




# Capturing & Exploring Datasets

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# Project Objectives

- To capture data sets and/or performing preliminary data explorations & analytics.
  - Outcome will lead to the set up of a centralised data repository/source for teaching & learning
- 

# Task Distribution



**Nicolas Tan**

Web Scraping  
Data Cleaning  
Mapping of Address



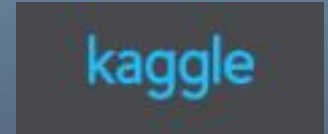
**Charlie Quah**

KNIME Exploration  
Data Visualisation

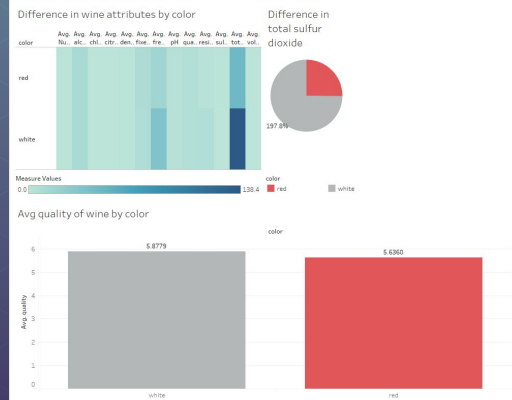
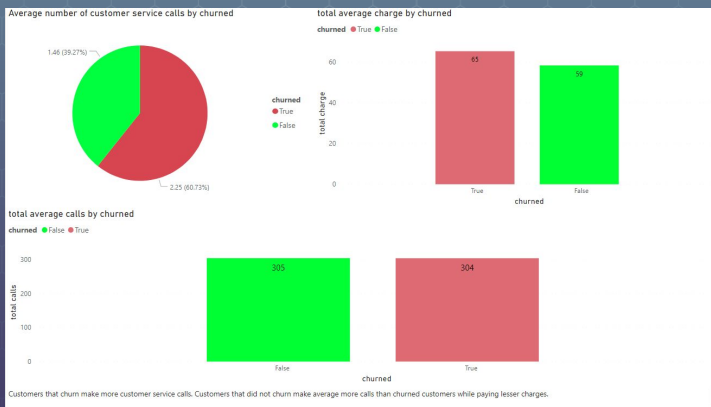
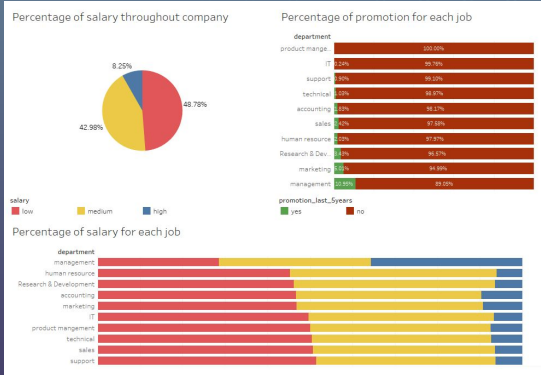
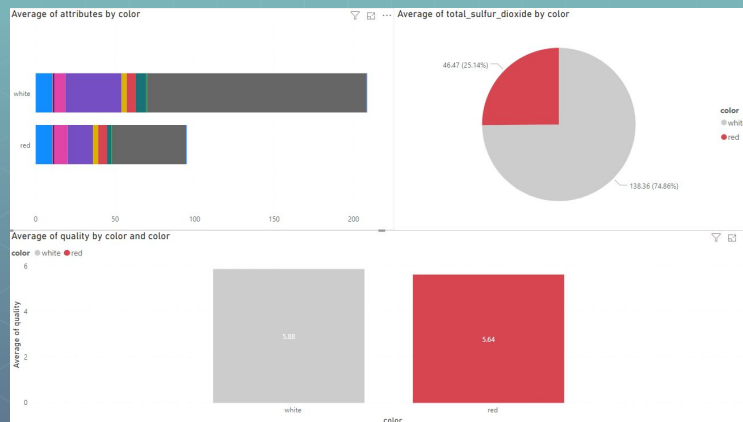
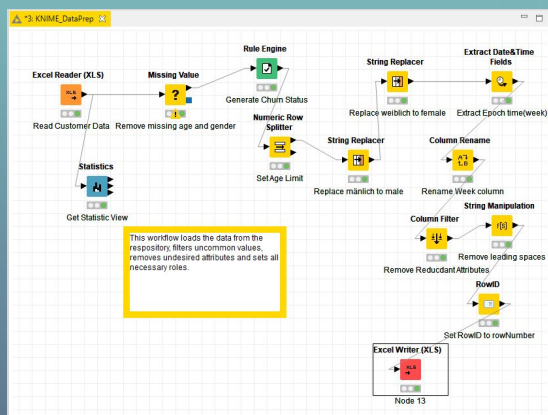
# Recap



- Knime data preparation
- Knime analytics (decision tree & linear regression)
- Knime mapping
- HR, Wine, Crime and Telecom dataset visualisation using Tableau and Power BI.



