

A Group Research Project entitled:

***“Urban Heat & Greenhouse Gas Trends in Filipino Urban
Areas: Pre- vs. Post-Quarantine”***

A research project presented to the:

INTERNATIONAL VIRTUAL SCIENCE SYMPOSIUM

By:

University of The Philippines’ EEEI - UP StEEEllar

Andrei Joshua A. Alayon, Lara Lista, Yohan Alkayde, Julienne Mikyla Cruz

Research Advisor:

Joan B. Callope

Title: *Urban Heat & Greenhouse Gas Trends in Filipino Cities: Pre- vs. Post-Quarantine*
Proponent's Name: Andrei Joshua A. Alayon, Lara Lista, Yohan Alkayde, Julienne Mikyla Cruz
Research Adviser: Joan B. Callope
School: University of the Philippines

ABSTRACT

This study examines the impact of quarantine on urban heat and greenhouse gas levels in the Philippines. The research measured air pollutants, Carbon Dioxide (CO₂), Carbon Monoxide (CO), Formaldehyde (HCHO), and Total Volatile Organic Compounds (TVOC), along with urban heat parameters before and after quarantine. Air pollution occurs when certain gases and particles, known as pollutants, accumulate in the atmosphere to levels that can cause harm to human health and the environment. The proponents visited several locations, namely IBP Road, San Mateo Bridge, and Don Antonio, to evaluate chemical pollutant levels and urban heat parameters before and after the nationwide lockdown. Meters were placed in these urban locations and exposed for an hour every day for a week. Data analysis involved Pearson's Correlation to test correlations between all sets of data and the T-Test to examine differences in data. The findings indicate a significant decline in greenhouse gas levels (except for HCHO, which showed a slight increase) post-quarantine, with CO₂, CO, and TVOC levels dropping across all test locations. However, despite reduced pollutant emissions, urban heat parameters, particularly surface and air temperatures, increased across all areas. The extraneous cloud data suggest that temperatures were lower on days with more cloud cover than that of when clouds were overcast, which corresponds to the higher humidity levels during those same days. The statistical analysis confirmed notable differences between pre- and post-quarantine data, with 61.47% of measured factors showing high variability. These results suggest that while reduced human activity lowered air pollution, it did not immediately mitigate rising urban temperatures. The study underscores the need for long-term, consistent efforts to combat climate change, as halting emissions alone is insufficient to reverse urban heat effects.

Keywords: Air, Chemical Pollutants, Urban Heat, Quarantine, Greenhouse Gas

Table of Figures

<i>Figure 1: Theoretical Framework</i>	12
<i>Figure 2: CO2 Emissions from fuel combustion in selected economies, 2000 – 2019 (retrieved 2020)</i>	13
<i>Figure 3: Jos Olivier, Jeroen Peters. Trends in Global CO2 and Greenhouse, Gas, Emissions 2012 (retrieved 2020)</i>	14
<i>Figure 4: Earth Systems Science Data 11, Top contributors to Carbon Dioxide Emissions, 2019 (retrieved 2020)</i>	14
<i>Figure 3: Research Set-up Table</i>	19
<i>Figure 4: NOAA’s CO2 Monitoring at Mauna Loa Observatory (Scripps Institution of Oceanography, NOAA Global Monitoring Observatory – 1960 to 2021).</i>	21
<i>Figure 5: CO2 Emissions per Country compiled by IEA (2018)</i>	27
<i>Figure 6: Air Quality Research Data Sheet Contents</i>	31
<i>Figure 7: Guidelines for interpreting regression graph (Garip, 2015)</i>	33

Table of Tables

<i>Table 1: Statistical Treatment Formulas</i>	32
<i>Table 2: Guidelines for interpreting correlation coefficient r</i>	33
<i>Table 3: Batasan Hills Overpass (bound BHNHS)</i>	35
<i>Table 4: San Mateo Northview (by the Marikina/ San Mateo River)</i>	44
<i>Table 5: Don Antonio Overpass (bound Shopwise and Petron)</i>	53
<i>Table 6: Batasan Hills Overpass After Quarantine (bound BHNHS)</i>	62
<i>Table 7: San Mateo Northview (by the Marikina/ San Mateo River)</i>	72
<i>Table 8: Don Antonio Overpass (bound Shopwise and Petron)</i>	82

Table of Charts

<i>Chart 1: Batasan Hills Overpass (bound BHNHS) Day 1</i>	36
<i>Chart 2: Batasan Hills Overpass (bound BHNHS) Day 2</i>	36
<i>Chart 3: Batasan Hills Overpass (bound BHNHS) Day 3</i>	37
<i>Chart 4: Batasan Hills Overpass (bound BHNHS) Day 4</i>	37
<i>Chart 5: Batasan Hills Overpass (bound BHNHS) Day 5</i>	38
<i>Chart 6.1: Batasan Hills Overpass Before Quarantine Matrix Plot</i>	42
<i>Chart 6.2 and 6.3: Batasan Hills Overpass Before Quarantine Matrix Plot (Pearson's)</i>	43
<i>Chart 7: San Mateo Northview (by the Marikina/ San Mateo River) Day 1</i>	45
<i>Chart 8: San Mateo Northview (by the Marikina/ San Mateo River) Day 2</i>	45
<i>Chart 9: San Mateo Northview (by the Marikina/ San Mateo River) Day 3</i>	46
<i>Chart 10: San Mateo Northview (by the Marikina/ San Mateo River) Day 4</i>	46
<i>Chart 11: San Mateo Northview (by the Marikina/ San Mateo River) Day 5</i>	47
<i>Chart 12.1: San Mateo Northview Road Before Quarantine Matrix Plot</i>	51
<i>Chart 12.2 and 12.3: San Mateo Northview Road Before Quarantine Matrix Plot</i> <i>(Pearson's)</i>	52
<i>Chart 13: Don Antonio Overpass (bound Shopwise and Petron) Day 1</i>	54
<i>Chart 14: Don Antonio Overpass (bound Shopwise and Petron) Day 2</i>	54
<i>Chart 15: Don Antonio Overpass (bound Shopwise and Petron) Day 3</i>	56
<i>Chart 16: Don Antonio Overpass (bound Shopwise and Petron) Day 4</i>	56
<i>Chart 17: Don Antonio Overpass (bound Shopwise and Petron) Day 5</i>	56
<i>Chart 18.1: Don Antonio Before Quarantine Matrix Plot</i>	60
<i>Chart 18.2 and 18.3: Don Antonio Before Quarantine Matrix Plot (Pearson's)</i>	61

<i>Chart 19: Batasan Hills Overpass (bound BHNHS) Day 1</i>	63
<i>Chart 20: Batasan Hills Overpass (bound BHNHS) Day 2</i>	63
<i>Chart 21: Batasan Hills Overpass (bound BHNHS) Day 3</i>	64
<i>Chart 22: Batasan Hills Overpass (bound BHNHS) Day 4</i>	64
<i>Chart 23: Batasan Hills Overpass (bound BHNHS) Day 5</i>	65
<i>Chart 24.1: Batasan Hills Overpass After Quarantine Matrix Plot</i>	70
<i>Chart 24.2 and 24.3: Batasan Hills Overpass after Quarantine Matrix Plot (Pearson's)</i>	71
.....	
<i>Chart 25: San Mateo Northview (by the Marikina/ San Mateo River) Day 1</i>	73
<i>Chart 26: San Mateo Northview (by the Marikina/ San Mateo River) Day 2</i>	73
<i>Chart 27: San Mateo Northview (by the Marikina/ San Mateo River) Day 3</i>	74
<i>Chart 28: San Mateo Northview (by the Marikina/ San Mateo River) Day 4</i> ...	74
<i>Chart 29: San Mateo Northview (by the Marikina/ San Mateo River) Day 5</i> ...	75
<i>Chart 30.1: San Mateo Northview Road After Quarantine Matrix Plot</i>	80
<i>Chart 30.2 and 30.3: San Mateo Northview Road After Quarantine Matrix Plot (Pearson's)</i>	81
<i>Chart 31: Don Antonio Overpass (bound Shopwise and Petron) Day 1</i>	83
<i>Chart 32: Don Antonio Overpass (bound Shopwise and Petron) Day 2</i>	83
<i>Chart 33: Don Antonio Overpass (bound Shopwise and Petron) Day 3</i>	84
<i>Chart 34: Don Antonio Overpass (bound Shopwise and Petron) Day 4</i>	84
<i>Chart 35: Don Antonio Overpass (bound Shopwise and Petron) Day 5</i>	85
<i>Chart 36.1: Don Antonio After Quarantine Matrix Plot</i>	91
<i>Chart 36.2 and 36.3: Don Antonio After Quarantine Matrix Plot (Pearson's)</i> ...	92
<i>Appendix</i>	127

CHAPTER I

INTRODUCTION

Air pollution occurs when certain gases and particles, known as pollutants, accumulate in the atmosphere to levels that can cause harm to human health and the environment. Some of the most common pollutants include Carbon dioxide (CO₂), Carbon monoxide (CO), Formaldehyde (HCHO), and Total Volatile Organic Compounds (TVOC).

In December 2019, COVID-19 caused a pandemic and claimed over 6 million lives worldwide.¹ However, although the disease has caused health concerns and economic crises, it proved advantageous to the planet environmentally as it kept people inside due to lockdowns. For instance, scientists at NASA's Ames Research Center discovered that emptier parking lots and cleaner air from fewer vehicles meant that solar heat radiating off dark asphalt and cement surfaces did not stay trapped near the ground which cooled the urban environment. Moreover, scientists found that large parking lots, highway corridors, and commercial rooftops were 10-15°F cooler from March to May 2020 compared to pre-pandemic data².

Before quarantine, scientists discovered an increase in harmful greenhouse gases globally, mainly Carbon dioxide (CO₂). CO₂ is a colorless, non-combustible gas mostly created by animal biological functions and petroleum derivatives³. CO₂ levels normally range from 300-400 ppm yet can be as high as 600-900 ppm in metropolitan regions. Using the Intergovernmental Panel on Climate Change (IPCC), information on CO₂ emissions from 1950 to recent years was used to estimate data until 2050. Unfortunately, the results of this endeavor showed the quickly rising levels of CO₂ gas as time went on.⁴

¹ W.H.O., 2022

² www.bgr.com, retrieved December 18, 2020

³ FSIS, retrieved March 19, 2021

⁴ Richard Schmalensee, Thomas M. Stoker, and Ruth A. Judson (retrieved 2021)

Methane (CO) is also a greenhouse gas. It can cause multiple respiratory diseases with extended exposure as it displaces oxygen in the blood. Another harmful chemical pollutant is Formaldehyde (HCHO). HCHO is used predominantly as a chemical intermediate. Major toxic effects caused by formaldehyde exposure are eye, nose, nasal cavity, and throat irritation. Other complications include coughing, wheezing, chest pains, and bronchitis.⁵ Lastly, Total Volatile Organic Compounds (TVOC) are chemicals that contain carbon and are found in all living things. Many VOCs are classified as known carcinogens, irritants, and toxicants.

During the pre-pandemic period, these gases played a role in global warming as they accumulated under the stratospheric ozone layer — increasing global temperatures and creating urban heat islands (UHI), which is considered one of the serious issues in the 21st century.⁶ Urban settings experience higher temperatures due to decreased vegetation, infrastructure geometry, and multiplied human activity.⁷ In fact, cities are often 18-27°F hotter than rural places, which is exacerbated during nighttime and winters.⁸ As a result, inhabitants of urban settings are the most affected in the growing problem of global warming and climate changes caused by pollutants.

As the population increases, analysts are presently thinking that urbanization further could influence worldwide outflows of CO₂ in the next 40 years.⁹ By mid-century, researchers estimated that the global population could rise by more than 3 billion people, with most of the increase occurring in urban areas. Generally, specialists found that if the populace follows one of the slower development ways predicted by United Nations demographers, there could be a 16-29% decrease in CO₂ emissions by 2050, thereby combating rising temperatures worldwide.

⁵ EPA, retrieved 2021

⁶ Oke, T. (n.d). 2018

⁷ EPA, retrieved 2025

⁸ EU Science Hub, retrieved 2025

⁹ Proceedings of the National Academy of Sciences, 2010

Statement of the Problem

The following are questions to be answered throughout the research and by the project's results. The main question to be answered:

1) What are the significant differences between the Urban Heat data from before, a few months after, and a few years after the Community Quarantine?

- This is important in order for us to understand the effect of prolonged decreased use of GHG emitting activities such as the use of vehicles, and other related outdoor activities which could contribute to the rapid climate change shifts.

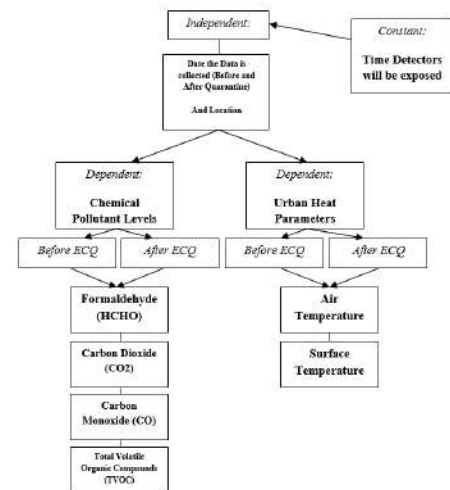
2) Is there a correlation between the Greenhouse Gas data and Urban Heat data? If so, (1) how were they correlated and (2) does this correlation change from before, a few months after, and a few years after the Community Quarantine?

- This would be able to show us the relationship of how the Greenhouse Gases could be contributing to the Urban Heat fluctuations observed through the collection of air and surface temperature data.

3) What are the average and highest levels of Urban Heat and Greenhouse Gas Parameters recorded?

- The data that will answer this question could give us a perspective on how high/low the fluctuations of each parameter mentioned.

From the figure, it is seen that the input or main cause of the creation of this research was from many problems caused by humans. This research model focuses on the rapidly growing urban areas, which, in turn caused a rise in population, resulting in more human activities such as gas combustion, automobile pollution, industrial emissions, increased ecological footprints, and so on. Even though these factors clearly affect us, such as when non-toxic yet



excess greenhouse gases are emitted, warming the planet, humans seem to disregard the destructive by-products of artificial works and overpopulation.

The *carbon footprint* data would reflect the "ignorance of man-made emissions" in the system, while *population* size will define the "rise in development of urban areas." To clarify the above-mentioned terminology, A carbon footprint is the total amount of greenhouse gas (GHG) emissions generated by a person, entity, event, or product, both directly and indirectly.¹⁰ A population is the group of people from which a statistical sample is taken for a study of statistics.¹¹

These two data sets (plus cloud data and relative humidity data as extraneous variables) to be collected from previous data (Carbon Footprints and Population) will be used as the basis for choosing the locations to be tested for. The location will be the main roads in Q.C. where most traffic passes through. These locations were chosen as they were the closest populations to each other, and considering the carbon footprints per location, they were all some of the highest. They were also the most convenient for testing and researching as they did not have strict protocols during the adjustment phase through the end of the quarantine.

Hypothesis

H_a – The Community Quarantine and stop on the GHG emitting activity had an observable effect on the fluctuations of the Urban Heat Parameters and Greenhouse Gas Emissions.

H_o – The Community Quarantine and stop on the GHG emitting activity had no effect on the fluctuations of the Urban Heat Parameters and Greenhouse Gas Emissions.

¹⁰ The Carbon Trust (2012) Carbon Footprinting

¹¹ <https://www.investopedia.com/terms/p/population.asp>, Momoh and Scott, 2021

Scope and Limitation

This study only included gathering data in three main locations in these different locations scattered across Q.C. The study also only considers 3 locations with similar urban development: similar high/low carbon footprints and similar population sizes. The researcher only collected surface temperature and air temperature as parameters for urban heat levels. It also only determines the average levels, highest levels and correlations of the chemical pollutants and urban heat parameters.

To answer the difference and effect of the collected results after the ECQ, data collected from a previous date before the quarantine or lockdown will be used to compare. No other data will be used other than the mentioned dates.

This study will not focus on other chemical pollutant measurements other than Formaldehyde (HCHO), Carbon dioxide (CO₂), Carbon monoxide (CO), and Total Volatile Organic Compounds (TVOC). This study is limited to only outdoor measurements of chemical pollutants and urban heat parameters.

The researcher also only used GLOBE protocols to make the data collection process be a great engaging activity and to help maintain consistency of methods. GLOBE's protocols have also been reviewed and made by professionals, ensuring that the data that will be collected would follow protocols made by those who are knowledgeable in the field.

Chapter 3

METHODOLOGY

This chapter presents the research design selection and the different procedures applied to the field of study. This chapter outlines the way in which research is to be undertaken and, among other things, identifies the methods to be used in it. It also includes an analysis of data with statistical treatment applied which were done for experimentation.

Research Design

The research design relies on *correlation research* which is a form of non-experimental study in which the researcher tests two variables and assesses their statistical relationship (i.e., the correlation) with little to no effort to control for extraneous variables.

This study is quantitative research, recording levels of the urban heat parameters (Carbon dioxide, Formaldehyde, and TVOC) and Urban Heat (Air and Surface Temperature, Humidity). It also delved into applying statistical treatment, specifically mean, measuring the average of the data collected. This research relies on measuring variables using a numerical system, analyzing measurements using a statistical model and reporting observable phenomena via mathematical or computational techniques.

Procedures

Settling Upon the Possible Locations for Testing

The researchers searched for the proposed locations of testing and evaluating the levels of chemical pollutants such as Formaldehyde (HCHO), Carbon dioxide (CO₂), Carbon monoxide (CO), and Total Volatile Organic Compounds (TVOC) and urban heat parameters such as Air Temperature and Surface Temperature. These locations are specifically 3 public roads in Q.C. The researchers also picked the locations of testing based on how urban they are. The researchers chose a busy area, a middle ground, and an urban area with a more rural surrounding for data variation.



IBP Road

IBP Road (14.689908, 121.094561) was chosen as one of the main areas for testing as it was a good middle ground for the locations. It had a similar size of road with the other locations, but still had a few spots with soil and plants, together with more urban buildings and surroundings.



San Mateo Bridge

The **San Mateo bridge** ($14.679415, 121.109877$) was chosen for its similar road size and the traffic that passes through it. It was mainly chosen as the other side of the urban spectrum, which is closer to a flowing river, with more soil and open land with plants and trees, with less buildings but still receives lots of carbon emissions due to the vehicles passing by.



Don Antonio (Holy Spirit Dr)

Don Antonio ($14.678123, 121.082559$) was chosen as it was the most convenient for the researchers and is also the most busy urban area among the three. With it having a similar size with the other roads, it was the perfect area for testing as the side of the urban spectrum with the busiest streets and most amount of urban buildings and activity.

Preparation of Greenhouse Gas Meters

Instruments such as the greenhouse gas; Formaldehyde (HCHO), Carbon dioxide (CO₂), Carbon monoxide (CO), and Total Volatile Organic Compounds (TVOC) meters and Air and Surface Thermometers were used in this study. The GLOBE observer app was also used during the more recent data collection rounds to collect cloud data as extraneous variables for a better understanding of what could've been influencing the parameter fluctuations. All materials were prepared onto the field every time data was collected from the time allocations from the set-up.

Calibration of Instruments

In calibrating the Air thermometer, the following steps were followed from the Air Thermometer Calibration Protocols on the GLOBE website. *If a digital thermometer is to be used, these steps are to be followed; Open the door to the instrument shelter and hang the calibration thermometer, the digital thermometer, and the soil sensor in the instrument shelter so that they have air flow all around them and do not contact the sides of the shelter. Close the door to the instrument shelter. Wait at least an hour and then open the door to the instrument shelter. Make sure that your digital thermometer is displaying the current temperature(s) (Neither 'MAX' or 'MIN' symbols should be displayed on the screen. (GLOBE.gov, 2014)*

If they are, press the MAX/MIN button until they disappear). Read the temperatures reported by the air sensor and the soil sensor of the digital thermometer and record them on a sheet of paper. Close the door of the instrument shelter. Repeat steps 2 to 5 four more times, waiting at least one hour between each set of readings. Try to space out the five sets of readings over as much of a day as possible. (GLOBE.gov, 2014)



detectors were not moved and only stayed in place the whole time they were exposed excluding the infrared sensor. For the Surface temperature data, 9 spots in each location were

chosen as points to be used to collect surface temperature in choice: following GLOBE Protocols.

In preparing the instruments, the sensors were placed beside each other, and the data collected were written down for statistical treatment. For the infrared sensor, a heat glove was made using an oven mitt following the GLOBE protocols and guidelines for Surface Temperature collection. The researcher also planned to have circumstantial information recorded each day. Namely, Relative Humidity, Cloud Formation and Weather (Rainy or Shiny).



The researcher will collect the data by filling up a data sheet prepared by the researcher. It includes spaces to be filled up for the Air Quality Data, Urban Heat Parameter Data and Circumstantial Data. The data collected spanned from February 27, 2018 - March 4, 2018 for the Pre-Quarantine data collection, February 27, 2022 - March 4, 2022 for the post quarantine data collection and February 27, 2025 - March 27, 2025 for the post-recovered quarantine data collection.

Analysis, and Statistical Treatment

The data recorded will go through analysis. This will help answer if the levels went up and down, or if the levels are constant during the hours of testing. The formula of mean will be used for getting the average levels, mode for the most frequent data, while the formula for linear regression will be used for correlating the data. Analyzing the data will also require Pearson's Correlation for testing correlations between ALL sets of data and T-test to test the difference in data.

For the surface temperature data, the researcher will apply mean deviation to get the average of the 9 spots per cycle of collection. Here are the different formulas and statistical instruments used on the data for the average, the most frequent data, and the correlation of each set of data.

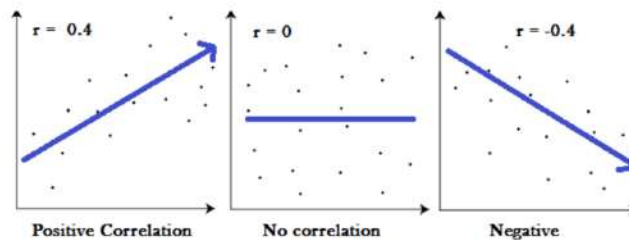
Statistical Treatment	Formula's and Tools to be Used
Mean	$\bar{x} = \frac{\sum x}{N}$ <p>$\sum x$ = the sum of x N = number of data</p> <p>Tool Used: Manual Arithmetic Solving</p>
Mode	<p>N/A (Ungrouped Data)</p> <p>Tool Used: Manual Arithmetic Solving</p>
Linear Regression	$Y_i = \beta_0 + \beta_1 X_i + \epsilon_i$ <p>Tool Used: Minitab 19</p>
Pearson's Correlation	$r = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum (x_i - \bar{x})^2 \sum (y_i - \bar{y})^2}}$ <p>Notes: r = Pearson Correlation Coefficient \sum = summation symbol \bar{x} = variable sample \bar{y} = variable sample \bar{x} = mean of values for x variable \bar{y} = mean of values for y variable</p> <p>Tool Used: Minitab 19</p>
T-Test	$t = \frac{\bar{x} - \mu}{\left(\frac{s}{\sqrt{n}}\right)}, \text{df} = n - 1$ <p>\bar{x} = sample mean difference μ = population mean difference s = sample difference standard deviation n = sample size</p> <p>Tool Used: Minitab 19</p>

The results will be between -1 and 1. You will very rarely see 0, -1 or 1. You'll get a number somewhere in between those values. The closer the value of r gets to zero, the greater the variation the data points are around the line of best fit.

High correlation: .5 to 1.0 or -0.5 to -1.0

Medium correlation: .3 to .5 or -0.3 to -.5

Low correlation: .1 to .3 or -0.1 to -0.3



Correlation Coefficient Scale

+ r values	Positive	- r values	Negative
1.0	Perfect +	-1.0	Perfect -
.8 to .99	Very strong+	-.8 to -.99	Very strong-
.6 to .8	Strong +	-.6 to -.8	Strong -
.4 to .6	Moderate +	-.4 to -.6	Moderate -
.2 to .4	Weak +	-.2 to -.4	Weak -
0 to .2	Very weak +	0 to -.2	Very weak -

CHAPTER IV

RESULTS, DATA ANALYSIS AND CONCLUSION

This chapter presents the data gathered, the results of the statistical analysis done, and interpretation of findings. These are presented in tables following the sequence of the specific research problem regarding the Correlation in between Chemical Pollutants in the air with Urban Heat in Q.C.

