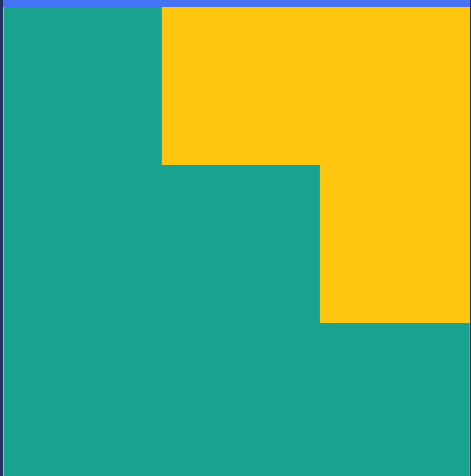
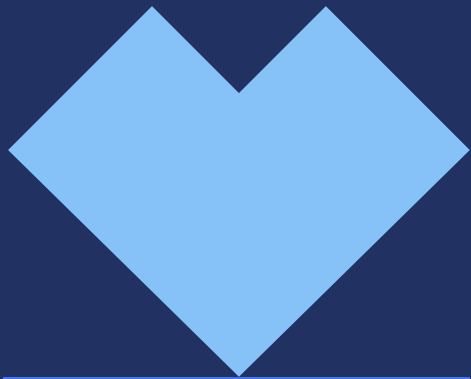


# Statistics, Math Models and Algorithms

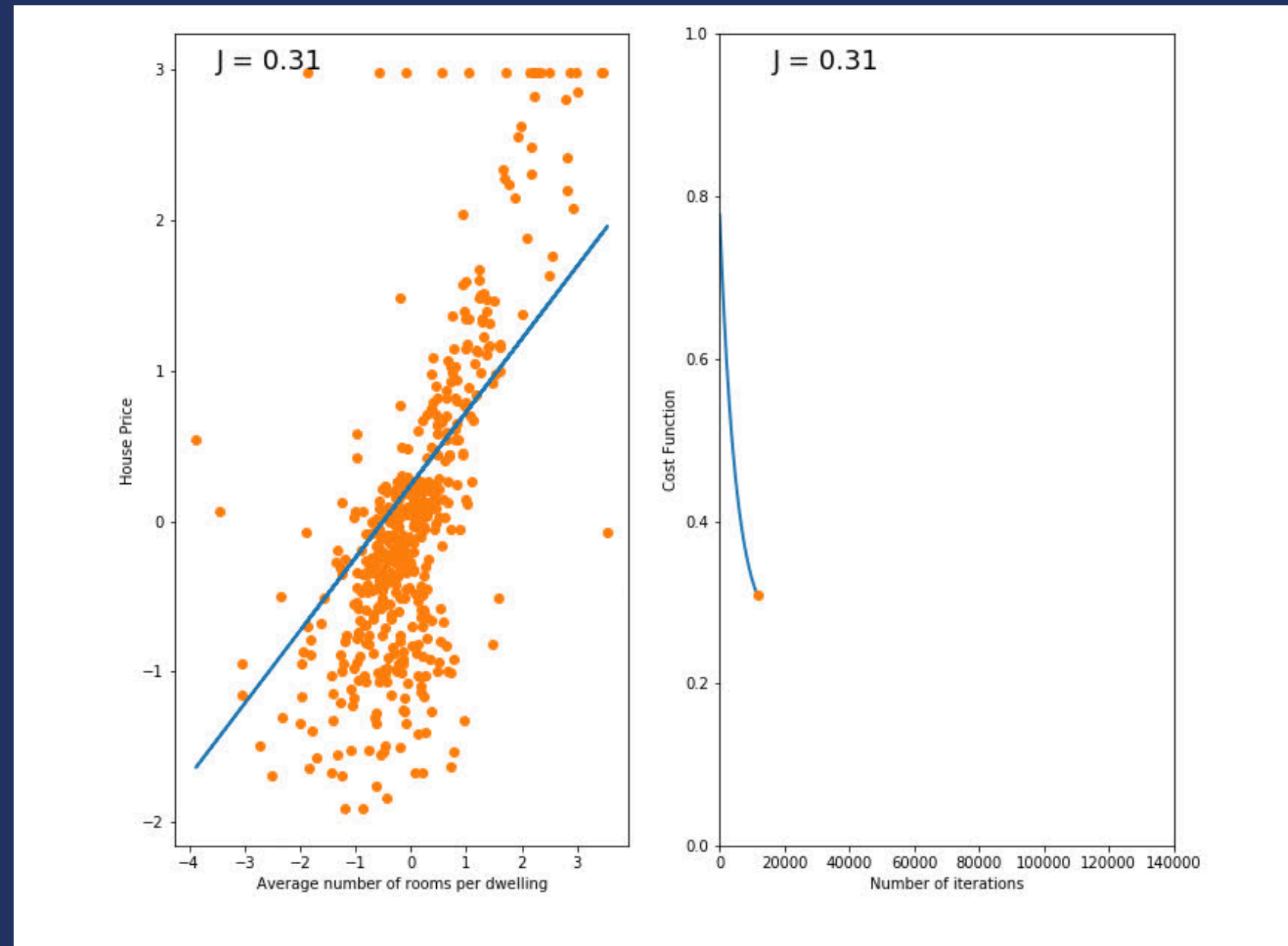
Fahad Faruqi, Ali Mohamed, Md Omit,  
Evan Perez, Jennifer Saeteros, Tak Kit Yeung

