 2016). A recent study focusing on the chemical characteristic of particulate matter monitored indoor and outdoor environments for a duration of two months in Qatar was published by (Saraga et al., 2017). The authors found that there is a positive correlation between indoor and outdoor pollution where pollutants could enter through ventilation system and window/cracks into the building. They concluded that the indoor PM concentrations can be influenced by the outdoor PM concentrations during dusty days. The elemental compositions of outdoor PM 2.5 and PM10 samples collected in Doha as reported in this study is provided in Table 7.

Concentration (ng m <sup>-3</sup> )						
	$PM_{10}$	PM <sub>2.5</sub>				
	outdoors	outdoors				
Al	4288	2876				
Fe	5165	3357				
Mn	138	88.8				
Ba	115	72.2				
Sr	165	111				
V	35.7	28.6				
Cr	39.3	19.8				
Rb	5.41	3.64				
Ni	36.6	21.1				
Zn	63.8	45.8				
Cu	53.7	31.86				
Co	3.59	2.24				
Ga	1.85	1.22				
Pb	20.6	17.77				
Cs	0.45	0.31				
As	2.03	1.39				
Cd	0.27	0.17				

**Table 7**Outdoor elemental composition of PM2.5 and PM10 (Saraga et al., 2017)

#### **1.8 Health Effects of PMs**

An association has been established between PM pollution and various kinds of health effects (Cascio, 2016). The WHO reported that three million deaths around the world are a result of air pollution (WHO, 2016). In Europe, PM2.5 pollution was related to 432,000 premature deaths in 2012 due to long-term air pollution exposure (EEA, 2015). PM could affect the cardiovascular system and result in sudden heart attacks (Chan et al., 2016) and irregular heartbeat. Respiratory effects include asthma (Baldacci et al., 2015) and decreased lung function (USEPA, 2016). There were 428 cases of people who contracted chronic obstructive pulmonary disease in Iran in 2009-2013 (21, 111, 94, 102, and 98 in each respective year) due to air pollution issues (Khaefi et al., 2017).

In recent years, studies have also identified a link between PM exposure and cancer (Raaschou-Nielsen et al., 2016; Loomis et al., 2013; Hamra et al., 2014). PM with different size fractions can cause direct damage to DNA, and changes in DNA could cause cancer when there is no DNA repair mechanism (Lynch et al., 2016). Characterization of the components of PM is important to identify potential risks to human health (Bari et al., 2016). PM2.5 can reach deeper parts of the lungs and cause much more serious health effects compared to PM10 (Khan et al., 2010). A study on more than three million people was carried out to correlate the PM components with cancer in different European countries (Raaschou-Nielsen et al., 2016). It was concluded that the elemental composition is a significant cause of cancer. The study focused on eight elements (Cu, Fe, K, Ni, S, Si, V, and Zn), and highlighted the high levels of S and Ni (Raaschou-Nielsen et al., 2016).

Chen et al. (2016) investigated the mortality and lung cancer with long-term exposure (12 years) to PM. They found that with every increase in PM10 concentration of  $10 \ \mu g/m^3$ , the probability of mortality by lung cancer increases by 3.4–6.0% (Chen et al., 2016). In addition, the concentration of PM is also significant because its individual components can lead to different health effects (Forsberg et al., 2005; Cassee et al., 2013; Peters et al., 2015). Outdoor air pollution, particularly with PM as a major component, is classified as a Group 1 pollutant (carcinogenic to humans) by the International Agency for Research on Cancer (IARC, 2013). Malley et al. (2017) investigated the correlation of
PM2.5 pollution with preterm birth in 183 countries around the world in 2010. The results showed that mothers who are exposed to more PM2.5 pollution have more risk factors that contribute to increasing preterm birth (Malley et al., 2017). In this region, Naimabadi et al.'s study (2016) is the only one which provides a detail information on the link between the composition of PM and negative health effects and cancer. The study did not find a significant correlation (P > 0.05) between the composition of PM and its effects on human on normal or dusty days (Naimabadi et al., 2016).

There are several methods and instruments for measuring particulate matter characteristics and concentration. There are two main measurements for PM, which are concentration and size distribution. Concentration measurement of particles has mainly three methods which are Gravimetric (using filters, impactor), optical (Scattering: using Photometer, OPC, and CPC; Extinction: Opacity meter; Absorption: Spotmeter, Aethalometer, PASS, LII) and microbalance. The size distribution measurement has five methods which are the microscopical (using Microscopy), impaction (using Impactor), diffusion (using Diffusion battery), charging (using DMA), and compete systems (using SMPS, CPMA, DMS, FIMS, ELPI, and EDB) (Amaral, de Carvalho, Costa, & Pinheiro, 2015). Most common technique that is used as a reference sampling of PM concentration is the gravimetric method which is also used in this study.

## **2.1 Justification**

Air pollution is known to be a major public health and environmental issue (due to decreased visibility effect) all around the world. The assessment of air quality has been carried out in developed nations, such as the USA, Canada, UK, Germany, etc. and

developing nations in Asia, Africa, and the Middle East (e.g. Lebanon, Saudi Arabia, Iraq, Kuwait). A substantial body of epidemiological evidence now exists that establishes a link between exposure to air pollution and increased mortality (especially premature death) and morbidity due to a wide range of adverse cardiovascular and respiratory problems (Lira et al., 2012; Yaacoub et al., 2013; Abdulaziz et al., 2015). The information on the levels of PM and their probable source is very limited in Qatar; hence, there is a need to have a more comprehensive study on the particulate matter in term of its elemental composition and its impact on air quality. Therefore, this study was designed to monitor the levels of PMs (PM2.5 and PM10) at different locations in Qatar by integrating emissions within a framework of geographic information system (GIS) through the use of different mapping techniques depending on the available data. The results obtained from this study will be helpful for decision makers to formulate and implement policies that are feasible and sustainable to protect public health and the environment in Qatar.

### 2.2 Objectives of the Study

- Monitor the concentration of particulate matter (PM2.5 and PM10) at different locations in Qatar.
- 2- Determine the elemental composition of PMs to identify their possible sources.
- 3- Create a map by using Geographical Information System (GIS) to show the air quality based on PM2.5 and PM10 concentrations at select locations in Qatar.

## **CHAPTER III: MATERIALS AND METHODS**

### 3.1 Study Site

Five main sampling locations were chosen in this study: Qatar University (QU), (25° 21' 29.8692" N, 51° 29' 34.5984" E), Aspire Zone (25° 16' 2.2332" N, 51° 27' 6.9012" E), Education City (EC) (25° 19' 25.6512" N, 51° 25' 58.3716" E), Wholesale market (WSM) (25° 14' 47.2452" N, 51° 28' 35.1768" E), and Al-Wakrah City (25° 9' 53.1792" N, 51° 35' 38.7456" E) (Figure 3). The main reason for selecting these locations was based on many criteria, such as land use, activities, and traffic density. The selected areas have different forms of land use and have major educational facilities, transportation facilities, industrial buildings, health centers, local schools, and residential buildings.

Qatar University has the largest number of students and faculty members among universities located in Qatar. There are about 14,000 students registered in Qatar University who are not using buses as their main transportation. The majority of students depend on personal transportation to reach the university. In addition to heavy traffic activities, there is a metro being constructed on the northern side of Qatar University, which may affect the concentration of suspended particles in the air.

The Aspire zone is a critical location with common shopping areas, sport facilities, and public parks. At Education city, the activities are similar to those at Qatar University in terms of traffic and construction (metro work) activities, with new building constructions of the Qatar foundation as well. The fourth sampling site is the Wholesale market (WSM), which includes several markets that sell products like fish, animals, vegetables/fruits, and home accessories. Qatar depends mainly on importing food, animals, and accessories, so big trucks and other vehicles of different sizes are common on this market area.



*Figure 3*. Map of sampling locations.

The fifth location is Al-Wakrah city, which is one of the fastest growing cities in Qatar in terms of residential and industrial activities. It has the oldest and main desalination/electrical plant in Qatar (Ras Abu Fontas, just 2 km away from the center of Al-Wakrah City). With the new marine port (Hamad Port), there is more construction of commercial offices and buildings that started in parallel with metro activities as well. Since being accepted for holding the world cup 2022, there have been increases in terms of population and construction activities in the state of Qatar.

## 3.2 Sample Collection and on-site measurements of PM2.5 and PM10

Air samples were collected monthly using Deployable Particle Samplers (DPS; Figure 4; SKC Inc., PA, USA) between September 2016 and December 2016. The DPS is a 24-hour Li-Ion battery-operated system that is easy to operate and portable. Five DPS pumps (3 for PM2.5 and 2 for PM10 measurements) were placed at each sampling location. Each DPS was equipped with a compact internal impactor comprising of a PM2.5 or PM10 inlet and outlet, and a 47-mm filter cassette.



Figure 4. Deployable Particle Sampler

The PTFE filters (SKC Omega Specialty Division, PTFE filters 2.0  $\mu$ m pore size, 47 mm diameter) were weighed by using a microbalance (METTLER TOLEDO XP2U) and conditioned before each sampling time using the method of California Air Resource Base (SOP MLD 055, 2014) with little modification for temperature and humidity based on the available lab condition in Qatar University. The simultaneous sampling on PTFE filters allowed the subsequent chemical determination of all macro-components of PM. The system was maintained at 10.0 L/min flow rate (Table 8) during 24 hours of sampling period once a month with three PM2.5 and two PM10 pumps located at each station.

# Table 8

Flow rate measurements of each PM2.5 and PM10 sampling device at the time of

calibration

Device	Flow Rate (L/min)	Accuracy
Device#1 PM2.5	9.76	±0.17
Device#2 PM2.5	10.07	±0.04
Device#3 PM2.5	10.13	±0.03
Device#4 PM10	10.09	$\pm 0.03$
Device#5 PM10	10.14	±0.05

Before each sampling period, each pump was calibrated to confirm the 10.0 L/min flow rate. Height of the sampling sites ranged from 3-15 meters for all five stations where pumps were placed.



Figure 5. Exploded view of IMPACT Sampler

The formation of ambient PM depends on an interrelated and complex system of emission rates, meteorological processes, and atmospheric chemistry. Thus, data on surface and atmospheric temperatures, dew point, relative humidity, wind speed, wind gust, and sea pressure were obtained from the Qatar Meteorology department. All these data were provided in Appendix A.

## **3.4 Calculation of PM Mass**

The calculation of the concentration of PM was carried out depending on the main equation that is used for PMs in the book of "Code of Federal Regulations Government: 1985-1999" (United States. Office of the Federal, 1994). The average sampling flow of each device was recorded initially and after 24 hrs. Total dust collected on filters were divided by the air volume of 24 hrs of sampling to have the PM2.5 and PM10 levels. The initial and final weight of samples recorded to get collected dust weight. The following formula (Actual PM concentration at field condition) was used to calculate the PM2.5 and PM10 concentrations (Appendix B).

$$C = PM_{act} (P_{std} / P_{act}) (T_{act} / T_{std})^*$$

(United States. Office of the Federal, 1994)

C= Actual concentration of PM at field conditions ( $\mu g/m^3$ ) PM<sub>std</sub> = Concentration at standard conditions ( $\mu g/m^3$ ) P<sub>act</sub> = Average barometric pressure at the field during sampling (mm Hg) P<sub>std</sub> = 760 mm Hg  $T_{act}$  = Average ambient temperature at the field conditions during the sampling period (K)  $T_{std}$  = 298 K

\* Detailed formula sequences is available in Appendix B.

### **3.5 Identification of the elemental composition of PMs**

The elemental compositions of all PM samples were determined using the USEPA method 200.7 Revision 4.4 (EPA, 1994). The following elements were targeted based on their known presence in PM samples: Al, Ca, Na, Mg, Fe, K, Cl, Li, P, Ti, V, Cr, Mn, Co, Ni, Zn, As, Se, Rb, Sr, Cd, Sn, Sb, Ba, Pb, and Hg. Calcium, Si, Fe, Al, K, and Ti are crustal elements; Mg, K, and Na come from sea salt; and heavy metals such as Cd, Hg, Pb, etc. are from traffic or industrial pollution. The PTFE filters were collected from different stations and weighed within available lab conditions following the California Air Resource Base (SOP MLD 055, 2014) with some modification for temperature and humidity (using available material at the university). After 24 hrs sampling, the filters were removed from the impactor, placed in a sterile plastic dishes and brought to the acid digestion lab in the Environmental Science Center at Qatar University. Filters were put in PTFE tubes, and HNO<sub>3</sub> 70-68% (12 ml) and HF 40% (3 ml) were added at different times. Digested samples were transferred to new tubes for analyses by Inductively Coupled Plasma-Optical Emission Spectrometry (ICP-OES) Model Optima 7300 DV (Perkin Elmer Inc., Waltham, MA, USA) located in the Central Lab Unit (CLU) at Qatar University. The instrumental characteristics and operating parameters of ICP-OES are summarized in Table 7. A Blank, Duplicate, and CRM (certified Reference Material) were included as quality control. The accuracy of heavy metal measurements was evaluated using the Multi Element standard

solution IV (Ag, Al, B, Ba, Bi, Ca, Cd, Co, Cr, Cu, Fe, Ga, In, K, Li, Mg, Mn, Na, Ni, Pb, Sr, Tl, Zn Fluka Analytical, Busch, Switzerland). Selection of the elements was based on the main heavy metals and other elements in the air as determined previously (Khan et al., 2010; Segalin et al., 2017). The standards were dissolved in 1% HNO<sub>3</sub> and four major concentrations were prepared to establish a calibration curve of the target elements.

# **Table 9***Tuning parameters of the ICP-OES*

Instrument	Optima 7300 DV
Nebulizer/Spray chamber	Meinhard/Cyclonic
Injector	Quartz 2.0 mm ID
Resolution	Normal
Read Time	20 sec (min) – 50 sec (max)
Resolution	Normal
Plasma Gas	15 L/min
Auxiliary Gas	0.2 L/min
Nebulizer Gas	0.6 L/min
Power	1400 W
Plasma View	Axial

## **3.6 Determination of PM sources using Enrichment factor (EF)**

The enrichment factor was used to highlight the possible source of particulate matter elemental composition. It is based on using a reference crustal element as natural source and comparing it with particulate matter composition. Al and Fe can be used as reference elements as previously reported (Chan et al., 1997).

The EF values were calculated using Al as a reference element that gotten from Rudnick and Gao (Rudnick & Gao, 2003), and applied on the following equation:

$$EF=[(Xa/Refa)_{Sample}/(Xc/Refc)_{Crustal}]$$
 (Chan et al., 1997).

Where:

Xa=Target element in air PM samples.

Refa=Reference element in air PM samples (ex. Al).

Xc=Reference element from crust like target element.

Refc=Reference element from crust (ex. Al).

## **3.7 Determination of Air Quality**

The air quality based on PM2.5 and PM10 concentrations was determined using the Air Quality Index (AQI) tool. This tool categorizes the health risk based on the following specifications: Good air quality (Index value 0-50), "Moderate" air quality (51-100), "Unhealthy for Sensitive Individuals" (101-150), "Unhealthy" (151-200), "Very Unhealthy" (201-300), and "Hazardous" air quality (301-500) (Table 10). Calculation of the AQI can be done by the official web site of AQI calculator (www.AirNow.gov) to convert the PM concentration to one of the AQI category. By using AQI, people can take a decision that is related to their living/working locations, and reduction of air pollution.

## Table 10

Air Quality Index		Level of Health		
	(AQI) values	Concern	Colors	
	0 to 50	Good	Green	
	51 to 100	Moderate	Yellow	
	101 to 150	Unhealthy for Sensitive Groups	Orange	
	151 to 200	Unhealthy	Red	
	201 to 300	Very Unhealthy	Purple	
	301 to 500	Hazardous	Maroon	

Air Quality Index categories by Level of Health concern and Colors (AirNow, 2016)

## 3.8 Mapping of AQI values calculated based on PM concentrations and Land use

ArcMap 9.3 software was used to compute Air Quality Index (AQI) values calculated based on the daily concentrations of PM2.5 and PM10 at each sampling location. The AQI tool is available on the USEPA's website which is used to categorize the health risks associated with different levels of PM2.5 and PM10. Furthermore, land use maps were also included in this study to highlight the main land use activity surrounding sampling stations.

## **3.9 Statistical Analysis**

Different statistical correlation models were used to determine the significant relations between PMs, elemental composition, and meteorological data. The Statistical Analysis System (SAS Institute Inc., Cary, NC, USA) software was used to apply generalized linear model (GLM) using ANOVA to determine the significant differences between PM concentrations and their elemental compositions at different sampling locations and months at p<0.05.

## **CHAPTER IV: RESULTS AND DISCUSSION**

## 4.1 PM Concentrations by Location and Month

A total of 100 samples (60 of PM2.5 and 40 PM10) were collected for four months starting from September to December 2016. The overall mean concentrations of PM2.5 and PM10 ranged from 50  $\mu$ g/m<sup>3</sup> to 64  $\mu$ g/m<sup>3</sup> and 127  $\mu$ g/m<sup>3</sup> to 185  $\mu$ g/m<sup>3</sup>, respectively (Table 11).

## Table 11

The mean concentrations of PM2.5 and PM10 samples collected from different Stations and Meteorological data

			Stations			
Parameters	Qatar	Education	Aspire Zone	Whole Sale	Al Wakrah	
1 al anicters	University	City		Market		
Average PM2.5	50 · <b>2</b> 45	50±2.47 65±1.90 55±2.24		<b>50 . 2 01</b>		
Conc. ( $\mu g/m^3$ )	50±2.47			59±3.01	57±2.08	
Average PM10	rage PM10					
$138\pm3.7$ Conc. (µg/m <sup>3</sup> )	138±3.72	<b>156±3.06</b> 127±3.90	185±3.64	160±2.79		
Temperature					5 26.47±1.25	
(°C)	26.90±1.16	27.92±1.18	8 26.97±1.15 28.00±1	28.00±1.26		
Wind Speed		75±0.32 1.80±0.26 1.80±0.35		2.10±0.52	4.15±0.60	
(m/s)	1.75±0.32		1.80±0.35			
Humidity (%)	60.00±1.78	59.25±1.97	61.00±2.02	50.00±1.01	62.00±1.24	

The concentrations of both PM2.5 and PM10 were relatively higher than the EPA standards (for PM10= 150  $\mu$ g/m<sup>3</sup> and PM2.5 = 35  $\mu$ g/m<sup>3</sup>) and WHO Standards (PM10= 50  $\mu$ g/m<sup>3</sup> 24 hr mean and PM2.5 = 25  $\mu$ g/m<sup>3</sup> for 24 hrs), as well as Qatari Ministry of Environment and Municipality (only 150  $\mu$ g/m<sup>3</sup> for PM10) standards (Figure 6, Tables 1 and 3).



*Figure 6*. The comparison of the USEPA standards and the average mean concentrations of PM2.5 and PM10 recorded at different locations during the study period.

These results were expected since Qatar is an arid region known for its frequent dust storms and dusty environment. In a recent study, Saraga et al. (2017) reported the indoor and outdoor PM2.5 and PM10 levels being higher than 25 and 50  $\mu$ g/m<sup>3</sup> (WHO standards) on most of the sampling days between April 22, 2015 and June 21, 2015 in Qatar. Similar findings were also recorded in Kuwait, indicating poor air quality based on PM2.5 levels (Brown et al., 2008).

A high concentration of PM10 was recorded at three stations: Wholesale Market, Al Wakrah City, and Education City, with concentrations of 185  $\mu$ g/m<sup>3</sup>, 160  $\mu$ g/m<sup>3</sup>, and 156  $\mu$ g/m<sup>3</sup>, respectively (Table 11). The concentration of PM2.5 reported at all five stations exceeded the EPA daily limit of 35  $\mu$ g/m<sup>3</sup> (Figure 6), indicating relatively poor air quality around these areas. The PM2.5 concentrations at Qatar University, Education City, WSM, Al-Wakrah City, and Aspire Zone were 50, 64, 59, 57, and 55  $\mu$ g/m<sup>3</sup>, respectively (Table 11). These high levels could be due to different human activities as observed in these stations.

The Wholesale Market (WSM) is an economic activity site where fish market, animal market, and produce market receive most of Qatar's food imports. Massive movement of trucks to transport food supplies to the market on unpaved road creates suspended particulate matter. There is an open area on the southern side of WSM and a semi-closed road that uptakes dust particles that come from the open area near the animal market when the wind is blowing from south of WSM which is the area with heavy traffic and logistic activities/industries (Figure 10). This could be one of the contributing factors for PM10 levels reaching considerably high concentrations compared to other sampling sites. Such findings were also supported by Zhu et al. (2015) and Patra et al. (2008) who mentioned that the concentration of PM10 close to roads is due to resuspended particles. Another factor which might result in high PM10 levels recorded at WSM is large cattle feedlots that supply the local market with cattle meat, which is under category of low to medium impact industries, and there is large logistic area for governmental ministries (Figure 10). The relationship between the cattle areas and high PM10 concentration was previously reported by (Guo et al., 2011).

In April 2017, the fish market and animal market at WSM was partially moved to UmSalal area located in the north of Qatar. As a result of this change, a reduction in PM10 levels is expected, but this hypothesis needs to be tested since this study only covered the period of September to December 2016.

The highest mean concentrations of PM2.5 and PM10 were monitored at concentrations of  $64 \,\mu\text{g/m}^3$  and  $185 \,\mu\text{g/m}^3$  at Education City and WSM, respectively (Table 10). Education city is recognized as one of the fastest growing sites in Qatar. The sampling station at EC was in the middle of Education City, which is surrounded by several universities and facilities that are still under construction -which considered as a temporary state-. During sampling, construction work was ongoing near the sampling station at EC, which may have contributed to high concentrations of both PM2.5 and PM10 as recorded at this location. The area is also housing many educational institutions, the largest convention center in Qatar, and many commercial sites. The vehicle movement in this area could also be another important factor contributing to the high concentrations of PMs (Araújo, Costa, & de Moraes, 2014).

#### 4.2 Air Quality Index Values Calculated based on PM2.5 and PM10 Concentrations

Air quality index (AQI) tool is recognized as one of indexes to determine health risks associated with PM levels based on the USEPA guidelines. It categorizes the air quality based on 24 hr PM concentrations using color codes, such as green for Good, yellow for Moderate, orange for Unhealthy for sensitive groups, red for Unhealthy, purple for Very unhealthy, and maroon for Hazardous air quality levels (AirNow, 2016). Air quality is progressively known as a serious issue for human health and is a subject for which comprehensive global emission data are missing. Using AQI, the quality of local air can be determined and a warning system can be created to inform the public, especially sensitive groups to protect their health.

The AQI values calculated based on PM2.5 concentration indicated that the air quality was in "moderate" range in September and December. While, the AQI levels exhibited "Unhealthy" rating during the months of October and November (Table 12). The 24 hrs mean levels of PM10 resulted in relatively better AQI levels which were reported mainly in the "Moderate" category during the entire sampling period, compared to the means of PM2.5 (Table 12) . These differences in AQI during sampling days are important to highlight human activities which were relatively low in September and December due to national holidays and late start of schools in September and winter breaks in December. On the other hand, human activities are at the highest levels in the months of October and November since schools/universities are open and running in full-term. Based on the AQI levels calculated using the concentrations of PM2.5 and PM10, there were 25%, 37.5%, and 37.5% of total days in the category of "Unhealthy", "Moderate", and "Unhealthy air quality for Sensitive Groups" during the study period,

respectively (Figure 7). It is noteworthy to mention that there was no single day with "good" air quality during the four months of sampling, demonstrating that air quality associated with PM air pollution could be a significant public health issue in Qatar. Even the PM levels are considered high in Qatar based on the findings obtained in this study and the recent report published by WHO (2016), the number of death and respiratory illness related to PM pollution is surprisingly low. This might create an opportunity for the public agencies to review current standards and establish more reasonable standards considering the arid environment of Qatar since most PM pollution is be realted to natural sources.

## Table 12

The Air Quality Index Values calculated based on the concentrations\* of PM2.5 and PM10 during the study period

	Air Quality Inde	dex Values		
Month	PM2.5	PM10		
September	104 Unhealthy for sensitive group (37 μg/m3)	93 Moderate (140 µg/m3)		
October	156 Unhealthy (66 µg/m3)	100 Moderate (154 µg/m3)		
November	73.01 Unhealthy (66 µg/m3)	111 Unhealthy for sensitive group (176 μg/m3)		
December	142 Unhealthy for sensitive group (52 $\mu g/m3$ )	95 Moderate (144 µg/m3)		

\*Values in parentheses are PM2.5 and PM10 concentrations.



*Figure 7.* The percentage distribution of different Air Quality Index ratings during the sampling months.

At Qatar University and Aspire Zone areas, the AQI levels indicated "Moderate" air quality for PM10 levels and "Unhealthy air quality for sensitive groups" for PM2.5 levels (Table 12). These two locations share similar land use exercises and have many green areas with dense tree population compared to other study sites (Figures 8 and 9). McDonald et al. (2007) investigated the positive changes in particulate matter of urban tree planting on the concentrations and depositions of particulate matter, and found that PM10 levels can be reduced up to 26% if tree density is increased. Trees also can support the air quality by removing 4.7 ton of PM2.5 annually (Nowak et al., 2013). Three of five

stations included in this study had "Unhealthy" air quality rating, which might negatively impact human health (Table 13). However, this correlation needs to be further investigated to establish the link between unhealthy air quality associated with high concentrations of PMs and their health effects on human, especially on sensitive groups by using data on health statistics.

## Table 13

Air Quality Index Values calculated based on the four month averages of PM2.5 and PM10 at different stations

	Air Quality Index			
Location	PM2.5*	PM10*		
Qatar University	136 Unhealthy for sensitive	92 Moderate (138 µg/m3)		
	Groups (50 µg/m3)			
Education City	155 Unhealthy (64µg/m3)	101 Unhealthy for sensitive Group		
2		(156 µg/m3)		
ADLQ (Aspire Zone)	150 Unhealthy for sensitive	86 Moderate (126.69 µg/m3)		
	Group (55 µg/m3)			
Whole Sale Market	153 Unhealthy (59 µg/m3)	115 Unhealthy for sensitive Group		
		(185 µg/m3)		
Al Wakrah	152 Unhealthy (57 µg/m3)	103 Unhealthy for sensitive Group		
		(160 µg/m3)		

\*Values in parentheses are the average concentrations of PM2.5 and PM10 collected for a period of four months.



Figure 8. Land use Illustration at Aspire Zone



Figure 9. Land Use Illustration at Qatar University



Figure 10. Land Use Illustration at Whole Sale Market



Figure 11. Land Use Illustration at Education City

## 4.2 Mapping of AQI Values Based on PM2.5 and PM10 Concentrations

The AQI values calculated based on PM concentrations at five stations during the moths of September-December 2016 were computed using GIS in combination with Google Earth mapping system. Figures 12 and 13 illustrate the AQI category for each site. Orange color was the dominant color in both maps, meaning that the air quality was considered "Unhealthy for sensitive groups" at these locations. This is a concern for part of the society who live in these areas since they can be directly affected by poor air quality. People with respiratory diseases, children and elderly are the groups who are at risk to be affected the most as a result of poor air quality as determined in this study. Such conditions might exacerbate likelihood of respiratory symptoms and aggravation of lung diseases, such as asthma. People who have heart and lung diseases could also be affected by poor air quality -specifically particulate matter pollution- as reported in previous studies (Gauderman et al., 2007; Pope et al., 2006; Zanobetti et al., 2000). Martins et al. (2004) reported that the mortality of elderly people increased from 1.4% to 14.2% in Brazil when the concentration of PM10 increased by 10  $\mu$ g/m<sup>3</sup>.

AQI values indicating "unhealthy" air quality (red color) based on the 24 hr mean PM2.5 concentrations were detected for three stations which were Education city, Whole Sale Market, and Al Wakrah city (Figure 12). This condition could lead to adverse health effects even in healthy people and could cause serious health consequences for sensitive groups. A similar pattern was also observed for PM10-based AQI levels at the same sampling station (Figure 13), meaning that there is a positive correlation between high concentrations of PM2.5 and PM10 and dangerous AQI categories (Mohan & Kandya, 2007).



Figure 12. PM2.5 based Air Quality Index Values at the sampling stations



Figure 13. PM10 based Air Quality Index Values at the sampling stations

#### **4.3 Particulate Matter Elemental Composition**

The total concentrations of elements detected in PM2.5 and PM10 samples collected from five stations on different days are listed in Tables 16 and 17. The concentrations of elements in PM samples were comparable to previous studies from the Middle East region (Brown et al., 2008; Saraga et al., 2017).

It is important to note that there was a significant correlation between elemental composition of PM2.5 and PM10 and sampling location. The concentration of elements detected in PM10 samples was significantly different (p<0.001) at each sampling location. The highest significant differences were observed among crustal and non-crustal elements, such as Al, Ca, Mg, Fe, Li, V, Cr, Mn, Ni, Co, , As, Sr, , Ba, Pb, while the concentration of Na, Zn , and Cd were not significantly different (p>0.05) at different sampling locations. There was also monthly variations in terms of elemental composition of PM10 samples, especially for (Al, Na, Mg, Fe, Li, V, Ba, and Pb) with (p<0.001) (Appendix C). (Cheung et al., 2011) also determined a significant correlation between the elemental composition of PM samples and sampling time of the year.

The highest concentrations of elements were detected in PM10 samples collected from WSM. Al, Ca, Na, Mg, Fe, Cr, Ni, Co, Sr, Cd, and Ba had significantly (p<0.05) higher levels compared to other stations (Table 17), while the concentrations of the same elements (Al, Ca, Mg, Fe, V, Ni, and Cd) were recorded at the highest level for PM2.5 samples collected from Education City (Table 16). It is expected that industrial and economic sites like the Wholesale market would normally have higher concentrations of metals such as Cr, Ni, and Ba as the land use illustration indicates (Table 16). It is important to compare the results obtained in this study with others from the same region to be sure that there is no increase of toxic elements in the air column. The highest concentrations of toxic non-crustal elements such as Cr, Ni, V, Pb, and Cd were 26.66, 21.87, 22.78, and 0.65, 4.91 ng/m<sup>3</sup> for PM10 and 17.48, 12.06, 14.77, 2.57, and 0.69 ng/m<sup>3</sup> for PM2.5, respectively (Tables 16 and 17).

A previous study conducted in Qatar by Saraga et al. (2017) recorded some of these elements: Cr (39.3 ng/m<sup>3</sup>), V (35.7 ng/m<sup>3</sup>), Cd (0.27 ng/m<sup>3</sup>), and Pb (20.6 ng/m<sup>3</sup>) in PM2.5 samples collected from a site known to have busy traffic. In another study, Naimabadi et al., 2016 reported the presence of similar elements in PM samples collected from a desert in Iran (another Middle East country), which has a similar climate to the Arabian region. The authors recorded high elemental concentrations of heavy metals during the dusty period, which is the most dominant time throughout the year in Qatar. The concentrations of Cr (72.62 ng/m<sup>3</sup>), Ni (74.34 ng/m<sup>3</sup>), V (82.96 ng/m<sup>3</sup>), and Pb (52.06 ng/m<sup>3</sup>) were three times higher than the values determined in this study (Naimabadi et al., 2016).

Investigation of the relation of the elements was based on grouping them according to their possible source. Determination of Ca, K, Al, and Fe in PM samples usually indicates the source as upper earth crust. Other elements like Cu, Zn, and Pb are mainly considered as indicators of traffic emission elements (Querol et al., 2001; Manoli et al., 2002; Fang et al., 2003). Elements such Ni and V are recognized as elements coming from burning fossil fuel, and other trace elements such as Cr and Cd result from burning coal. Ni, Cr, Cu, Sn, Zn, Mo, Sb, Pb, and Cd indicate the source as industrially related pollution (Das et al., 2015).

The concentrations of crustal elements such as Al and Fe were relatively high in all samples collected in this study, which is expected since these elements are the normal properties of desert dust. In all stations, the concentrations of non-crustal elements were lower compared to a previously reported study in Qatar (Saraga et al., 2017). This could be due to sampling location and duration. The locations chosen in this study had different activities and sampling was carried out during a warm season. While, Saraga et al. (2017) conducted PM sample collection during the hot season (July) in an area known to be highly crowded part of Doha.