Statistical Data and Graphs

The data collected were put into statistical treatment, specifically the Mean, used for getting the average scores per parameter, the Mode, used to get the highest level of parameter taken, and for testing the cause-and-effect of relationship of the two variables, no statistical treatment will be applied because according to McMahon(2020), There is no statistical analysis, by itself, that will demonstrate a cause and effect relationship.

The data recorded went through analysis. This helped answer if the levels went up and down, or if the levels are constant during the hours of testing. The formula of mean is used for getting the average levels, while the formula for linear regression is used for correlating the data. Analyzing the data also requires a T-Test to compare the data from before and after the quarantine.

Before ECQ Data

The following data represents the data collected from before the quarantine. This is prior to the data collection taking place during ECQ. This was on February 27, 2018 – March 4, 2018.

Table 3: Batasan Hills Overpass (bound BHNHS)

Parameter	Mean (Average)	Mode (Highest)
Surface Temperature	40.05 C°	48.2 C°
Air Temperature	31.566 C°	35 C°
Carbon Dioxide (CO ₂)	53.70ppm	58.90ppm
Carbon Monoxide (CO)	85ppm	90ppm
Humidity	52.4%	61%
HCHO (Formaldehyde)	0.413033ppm	0.477ppm
TVOC (Volatile Organic Compounds)	0.8307ppm	1.071ppm

Chart 1: Batasan Hills Overpass (bound BHNHS) Day 1

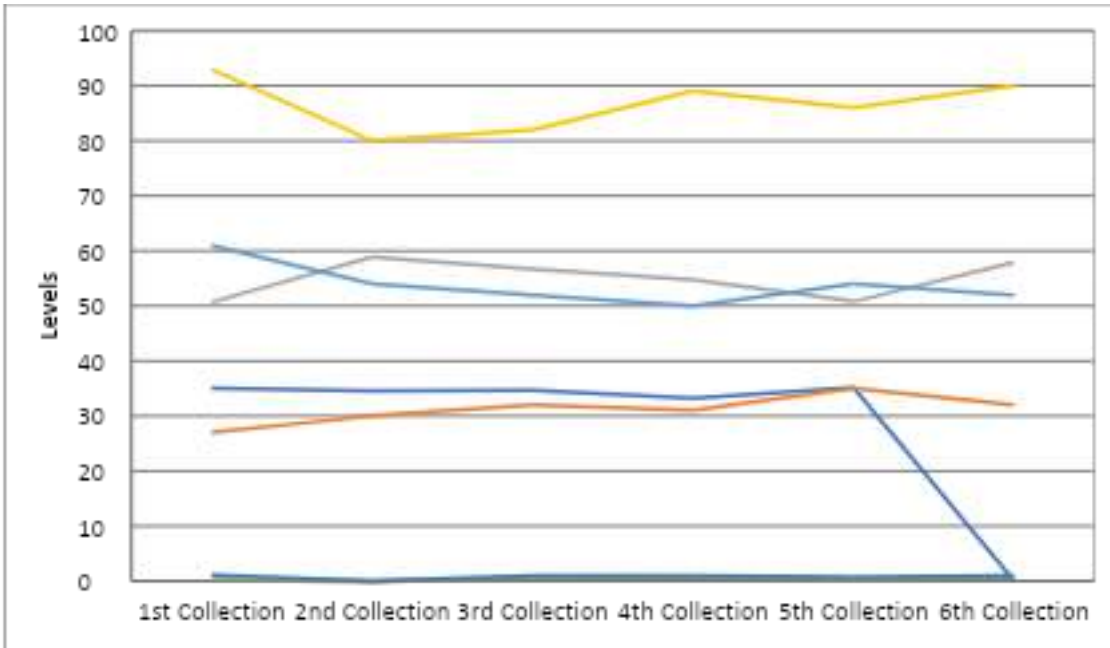


Chart 2: Batasan Hills Overpass (bound BHNHS) Day 2

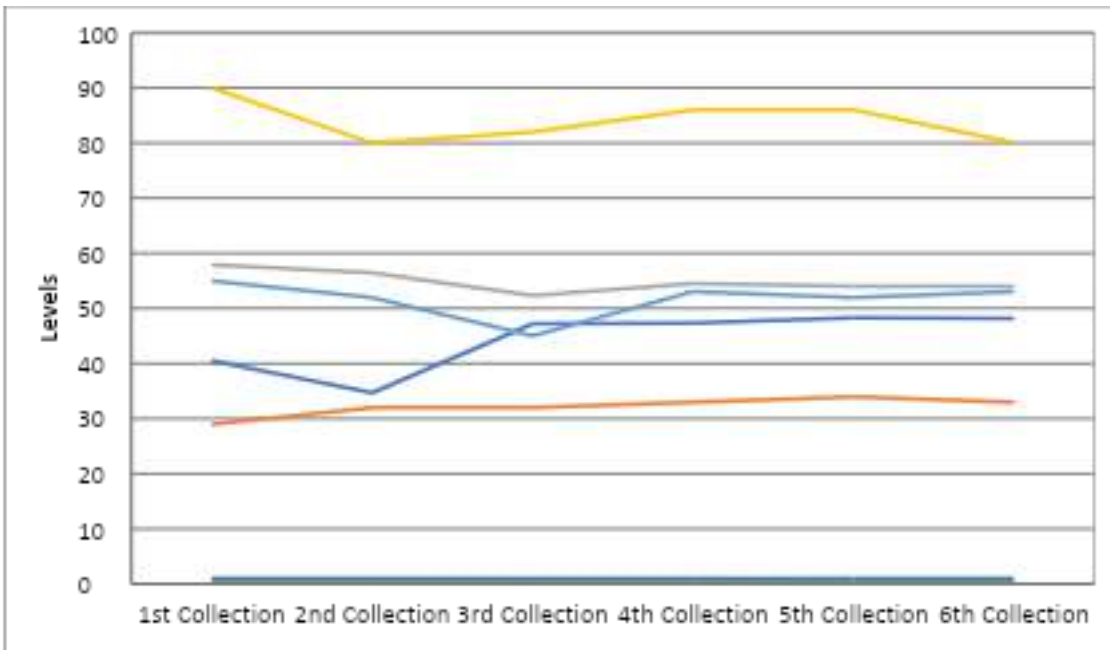


Chart 3: Batasan Hills Overpass (bound BHNHS) Day 3

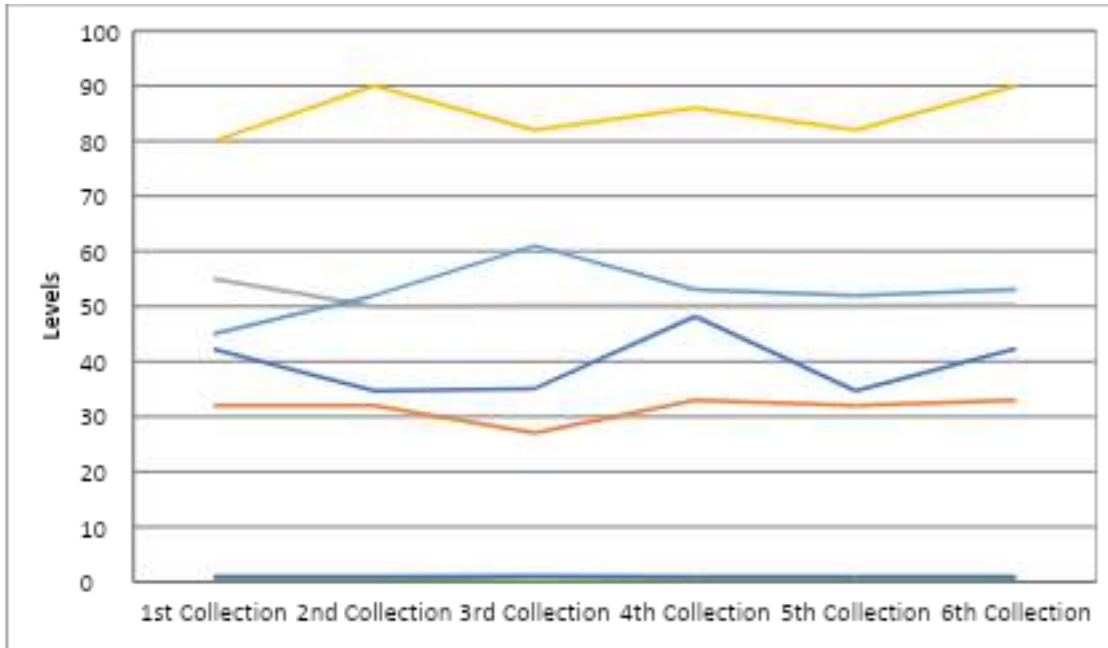


Chart 4: Batasan Hills Overpass (bound BHNHS) Day 4

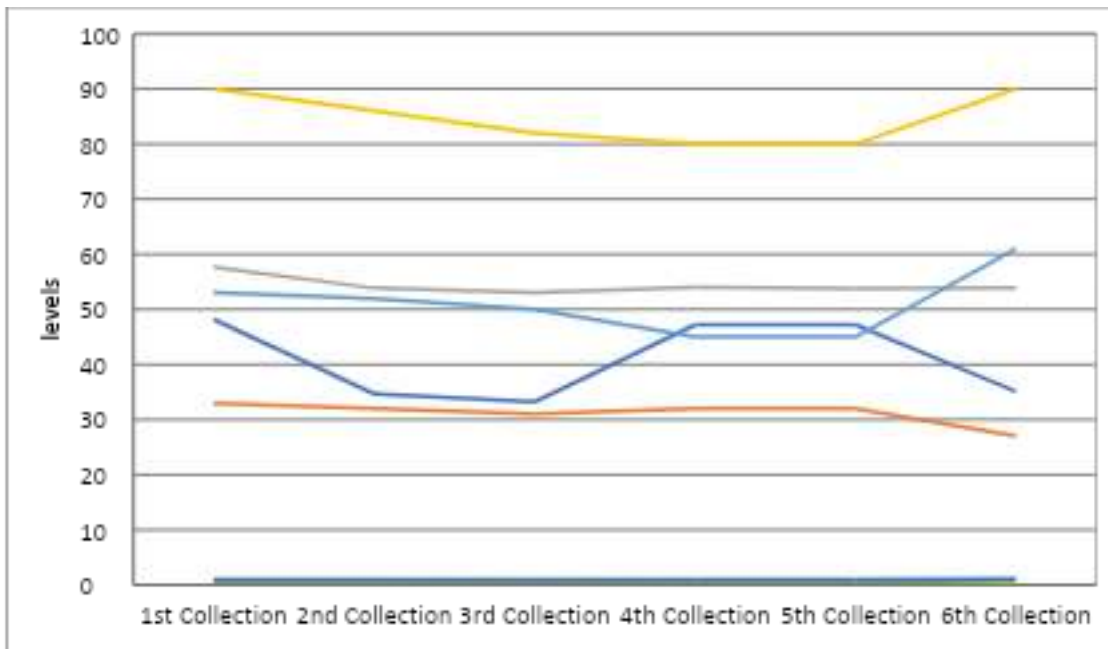
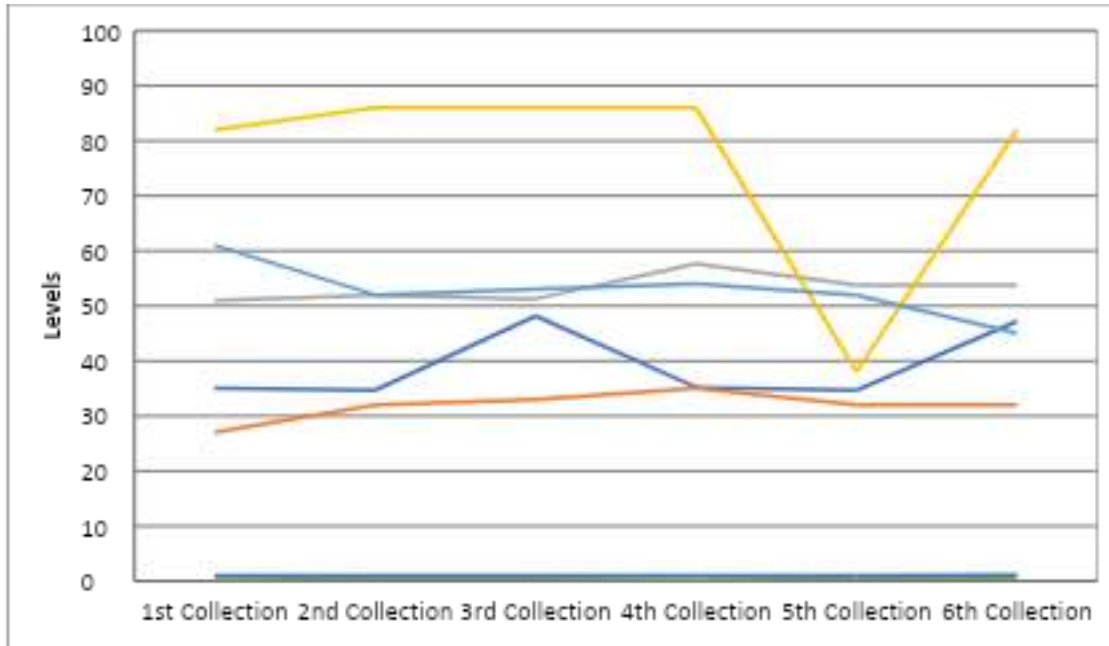


Chart 5: Batasan Hills Overpass (bound BHNHS) Day 5



Formaldehyde (HCHO) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{HCHO} = -0.411 + 0.00226 \text{ ST} + 0.0231 \text{ AT}$$

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.084508	0.042254	2.19	0.131
ST	1	0.005013	0.005013	0.26	0.614
AT	1	0.050765	0.050765	2.64	0.116
Error	27	0.519917	0.019256		
Lack-of-Fit	11	0.519917	0.047265	*	*
Pure Error	16	0.000000	0.000000		
Total	29	0.604425			

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	-0.411	0.410	-1.00	0.325	
ST	0.00226	0.00444	0.51	0.614	1.22
AT	0.0231	0.0142	1.62	0.116	1.22

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.138767	13.98%	7.61%	0.00%

Fits and Diagnostics for Unusual Observations

Obs	HCHO	Fit	Resid	Std Resid
2	0.0030	0.3607	-0.3577	-2.67 R
4	0.6680	0.3809	0.2871	2.15 R
5	0.2200	0.4779	-0.2579	-2.10 R
21	0.6680	0.3809	0.2871	2.15 R

R Large residual

Total Volatile Organic Compound (TVOC) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{TVOC} = 0.730 + 0.00555 \text{ ST} - 0.0002 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	0.730	0.566	1.29	0.208	
ST	0.00555	0.00612	0.91	0.372	1.22
AT	-0.0002	0.0196	-0.01	0.991	1.22

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.191459	3.55%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.03639	0.018193	0.50	0.614
ST	1	0.03017	0.030172	0.82	0.372
AT	1	0.00001	0.000005	0.00	0.991
Error	27	0.98972	0.036656		
Lack-of-Fit	11	0.94910	0.086282	33.98	0.000
Pure Error	16	0.04063	0.002539		
Total	29	1.02611			

Fits and Diagnostics for Unusual Observations

Obs	TVOC	Fit	Resid	Std Resid
2	0.0000	0.9145	-0.9145	-4.95 R

R Large residual

Carbon Monoxide (CO) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{CO} = 93.2 - 0.114 \text{ ST} - 0.112 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	93.2	11.7	7.98	0.000	
ST	-0.114	0.126	-0.90	0.376	1.22
AT	-0.112	0.406	-0.28	0.784	1.22

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
3.95602	4.72%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	20.915	10.457	0.67	0.521
ST	1	12.656	12.656	0.81	0.376
AT	1	1.196	1.196	0.08	0.784
Error	27	422.552	15.650		
Lack-of-Fit	11	165.728	15.066	0.94	0.531
Pure Error	16	256.824	16.051		
Total	29	443.467			

Carbon Dioxide (CO₂) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{CO}_2 = 5569 + 0.72 \text{ ST} - 7.0 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	5569	857	6.50	0.000	
ST	0.72	9.27	0.08	0.939	1.22
AT	-7.0	29.8	-0.24	0.816	1.22

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
290.048	0.21%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	4697	2348.6	0.03	0.972
ST	1	505	505.2	0.01	0.939
AT	1	4647	4646.7	0.06	0.816
Error	27	2271444	84127.6		
Lack-of-Fit	11	998802	90800.2	1.14	0.394
Pure Error	16	1272642	79540.1		
Total	29	2276141			

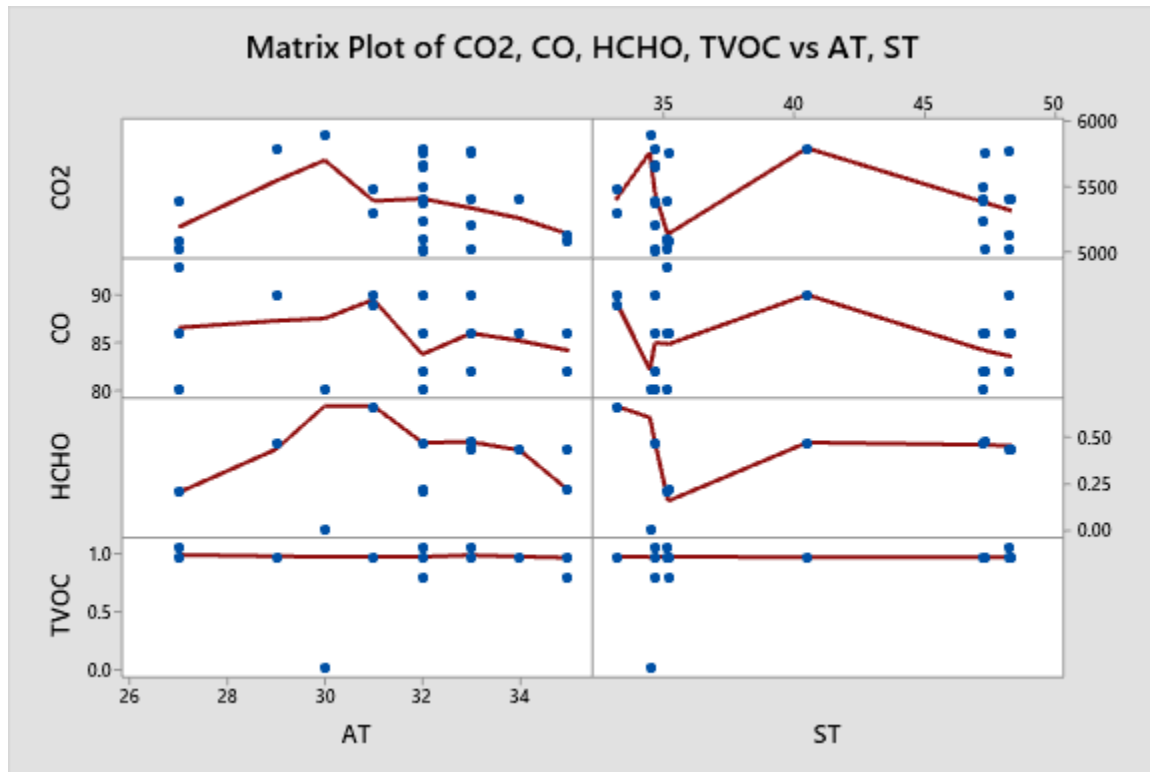
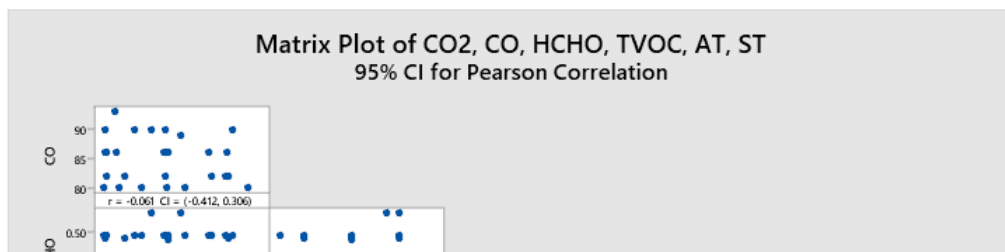


Chart 6.1: *Batasan Hills Overpass Before Quarantine Matrix Plot*



Method

Correlation type Pearson
Rows used 30

Correlations

	CO	HCHO	TVOC	AT
HCHO	-0.239			
TVOC	-0.070	0.501		
AT	-0.244	-0.103	-0.482	
ST	-0.042	0.238	0.126	0.183

Chart 6.2 and 6.3: Batasan Hills Overpass Before Quarantine Matrix Plot (Pearson's)

Table 4: San Mateo Northview (by the Marikina/ San Mateo River)

Parameter	Mean (Average)	Mode (Highest)
Surface Temperature	49.55 C ^o	49.3 C ^o
Air Temperature	32.233 C ^o	33 C ^o
Carbon Dioxide (CO2)	51.803ppm	57.76ppm
Carbon Monoxide (CO)	84.53ppm	86ppm
Humidity	49.8%	61%
HCHO (Formaldehyde)	0.41676ppm	0.478ppm
TVOC(Volatile Organic Compounds)	0.8344ppm	0.976ppm