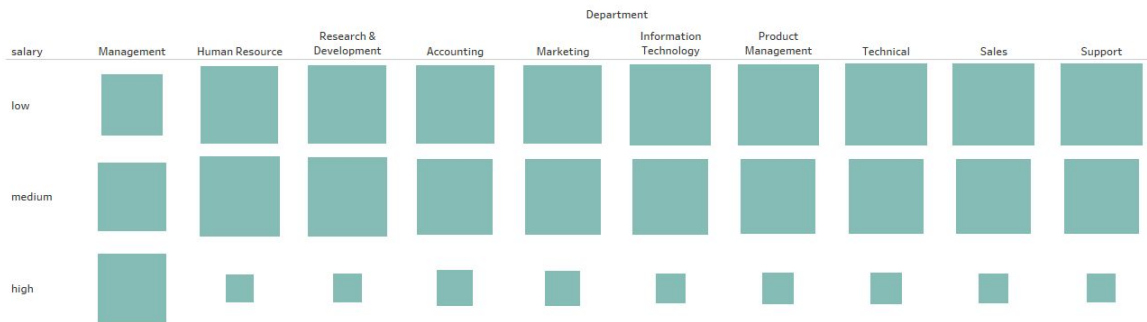
# Before



# HR dataset

## Job Information

### Percentage Distribution Of Salary



### Work Accident Count

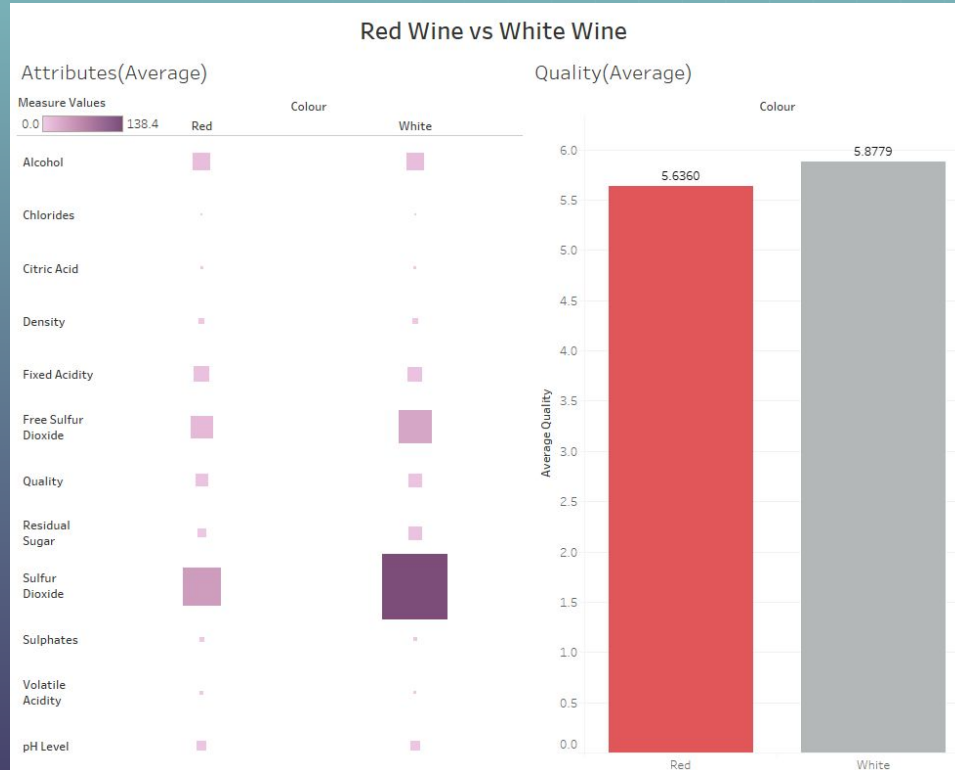
work_accident	Department	Count
yes	Sales	587
	Technical	381
	Support	345
	Information Technology	164
	Marketing	138
	Research & Development	134
	Product Management	132
	Management	103
	Accounting	96

### Promotion Percentage(Last 5 Years)

promotion..	Department									
	Management	Marketing	Research & Development	Sales	Human Resource	Accounting	Technical	Support	Information Technology	Product Management
yes	10.95%	5.01%	3.43%	2.42%	2.03%	1.83%	1.03%	0.90%	0.24%	
no	89.05%	94.99%	96.57%	97.58%	97.97%	98.17%	98.97%	99.10%	99.76%	100.00%

- View company insights.

# Wine dataset



- Compare red and white wine.

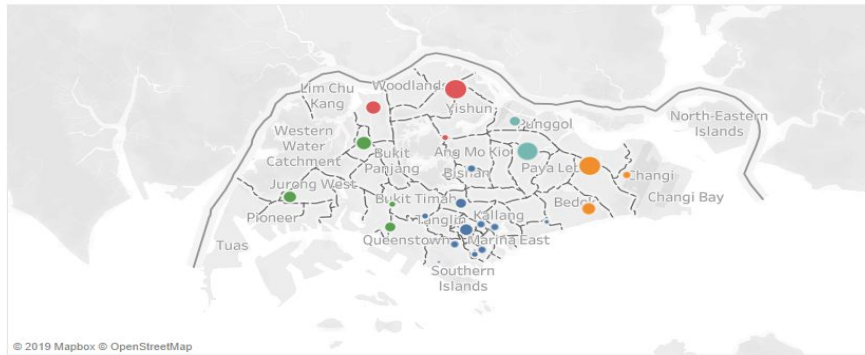
# Crime Dataset

## Crime Location Analysis

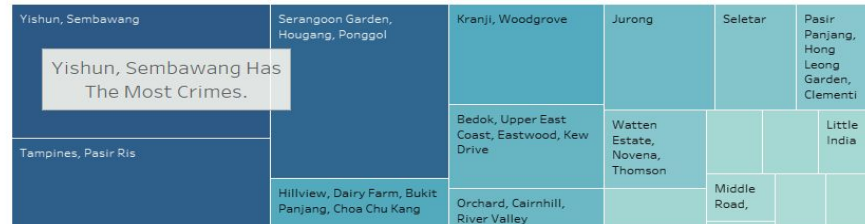
< Which Area Has The Most Crime?