# Table 14

	QU	EC	AZ	WSM	AW
Al	352.19±28	628.47±66	452.04±118	555.10±105	314.66±58
Ca	5176.29±529	6808.56±884	4820.03±942	5332.94±647	4168.10±451
Na	620.374±303	443.47±148	647.43±382	397.89±71	319.55±56
Mg	485.43±89	738.49±68	601.47±175	640.94±75	407.84±46
Fe	450.20±36	666.18±65	532.95±111	630.68±105	354.46±49
Li	0.10±0	0.24±0	0.32±0	0.24±0	0.64±0
V	11.9892±1	14.7705±1	9.42±1	11.91±1	11.02±0
Cr	14.72±0	15.2453±0	15.95±1	17.48±0	13.26±1
Mn	6.27±0	10.04±1	11.18±1	10.59±2	7.13±0
Ni	11.76±1	12.06±0	11.66±1	11.75±0	10.68±0
Zn	201.79±15	218.76±22	188.53±17	198.65±19	223.98±0
Sr	10.53±2	14.01±2	9.15±2	13.41±1	6.88±0
Cd	0.55±0	0.69±0	0.51±0	0.59±0	0.60±0
Ba	16.32±1	15.68±1	11.45±1	20.58±1	11.12±1
Pb	<b>2.57</b> ±1	ND*	1.74±0	1.86±0	1.56±0

Elemental composition concentration of PM2.5 samples collected from different locations

\* ND. Not detected elements.

# Table 15

PM10 Mean Elemental Composition Concentration (ng/m <sup>3</sup> )						
	QU	EC	AZ	WSM	AW	
	1275.32	2701.46	2215.90	2980.44	2106.97	
Al	±191	±410	±443	$\pm 685$	±389	
	10238.49	19238.23	17681.85	21286.56	20222.25	
Ca	±2705	±2909	±3479	±2766	±3078	
	1036.78	1241.33	1165.43	1260.99	1223.14	
Na	±176	±162	±234	±166	±166	
	1992.07	3814.10	3462.83	4507.67	3768.90	
Mg	±550	±507	±697	±650	±335	
	1645.29	2832.27	2425.55	3359.17	2258.90	
Fe	±246	±414	±385	±657	±302	
Li	$0.68\pm0$	2.02±0	1.59±0	2.20±0	2.42±0	
V	16.70±2	22.78±1	17.66±1	22.23±2	19.02±1	
Cr	18.98±0	22.28±1	22.66±1	26.66±1	21.97±1	
Mn	19.88±3	30.14±4	31.73±5	33.01±4	27.94±3	
Ni	17.36±1	20.33±0	18.17±1	21.87±1	16.97±1	
Zn	157.71±20	233.43±39	263.03±37	183.33±23	318.35±58	
Sr	36.41±9	61.31±9	47.13±8	74.88±11	49.41±4	
Cd	0.64±0	0.54±0	0.42±0	0.65±0	0.44±0	
Ba	51.17±11	54.84±5	47.92±5	80.57±10	53.34±5	
Pb	2.99±1	0.37±0	2.76±1	3.16±1	4.91±1	

Elemental composition concentration of PM10 samples collected from different locations

PM samples collected from Qatar University had much lower elemental concentrations compared to Education city. For PM10, just two peak values recorded for toxic non-crustal elements (Cr and Ba), while the concentrations of other toxic non crustal elements like Ni, Cd, and Pb were not significant comparing to other stations (Table 16). For PM2.5, there was a peak recorded for Pb, which is directly related to traffic pollution. Diesel fuel from trucks is the main source of Pb since most cars use lead- free gasoline in Qatar. The sampling location at Qatar University was right next to the main road (Figure 14), while at Education city, the station was not very close to the street (Figure 15).



Figure 14. Illustration of Land use at of Qatar University


Figure 15. Illustration of Land use at Education City

The spatial distribution of the elements varied by location based on the land use activities as illustrated in the study areas. Compared to other studies (Saraga et al., 2017; Hassan et al., 2016), the concentrations of heavy metals in PM2.5 and PM10 samples collected from different locations in Qatar were relatively low. These concentrations can give a general picture of the air quality level in the areas monitored in this study. Figures 16 and 17 show the concentrations of main elements related to the anthropogenic sources (Cr, Pb, Cr, Li, Cd, and V). Cr had the highest level in both PM samples collected from WSM station compared to other stations, exceeding 26 ng/m<sup>3</sup> and 17 ng/m<sup>3</sup> for PM10 and PM2.5, respectively. The highest concentration of chromium highlights the source as industrial origin.

Pb is recognized as one of the main indication of traffic related PM pollution due burning diesel in vehicle engines (Wang et al., 2003). The concertation of Pb recorded at the highest concentration in QU for PM2.5 and in Al-Wakrah for PM10. The lowest concentration of Pb recorded in EC. Since the sampling location at EC was not very near a roadside, PM pollution source is probably mainly due to construction activities.



*Figure 16*. Average concentrations of heavy metals in PM2.5 collected from different sampling stations.



*Figure 17*. Average concentrations of heavy metals in PM10 samples collected from different sampling stations.

#### 4.4 Possible sources of PMs based on Enrichment Factor Analysis

The enrichment factor (EF) analysis was used based on the concentration of individual elements in the air particulate samples compared with their concentrations in the crust (Rudnick and Gao, 2003). This analysis provides useful information in determining the sources of elements detected in PM samples. In this study, Al was used as a reference element assuming that Al will be present at low concentrations in air samples. Usually, high EF values indicate the origin of elements to be from non-crustal anthropogenic sources, while low EF values are indicative of earth-crust or soil as main source

In this study, the EF values were categorized into three main classes , for example, Fe and Mg are mainly crustal elements and have small EF values indicating the PM source as natural (Chan et al., 1997). The EF values ranging between 1-9 are indicative of nonanthropogenic sources like Fe, Na, Mg, Sr, Ba, and Li (Table 18). The EF values considered to be in the Moderate category range from 10 to 100. The elements in this category are Ni, Cr, and V which are recognized as anthropogenic sources, where Ni and V come from burning of fossil fuel (Zhang et al., 2009) and Cr is from industrial activity. The EF values higher than 100 are considered as significant anthropogenic sources, such as Zn, Cd, and Pb (Table 18).

The relatively high EF values for Zn, Cd, and Pb may be indication of significant anthropogenic contribution in both PM2.5 and PM10 samples collected in Qatar, even the elemental concentrations are lower than some of previous studies in the Middle East region.

Near to the results of EF were also reported in different areas (Jeddah city and Rabigh city) of Saudi Arabia where EF values of Pb and Cd were more than1500 for Pb and 8800 for Cd (Khodeir et al., 2012; Nayebare et al., 2016). This reported high EF values for Pb, Cd, and Zn trace the source of PM pollution to mainly traffic and industrial activities. It should be emphasized here that Pb and Cd are known to be highly toxic to humans, especially children. Therefore, a more comprehensive study including the collection of soil and more air samples needs to be carried out to determine the exact source of these two toxic metals in these locations.

### Table 16

Enrichment Factor (EF) Values for elements determined in PM2.5 and PM10 samples  $(\mu g/m^3)$ 

	PM2.	5	PM10					
	Elemental		Elemental					
	Concentration	EF	Concentration	EF				
	(Average)	24	(Average)					
	$(ng/m^3)$		$(ng/m^3)$					
Al	496.95	R.E*	2256.01	R.E*				
Fe	526.89	2.20	2504.23	2.30				
Na	485.74	2.76	1185.53	1.48				
Mg	574.83	6.30	3509.11	8.47				
Li	0.30	2.40	1.78	3.06				
V	11.82	19.98	19.67	7.32				
Cr	15.33	27.33	22.51	8.83				
Mn	9.04	1.91	28.54	1.33				
Ni	11.58	40.42	18.94	14.55				
Zn	206.34	505.11	231.17	124.65				
Sr	10.79	5.53	53.82	6.07				
Cd	0.58	1072.65	0.53	215.96				
Ba	15.03	5.78	57.56	4.88				
Pb	1.93	933.34	2.83	301.56				

\*Reference element used in EF calculation.

### **CHAPTER V: CONCLUSION**

The present study aimed at investigating PM pollution, chemical composition and source of PM and its impact on air quality at different locations in Qatar. The gravimetric measurements revealed that the four months average PM concentrations exceed the WHO and USEPA standards in some stations. The concentration of PM2.5 and PM10 in Qatar University, Education city, Aspire Zone, Whole Sale Market, and Al-Wakrah city were 50, 64, 55, 59, 57, 138, 156, 127, 185, and 160  $\mu$ g/m<sup>3</sup>, respectively. Overall, the mean concentrations of PM10 and PM2.5 were recorded as peak at Whole Sale Market (185  $\mu g/m^3$ ) and Education City (64  $\mu g/m^3$ ), respectively. Activities in these two stations were mainly industrial/trading (WSM) and construction in (EC). Having such concentration could be reduced after finishing the construction activities and more efficient of transportation management in industrial/trading areas. The Air Quality Index tool was also used in this study to categorize the health risk associated with different PM levels. The AQI values indicated that 37.5% and 25% days of "Moderate" air quality, "Unhealthy for Sensitive Group" air quality, and "Unhealthy" respectively. The concentrations of elements in PM samples were relatively low compared to previous studies. The enrichment factor analysis showed that high concentrations of Pb, Cd, and Zn were probably due to road traffic emission and activity relates to medium industrial activity near to WSM. The presence of these heavy metals may also influence the degraded air quality in the sampling area as confirmed by AQI values.

This study highlights the urgent need to establish a strategy for continuous monitoring and a reliable and real-time warning system to inform the public about the air

quality in Qatar. It is important to note that this study is considered as a pilot study to determine air quality based on PM pollution in Qatar. The findings obtained here provide important data which can be used to assist government agencies to establish air quality management system. However, this study was limited in terms of sampling and chemical composition analysis. Hence, future plans should include specific studies in determining long term effects of exposure to PM pollution in Qatar. Examples of such studies might be epidemiological studies using health statistics and PM pollution data, inclusion of a larger sampling area, and measurement of various PM sizes (like PM1) and different chemical components of PMs, such as PAHs and ions.

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## APPENDICES

QMD :	gi koji isi Ganada	ily Average of	Some Meteo	rological Dat	a At Qatar Univ	versity 🧲	arringer name
Date	Mean Temperature degree celsius	Mean Dew Point degree celsius	Mean Relative Humidity percentage	Total Rain Fall millimeters	Mean Wind Speed meters per second	Maximum Wind Gust meters per second	Mean Wind Direction 360 degree rose
6/1/2016	33.1	8.7	26	0	5.3	26.4	320
6/2/2016	34.6	4.9	18	0	3.8	19.2	300
6/3/2016	34.4	8	22	0	1.8	15.3	350
6/4/2016	33.2	14.4	36	0	2.7	13.2	20
6/5/2016	32.3	12.7	34	0	1.4	10.7	60
6/6/2016	33.6	16.4	40	0	1.1	7.8	100
6/7/2016	35.6	11.2	25	0	1.6	44.5	290
6/8/2016	37.5	9.3	21	0	5.2	25.7	330
6/9/2016	36.6	8.6	22	0	6.2	31.6	330
6/10/2016	35.2	15.5	33	0	3.2	30.6	350
6/11/2016	38.3	10.7	23	0	4.3	29.6	340
6/12/2016	36.1	8.3	22	0	7	17.5	350
6/13/2016	34.2	9	24	0	6.5	23.3	320
6/14/2016	34.4	3.9	20	0	6.5	28	320
6/15/2016	35	9.1	27	0	4.7	49	330
6/16/2016	33.7	16.4	41	0	2.4	13.6	20
6/17/2016	34.6	15.3	35	0	3.5	20.4	0
6/18/2016	34.9	11.2	27	0	4	16	360
6/19/2016	35.9	13.3	33	0	5	22.7	340
6/20/2016	37.7	9.3	22	0	5.2	28	340
6/21/2016	37.3	10.1	24	0	5.2	22.2	340
6/22/2016	35.2	16.7	35	0	2.7	12.7	20
6/23/2016	34.6	16.6	36	0	2.4	18.1	30
6/24/2016	34.9	16.4	37	0	2	19.6	50
6/25/2016	35.3	16.9	37	0	2	9.8	80
6/26/2016	35.9	15.4	36	0	1.2	20.4	60
6/27/2016	39	9.5	21	0	2.5	24.5	330
6/28/2016	36.6	19.7	41	0	3	25.7	30
6/29/2016	34.7	22.8	53	0	1.6	15	90
6/30/2016	36.5	18.9	40	0	1.3	15.8	50
7/1/2016	39.7	13.2	25	0	4.9	30.6	330
7/2/2016	38	13.1	26	0	6.5	29.6	330
7/3/2016	37.6	13	26	0	6.3	28.8	330
7/4/2016	38.2	10.7	23	0	3.6	22.2	340
7/5/2016	36.8	18.2	38	0	2.7	18.4	330
7/6/2016	36.6	14.3	28	0	2.4	36.2	10

7/7/2016	37.2	14.5	29	0	2.6	13	340
7/8/2016	38.8	14	26	0	3.1	20.8	340
7/9/2016	39.7	11.7	22	0	4.9	22.7	320
7/10/2016	39.6	9.6	20	0	6	28.8	320
7/11/2016	39.1	10.7	22	0	5.9	24.5	340
7/12/2016	37.8	10	23	0	5.7	22.7	320
7/13/2016	37.9	11.2	24	0	4.3	30.6	320
7/14/2016	36.1	20	45	0	2.8	9.3	20
7/15/2016	35.8	22.2	49	0	2.3	19.6	30
7/16/2016	35.2	23.8	54	0	3	29.6	20
7/17/2016	36.6	23.4	50	0	2.6	23.9	30
7/18/2016	35.8	26.5	62	0	2.6	28.8	90
7/19/2016	38	22.7	43	0	2.4	20.8	110
7/20/2016	39	21.5	39	0	2	10.2	150
7/21/2016	36.6	25	52	0	2.5	35	90
7/22/2016	37.3	25.1	53	0	1.7	15.5	80
7/23/2016	37.7	23.3	47	0	3.1	31.6	320
7/24/2016	35.1	27.4	65	0	2.8	26.4	40
7/25/2016	36.7	28.5	64	0	2.9	11.6	90
7/26/2016	36.2	28.7	66	0	2.9	28.8	90
7/27/2016	35.6	27.6	64	0	2.4	11.5	80
7/28/2016	35.6	27.1	62	0	2.8	28	100
7/29/2016	35.8	26.9	61	0	2.7	9.1	110
7/30/2016	35.4	25.8	59	0	1.3	28.8	80
7/31/2016	35.6	26.2	59	0	1.9	22.2	90
8/1/2016	36.9	22.7	47	0	1.2	11.8	70
8/2/2016	38.7	18.8	37	0	1.7	20	300
8/3/2016	37.8	23.5	49	0	2.6	19.6	70
8/4/2016	35.8	27.1	62	0	2.4	10.4	70
8/5/2016	35.7	26.8	61	0	2.5	22.7	40
8/6/2016	36.3	24.5	54	0	1.5	22.2	50
8/7/2016	36.7	24.6	53	0	1.9	12	70
8/8/2016	36.8	22.9	53	0	1.6	20	60
8/9/2016	35.8	25.5	58	0	1.4	27.2	40
8/10/2016	36.5	22.5	49	0	1.8	16	30
8/11/2016	35.9	26	59	0	1.9	31.6	90
8/12/2016	35.7	27.8	64	0	2.3	10.8	70
8/13/2016	36.3	27	64	0	2	17.5	60
8/14/2016	37.9	23.1	49	0	1.4	17.1	230
8/15/2016	38.1	22.3	44	0	1.2	12.7	120
8/16/2016	37	23.8	49	0	1.3	25.1	30

Appendix A: Meteorological Data

8/17/2016	37.5	22	45	0	3.3	29.6	350
8/18/2016	36.7	22.9	48	0	1.5	9.5	70
8/19/2016	35.4	25.9	60	0	2	12.7	70
8/20/2016	35.5	25	57	0	1.4	30.6	60
8/21/2016	35.4	25.7	60	0	1.2	28.8	80
8/22/2016	35.3	27.4	65	0	1.9	19.2	110
8/23/2016	35.9	26.1	60	0	2	12	100
8/24/2016	36.4	23.5	53	0	2.1	24.5	70
8/25/2016	36.3	24.4	54	0	2.4	26.4	80
8/26/2016	36.5	23.3	53	0	1.9	12.7	100
8/27/2016	36.1	25.1	56	0	1.3	9.8	70
8/28/2016	35.5	25.3	58	0	2	32.6	80
8/29/2016	35.9	24.5	56	0	2	9.5	80
8/30/2016	35.9	24.1	55	0	1.5	20.8	100
8/31/2016	36.4	21.5	48	0	1.6	18.1	70
9/1/2016	37.7	20.9	43	0	2	31.6	140
9/2/2016	37.6	21.3	45	0	2.6	20	130
9/3/2016	35	28	68	0	2.4	35	80
9/4/2016	35	26.8	63	0	1.9	8.6	60
9/5/2016	35.9	26.4	60	0	2.1	14.8	30
9/6/2016	36.5	22	45	0	2.7	29.6	350
9/7/2016	35.3	21	45	0	3.9	16	360
9/8/2016	34.1	21.4	49	0	1.4	25.1	20
9/9/2016	33.3	20.4	48	0	1.6	18.4	70
9/10/2016	32.9	23.3	58	0	1.5	21.7	60
9/11/2016	32.5	23.1	59	0	1.5	29.6	60
9/12/2016	33	24	61	0	2	23.3	60
9/13/2016	33.1	23.4	59	0	1.5	25.1	90
9/14/2016	33.8	26.5	68	0	1.9	9.8	70
9/15/2016	33.7	24.9	62	0	1.2	18.1	70
9/16/2016	33.4	22.8	56	0	1.5	14	70
9/17/2016	34.5	18.8	46	0	1.3	14.4	70
9/18/2016	34.3	22.5	56	0	2.2	28.8	350
9/19/2016	34.2	22	52	0	2.4	18.4	40
9/20/2016	34.1	13.6	33	0	3.6	25.7	340
9/21/2016	33.2	18	43	0	3.2	36.2	360
9/22/2016	32.4	23	61	0	1.5	21.7	60
9/23/2016	33.6	21.6	55	0	2.4	20.8	330
9/24/2016	33.7	19.9	47	0	2.8	18.4	330
9/25/2016	33.2	17.8	44	0	3.8	25.7	330
9/26/2016	31.7	20.7	54	0	1.8	9.3	0

Appendix A: Meteorological Data

9/27/2016	30.7	24.2	69	0	1.3	10.1	80
9/28/2016	32.9	19.4	51	0	1.8	28.8	360
9/29/2016	33.5	16.2	42	0	4	31.6	320
9/30/2016	31	14.8	40	0	5.1	24.5	320
10/1/2016	30.2	14.5	41	0	3	14.4	330
10/2/2016	29.4	13.4	41	0	1.8	20.8	20
10/3/2016	29.5	13.4	41	0	1.8	12.7	330
10/4/2016	29.6	16.4	49	0	2.2	14	320
10/5/2016	29.4	14.8	47	0	2.1	20.4	290
10/6/2016	30	12.2	39	0	2.2	15.8	280
10/7/2016	29.4	14.3	48	0	1.4	18.8	40
10/8/2016	29.1	18.7	58	0	1.3	10.3	40
10/9/2016	30.1	18	56	0	1.4	22.2	20
10/10/2016	30.1	17.7	57	0	1.4	13.8	80
10/11/2016	30.7	8.7	35	0	2.5	25.7	300
10/12/2016	30	14.8	49	0	1.7	28.8	310
10/13/2016	30.2	17	50	0	3.2	25.1	320
10/14/2016	29.4	20	60	0	1.4	20.8	30
10/15/2016	30	13.3	44	0	1.6	15	300
10/16/2016	29.8	15.4	50	0	2.5	19.2	330
10/17/2016	28.4	18.7	58	0	1.8	10	70
10/18/2016	29	19.6	60	0	1.9	18.4	70
10/19/2016	29.4	19.3	58	0	0.8	28	100
10/20/2016	30	13.7	42	0	2	33.7	300
10/21/2016	29.5	15.3	49	0	2.6	12.8	330
10/22/2016	28.1	19.4	61	0	1.8	12.7	20
10/23/2016	27.9	20	65	0	1.4	10.1	0
10/24/2016	27.9	21.2	70	0	1.4	31.6	40
10/25/2016	27.9	21.4	72	0	1	28	100
10/26/2016	28.7	23.3	73	0	2	30.6	100
10/27/2016	28.4	20.9	66	0	1.6	15.3	60
10/28/2016	28.5	18.1	58	0	1	11.9	50
10/29/2016	28.2	13.6	48	0	1.3	14.6	310
10/30/2016	27.6	15.1	50	0	1.5	28	320
10/31/2016	27.8	19.4	64	0	0.9	8.9	80
11/1/2016	27.6	22	73	0	1.3	18.4	100
11/2/2016	28.7	21.2	65	0	2.2	9.8	100
11/3/2016	28.6	20.8	64	0	1.7	9.8	20
11/4/2016	27.2	17.1	57	0	2.9	7.6	330
11/5/2016	27.1	18.2	62	0	2.3	6.9	320
11/6/2016	27.8	15.7	58	0	0.7	3.9	360

Appendix A: Meteorological Data

11/7/2016	26.9	17.2	57	0	2.8	10.1	350
11/8/2016	25.1	10.6	42	0	4	11.9	330
11/9/2016	25	13.8	51	0	4.1	8.9	330
11/10/2016	25.3	14.8	54	0	3.3	8.4	330
11/11/2016	24.6	17.4	66	0	1.5	5	0
11/12/2016	25.1	18.8	70	0	1	5.6	30
11/13/2016	24.7	18.4	71	0	0.9	4.6	50
11/14/2016	24.4	18.4	73	0	0.6	3.2	350
11/15/2016	24.9	17.9	69	0	1.1	4.9	300
11/16/2016	24.5	15.6	63	0	0.5	3.5	90
11/17/2016	25	19.2	71	0	1.3	5.3	110
11/18/2016	24.7	17.5	66	0	0.6	4.3	50
11/19/2016	24.6	16.1	64	0	2.9	6.9	310
11/20/2016	24.5	16.3	62	0	3.6	8.2	320
11/21/2016	24	14.7	58	0	2.1	5.9	330
11/22/2016	23.8	16.7	65	0	0.9	5.3	30
11/23/2016	23.9	18.7	74	0	1	6.4	30
11/24/2016	25.2	17.9	65	0	2.1	7.1	60
11/25/2016	23.9	17.8	69	4.1	3.4	10.2	70
11/26/2016	22	18.6	81	3.6	2.7	8.2	40
11/27/2016	22.8	19.5	82	14.1	2.2	11.6	30
11/28/2016	23	18.6	77	0.1	2.6	6	330
11/29/2016	22.9	16.8	70	0	2.9	8.2	320
11/30/2016	23.3	16.6	67	0	1.8	4.7	340
12/1/2016	24.8	18.1	68	0	1.8	5.2	100
12/2/2016	25.6	20.6	76	0	2.5	6.6	120
12/3/2016	25.9	19.9	72	0	2.2	5.6	320
12/4/2016	24.7	18.7	71	0	4.2	9.5	330
12/5/2016	24	18.4	73	0	3.2	7.7	330
12/6/2016	23.7	20.3	82	0	1.5	3.9	350
12/7/2016	23.3	16.8	68	0	3.2	8.2	330
12/8/2016	21	13.2	63	0	4	8.3	310
12/9/2016	20.2	11	57	0	3.8	9.4	320
12/10/2016	17.3	5.4	49	0	4.5	10.8	310
12/11/2016	16.8	4.6	45	0	4.4	8.5	310
12/12/2016	17.6	6.4	48	0	4.6	9.6	310
12/13/2016	17.9	8.2	54	0	1.9	7.4	310
12/14/2016	19.8	11.7	61	0	1.8	6.2	130
12/15/2016	22.1	12.3	57	0	3.3	7.8	170
12/16/2016	18.4	9.6	58	0	3.8	9	330
12/17/2016	18.2	10.6	63	0	3.8	8.2	320

Appendix A: Meteorological Data

12/18/2016	18.7	11.5	64	0	3.7	9.3	320
12/19/2016	19.2	12	65	0	2.6	6.4	320
12/20/2016	20.2	12.4	63	0	1.6	4.7	210
12/21/2016	19.7	13	66	0	3.1	7.8	340
12/22/2016	20.2	13.3	66	0	2.2	6.4	340
12/23/2016	21.4	14.3	66	0	2.9	9.8	330
12/24/2016	21.1	16	74	0	2.1	5.6	350
12/25/2016	21.2	18.1	84	0	1.3	5.2	100
12/26/2016	20.3	17.9	87	0	1.3	3.9	80
12/27/2016	19.7	17.5	88	0	1	3.2	80
12/28/2016	21.8	14.6	68	0	1.7	4.9	290
12/29/2016	21	16.3	76	0	2.8	7.5	330
12/30/2016	21.7	15.6	72	0	1.9	4.7	330
12/31/2016	21.8	12.5	60	0	1.6	4.6	280
1/1/2017	23	11.4	55	0	2.4	5.2	300
1/2/2017	21.2	14.7	70	0	1.8	5.1	300
1/3/2017	20.2	14.1	71	0	1.8	4.1	310
1/4/2017	20.2	12	61	0	3.1	7.1	320
1/5/2017	18.4	9.6	58	0	5	10.8	320
1/6/2017	17.2	9.7	62	0	4.4	10.5	310
1/7/2017	18	11.8	69	0	1.8	4.7	320
1/8/2017	19.1	9.4	60	0	1.3	4	160
1/9/2017	20.8	13.4	66	0	2.2	5.9	130
1/10/2017	20.4	16.1	78	0	1.5	4.2	160
1/11/2017	19.4	14.3	75	0	1.6	4.6	310
1/12/2017	20.1	11.7	62	0	3.3	8.1	320
1/13/2017	18.5	9	56	0	3.7	8.3	310
1/14/2017	17.5	6	49	0	3.9	10.8	320
1/15/2017	17.6	9.6	61	0	1.8	4.5	310
1/16/2017	18.2	13	73	0	1.9	4.7	80
1/17/2017	18.4	10.7	62	0	3.2	7.5	330
1/18/2017	17.7	10.8	65	0	1.9	5.6	350
1/19/2017	19.1	13.7	71	0	2.2	5.9	110
1/20/2017	21	13.3	63	0	2.4	6.9	150
1/21/2017	19.5	15.1	77	0	1.5	5.6	310
1/22/2017	20.7	17	80	0	2	5.8	90
1/23/2017	19.3	9.6	55	0	4.2	9.1	340
1/24/2017	16.2	5.2	49	0	4.2	10.1	310
1/25/2017	17.1	8.4	58	0	2.6	7.3	320
1/26/2017	18.3	11.7	67	0	1.6	5.6	120
1/27/2017	20.2	12.1	62	0	3.3	9.9	150

Appendix A: Meteorological Data

1/28/2017	22.1	14.3	63	0	3.3	7.4	130
1/29/2017	20.5	12.8	62	0	3.5	9.6	350
1/30/2017	16.8	6.2	51	0	3.9	9.3	340
1/31/2017	17.2	10.8	67	0	1.6	4.8	80
2/1/2017	20.2	13.3	65	0	3.4	6.9	50
2/2/2017	15.2	3.2	45	0	5.8	11.8	340
2/3/2017	10.8	-2.4	42	0	6.2	14.7	310
2/4/2017	13	3.1	52	0	3.9	8.8	320
2/5/2017	13.5	2.9	50	0	2.1	6.6	310
2/6/2017	14.8	5.5	58	0.6	2.9	7.4	310
2/7/2017	15.6	5.9	54	0	3.5	9.3	310
2/8/2017	16.8	9.1	63	0	2.1	5.9	30
2/9/2017	18.2	13.4	74	0	1.5	6.2	90
2/10/2017	20	15	74	0	2.6	6.6	110
2/11/2017	19.6	14.6	74	0.3	3.1	8.7	50
2/12/2017	21.2	15.6	71	0.1	5.9	13.5	90
2/13/2017	21.1	16.9	77	4.2	3.3	11.2	110
2/14/2017	18.4	16.5	89	19.2	2.9	7.8	10
2/15/2017	20.4	18	86	7.4	4.2	10.1	100
2/16/2017	20.9	17.8	83	14.2	3.6	8.3	140
2/17/2017	19.5	16.2	81	11.7	3.8	9.1	360
2/18/2017	14.5	6.5	59	0	4.3	10.6	320
2/19/2017	12.6	4.7	59	0	4.3	11.3	320
2/20/2017	13.8	7	65	2.3	3.1	7.7	300
2/21/2017	14.9	7.7	63	0.6	2.4	6.1	300
2/22/2017	17.3	11.5	71	0	1.3	5	350
2/23/2017	18.5	14.2	77	0.2	2.3	6.2	70
2/24/2017	19.6	14.2	73	0.1	3.5	8.8	350
2/25/2017	19.6	12.9	67	0.8	2.1	8.5	350
2/26/2017	17.8	6.6	52	0.7	2.1	6	290
2/27/2017	18.3	6.2	51	0	2	6.3	300
2/28/2017	18	8	56	0	1.4	4.8	340

Appendix A: Meteorological Data

QMD (standing) Daily Average of Some Meteorological Data At Alwakrah								
Date	Mean Temperature degree celsius	Mean Dew Point degree celsius	Mean Relative Humidity percentage	Total Rain Fall millimetes	Mean Wind Speed meters per second	Maximum Wind Gust meters per second	Mean Wind Direction 360 degree rose	
6/1/2016	33.2	7.8	23	0	6.2	13.7	320	
6/2/2016	34.1	4.3	18	0	5.4	12	300	
6/3/2016	32.6	14.9	39	0	2.5	8	130	
6/4/2016	31.9	16.7	44	0	4.6	10.1	10	
6/5/2016	31.1	17.2	45	0	2.3	6.5	50	
6/6/2016	31.1	20.9	55	0	2.6	5.7	120	
6/7/2016	32.5	18.1	46	0	2.5	6.2	160	
6/8/2016	36.8	10.9	22	0	6.6	12.8	330	
6/9/2016	37.7	5.9	16	0	7.5	13.4	330	
6/10/2016	35.2	15.6	35	0	5.1	11.2	350	
6/11/2016	36.5	13.9	29	0	5.9	12.3	340	
6/12/2016	35.9	9.8	21	0	8.2	15.7	350	
6/13/2016	34.8	5.9	18	0	7.3	13	320	
6/14/2016	34.4	1.4	16	0	7.2	13.6	310	
6/15/2016	33	14.7	40	0	6	12	340	
6/16/2016	31.9	22.1	58	0	3.4	6.4	20	
6/17/2016	32.8	19.3	46	0	5.8	11	0	
6/18/2016	33.9	13.8	31	0	6.5	10.8	360	
6/19/2016	34.1	17.8	40	0	6.1	10.5	360	
6/20/2016	36.3	14.5	29	0	7	12.4	360	
6/21/2016	34.9	16.3	35	0	6.5	14.5	360	
6/22/2016	33.8	19.3	45	0	4.8	8.7	10	
6/23/2016	33	20	49	0	3	7.1	20	
6/24/2016	32.7	21.9	54	0	2.9	6.4	70	
6/25/2016	32.9	21.6	53	0	2.9	6.8	50	
6/26/2016	33.5	23.5	57	0	2.2	6.2	100	
6/27/2016	36.4	16.8	35	0	3.2	8.7	350	
6/28/2016	34.4	22.4	52	0	3.7	8.7	20	
6/29/2016	32.8	25.5	66	0	2.8	6.4	90	
6/30/2016	33.5	24.1	59	0	2.2	5.8	110	
7/1/2016	39.2	14.4	25	0	6.4	13.7	330	
7/2/2016	38.4	11.7	22	0	7	13.4	320	
7/3/2016	37.1	13.3	25	0	7.4	15.5	340	
7/4/2016	34.8	20.9	47	0	4.6	10.7	0	
7/5/2016	34.6	21.6	52	0	3.5	8.7	340	
7/6/2016	33.6	20.4	48	0	3.4	7	90	