Chart 7: San Mateo Northview (by the Marikina/ San Mateo River) Day 1

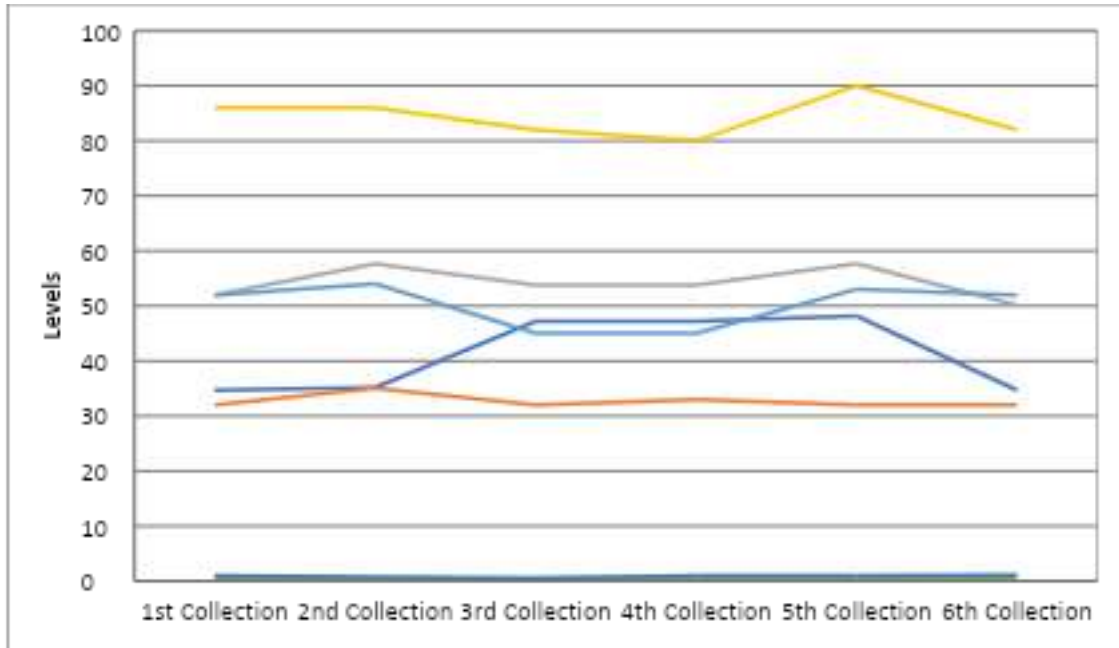


Chart 8: *San Mateo Northview (by the Marikina/ San Mateo River) Day 2*

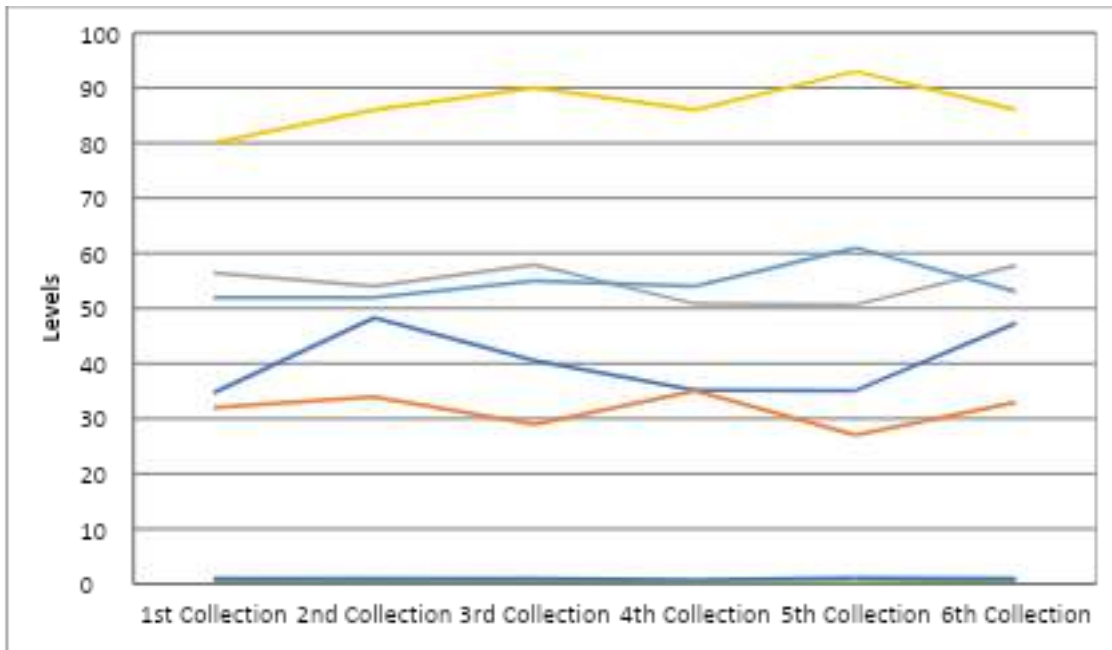


Chart 9: *San Mateo Northview (by the Marikina/ San Mateo River) Day 3*

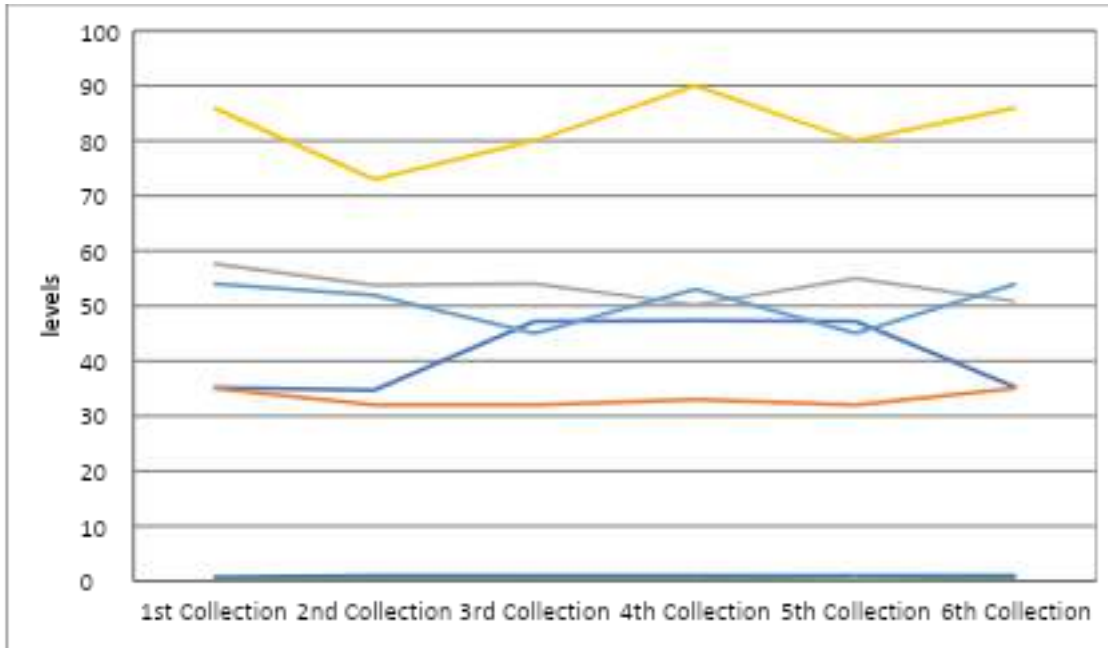


Chart 10: San Mateo Northview (by the Marikina/ San Mateo River) Day 4

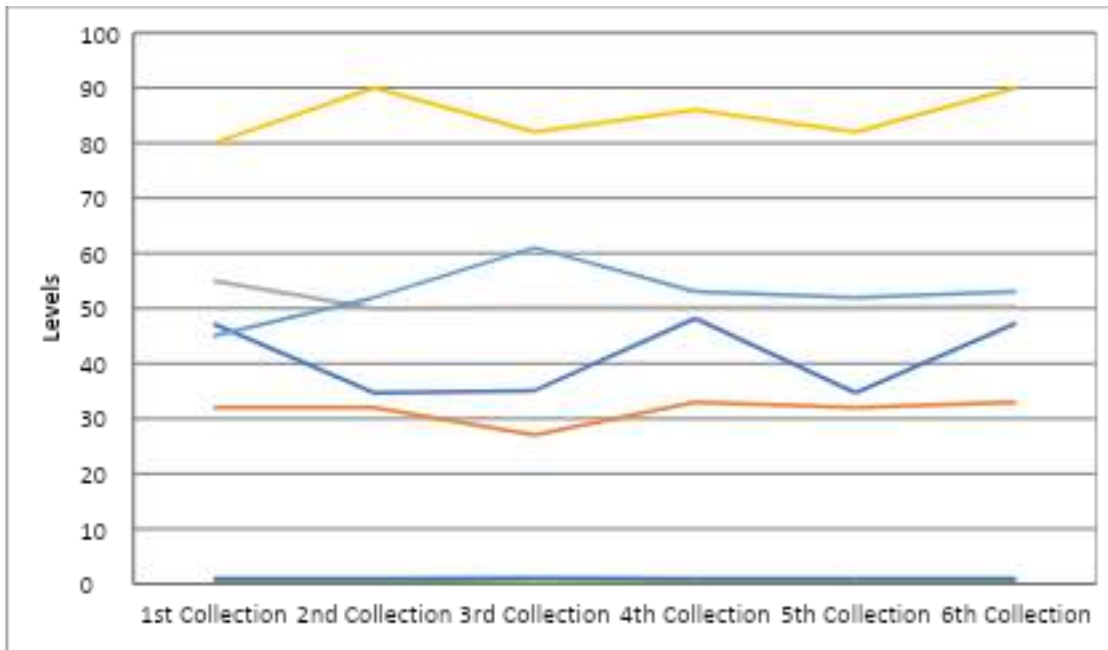
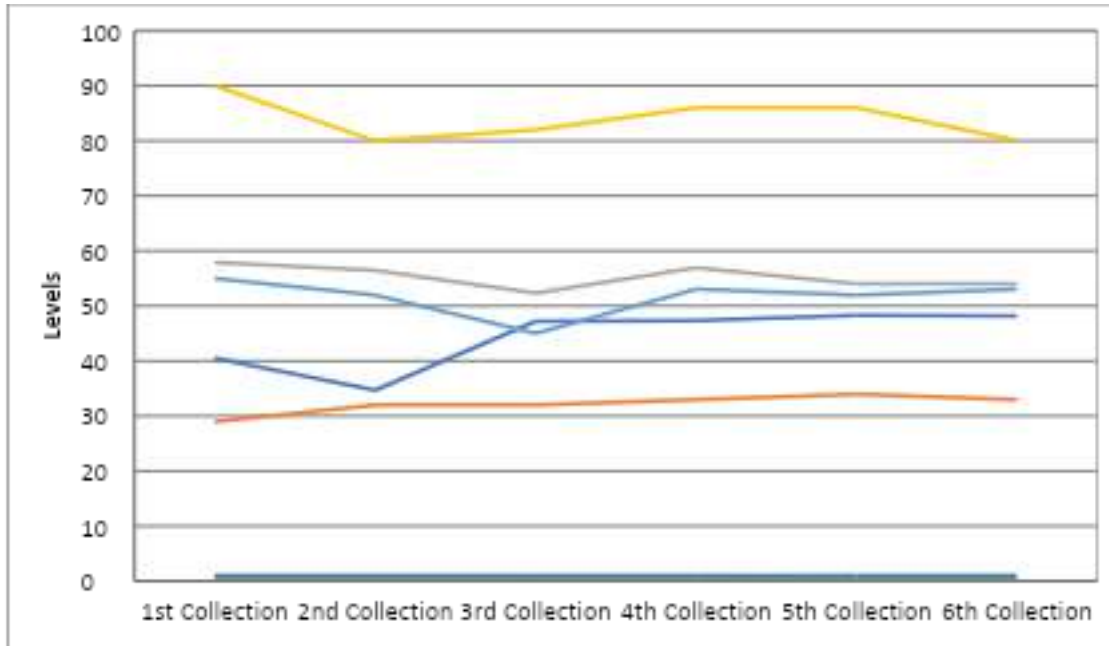


Chart 11: San Mateo Northview (by the Marikina/ San Mateo River) Day 5



Formaldehyde (HCHO) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{HCHO} = 0.502 + 0.00431 \text{ ST} - 0.00776 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	0.502	0.311	1.61	0.118	
ST	0.00431	0.00304	1.42	0.168	1.03
AT	-0.00776	0.00962	-0.81	0.427	1.03

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.102429	7.91%	1.08%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.024320	0.012160	1.16	0.329
ST	1	0.021072	0.021072	2.01	0.168
AT	1	0.006824	0.006824	0.65	0.427
Error	27	0.283274	0.010492		
Lack-of-Fit	18	0.283266	0.015737	17704.11	0.000
Pure Error	9	0.000008	0.000001		
Total	29	0.307593			

Total Volatile Organic Compound (TVOC) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{TVOC} = 1.453 + 0.00244 \text{ ST} - 0.01820 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	1.453	0.187	7.77	0.000	
ST	0.00244	0.00183	1.33	0.193	1.03
AT	-0.01820	0.00578	-3.15	0.004	1.03

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.0615677	28.01%	22.68%	12.05%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.039828	0.019914	5.25	0.012
ST	1	0.006746	0.006746	1.78	0.193
AT	1	0.037578	0.037578	9.91	0.004
Error	27	0.102346	0.003791		
Lack-of-Fit	18	0.102345	0.005686	51172.40	0.000
Pure Error	9	0.000001	0.000000		
Total	29	0.142174			

CO VS AT AND ST

Regression Equation

$$CO = 102.0 + 0.002 ST - 0.544 AT$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	102.0	13.7	7.46	0.000	
ST	0.002	0.134	0.01	0.988	1.03
AT	-0.544	0.423	-1.29	0.209	1.03

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
4.50042	5.95%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	34.613	17.3067	0.85	0.437
ST	1	0.005	0.0045	0.00	0.988
AT	1	33.591	33.5905	1.66	0.209
Error	27	546.853	20.2538		
Lack-of-Fit	18	421.187	23.3993	1.68	0.216
Pure Error	9	125.667	13.9630		
Total	29	581.467			

CO2 VS AT AND ST

Regression Equation

$$\text{CO2} = 4192 + 10.91 \text{ ST} + 23.3 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	4192	907	4.62	0.000	
ST	10.91	8.87	1.23	0.229	1.03
AT	23.3	28.0	0.83	0.413	1.03

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
298.710	8.99%	2.25%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	238008	119004	1.33	0.280
ST	1	135002	135002	1.51	0.229
AT	1	61619	61619	0.69	0.413
Error	27	2409148	89228		
Lack-of-Fit	18	1630769	90598	1.05	0.494
Pure Error	9	778379	86487		
Total	29	2647156			

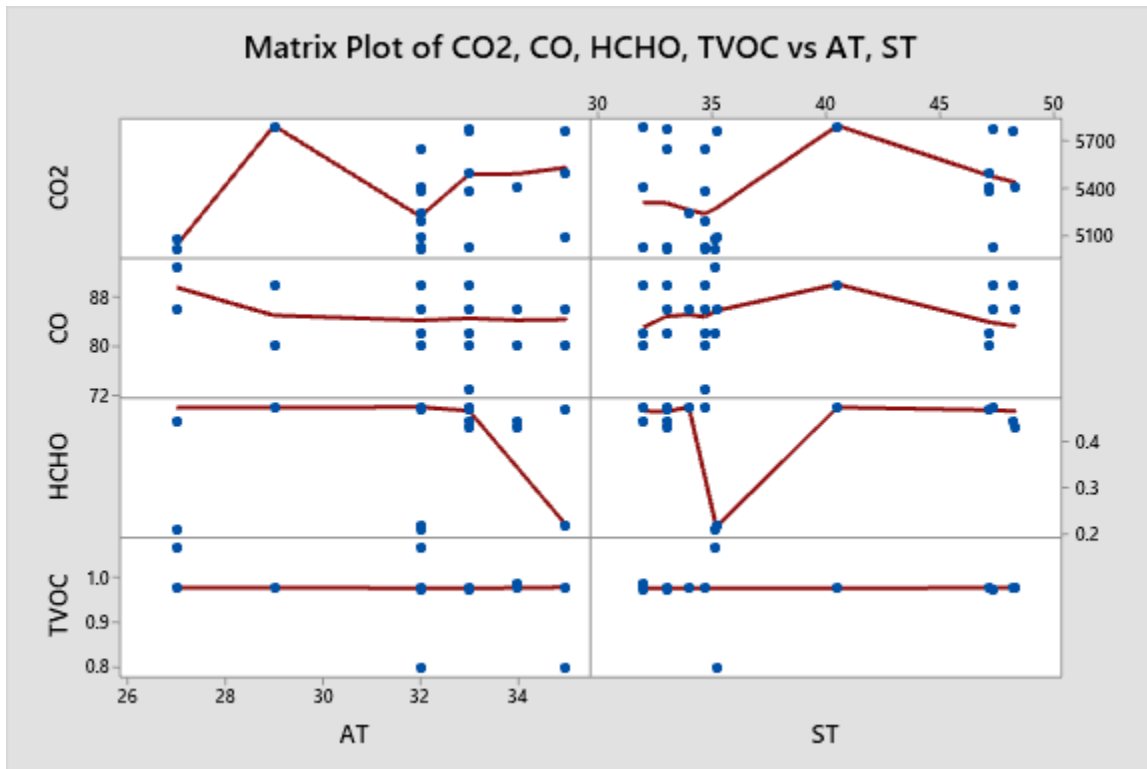
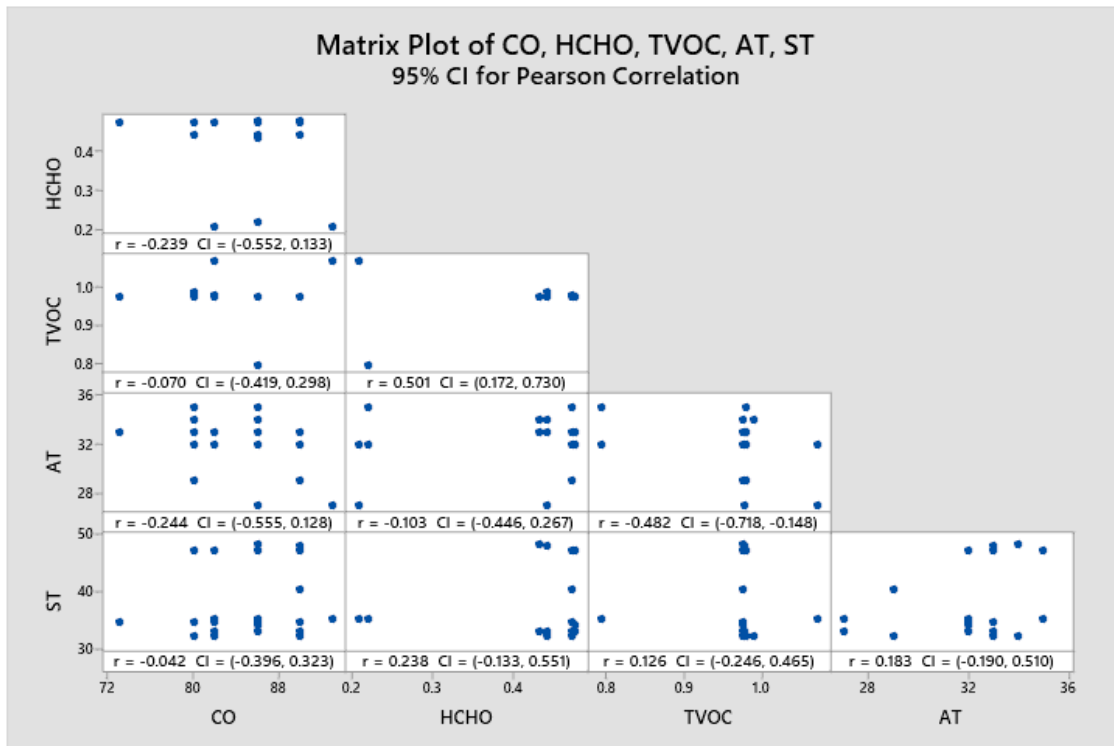


Chart 12.1: San Mateo Northview Road Before Quarantine Matrix Plot



Method

Correlation type Pearson
 Rows used 30

Correlations

	CO	HCHO	TVOC	AT
HCHO	-0.239			
TVOC	-0.070	0.501		
AT	-0.244	-0.103	-0.482	
ST	-0.042	0.238	0.126	0.183

Chart 12.2 and 12.3: *San Mateo Northview Road Before Quarantine Matrix Plot (Pearson's)*

Table 5: Don Antonio Overpass (bound Shopwise and Petron)

Parameter	Mean (Average)	Mode (Highest)
Surface Temperature	41.6133 C ^o	48.2 C ^o
Air Temperature	32.233 C ^o	34 C ^o
Carbon Dioxide (CO2)	53.30ppm	58.90ppm
Carbon Monoxide (CO)	85.09ppm	82ppm
Humidity	46.8%	59%
HCHO (Formaldehyde)	0.4788ppm	0.478ppm
TVOC(Volatile Organic Compounds)	0.834ppm	0.926ppm