Which Area Has The Highest Increasing Crime Rate? >

Crime count per Town



Crime Treemap



- View insights about location of crimes.



# Criminal dataset

## Criminals Arrested

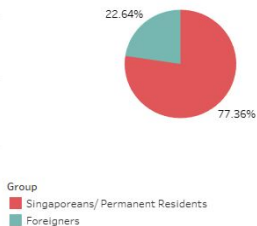
< Who Are Mostly Arrested & What For? Which Top Crimes Are On The Rise? >

### Statistics

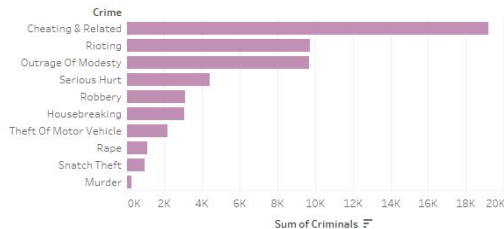
#### Over The Years



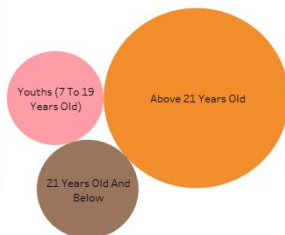
#### Nationality



#### Types Of Crime

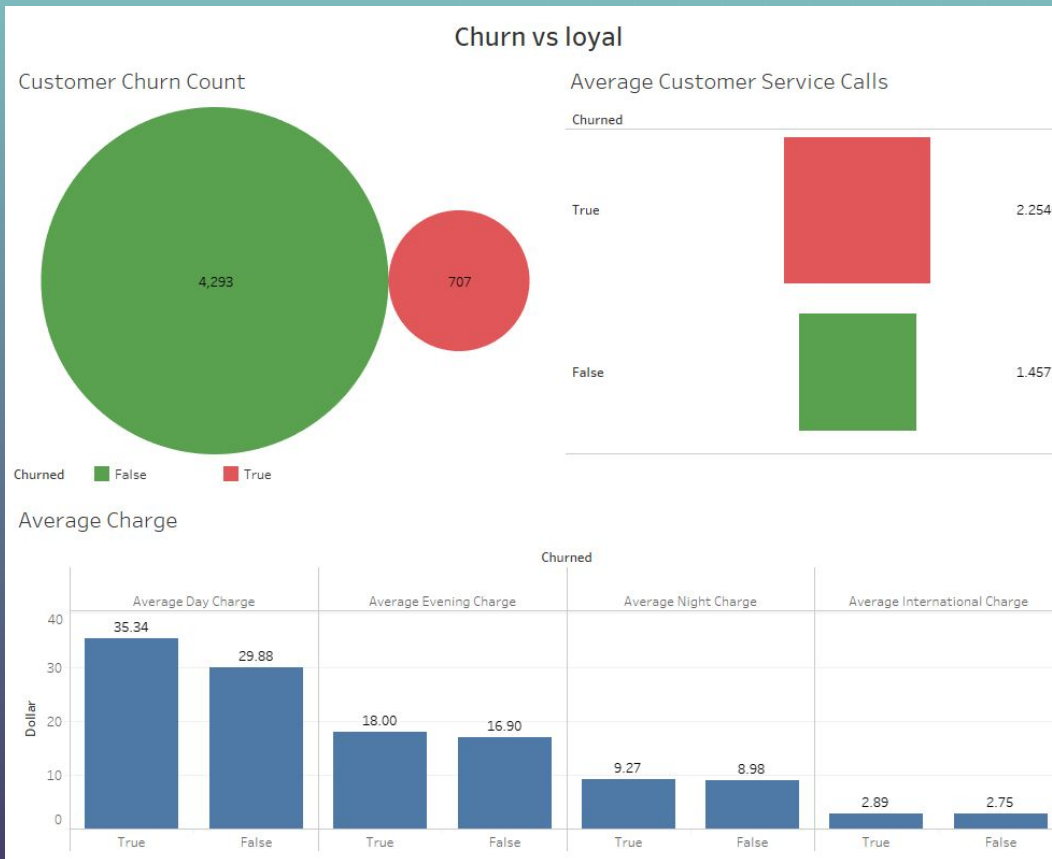


#### Age



- Get criminals insights.

# Telecom Customer Churn Dataset



- View insights about customer.

# World happiness dataset

## World Happiness

Top 10 corrupted countries



top 10 corrupted countries

- ☐ (All)
- ☐ False
- ☒ True

Top 10 countries with highest positive affect

Country (region)	
Turkey	154
Yemen	153
Afghanistan	152
Iraq	151
Lebanon	150
Belarus	149
Serbia	148
Tunisia	147
Egypt	146
Bangladesh	145

Top 10 countries with highest negative affect

Country (region)	
Iraq	154
Central African Republic	153
South Sudan	152
Chad	151
Iran	150
Sierra Leone	149
Benin	148
Togo	147
Liberia	146
Armenia	145