7/7/2016	34.7	19.7	46	0	3.5	8.4	360
7/8/2016	35.9	17.3	36	0	4	9	350
7/9/2016	39.4	9.9	19	0	5.6	12.3	320
7/10/2016	40	7.5	17	0	6.5	15.3	310
7/11/2016	39	10.3	19	0	7.3	14.7	340
7/12/2016	38.2	6.9	18	0	6.4	17	310
7/13/2016	37.4	11.4	23	0	5.4	11.6	320
7/14/2016	34.2	23	56	0	3.8	6.4	0
7/15/2016	33.9	26.6	67	0	3.6	8.1	40
7/16/2016	34.1	25.8	64	0	4.2	9.7	10
7/17/2016	34.6	25.5	62	0	3	9.5	80
7/18/2016	34.5	28.3	71	0	3.7	7.5	70
7/19/2016	35.6	28.5	67	0	3.8	6.3	100
7/20/2016	35.9	27.8	64	0	3.7	7.5	130
7/21/2016	35	27.7	67	0	4.6	8.9	90
7/22/2016	34.8	28.1	69	0	3.1	6.1	100
7/23/2016	35.6	26.7	63	0	4.7	8.9	360
7/24/2016	34.2	28.4	72	0	4.8	7.9	40
7/25/2016	35.4	28.8	69	0	4.1	9.2	100
7/26/2016	35	29.9	75	0	4.5	9.1	80
7/27/2016	34.6	28.8	72	0	4.3	7.1	80
7/28/2016	34.7	27.2	65	0	5	9	110
7/29/2016	34.9	26.9	64	0	4.4	7.6	110
7/30/2016	34.5	26.5	63	0	3	7.4	110
7/31/2016	34.7	27	65	0	3.6	6.2	90
8/1/2016	34.5	27.1	66	0	3.2	6.2	110
8/2/2016	36.2	24.2	53	0	3.4	6.7	130
8/3/2016	36.4	24.9	56	0	4.1	6.9	110
8/4/2016	35.4	27.8	65	0	4.3	8.7	60
8/5/2016	35	29.1	72	0	4.5	7.7	20
8/6/2016	35.3	26.1	60	0	3.2	9.2	40
8/7/2016	35.3	26.7	63	0	3.5	7.8	90
8/8/2016	35.1	27.1	65	0	3.6	6.4	70
8/9/2016	35	26.4	63	0	3.1	6	40
8/10/2016	35.2	24.7	58	0	3.9	7.6	20
8/11/2016	34.9	26.6	64	0	4.1	7.6	90
8/12/2016	35.2	28.9	70	0	3.8	9.7	70
8/13/2016	34.9	29.9	76	0	3.7	7.9	50
8/14/2016	35.8	27.3	63	0	3.3	5.6	130
8/15/2016	35.9	25	55	0	2.9	6.4	180
8/16/2016	35.7	25	55	0	3.1	7.9	10

8/17/2016	36	25.1	55	0	6.3	11.7	360
8/18/2016	35.7	25.4	56	0	3.7	6.9	50
8/19/2016	35.3	27.3	64	0	4.2	7.9	50
8/20/2016	35	26.5	62	0	3.3	6.7	50
8/21/2016	34.2	26.8	66	0	2.9	5.7	90
8/22/2016	34.7	28	69	0	3.7	8.2	120
8/23/2016	35.2	27.7	66	0	3.7	6.7	100
8/24/2016	35.9	25.8	58	0	4.6	12.5	100
8/25/2016	35.1	26.7	63	0	4.4	7.5	80
8/26/2016	35.3	26.8	62	0	3	5.6	90
8/27/2016	34.7	27.9	68	0	3.1	7.2	110
8/28/2016	34.7	26.4	63	0	4	8.5	90
8/29/2016	35.1	26.6	62	0	4	9.6	70
8/30/2016	35.3	26.2	60	0	3.8	8.2	100
8/31/2016	34.9	25.2	59	0	3.1	5.7	110
9/1/2016	35.6	24.3	57	0	3.8	6.4	170
9/2/2016	35.4	26.3	61	0	4	9.1	150
9/3/2016	34.8	28.3	69	0	4	7.1	70
9/4/2016	34.5	27.1	66	0	3.8	7.3	40
9/5/2016	35.3	26.5	61	0	4.1	8.9	10
9/6/2016	35.7	23.7	51	0	5.5	12.4	350
9/7/2016	34.8	22.9	51	0	7.1	12.9	0
9/8/2016	33.7	22.8	53	0	3.8	8.2	20
9/9/2016	32.8	21.6	52	0	3.3	7.3	60
9/10/2016	33.1	24.1	60	0	3.7	7.2	60
9/11/2016	32.2	24.1	63	0	2.6	7.1	40
9/12/2016	31.7	24.7	67	0	3.2	6.6	90
9/13/2016	32.4	24.7	65	0	3.4	7.1	80
9/14/2016	33.5	27.6	72	0	3.5	6.4	50
9/15/2016	33.3	26.3	67	0	2.9	5.3	60
9/16/2016	33.1	22.7	55	0	3	7.3	70
9/17/2016	33	22.8	57	0	2.5	4.8	100
9/18/2016	33.3	24.6	62	0	3.4	6.6	40
9/19/2016	33.9	23.4	55	0	3.7	8.6	10
9/20/2016	34	15.2	34	0	6	13	340
9/21/2016	32.4	21	51	0	5.9	10.7	350
9/22/2016	31.7	24.1	64	0	2.9	5.4	30
9/23/2016	32.6	22.7	58	0	4.3	8	330
9/24/2016	33.1	19.4	45	0	5.1	10	330
9/25/2016	33.7	15.9	37	0	6	12.7	320
9/26/2016	31.2	21.4	57	0	3.8	7.1	350

Appendix A: Meteorological Data

9/27/2016	30.5	24.6	71	0	2.6	6.3	60
9/28/2016	31.8	20.9	54	0	2.8	7	340
9/29/2016	32.9	17.6	43	0	6.2	12.9	330
9/30/2016	31	14	38	0	6.6	13	310
10/1/2016	29.8	16.6	46	0	5.7	10.1	350
10/2/2016	29	17	49	0	4.4	10.6	20
10/3/2016	28.7	16.3	48	0	4.8	9.3	350
10/4/2016	28.6	17.7	53	0	3.8	8.6	340
10/5/2016	28.1	16.5	52	0	2.8	6.7	320
10/6/2016	28.2	16.6	51	0	3.1	5.7	230
10/7/2016	29.5	16.9	50	0	3.8	6.6	30
10/8/2016	28.3	20	62	0	3	6.7	30
10/9/2016	29.2	20.7	62	0	2.6	5.8	80
10/10/2016	29	21.1	65	0	3	6	80
10/11/2016	29.8	12.7	42	0	3.3	9.7	290
10/12/2016	29	18.3	55	0	2.7	7.3	310
10/13/2016	29.6	18.8	54	0	4.5	9.8	340
10/14/2016	29	20.7	62	0	3.7	7.9	10
10/15/2016	28.8	17.9	56	0	2.6	6.7	250
10/16/2016	29.3	18.5	55	0	4.9	10.3	340
10/17/2016	28.8	19.7	59	0	3	5.8	60
10/18/2016	29.3	22.3	66	0	3.4	5.7	40
10/19/2016	29.7	21.4	62	0	2.4	4.9	90
10/20/2016					i i		
10/21/2016					1		
10/22/2016							
10/23/2016	27.9	21.1	67	0	3	6.2	50
10/24/2016	27.7	21.8	71	0	3.2	5.2	0
10/25/2016	27.8	22.7	74	0	2.9	5.4	90
10/26/2016	29.2	23.7	72	0	3.7	6.3	90
10/27/2016	29.4	22.8	68	0	4.4	6.8	40
10/28/2016	28.8	20.9	63	0	3.3	5.9	30
10/29/2016	27.5	15.1	50	0	2.9	6.9	320
10/30/2016	27.5	16.2	54	0	3	7.2	310
10/31/2016	27.7	20.5	67	0	3.4	6.5	50
11/1/2016	28.7	21.9	67	0	3.1	5.9	80
11/2/2016	29.1	21.5	64	0	4.1	6.9	90
11/3/2016	28.5	21.7	67	0	4.2	9.2	40
11/4/2016	27	17.8	58	0	6	9.3	340
11/5/2016	27.2	16.6	57	0	4.2	8.6	300
11/6/2016	27.4	19.6	64	0	2.5	4.9	110

11/7/2016	26.6	18	60	0	5.4	12.8	350
11/8/2016	25.4	10.1	39	0	6.5	13.1	320
11/9/2016	25	12.7	48	0	5.7	9.9	310
11/10/2016	24.9	15.6	57	0	5.8	11.2	340
11/11/2016	25	17.9	65	0	4	7.5	350
11/12/2016	25.3	19.2	70	0	3.1	5.7	10
11/13/2016	24.7	18.8	70	0	2.5	5.3	30
11/14/2016	23.7	17.4	70	0	2.4	5	300
11/15/2016	23.7	19	76	0	2.4	5.5	250
11/16/2016	24.1	18.2	70	0	2.3	4.3	130
11/17/2016	26.1	19.1	65	0	3.5	4.9	120
11/18/2016	24.8	17.3	63	0	2.3	5.6	60
11/19/2016	24.1	14.7	60	0	5	8.9	300
11/20/2016	24.5	15.3	59	0	5.7	10	310
11/21/2016	24.1	15.9	61	0	4.5	8.1	340
11/22/2016	24.4	17.8	67	0	4	8.9	360
11/23/2016	25	19.2	70	0	3.1	7	40
11/24/2016	25.6	18.8	66	0.1	5.1	11.2	100
11/25/2016	24.2	18.3	70	0.5	7.1	12.8	80
11/26/2016	23.7	19.3	77	18	6.5	11.2	40
11/27/2016	24.2	19.2	74	11.6	3.4	15.3	40
11/28/2016	23.1	17.7	72	0	4.6	8.5	330
11/29/2016	22.8	15.5	64	0	5.2	9.8	310
11/30/2016	23.1	16.4	66	0	3.6	7.3	340
12/1/2016	24.1	18.6	72	0	3.2	5.8	120
12/2/2016	25	22.2	84	0	3.1	5.3	130
12/3/2016	25	20.9	79	0	3.6	8.7	350
12/4/2016	24.8	18	67	0	6	10.9	320
12/5/2016	24.2	17.9	69	0	4.9	10	340
12/6/2016	23.5	19.9	80	0	2.1	5.3	30
12/7/2016	22.8	16.9	70	0	5.5	9.7	330
12/8/2016	20.8	12.5	61	0	5.6	10	300
12/9/2016	20.7	10.4	53	0	5.6	10.3	310
12/10/2016	17.3	3.7	44	0	6.8	12.1	310
12/11/2016	16.5	3.5	43	0	5.9	11.2	300
12/12/2016	17.2	5.7	47	0	5.9	11	300
12/13/2016	17.4	8.1	55	0	2.9	6.6	310
12/14/2016	19.5	13.3	68	0	3.6	7.4	140
12/15/2016	21.1	14.6	67	0	5.3	9.8	180
12/16/2016	18.5	9.3	56	0	6	10.7	330
12/17/2016	18	9.9	60	0	5.8	9.3	310

Appendix A: Meteorological Data

12/18/2016	18.6	10.7	61	0	5.8	10.5	310
12/19/2016	18.9	11.8	64	0	4.1	8.1	320
12/20/2016	19.1	13.4	70	0	2.2	5.6	130
12/21/2016	19.3	13.2	68	0	6.1	11.4	340
12/22/2016	19.4	13.3	68	0	4.1	8.9	340
12/23/2016	20.6	14.7	69	0	5	9.9	330
12/24/2016	20.9	16.3	76	0	3.8	8	350
12/25/2016	19.2	17.5	90	0	1.4	3.6	90
12/26/2016	18.4	17.3	94	0	1.6	4.5	120
12/27/2016	17.7	8.4	55	0	1.4	3.1	40
12/28/2016	19.1	12.8	66	0	2.9	6.8	240
12/29/2016	20.6	16.1	76	0	4.5	8.7	340
12/30/2016	21.4	14.8	68	0	3.3	6.6	350
12/31/2016	21.4	12.6	61	0	2.9	6	280
1/1/2017	21.5	12	59	0	3.2	6.9	300
1/2/2017	20.3	16.1	77	0	2.5	6.3	280
1/3/2017	19.8	15.2	76	0	2.9	6.9	310
1/4/2017	19.6	12.1	62	0	5.2	9	320
1/5/2017	18.1	8.8	56	0	6.7	11.8	310
1/6/2017	16.8	8.9	60	0	6.5	11.1	310
1/7/2017	17.8	11.6	68	0	3.6	6.1	330
1/8/2017	19	11.6	63	0	2.6	4.8	150
1/9/2017	20.1	15.8	77	0	3.1	5.1	140
1/10/2017	19.8	17.2	85	0	2.4	4.2	140
1/11/2017	19.3	14.5	74	0	2.6	6.2	340
1/12/2017	19.9	10.4	57	0	5.7	9.5	300
1/13/2017	18.6	8.4	53	0	5.5	8.7	300
1/14/2017	17.6	5.4	46	0	5.8	12	320
1/15/2017	17.2	9,8	62	0	3,4	6.4	300
1/16/2017	18.5	13.5	73	0	2.8	5.2	80
1/17/2017	18.2	10.9	63	0	5.3	9.9	330
1/18/2017	17.6	10.9	66	0	3.6	8.4	350
1/19/2017	18.9	14	73	0	3.3	7	120
1/20/2017	19.9	14.7	73	0	3.6	7.4	150
1/21/2017	19.5	14.8	75	0	2.9	6.1	310
1/22/2017	20.6	17.5	82	0	3.5	7.1	90
1/23/2017	19.2	9.8	55	0	6.5	12.3	340
1/24/2017	15.9	4.3	47	0	6.6	11.3	310
1/25/2017	16.7	8.3	59	0	4.5	8.5	320
1/26/2017	18.3	12.2	68	0	3.1	6.2	110
1/27/2017	19.7	13.9	69	0	6.3	8.7	140

Appendix A: Meteorological Data

		And a second					
1/28/2017	20.6	16.1	76	0	5	8.1	120
1/29/2017	20	13.4	67	0	4.6	12	0
1/30/2017	16.6	6.2	51	0	6.4	12.7	340
1/31/2017	16.9	11.9	72	0	3.6	5.9	110
2/1/2017	19.6	14.7	74	0	4.3	10.9	40
2/2/2017	15.3	3.4	45	0	8.3	15.9	340
2/3/2017	10.8	-3.7	39	0	8,8	18.5	310
2/4/2017	13.1	2.5	49	0	6.3	10.9	300
2/5/2017	13.4	2.6	49	0	3.3	7.5	310
2/6/2017	14.3	5.8	58	0.3	4.9	9.5	320
2/7/2017	15.6	4.6	50	0	6.1	10.3	310
2/8/2017	17.3	11.1	68	0	3.5	10	20
2/9/2017	18.2	14.8	81	0	2.7	7.1	50
2/10/2017	18.7	16	84	0.1	2.4	4.1	110
2/11/2017	19.2	15.3	78	0.4	5.1	10	40
2/12/2017	20.1	16.5	80	0.2	7.5	15.4	80
2/13/2017	20.5	16.6	79	2	5	12.3	100
2/14/2017	18.7	16.3	86	7.1	5.8	11.1	20
2/15/2017	19.7	17.4	87	3.7	5.3	9.8	80
2/16/2017	19.9	17.3	85	8.9	5.4	13.3	120
2/17/2017	19.3	15.4	78	5	5.6	13.7	20
2/18/2017	14.6	5.3	54	0	6.2	12.5	320
2/19/2017	12.8	3.5	54	0	6.7	10.7	310
2/20/2017	13.5	5.7	60	1.9	5.3	9.8	300
2/21/2017	14.5	6.6	61	1.4	3.5	7	290
2/22/2017	16.7	12.1	74	0	2.6	6.1	50
2/23/2017	18.4	14.9	80	1.2	3.7	6.7	50
2/24/2017	19.1	14.2	74	0	6	12	360
2/25/2017	18.8	13.2	70	1.6	3.7	10.3	10
2/26/2017	17.4	6.9	52	0.1	2.8	9.8	330
2/27/2017	18.3	6.5	50	0	3.5	7.7	300
2/28/2017	17.5	10.4	64	0	3.3	6.7	350

Appendix A: Meteorological Data

Date	Mean Temperature (degree celsius)	Mean Dew Point (degree celsius)	Mean Relative Humidity (percentage)	Total Rain Fall (millimeters)	Mean Wind Speed meters per second	Mean Wind Direction 360 degree rose
6/1/2016	33.9	5.5	20	0	3.9	320
6/2/2016	35.6	0.4	14	0	3	300
6/3/2016	35.9	3.3	14	0	1.9	0
6/4/2016	34.9	9.2	22	0	2.8	10
6/5/2016	34.6	7.6	22	0	1.6	40
6/6/2016	35.5	10.1	25	0	1.5	120
6/7/2016	36	9.6	24	0	1.9	160
6/8/2016	38.4	5	16	0	4	330
6/9/2016	37.5	4.1	16	0	4.8	320
6/10/2016	37.7	10.8	22	0	3.2	350
6/11/2016	39.3	8	17	0	4	340
6/12/2016	36.9	5.4	17	0	5.1	340
6/13/2016	35	5.1	18	0	4.7	320
6/14/2016	35.2	-1.2	14	0	4.6	310
6/15/2016	35.9	7.7	21	0	3.5	330
6/16/2016	35.6	12.4	28	0	2	10
6/17/2016	35.8	11.6	26	0	3.7	350
6/18/2016	36.1	7.7	20	0	3.8	350
6/19/2016	36.4	13.1	28	0	3.9	350
6/20/2016	38.4	7.7	18	0	4	340
6/21/2016	38.2	8	19	0	3.9	340
6/22/2016	36.9	12.6	24	0	3.1	10
6/23/2016	36.5	12.4	25	0	2.3	360
6/24/2016	36.8	12.5	25	0	2	30
6/25/2016	36.5	12.8	28	0	1.8	60
6/26/2016	37	12.5	27	0	1.3	70
6/27/2016	40.5	4.7	16	0	2.2	330
6/28/2016	37.9	15.7	30	0	2.6	20
6/29/2016	35.8	20.1	42	0	1.5	70

6/30/2016	36.8	19.4	40	0	1.8	80
7/1/2016	40.8	9.3	18	0	4.1	330
7/2/2016	38.7	9.7	20	0	4.9	320
7/3/2016	38.4	9.5	20	0	4.6	330
7/4/2016	39.2	8.8	18	0	3.1	350
7/5/2016	37.5	15.2	33	0	2.5	320
7/6/2016	37.5	11.3	24	0	2.2	340
7/7/2016	37.9	12.4	26	0	2.5	330
7/8/2016	39.5	10.4	21	0	2.6	340
7/9/2016	40.8	7.3	16	0	3.1	320
7/10/2016	40.4	4.9	15	0	4.4	310
7/11/2016	40	6.9	16	0	4	330
7/12/2016	38.6	5.2	16	0	4.4	310
7/13/2016	38.9	6.6	17	0	3.4	310
7/14/2016	37.7	15.9	33	0	2.3	0
7/15/2016	37.4	19.1	37	0	2	10
7/16/2016	36.7	21.8	44	0	2.7	0
7/17/2016	38.1	19.4	38	0	2.2	20
7/18/2016	37.3	23.8	49	0	2.3	80
7/19/2016	38.9	21.7	40	0	2.2	110
7/20/2016	39.8	18.9	34	0	1.8	150
7/21/2016	37.7	22.4	43	0	2.3	90
7/22/2016	38	23.6	47	0	1.9	80
7/23/2016	38.5	22	42	0	2.5	350
7/24/2016	36.3	25.4	55	0	2.7	30
7/25/2016	37.5	26.4	55	0	2.5	100
7/26/2016	37.2	26.7	56	0	2.6	80
7/27/2016	36.3	26	57	0	2.7	80
7/28/2016	35.9	25.5	55	0	2.7	100
7/29/2016	36.4	25.1	53	0	2.6	110
7/30/2016	36	24.4	52	0	2	90
7/31/2016	36.3	24.1	51	0	1.9	100
8/1/2016	37.3	22	45	0	1.7	90
8/2/2016	39	17.5	35	0	1.6	250
8/3/2016	38.7	20.2	40	0	2.5	60

Appendix A: Meteorological Data

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8/4/2016	37.1	25	51	0	2.6	50
8/5/2016	36.9	26.2	56	0	2.5	20
8/6/2016	37.4	22.2	45	0	1.8	30
8/7/2016	37.3	22.7	47	0	2	90
8/8/2016	37.6	21	46	0	1.8	70
8/9/2016	37.3	23.2	47	0	1.8	10
8/10/2016	37.4	21	44	0	2.4	10
8/11/2016	36.7	24.1	52	0	2.2	80
8/12/2016	36.7	26.8	58	0	2.2	70
8/13/2016	37.4	25.9	56	0	2.2	60
8/14/2016	38.1	22.5	47	0	1.8	110
8/15/2016	39	19.9	37	0	1.3	200
8/16/2016	38.7	21	39	0	1.9	340
8/17/2016	38.3	20.9	39	0	3.4	340
8/18/2016	37.9	19.7	38	0	2.1	60
8/19/2016	37	23.8	49	0	2.3	40
8/20/2016	37.1	23.7	49	0	1.9	50
8/21/2016	36.2	24.3	52	0	1.7	70
8/22/2016	36.1	25.8	57	0	2.1	110
8/23/2016	36.7	23.5	52	0	2.3	100
8/24/2016	37.9	20.3	43	0	1.9	110
8/25/2016	37	22.1	48	0	2.1	80
8/26/2016	37.8	21.3	45	0	1.9	90
8/27/2016	37.1	23.7	49	0	1.8	100
8/28/2016	36.4	23.6	51	0	2.1	70
8/29/2016	37.1	21.6	45	0	2.3	70
8/30/2016	37.7	20.8	42	0	2	80
8/31/2016	36.9	20.4	44	0	1.9	90
9/1/2016	37.4	18.9	44	0	1.9	150
9/2/2016	37.7	20.5	42	0	2.3	130
9/3/2016	36	26.8	60	0	2.4	60
9/4/2016	36.4	24.7	52	0	2.1	50
9/5/2016	37.6	23	46	0	2.1	350
9/6/2016	37.5	20.8	40	0	3	350
9/7/2016	36.1	20.8	42	0	4	360

Appendix A: Meteorological Data
9/8/2016	35.4	19.6	41	0	1.8	10
9/9/2016	34.4	18.2	41	0	1.8	50
9/10/2016	34.2	22.5	51	0	2.1	60
9/11/2016	33.7	22.2	52	0	1.8	30
9/12/2016	33.4	23.2	57	0	1.9	70
9/13/2016	34	22.5	53	0	1.7	70
9/14/2016	35.2	25	58	0	1.8	60
9/15/2016	35	23.7	54	0	1.9	70
9/16/2016	34.7	20.1	45	0	1.5	50
9/17/2016	35	18.7	43	0	1.4	100
9/18/2016	35.6	20.4	48	0	2.2	350
9/19/2016	35.6	19.6	42	0	2.8	20
9/20/2016	35.2	11.7	28	0	3.7	330
9/21/2016	33.9	17.3	39	0	3.5	360
9/22/2016	33.4	21.6	53	0	1.7	20
9/23/2016	34.6	19.5	48	0	2.3	320
9/24/2016	34.8	17.2	38	0	2.7	340
9/25/2016	34.3	15.2	36	0	3.4	320
9/26/2016	33	19.4	46	0	1.9	360
9/27/2016	32.2	22.9	59	0	1.3	70
9/28/2016	34.6	16.7	38	0	1.7	340
9/29/2016	34.3	14.6	36	0	3.7	320
9/30/2016	31.5	13.6	36	0	4.2	310
10/1/2016	31.1	13.9	37	0	2.9	340
10/2/2016	30.7	12.7	35	0	2.5	10
10/3/2016	29.7	13.1	38	0	2	330
10/4/2016	30.4	15	42	0	2.1	330
10/5/2016	29.9	11.2	39	0	1.6	300
10/6/2016	30.8	8.9	32	0	1.9	280
10/7/2016	31.2	9.5	35	0	1.6	10
10/8/2016	30.6	15.4	44	0	1.5	10
10/9/2016	31.5	13.2	44	0	1.5	10
10/10/2016	31.4	13.6	45	0	1.5	70
10/11/2016	32.5	2.4	25	0	2.2	290
10/12/2016	32	9.7	34	0	1.9	310

Appendix A: Meteorological Data

10/13/2016	31.2	14.5	41	0	2.7	320
10/14/2016	30.5	17.4	48	0	2.3	10
10/15/2016	31.3	9.8	35	0	1.6	300
10/16/2016	31.1	13	41	0	2.8	320
10/17/2016	29.6	17.5	50	0	1.8	60
10/18/2016	30.7	17.4	48	0	2	60
10/19/2016	30.7	16.4	48	0	1.1	100
10/20/2016	31.3	11	33	0	1.8	310
10/21/2016	30.3	14.3	43	0	2.7	330
10/22/2016	29.6	17.7	50	0	2.1	360
10/23/2016	28.8	18.5	56	0	1.6	340
10/24/2016	29.3	19.8	60	0	1.9	20
10/25/2016	29	20.5	63	0	1.5	110
10/26/2016	29.9	22.3	65	0	2.2	90
10/27/2016	30.8	19	52	0	2.3	40
10/28/2016	30.5	15	45	0	1.9	10
10/29/2016	29.5	12	41	0	1.7	300
10/30/2016	29	13.9	43	0	1.5	330
10/31/2016	29.1	18.7	58	0	1.4	40
11/1/2016	29	21.4	64	0	1.3	90
11/2/2016	29.4	20.5	59	0	2	110
11/3/2016	29.6	20	58	0	2.2	20
11/4/2016	28.1	16.4	52	0	2.7	330
11/5/2016	28.1	17.1	56	0	2.5	320
11/6/2016	28.7	15	52	0	1.3	340
11/7/2016	27.3	16.7	54	0	3.1	340
11/8/2016	25.7	10	38	0	4.1	320
11/9/2016	25.6	13	47	0	3	320
11/10/2016	26.2	14.1	49	0	3	330
11/11/2016	26.3	16.7	57	0	2.1	0
11/12/2016	26.9	16.9	58	0	1.8	0
11/13/2016	25.7	17.6	64	0	1.4	360
11/14/2016	25.3	16.6	63	0	1.3	310
11/15/2016	25.7	17.6	64	0	1.2	300
11/16/2016	25.3	16.2	61	0	1	80

Appendix A: Meteorological Data

11/17/2016	25.8	18.2	65	0	1.4	130
11/18/2016	25.8	16.4	58	0	1	40
11/19/2016	25.1	14.7	58	0	2.3	310
11/20/2016	24.9	15.8	59	0	3.2	310
11/21/2016	25	14.5	54	0	2.3	330
11/22/2016	25	17.1	62	0	1.9	360
11/23/2016	25.6	18.9	67	0	1.6	10
11/24/2016	25.7	17.7	63	0	2.2	60
11/25/2016	24.6	17.4	64	0	3.1	70
11/26/2016	23	18.8	78	36.3	2.9	30
11/27/2016	23.9	19.3	76	3.9	1.7	20
11/28/2016	23.8	17.9	70	0	2.3	330
11/29/2016	23.6	16.4	65	0	2.6	320
11/30/2016	24.1	16.1	62	0	1.8	340
12/1/2016	25.1	17.8	66	0	1.5	120
12/2/2016	26	20.8	76	0	2	120
12/3/2016	26	20.4	74	0	1.9	330
12/4/2016	25.1	18.6	68	0	3.7	320
12/5/2016	24.7	18.2	70	0	3.1	320
12/6/2016	24.7	20.3	78	0	1.4	350
12/7/2016	25	18.1	67	0	2.1	330
12/8/2016	21.7	12.4	57	0	3	310
12/9/2016	21	10.5	53	0	3.2	320
12/10/2016	17.7	4.2	44	0	4	310
12/11/2016	17.2	3	40	0	3.4	300
12/12/2016	18	5.3	44	0	3.6	310
12/13/2016	18.7	7.2	48	0	1.6	310
12/14/2016	20.2	11.9	60	0	1.6	140
12/15/2016	22.1	12.2	56	0	3.1	160
12/16/2016	18.8	9.3	55	0	3.2	330
12/17/2016	18.6	10.2	59	0	3.1	320
12/18/2016	18.9	11.2	62	0	3.2	320
12/19/2016	19.8	11.6	60	0	2.1	310
12/20/2016	20.4	11.7	60	0	1.5	140
12/21/2016	20.3	12.9	63	0	3	340

Appendix A: Meteorological Data

12/22/2016	20.6	12.9	63	0	2.5	350
12/23/2016	21.9	13.7	63	0	2.6	330
12/24/2016	22.1	15.8	69	0	1.8	350
12/25/2016	21.6	18	81	0	1.2	100
12/26/2016	20.6	17.5	84	0	1.1	100
12/27/2016	19.7	17	86	0	0.9	100
12/28/2016	21.6	14.5	69	0	1.7	290
12/29/2016	22	16.2	71	0	2.7	330
12/30/2016	22.8	14.8	65	0	1.9	340
12/31/2016	23.2	11.3	53	0	1.3	300
1/1/2017	23.9	9.9	48	0	2.4	300
1/2/2017	22.1	14.5	65	0	1.6	310
1/3/2017	21.3	13.7	66	0	1.6	310
1/4/2017	20.3	11.6	59	0	2.8	320
1/5/2017	18.7	9.2	55	0	4.1	320
1/6/2017	17.5	9	58	0	3.5	310
1/7/2017	19.1	11.4	62	0	1.6	340
1/8/2017	19.6	8.7	55	0	1.1	70
1/9/2017	20.9	12.6	64	0	2	130
1/10/2017	21.1	16.6	78	0	1.3	140
1/11/2017	20.7	13.4	65	0	1.6	310
1/12/2017	20.5	10.8	57	0	3	320
1/13/2017	19.1	8.5	52	0	2.9	310
1/14/2017	18	5.6	46	0	3	320
1/15/2017	17.9	8.9	57	0	1.5	330
1/16/2017	19.6	12.8	66	0	1.4	80
1/17/2017	19.2	10.5	59	0	2.6	320
1/18/2017	18.6	10.8	62	0	1.5	350
1/19/2017	19.4	13.3	68	0	1.8	120
1/20/2017	21.7	12.6	59	0	2.1	140
1/21/2017	20.4	14.8	71	0	1.4	310
1/22/2017	21.6	16.6	73	0	2.1	90
1/23/2017	19.8	9.3	53	0	3.9	330
1/24/2017	16.6	4.6	46	0	3.7	310
1/25/2017	17.7	8.3	55	0	2.3	330

## Appendix A: Meteorological Data

1/26/2017	18.9	11.2	63	0	1.7	120
1/27/2017	20.3	11.9	61	0	2.8	140
1/28/2017	22.1	14.1	63	0	2.8	120
1/29/2017	21.4	12.8	59	0	3	340
1/30/2017	17.4	5.6	48	0	3.1	330
1/31/2017	17.8	10.6	64	0	1.6	90
2/1/2017	21	12.6	61	0	2.9	20
2/2/2017	15.6	2.7	42	0	5.3	330
2/3/2017	11.1	-3.9	38	0	5.4	310
2/4/2017	13.4	2.7	49	0	3.3	310
2/5/2017	14.5	1.9	44	0	2	310
2/6/2017	15.1	4.1	53	0.7	2.6	320
2/7/2017	16.3	4.8	49	0	3	320
2/8/2017	18.1	7.7	54	0	2	10
2/9/2017	19.2	13.3	70	0	1.6	90
2/10/2017	20.8	14.9	71	0	2	110
2/11/2017	20.4	14.5	69	0.3	3.2	40
2/12/2017	21.3	16.5	74	0.4	5.4	80
2/13/2017	21.9	17.2	75	2.9	3	100
2/14/2017	19.2	17.3	89	12.8	3.3	10
2/15/2017	20.8	18.4	87	6.2	3.3	80
2/16/2017	21	18.3	85	14.7	3.2	120
2/17/2017	19.8	16.1	80	11.5	3.3	10
2/18/2017	15	5.3	53	0	4.1	320
2/19/2017	13.2	3.8	54	0	3.5	320
2/20/2017	14.1	6	60	3.2	2.6	300
2/21/2017	15.4	6.6	58	1.8	2.2	290
2/22/2017	18.1	11.1	66	0	1.6	350
2/23/2017	19.4	14.9	76	0.3	2.4	60
2/24/2017	20.2	14.2	70	0	3.5	350
2/25/2017	20	12.6	63	1.6	2.2	340
2/26/2017	18.3	5.9	47	0.4	1.9	290
2/27/2017	19.8	3.7	40	0	1.9	300
2/28/2017	19.5	5.9	46	0	1.3	340

Appendix A: Meteorological Data

#### Appendix B: PM Calculations

$$Q_{act} = F_i + F_f / 2$$

Where:  $Q_{act} = Average$  sampling flow rate at field sampling conditions

 $F_i$  = Initial actual flow rate (m<sup>3</sup>/min)

 $F_f$  = Final actual flow rate (m<sup>3</sup>/min)

 $V_{act} = Q_{act} x$  Sampling period

Vact = Volume of sampled air  $(m^3)$ 

To calculate PM concentration:

 $TSP = (W_f - W_i) \times 10^6 / V_{act}$ 

Where:

TSP = mass concentration of total suspended particulate matter ( $\mu g/m^3$ )

 $W_f$  = Initial weight of clean filter (g)

 $W_i$  = Final weight of exposed filter (g)

 $10^6$  = Conversion from g to µg

To calculate actual PM concentration at field condition:

 $TSP_{std} = TSP_{act} (P_{std} / P_{act}) (T_{act} / T_{std})$ 

 $TSP_{act} = Actual \text{ concentration of PM at field conditions } (\mu g/m^3)$ 

 $TSP_{std} = Concentration at standard conditions (\mu g/m<sup>3</sup>)$ 

 $P_{act}$  = Average barometric pressure at the field during sampling (mm Hg)

 $P_{std} = 760 \text{ mm Hg}$ 

 $T_{act}$  = Average ambient temperature at the field conditions during the sampling period (K)

 $T_{std}\!=\!298~K$ 

THE ANOVA MODEL FITTED IS:  $Y_i = \beta_0 + \beta_1 \text{SAMP}_\text{DATE} + \beta_2 \text{STATION} + \beta_3 \text{SAMP}_\text{DATE} \text{STATION} + \epsilon_i$ where  $Y_i$  is the reading of the ith element The SAS System 14:45 Friday, October 6, 2017 1

The SAS System 14:45 Friday, Octo
The GLM Procedure
Class Level Information
Class Levels Values
Samp Date 4 1 2 3 4
Station 5 1 2 3 4 5
Number of observations 60
ing values, only 45 observations can be used in this analysis.