Chart 13: Don Antonio Overpass (bound Shopwise and Petron) Day 1

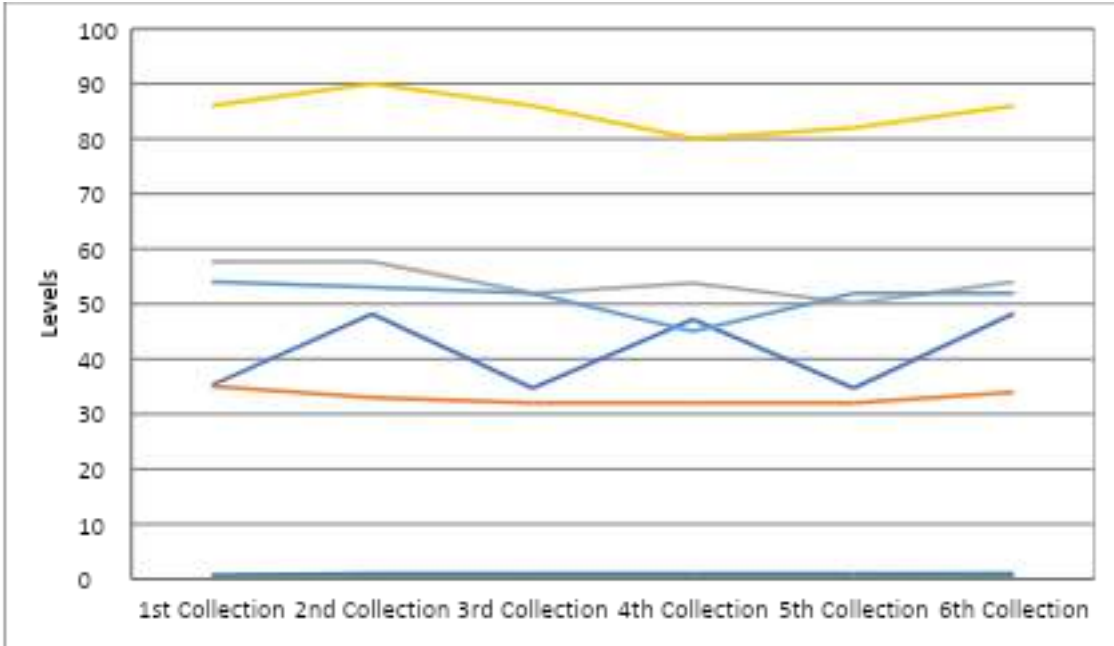


Chart 14: *Don Antonio Overpass (bound Shopwise and Petron) Day 2*

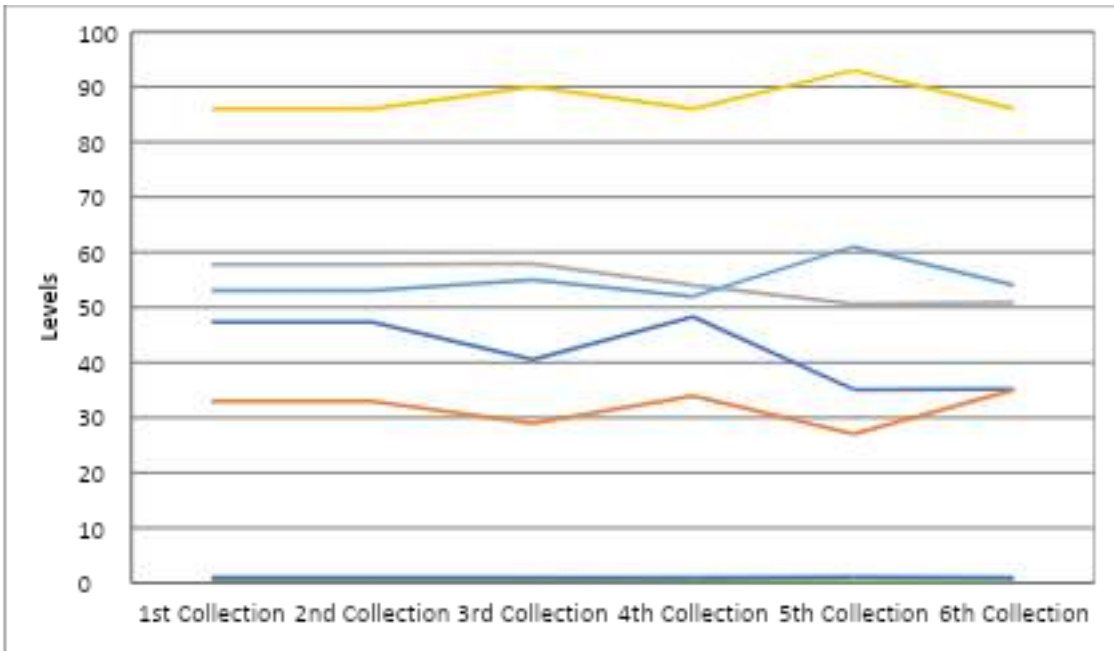


Chart 15: *Don Antonio Overpass (bound Shopwise and Petron) Day 3*

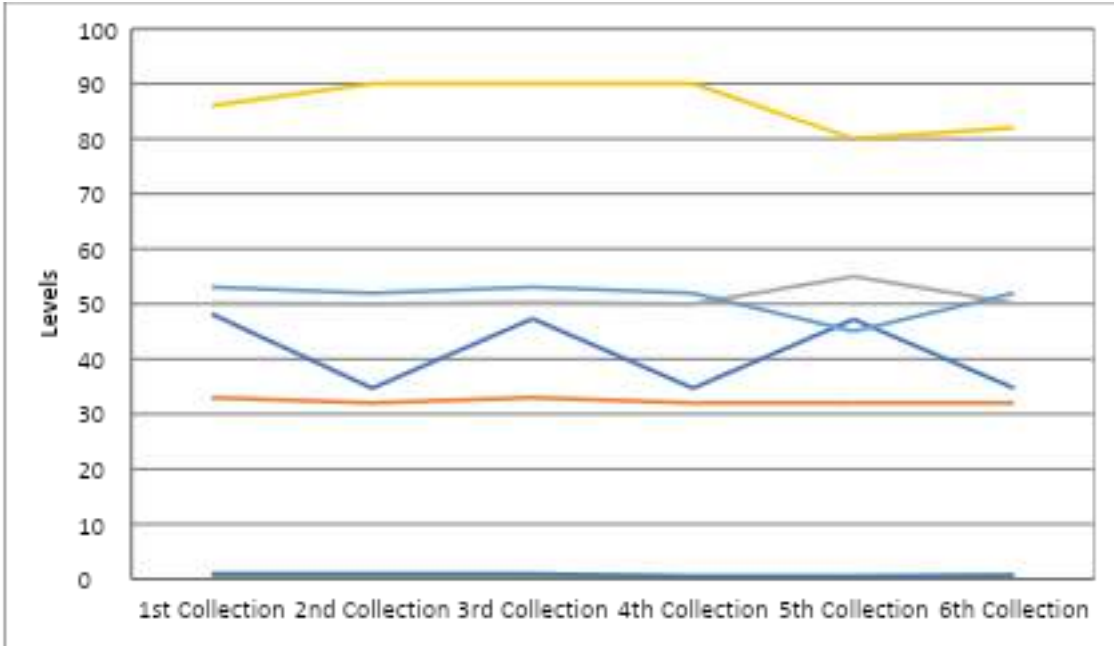


Chart 16: *Don Antonio Overpass (bound Shopwise and Petron) Day 4*

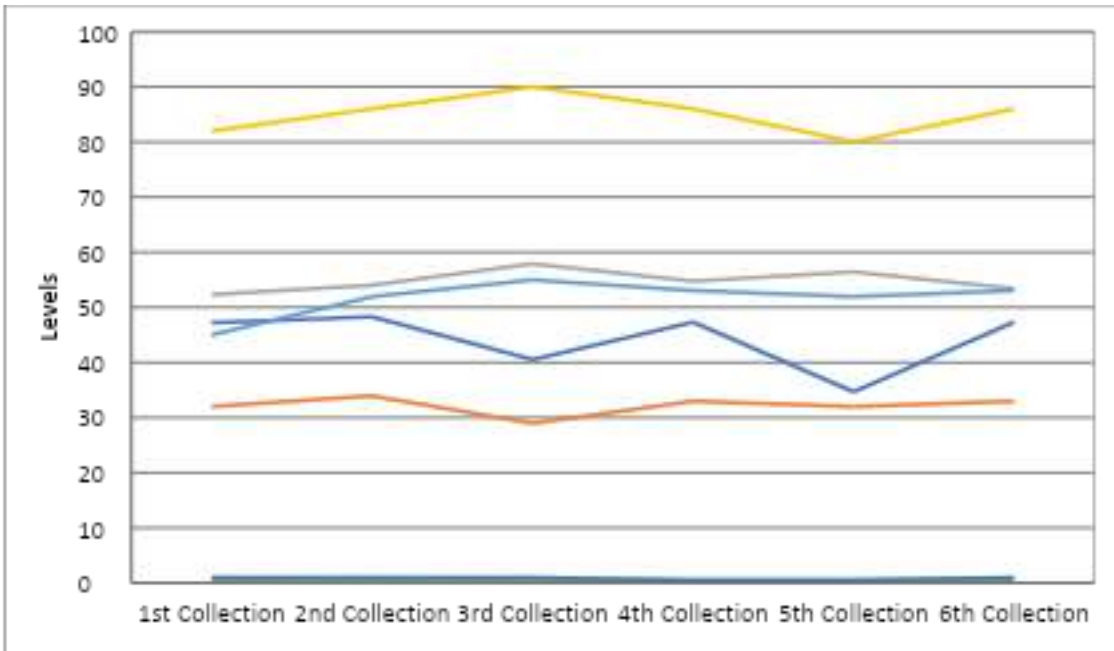
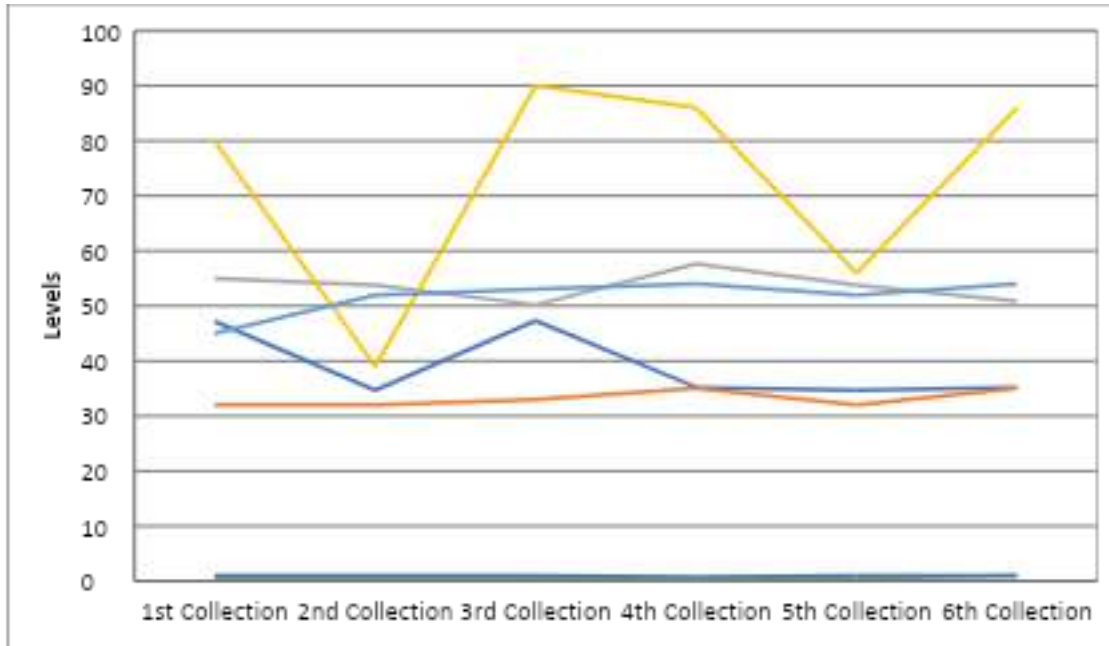


Chart 17: *Don Antonio Overpass (bound Shopwise and Petron) Day 5*



Formaldehyde (HCHO) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{HCHO} = 1.213 - 0.0065 \text{ ST} - 0.0158 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	1.213	0.994	1.22	0.233	
ST	-0.0065	0.0236	-0.28	0.784	1.04
AT	-0.0158	0.0236	-0.67	0.509	1.04

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.351188	2.23%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.07585	0.037927	0.31	0.738
ST	1	0.00945	0.009451	0.08	0.784
AT	1	0.05521	0.055210	0.45	0.509
Error	27	3.32999	0.123333		
Lack-of-Fit	20	2.62865	0.131433	1.31	0.375
Pure Error	7	0.70134	0.100191		
Total	29	3.40584			

Total Volatile Organic Compound (TVOC) vs Air Temperature (AT) and Surface Temperature (ST)

Regression Equation

$$\text{TVOC} = 1.65 - 0.0199 \text{ ST} - 0.0131 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	1.65	1.18	1.40	0.174	
ST	-0.0199	0.0280	-0.71	0.483	1.04
AT	-0.0131	0.0280	-0.47	0.642	1.04

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.416760	3.17%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.15343	0.07672	0.44	0.648
ST	1	0.08804	0.08804	0.51	0.483
AT	1	0.03831	0.03831	0.22	0.642
Error	27	4.68960	0.17369		
Lack-of-Fit	20	3.43856	0.17193	0.96	0.564
Pure Error	7	1.25104	0.17872		
Total	29	4.84303			

CO VS AT AND ST

Regression Equation

$$\text{CO} = 244 + 1.21 \text{ ST} - 5.36 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	244	369	0.66	0.515	
ST	1.21	8.76	0.14	0.891	1.04
AT	-5.36	8.76	-0.61	0.546	1.04

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
130.544	1.37%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	6379	3189.5	0.19	0.830
ST	1	325	325.1	0.02	0.891
AT	1	6369	6369.0	0.37	0.546
Error	27	460124	17041.6		
Lack-of-Fit	20	217884	10894.2	0.31	0.980
Pure Error	7	242240	34605.7		
Total	29	466503			

Fits and Diagnostics for Unusual Observations

Obs	CO	Fit	Resid	Std Resid
16	778.0	117.4	660.6	5.20 R

R Large residual

CO2 VS AT AND ST

Regression Equation

$$\text{CO2} = 2249 + 3.08 \text{ ST} + 91.33 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	2249	396	5.68	0.000	
ST	3.08	9.38	0.33	0.745	1.04
AT	91.33	9.38	9.73	0.000	1.04

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
139.799	78.66%	77.08%	72.82%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	1944965	972483	49.76	0.000
ST	1	2107	2107	0.11	0.745
AT	1	1851520	1851520	94.74	0.000
Error	27	527681	19544		
Lack-of-Fit	20	375646	18782	0.86	0.630
Pure Error	7	152035	21719		
Total	29	2472646			

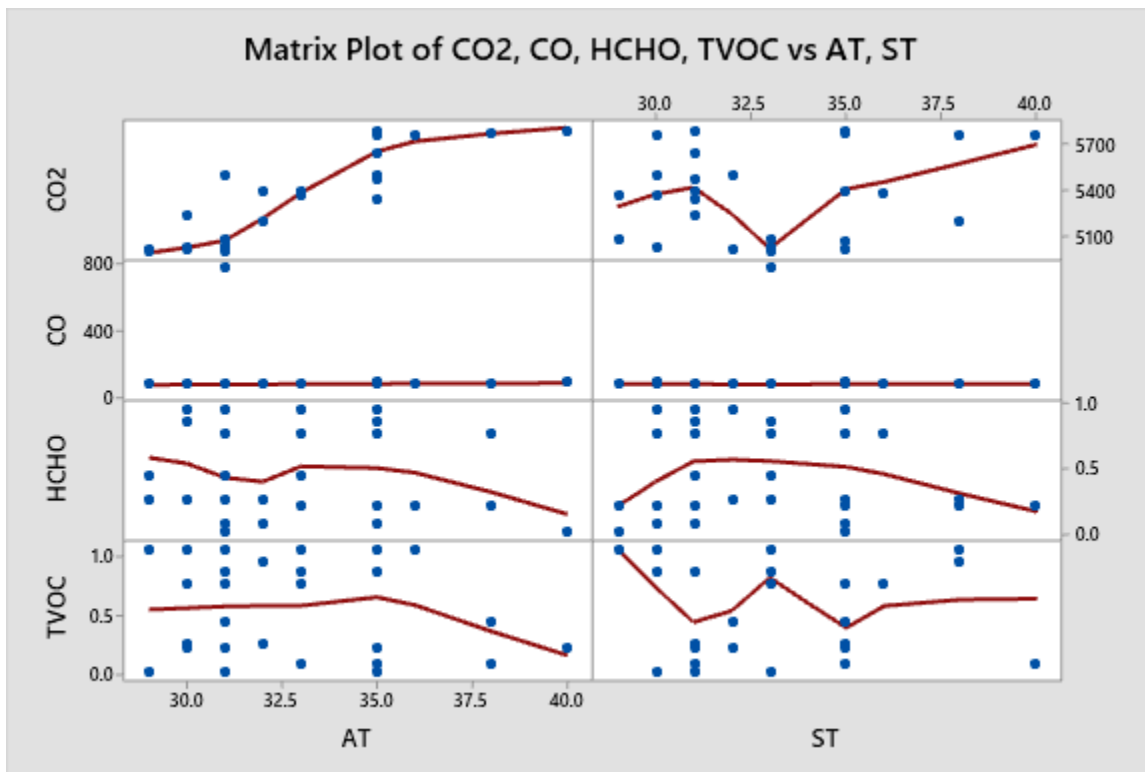
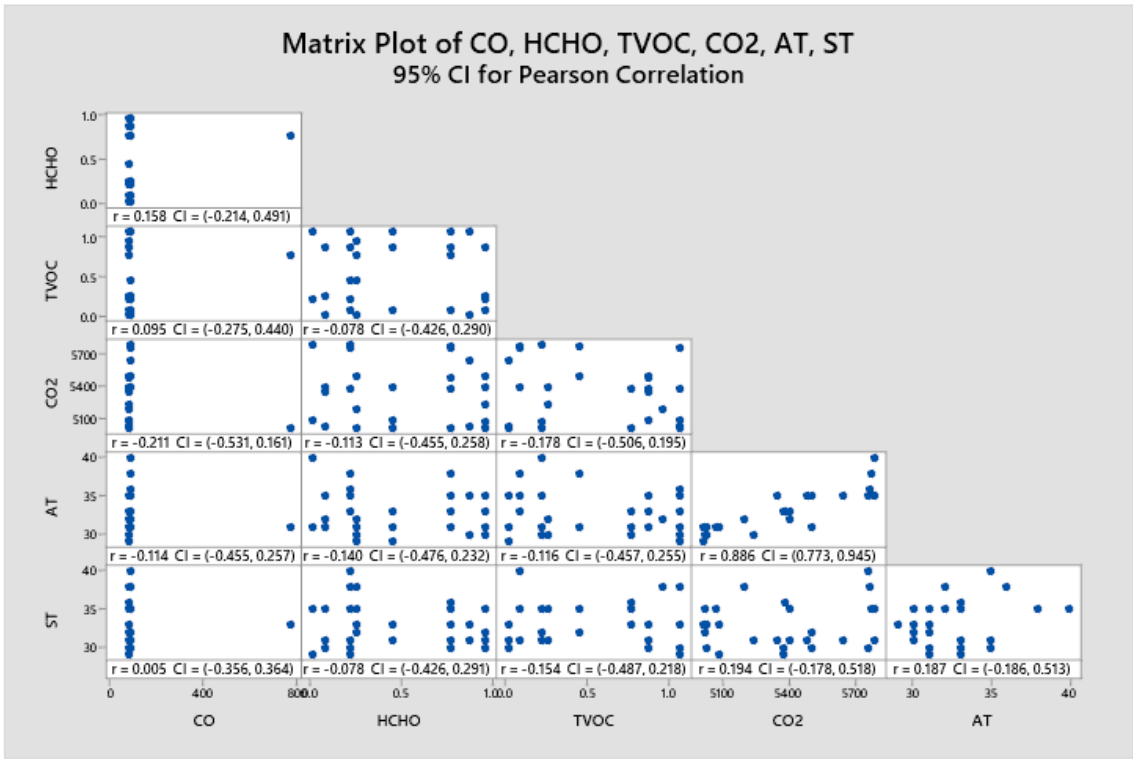


Chart 18.1: Don Antonio Before Quarantine Matrix Plot

Matrix Plot of CO, HCHO, TVOC, CO2, AT, ST
95% CI for Pearson Correlation



Method

Correlation type Pearson
Rows used 30

Correlations

	CO	HCHO	TVOC	CO2	AT
HCHO	0.158				
TVOC	0.095	-0.078			
CO2	-0.211	-0.113	-0.178		
AT	-0.114	-0.140	-0.116	0.886	
ST	0.005	-0.078	-0.154	0.194	0.187

Chart 18.2 and 18.3: *Don Antonio Before Quarantine Matrix Plot (Pearson's)*

After ECQ Data

The following data represents the combination of data collected from after the enhanced community quarantine and around three years after the said recovery period. This was taken once the alert in Metro Manila was lowered. This was on February 27, 2022 – March 4 2022 and February 27, 2025 - March 4, 2025.

Table 6: *Batasan Hills Overpass After Quarantine (bound BHNHS)*

Parameter	Mean (Average)	Mode (Highest)
Surface Temperature	45.95°C	48.3°C
Air Temperature	42.23°C	47°C
Carbon Dioxide (CO2)	43.68ppm	48.9ppm

Humidity	53.3%	61%
HCHO (Formaldehyde)	0.234ppm	0.325ppm
TVOC (Volatile Organic Compounds)	0.578ppm	0.943ppm

Chart 19: Batasan Hills Overpass (bound BHNHS) Day 1

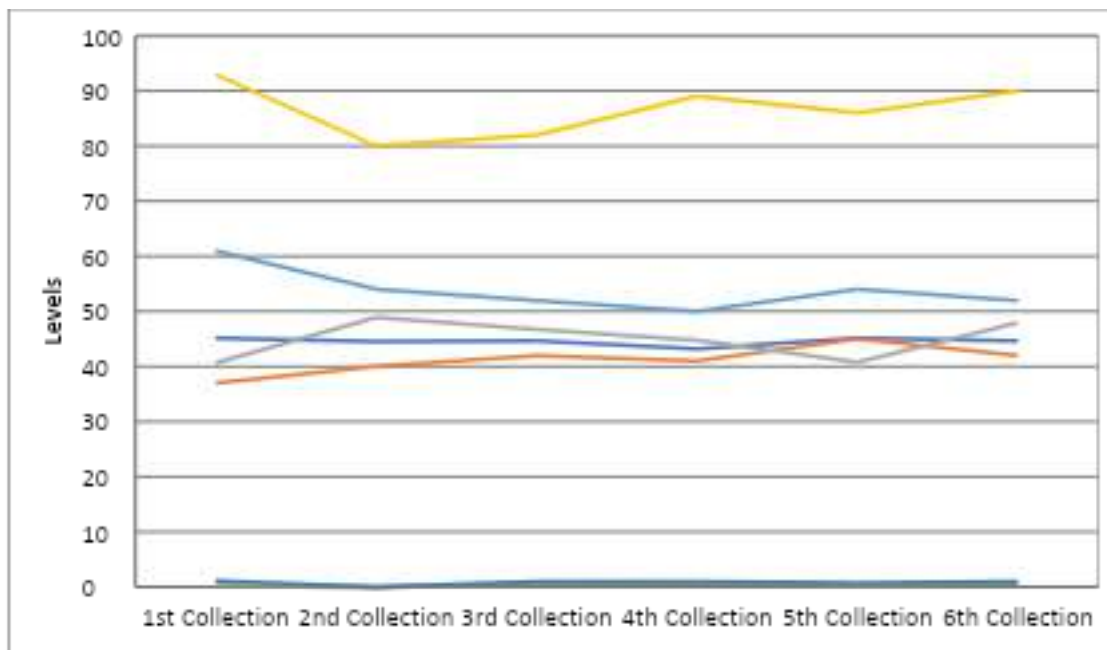


Chart 20: Batasan Hills Overpass (bound BHNHS) Day 2

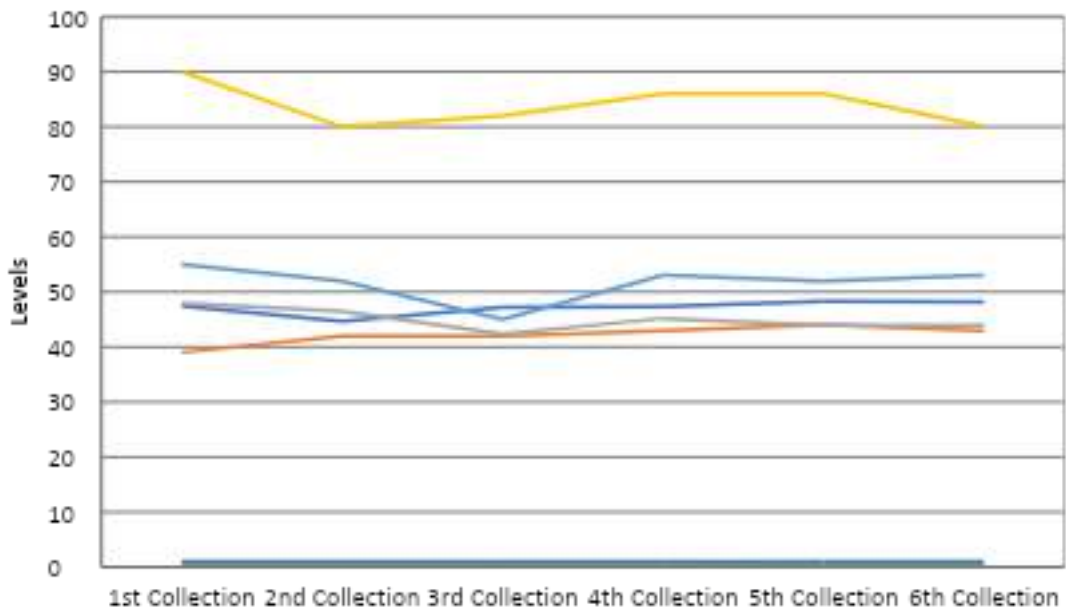


Chart 21: *Batasan Hills Overpass (bound BHNHS) Day 3*

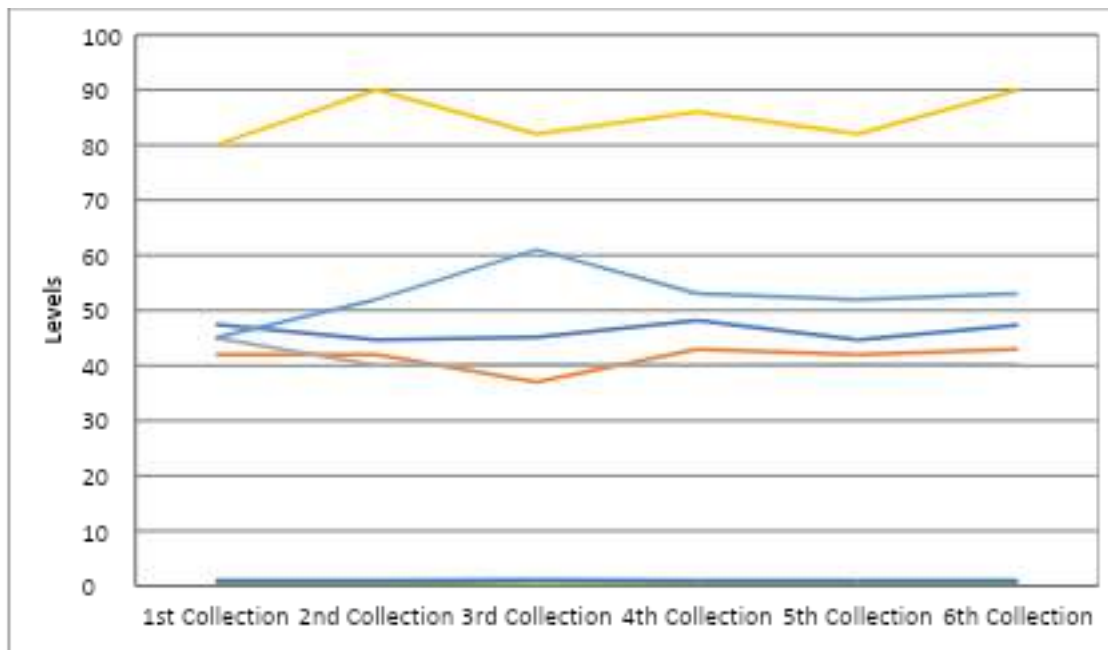


Chart 22: *Batasan Hills Overpass (bound BHNHS) Day 4*

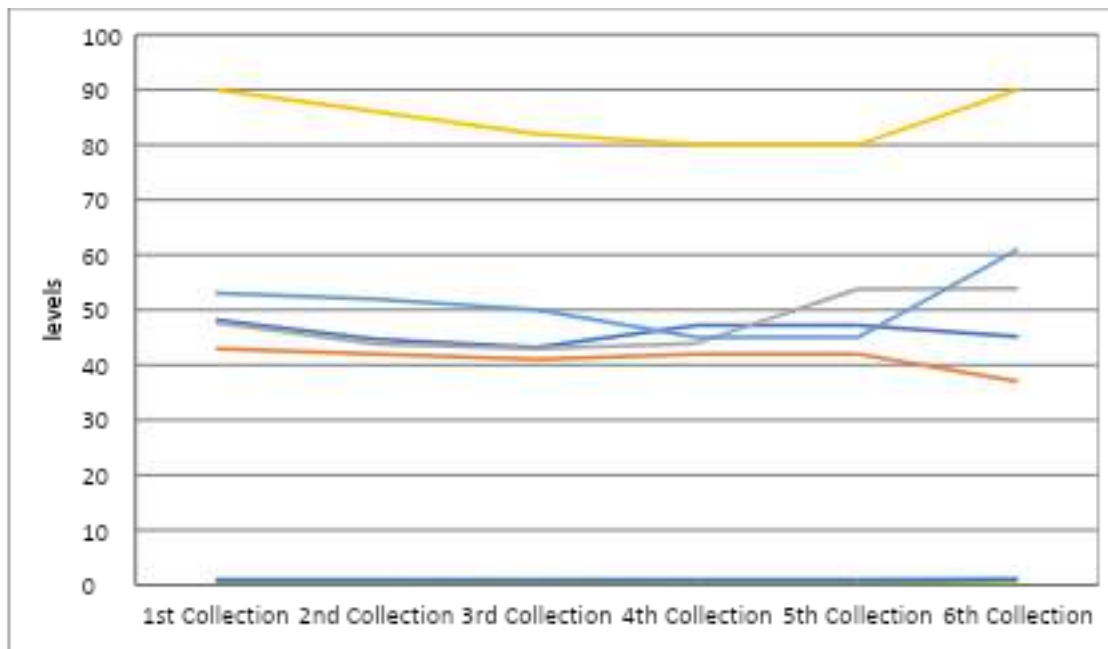
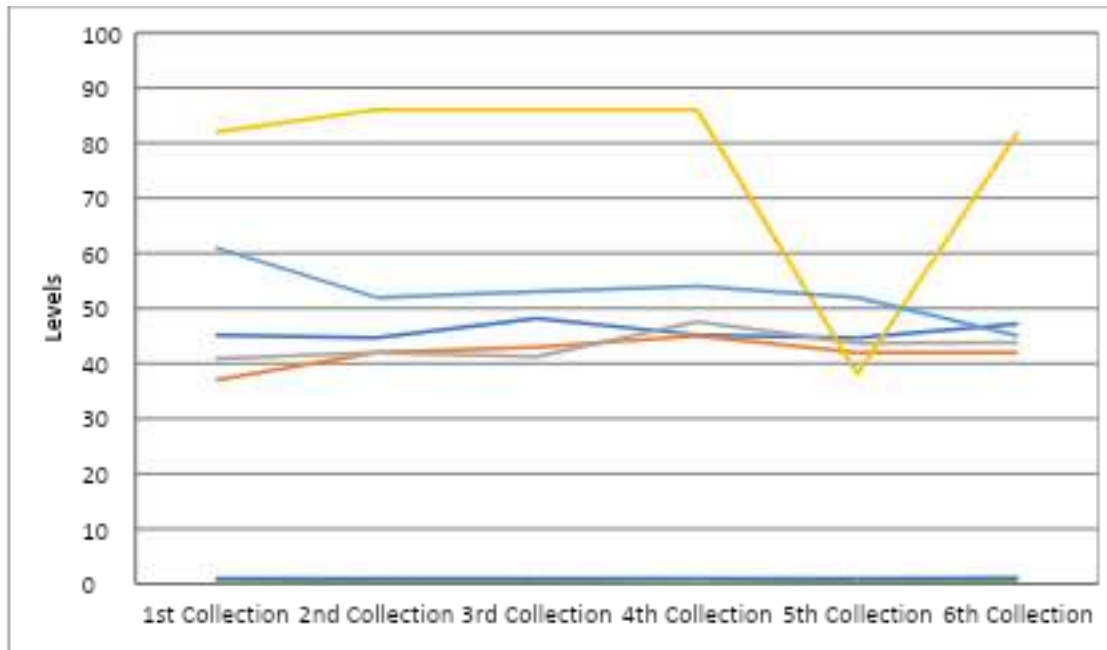