4219 Industries LLC

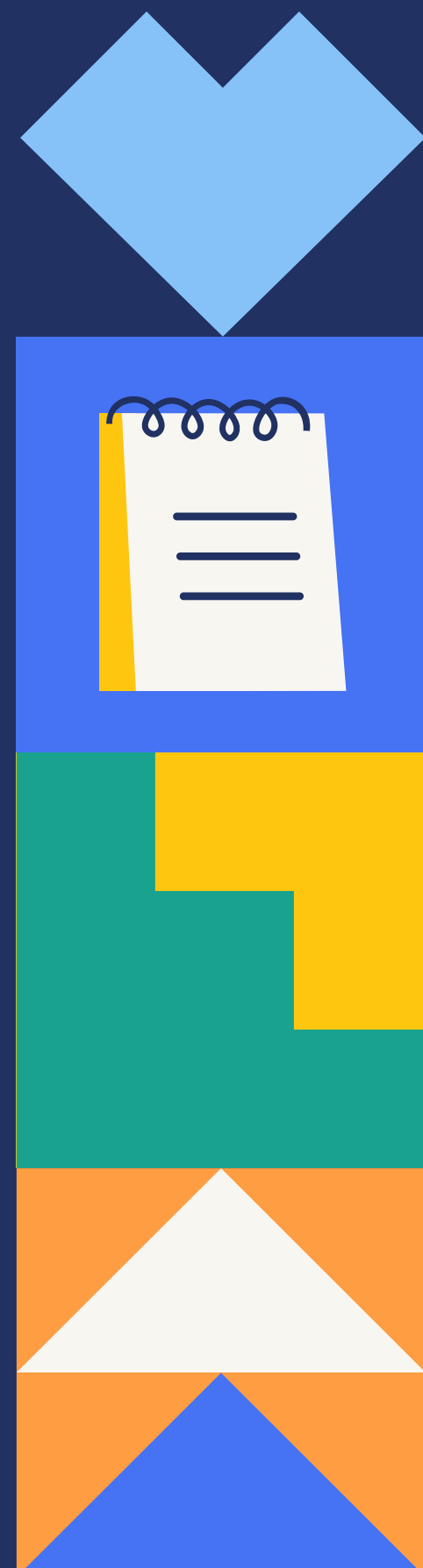




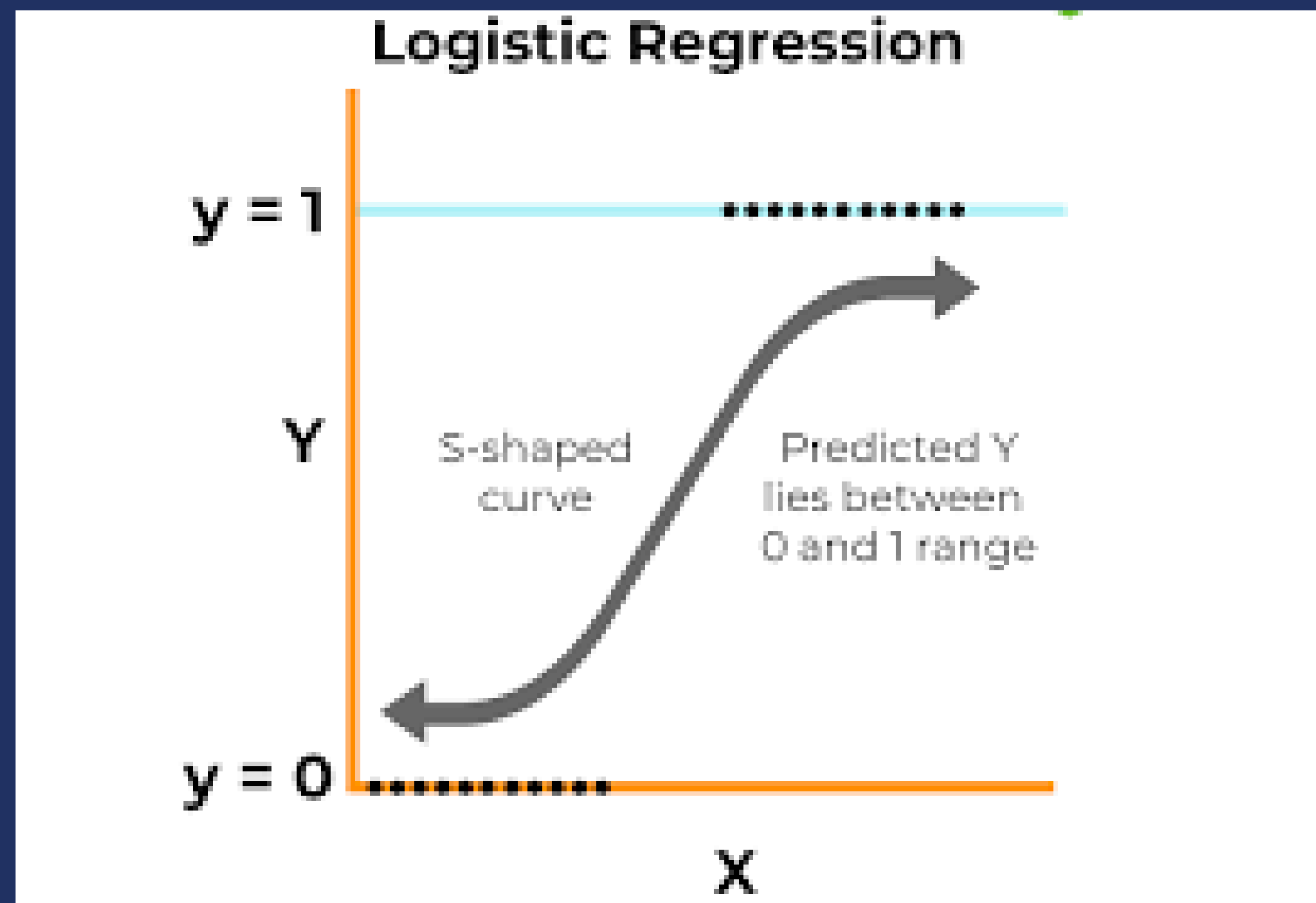
# Linear Regression



Finding a linear relationship  
between variables.