top 10 positive

- ☐ (All)
- ☐ False
- ☒ True

top 10 negative

- ☐ (All)
- ☐ False
- ☒ True

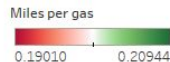
- Understand countries' placing

# Trucks dataset

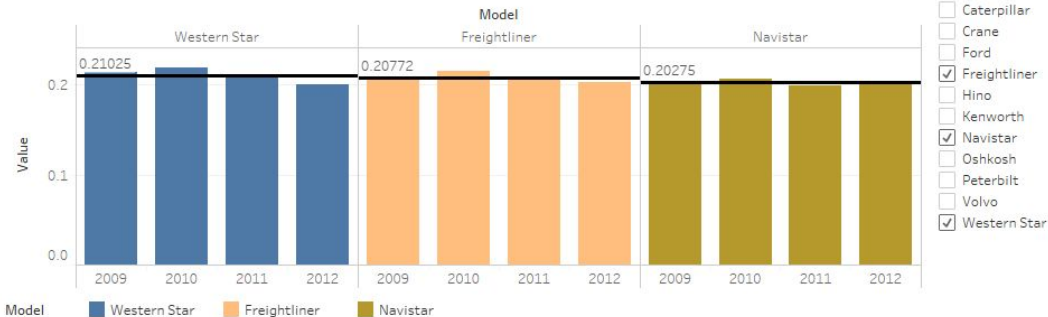
## Fuel Economy

Miles per gas unit

Model	
Western Star	0.20944
Freightliner	0.20572
Navistar	0.20335
Kenworth	0.20235
Peterbilt	0.20102
Hino	0.20074
Oshkosh	0.19960
Caterpillar	0.19797
Ford	0.19703
Volvo	0.19475
Crane	0.19010



Top 3 fuel efficient trucks(miles per gas)

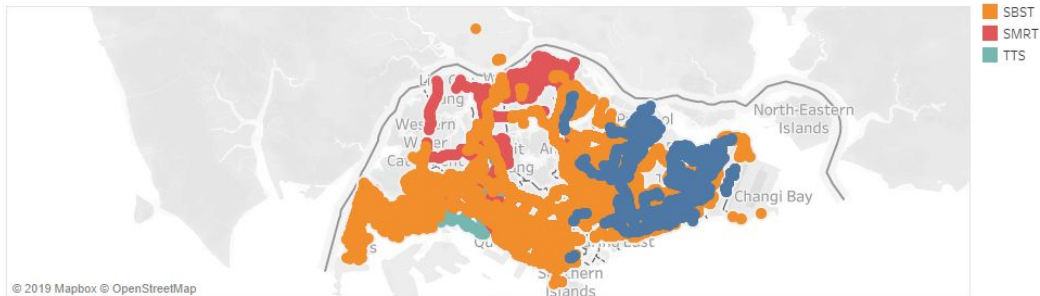


- View information about different trucks.

# Bus dataset

Locations Covered By Operators

Bus Stop Location



Number Of Bus Stops By Operator



Count of Bus Stop

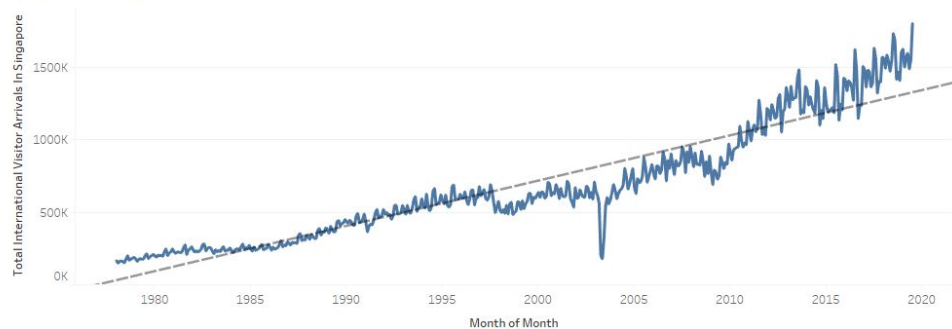


- Get bus services information.

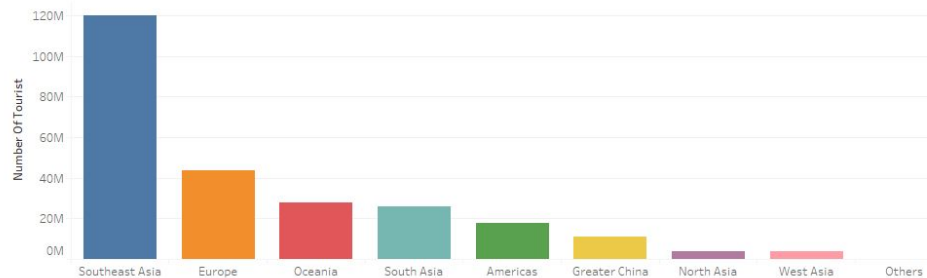
# Tourism dataset

Singapore Tourist

Number Of Tourists



Origin Of Tourist That Visit Singapore



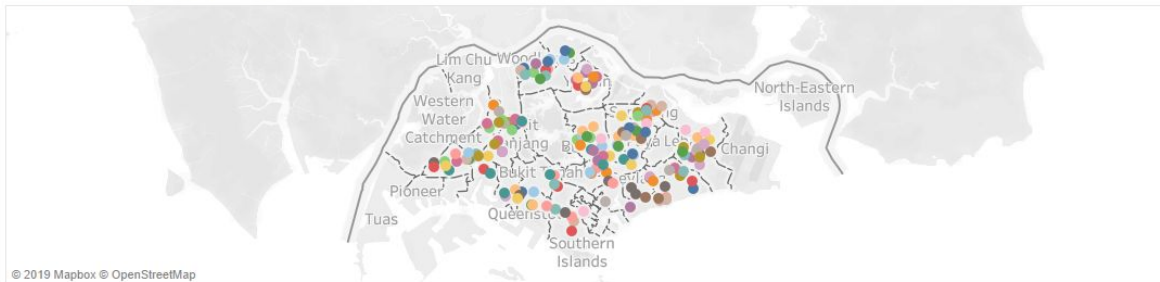
- View tourist information.