Chart 23: *Batasan Hills Overpass (bound BHNHS) Day 5*



CO2 VS AT AND ST

Regression Equation

$$\text{CO2} = 45.4 + 0.105 \text{ AT} - 0.115 \text{ ST}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	45.4	19.0	2.39	0.023	
AT	0.105	0.288	0.36	0.718	1.10
ST	-0.115	0.409	-0.28	0.780	1.10

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
3.60092	0.50%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	2.134	1.067	0.08	0.921
AT	1	1.715	1.715	0.13	0.718
ST	1	1.027	1.027	0.08	0.780
Error	33	427.897	12.967		
Lack-of-Fit	8	75.294	9.412	0.67	0.715
Pure Error	25	352.603	14.104		
Total	35	430.031			

CO VS AT AND ST

Regression Equation

$$\text{CO} = 48.7 - 0.480 \text{ AT} + 1.193 \text{ ST}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	48.7	46.3	1.05	0.301	
AT	-0.480	0.702	-0.68	0.498	1.10
ST	1.193	0.998	1.20	0.241	1.10

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
8.78355	4.47%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	119.03	59.51	0.77	0.471
AT	1	36.16	36.16	0.47	0.498
ST	1	110.21	110.21	1.43	0.241
Error	33	2545.97	77.15		
Lack-of-Fit	8	212.27	26.53	0.28	0.965
Pure Error	25	2333.70	93.35		
Total	35	2665.00			

HCHO VS AT AND ST

Regression Equation

$$\text{HCHO} = -0.449 + 0.0154 \text{ AT} + 0.0050 \text{ ST}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	-0.449	0.692	-0.65	0.521	
AT	0.0154	0.0105	1.47	0.151	1.10
ST	0.0050	0.0149	0.33	0.741	1.10

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.131284	7.90%	2.32%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.048807	0.024404	1.42	0.257
AT	1	0.037281	0.037281	2.16	0.151
ST	1	0.001914	0.001914	0.11	0.741
Error	33	0.568772	0.017236		
Lack-of-Fit	8	0.350252	0.043781	5.01	0.001
Pure Error	25	0.218520	0.008741		
Total	35	0.617579			

TVOC VS AT AND ST

Regression Equation

$$\text{TVOC} = 0.155 - 0.0033 \text{ AT} + 0.0204 \text{ ST}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	0.155	0.906	0.17	0.865	
AT	-0.0033	0.0137	-0.24	0.810	1.10
ST	0.0204	0.0195	1.05	0.303	1.10

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.171847	3.23%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.03252	0.016260	0.55	0.582
AT	1	0.00174	0.001735	0.06	0.810
ST	1	0.03234	0.032338	1.10	0.303
Error	33	0.97453	0.029531		
Lack-of-Fit	8	0.93035	0.116294	65.81	0.000
Pure Error	25	0.04418	0.001767		
Total	35	1.00705			

TVOC VS AT AND ST

Regression Equation

$$\text{Humidity} = 68.7 - 0.969 \text{ AT} + 0.521 \text{ ST}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	68.7	20.8	3.30	0.002	
AT	-0.969	0.315	-3.08	0.004	1.10
ST	0.521	0.448	1.16	0.253	1.10

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
3.94298	22.41%	17.71%	6.90%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	148.17	74.08	4.77	0.015
AT	1	147.26	147.26	9.47	0.004
ST	1	21.04	21.04	1.35	0.253
Error	33	513.05	15.55		
Lack-of-Fit	8	136.05	17.01	1.13	0.379
Pure Error	25	377.00	15.08		
Total	35	661.22			

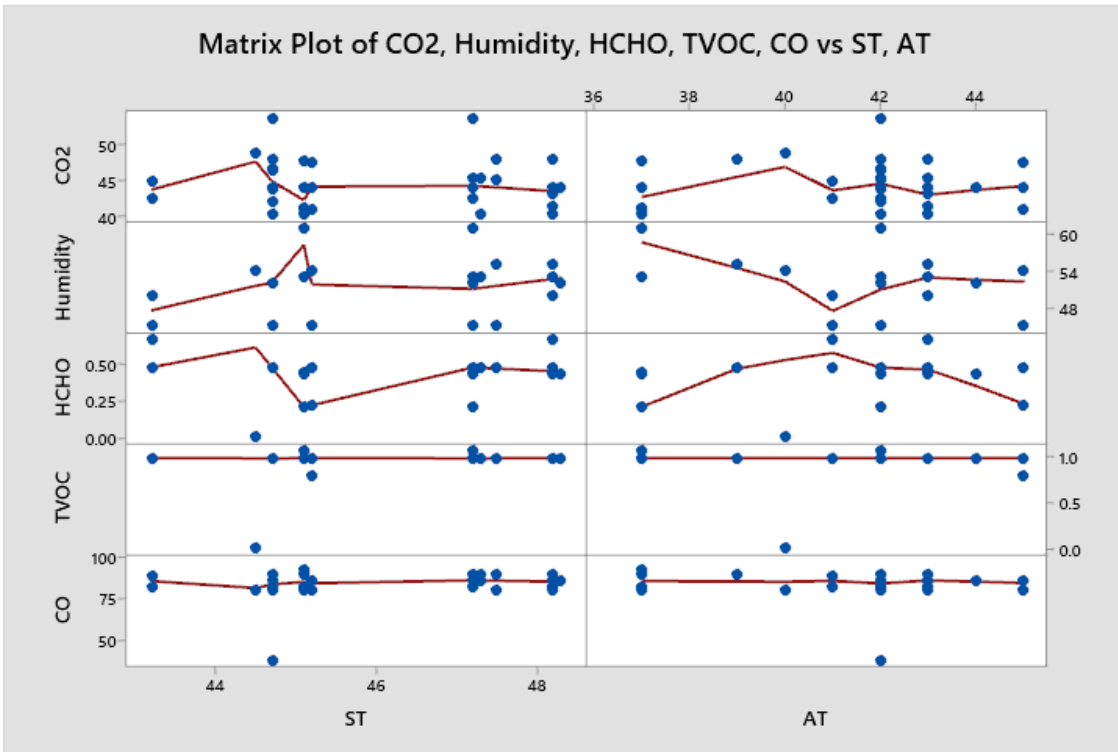
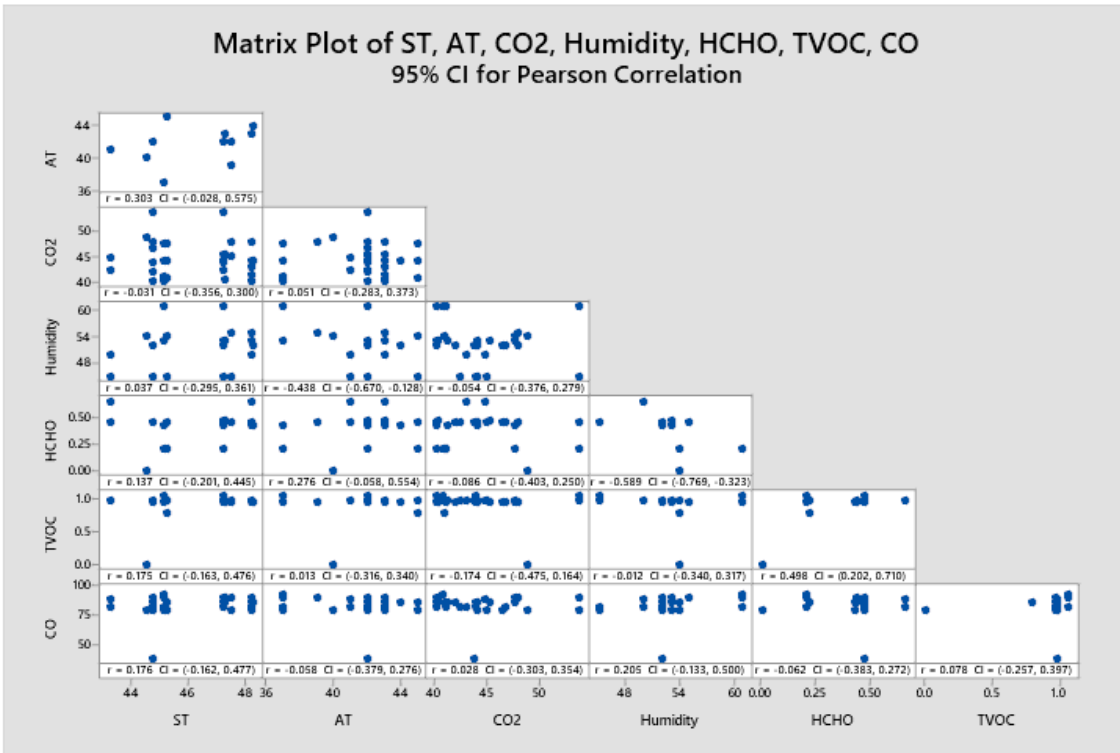


Chart 24.1: *Batasan Hills Overpass After Quarantine Matrix Plot*



Method

Correlation type Pearson
 Rows used 36

Correlations

	ST	AT	CO2	Humidity	HCHO	TVOC
AT	0.303					
CO2	-0.031	0.051				
Humidity	0.037	-0.438	-0.054			
HCHO	0.137	0.276	-0.086	-0.589		
TVOC	0.175	0.013	-0.174	-0.012	0.498	
CO	0.176	-0.058	0.028	0.205	-0.062	0.078

Chart 24.2 and 24.3: *IBP Road After Quarantine Matrix Plot (Pearson's)*

Table 7: San Mateo Northview (by the Marikina/ San Mateo River)

Parameter	Mean (Average)	Mode (Highest)
Surface Temperature	51.62°C	58.3°C
Air Temperature	42.9°C	49°C
Carbon Dioxide (CO2)	43.895ppm	49.8ppm
Humidity	49.8%	61%
HCHO (Formaldehyde)	0.3436ppm	0.3578ppm
TVOC (Volatile Organic Compounds)	0.7345ppm	0.9632ppm

Chart 25: San Mateo Northview (by the Marikina/ San Mateo River) Day 1

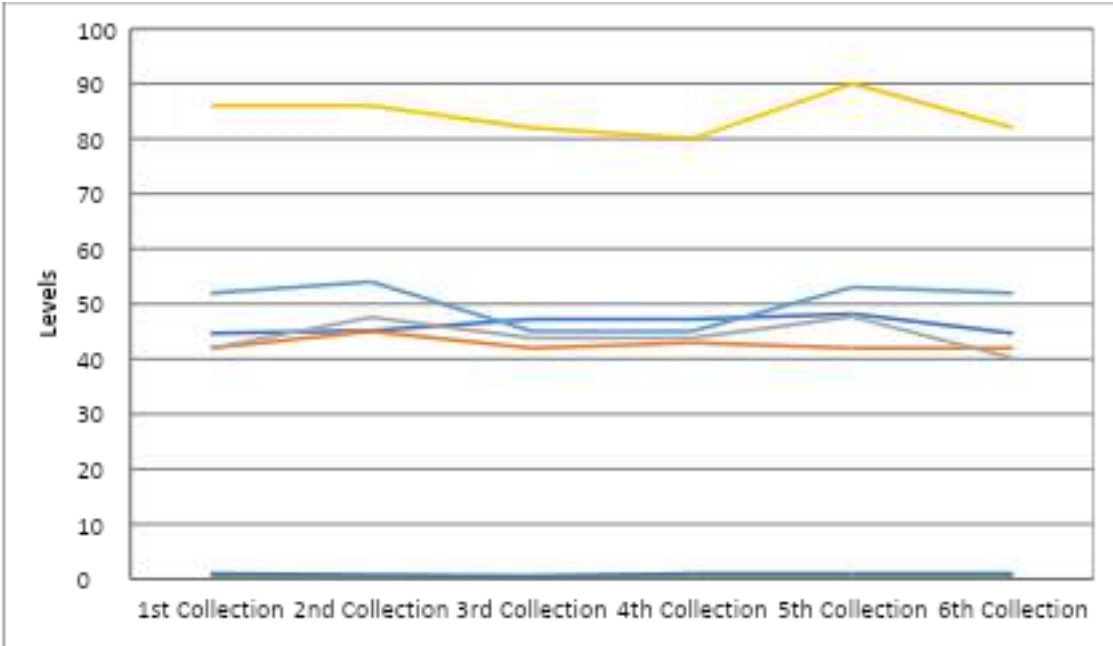


Chart 26: *San Mateo Northview (by the Marikina/ San Mateo River) Day 2*

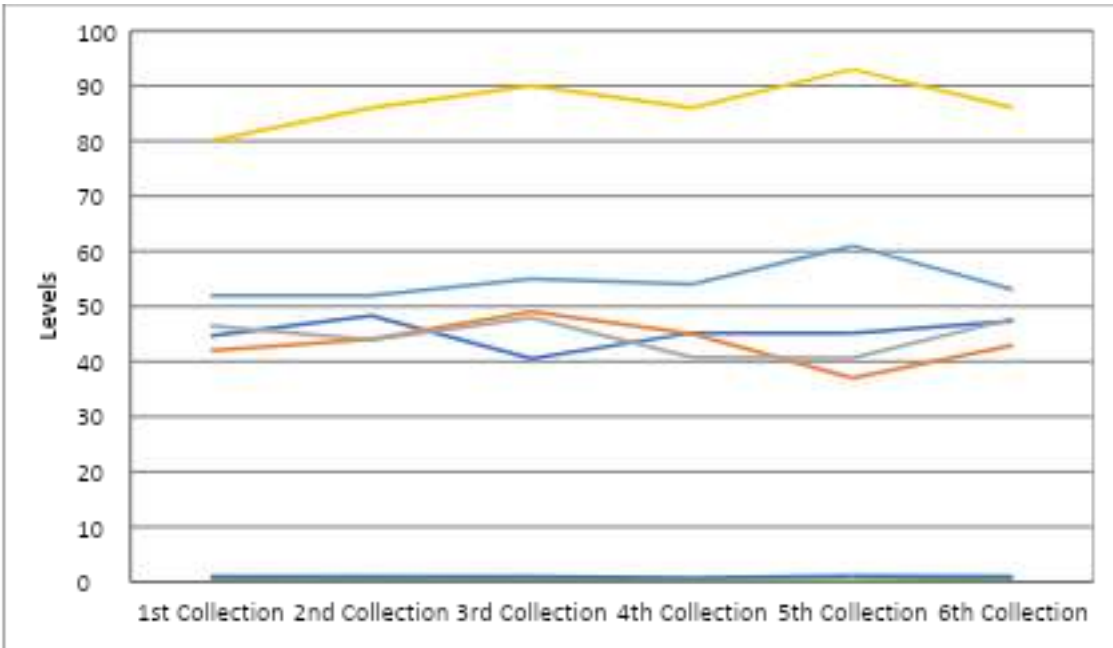


Chart 27: *San Mateo Northview (by the Marikina/ San Mateo River) Day 3*

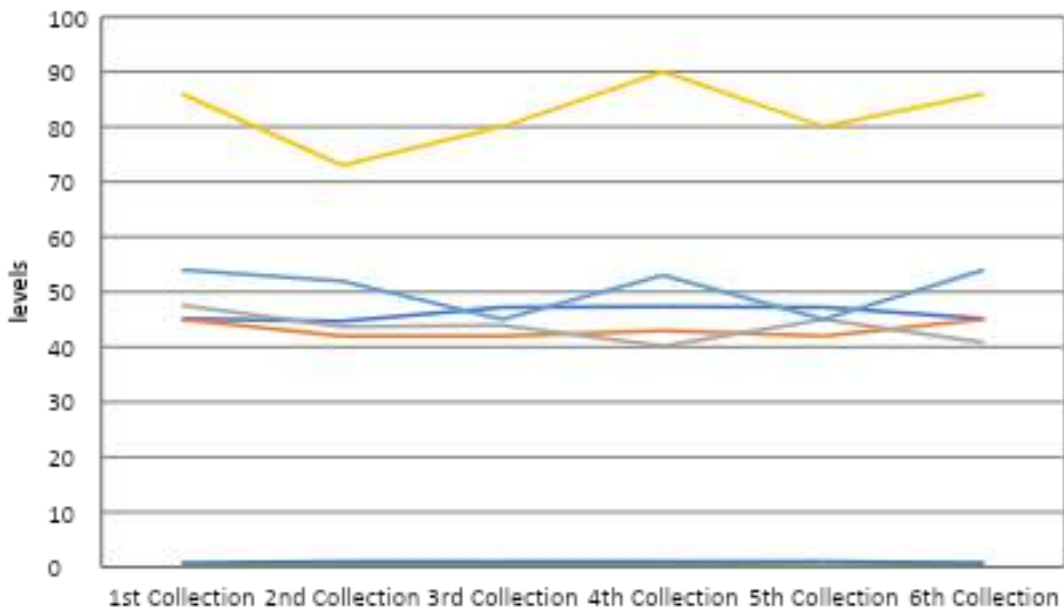


Chart 28: San Mateo Northview (by the Marikina/ San Mateo River) Day 4

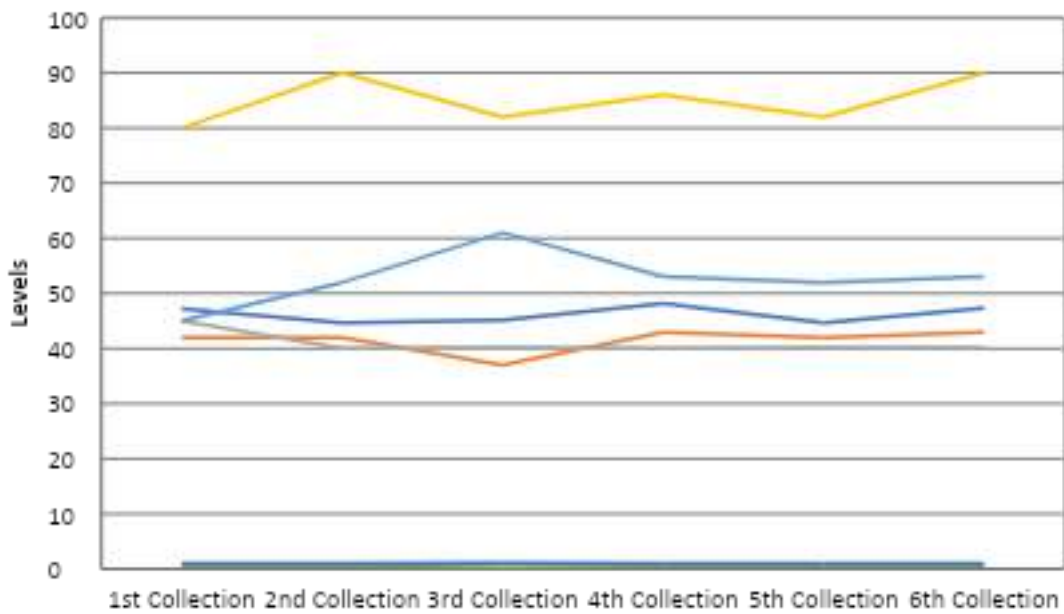
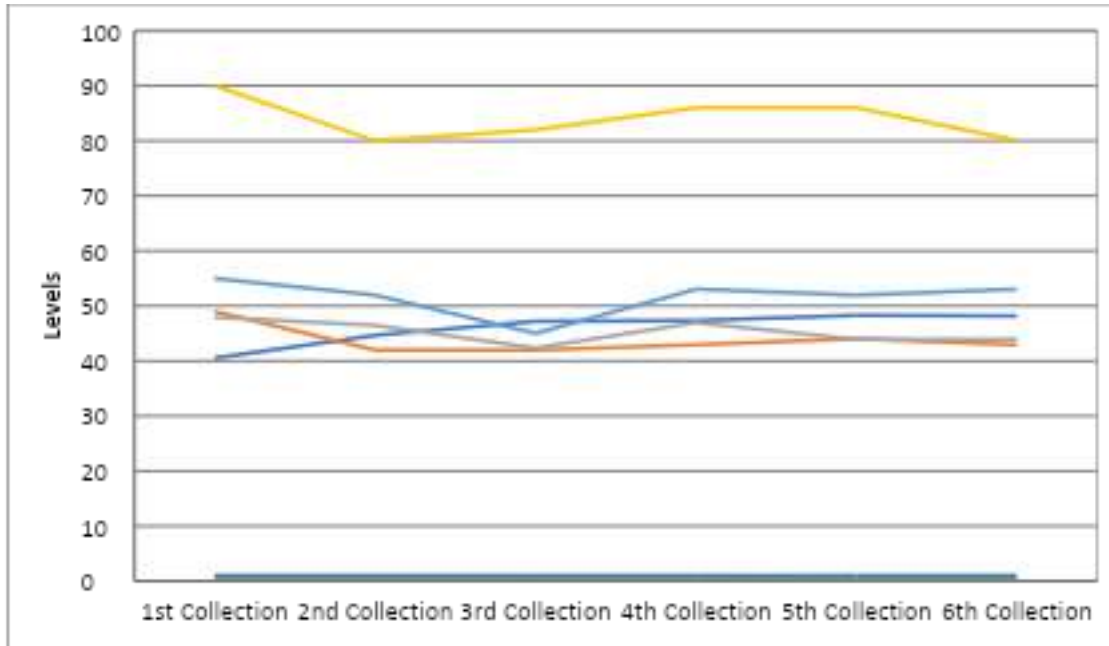


Chart 29: San Mateo Northview (by the Marikina/ San Mateo River) Day 5



CO2 VS AT AND ST

Regression Equation

$$\text{CO2} = 18.2 + 0.032 \text{ ST} + 0.560 \text{ A.T.}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	18.2	18.4	0.99	0.330	
ST	0.032	0.270	0.12	0.907	1.18
A.T.	0.560	0.221	2.54	0.017	1.18

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
2.68506	21.38%	15.56%	9.66%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	52.949	26.4745	3.67	0.039
ST	1	0.100	0.1001	0.01	0.907
A.T.	1	46.343	46.3430	6.43	0.017
Error	27	194.658	7.2095		
Lack-of-Fit	7	33.913	4.8447	0.60	0.747
Pure Error	20	160.745	8.0372		
Total	29	247.607			

CO VS AT AND ST

Regression Equation

$$\text{CO} = 77.2 - 0.192 \text{ ST} + 0.377 \text{ A.T.}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	77.2	30.7	2.52	0.018	
ST	-0.192	0.452	-0.42	0.675	1.18
A.T.	0.377	0.369	1.02	0.317	1.18

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
4.48971	6.40%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	37.215	18.608	0.92	0.409
ST	1	3.629	3.629	0.18	0.675
A.T.	1	20.958	20.958	1.04	0.317
Error	27	544.252	20.157		
Lack-of-Fit	7	276.095	39.442	2.94	0.027
Pure Error	20	268.157	13.408		
Total	29	581.467			

HCHO VS AT AND ST

Regression Equation

$$\text{HCHO} = -0.561 + 0.00841 \text{ ST} + 0.01386 \text{ A.T.}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	-0.561	0.653	-0.86	0.398	
ST	0.00841	0.00961	0.88	0.389	1.18
A.T.	0.01386	0.00786	1.76	0.089	1.18

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.0954903	10.45%	3.82%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.028734	0.014367	1.58	0.225
ST	1	0.006986	0.006986	0.77	0.389
A.T.	1	0.028381	0.028381	3.11	0.089
Error	27	0.246197	0.009118		
Lack-of-Fit	7	0.149787	0.021398	4.44	0.004
Pure Error	20	0.096410	0.004821		
Total	29	0.274931			

TVOC VS AT AND ST

Regression Equation

TVOC = 2.050 - 0.0095 ST - 0.01569 A.T.