		The SA: The GLM	S System Procedure	14:45	Friday, Oct	tober 6, 2017	2
Dependent Variable: PM	25 PM25						
_		SI	um of				
Source	1	DF Sq	uares M	lean Square	F Value	Pr > F	
Model		19 22965 <b>.</b>	98953	1208.73629	3.86	0.0010	
Error	-	25 /833.	22551	313.32902			
Corrected Total		44 30799.	21503				
	R-Square 0.745668	Coeff Var 30.03388	Root MS 17.7011	SE PM25	Mean 93711		
Source	1	DF Type	ISS M	lean Square	F Value	Pr > F	
Samp_Date		3 7598.	74765	2532.91588	8.08	0.0006	
Station		4 1154.	69557	288.67389	0.92	0.4672	
Samp_Date*Statio	n :	12 14212.	54631	1184.37886	3.78	0.0024	
Source	1	DF Type I	II SS M	lean Square	F Value	Pr > F	
Samp Date		3 8255.	42007	2751.80669	8.78	0.0004	
Station		4 687.	86820	171.96705	0.55	0.7016	
Samp_Date*Statio	n i	12 14212.	54631	1184.37886	3.78	0.0024	
		The SA:	S System	14:45	Friday, Oct	tober 6, 2017	3
		Least Sour	res Means				
	Adjustment i	for Multiple (	Comparisons	: Tukey-Kra	amer		
	Samo I	Date PM25	L.SMEAN	Number			
	1	36.	8070000	1			
	2	66.	0682778	2			
	3	72.	8567062	3			
	4	54.	5246667	4			
	Least S Pr >	Squares Means  t  for H0: 1 Dependent V	for effect LSMean(i)=L	Samp_Date SMean(j)			
i/j	1	Dopondono V	2	3	4		
1		0.0	056	0.0003	0.1308		
2	0.005	56		0.7938	0.4621		
3	0.000	0.7	938		0.0783		
4	0.130	0.4	621	0.0783			
	The SAS S	ystem The GLM 1	14:45 Frid Procedure	ay, October	6, 2017	4	
		Least Squa	ares Means				
	Adjustment i	for Multiple (	Comparisons	: Tukey-Kra LSMEAN	amer		
	Stat	ion PM25	LSMEAN	Number			
	1	56.3	333828	1			
	2	55.4	754167	2			
	3	58.8	454167	3			
	4	64.2	321699	4			
	5	52.8	344274	5			
	Least Pr >	Squares Means	s for effec LSMean(i)=L	t Station SMean(j)			
		Dependent Va	ariable: PM	125			
i/j	1	2	3	4		5	
1	1 0000	1.0000	0.9983	0.87	/93 (	).9944	
2	1.0000	0.0011	0.9944	0.82	230 (	J.9980	
د .	0.9983	0.9944	0 0 0 2 7	0.96	os/ (	J. YOJX D. 6475	
4	0.0193	0.0230	0.963/	0.00	175	.04/3	
5	0.9944	0.3980	0.9038	0.64	± / J		

101

The SAS System 14:45 Friday, October 6, 2017 5

			The GLM	Procedure					
		C	lass Leve	l Informat:	lon				
		Class	Le	evels Va	alues				
		Samp_Da	te	4 1	234				
		Station		5 1	2345				
		Num	ber of obs	servations	60				
			The SA	AS System	14:	:45 Fi	riday, Octo	ber 6, 2017	6
			The GLM	Procedure					
Depe	ndent Variable: Al	Al							
			5	Sum of					
	Source	DF	S	quares	Mean Squa	are	F Value	Pr > F	
	Model	19	14340	50.142	75476.3	323	8.79	<.0001	
	Error	40	3435	67.439	8589.2	186			
	Corrected Total	59	17776	17.581					
						_			
		R-Square C	oeff Var	Root I	4SE	Al Me	ean		
		0.806726	34.88243	92.67	/86	265.6	863		
	-		_						
	Source	DF	Type	e I SS	Mean Squa	are	F Value	Pr > F	
	Samp_Date	3	1449	18.3/4	48306.	125	5.62	0.0026	
	Station	4	2755	01.076	68875.2	269	8.02	<.0001	
	Samp_Date*Statior	n 12	10136	30.692	84469.2	224	9.83	<.0001	
	0								
	Source	DF	Type .	III SS	Mean Squa	are	F Value	Pr > F	
	Samp_Date	3	1449	18.3/4	48306.	125	5.62	0.0026	
	Station	4	2/55	01.076	688/5.2	269	8.02	<.0001	
	Samp_Date^Station	1 12	10136	30.692	84469.4	224	9.83	<.0001	-
			The SA	AS System	14:	:45 F1	riday, Octo	ber 6, 201/	/
			ml ot M	<b>D</b>					
			The GLM	Procedure					
		7 -1	Least Squ	lares Mean:	3 	. 1			
		Adjustment	IOF MUIL	ipie compa	LISONS: TU	икеу			
		Comp. Dot	~ <sup>^</sup>	LICMEAN	LSMEAD	N			
		samp_bac	e A.	1 LOMEAN	NULLDEI	1			
		1	221	1 102222	-	1 2			
		2	22.	2 250667	4	2			
		3	212	2.230007	-	2			
		7	510	5.001555		1			
		Toast Sou	ares Mean	for offo	at Samp Da	- t o			
		Pr >  +	l for HO.	I.SMean(i):	=LSMean(i)	100			
		11 / 10	Dependent	Wariahle.		,			
	i/i	1	Dependenc	2	3		4		
	-, 5	-	0.0	0543	0.0289		0.9954		
	2	0.0543			0.9934		0.0312		
	- 3	0.0289	0.0	9934			0.0160		
	4	0.9954	0.0	0312	0.0160				
			The SA	AS System	14:	:45 F1	riday. Octo	ber 6, 2017	8
			The GLM	Procedure					
			Least Squ	uares Mean:	3				
		Adjustment	for Mult:	iple Compa:	risons: Tu	ukey			
		-			LSMEAN	-			
		Station	Al	LSMEAN	Number				
		1	182	.445000	1				
		2	260	.661667	2				
		3	321	.908333	3				
		4	360	.225000	4				
		5	203	.191667	5				
		Least Sq	uares Mean	ns for effe	ect Statio	on			
		Pr >  t	for HO:	LSMean(i):	=LSMean(j)	)			
			Dependent	Variable:	Al				
	i/j	1	2	3		4		5	
	1		0.2541	0.005	3 (	0.000	3 0.	9815	
	2	0.2541		0.494	5 (	0.083	ó 0.	5566	
	3	0.0058	0.4946		(	0.8480	) 0.	0251	
	4	0.0003	0.0836	0.8480	)		0.	0015	
	.5	0.9815	0.5566	0.025	(	0.001	0		

	ſ	The SAS The GLM H	S System Procedure	14:45	Friday, Octo	ober 6, 2017	9
	Class	Jass Level	vels Va	alues			
	Samp_Da Statior	ate 1	4 1 5 1	2 3 4 2 3 4 5			
	Nur	nber of obse	ervations	60			
		The SAS The GLM H	S System Procedure	14:45	Friday, Octo	ober 6, 2017	10
Dependent Variable: Ca	a Ca	S1	um of				
Source	DF	Squ	uares	Mean Square	F Value	Pr > F	
Model	19	37859	671.6	1992614.3	0.94	0.5454	
Error Corrected Total	40	849944 1228541	409.3 n81 n	2124860.2			
corrected rotar		122001					
	R-Square 0.308168	Coeff Var 48.14009	Root N 1457.6	1SE Ca 590 302	Mean 8.017		
Source	DF	Туре	I SS	Mean Square	F Value	Pr > F	
Samp_Date	3	63905	77.65	2130192.55	1.00	0.4017	
Station	4	139215	55.90	3480388.97	1.64	0.1836	
Samp_Date^Statt	511 12	1/54/5.	38.10	1402294.84	0.69	0.7526	
Source	DF	Type II	II SS	Mean Square	F Value	Pr > F	
Samp_Date	3	63905	77.65	2130192.55	1.00	0.4017	
Station Samp Date*Stati		139215:	38.10	1462294 84	1.64	0.1836	
Samp_Sacc Scace		The SAS	S System	14:45	Friday, Octo	ober 6, 2017	11
		The CIM I	) no co du no				
		Least Squa	ares Means	3			
	Adjustment	t for Multip	ole Compar	isons: Tuke	У		
	Comp. Dot		TOMEAN	LSMEAN			
	Samp_Dat	Le Ca 3548	LSMEAN 3.40000	Number 1			
	2	2815	5.26667	2			
	3	2697	7.20000	3			
	4	3051	1.20000	4			
	Least Squ Pr >  t	ares Means   for HO: I Dependent \	for effec LSMean(i)= /ariable:	ct Samp_Date =LSMean(j) Ca			
i/j	1	2		3	4		
	1 0 5000	0.52	206	0.3906	0.7868		
	2 0.5206 3 0.3906	0.90	961	0.9961	0.9705		
	4 0.7868	0.97	705	0.9096			
		The SAS The GLM H	S System Procedure	14:45	Friday, Octo	ober 6, 2017	12
	Adjustment	Least Squa for Multip	ares Means ble Compar	isons: Tuke	У		
	Station	n Cal	SMEAN	Number			
	1	2416.	25000	1			
	2	2776.	.91667	2			
	3	3087.	.50000	3			
	5	2984.	.41667	5			
	Least So Pr >  †	quares Means	s for effe LSMean(i)=	ect Station =LSMean(i)			
		Dependent V	/ariable:	Ca		_	
i/j	1	2	0 7000	3	4	5	
1	0 9733	0.9/33	0.7909	ע.1 ה ס א	∠∠ờ Ü 629 N	.0/34 9967	
3	0.7909	0.9846	0.0010	0.6	786 0	.9998	
4	0.1228	0.3629	0.6786	5	0	.5707	
5	0.8734	0.9967	0.9998	3 0.5	707		

		Cla	The SAS S The GLM Pro	System ocedure	14:45	Friday, Octo	ober 6, 2017 13
		Class Samp_Date	Level	s Valu 4 1 2	ies 3 4		
		Numbe	er of observ	ations	60		
			The SAS S The GLM Pro	System Scedure	14:45	Friday, Octo	ober 6, 2017 14
Depen	dent Variable: Na	Na	Siim	of			
	Source	DF	Squai	res Me	ean Square	F Value	Pr > F
	Model	19	3496804.	.53	184042.34	0.84	0.6509
	Corrected Total	59	12264025.	.14	219100.92		
		R-Square Coe 0.285127 16	eff Var 66.6504	Root MSH 468.1672	E Na 2 280.	Mean .9277	
	Source	DF	Type I	SS Me	ean Square	F Value	Pr > F
	Samp_Date	3	1345329.0	)86 4	148443.029	2.05	0.1228
	Station	4	327065.6	541 207 1	81766.410	0.37	0.8264
	Samp_Date"Station	1 12	1024409.0	507 1	132034.131	0.09	0.7470
	Source	DF	Type III	SS Me	ean Square	F Value	Pr > F
	Station	3	327065.0	541 4	81766.410	2.05	0.1228
	Samp_Date*Station	12	1824409.8	307 1	152034.151	0.69	0.7476
			The SAS S The GLM Pro	System ocedure	14:45	Friday, Octo	ober 6, 2017 15
		1	Least Square	es Means			
		Adjustment i	tor Multiple	e Comparis	LSMEAN	7	
		Samp_Date	Na LS	SMEAN	Number		
		1	530.94	16667	1		
		2	261.22	20000	2		
		4	177.37	0667	4		
		Least Squar Pr >  t  De	res Means fo for HO: LSN	or effect Mean(i)=LS	Samp_Date SMean(j)		
	i/j	1	2		3	4	
	1	0 4005	0.4025	5 (	0.1396	0.1810	
	2 3	0.4025	0 9230	)	0.9230	0.9608	
	4	0.1810	0.9608	3 (	0.9991	0.0001	
			The SAS S The GLM Pro	System	14:45	Friday, Octo	ober 6, 2017 16
		Adiustment 1	Least Square for Multiple	es Means Comparis	sons. Tukey	7	
		najusemene i	tor narcipic	comparts	LSMEAN		
		Station	Na LSM	IEAN	Number		
		1	185.224	1167 1833	1		
		3	230.225	5000	3		
		4	255.483	3333	4		
		J	509.575		5		
		Least Squa	for HO. ICN	for effect	t Station		
		De	ependent Var	iable: Na	a		
	i/j	1	2	3	4		5
	1	0.8585	.8585	0.9993	0.99	159 0. 107 1	.8909
	3	0.9993 0.	.9421	0.9421	0.99	99 0.	.9605
	4	0.9959 0.	.9707	0.9999		0 .	.9821
	5	U.8909 1.	.0000	0.9605	0.98	21	

		Clas	The SAS he GLM P	System rocedure Informati	14:	45 Friday	, Octok	per 6,	2017	17
		Class Samp_Date Station	Leve	els Va 4 1 5 1	lues 2 3 4 2 3 4 5					
		Number	of obset The SAS	rvations System	60 14:	45 Friday	, Octob	per 6,	2017	18
Dependent Variable:	Ma Ma	Т	he GLM P	rocedure						
•	5		Su	n of						
Source		DF	Squ	ares	Mean Squa	re Fl	/alue	Pr >	F	
Model		19	1296727	.676	68248.8	25 16	2.19	0.018	33	
Corrected Tot	al	59	2542480	.328	31143.0	10				
	R-Squar	e Coef	f Var	Root M	ISE I	Mg Mean				
	0.51002	5 53.	23480	176.47	761 3	31.5052				
Source		DF	Туре	I SS	Mean Squa	re FV	/alue	Pr >	F	
Samp_Date		3	396543.	0896	132181.02	99	4.24	0.010	8	
Station		4	264266.	5841	66066.64	60	2.12	0.096	50	
Samp_Date*Sta	ition	12	635918.	0028	52993.16	69	1.70	0.103	31	
Source		DF	Type II	I SS	Mean Squa	re FV	alue	Pr >	F	
Samp Date		3	396543.	0896	132181.02	99	4.24	0.010	8	
Station		4	264266.	5841	66066.64	60	2.12	0.096	50	
Samp_Date*Sta	ution	12	635918.	0028	52993.16	69	1.70	0.103	1	
		т	The SAS he GLM P:	System rocedure	14:	45 Friday	, Octob	per 6,	2017	19
		Le	ast Squa	res Means	3					
	Adj	ustment fo	r Multip	le Compar	isons: Tu	key				
	S	amp Date	Ma	LSMEAN	Number					
	1	amp_bace	465.	553333	Number 1					
	2		286.	806667	2					
	3		251.	633333	3					
	4		322.1	JZ/333	4					
	Le	ast Square Pr >  t  f Dep	s Means or HO: L endent V	for effec SMean(i)= ariable:	t Samp_Da LSMean(j) Mg	te				
		-1			2					
i	./j	1	2		3	4				
	1	0 0401	0.04	01	0.0100	0.1	.333			
	2	0.0401	0 94	71	0.94/1	0.9	1469 1962			
	4	0.1333	0.94	69	0.6962	0.0	,502			
		т	The SAS he GLM P:	System rocedure	14:	45 Friday	, Octob	per 6,	2017	20
		Le	ast Squa	res Means	3					
	Adj	ustment fo	r Multip	le Compar	isons: Tu	key				
		Station	Ma L	SMEAN	LSMEAN					
		1	235.8	25833	1					
		2	346.5	16667	2					
		3	371.42	25000	3					
		4	423.3	58333	4					
		J	280.4	00000	5					
	L	east Squar Pr >  t  f	es Means or HO: L	for effe SMean(i)=	ect Statio: =LSMean(j)	n				
		Dep	endent V	ariable:	Mg			-		
i/j	1	0 5	2	3		4	0.0	5		
1	0 5457	0.5	43/	0.3433	0	.U89U 8223	0.9	2/12 2884		
2	0.3433	0.9	968	0.9900	, U	.9504	0.0	/146		
4	0.0890	0.8	223	0.9504	ļ		0.2	2920		
5	0.9712	0.8	884	0.7146	5 0	.2920				

	Cla	The SAS System The GLM Procedur	14:45 re	Friday, Octo	ber 6, 2017 21
	Class	Levels	Values		
	Samp_Date	4	1 2 3 4		
	Station Numbe	5 r of observation	12345 s 60		
	wande	I OI ODSEIVACION	.5 00		
		The SAS System	14:45	Friday, Octo	ber 6, 2017 22
Dependent Variable. Fo	Fo	The GLM Procedur	е		
Dependent Vallable: Fe	re	Sum of			
Source	DF	Squares	Mean Square	F Value	Pr > F
Model	19	1294135.415	68112.390	7.32	<.0001
Error	40	372021.171	9300.529		
Corrected Total	59	1000130.380			
	R-Square Coe	ff Var Root	MSE Fe	Mean	
	0.776719 31	.74446 96.4	13925 303	.7987	
Source	DF	Type I SS	Mean Square	F Value	Pr > F
Samp Date	3	103468.1697	34489.3899	3.71	0.0191
Station	4	259089.8658	64772.4664	6.96	0.0002
Samp_Date*Statior	n 12	931577.3799	77631.4483	8.35	<.0001
Source	DF	Type III SS	Mean Square	F Value	Pr > F
Samp_Date	3	103468.1697	34489.3899	3.71	0.0191
Station	4	259089.8658	64772.4664	6.96	0.0002
Samp_Date*Statior	n 12	931577.3799	77631.4483	8.35 Erridari Octo	<.0001
		INE SAS SYSCEN	14:45	FIIday, Octo	Der 0, 2017 25
		The GLM Procedur	е		
	L	east Squares Mea	ins		
	Adjustment f	or Multiple Comp	arisons: Tukey	ł	
	Samp Date	FO LOMEAN	LSMEAN		
	1	316.786667	1		
	2	264.153333	2		
	3	268.126667	3		
	4	366.128000	4		
	Least Squar	es Means for eff	ect Samp Date		
	Pr >  t	for HO: LSMean(i	)=LSMean(j)		
	De	pendent Variable	: Fe		
i/j	1	2	3	4	
1	0 4501	0.4501	0.5179	0.5060	
3	0.5179	0.9995	0.9990	0.0393	
4	0.5060	0.0299	0.0393		
		The SAS System	14:45	Friday, Octo	ber 6, 2017 24
	L	east Squares Mea	ins		
	Adjustment f	or Multiple Comp	arisons: Tukey	7	
	Q+ + + + + + + +	FO TOMEAN	LSMEAN		
	1	205 176667	Number 1		
	2	306.933333	2		
	3	365.616667	3		
	4	381.733333	4		
	5	259.533333	5		
	Least Squa Pr >  +	res Means for ef for HO: LSMean(i	fect Station		
	De	pendent Variable	: Fe		
i/j	1	2 3	4	1	5
1	0.	0.00	19 0.00	0.	6433
2	0.0926	0.57 5745	45 0.33	340 O. 939 A	/491 0726
4	0.0006 0.	3340 0.99	39		0273
5	0.6433 0.	7491 0.07	26 0.02	273	

		Class Samp_Da Statior Nur	The SJ The GLM Class Leve L ate 1 nber of ob:	AS System Procedure 1 Informati evels Va 4 1 5 1 servations	1 alues 2 3 4 2 3 4 5 60	4:45 Fr	riday, O	October 6,	2017 25
			The SI The GLM	AS System Procedure	1	4:45 Fr	riday, O	)ctober 6,	2017 26
Depen	dent Variable: Li	Li		Sum of					
	Source	DF	S	quares	Mean Sq	uare	F Valu	le Pr>	F
	Model	19	2.00	125525	0.1053	2922	5.5	55 <.000	1
	Error Corrected Total	40 59	0.75 2.75	865133 990658	0.0189	6628			
		R-Square (	Coeff Var	Root N	MSE 710	Li Me	ean		
		0.725117	/0.08/39	0.137	/18	0.1/95	000		
	Source	DF	Тур	e I SS	Mean Sq	uare	F Valu	le Pr>	F
	Samp_Date	3	0.18	303072	0.0610	1024	3.2	22 0.032	8
	Samp Date*Station	n 12	1.16	061470	0.1044	0240 1789	5.1	.0 <.000	1
	Source	DF	Туре	III SS	Mean Sq	uare	F Valu	le Pr>	F
	Samp_Date	3	0.18	303072	0.0610	1024	3.2	0.032	8
	Station Samp Date*Station	4 12	0.65	760983 061470	0.1644	0246 1789	8.6 5.1		1
	bamp_bacc beactor								-
			The SI The GLM	AS System Procedure	1	4:45 Fr	iday, O	October 6,	2017 27
			Least Sq	uares Means	5				
		Adjustment	for Mult	iple Compan	risons:	Tukey			
		Samp Dat	te Li	i LSMEAN	LSME. Numb	an er			
		1	0.	18260000		1			
		2	0.	24426667		2			
		3 4	0.	19920000 09226667		3 4			
		Least Squ Pr >  t	ares Mean   for H0: Dependent	s for effec LSMean(i)= Variable:	ct Samp_ =LSMean( Li	Date j)			
	i/j	1	<u>^</u>	2	3		4		
	1	0 6141	0.	6141	0.98/4		0.2901	-	
	3	0.9874	0.	8068	0.0000		0.1623	3	
	4	0.2901	0.	0217	0.1623				
			The SI	AS System	1	4:45 Fr	iday, O	October 6,	2017 28
			Least Sq	uares Means	5				
		Adjustment	for Mult	iple Compan	risons:	Tukey			
		Station	. т.	TOMEAN	LSMEA	N			
		1	0.3	7225000	nunne	1			
		2	0.1	8558333		2			
		3	0.1	4141667		3			
		4 5	0.0	4000000 5866667		4 5			
	: / <del>.</del>	Least Sc Pr >  t	quares Meas   for H0: Dependent	ns for effe LSMean(i)= Variable:	ect Stat =LSMean( Li	ion j)		F	
	ر / ـــــــــــــــــــــــــــــــــــ	Ť	0.0156	0.0017	7	0.0016	5	<.0001	
	2	0.0156		0.9333	3	0.9258	3	0.1802	
	3	0.0017	0.9333	1 0000	<b>`</b>	1.0000	)	0.5862	
	4 5	<.0001	0.9238	0.5862	2	0.6020	)	0.0020	

		Class Samp_Da Station	The SA The GLM Class Level Le	AS System Procedure 1 Informat evels V 4 1 5 1	14 ion alues 2 3 4 2 3 4 5	:45 Fric	lay, Octok	ber 6, 2017	729
		Num	ber of obs	servations	60				
			The SA	AS System	14	:45 Frid	lay, Octob	per 6, 2017	7 30
Depen	dent Variable: V	V	THE GER	niocedure					
	Source	DF	S	quares	Mean Squa	are B	7 Value	Pr > F	
	Model Error	19 40	397.9 56.0	842309 448753	20.9465. 1.40112	385 219	14.95	<.0001	
	Corrected Total	59	454.03	291062					
		R-Square C 0.876561	oeff Var 17.35900	Root 1.183	MSE 690	V Mear 6.818883	1 3		
	Source	DF	Туре	e I SS	Mean Squa	are E	7 Value	Pr > F	
	Samp_Date Station	3	173.7	653430 902091	57.92178	810 523	41.34 10.60	<.0001 <.0001	
	Samp_Date*Station	12	164.8	286788	13.73572	232	9.80	<.0001	
	Source	DF	Type 1	III SS	Mean Squa	are E	7 Value	Pr > F	
	Samp_Date	3	173.7	653430 902091	57.92178	810 523	41.34 10.60	<.0001	
	Samp_Date*Station	ı 12	164.8	286788	13.73572	232	9.80	<.0001	
			The SA The GLM	AS System Procedure	14	:45 Fric	lay, Octob	ber 6, 2017	7 31
		Adjustment	Least Squ for Multi	uares Mean iple Compa	s risons: Tu	ukey			
		Camp Dat		T CMEAN	LSMEAL	N			
		1	.e 4.	74113333	Numbe.	1			
		2	7.	99040000	:	2			
		3 4	8. 5.	92793333 61606667		3 4			
		Least Squ Pr >  t	ares Means	s for effe LSMean(i)	ct Samp_Da =LSMean(j)	ate )			
	: / -	1	Dependent	Variable:	V		4		
	1/1	Ţ	<.(	2001	<.0001	C	.1963		
	2	<.0001	0.1	1405	0.1495	<	.0001		
	5 4	0.1963	<.(	0001	<.0001		.0001		
			The S <i>I</i> The GLM	AS System Procedure	14	:45 Fric	lay, Octob	ber 6, 2017	7 32
		Adjustment	Least Squ for Multi	uares Mean iple Compa	s risons: Ti	ukey			
		Station	u v	LSMEAN	LSMEAN Number				
		1 2	6.3 5.4	8116667 2833333	1				
		3	6.8	8808333	3				
		4	8.4	9591667 0091667	4				
		Least Sq Pr > It	uares Mear	ns for eff LSMean(i)	ect Static	on )			
			Dependent	Variable.	V	-			
	i/j	1	2	3 3	×	4		5	
	1	0 2990	0.2980	0.830	9 1	0.0008	0.8	3179 314	
	3	0.8309	0.0336	0.033		0.0153	1.0	0000	
	4	0.0008	<.0001	0.015	3	0 0164	0.0	0164	
	5	····	0.0017	1.000	- · · ·				

			The SAS Sy	stem	14:45	Friday, Oc	tober 6, 2017	33
		<b>C1</b>	The GLM Proc	edure				
		Class	ass Level ini Levels	ormation Value	9			
		Samp Dat	e 4	1 2 3	4			
		Station	5	1 2 3	4 5			
		Numb	er of observa	tions	60			
			The SAS Sy	stem	14:45	Friday, Oc	tober 6, 2017	34
Depen	dent Variable. Cr	Cr	The GLM Proc	edure				
Dopon		01	Sum o	f				
	Source	DF	Square	s Mean	n Square	F Value	Pr > F	
	Model	19	70.049689	7 3	.6868258	0.56	0.9091	
	Error	40	261.072793	36	5.5268198			
	Corrected Total	59	331.122483	0				
		R-Square Co	eff Var	Root MSE	Cr	Mean		
		0.211552 2	8.84941	2.554764	8.85	55517		
	Source	DF	Type I S	S Mean	n Square	F Value	Pr > F	
	Samp Date	3	1.2254621	8 0.4	40848739	0.06	0.9792	
	Station	4	38.9190430	7 9.7	72976077	1.49	0.2232	
	Samp_Date*Station	n 12	29.9051844	0 2.4	49209870	0.38	0.9625	
	~			~	~			
	Source	DF	Type III S	S Meai	n Square	F Value	Pr > F	
	Samp_Date	3	20 0100420	8 U.4 7 Q.4	40848/39	1.00	0.9792	
	Samp Date*Station	12 <sup>4</sup>	29.9051844	0 2.4	49209870	0.38	0.9625	
	··· · ·							
			The SAS Sy	stem	14:45	Friday, Oc	tober 6, 2017	35
			The GLM Proc	edure				
		Adiustment	for Multiple	Compariso	ns: Tukey	7		
		)		]	LSMEAN			
		Samp_Date	Cr LSM	EAN	Number			
		1 -	8.93906	667	1			
		2	8.64406	667	2			
		3	8.81500	000	3			
		4	9.02393	333	4			
		Least Squa	res Means for	effect Sa	amp Date			
		Pr >  t	for H0: LSMe	an(i)=LSM	ean(j)			
		D	ependent Vari	able: Cr				
	i/j	1	2		3	4		
	1	0 0000	0.9889	0.	9991	0.9997		
	23	0.9889	0 0070	0.	9978	0.9769		
	4	0.9997	0.9769	0.	9960	0.9900		
			The SAS Sy	stem	14:45	Friday, Oc	tober 6, 2017	36
			The GLM Proc	edure				
		Adjustment	Least Squares for Multiple	Means	ne. Tukot	7		
		Adjustment	ioi Muicipie	COMPATISO L	SMEAN	(		
		Station	Cr LSME	AN N	umber			
		1	7.68116	67	1			
		2	9.19666	67	2			
		3	10.12533	33	3			
		4	8.77866	67	4			
		5	8.49575	00	5			
		Least Sou	ares Means fo	r effect :	Station			
		Pr >  t	for H0: LSMe	an(i)=LSM	ean(j)			
		D	ependent Vari	able: Cr				
	i/j	1	2	3	4	1	5	
	1	0	.5980	0.1526	0.82	293	0.9346	
	2	0.5980		0.8988	0.99	943	0.9613	
	3	U.1526 0	.8988	0 0000	0.69	980	0.5294	
	4	0.0246	. 9943	0.6980	0.00	100	0.9988	
	Э	U. 2340 U	. 2013	U.JZ 74	0,99	200		

		Cla	The SAS The GLM Pr ass Level I	System ocedure informati	14:45 on	Friday, Oo	ctober 6, 2017 37
		Class Samp_Date Station	Leve	els Va 4 1 5 1	lues 2 3 4 2 3 4 5		
		NUMDE	er of obser	Valions	60		
			The SAS The GLM Pr	System cocedure	14:45	Friday, Oc	ctober 6, 2017 38
Depen	dent Variable: Mn	Mn	Sum	of			
	Source	DF	Squa	ires	Mean Square	F Value	e Pr > F
	Model	19	339.2246	5829	17.8539307	3.4	7 0.0004
	Corrected Total	40 59	544.7665	5402	J.130J404		
		R-Square Coe	eff Var	Root M	ISE Mn	Mean	
		0.622697 43	3.41669	2.2668	36 5.2	221117	
	Source	DF	Type I	SS	Mean Square	F Value	e Pr > F
	Samp_Date	3	6.7946	5190	2.2648730	0.44	4 0.7251
	Station	4	255 1045	5164 5474	19.3313791	3.70	6 0.0109 4 0.0003
	Samp_Date~Station	1 12	233.1045	9474	21.230/123	4.1	4 0.0005
	Source	DF	Type III	SS	Mean Square	F Value	e Pr > F
	Samp_Date	3	6.7946	5190	2.2648730	0.44	4 0.7251
	Samp Date*Station	12 <sup>4</sup>	255.1045	5474	21.2587123	4.14	4 0.0003
	* _		The SAS The GLM Pr	System	14:45	Friday, Oc	ctober 6, 2017 39
		1	Least Squar	es Means			
		Adjustment :	for Multipl	e Compar	isons: Tuke	У	
		Samp Date	Mn I	SMEAN	Number		
		1	5.317	53333	1		
		2	4.662	26667	2		
		4	5.342	60000	4		
		Least Squa: Pr >  t  De	res Means f for H0: LS ependent Va	for effec Mean(i)= wriable:	t Samp_Date LSMean(j) Mn	:	
	i/j	1	2		3	4	
	1	0 0577	0.857	7	0.9909	1.0000	
	2 3	0.8577	0.699	1	0.6991	0.8436	
	4	1.0000	0.843	6	0.9934		
			The SAS The GLM Pr	System	14 <b>:</b> 45	Friday, Oc	ctober 6, 2017 40
		Adjustment :	Least Squar for Multipl	es Means e Compar	isons. Tuke		
		naj as chierre	LOI HUICIPI	.e compar	LSMEAN	. <u>y</u>	
		Station	Mn LS	MEAN	Number		
		1	4.1233	3333	1		
		3	6.1504	1667	3		
		4	5.7645	8333	4		
		5	3.6177	5000	5		
		Least Squa Pr >  t	ares Means for H0: LS	for effe Mean(i)=	ct Station LSMean(j)		
	i/i	De 1	ependent Va 2	riable:	Mn	4	5
	1	± 0	.1078	0.2042	0.4	027	0.9818
	2	0.1078		0.9975	0.9	457	0.0305
	3	0.2042 0	.9975	0 0004	0.9	934	0.0660
	4 5	0.402/ 0	.9437	0.9934	0.1	598	0.1330