# School dataset

## Subjects Offered By School

Schools & Subjects Free Text Search

School Map(Click To Filter)



## Subject Count

Subject Desc

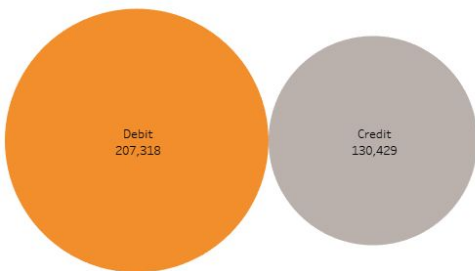
CIVICS & MORAL EDUCATION	2,747
MATHEMATICS	2,662
ENGLISH LANGUAGE	2,662
CHINESE	2,448
HIGHER CHINESE	2,421
ART	2,262
PHYSICAL EDUCATION	2,209
MALAY	2,204
MUSIC	2,101
SCIENCE	1,681
DRUGS	1,681

- Find information about schools.

# Banking dataset

## Deposit Account

Debit Vs Credit Transaction Count



Count Of Different Transaction Type

Transaction Type	Count
AWL	68,449
ATR	41,624
CAM	36,999
ADP	27,606
ITR	27,316
CDM	25,233
EZT	24,046
TTR	19,392
ACCT	18,782
NWL	18,709
GR	14,858
BILL	14,733

Accounts With I Banking



Accounts With P Banking



- View insights about bank accounts.

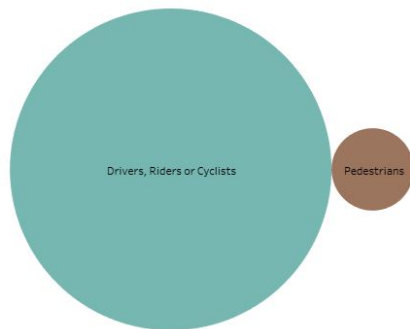


# Road accident dataset

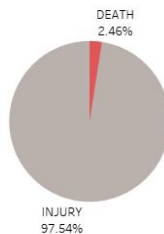


## Road Accident Statistics

Road User Involved



Injury Vs Death



Cause Of Accident

### Causes Of Accident

Failing to Keep a Proper Lookout	17,278
Failing to Have Proper Control	7,679
Failing to Give Way to Traffic with Right of Way	5,923
Other causes attributed to drivers, riders or pedal cyclists	3,926
Changing Lane without Due Care	3,449
Turning Without Due Care	1,304
Disobeying Traffic Light Signals Resulting in Accidents with Vehicle	1,192
Crossing Headless of Traffic	945
Other Causes of Accidents Attributed to Pedestrians	799
Driving under the Influence of Alcohol	613
Overtaking without Due Care	605
Turning Vehicle & Failing to Give Way to Pedestrian During Green Man	450
Crossing Within Pedestrian Crossing When Red Man Lighted	379
Other Causes	333
Failing to Use Available Pedestrian Crossings	222

- Get insights on road accidents.

# Address Web Scrape

Why?

Required a set of local addresses which are tied to a region for my mapper



# SRX

## What is SRX

- SRX is a property website (Sale/Rent)

## How?

- Used **python** (programming language) and 2 of libraries(**Pandas** & **BeautifulSoup**) on **Jupyter Notebook** (open source application for coding)
- Collected Addresses & their associated region
- Collected a total of approximately 10,000 Rows of Address Data.



# Outcome of SRX Scrape

Region	Street Name	Address	Postal Code
Ang Mo Kio	Ang Mo Kio Avenue 1	305 Ang Mo Kio Avenue 1	560305
Ang Mo Kio	Ang Mo Kio Avenue 1	332 Ang Mo Kio Avenue 1	560332
Ang Mo Kio	Ang Mo Kio Avenue 1	205 Ang Mo Kio Avenue 1	560205
Ang Mo Kio	Ang Mo Kio Avenue 1	321 Ang Mo Kio Avenue 1	560321
Ang Mo Kio	Ang Mo Kio Avenue 1	306 Ang Mo Kio Avenue 1	560306
Ang Mo Kio	Ang Mo Kio Avenue 1	333 Ang Mo Kio Avenue 1	560333
Ang Mo Kio	Ang Mo Kio Avenue 1	303 Ang Mo Kio Avenue 1	560303
Ang Mo Kio	Ang Mo Kio Avenue 1	319 Ang Mo Kio Avenue 1	560319
Ang Mo Kio	Ang Mo Kio Avenue 1	303 Ang Mo Kio Avenue 1	560303
Ang Mo Kio	Ang Mo Kio Avenue 1	308B Ang Mo Kio Avenue 1	562308
Ang Mo Kio	Ang Mo Kio Avenue 1	333 Ang Mo Kio Avenue 1	560333
Ang Mo Kio	Ang Mo Kio Avenue 1	319 Ang Mo Kio Avenue 1	560319
Ang Mo Kio	Ang Mo Kio Avenue 1	225 Ang Mo Kio Avenue 1	560225
Ang Mo Kio	Ang Mo Kio Avenue 1	219 Ang Mo Kio Avenue 1	560219
Ang Mo Kio	Ang Mo Kio Avenue 1	320 Ang Mo Kio Avenue 1	560320
Ang Mo Kio	Ang Mo Kio Avenue 1	226 Ang Mo Kio Avenue 1	560226
Ang Mo Kio	Ang Mo Kio Avenue 1	307C Ang Mo Kio Avenue 1	563307

# Mapping Foreign Datasets to Local Context



The datasets we are working on may not be local datasets. Hence the addresses in the datasets will be hard for students/adult learners to understand/visualize due to the unfamiliarity of the locations.