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	2.050	0.757	2.71	0.012	
ST	-0.0095	0.0111	-0.85	0.403	1.18
A.T.	-0.01569	0.00912	-1.72	0.097	1.18

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.110834	9.98%	3.31%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.036776	0.018388	1.50	0.242
ST	1	0.008852	0.008852	0.72	0.403
A.T.	1	0.036347	0.036347	2.96	0.097
Error	27	0.331673	0.012284		
Lack-of-Fit	7	0.126844	0.018121	1.77	0.150
Pure Error	20	0.204829	0.010241		
Total	29	0.368449			

Humidity VS AT AND ST

Regression Equation

Humidity = 118.8 - 1.027 ST - 0.458 A.T.

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	118.8	26.0	4.58	0.000	
ST	-1.027	0.382	-2.69	0.012	1.18
A.T.	-0.458	0.313	-1.47	0.154	1.18

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
3.79836	21.57%	15.76%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	107.157	53.579	3.71	0.038
ST	1	104.311	104.311	7.23	0.012
A.T.	1	31.030	31.030	2.15	0.154
Error	27	389.543	14.428		
Lack-of-Fit	7	389.543	55.649	*	*
Pure Error	20	0.000	0.000		
Total	29	496.700			

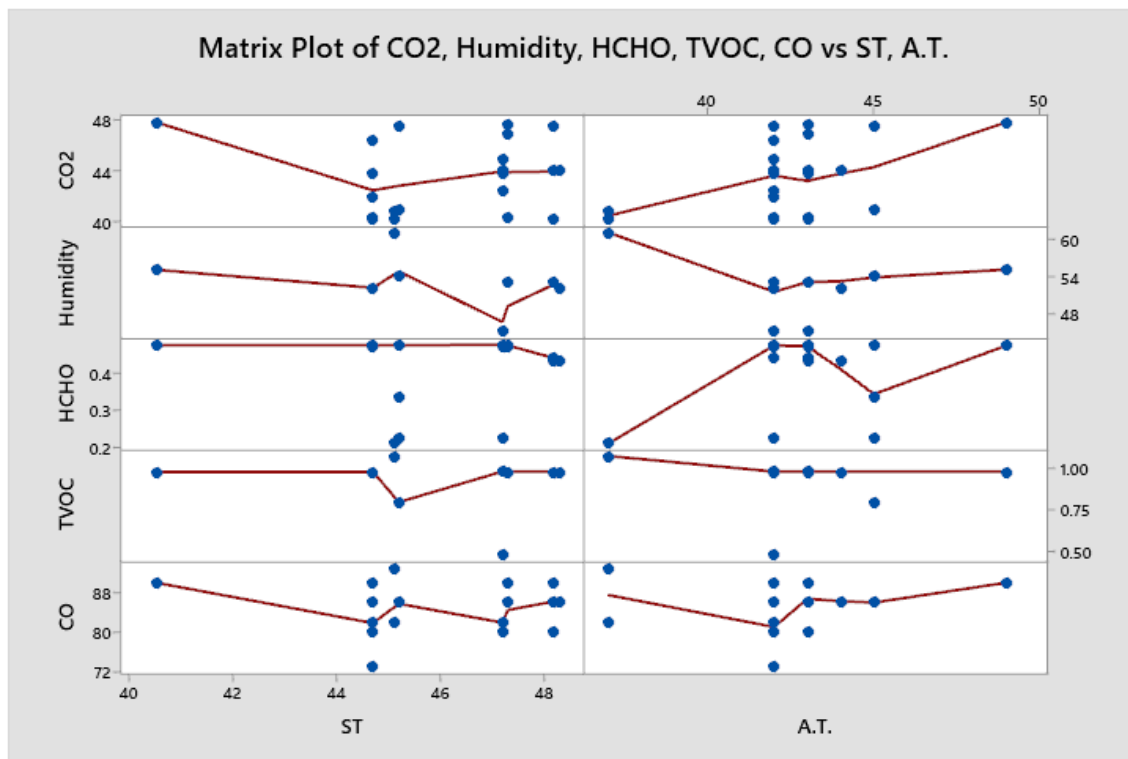
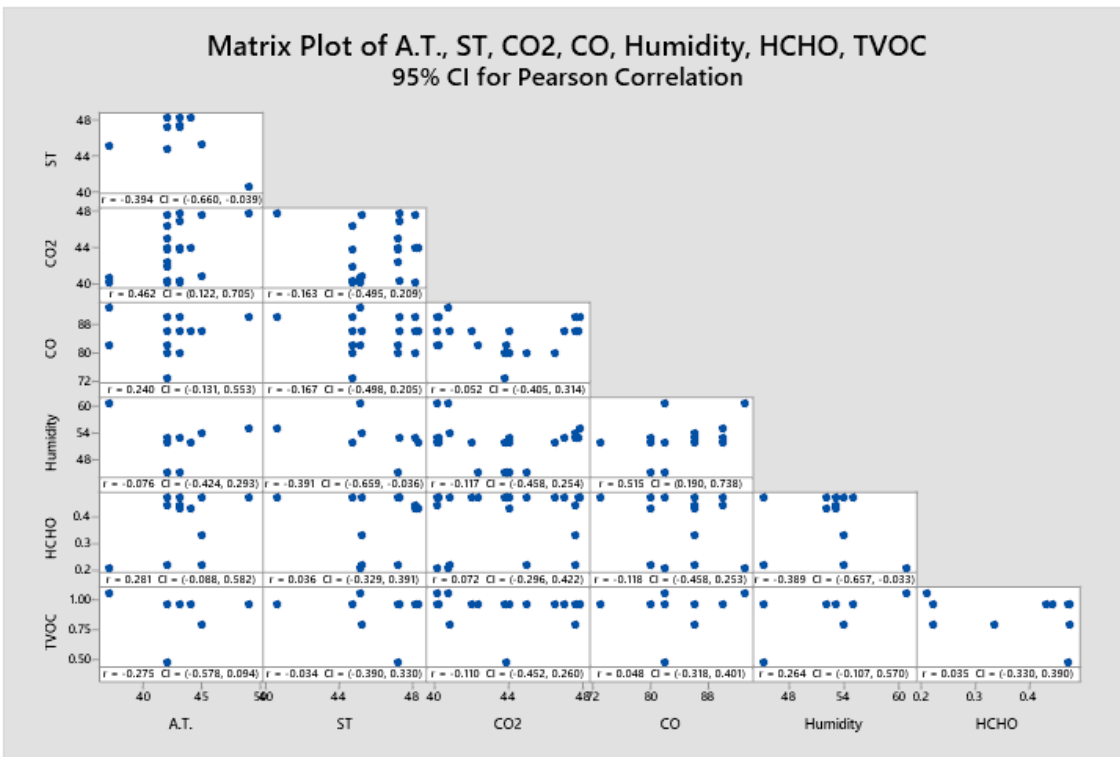


Chart 30.1: San Mateo After Quarantine Matrix Plot



Method

Correlation type Pearson
 Rows used 30

Correlations

	A.T.	ST	CO2	CO	Humidity	HCHO
ST	-0.394					
CO2	0.462	-0.163				
CO	0.240	-0.167	-0.052			
Humidity	-0.076	-0.391	-0.117	0.515		
HCHO	0.281	0.036	0.072	-0.118	-0.389	
TVOC	-0.275	-0.034	-0.110	0.048	0.264	0.035

Chart 30.2 and 30.3: *San Mateo After Quarantine Matrix Plot (Pearson's)*

Table 8: *Don Antonio Overpass (bound Shopwise and Petron)*

Parameter	Mean (Average)	Mode (Highest)
Surface Temperature	51.88°C	58.3°C
Air Temperature	42.43°C	45 °C
Carbon Dioxide (CO2)	43.340ppm	48.3240ppm
Humidity	48.8%	49%
HCHO (Formaldehyde)	0.3358ppm	0.2478ppm
TVOC (Volatile Organic Compounds)	0.6234ppm	0.7526ppm

Chart 31: *Don Antonio Overpass (bound Shopwise and Petron) Day 1*

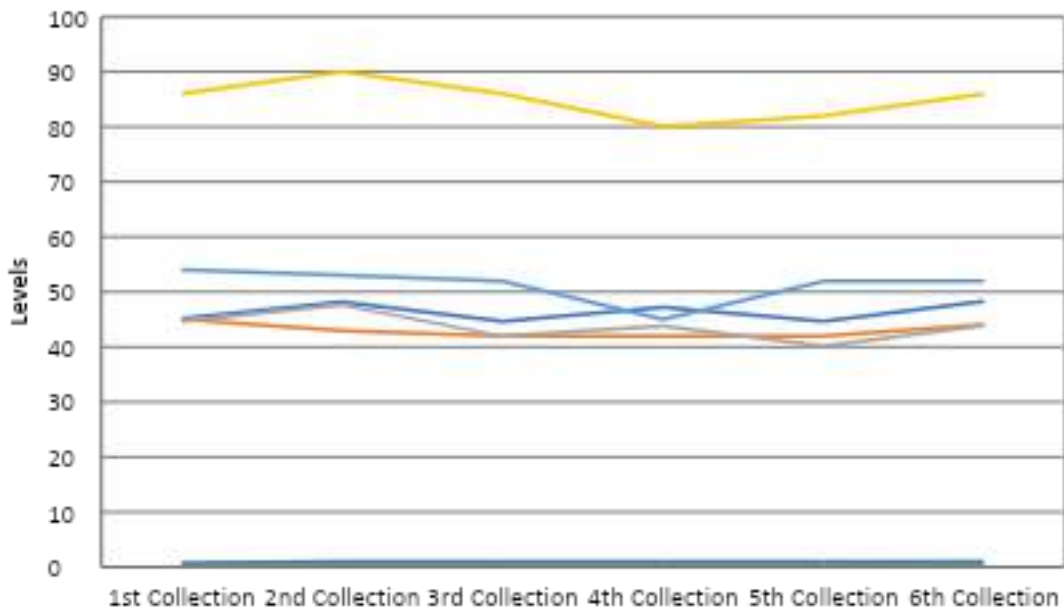


Chart 32: *Don Antonio Overpass (bound Shopwise and Petron) Day 2*

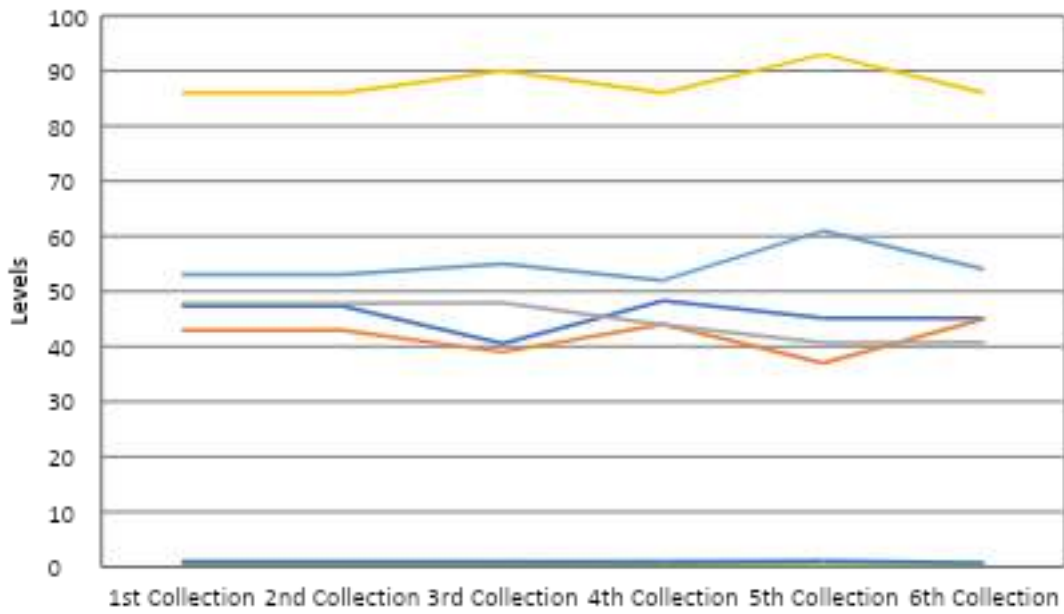


Chart 33: *Don Antonio Overpass (bound Shopwise and Petron) Day 3*

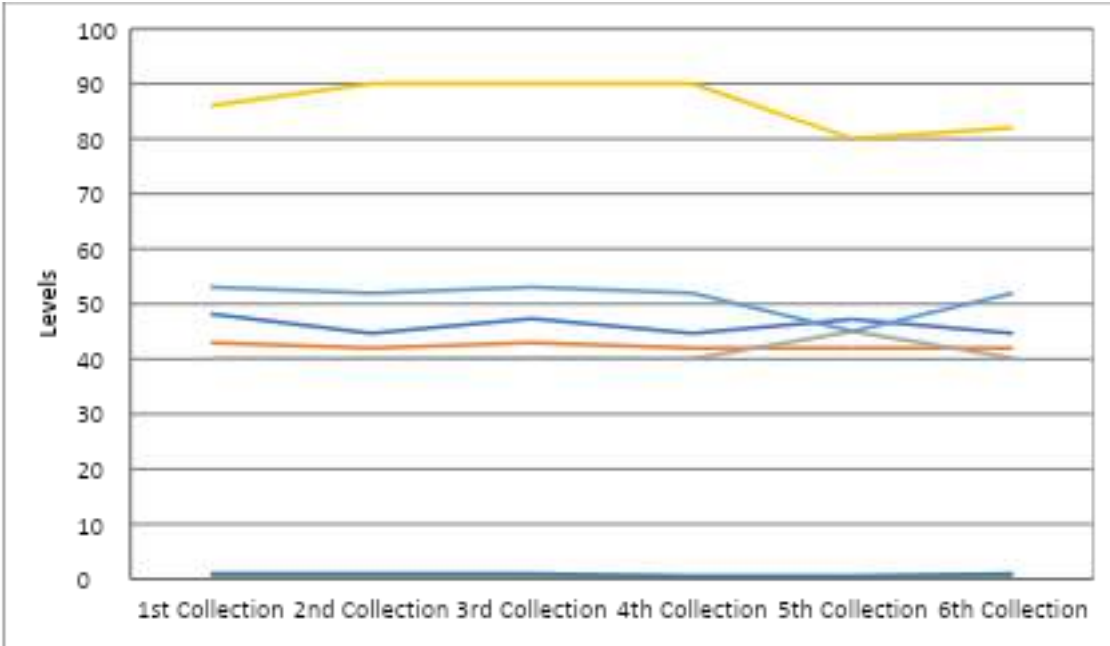


Chart 34: *Don Antonio Overpass (bound Shopwise and Petron) Day 4*

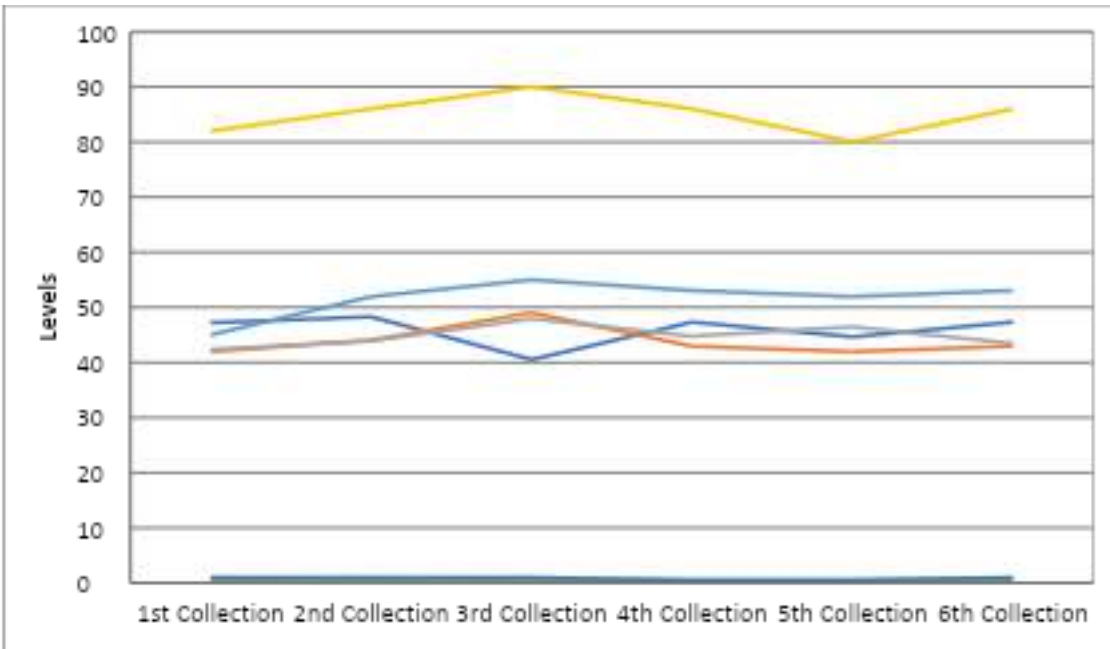
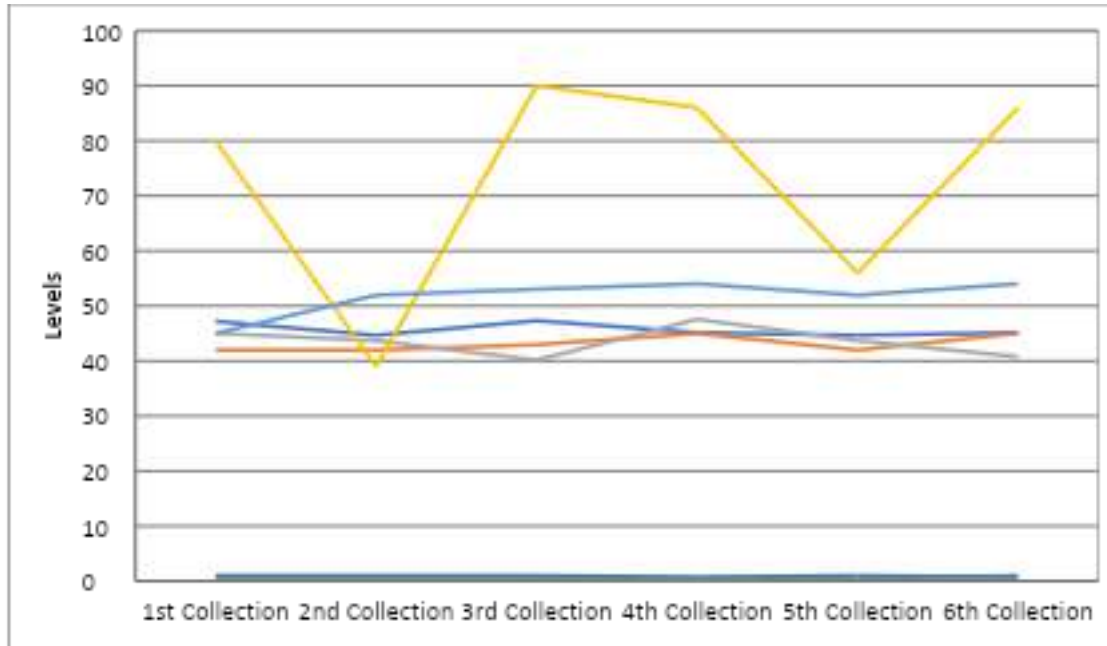


Chart 35: *Don Antonio Overpass (bound Shopwise and Petron) Day 5*



CO2 VS AT and ST

Regression Equation

$$\text{CO2} = 37.5 - 0.160 \text{ ST} + 0.312 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	37.5	16.4	2.29	0.030	
ST	-0.160	0.261	-0.61	0.544	1.00
AT	0.312	0.261	1.19	0.243	1.00

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
2.83914	6.24%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	14.489	7.245	0.90	0.419
ST	1	3.037	3.037	0.38	0.544
AT	1	11.475	11.475	1.42	0.243
Error	27	217.639	8.061		
Lack-of-Fit	6	53.088	8.848	1.13	0.380
Pure Error	21	164.551	7.836		
Total	29	232.129			

CO VS AT and ST

Regression Equation

$$\text{CO} = 44.6 + 0.39 \text{ ST} + 0.49 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	44.6	62.9	0.71	0.485	
ST	0.39	1.00	0.39	0.701	1.00
AT	0.49	1.00	0.49	0.629	1.00

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
10.9084	1.42%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	46.38	23.19	0.19	0.824
ST	1	17.92	17.92	0.15	0.701
AT	1	28.38	28.38	0.24	0.629
Error	27	3212.82	118.99		
Lack-of-Fit	6	832.61	138.77	1.22	0.333
Pure Error	21	2380.21	113.34		
Total	29	3259.20			

HCHO VS AT and ST

Regression Equation

$$\text{HCHO} = 0.143 + 0.0001 \text{ ST} + 0.0071 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	0.143	0.786	0.18	0.857	
ST	0.0001	0.0125	0.01	0.996	1.00
AT	0.0071	0.0125	0.57	0.575	1.00

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.136240	1.18%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.005984	0.002992	0.16	0.852
ST	1	0.000001	0.000001	0.00	0.996
AT	1	0.005983	0.005983	0.32	0.575
Error	27	0.501154	0.018561		
Lack-of-Fit	6	0.072502	0.012084	0.59	0.733
Pure Error	21	0.428652	0.020412		
Total	29	0.507137			