			The SAS Sy: The GLM Proc	stem edure	14:45	Friday, Oct	tober 6, 2017	41
		Cla	ss Level Inf	ormation				
		Class	Levels	Value	es			
		Samp_Date	4	123	34			
		Station	5	123	345			
		Numbe	r of observa The SAS Sys	tions stem	60 14:45	Friday, Oct	tober 6, 2017	7 42
Janar	ndent Variable: Ni	Ni	The GLM Proc	edure				
Deper	ndent variable. Ni	IVI	Sum o	f				
	Source	DF	Square	s Mea	an Square	F Value	Pr > F	
	Model	19	107.151544	7	5.6395550	2.38	0.0103	
	Error	40	94.615266	7	2.3653817			
	Corrected Total	59	201.766811	3				
		R-Square Coe 0.531066 23	ff Var 1 .02135	Root MSE 1.537980	Ni 6.6	Mean 80667		
	<b>Q</b>	22						
	Source	DE	Type I S	S Mea	in Square	F Value	Pr > F	
	Samp_Date	3	39.0031/02 4 1102251	/ 13. 7 1	.22//2609	5.59	0.0027	
	Station	. 10	4.1192231	/ 1. 2 5	.02980629	0.44	0.7822	
	Samp_Date^Station	1 12	03.3491412	з э.	.27909510	2.23	0.0287	
	Source	DF	Type III S	s Mea	an Square	F Value	Pr > F	
	Samp Date	3	39.6831782	7 13.	.22772609	5.59	0.0027	
	Station	4	4.1192251	71.	.02980629	0.44	0.7822	
	Samp_Date*Station	n 12	63.3491412	3 5.	.27909510	2.23	0.0287	
			The SAS Sys The GLM Proce	stem	14:45	Friday, Oct	tober 6, 2017	43
		L	east Squares	Means				
		Adjustment f	or Multiple	Compariso	ons: Tukey	7		
		0	N		LSMEAN			
		Samp_Date	NI LSM	EAN	Number			
		1	7.10060	000	1			
		2	5 69773	333	2			
		4	6.14240	000	4			
		Least Squar	es Means for	offort 9	Samo Date			
		Pr >  t	for HO: LSMe	an(i)=LSN	Mean(i)			
		De	pendent Vari	able: Ni	10411())			
	i/j	1	2		3	4		
	1		0.6798	0.	.0036	0.0309		
	2	0.6798		0.	.0641	0.3063		
	3	0.0036	0.0641			0.8495		
	4	0.0309	0.3063	0.	.8495			
			The SAS Sy: The GLM Proc	stem edure	14:45	Friday, Oct	tober 6, 2017	7 44
		L	east Squares	Means				
		Adjustment f	or Multiple (	Compariso	ons: Tukey	ł		
		0++++++	NE LONE	I	LSMEAN			
		1	6 177933	33	Nulliber 1			
		2	6 708666	67	2			
		3	6 796666	67	3			
		4	6 936916	67	4			
		5	6.783250	00	5			
		Least Squa	res Means fo	r effect	Station			
		rr >  t	LUI HU: LSMe nendent Vari	anı(ı)=LSN able: N;	rean(])			
	i/i	1 De	2	3	4	1	5	
	- / J 1	- ^	9146	0.8602	0 74	162 I	0.8694	
	2	0.9146		0.99999	0.90	961	1.0000	
	3	0.8602 0	9999		0.90	994	1.0000	
	4	0.7462 0.	9961	0.9994	0.00		0.9992	
	5	0.8694 1.	0000	1.0000	0.90	992		

		Class Samp_Date Station	The SAS The GLM Pr ass Level I Leve	System ocedur nforma ls 4 5	te tion Values 1234 1234	14:45	Friday,	October	6,	2017	45
		Numbe	er of obser	vation	is 60	)					
_		_	The SAS The GLM Pr	System ocedur	ı Te	14:45	Friday,	October	6,	2017	46
Depen	dent Variable: Co	Со	Sum	of							
	Source	DF	Squa	res	Mean	Square	F Va	lue P	r >	F	
	Model	19		0		0		•	•		
	Corrected Total	40		0		0					
	0011000000 100011	00		0							
		R-Square Coe 0.000000	eff Var •	Root	MSE 0	Co	Mean 0				
	Source	DF	Type I	SS	Mean	Square	F Va	lue P	r >	F	
	Samp_Date	3		0		0			•		
	Station Samp Date*Station	4		0		0		•	•		
	Samp_Date Station	12		0		0		•	•		
	Source	DF	Type III	SS	Mean	Square	F Va	lue P	r >	F	
	Samp_Date	3		0		0		•	•		
	Samp Date*Station	12		0		0			•		
		I Adjustment f	The SAS The GLM Pr Least Squar for Multipl	System ocedur es Mea e Comp	n Te Ans Darisons	14:45 s: Tuke	Friday, y	October	6,	2017	47
		Samp Date	Co. L	SMEAN	LS	SMEAN					
		1	C0 1	0	140	1					
		2		0		2					
		3		0		3					
		4		0		4					
		Least Squar Pr >  t	res Means f for HO: LS	or eff Mean(i	fect Sam )=LSMea	np_Date in(j)					
	i/j	1	2	LIADIE	3: 00	3	4				
	1										
	2	•			•		•				
	4	•					•				
			The SAS The GLM Pr	System	ı ce	14:45	Friday,	October	6,	2017	48
		I Adjustment f	least Squar for Multipl	es Mea e Comp	ans Darisons LSN	: Tuke	У				
		Station	Co LS	MEAN	Nur	nber					
		1		0		1					
		2		0		2					
		4		0		4					
		5		0		5					
		Least Squa Pr >  t  De	for HO: LS	for ef Mean(i riable	fect St )=LSMea	tation an(j)					
	i/j	1	2		3		4		5		
	1			•		•		•			
	∠ 3			•		•		•			
	4										
	5										

		The SA The GLM Class Level	S System Procedure Informatic	14:45	Friday, Octo	ober 6, 2017 49
	Class	Le	vels Val	Lues		
	Samp_I Statio	Date	4 1 2	234 2345		
	Ni	umber of obs	ervations	60		
		The SA The GLM	S System Procedure	14:45	Friday, Octo	ober 6, 2017 50
Dependent Variable: Zi	n Zn	ŝ	um of			
Source	DI	F Sq	uares N	Mean Square	F Value	Pr > F
Model	19	9 37785.	39555	1988.70503	1.54	0.1225
Corrected Total	4 ( 5 !	9 89343.	08909	1288.94234		
	P. Coulone	Cooff Vor	Doot M	70 70	Moon	
	0.422925	30.15175	35.9018	34 119	.0705	
Source	DI	F Туре	I SS N	Mean Square	F Value	Pr > F
Samp_Date Station	2	3 22765. 4 3363.	70163	840.94462	0.65	0.0020
Samp_Date*Stati	on 12	2 11655.	91543	971.32629	0.75	0.6920
Source	ות	F Time T	TT SS N	Aan Square	F Value	Dr > F
Samp Date		3 22765.	70163	7588.56721	5.89	0.0020
Station	2	4 3363.	77849	840.94462	0.65	0.6286
Samp_Date*Statio	on 1.	2 11655.	91543	9/1.32629	0.75	0.6920
		The SA The GLM	S System Procedure	14:45	Friday, Octo	ober 6, 2017 51
	Adjustmer	Least Squ nt for Multi	ares Means ple Compari	isons: Tukey	7	
	5			LSMEAN		
	Samp_Da 1	ate Zn 144	LSMEAN 717333	Number 1		
	2	127	.749333	2		
	3	92	.020667	3		
	4	111	./9400/	4		
	Least So Pr >	quares Means  t  for HO: Dependent	for effect LSMean(i)=I Variable: 2	: Samp_Date LSMean(j)		
i/j	1	2		3	4	
	0 5710	0.5	718	0.0014	0.0734	
	3 0.0014	4 0.0	450	0.0450	0.4421	
	4 0.0734	4 0.6	199	0.4421		
		The SA The GLM	S System Procedure	14:45	Friday, Octo	ober 6, 2017 52
	Adjustmer	Least Squ nt for Multi	ares Means ple Compari	isons: Tukey	7	
	Statio	on Zn	LSMEAN	Number		
	1	129.	948333	1		
	2	108.	966667 973333	2		
	4	125.	067500	4		
	5	116.	396667	5		
	Least S Pr >	Squares Mean  t  for H0:	s for effec LSMean(i)=I	ct Station LSMean(j)		
: /-:	1	Dependent	Variable: 2	Zn /	I	5
1/J	Ť	0.6115	د 0.8439	0.99	972 0.	.8857
2	0.6115		0.9938	0.80	062 0.	.9862
3	0.8439	0.9938	0 0570	0.95	578 1.	.0000
4	0.9972	0.9862	1.0000	0.97	U. 756	0010

		Cla	The SAS Syst The GLM Proced ss Level Infor	em ure mation	14:45	Friday,	October	6,	2017	53
		Class	Levels	Values						
		Samp_Date	4	123	4					
		Station	5	123	45					
		Numbe	r oi observati	ons 6	0					
			The SAS Syst	em	14:45	Friday,	October	6,	2017	54
Donon	dont Variable. No	2	The GLM Proced	ure						
Depend	uent variable: AS	AS	Sum of							
	Source	DF	Squares	Mean	Square	F Va	lue Pr	>	F	
	Model	19	0		0					
	Error	40	0		0					
	Corrected Total	29	U							
	1	R-Square Coe 0.000000	ff Var Ro •	ot MSE O	As	Mean 0				
	Source	DF	Type I SS	Mean	Square	F Va	lue Pr	>	F	
	Samp Date	3	1900 1 00	nean	0	1 14			-	
	Station	4	0		0					
	Samp_Date*Station	12	0		0					
	Source	DF	Type III SS	Mean	Square	F Va	lue Pr	>	F	
	Samp Date	3	0		0	1 10		,	-	
	Station	4	0		0					
	Samp_Date*Station	12	0		0		• •			
		т	The SAS Syst The GLM Proced	em ure	14:45	Friday,	October	6,	2017	55
		Adjustment f	or Multiple Co	mparison	s: Tuke	У				
				L	SMEAN					
		Samp_Date	AS LSMEA	N N 0	umber 1					
		2		0	2					
		3		0	3					
		4		0	4					
		Least Squar Pr >  t	es Means for e for HO: LSMean	ffect Sar (i)=LSMea	mp_Date an(j)					
	i/i	1	2	IE: AS	3		4			
	1									
	2			•		•				
	3	•	•			•				
	4	•	•	•						
		T.	The SAS Syst The GLM Proced east Squares M	em ure eans	14:45	Friday,	October	6,	2017	56
		_ Adjustment f	or Multiple Co	mparison	s: Tuke	У				
				LSI	MEAN					
		Station	As LSMEAN	Nu	mber 1					
		2	0		2					
		3	0		3					
		4	0		4					
		5	0		5					
		Least Squa Pr >  t  De	res Means for for HO: LSMean pendent Variab	effect S (i)=LSMe	tation an(j)					
	i/j	1	2	3		4	5			
	1									
	2	•	•		•		•			
	3 4		-		•		•			
	5									

		The SA The GLM Class Level	S System Procedure	14:45	Friday, Octo	ber 6, 2017 57
	Class Samp	Date	vels Val 4 1 2	ues 34		
	Stati	on Number of obs	5 1 2 ervations	3 4 5 60		
		The SA	S System	14:45	Friday, Octo	ber 6, 2017 58
Dependent Variable: S	r Sr	The GLM	um of			
Source	I	DF Sq	quares M	ean Square	F Value	Pr > F
Model	1	11510.	33208	605.80695	0.93	0.5575
Error Corrected Total	4	10 26159. 59 37669	21268	653.98032		
	R-Square 0.305561	Coeff Var 271.0621	Root MS 25.5730	E Sr 4 9.43	Mean 84383	
Source	Ι	)F Type	ISS M	ean Square	F Value	Pr > F
Samp_Date		3 2362.6	32903	787.544301	1.20	0.3206
Station Samp Dato*Stati	on 1	4 2581.3	85496	645.346374 547 192906	0.99	0.4258
Samp_Date Statt	.011 1	0000.3	120/0	547.192000	0.04	0.0133
Source	I	DF Type I	II SS M	ean Square	F Value	Pr > F
Samp_Date		3 2362.6	532903	787.544301	1.20	0.3206
Station Samp Date*Stati	on 1	4 2001.3 2 6566.3	813678	547.192806	0.99	0.4258
1		The SA	S System	14:45	Friday, Octo	ober 6, 2017 59
		The GLM Least Sou	Procedure ares Means			
	Adjustme	ent for Multi	ple Compari	sons: Tukey		
	0		TOMPAN	LSMEAN		
	1 Samp_L	19.	9339333	Number 1		
	2	5.	1632000	2		
	3	4.	1260000	3		
	4	8.	5144000	4		
	Least S Pr >	Squares Means  t  for H0: Dependent	for effect LSMean(i)=L Variable: S	Samp_Date SMean(j) r		
i/j	1		2	3	4	
	1	0.4	003	0.3408	0.6162	
	2 0.400	13 18 0.9	995	0.9995	0.9839	
	4 0.616	52 0.9	839	0.9652	0.0002	
		The SA The GLM	S System Procedure	14:45	Friday, Octo	ober 6, 2017 60
		Least Squ	ares Means			
	Adjustme	ent for Multi	ple Compari	sons: Tukey		
	Stati	on Sr	LSMEAN	LSMEAN Number		
	1	3.9	870833	1		
	2	5.2	644167	2		
	3	7.7	720000	3		
	4	22.2	441667	4		
	Least Pr >	Squares Mean  t  for H0:	s for effec LSMean(i)=L	t Station SMean(j) r		
i/j	1	2	S 3	- 4		5
1		0.9999	0.9962	0.99	54 0.	4191
2	0.9999	0.0000	0.9992	0.99	90 0.	4922
3 4	0.9962 0.9954	0.9992	1 0000	1.00	υυ 0. Λ	0422 6524
5	0.4191	0.4922	0.6422	0.65	24	~~£ 1

		Cla	The SAS S The GLM Pro	System ocedure oformatio	14:45	Friday, Oct	ober 6, 2017 61
		Class Samp_Date Station	Level	ls Val 4 1 2 5 1 2	lues 2 3 4 2 3 4 5		
		Numbe	er of observ	vations	60		
			The SAS S The GLM Pro	System Docedure	14:45	Friday, Oct	ober 6, 2017 62
Depend	dent Variable: Cd	Cd	Sum	of			
	Source	DF	Squai	res N	Mean Square	F Value	Pr > F
	Model Error	19	0.337440	518 267	0.01776033	1.31	0.2326
	Corrected Total	59	0.880958	385	0.01000702		
		R-Square Coe 0.383044 34	eff Var 1.03907	Root MS 0.11650	SE Cd 67 0.3	Mean 42450	
	Source	DF	Type I	SS N	Mean Square	F Value	Pr > F
	Samp_Date	3	0.140163	378	0.04672126	3.44	0.0257
	Station Samp Date*Station	4	0.068010	563	0.01700419	1.25	0.3050
	Samp_Date Station	1 12	0.12920.	505	0.01077214	0.75	0.0000
	Source	DF	Type III	SS N	Mean Square	F Value	Pr > F
	Station	4	0.068010	570 577	0.01700419	1.25	0.3050
	Samp_Date*Station	12	0.129265	563	0.01077214	0.79	0.6550
			The SAS S The GLM Pro	System ocedure	14:45	Friday, Oct	ober 6, 2017 63
		I Adjustment f	least Square for Multiple	es Means e Compari	isons: Tukev	7	
		5	-	1	LSMEAN	•	
		Samp_Date	Cd LS	SMEAN	Number 1		
		2	0.3413	33333	2		
		3	0.3762	20000	3		
		4	0.3000	00007	4		
		Least Squar	es Means fo	or effect	t Samp_Date		
		Pr >  L  De	pendent Vai	riable: (	Cd		
	i/j	1	2		3	4	
	1	0 2827	0.2827	7	0.0560	0.0288	
	3	0.0560	0.8451	L	0.0431	0.9923	
	4	0.0288	0.6928	3	0.9923		
			The SAS S The GLM Pro	System ocedure	14:45	Friday, Oct	ober 6, 2017 64
		I Adiustment f	least Square for Multiple	es Means e Compari	isons: Tukev	7	
					LSMEAN	<u>r</u>	
		Station 1	Cd LSN 0 35191	IEAN 667	Number 1		
		2	0.29775	5000	2		
		3	0.34725	5000	3		
		4 5	0.39700	3333	4		
		Least Squa Pr >  +	res Means i for H0: LSM	for effec Mean(i)=1	ct Station LSMean(i)		
		De	ependent Vai	ciable: (	Cd		_
	i/j 1	1	2 7854	3	4	765 0	5
	2	0.7854	1004	0.8352	0.8	162 (	).9924
	3	1.0000 0.	8352		0.83	326 0	.9731
	4	U.8765 0. 0.9540 0.	2462 9924	0.8326 0.9731	0.4	( 736	1.4/36

		Class Samp_Dat Station	The SAS The GLM Pr lass Level I Leve	System ocedure nformati ls Va 4 1 5 1	14:45 on lues 2 3 4 2 3 4 5	Friday,	October 6,	2017 65	
		Numk	per of obser	vations	60				
			The SAS The GLM Pr	System ocedure	14:45	Friday,	October 6,	2017 66	
Depen	dent Variable: Ba	Ba							
	Source Model Error Corrected Total	DF 19 40 59	Sum Squa 465.6023 284.8009 750.4033	1 01 1res 1754 1553	Mean Square 24.5053882 7.1200239	F Val 3.	lue Pr > .44 0.000	F )5	
	0011000000 10001	R-Square Co	Deff Var	Root M	ISE Ba	Mean			
		0.020409		2.0003				-	
	Source Samp_Date	DF 3	16.6169	202	5.5389734	e Fval	.78 0.513	£ 32	
	Samp_Date*Station	12 <sup>4</sup>	246.0103	1570	16.9145881	2.	.38 0.020	)2	
	Source Samp_Date Station	DF 3 4	Type III 16.6169 246.0103	SS 9202 982	Mean Square 5.5389734 61.5025996	F Val 0. 8.	Lue Pr > .78 0.513 .64 <.000	F 32 01	
	Samp_Date*Station	n 12	202.9750	1570	16.9145881	2.	.38 0.020	JZ	
			The SAS The GLM Pr	System	14:45	Friday,	October 6,	201/6/	
		Adjustment	Least Squar for Multipl	es Means e Compar	isons: Tuke	У			
		Samp_Date	e BaL	SMEAN	LSMEAN Number				
		1 2	9.093 9.018	06667 53333	1 2				
		3 4	8.801 7.781	06667 06667	3 4				
		Least Squa Pr >  t	ares Means f   for H0: LS	or effec Mean(i)=	t Samp_Date LSMean(j)	•			
	i/j	1	2		3	4	-		
	1 2 2	0.9998	0.999	.0	0.9905 0.9960	0.539	59		
	4	0.5395	0.586	9	0.7232	0.723	2		
			The SAS The GLM Pr	System ocedure	14:45	Friday,	October 6,	2017 68	
		Adjustment	for Multipl	e Compar	isons: Tuke	зy			
		Station	Ba LS	MEAN	LSMEAN Number				
		2	6.597	0000	2				
		3 4	11.914 9.000	4167 4167	3 4				
		5	9.422	0000	5				
		Least Squ Pr >  t I	uares Means   for H0: LS Dependent Va	for effe Mean(i)= riable:	ct Station LSMean(j) Ba				
	i/j 1	1 (	2 ).9999	3 0.0001	0.1	4 487	5 0.0650		
	2 3	0.9999 0.0001 (	0.0002	0.0002	0.1	.983 1759	0.0908 0.1700		
	4 5	0.1487 0	).1983 ).0908	0.0759 0.1700	0.9	951	0.9951		

		The SA The GLM	S System Procedure	14:45 E	Friday, Octo	ber 6, 2017 69
	Class	Les Tever	vels Valu	100		
	Samp	Date	4 1 2	3 4		
	Stati	on	5 1 2	3 4 5		
	N	umber of obse	ervations	60		
		The SA The GLM	S System Procedure	14:45 B	Friday, Octo	ber 6, 2017 70
Dependent Variable: P	b Pb					
Source	D	IF Sa	umoi Nares Me	ean Square	F Value	Pr > F
Model	1	9 146.35	19783	7.7027357	3.00	0.0017
Error	4	0 102.63	58473	2.5658962		
Corrected Total	. 5	9 248.98	78257			
	R-Square 0 587788	Coeff Var 178 5478	Root MSE 1 601841	E Pb N 1 0.897	1ean 7150	
	0.307700	1/0.54/0	1.001041	1 0.05	1100	
Source	D	F Type	I SS Me	ean Square	F Value	Pr > F
Samp_Date		3 13.84	64565	4.6154855	1.80	0.1629
Station	1	4 14.61	84996	3.6546249	1.42	0.2436
Samp_Date*Stati	on 1	2 117.88	10223	9.8239185	3.83	0.0007
Source	D	F Type I	II SS Me	ean Square	F Value	Pr > F
Samp_Date		3 13.84	64565	4.6154855	1.80	0.1629
Station		4 14.61	84996	3.6546249	1.42	0.2436
Samp_Date*Stati	on 1	2 117.88	70223	9.8239185	3.83	0.0007
		The SA The GLM	S System Procedure	14:45 E	Friday, Octo	ber 6, 2017 71
		Least Squ	ares Means			
	Adjustme	nt for Multip	ple Comparis	sons: Tukey		
	Samp D	ate Ph	LSMEAN	Number		
	1	0.2	2626667	1		
	2	0.8	4506667	2		
	3	1.5	8186667	3		
	4	0.9	3540000	4		
	Least S Pr >	quares Means	for effect LSMean(i)=LS	Samp_Date SMean(j)		
i/j	1	Dopondonio	2	3	4	
	1	0.7	167 (	0.1110	0.6228	
	2 0.716	7	(	0.5934	0.9987	
	3 0.111	0 0.5	934 987 (	1 6884	0.6884	
	- 0.022	0.5	507 0			
		The SA The GLM	S System Procedure	14:45 E	Friday, Octo	ber 6, 2017 72
		Least Squ	ares Means	-		
	Adjustme	nt for Multi	ple Comparis	sons: Tukey		
	Stati	on Ph	LSMEAN	LSMEAN		
	1	0.89	975000	1		
	2	1.00	758333	2		
	3	1.07	275000	3		
	4	0.00	000000	4		
	5	1.50	20000/	5		
	Least Pr >	Squares Mean  t  for H0: 1	s for effect LSMean(i)=LS	t Station SMean(j)		
		Dependent '	Variable: Ph	с.		-
i/j 1	1	2	3	4	· · ·	5
1 2	0.9998	0.3330	1.0000	0.040	29 N	9400
3	0.9989	1.0000	1.0000	0.481	13 0.	9633
4	0.6462	0.5429	0.4813		0.	1653
5	0.8849	0.9400	0.9633	0.165	53	

NOTE: I COMPUTED THE MEANS & STANDARD ERRORS. YOU CAN USE THEM FOR YOUR REPORTING BUT USE THE TUKEY TEST STATISTICAL DIFFERENCES TO COMPARE THESE MEANS The SAS System 18:34 Sunday, October 8, 2017 1

 			Samp	ling_Date=1			
			The MEAN	NS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	10	140.9667542	10.7478976	87.6500000	186.8588443	
Al	Al	10	1303.01	191.9381384	493.0000000	2395.00	
Ca	Ca	10	10108.80	1357.43	3727.00	19110.00	
Na	Na	10	814.6500000	70.0205958	592.9000000	1123.00	
Mg	Mg	10	2093.61	223.7289603	763.1000000	2855.00	
Fe	Fe	10	1330.45	175.8242274	540.2000000	2311.00	
L1 V	11	10	0.8346000	0.1315472	0.2440000	1.5580000	
Cr	Cr	10	12 8640000	0.4709019	10 7200000	15 6500000	
Mn	Mn	10	17 4878000	2 2752330	6 9160000	29 3000000	
Ni	Ni	10	10.7868000	0.6527726	7.8520000	15.0100000	
Co	Со	10	0.0764000	0.0519913	0	0.4520000	
Zn	Zn	10	148.9410000	26.5477089	52.5800000	279.0000000	
As	As	10	0	0	0	0	
Sr	Sr	10	29.2990000	1.6282161	19.6000000	36.2700000	
Cd	Cd	10	0.2418000	0.0446624	0	0.5290000	
Ba	Ba	10	24.7770000	1.8981881	14.1900000	30.5400000	
Pb	Pb	10	0.6581000	0.4484166	0	4.2470000	
 			Somr	ling Dato=2			
 			Samp	Diing_Date=2			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	9	153.6347835	13.6538909	106.6215558	213.9679639	
AL	AL	9	1259.47	1/8.1916634	597.6000000	2124.00	
Ca	Ca	9	9841.36	15/9.14	4857.00	19840.00	
Ma	Ma	9	1858 47	238 6053617	857 9000000	3027 00	
Fe	Fe	9	1394.10	156.2810662	667.6000000	1989.00	
Li	Li	9	1.1930000	0.3045770	0.2940000	2.9710000	
V	V	9	12.8911111	0.3935749	10.5300000	13.9400000	
Cr	Cr	9	13.6971111	0.7752978	9.1540000	17.4900000	
Mn	Mn	9	15.7866667	1.8714766	7.2390000	21.1200000	
Ni	Ni	9	11.9588889	0.3711485	10.0700000	13.6000000	
Co	Co	9	0.1557778	0.1007623	0	0.9100000	
Zn	Zn	9	132.9433333	26.1436501	70.6900000	323.2000000	
As	As	9	0 015555	2 0212212	14 2000000	E2 100000	
SI	ST Cd	9	29.010000	0.0204416	14.3000000	52.1800000	
Cu Bo	Cu Po	9	21 2700000	4 2761520	17 5400000	56 9400000	
Ph	Ph	9	2 7748889	0 8847971	17.5400000	7 3410000	
			The S	SAS System	18:34 Sunday,	October 8, 2017	2
 			Samp	oling_Date=3			
			The MEAN	IS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	10	190.0108368	11.9337354	115.4979285	239.8255055	
Al	Al	10	1248.29	208.2876214	531.7000000	2465.00	
Ca	Ca	10	11544.50	1660.49	2777.00	19060.00	
Na	Na	10	666.6000000	102.4323929	242.300000	1213.00	
Mg	Mg	10	2302.45	296.2850344	627.9000000	3436.00	
Fe	Fe	10	1515.58	178.4527835	912.6000000	2499.00	
L1 	Ll	10	1.1006000	0.1701836	0.2020000	1.9500000	
v	V	10	13.9810000	0.8444281	10.200000	17 4100000	
U⊥ Mn	Mn	10	17 3597000	1 9661932	8 3500000	26 4600000	
Ni	Ni	10	11.0038000	0.7838067	8.4010000	15.8300000	
Co	Co	10	0.0537000	0.0428123	0	0.4260000	
Zn	Zn	10	124.5420000	15.7734028	58.7000000	204.2000000	
As	As	10	0	0	0	0	
Sr	Sr	10	34.5945000	4.9800596	9.8950000	56.0700000	

Cd	Cd	10	0.4128000	0.0246630	0.2780000	0.4950000	
Ba	Ba	10	43.2930000	2.3764792	35.5500000	57.7400000	
Pb	Pb	10	2.5442000	0.8/88419		6.8460000	
			Samp	ling_Date=4			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	10	145.2970000	16.3851203	77.9800000	214.0500000	
Ca	C a	10	9340 00	2043 45	3791 00	20480.00	
Na	Na	10	758 9000000	125 0347463	120 9000000	1252 00	
Ma	Ma	10	1847 94	472 2598450	378 2000000	4410 00	
Fe	Fe	10	1568.23	388.1998586	499.0000000	3752.00	
Li	Li	10	1.0053000	0.3247419	0.0320000	2.6950000	
V	V	10	10.1890000	1.1608898	5.6490000	14.7100000	
Cr	Cr	10	12.0682000	1.2006303	8.1840000	19.2800000	
Mn	Mn	10	15.2171000	3.1697469	4.9830000	32.2400000	
Ni	Ni	10	10.0195000	1.2335396	4.9850000	15.5200000	
Со	Co	10	0.3157000	0.2011054	0	1.5810000	
Zn	Zn	10	123.1600000	16.4520851	59.5100000	237.5000000	
As	As	10	0	0	0	0	
Sr	Sr	10	31.4608000	8.2691083	6.4080000	73.9300000	
Cd	Cd	10	0.3508000	0.0364989	0.1250000	0.5410000	
Ba	Ba	10	30.2340000	6.3463059	14.6600000	65.3100000	
Pb	Pb	10	0.6315000	0.3583771	0	3.1560000	
			The S	AS System	18:34 Sunday,	October 8, 2017	3
			Sta	ation=1			
			The MEAN	S Procedure			
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	163 5506531	3 8776275	159 6730256	167 4282806	
A1	A1	2	1125.50	7.5000000	1118.00	1133.00	
Ca	Ca	2	11160.00	1070.00	10090.00	12230.00	
Na	Na	2	632.3000000	10.0000000	622.3000000	642.3000000	
Mg	Mg	2	2505.00	20.000000	2485.00	2525.00	
Fe	Fe	2	1235.50	21.5000000	1214.00	1257.00	
Li	Li	2	0.7935000	0.0105000	0.7830000	0.8040000	
V	V	2	9.9305000	0.0665000	9.8640000	9.9970000	
Cr	Cr	2	12.4500000	0.1300000	12.3200000	12.5800000	
Mn	Mn	2	20.0650000	0.0950000	19.9700000	20.1600000	
Ni	Ni	2	10.6300000	0.1300000	10.5000000	10.7600000	
Со	Co	2	0	0	0	0	
Zn	Zn	2	225.4000000	53.6000000	171.8000000	279.0000000	
As	As	2	0	0	0	0	
Sr	Sr	2	33.1150000	0.2550000	32.8600000	33.3700000	
Cd	Cd	2	U.2035000	0.0275000	0.1760000	0.2310000	
Ba	Ba	2	26.0100000	0.8200000	25.1900000	26.8300000	
			J.IQJJUUU	1.0000000	2.0000000	4.24/0000	
			Sta	tion=2			
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum	
 РМ10	 ₽M1∩	2	158 4813993	5 2004007	153 2809986	163 6818000	
Al	Al	2	2254 00	141 0000000	2113 00	2392 00	
Ca	Ca	2	10565.00	1085.00	9480.00	11650.00	
Na	Na	2	1071.00	52.0000000	1019.00	1123.00	
Mg	Mg	2	2427.50	165.5000000	2262.00	2593.00	
Fe	Fe	2	2194.00	117.0000000	2077.00	2311.00	
Li	Li	2	1.4605000	0.0975000	1.3630000	1.5580000	
V	V	2	9.0335000	0.5305000	8.5030000	9.5640000	
Cr	Cr	2	14.8900000	0.7600000	14.1300000	15.6500000	
Mn	Mn	2	27.1750000	2.1250000	25.0500000	29.300000	
Ni	Ni	2	14.0150000	0.9950000	13.0200000	15.0100000	
Co	Co	2	0.3820000	0.0700000	0.3120000	0.4520000	
Zn	Zn	2	179.0000000	59.2000000	119.8000000	238.2000000	
As	As	2	0	0	0	0	
Sr	Sr	2	30 1000000	1 4500000	28 6500000	31 5500000	
~ 1	01 01	2	30.1000000	1.4500000	20.0500000	51.5500000	
Cd	Cd	2	0.1735000	0.0105000	0.1630000	0.1840000	