The purpose of mapping the datasets to singapore region is so that the user of the datasets can understand/visualize the datasets better  
(User of the data will know where “Yishun” is instead of “TX” or “TEXAS”)

# Recap of Previous Mapping Program

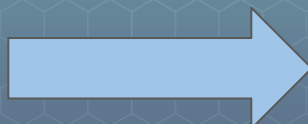


- ◀ Programmed using Python with the use of only “Pandas” Library
- ◀ Can only map Singapore Regions ( Yishun, Yio Chu Kang, Sengkang etc. )
- ◀ Every dataset would require the modification of codes before being able to execute the program.



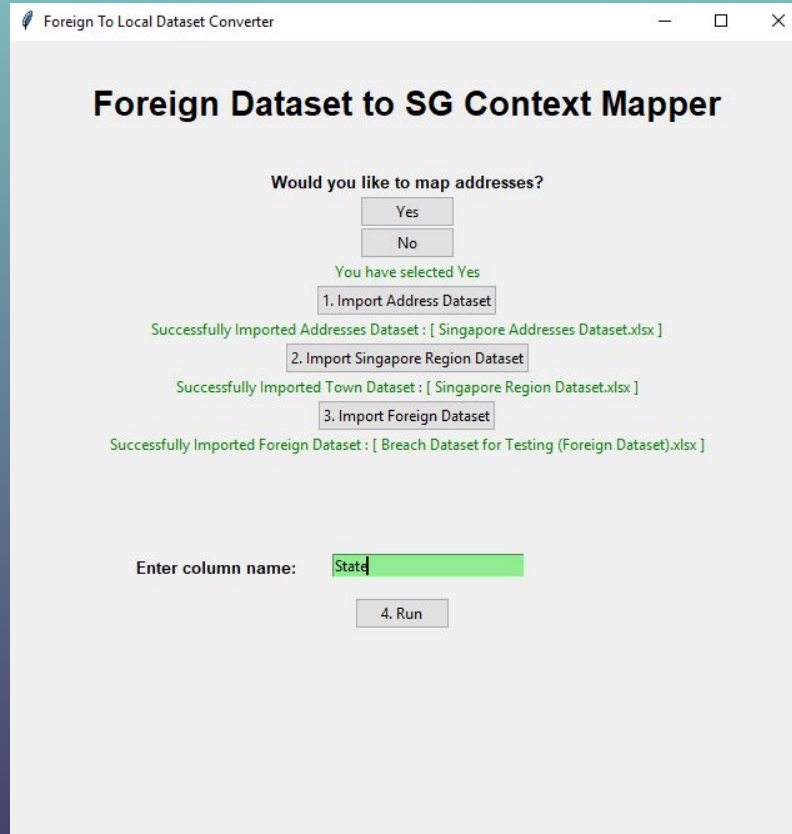
# Results of old mapper

Number	Name	State	Individual	Date_of_Breach	Type_of_B
1	Brooke	TX	1000	10/16/2009	Theft
2	Mid Am	MO	1000	9/22/2009	Theft
3	Alaska	AK	501	12/10/2009	Theft
4	Health	DC	3800	9/10/2009	Loss
5	Cogent	TN	6400	11/10/2009	Theft
6	Univers	NY	83000	12/11/2009	Other
7	Keith W	NC	2000	8/12/2009	Hacking/I
8	Detroit	MI	10000	10/22/2009	Theft
9	Detroit	MI	646	11/26/2009	Theft
10	Daniel J	MA	1860	11/12/2009	Theft
11	BlueCro	DC	3400	10/26/2009	Theft
12	Kaiser P	CA	15500	1/12/2009	Theft
13	Blue Isl	IL	2562	9/12/2009	Theft
14	Concer	TX	900	11/19/2009	Theft
15	Ashley	MO	9309	10/1/2010	Theft
16	Advocat	IL	812	11/24/2009	Theft
17	Carle C	IL	1300	1/13/2010	Theft
18	Educato	UT	5700	12/27/2009	Theft
19	Univers	NV	5103	10/31/2009	Theft
20	Brown	URI	528	11/12/2009	Other
21	Univers	NM	1900	8/2/2010	Other
22	Advanc	CA	3500	12/30/2009	Theft
23	Aspen	CO	2500	4/10/2009	Theft



Number	Name	Region	Individual	Date_of_E	Type_of_E
1	Brooke	Yishun	1000	#####	Theft
2	Mid Am	Queenstown	1000	#####	Theft
3	Alaska	Woodlands	501	#####	Theft
4	Health	Tampines	3800	#####	Loss
5	Cogent	Bishan	6400	#####	Theft
6	Univers	Sengkang	83000	#####	Other
7	Keith W	Bukit Panjang	2000	#####	Hacking/I
8	Detroit	Newton	10000	#####	Theft
9	Detroit	Newton	646	#####	Theft
10	Daniel J	Paya Lebar	1860	#####	Theft
11	BlueCro	Tampines	3400	#####	Theft
12	Kaiser P	Yio Chu Kang	15500	#####	Theft
13	Blue Isl	Boon Lay	2562	#####	Theft
14	Concer	Yishun	900	#####	Theft
15	Ashley	Queenstown	9309	#####	Theft
16	Advocat	Boon Lay	812	#####	Theft
17	Carle C	Boon Lay	1300	#####	Theft
18	Educato	Farrer Park	5700	#####	Theft
19	Univers	Hougang	5103	#####	Theft
20	Brown	Pasir Ris	528	#####	Other
21	Univers	Choa Chu Kang	1900	#####	Other
22	Advanc	Yio Chu Kang	3500	#####	Theft
23	Aspen	Redhill	2500	#####	Theft

# Updated Mapping Program



Foreign To Local Dataset Converter

## Foreign Dataset to SG Context Mapper

Would you like to map addresses?