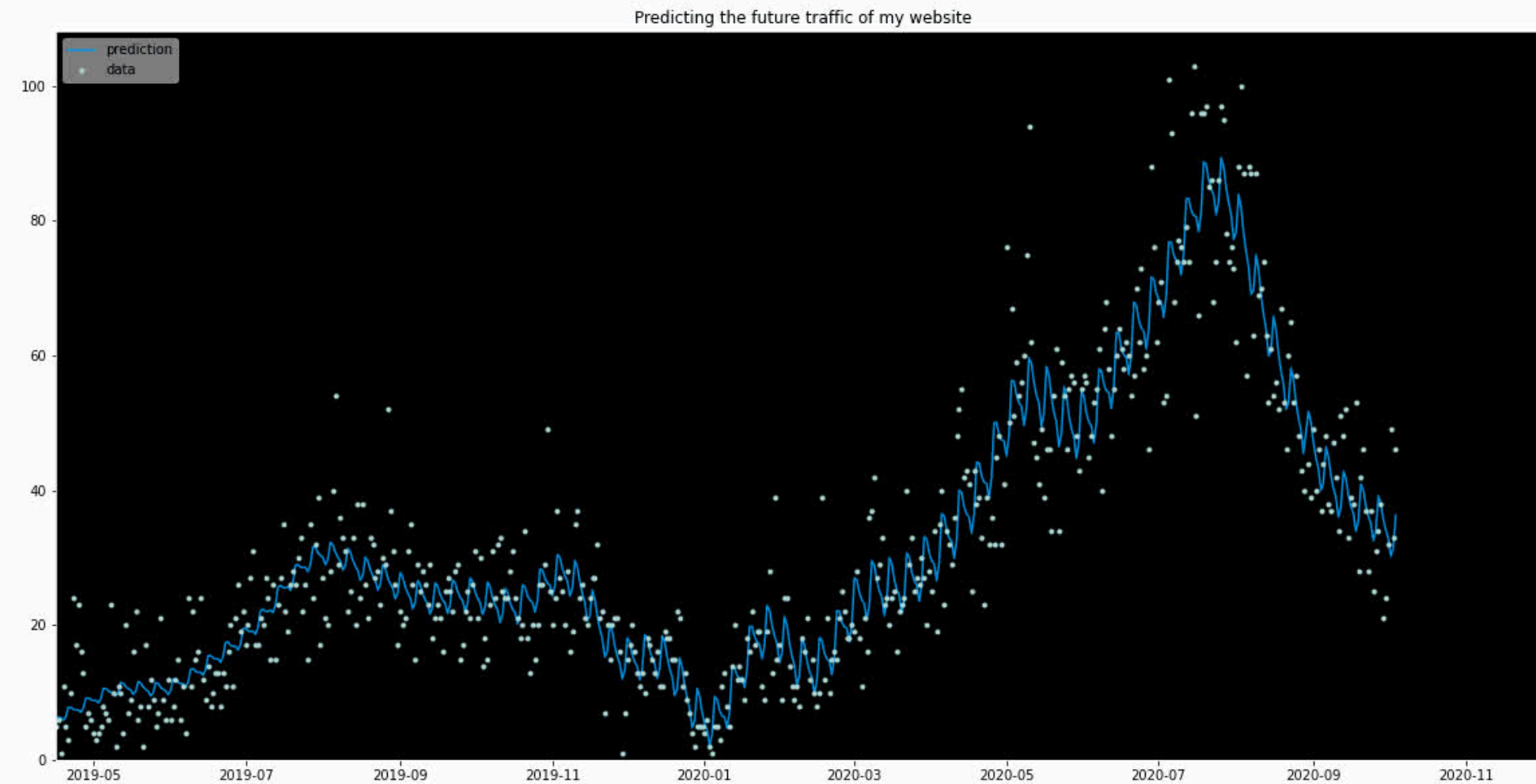
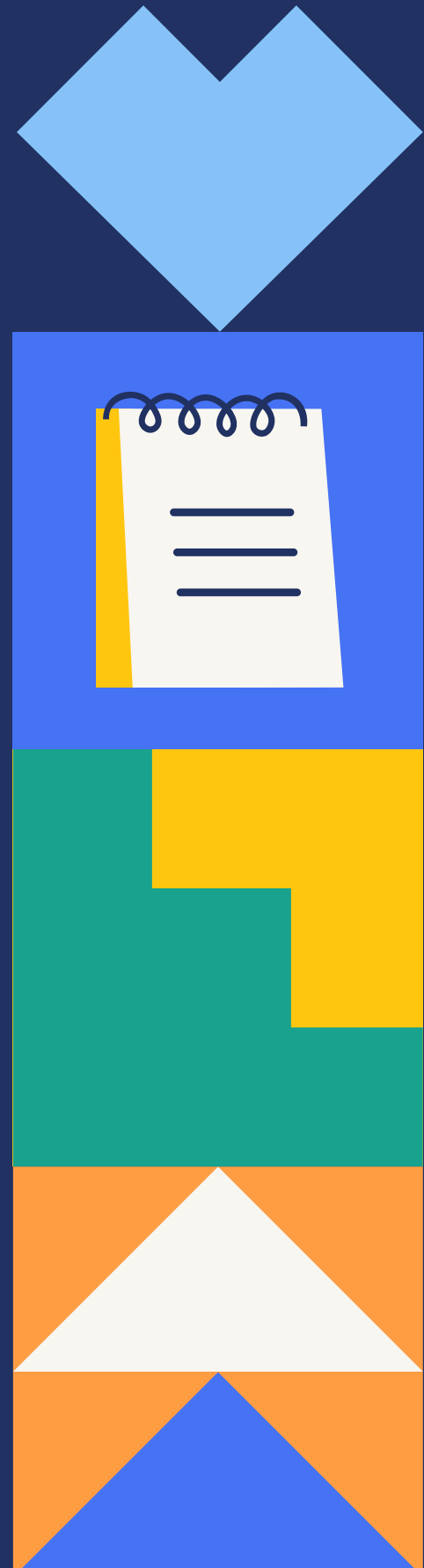