TVOC VS AT and ST

Regression Equation

$$\text{TVOC} = 1.08 + 0.0035 \text{ ST} - 0.0082 \text{ AT}$$

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	1.08	1.06	1.02	0.318	
ST	0.0035	0.0169	0.21	0.836	1.00
AT	-0.0082	0.0169	-0.48	0.632	1.00

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.183628	1.02%	0.00%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	0.009371	0.004685	0.14	0.871
ST	1	0.001465	0.001465	0.04	0.836
AT	1	0.007919	0.007919	0.23	0.632
Error	27	0.910418	0.033719		
Lack-of-Fit	6	0.130464	0.021744	0.59	0.738
Pure Error	21	0.779954	0.037141		
Total	29	0.919789			

Humidity VS AT AND ST

Regression Equation

Humidity = 79.9 - 0.579 ST - 0.028 AT

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	79.9	18.5	4.31	0.000	
ST	-0.579	0.295	-1.96	0.060	1.00
AT	-0.028	0.296	-0.10	0.924	1.00

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
3.21348	12.52%	6.03%	0.00%

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	39.885	19.9427	1.93	0.164
ST	1	39.782	39.7816	3.85	0.060
AT	1	0.096	0.0958	0.01	0.924
Error	27	278.815	10.3265		
Lack-of-Fit	6	278.815	46.4691	*	*
Pure Error	21	0.000	0.0000		
Total	29	318.700			

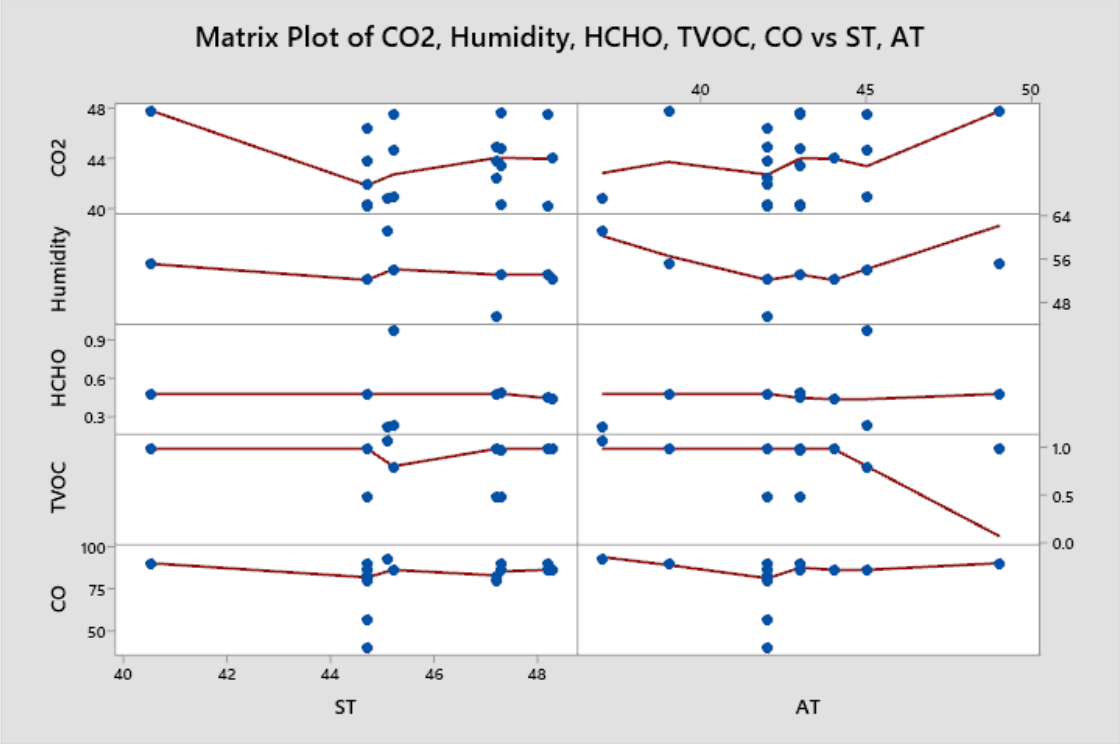
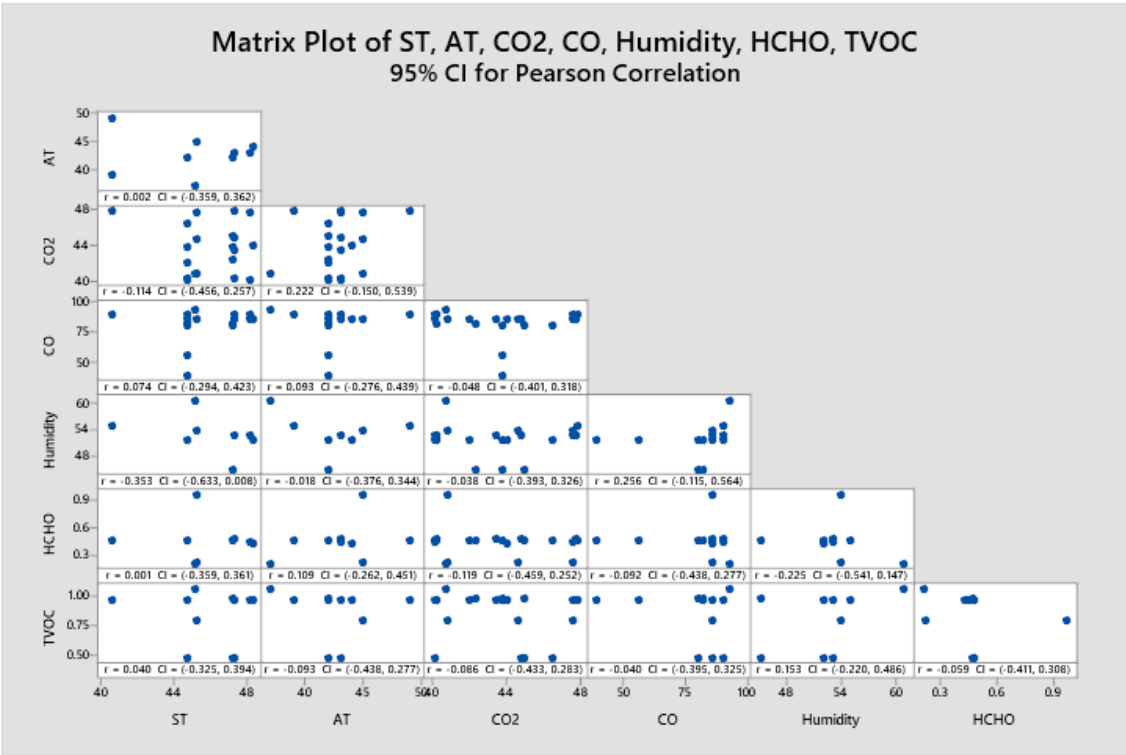


Chart 36.1: Don Antonio After Quarantine Matrix Plot



Method

Correlation type Pearson
 Rows used 30

Correlations

	ST	AT	CO2	CO	Humidity	HCHO
AT	0.002					
CO2	-0.114	0.222				
CO	0.074	0.093	-0.048			
Humidity	-0.353	-0.018	-0.038	0.256		
HCHO	0.001	0.109	-0.119	-0.092	-0.225	
TVOC	0.040	-0.093	-0.086	-0.040	0.153	-0.059

Chart 36.2 and 36.3: *Don Antonio After Quarantine Matrix Plot (Pearson's)*

Before and After Quarantine Comparison

These set of data are the comparisons received from entering the data into a T-Test formula through a program called Minitab 19.

Batasan Hills Overpass

ST before VS ST After

Surface Temperature from before quarantine is denoted as ST_1; Meanwhile, Surface Temperature from after quarantine is denoted as ST.

Descriptive Statistics

<u>Sample</u>	<u>N</u>	<u>Mean</u>	<u>StDev</u>	<u>SE Mean</u>
ST	30	45.96	1.60	0.29
ST_1	30	40.05	6.41	1.17

Estimation for Paired Difference

95% CI for			
<u>Mean</u>	<u>StDev</u>	<u>SE Mean</u>	<u>$\mu_{\text{difference}}$</u>
5.910	4.917	0.898	(4.074, 7.746)

$\mu_{\text{difference}}$: mean of (ST - ST_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$
Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

<u>T-Value</u>	<u>P-Value</u>
6.58	0.000

AT before VS AT After

Air Temperature from before quarantine is denoted as AT_1; Meanwhile, Air Temperature from after quarantine is denoted as AT.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
AT	30	41.567	2.176	0.397
AT_1	30	31.733	1.999	0.365

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
9.833	1.147	0.209	(9.405, 10.262)

$\mu_{\text{difference}}$: mean of (AT - AT_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
46.95	0.000

CO2 before VS CO2 After

Carbon Dioxide (CO2) from before quarantine is denoted as CO2_1; Meanwhile, Carbon Dioxide (CO2) from after quarantine is denoted as CO2.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
CO2	30	44.705	3.658	0.668
CO2_1	30	53.761	2.802	0.511

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-9.056	2.962	0.541	(-10.162, -7.950)

$\mu_{\text{difference}}$: mean of (CO2 - CO2_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$
Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-16.75	0.000

CO before VS CO After

Carbon Monoxide (CO) from before quarantine is denoted as CO_1; Meanwhile, Carbon Monoxide (CO) from after quarantine is denoted as CO.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
CO	30	84.867	4.216	0.770
CO_1	30	85.133	3.910	0.714

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-0.267	4.863	0.888	(-2.083, 1.549)

$\mu_{\text{difference}}$: mean of (CO - CO_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-0.30	0.766

HCHO before VS HCHO After

Formaldehyde (HCHO) from before quarantine is denoted as HCHO_1; Meanwhile, Formaldehyde (HCHO) from after quarantine is denoted as HCHO.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
HCHO	30	0.4288	0.1341	0.0245
HCHO_1	30	0.4134	0.1444	0.0264

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
0.0154	0.1057	0.0193	(-0.0241, 0.0549)

$\mu_{\text{difference}}$: mean of (HCHO - HCHO_1)

Test

Null hypothesis	$H_0: \mu_{\text{difference}} = 0$
Alternative hypothesis	$H_1: \mu_{\text{difference}} \neq 0$
T-Value	P-Value
0.80	0.432

TVOC before VS TVOC After

Total Volatile Organic Compounds (TVOC) from before quarantine is denoted as TVOC_1; Meanwhile, Total Volatile Organic Compounds (TVOC) from after quarantine is denoted as TVOC.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
TVOC	30	0.9479	0.1847	0.0337
TVOC_1	30	0.9449	0.1881	0.0343

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
0.00297	0.05215	0.00952	(-0.01651, 0.02244)

$\mu_{\text{difference}}$: mean of (TVOC - TVOC_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
0.31	0.758

San Mateo Road

ST before VS ST After

Surface Temperature from before quarantine is denoted as ST_1; Meanwhile, Surface Temperature from after quarantine is denoted as ST.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
ST	30	45.947	2.024	0.369
ST_1	30	33.000	2.816	0.514

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
12.947	3.365	0.614	(11.690, 14.203)

$\mu_{\text{difference}}$: mean of (ST - ST_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
21.07	0.000

AT before VS AT After

Air Temperature from before quarantine is denoted as AT_1; Meanwhile, Air Temperature from after quarantine is denoted as AT.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
AT	30	42.833	2.019	0.369
AT_1	30	33.000	2.816	0.514

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
9.833	3.374	0.616	(8.573, 11.093)

$\mu_{\text{difference}}$: mean of (AT - AT_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
15.96	0.000

CO2 before VS CO2 After

Carbon Dioxide (CO2) from before quarantine is denoted as CO2_1; Meanwhile, Carbon Dioxide (CO2) from after quarantine is denoted as CO2.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
CO2	30	43.5	2.8	0.5
CO2_1	30	5364.7	292.0	53.3

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-5321.2	289.2	52.8	(-5429.2, -5213.2)

$\mu_{\text{difference}}$: mean of (CO2 - CO2_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-100.77	0.000

CO before VS CO After

Carbon Monoxide (CO) from before quarantine is denoted as CO_1; Meanwhile, Carbon Monoxide (CO) from after quarantine is denoted as CO.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
CO	30	83.4	10.6	1.9
CO_1	30	106.7	126.8	23.2

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-23.3	126.0	23.0	(-70.3, 23.8)

$\mu_{\text{difference}}$: mean of (CO - CO_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-1.01	0.320

HCHO before VS HCHO After

Formaldehyde (HCHO) from before quarantine is denoted as HCHO_1; Meanwhile, Formaldehyde (HCHO) from after quarantine is denoted as HCHO.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
HCHO	30	0.4508	0.1322	0.0241
HCHO_1	30	0.4776	0.3427	0.0626

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-0.0269	0.3991	0.0729	(-0.1759, 0.1222)

$\mu_{\text{difference}}$: mean of (HCHO - HCHO_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-0.37	0.715

TVOC before VS TVOC After

Total Volatile Organic Compounds (TVOC) from before quarantine is denoted as TVOC_1; Meanwhile, Total Volatile Organic Compounds (TVOC) from after quarantine is denoted as TVOC.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
TVOC	30	0.8881	0.1781	0.0325
TVOC_1	30	0.5559	0.4087	0.0746

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
0.3322	0.4553	0.0831	(0.1622, 0.5022)

$\mu_{\text{difference}}$: mean of (TVOC - TVOC_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
4.00	0.000

Don Antonio Overpass

ST before VS ST After

Surface Temperature from before quarantine is denoted as ST_1; Meanwhile, Surface Temperature from after quarantine is denoted as ST.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
ST	30	45.95	2.01	0.37
ST_1	30	38.20	6.36	1.16

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
7.75	5.89	1.07	(5.55, 9.94)

$\mu_{\text{difference}}$: mean of (ST - ST_1)

Test

Null hypothesis	$H_0: \mu_{\text{difference}} = 0$
Alternative hypothesis	$H_1: \mu_{\text{difference}} \neq 0$
T-Value	P-Value
7.21	0.000

AT before VS AT After

Air Temperature from before quarantine is denoted as AT_1; Meanwhile, Air Temperature from after quarantine is denoted as AT.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
AT	30	42.900	2.454	0.448
AT_1	30	32.233	2.012	0.367

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
10.667	2.758	0.504	(9.637, 11.697)

$\mu_{\text{difference}}$: mean of (AT - AT_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
21.18	0.000

CO2 before VS CO2 After

Carbon Dioxide (CO2) from before quarantine is denoted as CO2_1; Meanwhile, Carbon Dioxide (CO2) from after quarantine is denoted as CO2.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
CO2	30	44.705	3.658	0.668
CO2_1	30	53.761	2.802	0.511

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-9.056	2.962	0.541	(-10.162, -7.950)

$\mu_{\text{difference}}$: mean of (CO2 - CO2_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-16.75	0.000

CO before VS CO After

Carbon Monoxide (CO) from before quarantine is denoted as CO₁; Meanwhile, Carbon Monoxide (CO) from after quarantine is denoted as CO.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
CO	30	84.867	4.216	0.770
CO_1	30	85.133	3.910	0.714

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
-0.267	4.863	0.888	(-2.083, 1.549)

$\mu_{\text{difference}}$: mean of (CO - CO_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-0.30	0.766

HCHO before VS HCHO After

Formaldehyde (HCHO) from before quarantine is denoted as HCHO_1; Meanwhile, Formaldehyde (HCHO) from after quarantine is denoted as HCHO.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
HCHO	30	0.4204	0.0974	0.0178
HCHO_1	30	0.4168	0.1030	0.0188

Estimation for Paired Difference

Mean	StDev	SE Mean	95% CI for $\mu_{\text{difference}}$
0.0036	0.0703	0.0128	(-0.0226, 0.0299)

$\mu_{\text{difference}}$: mean of (HCHO - HCHO_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
0.28	0.779

TVOC before VS TVOC After

Total Volatile Organic Compounds (TVOC) from before quarantine is denoted as TVOC_1; Meanwhile, Total Volatile Organic Compounds (TVOC) from after quarantine is denoted as TVOC.

Descriptive Statistics

Sample	N	Mean	StDev	SE Mean
TVOC	30	0.9418	0.1127	0.0206
TVOC_1	30	0.9591	0.0700	0.0128

Estimation for Paired Difference

95% CI for			
Mean	StDev	SE Mean	$\mu_{\text{difference}}$
-0.0173	0.0923	0.0169	(-0.0518, 0.0172)

$\mu_{\text{difference}}$: mean of (TVOC - TVOC_1)

Test

Null hypothesis $H_0: \mu_{\text{difference}} = 0$

Alternative hypothesis $H_1: \mu_{\text{difference}} \neq 0$

T-Value	P-Value
-1.03	0.313

Discussion

The results show that most of the levels of chemical pollutants are high in the roads chosen when we look at the data from before the quarantine. For the **IBP Road**, the Surface temperature had an average score of 40.05C and a modal score of 48.2C, air temperature had an average level of 31.566C and a modal score of 35C, Carbon dioxide (CO₂) had an average score of 5370ppm and a modal score of 5890ppm, Carbon monoxide (CO) has an average score of 85ppm and a modal score 90ppm. Other chemical pollutants such as Formaldehyde (HCHO) had an average score of 0.413033mg/m³ and a modal score of 0.477mg/m³, while Total Volatile Organic Compounds (TVOC) has an average level of 0.8307ppm and a modal score of 1.071ppm.

For the **San Mateo Northview Road**, the Surface temperature had an average score of 49.55C and a modal score of 49.3 degrees Celsius, air temperature had an average level of 32.233C and a modal score of 33C, Carbon dioxide (CO₂) had an average score of 5180.3ppm and a modal score of 5776ppm, Carbon monoxide (CO) has an average score of 84.533ppm and a modal score 86ppm. Other chemical pollutants such as Formaldehyde (HCHO) had an average score of 0.41676ppm and a modal score of 0.478, while Total Volatile Organic Compounds (TVOC) has an average level of 0.8344ppm and a modal score of 0.976ppm.

For the **Don Antonio Overpass**, the Surface temperature had an average score of 41.6133 C and a modal score of 48.2C, air temperature had an average level of 32.233C and a modal score of 34C, Carbon dioxide (CO₂) had an average score of 5330ppm and a modal score of 5890ppm, Carbon monoxide (CO) has an average score of 85.09ppm and a modal score 82ppm. Other chemical pollutants such as Formaldehyde (HCHO) had an average score of

0.4788ppm and a modal score of 0.478ppm, while Total Volatile Organic Compounds (TVOC) has an average level of 0.834ppm and a modal score of 0.926ppm.

Meanwhile, the results show that most of the levels of chemical pollutants are lower in the roads chosen when we look at the data from after the quarantine excluding the temperatures. For the **IBP Road**, the Surface temperature had an average score of 45.95C and a modal score of 48.3C, air temperature had an average level of 42.23C and a modal score of 47C, Carbon dioxide (CO₂) had an average score of 43.68ppm and a modal score of 48.9ppm. Other chemical pollutants such as Formaldehyde (HCHO) and Total Volatile Organic Compounds (TVOC) had fairly the same average and modal score as the first data sets.

For the **San Mateo Northview Road**, the Surface temperature had an average score of 51.62C and a modal score of 589.3 degrees Celsius, air temperature had an average level of 42.9C and a modal score of 49C, Carbon dioxide (CO₂) had an average score of 43.895ppm and a modal score of 49.8ppm. Other chemical pollutants such as Formaldehyde (HCHO) and Total Volatile Organic Compounds (TVOC) had fairly the same average and modal score as the first data sets.

For the **Don Antonio Overpass**, the Surface temperature had an average score of 51.88C and a modal score of 58.3C, air temperature had an average level of 42.43C and a modal score of 45C, Carbon dioxide (CO₂) had an average score of 43.340ppm and a modal score of 48.3240ppm. Other chemical pollutants such as Formaldehyde (HCHO) and Total Volatile Organic Compounds (TVOC) had fairly the same average and modal score as the first data sets.

In comparing previous data sets recorded from Pre- and Post Quarantine, the t-test experiment displayed the following results.

For the **Batasan Hills Overpass area**, the Urban Heat Parameters (Surface Temperature and Air Temperature), and CO₂ showed high intensity differences in terms of data. The Urban Heat Parameters showed positive differences, meaning their values were higher than the older data. Meanwhile, CO₂ levels showed negative differences, meaning their values drifted lower when compared to different data.

Meanwhile, the other Greenhouse Gas parameters (CO, HCHO and TVOC) showed low intensity differences in terms of data. CO and TVOC had negative differences, meaning their values were lower than old data, while HCHO on the other hand became higher, having a positive value.

For the **San Mateo Road area**, the Urban Heat Parameters (Surface Temperature and Air Temperature), and CO₂ also showed high intensity differences in terms of data. The Urban Heat Parameters showed positive differences again, meaning their values were higher than the older data. Meanwhile, CO₂ levels showed negative differences also, meaning their values drifted lower when compared to different data.

Meanwhile, the other Greenhouse Gas parameters (CO, HCHO and TVOC) showed low intensity differences in terms of data. CO and HCHO had negative differences, meaning their values were lower than old data, while TVOC on the other hand became higher, having a positive value.

For the **Don Antonio Overpass**, the Urban Heat Parameters (Surface Temperature and Air Temperature), and CO₂ showed high intensity differences in terms of data once again. The Urban Heat Parameters showed positive differences, meaning their values were higher than the older data only this time—Surface Temperature had a lower difference between pre and current

quarantine timelines. Meanwhile, CO₂ levels showed negative differences again, meaning their values drifted lower when compared to different data. Meanwhile, the other Greenhouse Gas parameters (CO, HCHO and TVOC), like the other locations, showed low intensity differences in terms of data. CO and TVOC had negative differences, meaning their values were lower than old data, while HCHO on the other hand became higher again, having a positive value.

Here is a diagram showing a clearer explanation of the relationships between the data of each parameter from pre- and post quarantine timelines.

Batasan Hills Overpass

<i>Parameter</i>	Positive or Negative Difference	Height of Difference
ST	+	High
AT	+	High
CO	-	High
CO ₂	-	Low
HCHO	+	Low
TVOC	-	Low

San Mateo Road

<i>Parameter</i>	Positive or Negative Difference	Height of Difference
ST	+	High
AT	+	High
CO	-	High
CO ₂	-	Low
HCHO	-	Low
TVOC	+	Low

Don Antonio Overpass

<i>Parameter</i>	Positive or Negative Difference	Height of Difference
ST	+	Low
AT	+	High
CO	-	High
CO ₂	-	Low
HCHO	+	Low
TVOC	-	Low

Data T-Test Interpretation

The extraneous cloud data suggest that during the days with less cloud cover, temperatures were less than that of when clouds were overcast, which corresponds to the higher humidity levels during those same days.

Conclusion

The investigation has led to the discovery of multiple things. First, the data from the first collection shows the high levels of Urban Heat temperatures and Greenhouse Gases. Compared to the second collection, Urban Heat temperatures have risen, meanwhile Greenhouse Gases have decreased. The researcher notices there were big differences between the two data sets and most of these differences were found with the Urban Heat Parameters rising and the Greenhouse Gas Parameters dropping.

Overall, across all data sets, 44.44% of the factors had high differences (both increased and decreased) when compared to old data. This had increased to 61.47% after around three to four years.

This tells us that although there is a significance between the two factors (considering how heat affects diffusion rate of these gases and how the quarantine period greatly decreased the GHG emissions), we can't solve both problems at once even if we tried. To explain further, heat is expected to come from the built-up gases in the atmosphere, but because these gases were produced excessively and suddenly cut off because of the quarantine, we expected climate change to at least decrease.

This research shows that because we had already done damage by producing so many chemically dangerous gases in the air in the past, simply stopping the production of these gases won't solve all of our issues with temperature. The data shows that because of quarantine, we produced less dangerous gases, meaning that if we all collectively worked together to decrease the activities that may produce these gases, we can hopefully slow down climate change even if we cannot solve it immediately.

Recommendation

When creating research in collecting data as such, it is recommended that researchers and students follow specific protocol to get accurate data. It's recommended also that the researcher use statistical treatment through software that produces graphs for the data to help the researcher create matrix plots of the correlations of their data.

The researcher recommends conducting the research with caution, of course, the researcher should wear a mask because inhalation of these chemicals, although small amounts, can still lead to a few complications in terms of health. It is also recommended to collect data in more urban areas when conducting trials and collecting data to help explore more of the pollutant production around the globe to find out how much pollutants and heat are being excreted by these locations.