Yes  
No

You have selected Yes

1. Import Address Dataset  
Successfully Imported Addresses Dataset : [ Singapore Addresses Dataset.xlsx ]

2. Import Singapore Region Dataset  
Successfully Imported Town Dataset : [ Singapore Region Dataset.xlsx ]

3. Import Foreign Dataset  
Successfully Imported Foreign Dataset : [ Breach Dataset for Testing (Foreign Dataset).xlsx ]

Enter column name: State

4. Run

- ◀ Programmed using Python with the use of “Pandas” & “Tkinter” Library.
- ◀ Can map to local Regions & assign them new columns such as “Address” & “Postal Code” to make dataset richer for analysis purposes.
- ◀ Graphical User interface which allows mapping process to be executed without any modification of codes (easier to use)

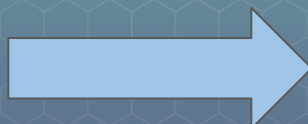


# Mapper Demo



# Results of new mapper

breach_start	year	Number	State
10/16/2009	2009	1	TX
9/22/2009	2009	2	MO
10/12/2009	2009	3	AK
10/9/2009	2009	4	DC
10/11/2009	2009	5	TN
11/12/2009	2009	6	NY
12/8/2009	2009	7	NC
10/22/2009	2009	8	MI
11/26/2009	2009	9	MI
12/11/2009	2009	10	MA
10/26/2009	2009	11	DC
12/1/2009	2009	12	CA
12/9/2009	2009	13	IL
11/19/2009	2009	14	TX
1/10/2010	2010	15	MO
11/24/2009	2009	16	IL
1/13/2010	2010	17	IL
12/27/2009	2009	18	UT
10/31/2009	2009	19	NV
12/11/2009	2009	20	RI



Number	Region	Postal Code	Address
1	Yishun	763336	336C Yishun Street 31
2	Queensto	141168	168 Stirling Road
3	Woodlanc	730679	679 Woodlands Avenue 6
4	Tampines	520312	312 Tampines Street 33
5	Bishan	570128	128 Bishan Street 12
6	Sengkang	530972	972 Hougang Street 91
7	Bukit Panj	672635	635B Senja Road
8	Newton	298130	376 Thomson Road
9	Newton	307740	1 Surrey Road
10	Paya Leba	381121	121 Paya Lebar Way
11	Tampines	520839	839 Tampines Street 83
12	Yio Chu Ka	807012	3 Seletar Road
13	Boon Lay	640812	812 Jurong West Street 81
14	Yishun	760784	784 Yishun Avenue 2
15	Queensto	141086	86 Dawson Road
16	Boon Lay	643197	197C Boon Lay Drive
17	Boon Lay	640186	186 Boon Lay Avenue
18	Farrer Par	190468	468 North Bridge Road
19	Hougang	530231	231 Hougang Street 21
20	Pasir Ris	512528	528B Pasir Ris Street 51

# How it works?

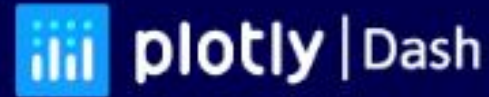


- ◀ 1. Application would gather all the required files
  - ◀ Singapore Address Dataset and/or Singapore Region Dataset - Web Scrapped Singapore Addressed from SRX property website.
  - ◀ Foreign Dataset
- ◀ 2. Read in all the dataset using Pandas
- ◀ 3. Assign an index to the unique values of foreign dataset (e.g. State) → [TX = 0, MO = 1, AK = 2]
- ◀ 4. Assign an index to the unique values of Singapore Dataset (Region) → [Yio Chu Kang = 0, Hougang = 1, Sembawang = 2]
- ◀ 5. Match the index together and update the dataset. [TX becomes Yio Chu Kang, MO becomes Hougang, AK becomes Sembawang]
- ◀ 6. IF user selects that they would like address to be mapped,
  - ◀ 6a. The program loops through every row in the updated Singapore Context dataset.
  - ◀ 6b. Reads in the region of current row. [e.g. Ang Mo Kio]
  - ◀ 6c. Filters the Singapore Address Dataset to only "Ang Mo Kio"
  - ◀ 6d. Randomly select one of the address of region "Ang Mo Kio" and assign it.