# Logistic Regression



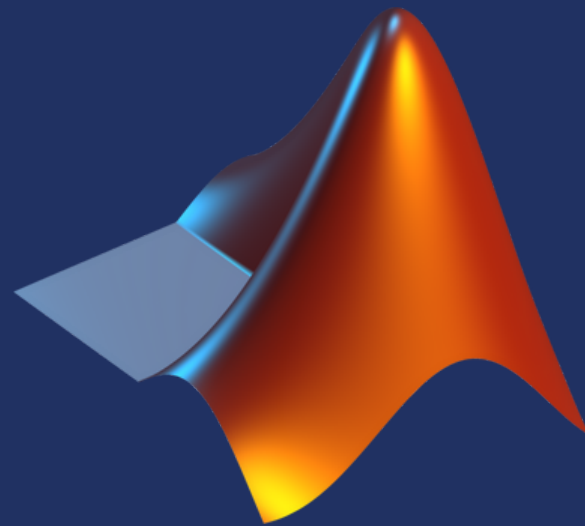
Determines probability that an event will occur.

# Time Series



Find patterns and forecast future values  
based on historical data.

# Common Statistical Softwares



Matlab



PowerBI



JMP

## Median

Set A: 8, 3, 4, 9, 6

3, 4, 6, 8, 9  
↓

Median

Set B: 11, 17, 3, 14, 19, 7

3, 7, 11, 14, 17, 19

$$\text{Median} = \frac{11 + 14}{2} = 12.5$$



## Mean

Set A: 8, 3, 4, 9, 6

6



Mean

Set B: 11, 17, 3, 14, 19, 7

$$\text{Mean} = \frac{11 + 17 + 3 + 14 + 19 + 7}{6} = 11.8$$



# Common Statistical Metric

$R^2$

correlation  
between two  
datasets  
high  $R^2$  == high  
correlation

RMSE

differences  
between true or  
predicted values

Recall,  
Precision,  
**Accuracy**

true positive rate,  
how closely the  
predictions group  
together,  
percentage  
correct prediction