Appendix C	Statistical	Analysis
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Pb	Pb	2	0.1270000	0.1270000	0	0.2540000	
			The S	AS System	18:34 Sunday,	October 8, 2017	4
			Sta	ation=3			
			The MEAN	IS Procedure			
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	119.5699325	8.1441752	111.4257573	127.7141077	
Al	Al	2	1038.50	10.5000000	1028.00	1049.00	
Ca	Ca	2	9048.00	281.0000000	8767.00	9329.00	
Na	Na	2	846.2500000	210.7500000	635.5000000	1057.00	
Mg	Mg	2	1912.50	71.5000000	1841.00	1984.00	
Fe	Fe	2	1120.00	2.0000000	1118.00	1122.00	
Li	Lı	2	0.6430000	0.0270000	0.6160000	0.6700000	
Cr	0 Cr	2	13 8750000	1 5050000	12 3700000	15 3800000	
Mn	Mn	2	13.6450000	0.3950000	13.2500000	14.0400000	
Ni	Ni	2	9.9080000	0.0490000	9.8590000	9.9570000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	74.2650000	7.5850000	66.6800000	81.8500000	
As	As	2	0	0	0	0	
Sr	Sr	2	28.2900000	0.6200000	27.6700000	28.9100000	
Cd	Cd	2	0.2370000	0.0660000	0.1710000	0.3030000	
Ba	Ba	2	23.3450000	1.0650000	22.2800000	24.4100000	
PD 	04 						
			Sta	tion=4			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	173.1228166	13.7360277	159.3867889	186.8588443	
Al	Al	2	1548.00	17.0000000	1531.00	1565.00	
Ca	Ca	2	15610.00	3500.00	12110.00	19110.00	
Na	Na	2	701.7500000	90.4500000	611.3000000	792.2000000	
Mg	Mg	2	2/38.50	22 0000000	2622.00	2855.00	
re Li	re Li	2	0 9925000	0 0895000	0 9030000	1 0820000	
V	V	2	10.0540000	0.4060000	9.6480000	10.4600000	
Cr	Cr	2	12.0250000	0.3750000	11.6500000	12.4000000	
Mn	Mn	2	19.4850000	0.6750000	18.8100000	20.1600000	
Ni	Ni	2	11.1750000	0.1050000	11.0700000	11.2800000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	187.5000000	76.4000000	111.1000000	263.9000000	
As	As	2	0	0 1500000	0	0	
Sr	Sr	2	34.1200000	2.1500000	31.9/00000	36.2700000	
Ba	Ba	2	30 4700000	0.2843000	30 4000000	30 5400000	
Ph	Ph	2	30.4700000	0.0700000	30.4000000	30.3400000	
							_
			The S	AS System	18:34 Sunday,	October 8, 2017	5
			Sampling_	Date=1 Station=	1		
			The MEAN	IS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	163.5506531	3.8776275	159.6730256	167.4282806	
Al	Al	2	1125.50	7.5000000	1118.00	1133.00	
Ca	Ca	2	11160.00	1070.00	10090.00	12230.00	
Na	Na	2	632.3000000	10.000000	622.3000000	642.3000000	
Mg	Mg	2	2505.00	20.0000000	2485.00	2525.00	
F.e	re	2	1235.50	21.5000000	1214.00	1257.00	
11 V	11 V	2	0.1933000	0.0103000	0./830000	0.8040000	
v Cr	v Cr	2	9.9303000 12 4500000	0.00000000	9.8040000 12 3200000	3.99/0000 12 5800000	
Mn	Mn	2	20.0650000	0.0950000	19.9700000	20.1600000	
Ni	Ni	2	10.6300000	0.1300000	10.5000000	10.7600000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	225.4000000	53.6000000	171.8000000	279.0000000	
As	As	2	0	0	0	0	
Sr	Sr	2	33.1150000	0.2550000	32.8600000	33.3700000	
Cd	Cd	2	0.2035000	0.0275000	0.1760000	0.2310000	
Ba	Ba	2	26.0100000	U.8200000	25.1900000	26.8300000	

Pb	Pb	2	3.1635000	1.0835000	2.0800000	4.2470000	
					<u>`</u>		
 			Sampling	g_Date=1 Station=2	2		
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	158.4813993	5.2004007	153.2809986	163.6818000	
Al	Al	2	2254.00	141.0000000	2113.00	2395.00	
Ca	Ca	2	10565.00	1085.00	9480.00	11650.00	
Na	Na	2	1071.00	52.0000000	1019.00	1123.00	
Mg	Mg	2	2427.50	165.5000000	2262.00	2593.00	
Fe	Fe	2	2194.00	117.0000000	2077.00	2311.00	
	L1 17	2	1.4605000	0.0975000	1.3630000	1.5580000	
V Cr	V Cr	2	9.0335000	0.5305000	14 1300000	9.5640000	
Mn	Mn	2	27.1750000	2.1250000	25.0500000	29.3000000	
Ni	Ni	2	14.0150000	0.9950000	13.0200000	15.0100000	
Со	Со	2	0.3820000	0.0700000	0.3120000	0.4520000	
Zn	Zn	2	179.0000000	59.2000000	119.8000000	238.200000	
As	As	2	0	0	0	0	
Sr	Sr	2	30.1000000	1.4500000	28.6500000	31.5500000	
Cd	Cd	2	0.1/35000	0.0105000	0.1630000	0.1840000	
Ba Ph	Ba Ph	2	29.3700000	1.1200000	28.2500000	0 2540000	
			The	SAS System	18:34 Sunday,	October 8, 2017	6
 			Sampling	g_Date=1 Station=3	3		
			The MEA	ANS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	119.5699325	8.1441752	111.4257573	127.7141077	
Al	Al	2	1038.50	10.5000000	1028.00	1049.00	
Ca	Ca	2	9048.00	281.0000000	8767.00	9329.00	
Na	Na	2	846.2500000	210.7500000	635.5000000	1057.00	
Mg	Mg	2	1912.50	2 0000000	1841.00	1984.00	
Li	Li	2	0.6430000	0.0270000	0.6160000	0.6700000	
V	V	2	7.0540000	0.0620000	6.9920000	7.1160000	
Cr	Cr	2	13.8750000	1.5050000	12.3700000	15.3800000	
Mn	Mn	2	13.6450000	0.3950000	13.2500000	14.0400000	
Ni	Ni	2	9.9080000	0.0490000	9.8590000	9.9570000	
Co	Co	2	74 2650000	7 5050000	0	01 0500000	
20	20	2	/4.2650000	7.5850000	00.0800000	81.8500000	
Sr	Sr	2	28.2900000	0.6200000	27.6700000	28,9100000	
Cd	Cd	2	0.2370000	0.0660000	0.1710000	0.3030000	
Ba	Ba	2	23.3450000	1.0650000	22.2800000	24.4100000	
Pb	Pb	2	0	0	0	0	
 			Sampling	g_Date=1 Station=4	1		
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	173.1228166	13.7360277	159.3867889	186.8588443	
Al	Al	2	1548.00	17.0000000	1531.00	1565.00	
Ca	Ca	2	15610.00	3500.00	12110.00	19110.00	
Na	Na	2	701.7500000	90.4500000	611.3000000	792.2000000	
Mg Fo	Mg Fo	2	∠/38.50 1500 00	110.000000	2622.00	2855.00	
Li	Li	2	0.9925000	0.0895000	0.9030000	1.0820000	
v	V	2	10.0540000	0.4060000	9.6480000	10.4600000	
Cr	Cr	2	12.0250000	0.3750000	11.6500000	12.4000000	
Mn	Mn	2	19.4850000	0.6750000	18.8100000	20.1600000	
Ni	Ni	2	11.1750000	0.1050000	11.0700000	11.2800000	
CO	Co	2	107 5000000	0	111 1000000	262 0000000	
۵11 ک	211 2 c	2	0000000, 0±	/0.400000	UTTT.T000000	203.9000000 ^	
Sr	Sr	2	34.1200000	2.1500000	31.9700000	36.2700000	
Cd	Cd	2	0.2645000	0.2645000	0	0.5290000	
Ba	Ba	2	30.4700000	0.0700000	30.4000000	30.5400000	

Pb	Pb	2	0	0	0	0	
			The	SAS System	18.34 Sunday.	October 8, 2017	7
			1110	, one of com	10.01 ballaaj,	0000001 0, 2017	
 			Samplin	g_Date=1 Station=5			
			The ME	ANS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	рм1 0	2	90 1089693	2 /589693	87 6500000	92 5679386	
Al	Al	2	549.0500000	56.0500000	493.0000000	605.1000000	
Ca	Ca	2	4161.00	434.0000000	3727.00	4595.00	
Na	Na	2	821.9500000	229.0500000	592.9000000	1051.00	
Mg	Mg	2	884.5500000	121.4500000	763.1000000	1006.00	
Fe	Fe	2	593.7500000	53.5500000	540.2000000	647.3000000	
L1 V	Ll V	2	6 7630000	0.0395000	0.2440000	0.3230000	
Cr	Cr	2	11.0800000	0.3600000	10.7200000	11.4400000	
Mn	Mn	2	7.0690000	0.1530000	6.9160000	7.2220000	
Ni	Ni	2	8.2060000	0.3540000	7.8520000	8.5600000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	78.5400000	25.9600000	52.5800000	104.5000000	
AS	AS	2	20 8700000	1 2700000	19 600000	22 1400000	
Cd	Cd	2	0.3305000	0.0065000	0.3240000	0.3370000	
Ba	Ba	2	14.6900000	0.5000000	14.1900000	15.1900000	
Pb	Pb	2	0	0	0	0	
 			Samplin	g Date=2 Station=1			
				·			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	DM10	2	191 9873670	3 5726478	188 /1/7192	195 5600148	
Al	Al	2	2070.50	53.5000000	2017.00	2124.00	
Ca	Ca	2	14688.50	5151.50	9537.00	19840.00	
Na	Na	2	784.8000000	55.1000000	729.7000000	839.9000000	
Mg	Mg	2	2432.50	227.5000000	2205.00	2660.00	
Fe	Fe	2	1960.50	28.5000000	1932.00	1989.00	
Ll	Ll	2	2.7050000	0.2660000	2.4390000	2.9/10000	
Cr	Cr	2	15.5600000	1.9300000	13.6300000	17.4900000	
Mn	Mn	2	20.8200000	0.3000000	20.5200000	21.1200000	
Ni	Ni	2	12.1300000	1.4700000	10.6600000	13.6000000	
Co	Co	2	0.2460000	0.0120000	0.2340000	0.2580000	
Zn	Zn	2	220.4000000	102.8000000	117.6000000	323.2000000	
AS	AS	2	34 5150000	2 3650000	32 1500000	36 8800000	
Cd	Cd	2	0.2620000	0.0120000	0.2500000	0.2740000	
Ba	Ba	2	35.1650000	2.2050000	32.9600000	37.3700000	
Pb	Pb	2	2.1275000	1.5785000	0.5490000	3.7060000	
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			The	e SAS System	18:34 Sunday,	October 8, 2017	8
 			Samplin	g Date=2 Station=2			
				·			
			The ME	ANS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Mavimum	
PM10	PM10	2	117.3200295	5.5700295	111.7500000	122.8900590	
Al	Al	2	1268.50	19.5000000	1249.00	1288.00	
Ca	Ca	2	8014.50	323.5000000	7691.00	8338.00	
Na	Na	2	566.2500000	7.2500000	559.0000000	573.5000000	
MY Fe	mg Fe	2	1/31.30 1482 50	18.5000000 4 5000000	1478 NO	1//U.UU 1487 00	
Li	Li	2	1.0670000	0.0240000	1.0430000	1.0910000	
V	V	2	13.7700000	0.0600000	13.7100000	13.8300000	
Cr	Cr	2	14.0550000	0.8550000	13.2000000	14.9100000	
Mn	Mn	2	20.3150000	0.0250000	20.2900000	20.3400000	
Ni	Ni	2	12.0950000	0.3850000	11.7100000	12.4800000	
Co	Со	2	0	0	0	0	
Zn Na	Zn No	2	82.9400000	12.2500000	/0.6900000	95.1900000	
Sr	Sr	2	24.7100000	0.3500000	24.3600000	25.0600000	

Cd Ba Pb	Cd Ba Pb	2 2 2	0.1445000 29.8800000 6.3000000	0.0455000 0.4200000 1.0410000	0.0990000 29.4600000 5.2590000	0.1900000 30.3000000 7.3410000	
 			Sampling_	Date=2 Station=3	3		
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	165.0029303	4.2750001	160.7279302	169.2779305	
Al	Al	2	1051.50	31.5000000	1020.00	1083.00	
Ca	Ca	2	8968.50	461.5000000	8507.00	9430.00	
Na	Na	2	462.1000000	7.5000000	444.6000000	4/9.6000000	
Mg Fe	Mg Fe	2	1300 00	39 0000000	1261 00	1339.00	
Li	Li	2	0.8255000	0.1665000	0.6590000	0.9920000	
V	V	2	11.9900000	0.2700000	11.7200000	12.2600000	
Cr	Cr	2	14.4250000	1.4050000	13.0200000	15.8300000	
Mn	Mn	2	12.9250000	0.2650000	12.6600000	13.1900000	
Ni	Ni	2	12.2550000	0.7650000	11.4900000	13.0200000	
Co	Co	2	0.4550000	0.4550000	72 5400000	0.9100000	
20	Zn As	2	109.8/00000	37.3300000	/2.5400000	147.2000000	
Sr	Sr	2	30.4950000	0.0750000	30.4200000	30.5700000	
Cd	Cd	2	0.4375000	0.0455000	0.3920000	0.4830000	
Ba	Ba	2	43.4650000	4.2350000	39.2300000	47.7000000	
Pb	Pb	2	1.2405000	1.2405000	0	2.4810000	
			The S	AS System	18:34 Sunday,	October 8, 2017	9
 			Sampling_	Date=2 Station=4	4		
			The MEAN	IS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	110.0622168	3.4406610	106.6215558	113.5028779	
Al	Al	2	601.1000000	3.5000000	597.6000000	604.6000000	
Ca	Ca	2	5135.50	278.5000000	4857.00	5414.00	
Na	Na	2	439.4000000	60.2000000	379.2000000	499.6000000	
Mg	Mg	2	8/4.1000000	16.2000000	857.9000000	710 3000000	
re Li	re Li	2	0 3215000	0 0275000	0 2940000	0 3490000	
V	V	2	13.9400000	0.0275000	13.9400000	13.9400000	
Cr	Cr	2	10.8120000	1.6580000	9.1540000	12.4700000	
Mn	Mn	2	7.5850000	0.3460000	7.2390000	7.9310000	
Ni	Ni	2	11.1000000	1.0300000	10.0700000	12.1300000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	11/.5350000	39.6650000	//.8/00000	157.2000000	
AS Sr	AS Sr	2	14 7600000	0 4600000	14 3000000	15 2200000	
Cd	Cd	2	0.2220000	0.0190000	0.2030000	0.2410000	
Ва	Ba	2	17.7250000	0.1850000	17.5400000	17.9100000	
Pb	Pb	2	0.3890000	0.0530000	0.3360000	0.4420000	
 			Sampling	Date=2 Station=	5		
				-			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	1	213.9679639		213.9679639	213.9679639	
AL	A1 Co	1	1352.00	•	1352.00	1352.00	
Cd Na	Ud. Na	⊥ 1	14960.00 561 1000000	•	14960.00 561 1000000	14960.00 561 1000000	
Ma	Mq	1	3027.00	•	3027.00	3027.00	
Fe	Fe	1	1683.00		1683.00	1683.00	
Li	Li	1	0.8990000		0.8990000	0.8990000	
V	V	1	10.5300000	•	10.5300000	10.5300000	
Cr	Cr	1	13.5700000	•	13.5700000	13.5700000	
PIII Ni	MII	⊥ 1	12 4700000	•	12 4700000	12 /700000	
Co	Co	⊥ 1	12.4/00000	•	12.4/00000 N	12.4/00000 N	
Zn	Zn	1	135.0000000		135.0000000	135.0000000	
As	As	1	0		0	0	
Sr	Sr	1	52.1800000		52.1800000	52.1800000	

Cd	Cd	1	0.1580000	•	0.1580000	0.1580000	
Ba	Ba	1	56.9400000	•	56.9400000	56.9400000	
Pb	Pb	1	4.8600000	•	4.8600000	4.8600000	
			The S	SAS System	18:34 Sunday,	October 8, 2017	10
 			Sampling	_Date=3 Station=1			
			The MEAN	NS Procedure			
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	195 3131784	1 6131784	193 7000000	196 9263568	
A 1	A 1	2	739 4500000	9 8500000	729 6000000	749 3000000	
C 2	C 2	2	12900 00	1760 00	11130 00	14650 00	
Ca Na	Ca N-	2	12090.00	12 5500000	110.00	14050.00	
Na	Na	2	461.1500000	12.5500000	448.6000000	4/3./000000	
Mg	мg	2	2187.00	1.0000000	2186.00	2188.00	
Fe	F.e	2	1009.60	11.4000000	998.2000000	1021.00	
Li	Li	2	1.2555000	0.0015000	1.2540000	1.2570000	
V	V	2	12.0350000	0.0750000	11.9600000	12.1100000	
Cr	Cr	2	11.1400000	0.200000	10.9400000	11.3400000	
Mn	Mn	2	12.3400000	0.500000	11.8400000	12.8400000	
Ni	Ni	2	8.6785000	0.2775000	8.4010000	8.9560000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	149 1350000	55 0650000	94 0700000	204 2000000	
As	Δs	2	0	0	0	0	
Sr	Sr	2	25 5050000	0 2350000	25 2700000	25 7400000	
Cd	Cd	2	0.3675000	0.2550000	0 3010000	0 4340000	
Da	Cu Do	2	20.0550000	1 0150000	20.000000	40.0700000	
Da	Dd	2	39.0330000	1.0130000	38.0400000	40.0700000	
PD	PD	2	4.02/3000	0.4903000	4.1290000	3.1200000	
 			Sampling	Date=3 Station=2			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	195.9670936	0.3807126	195.5863811	196.3478062	
Al	Al	2	1252.50	31.5000000	1221.00	1284.00	
Ca	Ca	2	18500.00	560.0000000	17940.00	19060.00	
Na	Na	2	918.8500000	6.9500000	911.9000000	925.8000000	
Ма	Ma	2	3391.50	44.5000000	3347.00	3436.00	
Fe	Fe	2	1390.00	33,0000000	1357.00	1423.00	
T.i	T.i	2	1 1165000	0 1385000	0 9780000	1 2550000	
V	V	2	11 1900000	0 4200000	10 7700000	11 6100000	
C.m.	Ç.~	2	14 9700000	0.0600000	12 0100000	15 9300000	
C1 Ma	C1	2	14.0700000	0.9000000	10 000000	10.000000	
MII	PIII NI	2	20.4030000	0.4230000	19.9000000	20.8500000	
NL	NI	2	9.3430000	0.0140000	9.3290000	9.3570000	
0	0	2	1.4.0.45.0.0.0.0	10.0500000	101 0000000	1.67 700000	
Zn	Zn	2	149.4500000	18.2500000	131.2000000	167.7000000	
As	As	2	0	0	0	0	
Sr	Sr	2	46.5700000	0.3900000	46.1800000	46.9600000	
Cd	Cd	2	0.3545000	0.0765000	0.2780000	0.4310000	
Ba	Ba	2	36.4550000	0.9050000	35.5500000	37.3600000	
Pb	Pb	2	0	0	0	0	
					10 04 0 11	0.1.1.10.0017	1 1
			The	SAS System	18:34 Sunday,	October 8, 2017	11
 			Sampling	_Date=3 Station=3			
			The MEAN	NS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Mayimum	
PM10	PM10	2	239.5947958	0.2307097	239.3640862	239.8255055	
Al	Al	2	1282.50	42.5000000	1240.00	1325.00	
Ca	Ca	2	12270.00	1660.00	10610.00	13930.00	
Na	Na	2	569.3500000	49.6500000	519.7000000	619.0000000	
Ma	Mq	2	2467.00	30.000000	2437.00	2497.00	
Fe	Fe	2	1718.00	19.0000000	1699.00	1737.00	
T.i	T. i	2	1 0395000	0 0265000	1 0130000	1 0660000	
V	V	2	18.2650000	0.3450000	17.9200000	18,6100000	
Cr	Cr	2	14 8250000	0 8850000	13 0400000	15 7100000	
C⊥ Mp	UL Mn	2	10 5450000	1 0750000	19 /700000	20 6200000	
PHI I	MIII NI-	2	12 0250000	1 0050000	10.4/00000	20.0200000	
1N T	IN L	2	T3.3320000	T.0ADAAAA	12.0400000	T3.0200000	
0	C0 7 m	2	122 2650000	42 0350000	01 2200000	175 200000	
<u>41</u>	211	2	⊥33.∠050000	42.0350000	91.2300000	113.3000000	

125

As	As	2	0	0	0	0	
Sr	Sr	2	42.9000000	7.6800000	35.2200000	50.5800000	
Cd	Cd	2	0.4700000	0.0250000	0.4450000	0.4950000	
Ba	Ba	2	56 2700000	1 4700000	54 8000000	57 7400000	
Pb	Ph	2	6 0815000	0 7645000	5 3170000	6 8460000	
 			Sampling_	Date=3 Station=4	1		
Variable	Tabal	NT	Moon	Ctd Error	Minimum	Marrimum	
variabie	Label	IN	Mean		MIIIIIIIUIII	Maximum	
PM10	PM10	2	189 9572381	0 1527619	189 8044761	190 1100000	
71	Z I	2	2369 00	97 0000000	2271 00	2465 00	
Co	Ca	2	10399 00	1761 00	2271.00	12150 00	
No	No	2	1103 00	100 1000000	901 9000000	1213 00	
Ma	Ma	2	2747 00	100.000000	2647 00	2847 00	
Fo	Fo	2	2/4/ 50	57 5000000	2384 00	2/99 00	
Li	Li	2	1 8305000	0 1195000	1 7110000	1 9500000	
V	17	2	14 6450000	0.0550000	14 5900000	14 7000000	
Cr	Ċr	2	16 6950000	0.7150000	15 9800000	17 4100000	
Mn	Mn	2	25 3550000	1 1050000	24 2500000	26.4600000	
Ni	Ni	2	13 2800000	0 7400000	12 5400000	14 0200000	
CO	C 0	2	0 2685000	0 1575000	1110000	1 1260000	
20 Zn	2n	2	109 4000000	50 7000000	58 700000	160 100000	
As	As	2	T02.4000000	0.,000000	000000	100.1000000	
Sr	Sr	2	46 3400000	9 7300000	36 610000	56 070000	
Cd	Cd	2	0 4235000	0 0655000	0 3580000	0 //200000	
Ba	Ba	2	41 530000	0.0000000	40 7100000	42 3500000	
Ph	Ph	2	0 0260000	0.0260000	40.7100000	42.3300000	
						0.0320000	
			The S	AS System	18:34 Sunday,	October 8, 2017	12
 			Sampling_	Date=3 Station=5	·		
			The MEAN	IS Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	129.2218781	13.7239496	115.4979285	142.9458277	
Al	Al	2	599.0000000	67.3000000	531.7000000	666.3000000	
Ca	Ca	2	3673.50	896.5000000	2777.00	4570.00	
Na	Na	2	279.7500000	37.4500000	242.300000	317.2000000	
Mg	Mg	2	719.7500000	91.8500000	627.9000000	811.6000000	
Fe	Fe	2	1018.80	106.2000000	912.6000000	1125.00	
Li	Li	2	0.2610000	0.0590000	0.2020000	0.3200000	
V	V	2	13.7700000	1.1700000	12.6000000	14.9400000	
Cr	Cr	2	10.5050000	0.1250000	10.3800000	10.6300000	
Mn	Mn	2	9.1535000	0.8035000	8.3500000	9.9570000	
Ni	Ni	2	9.7825000	0.0135000	9.7690000	9.7960000	
Co	Co	2	0	0	0	0	
Zn	Zn	2	81.4600000	10.3300000	71.1300000	91.7900000	
As	As	2	0	0	0	0	
Sr	Sr	2	11.6575000	1.7625000	9.8950000	13.4200000	
ud D-	Ca	2	0.4485000	0.0455000	0.4030000	0.4940000	
Ba	Ba	2	43.1550000	3.6350000	39.5200000	46./900000	
PD	PD		1.9860000	1.9860000	0	3.9720000	
				·			
 			Sampling_	Date=4 Station=1			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	117.1350000	6.2750000	110.8600000	123.4100000	
Al	Al	2	704.6000000	12.2000000	692.4000000	716.8000000	
Ca	Ca	2	4446.50	34.5000000	4412.00	4481.00	
Na	Na	2	1232.00	20.000000	1212.00	1252.00	
Mg	Mg	2	1023.50	1.5000000	1022.00	1025.00	
Fe	Fe	2	760.0500000	0.5500000	759.5000000	760.6000000	
Li	Li	2	0.3910000	0.0270000	0.3640000	0.4180000	
V	V	2	7.0495000	0.0425000	7.0070000	7.0920000	
Cr	Cr	2	10.2155000	0.6545000	9.5610000	10.8700000	
Mn	Mn	2	7.2270000	0.3310000	6.8960000	7.5580000	
	37.	~	E 0880000	A **AAAAA			
Ni	Ni	2	5.8770000	0.1190000	5.7580000	5.9960000	
Ni Co	Ni Co	2	5.8770000	0.1190000	5.7580000	5.9960000	

 As Sr Cd Ba Pb	As Sr Cd Ba Pb	2 2 2 2 2	0 14.000000 0.2275000 15.8300000 0 The S.	0 0.1000000 0.1025000 0.1600000 0 AS System	0 13.900000 0.1250000 15.6700000 0 18:34 Sunday,	0 14.100000 0.330000 15.9900000 0 October 8, 2017	13
			bampiing_		-		
			The MEAN	S Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	80.5100000	2.5300000	77.9800000	83.0400000	
Al	Al	2	365.1500000	53.5500000	311.6000000	418.7000000	
Ca	Ca	2	4030.00	239.0000000	3791.00	4269.00	
Ma	Ma	2	470 7500000	92 5500000	378 2000000	563 3000000	
Fe	Fe	2	566.0000000	67.0000000	499.0000000	633.0000000	
Li	Li	2	0.0625000	0.0305000	0.0320000	0.0930000	
V	V	2	7.1715000	1.5225000	5.6490000	8.6940000	
Cr	Cr	2	8.9155000	0.7315000	8.1840000	9.6470000	
Mn	Mn	2	5.7790000	0.7960000	4.9830000	6.5750000	
Ni	N1	2	6.8495000	1.8645000	4.9850000	8.7140000	
C0 7n	CO Zn	2	202 500000	35 0000000	167 5000000	237 5000000	
As	As	2	202.3000000	33.0000000	107.3000000	237.3000000	
Sr	Sr	2	8.1540000	1.7460000	6.4080000	9.9000000	
Cd	Cd	2	0.3315000	0.1055000	0.2260000	0.4370000	
Ba	Ba	2	15.8000000	1.1400000	14.6600000	16.9400000	
Pb	Pb	2	0	0	0	0	
 			Sampling_	Date=4 Station=3	3		
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum	
PM10	DM10	2	206 1950000	7 8550000	198 3400000	214 0500000	
Al	Al	2	3572.00	88.0000000	3484.00	3660.00	
Ca	Ca	2	19260.00	1220.00	18040.00	20480.00	
Na	Na	2	1055.50	23.5000000	1032.00	1079.00	
Mg	Mg	2	4323.50	86.5000000	4237.00	4410.00	
Fe	Fe	2	3687.00	65.0000000	3622.00	3752.00	
Li	Li	2	2.6210000	0.0740000	2.5470000	2.6950000	
V	V Cm	2	14.4150000	0.2950000	14.1200000	14./100000	
Mn	Mn	2	30 7400000	1 5000000	29 2400000	32 2400000	
Ni	Ni	2	14.7600000	0.7600000	14.0000000	15.5200000	
Со	Со	2	1.5190000	0.0620000	1.4570000	1.5810000	
Zn	Zn	2	108.7600000	16.2400000	92.5200000	125.0000000	
As	As	2	0	0	0	0	
Sr	Sr	2	72.6850000	1.2450000	71.4400000	73.9300000	
Ca Ba	Ba	2	64 4300000	0.0330000	63 5500000	65 3100000	
Pb	Pb	2	0.0395000	0.0395000	03.3300000	0.0790000	
			The S.	AS System	18:34 Sunday,	October 8, 2017	14
 			Sampling_	Date=4 Station=4	1		
			The MEAN	S Procedure			
Variable	Label	N	Mean	Std Error	Minimum	Maximum	
PM10	PM10	2	197.6550000	10.3050000	187.3500000	207.9600000	
Al	Al	2	1769.50	19.5000000	1750.00	1789.00	
Ca	Ca	2	13565.00	1395.00	12170.00	14960.00	
Na	Na	2	641.0000000	12.3000000	628.7000000	653.3000000	
Mg	Mg	2	2510.00	21.0000000	2489.00	2531.00	
re Ti	re	2	1 5655000	24.3000000	1 1600000	1 0710000	
U V	V	2	14 2700000	0.4055000	13 0200000	14 6200000	
Čr	Čr	2	12.2500000	0.0400000	12.2100000	12.2900000	
Mn	Mn	2	17.6850000	0.6850000	17.0000000	18.3700000	
Ni	Ni	2	11.6600000	0.5300000	11.1300000	12.1900000	

Со	Co	2	0.0595000	0.0595000	0	0.1190000
Zn	Zn	2	127.0000000	22.6000000	104.4000000	149.6000000
As	As	2	0	0	0	0
Sr	Sr	2	47.4150000	0.8150000	46.6000000	48.2300000
Cd	Cd	2	0.3580000	0.0480000	0.3100000	0.4060000
Ba	Ba	2	37.8150000	0.1550000	37.6600000	37.9700000
Pb	Pb	2	0.4465000	0.4465000	0	0.8930000

			Sampling_	_Date=4 Station=5	)	
Variable	Label	Ν	Mean	Std Error	Minimum	Maximum
 РМ10	PM10	2	124.9900000	3.8500000	121.1400000	128.8400000
Al	Al	2	757.3000000	6.8000000	750.5000000	764.1000000
Ca	Ca	2	5398.50	691.5000000	4707.00	6090.00
Na	Na	2	718.4000000	26.0000000	692.4000000	744.400000
Mg	Mg	2	911.9500000	26.0500000	885.9000000	938.0000000
Fe	Fe	2	876.6000000	9.600000	867.0000000	886.200000
Li	Li	2	0.3865000	0.0235000	0.3630000	0.4100000
V	V	2	8.0390000	0.1330000	7.9060000	8.1720000
Cr	Cr	2	10.0900000	0.3900000	9.700000	10.4800000
Mn	Mn	2	14.6545000	6.9555000	7.6990000	21.6100000
Ni	Ni	2	10.9510000	3.2990000	7.6520000	14.2500000
Co	Co	2	0	0	0	0
Zn	Zn	2	91.9550000	32.4450000	59.5100000	124.4000000
As	As	2	0	0	0	0
Sr	Sr	2	15.0500000	0.6100000	14.4400000	15.6600000
Cd	Cd	2	0.4530000	0.0880000	0.3650000	0.5410000
Ba	Ba	2	17.2950000	0.4550000	16.8400000	17.7500000
Pb	Pb	2	2.6715000	0.4845000	2.1870000	3.1560000

The SAS System 18:34 Sunday, October 8, 2017 15

The GLM Procedure

#### Class Level Information

Class	Levels	Values					
Sampling_Date		4	1	2	3	4	
Station	5	1 2	3	4	5		

#### Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 16

#### The GLM Procedure

Dependent Variable: PM10 PM10

Source		DF	Sum of Squares	Mean Square	F Value	Pr > F
Model		19	74132.18035	3901.69370	46.52	<.0001
Error		19	1593.39854	83.86308		
Corrected Total		38	75725.57889			
	R-Square 0.978958	Coef 5.8	f Var Root 811600 9.15	MSE PM10 M 7679 157.5	ean 759	
Source		DF	Type I SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	3 4 12	14926.39396 15609.27417 43596.51222	4975.464 3902.31854 3633.042	65 59. 46.53 69 43.	33 <.0001 <.0001 32 <.0001
Source		DF	Type III SS	Mean Square	F Value	Pr > F