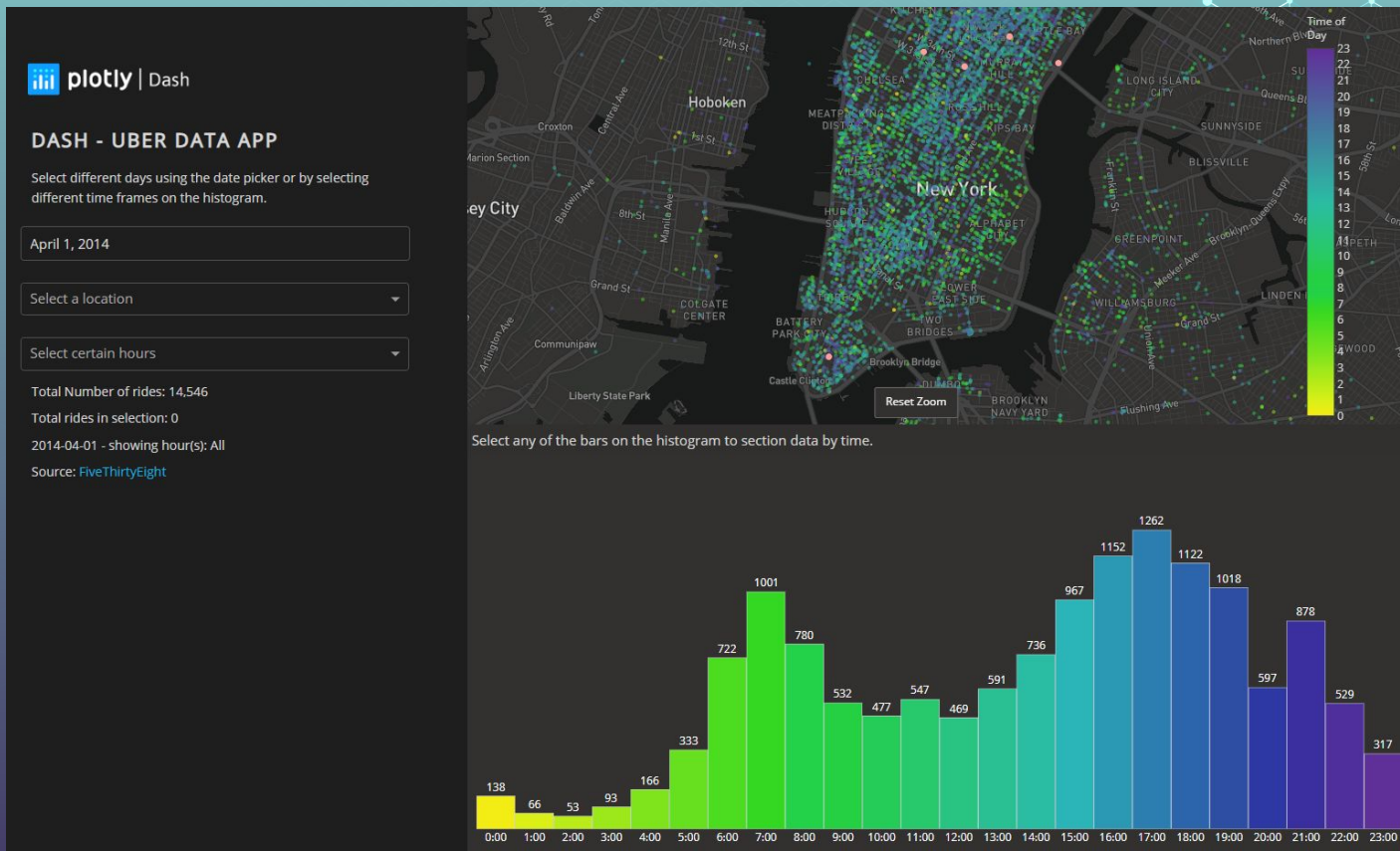
# Research on Plotly Dash

# What is Plotly Dash



- ◀ An open source Python framework for building responsive analytical web applications which do not require any JavaScript/HTML/CSS coding.
- ◀ Plotly Dash is built on top of :
  - ◀ **Flask** - A web framework which allows building of web applications without HTML/CSS
  - ◀ **Plotly.js** - A high-level, declarative charting library with over 40 chart types.
  - ◀ **React.js** - A JavaScript library for building user interfaces

# Plotly Dashboard Example





# My attempt on Plotly Dash



# Thoughts of Plotly Dash

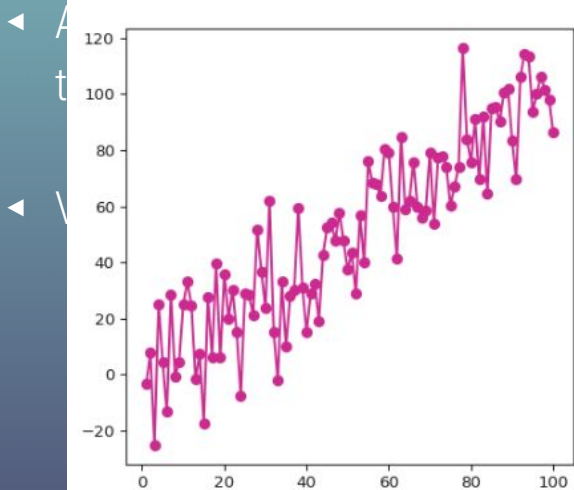


- ◀ Nice looking
- ◀ Very Flexible
- ◀ Not many resources available explaining how to use Plotly.
- ◀ Time consuming compared to visualization applications (Tableau/SAS etc)
- ◀ Datasets requires a lot of data preparation before being able to be used for charting. ( Very Tedious )



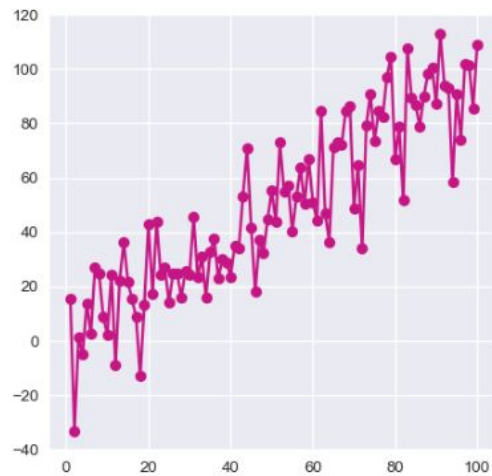
# Other Open Sourced Charting Frameworks?

## Matplotlib



#106 Matplotlib style

## Seaborn



#106 .. with Seaborn style

# Why use Plotly?



- ◀ Plotly has a wide variety of chart types which looks appealing.
- ◀ The charts are not rendered as an Image ( Making it interactive )
- ◀ The charts can be published onto a web server making it visible to anyone with the link.
  - ◀ Includes viewing on mobile devices
- ◀ **FREE**, does not require any licensing.

**THANK YOU**