\

CHAPTER 5

BIBLIOGRAPHY

This part of the research is a list of books, scholarly articles, speeches, private records, diaries, interviews, laws, letters, websites, and other sources the researcher used when researching the topic and writing the paper.

Bibliography/References:

1. Naghizadeh, A., Sharifzadeh, G., Tabatabaei, F. et al. Environ Geochem Health. *Pitfalls in diagnosis and management of carbon monoxide poisoning*. (2019). <https://doi.org/10.1007/s10653-018-0226-5>
2. Klepeis, N.E.; Nelson, W.C.; Ott, W.R.; Robinson, J.P.; Tsang, A.M.; Switzer, P.; Behar, J.V.; Hern, S.C.; Engelmann, W.H. The National Human Activity Pattern Survey (NHAPS): A resource for assessing exposure to environmental pollutants. *J. Expo. Sci. Environ. Epidemiol.* 2001, 11, 231–252.
3. Sexton, K.; Gong, H., Jr.; Bailar, J.C.; Ford, J.G.; Gold, D.R.; Lambert, W.E.; Utell, M.J. Air Pollution Health Risks: Do Class and Race Matter? *Toxicol. Ind. Health* 1993, 9, 843–878
4. Salthammer, T. (2013). Formaldehyde in the ambient atmosphere - from an indoor pollutant to an outdoor pollutant? *ISEE Conference Abstracts*, 2013(1), 3365. doi: 10.1289/isee.2013.o-4-22-01

5. Gustafson, P., Barregård, L., Lindahl, R., & Sällsten, G. (2004). Formaldehyde levels in Sweden: personal exposure, indoor, and outdoor concentrations. *Journal of Exposure Science & Environmental Epidemiology*, 15(3), 252–260. doi: 10.1038/sj.jea.7500399
6. Oke, T. (n.d.). Urban heat islands. *The Routledge Handbook of Urban Ecology*. doi: 10.4324/9780203839263.ch11
7. EPA (2001) Environmental Protection Agency. Website: <http://www.epa.gov/airtrends/>
8. Sexton, K.; Gong, H., Jr.; Bailar, J.C.; Ford, J.G.; Gold, D.R.; Lambert, W.E.; Utell, M.J. Air Pollution Health Risks: Do Class and Race Matter? *Toxicol. Ind. Health* 1993, 9, 843–878
9. Jetter, J.J.; Kariher, P. Solid-fuel household cooking stoves: Characterization of performance and emissions. *Biomass Bioenergy* 2009, 33, 294–305.
10. Hulin M, Caillaud D, Annesi-Maesano I (2010) Indoor air pollution and childhood asthma: variations between urban and rural areas. *Indoor Air* 20:502–514
11. Balzan, M.V., Agius, G., Debono, A.G., 1996. Carbon monoxide poisoning: easy to treat but difficult to recognize. *Postgrad. Med. J.* 72, 470–473.
12. Coburn, R.F. (Ed.), 1970. The carbon monoxide body stores. In: *Biological Effects of Carbon Monoxide*. *Ann. N.Y. Acad. Sci.* 174, 11–22.
13. Federal Register, 1994. National ambient air quality standards for carbon monoxide — final decision. *F. R.* (August 1) 59, 38, 906–938, 917.

14. Garland, A., Pearce, J., 1967. Neurological complications of carbon monoxide poisoning. *Q. J. Med.* 36, 445–451.
15. Garland, A., Pearce, J., 1967. Neurological complications of carbon monoxide poisoning. *Q. J. Med.* 36, 445–451.
16. Roy, B., Crawford, R., 1996. Pitfalls in diagnosis and management of carbon monoxide poisoning. *J. Accid. Emerg. Med.* 9, 62–63.
17. Scheinkestel, C.D., Bailey, M., Myles, P.S., Jones, K., Cooper, D.J., Millar, I.L., Tuxen, D.V., 1999. Hyperbaric or normobaric oxygen for acute carbon monoxide poisoning: a randomized controlled clinical trial. *Med. J. Aust.* 170, 203–210 (available as of 5 January 2000 at: www.mja.com.au/public/issues/mar1/scheink/scheink.html).
18. Klepeis, N.E.; Nelson, W.C.; Ott, W.R.; Robinson, J.P.; Tsang, A.M.; Switzer, P.; Behar, J.V.; Hern, S.C.; Engelmann, W.H. The National Human Activity Pattern Survey (NHAPS): A resource for assessing exposure to environmental pollutants. *J. Expo. Sci. Environ. Epidemiol.* 2001, 11, 231–252.
19. Sexton, K.; Gong, H., Jr.; Bailar, J.C.; Ford, J.G.; Gold, D.R.; Lambert, W.E.; Utell, M.J. Air Pollution Health Risks: Do Class and Race Matter? *Toxicol. Ind. Health* 1993, 9, 843–878
20. Gustafson, P., Barregård, L., Lindahl, R., & Sällsten, G. (2004). Formaldehyde levels in Sweden: personal exposure, indoor, and outdoor concentrations. *Journal of Exposure Science & Environmental Epidemiology*, 15(3), 252–260. doi: 10.1038/sj.jea.7500399

21. Oke, T. (n.d.). Urban heat islands. *The Routledge Handbook of Urban Ecology*.
doi: 10.4324/9780203839263.ch11
22. EPA (2001) Environmental Protection Agency. Website: <http://www.epa.gov/airtrends/>.
23. Sexton, K.; Gong, H., Jr.; Bailar, J.C.; Ford, J.G.; Gold, D.R.; Lambert, W.E.; Utell, M.J. Air Pollution Health Risks: Do Class and Race Matter? *Toxicol. Ind. Health* 1993, 9, 843–878
24. Jetter, J.J.; Kariher, P. Solid-fuel household cooking stoves: Characterization of performance and emissions. *Biomass Bioenergy* 2009, 33, 294–305.
25. Hulin M, Caillaud D, Annesi-Maesano I (2010) Indoor air pollution and childhood asthma: variations between urban and rural areas. *Indoor Air* 20:502–514
26. Taha H., Konopacki S. and Gaberseck S. (1996) Modeling the Rosenfeld A. H., Romm J. J., Akbari H. and Pomerantz M. Meteorological and Energy Effects of Urban Heat Islands (1998) *Cool communities: strategies for heat islands mitigation and their Mitigation: A 10-Region Study*, Lawrence Berkeley and smog reduction. *Energy and Buildings* 28, 51–62. keley Laboratory Report LBL-38667, Berkeley, CA.
27. Stock, T.H., 1987. Formaldehyde concentrations inside conventional housing. *Journal of the Air Pollution Control Association* 37, 913–918.
28. Milford, J.B., G.R. Armistead, and G.J. McRae. 1989. "A New Approach to Photochemical Pollution Control: Implications of Spatial Patterns in Pollutant

Responses to Reduction in Nitrogen Oxides and Reactive Organic Emissions,"
Environmental Science and Technology, 23, pp. 1290-1301.

29. Baek, S.-O., Kim, Y.-S., Perry, R., 1997. Indoor air quality in homes, offices, and restaurants in Korean urban areas— indoor/outdoor relationships. *Atmospheric Environment* 31, 529–544.
30. Brauer, B., K. Bartlett, J. Regaldo-Pineda, R. Perez-Padilla. 1996. "Assessment of Particulate Concentrations from Domestic Biomass Combustion in Rural Mexico." *Environmental Science and Technology* 30: 104–9.
31. Smith, K.R. (2000). "National Burden of Disease in India from Indoor Air Pollution," *Proceedings of the National Academy of Sciences of the United States of America*. 97(24): 13286-13293.
32. Persily, A. 1997. Evaluating building IAQ and ventilation with indoor carbon dioxide. *ASHRAE Transactions* 102(2).
33. JOUR, Grierson, David, 2007/12/01 "The Urban Environment: Agendas and Problems", *The International Journal of Environmental, Cultural, Economic and Social Sustainability*. 10.18848/1832-2077/GP/v03i01/54314.
34. Richard Schmalensee, Thomas M. Stoker, Ruth A. Judson; World Carbon Dioxide Emissions: 1950–2050. *The Review of Economics and Statistics* 1 February 1998; 80 (1): 15–27. doi: <https://doi.org/10.1162/003465398557294>
35. Carbon Dioxide Health Hazard Information Sheet [PDF]. (n.d.). FSIS Environmental, Safety and Health Group. Retrieved March 19, 2021
36. Material Safety Data Sheet for Methane [PDF]. (2002). North Branch, New Jersey: © Voltaix, Inc.

37. Cappel, K. (n.d.). Methane Emissions [PPT]. US EPA.
38. Riverside Resource Recovery emission report – October 2020 [PDF]. (2020). Cory Energy.
39. Lorente, A., Boersma, K.F., Eskes, H.J. et al. Quantification of nitrogen oxides emissions from build-up of pollution over Paris with TROPOMI. *Sci Rep* 9, 20033 (2019). <https://doi.org/10.1038/s41598-019-56428-5>
40. Ozone trends: A review [PDF]. (retrieved march 19, 2021). US EPA.
41. Ahmed Memon RIZWAN, Leung Y.C. DENNIS, Chunho LIU, A review on the generation, determination and mitigation of Urban Heat Island, *Journal of Environmental Sciences*, Volume 20, Issue 1, 2008, Pages 120-128, ISSN 1001-0742, [https://doi.org/10.1016/S1001-0742\(08\)60019-4](https://doi.org/10.1016/S1001-0742(08)60019-4). (<https://www.sciencedirect.com/science/article/pii/S1001074208600194>)
42. Darkwah, Williams Kweku & Odum, Bismark & Addae, Maxwell & Koomson, Desmond & Kwakye Danso, Benjamin & Oti-Mensah, Ewurabena & Asenso, Theophilus & Buanya, Beryl. (2018). Greenhouse Effect: Greenhouse Gases and Their Impact on Global Warming. *Journal of Scientific Research and Reports*. 17. 1-9. 10.9734/JSRR/2017/39630.
43. Lu, Z., Streets, D. G., de Foy, B., Lamsal, L. N., Duncan, B. N., and Xing, J.: Emissions of nitrogen oxides from US urban areas: estimation from Ozone Monitoring Instrument retrievals for 2005–2014, *Atmos. Chem. Phys.*, 15, 10367–10383, <https://doi.org/10.5194/acp-15-10367-2015>, 2015.
44. The Carbon Trust (2012) Carbon Footprinting.

45. Jones C., Kammen D. (2011) “Quantifying Carbon Footprint Reduction Opportunities for U.S. Households and Communities.”
46. Heller, M.C., et al. (2018). Greenhouse gas emissions and energy use associated with production of individual self-selected US diets. *Environmental Research Letters*, 13(4), 044004.
47. Boehm R., et al. (2018) “A Comprehensive Life Cycle Assessment of Greenhouse Gas Emissions from U.S. Household Food Choices.”
48. Weber, C. and H. Matthews (2008) “Food miles and the Relative Climate Impacts of Food Choices in the United States.” *Environmental Science & Technology*, 42(10): 3508-3513.
49. U.S. Environmental Protection Agency (EPA) (2020) Inventory of U.S. Greenhouse Gas Emissions and Sinks 1990 - 2018.
50. Heller, M., et al. (2020). Implications of Future US Diet Scenarios on Greenhouse Gas Emissions.
51. U.S. EPA (2020) “Emissions & Generation Resource Integrated Database (eGRID).”
52. U.S. Energy Information Administration (EIA) (2020) Electric Power Monthly with Data from February 2020.
53. U.S. EIA (2020) Annual Energy Outlook 2020.
54. U.S. EIA (2020) Residential Energy Consumption Survey 2015.
55. Mars C., (2016) Benefits of Using Cold Water for Everyday Laundry in the U.S.
56. Heller, M. and G. Keoleian. (2014) Greenhouse gas emissions estimates of U.S. dietary choices and food loss. *Journal of Industrial Ecology*, 19 (3): 391-401.

57. U.S. EPA (2020) The 2019 EPA Automotive Trends Report: Greenhouse Gas Emissions, Fuel Economy, and Technology since 1975.
58. U.S. Department of Energy (DOE), Oak Ridge National Lab (2020) Transportation Energy Data Book: Edition 38.1.
59. Pero, F. et al. (2018) Life Cycle Assessment in the automotive sector: a comparative case study of Internal Combustion Engine and electric car. 17. U.S. EIA (2016) “How Much Carbon Dioxide is Produced by Burning Gasoline and Diesel Fuel.”
60. Center for Sustainable Systems, University of Michigan. 2020. “Carbon Footprint Factsheet.” Pub. No. CSS09-05.

Appendix

Don Antonio Post-Quarantine(2 Years Later) Raw Data Digitized:

<https://acesse.one/donantonioNEW>

IBP Road Post-Quarantine(2 Years Later) Raw Data Digitized: <https://link.dev/ibproadNEW>

San Mateo Road Post-Quarantine(2 Years Later) Raw Data Digitized:

<https://acesse.one/sanmateoNEW>

Don Antonio Pre-Quarantine Raw Data Digitized: <https://link.dev/donantoniobefore>

IBP Pre-Quarantine Raw Data Digitized: <https://acesse.one/IBProadBEFORE>

San Mateo Pre-Quarantine Raw Data Digitized: <https://acesse.one/sanmateobefore>

Materials Links:

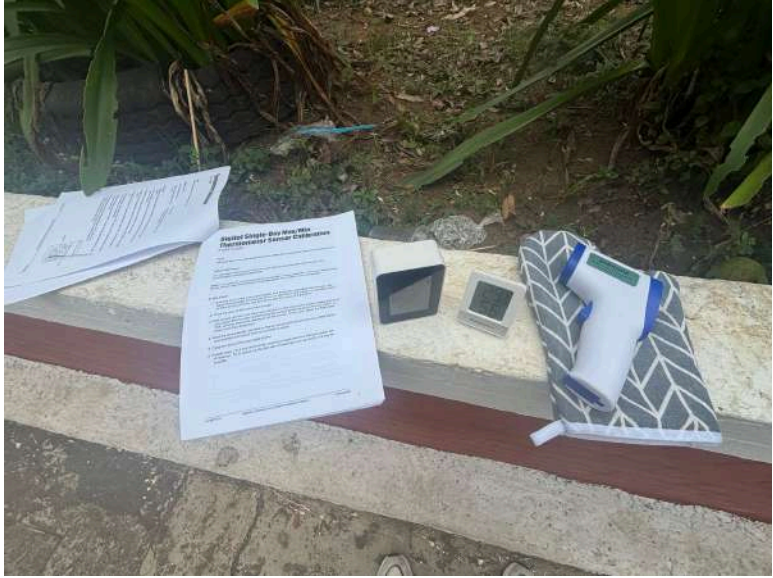
Infrared Thermometer - <https://tinyurl.com/3tpbswhc>

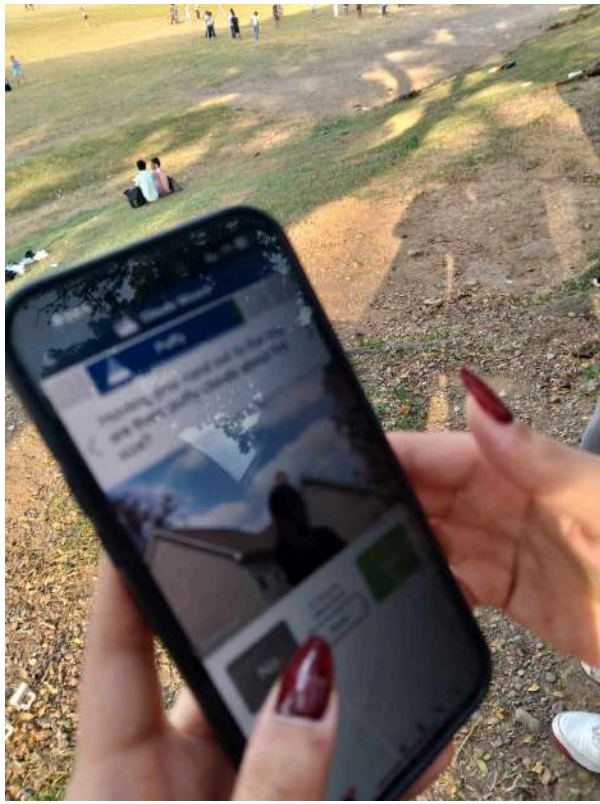
Greenhouse Gas Detector - <https://tinyurl.com/y3s555hw>

Air Thermometer - <https://tinyurl.com/2n3r6eu3>



Documentation







Infrared Thermometer

Carbon Monoxide (CO) Detector

TVOC (Total Volatile Organic Compound) and Formaldehyde (HCHO) Detector

Air Thermometer, Barometer, Humidity Sensor and Carbon Dioxide Level Detector



ibp road

day 1



day 2



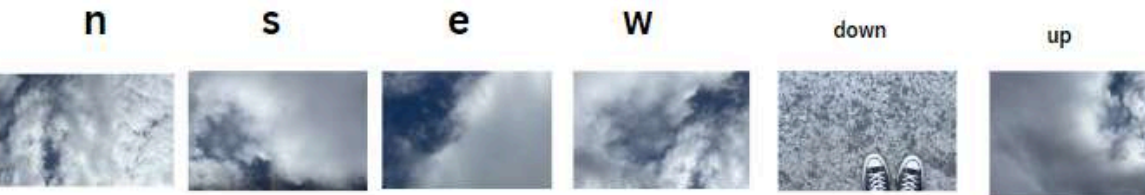
day 3



day 4

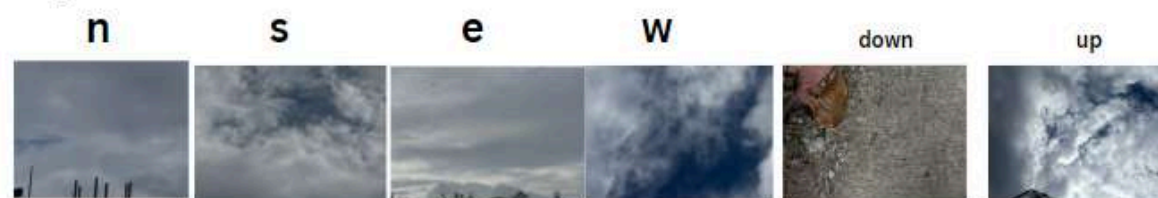


day 5

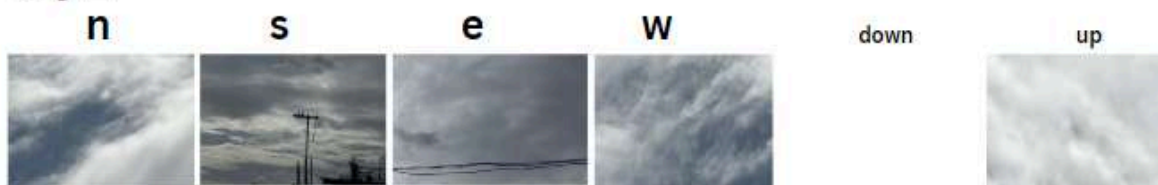


don antonio

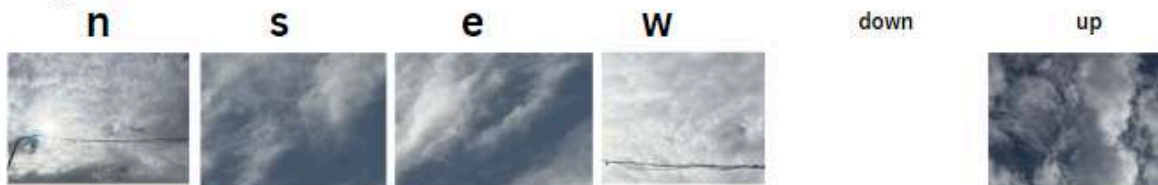
day 1



day 2



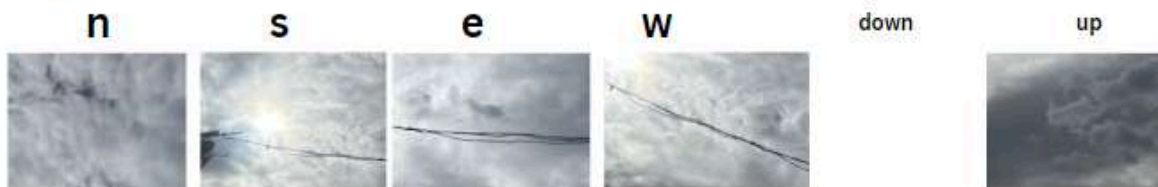
day 3



day 4



day 5



san mateo

day 1

n

s

e

w

down

up



day 2

n

s

e

w

down

up



day 3

n

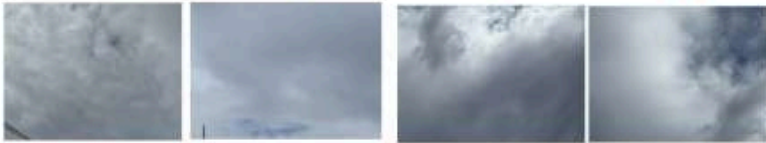
s

e

w

down

up



day 4

n

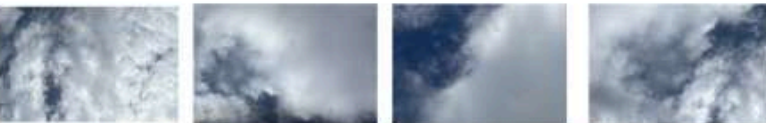
s

e

w

down

up



day 5

n

s

e

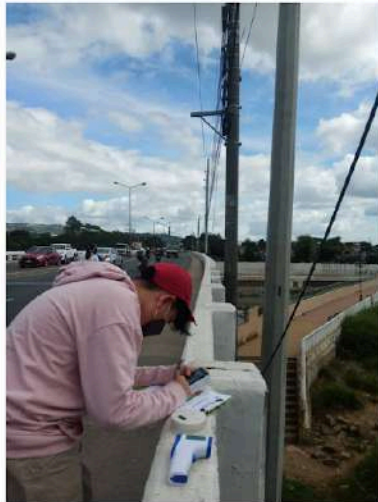
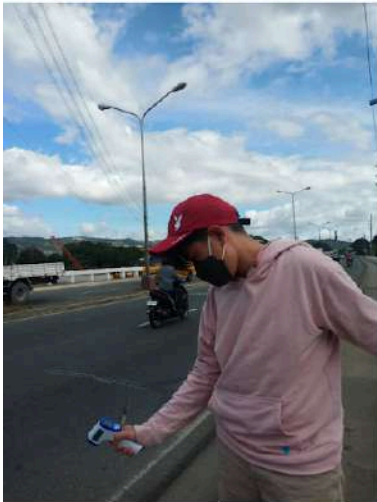
w

down

up



San Mateo Road Data Collection



Time/Day		CO2	CH4	N2O	Relative Humidity	Cloud Formation	Weather	Air Temperature	Surface Temperature
Day	1								
3:00		472	470	463	461	453	449	403	300
3:10		485	482	474	470	462	458	412	305
3:20		476	473	465	461	453	449	403	300
3:30		480	477	469	465	457	453	417	305
3:40		485	482	474	470	462	458	412	305
4:00		472	470	463	461	453	449	403	300
Day	2								
3:00		470	467	459	455	447	443	397	295
3:10		480	477	469	465	457	453	417	305
3:20		475	472	464	460	452	448	402	298
3:30		480	477	469	465	457	453	417	305
3:40		485	482	474	470	462	458	412	305
4:00		470	467	459	455	447	443	397	295

Time/Day		CO2	CH4	N2O	Relative Humidity	Cloud Formation	Weather	Air Temperature	Surface Temperature
Day	3								
3:00		463	461	453	449	441	437	391	290
3:10		474	471	463	459	451	447	401	295
3:20		479	476	468	464	456	452	406	300
3:30		484	481	473	469	461	457	411	305
3:40		489	486	478	474	466	462	416	310
4:00		474	471	463	459	451	447	395	290
Day	4								
3:00		460	457	449	445	437	433	387	285
3:10		470	467	459	455	447	443	397	290
3:20		475	472	464	460	452	448	402	295
3:30		480	477	469	465	457	453	417	305
3:40		485	482	474	470	462	458	412	305
4:00		470	467	459	455	447	443	397	295