<pre>atlon</pre>	Sampling_Date		3 14750.06710	4916.6890	58.63	<.0001
The GLM Procedure Least Squares Means for effect Station Pr > [t] 106.99653 1 Least Squares Means for effect Station Pr > [t] for H0.15MEAN (001 0.7187 2 19.668102 2 3 100.01037 3 4 165.237000 4 Least Squares Means for effect Sampling_Date Pr > [t] for H0.15MEAN (001 0.7187 2 0.0018 <.0001 0.0184 1 0.0018 <.0001 0.0164 2 0.0018 <.0001 0.0164 4 0.7187 0.0164 <.0001 0.0164 3 0.0011 0.0164 4 0.7187 0.0164 <.0001 0.0164 3 0.0011 0.0018 4 0.7187 0.0164 <.0001 0.0164 4 0.7187 0.0016 <.0001 0.0164 5 138.069631 2 3 182.590655 3 4 165.99555 0 2 1 166.99555 0 2 1 166.99553 0 1 1 166.99555 0 2 1 167.001 5.0001 0.0219 0.9999 0.0002 3 182.590655 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr > [t] for H0.15MEan(j) Dependent Variable: PMIO 1/j 1 2 3 4 5 1 c.0001 <.0001 0.0219 0.9999 0.0002 2 c.0001 0.0219 0.9999 0.0002 3 0.0219 c.0001 0.0303 0.0001 The SAS System 18134 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	tation Sampling Date*Sta	tion 1	11465.15025 2 /3596 51222	2866.28/56	34.18 <.1	JUUI < 0001
The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukky-Kramer 1 140.966754 1 2 159.665102 2 3 190.010337 3 4 1152.23700 4 Least Squares Means for effect Sampling_Date Pr > 1t1 for NO: LSMean(i) Dependent Variable: PMIO 1/j 1 2 3 4 1 0.0018 <.0001 0.0164 3 0.0011 <.0001 0.0164 3 0.0011 <.0001 0.0164 3 0.0011 Conol 0.0164 3 0.0011 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means for effect Station Pr > 1t1 for NU: LSMean(i)=LSMEAN 1 166.996550 1 2 0.001 3 182.590655 3 4 183.066651 2 3 182.590655 3 4 166.99550 1 2 3 182.590655 3 4 5 133.572203 5 Least Squares Means for effect Station Pr > 1t1 for NU: LSMean(i)=LSMean(j) Dependent Variable: PMIO 1 2 3 4 5 1 2 < 0001 <.0001 0.0219 2 <.0001 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 0.0219 0.9999 0.0002 3 0.0223 <.0001 0.0203 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The SAS System 18:34 Sunday, October 8, 2017 1 1/j 1 2 3 4 5 1 2 < .0001 <.0001 0.0219 0.99978 3 0.0223 <.0001 0.0203 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The SAS System 18:34 Sunday, October 8, 2017 1 Sampling_Date 4 1.2.3.4 5 Station 5 1.2.3.4.5	ampiing_bate~5te	101011 1	The SAS System	18:34 Su	inday, October	8, 2017 1
The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer 1 140.966754 1 2 159.668102 2 3 150.01037 3 4 145.237000 4 Least Squares Means for effect Sampling_Date Pr > ht for NO: LSMean(j) Dependent Variable: PMIO 1/j 1 2 3 4 1 0.0018 <.0001 0.0164 <.0001 0.0164 <.0001 0.0164 <.0001 0.0164 <.0001 0.0164 4 0.7187 0.0164 <.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means for effect Station Pr > ht for Multiple Comparisons: Tukey-Kramer 1 166.996550 1 2 139.656651 2 3 123.066651 2 3 123.059655 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr > ht for H0: LSMean(j)=LSMean(j) Dependent Variable: PMIO 1/j 1 2 3 4 5 1 30.669651 2 3 123.972203 5 Least Squares Means for effect Station Pr > ht for H0: LSMean(j)=LSMean(j) Dependent Variable: PMIO 1/j 1 2 3 4 5 1 30.0219 0.9999 0.0002 2 <.0001 <.0001 0.0303 0.0303 0.0001 3 0.0223 <.0001 0.0303 0.0303 0.0001 5 0.0002 0.0011 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Level Information Class Level Svalues Sampling_Date 4 1234 5			*		- ·	
Last Squares Means       LSMEAN         LSMEAN         Sampling_Date       ISMEAN         Sampling_Date         1       10.966750       2         3       190.00037       3         4         Least Squares Means for effect Sampling_Date         Pr > [t] for H0: LSMean(i)=LSMean(j)         Dependent Variable: PM10         1         1         0.0018            Colspan="2">Colspan="2"Colspa="2"Colspa="2"Colspan="2"Colspan="2"Colspan="2"Colspan="		Ŧ	The GLM Procedure	_		
LSMEAN LABOR LABOR LABOR Number 1 100.966754 1 2 159.666102 2 3 100.010337 3 4 145.297000 4 Least Squares Means for effect Sampling_Date Fr > [t] for H0: LSMean()] Dependent Variable: FM10 i/j 1 2 3 4 1 0.0018 <.0001 0.7187 2 0.0018 <.0001 0.0164 3 <.0001 <.0001 0.0164 4 0.7187 0.0164 <.0001 0.0164 3 <.0001 <.0001 0.0164 4 0.7187 0.0164 <.0001 0.0164 4 0.7187 0.0164 <.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer 1 166.996550 1 2 139.666510 2 3 102.390665 3 4 167.693128 4 5 139.572203 5 Least Squares Means for effect Station Fr > [t] for H0: LSMean ()] Dependent Variable: FM10 i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 0.0219 0.9999 0.0002 2 <.0001 0.0219 0.9999 0.0002 2 <.0001 0.0219 0.9999 0.0002 3 0.0219 <.0001 0.0203 0.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0303 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 12.3.4 Station 5 12.3.4.5		Adjustment for M	ultiple Compariso	ns: Tukev-Krame	r	
$\begin{tabular}{ c c c c c c } \hline LSMEAN & Number \\ \hline & LSMEAN & Number \\ \hline & 1 & 140.966754 & 1 \\ 2 & 159.668102 & 2 \\ 3 & 190.01037 & 3 \\ 4 & 145.237000 & 4 \\ \hline & Least Squares Means for effect Sampling_Date \\ Pr >  t  for H0: LSMean(i)=LSMean(j) \\ \hline & Dependent Variable: PM10 \\ \hline & I/j & 1 & 2 & 3 & 4 \\ \hline & 0.0018 & <.0001 & 0.7187 \\ 2 & 0.0018 & <.0001 & 0.0164 \\ 3 & <.0001 & 0.0164 & <.0001 \\ \hline & 0.7187 & 0.0164 & <.0001 \\ \hline & 0.7187 & 0.0164 & <.0001 \\ \hline & 0.7187 & 0.0164 & <.0001 \\ \hline & The SAS System & 18:34 Sunday, October 8, 2017 18 \\ \hline & Least Squares Means \\ Adjustment for Multiple Comparisons: Tukey-Kramer \\ \hline & Least Squares Means for effect Station \\ PM10 LSMEAN & Number \\ \hline & 1 & 166.996550 & 1 \\ 2 & 139.069651 & 2 \\ 3 & 182.590665 & 3 \\ 4 & 167.699318 & 4 \\ 5 & 139.572203 & 5 \\ \hline & Least Squares Means for effect Station \\ Pr >  t  for H0: LSMean(j) - LSMean(j) \\ \hline & Dependent Variable: PM10 \\ \hline & 1/j & 1 & 2 & 3 & 4 & 5 \\ \hline & (.001 & 0.0219 & 0.9999 & 0.0002 \\ 2 & (.001 & 0.0219 & 0.9999 & 0.0002 \\ 2 & (.001 & 0.0219 & 0.0303 & 0.0001 \\ \hline & 0.0020 & 0.9978 & <.0001 & 0.0303 & 0.0001 \\ \hline & The SAS System & 18:34 Sunday, October 8, 2017 1 \\ \hline & The GLM Procedure \\ \hline & Class Levels Values \\ \hline & Sampling_Date & 4 & 1.2.3.4 \\ \hline & Station & 5 & 1.2.3.4.5 \\ \hline \end{cases}$						
1       140.966754       1         2       159.668102       2         3       190.010837       3         4       145.237000       4         Least Squares Means for effect Sampling_Date Pr > Itl for B0: LSWean(j)=LSMean(j)         Dependent Variable: PM10         1/j       1       2       3       4         1       0.0018       <.0001		Compling D	ata DM10 ICME	LSMEAN		
<pre>1 1 140.966754 1 2 159.668102 2 3 190.010337 3 4 145.297000 4 Least Squares Means for effect Sampling_Date Pr &gt; [t] for H0: LSMean(j)=LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 1 0.0018 &lt;.0001 0.0184 3 0.0011</pre>		Sampiing_D	ate PMIU LOME	AN NUMBEL		
2 159.668102 2 3 190.010837 3 4 145.297000 4 Least Squares Means for effect Sampling_Date Pr > [t] for H0: LSMean(j)=LSMean(j) Dependent Variable: PMI0 i/j 1 2 3 4 1 0.0018 <.0001 0.7187 2 0.0018 <.0001 0.0164 3 <.0001 0.0164 4 0.7187 0.0164 <.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer 1 166.996550 1 2 133.069651 2 3 182.590665 3 4 107.187 LSMEAN Station PM10 LSMEAN Number 1 166.995316 4 5 139.572203 5 Least Squares Means for effect Station Pr > [t] for H0: LSMean(j)=LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 < .0001 0.0219 0.9998 0.0002 2 <.0001 0.0201 0.9978 3 0.0213 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0303 0.0001 The GLM Procedure Class Levels Values Sampling_Date 4 12.34 5 12.345		1	140.966754	1		
<pre>3 19.000337 3 4 145.237000 4 Least Squares Means for effect Sampling_Date Pr &gt; 1t1 for H0: LSWean(i)=LSMean(c) Dependent Variable: PM10 i/j 1 2 3 4 1 0.0018 &lt;.0001 0.0187 2 0.0018 &lt;.0001 0.0144 3 &lt;.0001 0.0164 &lt;.0001 4 0.7187 0.0164 &lt;.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer 1 166.996550 1 2 138.069655 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr &gt; 1t1 for H0: LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 &lt;.0001</pre>		2	159.668102	2		
Least Squares Means for effect Sampling_Date Pr > [t] for H0: LSMean(i)=LSMean(j) Dependent Variable: PMIO i/j 1 2 3 4 1 0.0018 <.0001 0.7187 2 0.0018 <.0001 0.0164 3 <.0001 0.0164 3 <.0001 0.0164 3 <.0001 0.0164 3 <.0001 0.0164 4 0.7187 0.0164 Class Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer LSMEAN Station PMIO LSMEAN Number 1 166.996550 1 2 138.069631 2 3 122.590665 3 4 167.693318 4 5 139.572203 5 Least Squares Means for effect Station Pr > [t] for H0: LSMean(i)=LSMean(j) Dependent Variable: PMIO i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 0.0219 0.9999 0.0001 2 <.0001 0.0201 0.0303 0.0001 3 0.0219 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 12334 Station 5 12345		3	145.297000	3		
Least Squares Means for effect Sampling_Date Pr > [t] for H0: LSMean(i)=LSMean(j) Dependent Variable: FMIO i/j 1 2 3 4 1 0.0018 <.0001 0.7187 2 0.0018 <.0001 0.0164 3 <.0001 <.0001 4 0.7187 0.0164 <.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer 1 166.996550 1 2 138.069651 2 3 142.590655 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr > [t] for H0: LSMean(i)=LSMean(j) Dependent Variable: FMIO i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 0.0303 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Least Squares Means for effect Station Pr > [t] for H0: LSMean(i)=LSMean(j) Dependent Variable: FMIO i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 0.0303 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 12 3 4 Station 5 12 3 4 5		-		-		
i/j       1       2       3       4         1       0.0018       <.0001		Least Squar Pr >  t	es Means for effe for HO: LSMean(i)	ct Sampling_Dat =LSMean(j)	.e	
<pre>i/j 1 2 3 4 1 0.018 &lt;.001 0.7187 2 0.0018 &lt;.001 0.0154 3 0.001</pre>		Dep	endent Variable:	PM10		
1       0.0018       <.0001	i/j	1	2	3	4	
2 0.0018 (.0001 0.0164 3 (.0001 0.0164 (.0001 4 0.7187 0.0164 (.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer 1 166.996550 1 2 138.069651 2 3 182.590665 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr > [t] for H0: LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 (.0001 0.0219 0.9999 0.0002 2 (.0001 (.0001 0.0001 0.9978 3 0.0219 (.0001 0.0001 0.0303 0.0001 4 0.9999 (.0001 0.0303 0.0001 5 0.0002 0.9978 (.0001 0.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 1234 Station 5 12345	1		0 0018	< 0001	0 7187	
3 <.0001 <.0001 <.0001 4 0.7187 0.0164 <.0001 The SAS System 18:34 Sunday, October 8, 2017 18 The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer Last Squares Means Interpreter Station PMIO LSMEAN Number 1 166.996550 1 2 138.069631 2 3 182.590665 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j) Dependent Variable: PMIO i/j 1 2 3 4 5 (.0001 0.0219 0.9999 0.0002 <.0001 <.0001 0.0303 0.0001 3 0.0219 0.0001 0.0303 0.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	2	0.0018	0.0010	<.0001	0.0164	
4       0.0187       0.0144       <.0001	3	<.0001	<.0001		<.0001	
The GLM Procedure       Torse Schudy, October 6, Fork 1         The GLM Procedure         Least Squares Means         Adjustment for Multiple Comparisons: Tukey-Kramer         1       166.996550         2       138.069651         2       138.059655         3       182.590665         4       167.699318         4       167.699318         5       139.572203         Least Squares Means for effect Station         Pr >  t  for H0: LSMean(i)=LSMean(j)         Dependent Variable: PM10         i/j       1       2       3       4       5         1       2       3       4       5         2       <.0001	4	0.7187	U.U164 The SAS System	<.0001	ndav October	8 2017 18
The GLM Procedure Least Squares Means         Adjustment for Multiple Comparisons: Tukey-Kramer         LSMEAN Station       PMIO LSMEAN Number         1       166.996550       1         2       138.069631       2         3       182.590665       3         4       167.699318       4         5       139.572203       5         Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j)         Dependent Variable: PMIO         i/j       1       2       3       4       5         1       <0001       0.0219       0.9999       0.0002         2       <0001       0.0219       0.9999       0.0001         2       <0001       0.0303       0.0001       0.9978         3       0.0219       <.0001       0.0303       0.0001         The GLM Procedure         Class Levels Values         Sampling_Date       4       1 2 3 4         Sampling_Date       4       1 2 3 4         Station       5       1 2 3 4 5			THE SKS SYSCEM	10.34 50	inday, occober	5, 2017 10
Least squares means         Last squares means         LSMEAN         Station         Station         Station         Station         Image: Station         Station         Image: Station         2         3       102.590665       3         4       167.699318       4         A dot for 699318       4         5       Least Squares Means for effect Station         Pr >  t  for H0: LSMean(i)=LSMean(j)         Dependent Variable: PMIO         i/j       1       2       0.0001       0.0002         2       0.0001       0.0002         2       0.0001       0.0001         2       0.0001       0.0001         2       0.001       0.0001       0.0001         2       0.001       0.0001       0.0001       0.0001 <th< td=""><td></td><td>Ŧ</td><td>The GLM Procedure</td><td>_</td><td></td><td></td></th<>		Ŧ	The GLM Procedure	_		
Line Line Line Line Line Line Line Line		Adjustment for M	ultiple Compariso	ns: Tukey-Krame	r	
Station     PM10 LSMEAN     Number       1     166.996550     1       2     138.069631     2       3     162.590665     3       4     167.699318     4       5     139.572203     5       Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j)       Dependent Variable: PM10       i/j     1     2     3     4     5       1      <.0001		-				
<pre> 1 166.996550 1 2 138.069631 2 3 182.590665 3 4 167.699318 4 5 139.572203 5  Least Squares Means for effect Station Pr &gt;  t  for H0: LSMean(i)=LSMean(j)  Dependent Variable: PM10  i/j 1 2 3 4 5 1 &lt;.0001 0.0219 0.9999 0.0002 2 &lt;.0001 &lt;.0001 &lt;.0001 0.9978 3 0.0219 &lt;.0001 0.0303 0.0001 2 &lt;.0001 0.0303 0.0001 3 0.0219 &lt;.0001 0.0303 0.0001 5 0.0002 0.9978 &lt;.0001 0.0303 0.0001 5 0.0002 0.9978 &lt;.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1234 Station 5 12345 </pre>		Station	PM10 LSMEAN	Number		
1 166.99650 1 2 138.069631 2 3 182.590665 3 4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 <.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5						
<pre>     162.590665 3     162.590665 3     4     167.699318 4     5     139.572203 5      Least Squares Means for effect Station     Pr &gt;  t  for H0: LSMean(i)=LSMean(j)     Dependent Variable: PM10  i/j 1 2 3 4 5      (.0001 0.0219 0.9999 0.0002     &lt;.0001 &lt;.0001 0.09978     3 0.0219 &lt;.0001 0.0303 0.0001     4 0.9999 &lt;.0001 0.0303 0.0001     5 0.0002 0.9978 &lt;.0001 0.0001     The SAS System 18:34 Sunday, October 8, 2017 1     The GLM Procedure     Class Levels Values     Sampling_Date 4 1234     Station 5 12345 </pre>		1	166.996550	1		
4 167.699318 4 5 139.572203 5 Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 <.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5		3	182.590665	3		
5       139.572203       5         Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j)         Dependent Variable: PM10         i/j       1       2       3       4       5         1       <.0001		4	167.699318	4		
Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 <.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5		5	139.572203	5		
Pr >  t  for H0: LSMean(i)=LSMean(j) Dependent Variable: PM10 i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 0.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5		Least Squa	res Means for eff	ect Station		
i/j       1       2       3       4       5         1       <.0001		Pr >  t	for HO: LSMean(i)	=LSMean(j)		
i/j 1 2 3 4 5 1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 0.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5		Dep	endent Variable:	РМ10		
1 <.0001 0.0219 0.9999 0.0002 2 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 <.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	i/j	1	2	3 4	5	
2 <.0001 <.0001 <.0001 0.9978 3 0.0219 <.0001 0.0303 <.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	1	<.	0001 0.021	0.999	0.0002	
3 0.0219 <.0001 0.0303 <.0001 4 0.9999 <.0001 0.0303 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	2	<.0001	<.000	01 <.000	L 0.9978	
5 0.0002 0.9978 <.0001 0.0001 5 0.0002 0.9978 <.0001 0.0001 The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	3	0.0219 <.	0001	0.0303	3 <.0001	
The SAS System 18:34 Sunday, October 8, 2017 1 The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	4 5	0.0002 0	9978 <_000	1 0.0001	0.0001	
The GLM Procedure Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5	0	0.0002 0.	The SAS System	18:34 Su	inday, October	8, 2017 1
Class Level Information Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5			The GLM Procedure			
Class Levels Values Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5		Cla	ss Level Informat	ion		
Sampling_Date 4 1 2 3 4 Station 5 1 2 3 4 5		Class	Levels V	alues		
Station         5         1         2         4         5		Sampling	Date 4	1 2 3 4		
Station 5 1 2 5 4 5		Station		2315		
		SLALION	5 I	2343		
		Mullbe	- or observations	40		

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 20

The GLM Procedure

Dependent Variable: Al Al

Sum of

Source	DF	Squares	Mean Square	F Value	Pr > F		
Model	19	23263547.90	1224397.26	208.66	<.0001		
Error	19	111491.69	5867.98				
Corrected Total	38	23375039.60					
	R-Square C	oeff Var Roc	t MSE AI M	Mean			
	0.995230	5.836652	76.60277	4			
Source	DF	Type I SS	Mean Square	F Value	Pr > F		
Sampling_Date Station Sampling_Date*Sta	4 ation	3 214361. 4566524.77 12 18482661.	17 71453. 1141631.19 96 1540221.	.72 12.2 194.55 83 262.4	18 0.0001 <.0001 48 <.0001		
Source	DF	Type III SS	Mean Square	F Value	Pr > F		
Sampling_Date Station Sampling_Date*Sta	4 ation	3 204703. 3735120.45 12 18482661. The SAS Syste	70 68234 933780.11 96 1540221 m 18:34 5	.57 11. 159.13 .83 262. Sunday, Octo	63 0.0001 <.0001 48 <.0001 ber 8, 2017 21		
	Adjustment for	The GLM Procedu Least Squares Me Multiple Compari	re ans sons: Tukey-Kram	ner			
			LSMEAN				
	Sampling	_Date Al LS	MEAN Number	r			
	1 2	1303.01000 1268.72000	1 2				
	3 4	1248.29000 1433.71000	3 4				
	Least Squ Pr >  t	ares Means for ef   for HO: LSMean(	fect Sampling_Da i)=LSMean(j)	ate			
		Dependent Variabl	e: Al				
i/j	1	2	3	4			
1 2	0.7763	0.7763	0.4036 0.9402	0.0059 0.0010			
3	0.4036	0.9402	0 0002	0.0002			
T	0.0000	The SAS Syste	m 18:34 s	Sunday, Octo	ber 8, 2017 22		
	Adjustment for	The GLM Procedu Least Squares Me Multiple Compari	re ans sons: Tukey-Kram	ner			
	Station	Al LSMEAN	LSMEAN Number				
	1	1160.01250	1				
	2 3	1285.03750 1736.12500	2 3				
	4	1571.65000 814 33750	4				
Least Squares Means for effect Station Pr >  t  for H0: LSMean(i)=LSMean(j) Dependent Variable: Al							
i/j	1	2	3	4	5		
- 1		0.0295 < 0	1001 < 000	01 <	0001		
2	0.0295	<.(	0001 <.000	01 <.	0001		
3	<.0001 <.0001	<.0001 <.0001 0.0	0.003	32 <. <.	0001		
5	<.0001	<.0001 <.0	0001 <.000	01			
The SAS System

18:34 Sunday, October 8, 2017 23

The GLM Procedure

#### Class Level Information

Class Levels Values Sampling\_Date 4 1 2 3 4

Station 5 1 2 3 4 5

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 24

The GLM Procedure

Dependent Variable: Ca Ca

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	19	884411098.9	46547952.6	7.92	<.0001
Error	19	111630265.5	5875277.1		
Corrected Total	38	996041364.4			

R-Square Coeff Var Root MSE Ca Mean

#### 0.887926 23.72154 2423.897

Source	DF	Type I SS	Mean Square F	Value Pr > F
Sampling_Date	4	3 26699502.0	8899834.0	1.51 0.2429
Station		181802359.7	45450589.9	7.74 0.0007
Sampling_Date*Station		12 675909237.2	56325769.8	9.59 <.0001

Source	DF	Type III SS	Mean Square F	Value Pr	> F
Sampling_Date	4	3 25043603.8	8347867.9 27360953 2	1.42	0.2677
Sampling_Date*Station	1	12 675909237.2	56325769.8	9.59	<.0001
		The SAS System	18:34 Sunda	ay, October 8	8, 2017 25

The GLM Procedure Least Squares Means

Adjustment for Multiple Comparisons: Tukey-Kramer

Sampling_Date	Ca LSMEAN	LSMEAN Number
1	10108.8000	1
2	10353.4000	2
3	11544.5000	3
4	9340.0000	4

Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Ca

i/j	1	2	3	4		
1		0.9964	0.5594	0.8922		
2	0.9964		0.7242	0.8094		
3	0.5594	0.7242		0.2109		
4	0.8922	0.8094	0.2109			
		The SAS System	18:34	4 Sunday, October	8, 2017	26

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Ca LSMEAN	LSMEAN Number
1 2	10796.2500 10277.3750	1 2
3	12386.6250	3
4	11174.8750	4
5	7048.2500	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Ca

i/j	1	2	3	4	5	
1		0.9924	0.6872	0.9977	0.0600	
2	0.9924		0.4346	0.9441	0.1295	
3	0.6872	0.4346		0.8522	0.0043	
4	0.9977	0.9441	0.8522		0.0330	
5	0.0600	0.1295	0.0043	0.0330		
		The SA	AS System	18:34 Sunday,	October 8, 2017	27

#### The GLM Procedure

#### Class Level Information

Class	Levels	Values					
Sampling_Date		4		1	2	3	4
Station	5	1	2	3	4	5	

#### Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 28

The GLM Procedure

Dependent Variable: Na Na

Source		DF	Sum of Squares	Mean Square	F Value	Pr > F
Model		19	3034984.100	159736.005	11.38	<.0001
Error		19	266704.400	14037.074		
Corrected Total		38	3301688.500			
	R-Square 0.919222	Coef 16.	f Var Root 2 82212 118.4	MSE Na Me 782 704.30	an 000	
Source		DF	Type I SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	3 4 12	345713.086 166476.285 2522794.729	115237.69 41619.071 210232.89	95 8.2 2.96 94 14.9	21 0.0010 0.0464 98 <.0001
Source		DF	Type III SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	3 4 12 T Le	332588.518 134491.488 2522794.729 The SAS System he GLM Procedure ast Squares Mean	110862.83 33622.872 210232.89 18:34 Su	2.40 2.40 4 14.9 240 2.40 24 2.40	90 0.0013 0.0865 98 <.0001 per 8, 2017 29

Adjustment for Multiple Comparisons: Tukey-Kramer

			LSMEAN
Sampling_Date	Na	LSMEAN	Number

1	814.650000	1
2	562.730000	2
3	666.600000	3
4	758.900000	4

Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Na

i/j	1	2	3		4			
1		0.0012	0.0518	0.721	16			
2	0.0012		0.2738	0.011	10			
3	0.0518	0.2738		0.330	)7			
4	0.7216	0.0110	0.3307					
		The SAS System	18:34	Sunday,	October	8,	2017	30

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Na LSMEAN	LSMEAN Number
1	777.562500	1
2	675.925000	2
3	733.300000	3
4	721.512500	4
5	595.300000	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Na

i/j	1	2	3	4	5	
1		0.4483	0.9424	0.8752	0.0618	
2	0.4483		0.8658	0.9363	0.7042	
3	0.9424	0.8658		0.9996	0.2233	
4	0.8752	0.9363	0.9996		0.2994	
5	0.0618	0.7042	0.2233	0.2994		
		The SA	AS System	18:34 Sunday,	October 8, 2017	31

The GLM Procedure

Class Level Information

Class	Levels	evels Values					
Sampling_Date		4	1	2	3	4	
Station	5	1 2	3	4	5		

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 32

#### The GLM Procedure

Dependent Variable: Mg Mg

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	19	37652101.01	1981689.53	123.75	<.0001
Error	19	304258.64	16013.61		
Corrected Total	38	37956359.65			

R-Square Coeff Var Root MSE Mg Mean

	0.991984	6.234038	126.5449	2029.903			
Source	DI	Type I	SS Mea	n Square – F V	/alue Pr >	> F	
Sampling_Date Station Sampling_Date*Sta	ation	3 1379 8870803 12 27402	0023.57 83 22 2273.60	459674.52 17700.96 13 2283522.80 142	28.71 38.49 <.00 2.60 <.00	<.0001 )01 )01	
Source	DI	Type III	SS Mea	n Square – F V	/alue Pr >	> F	
Sampling_Date Station Sampling_Date*Sta	ation	3 110 5652753 12 27402 The SAS 5	8454.15 09 14 2273.60 System	369484.72 13188.27 { 2283522.80 18:34 Sunday	23.07 88.25 <.00 142.60 7, October 8,	<.0001 )01 <.0001 , 2017 33	
	Adjustment fo	The GLM Pro Least Square or Multiple Cor	ocedure es Means mparisons:	Tukey-Kramer			
				LSMEAN			
	Samplin	ng_Date M	ig LSMEAN	Number			
	1	2093.0	51000	1			
	2	1975.3	32000	2			
	3	2302.4	15000	3			
	4	1847.9	94000	4			
	Least So Pr >	quares Means fo t  for H0: LSM	or effect S Mean(i)=LSM	ampling_Date ean(j)			
		Dependent Vai	riable: Mg				
i/j	:	. 2	2	3	4		
1	0 225	0.2254	۱ 0. ٥	0077 0.	0018		
2 3	0.2254	· 0.000	υ.	UUUI U.	1/46		
4	0.0018	0.174		0001	0001		
		The SAS S	System	18:34 Sunday	, October 8,	, 2017 34	
		The GLM Pro Least Square	ocedure es Means				
	Adjustment fo	or Multiple Cor	mparisons:	Tukey-Kramer			
			L	SMEAN			
	Statio	on Mg LSN	iean n	umber			
	1	2037.00	0000	1			
	2	2010.33	250	2			
	3	2623.62	2500	3			
	4	2217.40	250	4			
	5	1000.01	200	5			
	Least S Pr >	Squares Means : t  for H0: LSM	for effect Mean(i)=LSM	Station ean(j)			
		Dependent Va	iable: Mg				
i/j	1	2	3	4	5		
1		0.9928	<.0001	0.0682	<.0001		
2	0.9928		<.0001	0.0290	<.0001		
3	<.0001	<.0001		<.0001	<.0001		
4	0.0682	0.0290	<.0001		<.0001		
5	<.0001	<.0001	<.0001	<.0001	· Oatsberr O	2017 25	
		The SAS S	ystem	10:34 Sunday	, uctoper 8,	ZUI1 35	
		The GLM Pro	ocedure				

Class Level Information

Class	Levels	Values					
Sampling_Date		41	2	3	4		
Station	5	1	2	3	4	5	

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 36 The GLM Procedure

Dependent Variable: Fe Fe

Source		DF	Sum of Squares	Mean Square	F Value	Pr > F
Model		19	21231355.17	1117439.75	231.34	<.0001
Error		19	91775.08	4830.27		
Corrected Total		38	21323130.25			
	R-Square 0.995696	Coef 4.	ff Var Roc 781317 69	ot MSE Fe M .50012 1453.	ean 577	
Source		DF	Type I SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	3 4 12	3 353337. 4543275.31 2 16334742.	03 117779. 1135818.83 84 1361228.	01 24. 235.15 57 281.	38 <.0001 <.0001 81 <.0001
Source		DF	Type III SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	3 4 12	3 327244. 3757882.32 2 16334742. The SAS Syste	74 109081. 939470.58 84 1361228. em 18:34 S	58 22. 194.50 57 281. unday, Octo	58 <.0001 <.0001 81 <.0001 ber 8, 2017 37

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Sampling_Date	Fe LSMEAN	LSMEAN Number
1 2 3 4	1330.45000 1422.99000 1515.58000 1568.23000	1 2 3 4

Least Squares Means for effect Sampling\_Date
 Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Fe

i/j	1	2	3	4	
1		0.0474	<.0001	<.0001	
2	0.0474		0.0472	0.0014	
3	<.0001	0.0472		0.3540	
4	<.0001	0.0014	0.3540		
		The SAS System	18:34	4 Sunday, Oc	to

18:34 Sunday, October 8, 2017 38

#### The GLM Procedure

Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Fe LSMEAN	LSMEAN Number
1	1241.41250	1
2	1408.12500	2
3	1956.25000	3
4	1647.73750	4
5	1043.03750	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Fe

i/j	1	2	3	4	5		
1		0.0011	<.0001	<.0001	0.0003		
2	0.0011		<.0001	<.0001	<.0001		
3	<.0001	<.0001		<.0001	<.0001		
4	<.0001	<.0001	<.0001		<.0001		
5	0.0003	<.0001	<.0001	<.0001			
		The SA	AS System	18:34 Sunday,	October 8,	2017	39

The GLM Procedure

Class Level Information

Class Levels Values

Sampling\_Date 4 1 2 3 4

Station 5 1 2 3 4 5

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 40

The GLM Procedure

Dependent Variable: Li Li

Source	DF	Squares	Mean Square	F Value	Pr > F
Model	19	20.35230340	1.07117386	30.88	<.0001
Error	19	0.65898450	0.03468339		
Corrected Total	38	21.01128790			

R-Square	Coeff Var	Root MSE	Li Mean
0 968637	18 09366	0 186235	1 029282

Source	DF	Type I SS	Mean Square	F Value Pr > F
Sampling_Date Station Sampling_Date*Station	4	3 0.676857 3.95298271 12 15.722463	00 0.22561900 0.98824568 69 1.31020531	6.51 0.0033 28.49 <.0001 37.78 <.0001
Source	DF	Type III SS	Mean Square	F Value Pr > F

Sampling Date		3 0.58269736	0.19423245	5.60	0.0063	
Station	4	3.37661097	0.84415274	24.34 <.	.0001	
Sampling_Date*Station	1	12 15.72246369	1.31020531	37.78	<.0001	
		The SAS System	18:34 Sunda	y, October	8, 2017	41

#### The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Sampling_Date	Li LSMEAN	LSMEAN Number
1	0.83460000	1
2	1.16360000	2
3	1.10060000	3
4	1.00530000	4

Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Li

i/j	1	2	3	4			
1		0.0066	0.0227	0.2054			
2	0.0066		0.8874	0.2984			
3	0.0227	0.8874		0.6676			
4	0.2054	0.2984	0.6676				
		The SAS System	18:3	4 Sunday, Oct	ober 8, 2	2017	42

#### The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Li LSMEAN	LSMEAN Number
1	1.28625000	1
2	0.92662500	2
3	1.28225000	3
4	1.17750000	4
5	0.45750000	5

#### Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Li

i/j	1	2	3	4	5	
1		0.0082	1.0000	0.7689	<.0001	
2	0.0082		0.0090	0.0924	0.0012	
3	1.0000	0.0090		0.7916	<.0001	
4	0.7689	0.0924	0.7916		<.0001	
5	<.0001	0.0012 The SA	<.0001 AS System	<.0001 18:34 Sunday,	October 8, 2017	43

#### The GLM Procedure

#### Class Level Information

Class	Levels	Val	ue	s		
Sampling_Date		4	1	2	3	4

#### Station 5 1 2 3 4 5

#### Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 44

#### The GLM Procedure

Dependent Variable: V V

Source		DF	Sum of Squares	Mean Square	F Value	Pr > F
Model		19	389.1924830	20.4838149	40.60	<.0001
Error		19	9.5865680	0.5045562		
Corrected Total		38	398.7790510			
	R-Square 0.975960	Coe 6.	eff Var Root .247892 0.71	MSE V M 0321 11.368	ean 397	
Source		DF	Type I SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	tation	4	3 181.5128801 77.9308922 2 129.7487107	60.50429 19.4827231 7 10.81239	34 119. 38.61 26 21.	92 <.0001 <.0001 43 <.0001
Source		DF	Type III SS	Mean Square	F Value	Pr > F

Sampling_Date Station Sampling_Date*Station	3 4 78. 12 The	174.210932 4454226 129.748710 SAS System	19.611 19.611 17 10	3.0703107 L3556 D.8123926 L8:34 Sunda	115.09 < 38.87 <.000 21.43 < ay, October 8,	<.0001 )1 <.0001 2017 45
Adjust	The GI Least S ment for Multipl	.M Procedur Squares Mea .e Comparis	e ins ions: Tuke	ey-Kramer		
-	-	-		-		
\$	Sampling_Date	V LSM	LSME IEAN	EAN Number		
<u>:</u>	1	8.5670000		1		
	2 1	2.6550000		2		
	4 1	0.1890000		4		
Le	east Squares Mea Pr >  t  for HC	ns for eff : LSMean(i	ect Sampl )=LSMean(	Ling_Date (j)		
	Depender	nt Variable	: V			
i/j	1	2	3	3	4	
1		< 0001	< 0001	0	0003	
2	<.0001	<	0.0041	L <	.0001	
3	<.0001 (	0.0041		<	.0001	
4	0.0003 ·	<.0001	<.0001	1 19:34 gunda	W October 9	2017 46
	1116	SAS SYSCEN	. 1	10.34 Sulluc	ay, Occober 0,	201/ 40
Adjust	The GI Least S nent for Multipl	M Procedur Squares Mea .e Comparis	e ins ions: Tuke	ey-Kramer		
			LSMEA	AN		
	Station	V LSMEAN	Numbe	er		
	1 1	0.5150000		1		
	2 1	0.2912500		2		
	3 1	2.9310000		3		
	4 1. 5	9.7755000		4 5		
1	Least Squares Me Pr >  t  for HC	eans for ef ): LSMean(i	fect Stat )=LSMean	cion (j)		
	Depender	nt Variable	e: V			
i/j 1	2		3	4	5	
1	0.9683	<.00	01	<.0001	0.3203	
2 0.9683	< 0.001	<.00	01	<.0001	0.6535	
4 < 0001	< 0001	0 91	66	0.9100	< 0001	
5 0.3203	0.6535	<.00	01	<.0001	(.0001	
	The	SAS System	ı 1	L8:34 Sunda	ay, October 8,	2017 47
	The GI	M Procedur	е			
	Class Lev	vel Informa	tion			
	Class	Levels	Values			
	Sampling_Date	4	123	3 4		
	Station	5	12345	5		

#### Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 48

The GLM Procedure

Dependent Variable: Cr Cr

Source	DF	Sur Squa	n of ares 1	Mean Square	F Value	Pr > F
Model	19	243.955	9309	12.8397858	7.69	<.0001
Error	19	31.731	2030	1.6700633		
Corrected Total	38	275.687	1339			
	D. 0.		Dist			
	R-Square	Coeff Var	Root M	3E Cr Me	an	
	0.884901	9.908282	1.2923	19 13.042	. 1 2	
Source	DF	Type 1	ISS 1	Mean Square	F Value	Pr > F
Sampling_Date	4	3 16 79 708	.8544694	5.618156	5 3. 11.93	36 0.0403 < 0001
Sampling_Date*St	tation	12 147	.3929543	12.282746	52 7.	35 <.0001
Source	DF	Type III	I SS 1	4ean Square	F Value	Pr > F
Sampling_Date		3 16	.4240258	5.474675	3 3.	28 0.0435
Station Sampling_Date*St	4 tation	71.3278 12 147	3988 .3929543	17.8319747 12.282746	10.68 52 7.	0.0001 35 <.0001
		The SAS	System	18:34 Su	inday, Octo	ber 8, 2017 49
		The GLM Pi Least Squai	rocedure res Means			
	Adjustment fo	r Multiple Co	omparison	s: Tukey-Krame	r	
	Samplin	g_Date	Cr LSMEAN	LSMEAN N Number		
	1	12.80	540000	1		
	2 3	13.68 13.60	344000 )70000	2 3		
	4	12.00	582000	4		
	Least Sq Pr >	uares Means f t  for H0: LS	for effect SMean(i)=1	Sampling_Dat SMean(j)	e	
		Dependent Va	ariable: (	Cr		
i/j	1		2	3	4	
1	L 2 0.5420	0.542	20	0.5827	0.5281	
3	3 0.5827 1 0.5281	0.99	92 58	0.0673	0.0673	
		The SAS	System	18:34 Su	unday, Octo	ber 8, 2017 50
		The GLM Pr	cocedure			
	Adjustment fo	r Multiple Co	omparison	s: Tukey-Krame	er	
	Statio	n CrLS	SMEAN	LSMEAN Number		
	1	12.341	L3750	1		
	2 3	13.182 15.498	26250 37500	2 3		
	4	12.945	55000	4		
	5	11.31.	12500	5		
	Least S Pr >	quares Means t  for H0: LS	for effe SMean(i)=1	ct Station LSMean(j)		
		Dependent Va	ariable: (	Cr		
i/j	1	2	3	4		5
1		0.6933	0.0009	0.8797	0.	5729
2 3	0.6933 0.0009	0.0150	0.0150	0.9958 0.0068	0. <.	0862 0001
4	0.8797	0.9958	0.0068		0.	1625

0.0862 5 0.5729 <.0001 0.1625 18:34 Sunday, October 8, 2017 51 The SAS System The GLM Procedure Class Level Information Class Levels Values Sampling\_Date 4 1 2 3 4 5 1 2 3 4 5 Station Number of observations 40 NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, Oct 18:34 Sunday, October 8, 2017 52 The GLM Procedure Dependent Variable: Mn Mn Sum of Source DF Squares Mean Square F Value Pr > F 19 1886.967016 99.314053 15.53 <.0001 Model Error 19 121.469089 6.393110 Corrected Total 38 2008.436105 R-Square Coeff Var Root MSE Mn Mean 0.939521 15.34245 2.528460 16.48015 Type I SS Mean Square F Value Pr > F Source DF 12.723631 1.99 0.1496 73.904027 11.56 <.0001 129.431668 20.25 <.0001 Sampling Date 3 38.170892 Station 4 295.616110 Sampling\_Date\*Station 12 1553.180014 Type III SS Mean Square F Value Pr > F Source DF 34.90130311.6337681.820.177821637854.3040958.490.00041553.180014129.43166820.25<.0001</td>AS System18:34 Sunday, October 8, 201753 3 Sampling\_Date 217.216378 12 1553.180014 Station 4 Sampling\_Date\*Station The SAS System The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer LSMEAN Mn LSMEAN Sampling\_Date Number 1 17.4878000 1 2 16.0870000 2 17.3597000 3 3 15.2171000 4 4 Least Squares Means for effect Sampling\_Date
Pr > |t| for H0: LSMean(i)=LSMean(j) Dependent Variable: Mn i/j 1 3 2 4 0.2199 0.6456 0.9995 1 2 0.6456 0.7095 0.8824 3 0.9995 0.7095 0.2632 0.2632 4 0.2199 0.8824 18:34 Sunday, October 8, 2017 54 The SAS System The GLM Procedure

Least Squares Means

Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Mn LSMEAN	LSMEAN Number
1 2 2	15.1130000 18.4185000	1 2
3 4	17.5275000	3
5	12.4167500	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Mn

i/j	1	2	3	4	5	
1		0.1073	0.0308	0.3458	0.2985	
2	0.1073		0.9684	0.9529	0.0021	
3	0.0308	0.9684		0.6745	0.0006	
4	0.3458	0.9529	0.6745		0.0092	
5	0.2985	0.0021	0.0006	0.0092		
		The SA	AS System	18:34 Sunday,	October 8, 201	7 55

The GLM Procedure

Class Level Information

Class	Levels	Values					
Sampling_Date		4		1	2	3	4
Station	5	1	2	3	4	5	

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 56

The GLM Procedure

Dependent Variable: Ni Ni

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	19	209.4767362	11.0250914	4.27	0.0014
Error	19	49.0986975	2.5841420		
Corrected Total	38	258.5754337			

R-Square	Coeff Var	Root MSE	Ni Mean
0.810118	14.72609	1.607527	10.91618

Source	DF	Тγ	pe I SS	Mean Square	F Value Pr	> F
Sampling_Date Station Sampling_Date*Station	4	3 56. 12	18.0696912 9170730 134.4899721	6.0232304 14.2292683 11.207497	4 2.33 5.51 0.0 7 4.34	0.1067 0041 0.0023
Source	DF	Туре	III SS	Mean Square	F Value Pr	> F
Sampling Date		З	18 2529754	6 084325	1 2 35	0 1042

		-						
Station	4	55.1	.353411	13.7838353		5.33 0	.0047	
Sampling_Date*Station		12	134.4899721	11.2074	977	4.34	0.0023	3
		The S	SAS System	18:34	Sunday	, October	8, 2017	57

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

LSMEAN

Sampling_Date	Ni LSMEAN	Number
1	10.7868000	1
2	12.0100000	2
3	11.0038000	3
4	10.0195000	4

# Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Ni

i/j	1	2	3		4			
1		0.3904	0.9901	0.71	29			
2	0.3904		0.5534	0.07	03			
3	0.9901	0.5534		0.53	27			
4	0.7129	0.0703	0.5327					
		The SAS System	18:34	Sunday,	October	8,	2017	58

#### The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Ni LSMEAN	LSMEAN Number
1	9.3288750	1
2	10.5756250	2
3	12.7145000	3
4	11.8037500	4
5	10.3523750	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Ni

i/j	1	2	3	4	5	
1		0.5441	0.0038	0.0432	0.7511	
2	0.5441		0.0983	0.5579	0.9989	
3	0.0038	0.0983		0.7874	0.0798	
4	0.0432	0.5579	0.7874		0.4558	
5	0.7511	0.9989	0.0798	0.4558		
		The SA	AS System	18:34 Sunday,	October 8, 2017	59

#### The GLM Procedure

#### Class Level Information

Class	Levels	vels Values				
Sampling_Date		4	1	2	3	4
Station	5	1 2	3	4	5	

#### Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 60

#### The GLM Procedure

#### Dependent Variable: Co Co

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	19	4.71241444	0.24802181	9.65	<.0001
Error	19	0.48851900	0.02571153		
Corrected Total	38	5.20093344			

Coeff Var Root MSE R-Square Co Mean

0.906071 106.7163 0.160348 0.150256 Source DF Type I SS Mean Square F Value Pr > F 
 3
 0.42176928
 0.14058976
 5.47
 0.0070

 1.22506957
 0.30626739
 11.91
 <.0001</td>

 12
 3.06557559
 0.25546463
 9.94
 <.0001</td>
 Sampling\_Date Station 4 Sampling\_Date\*Station Type III SS DF Mean Square F Value Pr > F Source 
 3
 0.42187277
 0.14062426
 5.47
 0.0070

 1.21108958
 0.30277240
 11.78
 <.0001</td>

 12
 3.06557559
 0.25546463
 9.94
 <.0001</td>

 The SAS System
 18:34
 Sunday, October 8, 2017
 61
 Sampling\_Date 4 Station Sampling\_Date\*Station The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer LSMEAN Co LSMEAN Sampling Date Number 0.07640000 1 1 0.14020000 0.05370000 2 2 3 3 0.31570000 4 4 Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j) Dependent Variable: Co 2 3 i/j 1 4 0.9887 0.6641 0.8308 0.0167 1 2 0.8308 0.1255 3 0.9887 0.6641 0.0084 4 0.0167 0.1255 0.0084 18:34 Sunday, October 8, 2017 62 The SAS System The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer LSMEAN Co LSMEAN Station Number 0.06150000 1 1 2 0.09550000 2

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

0.49350000

0.08200000

0.00000000

3

4

5

3

4

5

#### Dependent Variable: Co

i/j	1	2	3	4	5	
1		0.9927	0.0003	0.9990	0.9485	
2	0.9927		0.0007	0.9998	0.7926	
3	0.0003	0.0007		0.0005	0.0001	
4	0.9990	0.9998	0.0005		0.8676	
5	0.9485	0.7926	0.0001	0.8676		
		The SA	AS System	18:34 Sunday,	October 8,	2017

#### The GLM Procedure

#### Class Level Information

#### Class Levels Values

#### Sampling\_Date 4 1 2 3 4

63

Station 5 1 2 3 4 5

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 64

The GLM Procedure

Dependent Variable: Zn Zn

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	19	88168.4780	4640.4462	1.17	0.3687
Error	19	75435.4803	3970.2884		
Corrected Total	38	163603.9583			

R-Square	Coeff Var	Root MSE	Zn Mean
0.538914	47.59707	63.01022	132.3826

132.3826

Source	DF		Type I SS	Mean	Square 1	7 Value	Pr >	F
Sampling_Date Station	4	3	4209.94949 32294.14010	8073	1403.31650 3.53503	0. 2.03	35 C 0.130	.7872 3
Sampling_Date*Station		12	51664.38845		4305.36570	1.	08 C	.4235
Source	DF		Type III SS	Mean	Square 1	7 Value	Pr >	F

			11	1			
Sampling Date		3	4211.96769	1403.98923	0.35	0.7870	
Station _	4		28409.23062	7102.30766	1.79 0.	1728	
Sampling Date*Station		12	51664.38845	4305.36570	1.08	0.4235	
-			The SAS System	18:34 Sunda	y, October	8, 2017	65

The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

		LSMEAN
Sampling_Date	Zn LSMEAN	Number
1	148.941000	1
2	133.149000	2
3	124.542000	3
4	123.160000	4

Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Zn

i/j	1	2	3		4			
1		0.9496	0.8222	0.79	72			
2	0.9496		0.9911	0.98	63			
3	0.8222	0.9911		1.00	00			
4	0.7972	0.9863	1.0000					
		The SAS System	18:3	4 Sunday,	October	8,	2017	66

### The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Zn LSMEAN	LSMEAN Number
1	170.130000	1
2	153.472500	2
3	106.540000	3
4	135.358750	4
5	96.738750	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Zn

i/j	1	2	3	4	5	
1		0.9832	0.2951	0.8025	0.2233	
2	0.9832		0.5810	0.9772	0.4584	
3	0.2951	0.5810		0.8877	0.9982	
4	0.8025	0.9772	0.8877		0.7754	
5	0.2233	0.4584	0.9982	0.7754		
		The SA	AS System	18:34 Sunday,	October 8, 2	2017 67

The GLM Procedure

Class Level Information

Class	Levels	Vá	alı	les	3		
Sampling_Date		41	2	3	4		
Station	5	1	2	3	4	5	

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 68

The GLM Procedure

Dependent Variable: As As

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	19	0	0		
Error	19	0	0		
Corrected Total	38	0			

R-Square	Coeff Var	Root MSE	As Mean
0.000000		0	0

Source	DF		Туре І	SS	Mear	n Square	F	Value	Pr >	F
Sampling_Date Station Sampling_Date*Station	4	3 12		0	0 0	0	0 0			
Source	DF	Т	ype III	SS	Mear	n Square	F	Value	Pr >	F

Sampling Date	3	0	0			
Station	4	0	0			
Sampling_Date*Station	12	0	0			
	The	SAS System	18:34 Sunday,	October 8,	2017	69

#### The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Sampling_Date	As LSMEAN	LSMEAN Number
1	0	1
2	0	2
3	0	3
4	0	4

Least Squares Means for effect Sampling\_Date
 Pr > |t| for H0: LSMean(i)=LSMean(j)

		Depende	nt Variable	: As				
i/j		1	2		3	4		
1								
3	•			•		•		
4		The	SAS System	•	18:34 S	Sunday, Oc	ctober 8, 2017	70
	Adjustment	The G Least for Multip	LM Procedure Squares Mean le Compariso	e ns ons: Tu	key-Kram	ner		
	Stat	tion	As LSMEAN	LSM Num	EAN ber			
	1		0		1			
	2 3		0		2 3			
	4 5		0 0		4 5			
	Least Pr 2	: Squares M >  t  for H	Weans for efi 10: LSMean(i)	fect St )=LSMea	ation n(j)			
		Depende	nt Variable	: As				
i/j	1	2		3		4	5	
1		•	•		•			
3	•		•				•	
4 5	•	•	•				•	
		The	SAS System		18:34 S	Sunday, Oc	ctober 8, 2017	71
		The G	LM Procedure	е				
		Class Le	evel Informat	tion				
	Clas	35	Levels V	Values				
	Sam	pling_Date	4	1234				
	Stat	zion	5 2	1234	5			
		Number of	observations	в 40				
NOTE: Due to missing v	alues, only	39 observa The	tions can be SAS System	e used	in this 18:34 S	analysis. Sunday, Oc	ctober 8, 2017	72
		The G	LM Procedure	e				
Dependent Variable: Sr	Sr							
Source		DF	Sum of Squares	Mean	Square	F Value	e Pr > F	
Model		19 952	1.768429	501.	145707	26.85	5 <.0001	
Error		19 35	4.668344	18.	666755			
Corrected Total		38 987	6.436774					
	R-Square	Coeff Va	r Root	MSE	Sr M	lean		
	0.964089	13.8719	4.32	0504	31.14	1572		
Source		DF T	ype I SS	Mean	Square	F Value	e Pr > F	
Sampling_Date Station Sampling_Date*St	ation	3 4 243 12	194.875742 2.757589 6894.135099	2 608. 9	64.9585 189397 574.5112	32.58 32.58	3.48 0.036 3 <.0001 30.78 <.000	3 1

Source	DF Type	III SS	Mean Square	F Value Pr	c > F
Sampling_Date Station Sampling_Date*Station	3 4 1889 12 The	143.122858 .318109 6894.135099 SAS System	47.70761 472.329527 574.51125 18:34 Sur	2.56 25.30 <. 30.78 nday, October	0.0857 .0001 <.0001 8, 2017 73
Adjus	The GL Least S tment for Multipl	M Procedure quares Means e Comparisor	s is: Tukey-Krame	ŝ	
			LSMEAN		
	Sampling_Date	Sr LSMEA	N Number		
	1 2	9.2990000	1		
	3 3	4.5945000	3		
	4 3	1.4608000	4		
	Least Squares Mea Pr >  t  for HO	ns for effec : LSMean(i)=	t Sampling_Date LSMean(j)	2	
	Dependen	t Variable:	Sr		
i/j	1	2	3	4	
1	0 7404	.7494	0.0577	0.6827	
3	0.0577 (	.3969	0.3909	0.3906	
4	0.6827 0	.9999	0.3906		
	The	SAS System	18:34 Sur	nday, October	8, 2017 74
Adjus	The GL Least S tment for Multipl	M Procedure quares Means e Comparisor	s as: Tukey-Krames	2	
	Station S	r LSMEAN	LSMEAN Number		
	1 26	.7837500	1		
	2 27	.3835000	2		
	3 43 4 35	.6587500	3 4		
	5 24	.9393750	5		
	Least Squares Me Pr >  t  for HO	ans for effe : LSMean(i)=	ect Station LSMean(j)		
	Dependen	t Variable:	Sr		
i/j	1 2	3	3 4	ţ	5
1	0.9986	<.000	1 0.0048	0.9259	9
2 0.998	6	<.0001	0.0088	0.8209	9
4 0.004	8 0.0088	0.012	4	0.0014	1
5 0.925	9 0.8209 The	<.0001 SAS System	0.0014 18:34 Sur	nday, October	8,2017 75
	The GL	M Procedure			
	Class Lev	el Informati	.on		
	Class	Levels Va	lues		
	Sampling_Date	4	1 2 3 4		
	Station	5 1	2 3 4 5		

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 76

The GLM Procedure

Dependent Variable: Cd Cd

Source		DF	Sum of Squares	Mean Square	F Value	Pr > F	
Model		19	0 40132374	0 02112230	1 54	0 1757	
Reven		10	0.05000000	0.01267411	1.54	0.1737	
Error		19	0.25980800	0.0136/411			
Corrected Total		38	0.66113174				
	R-Square	Coeff '	Var Root	MSE Cd	Mean		
	0.607025	36.94	521 0.110	5936 0.31	6513		
Source		DF	Type I SS	Mean Square	F Value	Pr > F	
Sampling_Date Station Sampling_Date*St	ation	3 4 12	0.19496072 0.10711360 0.09924942	2 0.06498 0.02677840 2 0.00827	1.96 1.96 1.96	4.75 0.0123 0.1421 0.60 0.8125	
Source		DF T	ype III SS	Mean Square	F Value	Pr > F	
Sampling_Date Station Sampling_Date*St	cation	3 4 ( 12 Tl	0.20310478 0.09492031 0.09924942 ne SAS System	0.06770 0.02373008 0.00827 18:34	159 1.74 1.74 079 ( Sunday, Oct	4.95 0.0105 0.1838 0.60 0.8125 cober 8, 2017 77	
	Adjustment	The Leas for Mult:	GLM Procedure t Squares Mear iple Compariso	e is ons: Tukey-Kra	mer		
				LSMEAN			
	Samp	ling_Date	Cd LSME	AN Numbe	r		
	1		0.24180000	1			
	3		0.41280000	3			
	4		0.35080000	4			
Least Squares Means for effect Sampling_Date Pr >  t  for H0: LSMean(i)=LSMean(i)							
	Least Pr 3	Squares 1 >  t  for	Means for effe HO: LSMean(i)	ect Sampling_D =LSMean(j)	ate		
	Least Pr :	Squares I >  t  for Depend	Means for effe HO: LSMean(i) dent Variable:	ect Sampling_D =LSMean(j) Cd	ate		
i/j	Least Pr :	Squares I >  t  for Depend	Means for effe H0: LSMean(i) dent Variable: 2	ect Sampling_D =LSMean(j) Cd 3	ate 4		
i/j 1	Least Pr	Squares 1 >  t  for Depend 1	Means for effe H0: LSMean(i) dent Variable: 2 0.9999	ect Sampling_D =LSMean(j) Cd 3 0.0193	ate 4 0.1939		
i/j 1 2 3	Least Pr : 2 0.9	Squares I >  t  for Depend 1 999	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298	ct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298	ate 4 0.1939 0.2481 0.6429		
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1	Squares 1 >  t  for Depend 1 999 193 939	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481	<pre>cct Sampling_D =LSMean(j) Cd</pre>	4 0.1939 0.2481 0.6429		
i/j 1 2 3 4	Least Pr 5 2 0.9 3 0.0 4 0.1	Squares I >  t  for Depend 1 999 193 939 TJ	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System	ect Sampling_D =LSMean(j) Cd 0.0193 0.0298 0.6429 18:34	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1	Squares I >  t  for Depend 1 999 193 939 Tl The Leas	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System GLM Procedure t Squares Mear	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1 Adjustment	Squares N >  t  for Depend 1 999 193 939 The Leas for Mult:	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System GLM Procedure t Squares Mear iple Compariso	ect Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 ess pns: Tukey-Kra	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1 Adjustment Sta	Squares N >  t  for Depend 1 999 193 939 The Leas: for Mult: tion	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 ens: Tukey-Kra LSMEAN Number	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1 Adjustment Sta	Squares N >  t  for Depend 1 999 193 939 The Leas for Mult: tion	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN 0.26512500	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 e is ons: Tukey-Kra LSMEAN Number 1	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1 Adjustment Sta 1 2	Squares N >  t  for Depend 1 999 193 939 The Leas for Mult: tion	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.25100000	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 e is ons: Tukey-Kra LSMEAN Number 1 2	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 4 0.1 Adjustment Sta 1 2 3	Squares N >  t  for Depend 1 999 193 939 The Leas for Mult: tion	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 he SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.25100000 0.38212500	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 ens: Tukey-Kra LSMEAN Number 1 2 3	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 5 2 0.9 3 0.0 4 0.1 Adjustment Sta 1 2 3 4 5	Squares N >  t  for Depend 1 999 193 939 The Leas: for Mult: tion	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 he SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.25100000 0.38212500 0.31700000	ect Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 9 sons: Tukey-Kra LSMEAN Number 1 2 3 4 5	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 2 0.9 3 0.0 1 0.1 Adjustment Sta 1 2 3 4 5 Leas Pr	Squares I >  t  for Depend 1 999 193 939 The Leas: for Mult. tion t Squares >  t  for	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 he SAS System GLM Procedures t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.3270000 0.31700000 0.31750000 Means for eff H0: LSMean(i)	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 s s s ns: Tukey-Kra LSMEAN Number 1 2 3 4 5 5 cect Station =LSMean(j)	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 3 4	Least Pr 2 0.9 0.0 0 1 0.1 Adjustment Sta 1 2 3 4 5 Leas Pr	Squares N >  t  for Depend 1 999 193 939 The Leas for Mult: tion t Squares >  t  for Depend	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 he SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.35100000 0.38212500 0.31700000 0.34750000 Means for eff H0: LSMean(i) dent Variable:	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 5 sons: Tukey-Kra LSMEAN Number 1 2 3 4 5 Sect Station =LSMean(j) Cd	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4	Least Pr 0.9 0.0 Adjustment Sta 1 2 3 4 5 Leas Pr	Squares I >  t  for Depend 1 999 193 939 The Leas for Mult. tion t Squares >  t  for Depend	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 me SAS System GLM Procedure t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.325100000 0.38212500 0.31700000 0.34750000 Means for eff H0: LSMean(i) dent Variable: 2	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 s s s s s ts LSMEAN Number 1 2 3 4 5 Sect Station =LSMean(j) Cd 3	4 0.1939 0.2481 0.6429 Sunday, Oct	cober 8, 2017 78	
i/j 1 2 3 4 4	Least Pr 2 0.9 0.0 0.1 Adjustment Sta 1 2 3 4 5 Leas Pr 1	Squares I >  t  for Depend 1 999 193 939 The Leas: for Mult: tion t Squares >  t  for Depend 	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 ne SAS System GLM Procedures t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.25100000 0.38212500 0.31700000 0.34750000 Means for eff H0: LSMean(i) dent Variable: 2 0.302	cct Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 ens: Tukey-Kra LSMEAN Number 1 2 3 4 5 Fect Station =LSMean(j) Cd 3 8 0.89	4 0.1939 0.2481 0.6429 Sunday, Oct mer 4	5 0.6778	
i/j 1 2 3 4 4 1 2 2	Least Pr : 2 0.9 3 0.0 4 0.1 Adjustment Sta 1 2 3 4 5 Leas Pr : 1 0.9992 0.0000	Squares N >  t  for Depend 1 999 193 939 The Leas for Mult: tion t Squares >  t  for Depend 2 0.999 2 0.999	Means for effe H0: LSMean(i) dent Variable: 2 0.9999 0.0298 0.2481 he SAS System GLM Procedures t Squares Mear iple Compariso Cd LSMEAN 0.26512500 0.25100000 0.38212500 0.31700000 0.34750000 Means for eff H0: LSMean(i) dent Variable: 2 2 0.302 0.206	ect Sampling_D =LSMean(j) Cd 3 0.0193 0.0298 0.6429 18:34 5 sons: Tukey-Kra LSMEAN Number 1 2 3 4 5 Sect Station =LSMean(j) Cd 3 28 0.89 59 0.78	4 0.1939 0.2481 0.6429 Sunday, Oct mer 4	5 0.6778 0.571	

0.7973 0.9795 0.8981 0.7896 0.9872 4 0.9872 5 0.6778 0.5411 The SAS System 18:34 Sunday, October 8, 2017 79 The GLM Procedure Class Level Information Levels Values Class Sampling Date 4 1 2 3 4 5 1 2 3 4 5 Station Number of observations 40 NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 80 The GLM Procedure Dependent Variable: Ba Ba Sum of F Value Pr > F Source DF Squares Mean Square 7509.320847 395.227413 80.66 Model <.0001 19 19 93.100450 4.900024 Error 7602.421297 Corrected Total 38 Coeff Var Root MSE R-Square Ba Mean 0.987754 6.679592 2.213600 33.13974 Type I SS Mean Square F Value Pr > F Source DF 3 1828.493749 609.497916 124.39 <.0001 Sampling\_Date 
 1960.283061
 490.070765
 100.01
 <.0001</th>

 12
 3720.544037
 310.045336
 63.27
 <.0001</td>
 Station 4 Sampling\_Date\*Station DF Type III SS Mean Square F Value Pr > F Source 129.80 <.0001 Sampling\_Date 3 1908.052679 636.017560 466.311904 95.17 <.0001 310.045336 63.27 <.0001 18:34 Sunday, October 8, 2017 81 4 1865.247614 Station Sampling Date\*Station 12 3720.544037 The SAS System The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer LSMEAN Sampling\_Date Ba LSMEAN Number 24.7770000 1 1 36.6350000 43.2930000 2 2 3 3 30.2340000 4 4 Least Squares Means for effect Sampling\_Date
Pr > |t| for H0: LSMean(i)=LSMean(j) Dependent Variable: Ba 3 1 2 i/j 4 <.0001 <.0001 0.0001 1 2 <.0001 <.0001 <.0001 <.0001 3 <.0001 <.0001 <.0001 4 0.0001 <.0001 18:34 Sunday, October 8, 2017 82 The SAS System The GLM Procedure

Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Ba LSMEAN	LSMEAN Number
1	29.0150000	1
2	27.8762500	2
3	46.8775000	3
4	31.8850000	4
5	33.0200000	5

Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

Dependent Variable: Ba

i/j	1	2	3	4	5	
1		0.8390	<.0001	0.1116	0.0216	
2	0.8390		<.0001	0.0138	0.0026	
3	<.0001	<.0001		<.0001	<.0001	
4	0.1116	0.0138	<.0001		0.8666	
5	0.0216	0.0026	<.0001	0.8666		
		The SA	AS System	18:34 Sunday,	October 8, 20	17 83

#### The GLM Procedure

Class Level Information

Class	Levels	Valu			
Sampling_Date		4	1 2	3	4
Station	5	1 2	34	5	

Number of observations 40

NOTE: Due to missing values, only 39 observations can be used in this analysis. The SAS System 18:34 Sunday, October 8, 2017 84

The GLM Procedure

Dependent Variable: Pb Pb

Source		DF	Sum of Squares	Mean Square	F Value	Pr > F
Model		19	172.0626412	9.0559285	7.47	<.0001
Error		19	23.0411820	1.2126938		
Corrected Total		38	195.1038232			
	R-Square	Coe	ff Var Root	MSE Pb Me	ean	
	0.881903	67	.83505 1.103	1224 1.623	385	
Source		DF	Type I SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	4	3 39.5687633 24.3059398 2 108.1879381	3 13.18958 6.0764849 L 9.015661	78 10.8 5.01 .57.43 <.	38 0.0002 0.0063 0001
Source		DF	Type III SS	Mean Square	F Value	Pr > F
Sampling_Date Station Sampling_Date*St	ation	4	3 43.0242452 25.6547376 2 108.1879381 The SAS System	2 14.34141 6.4136844 1 9.01566 18:34 St	51 11. 5.29 15 7. unday, Octol	83 0.0001 0.0049 43 <.0001 ber 8, 2017 85
			The GLM Procedure	e		

Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

		LSMEAN
Sampling_Date	Pb LSMEAN	Number
1	0.65810000	1
2	2.98340000	2
3	2.54420000	3
4	0.63150000	4

# Least Squares Means for effect Sampling\_Date Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Pb

i/j	1	2	3		4			
1		0.0013	0.0057	0.99	99			
2	0.0013		0.8298	0.00	11			
3	0.0057	0.8298		0.00	51			
4	0.9999	0.0011	0.0051					
		The SAS System	18:34	Sunday,	October	8,	2017	86

#### The GLM Procedure Least Squares Means Adjustment for Multiple Comparisons: Tukey-Kramer

Station	Pb LSMEAN	LSMEAN Number
1	2.47962500	1
2	1.60675000	2
3	1.84037500	3
4	0.21537500	4
5	2.37937500	5

#### Least Squares Means for effect Station Pr > |t| for H0: LSMean(i)=LSMean(j)

#### Dependent Variable: Pb

i/j	1	2	3	4	5
1		0.5238	0.7726	0.0047	0.9998
2	0.5238		0.9927	0.1261	0.6809
3	0.7726	0.9927		0.0559	0.8845
4	0.0047	0.1261	0.0559		0.0115
5	0.9998	0.6809	0.8845	0.0115	