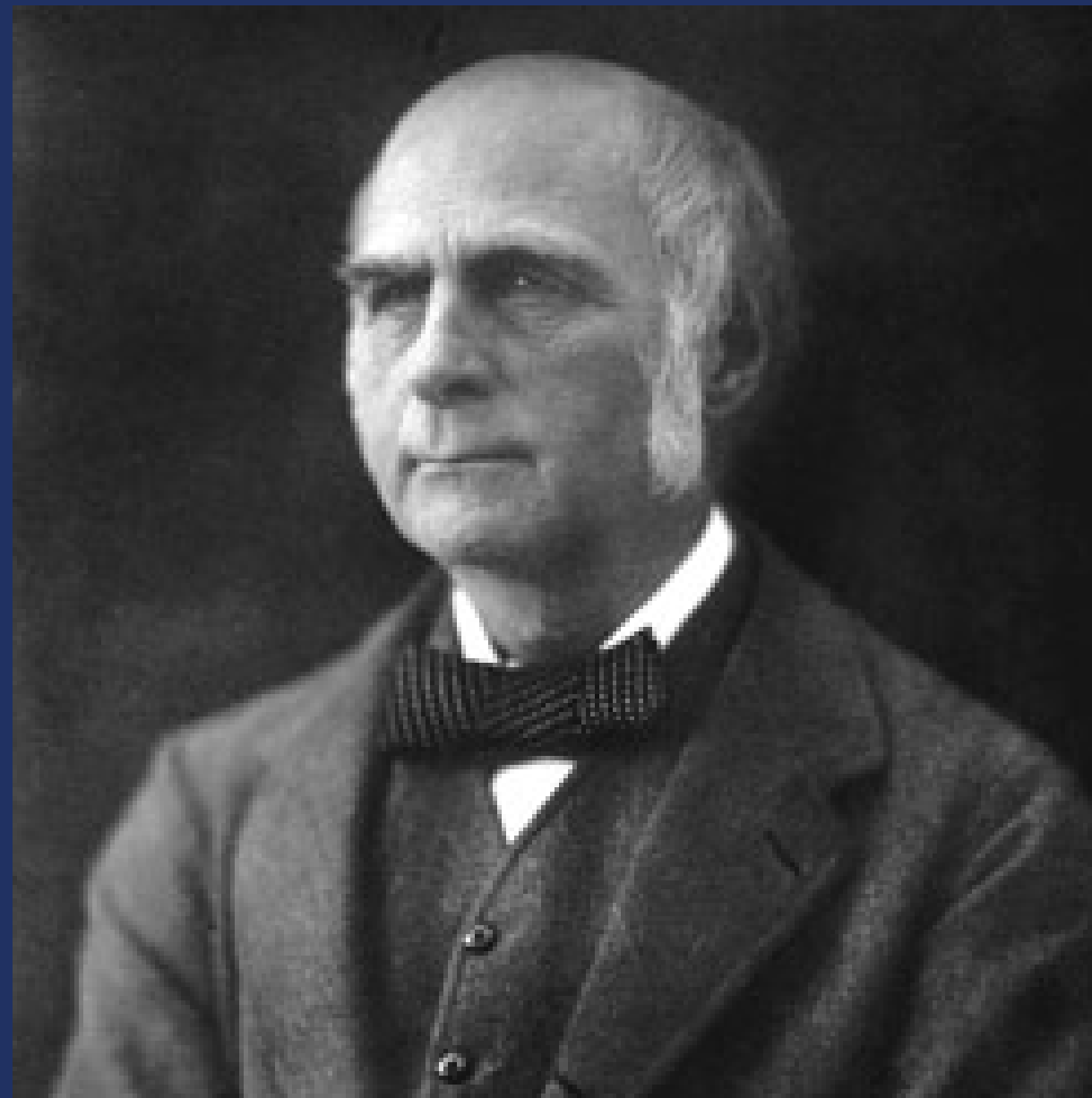


# History

## Linear Regression Model

Introduced in 1894.

Derived by Sir Francis Galton after conceptualizing regression towards the mean.



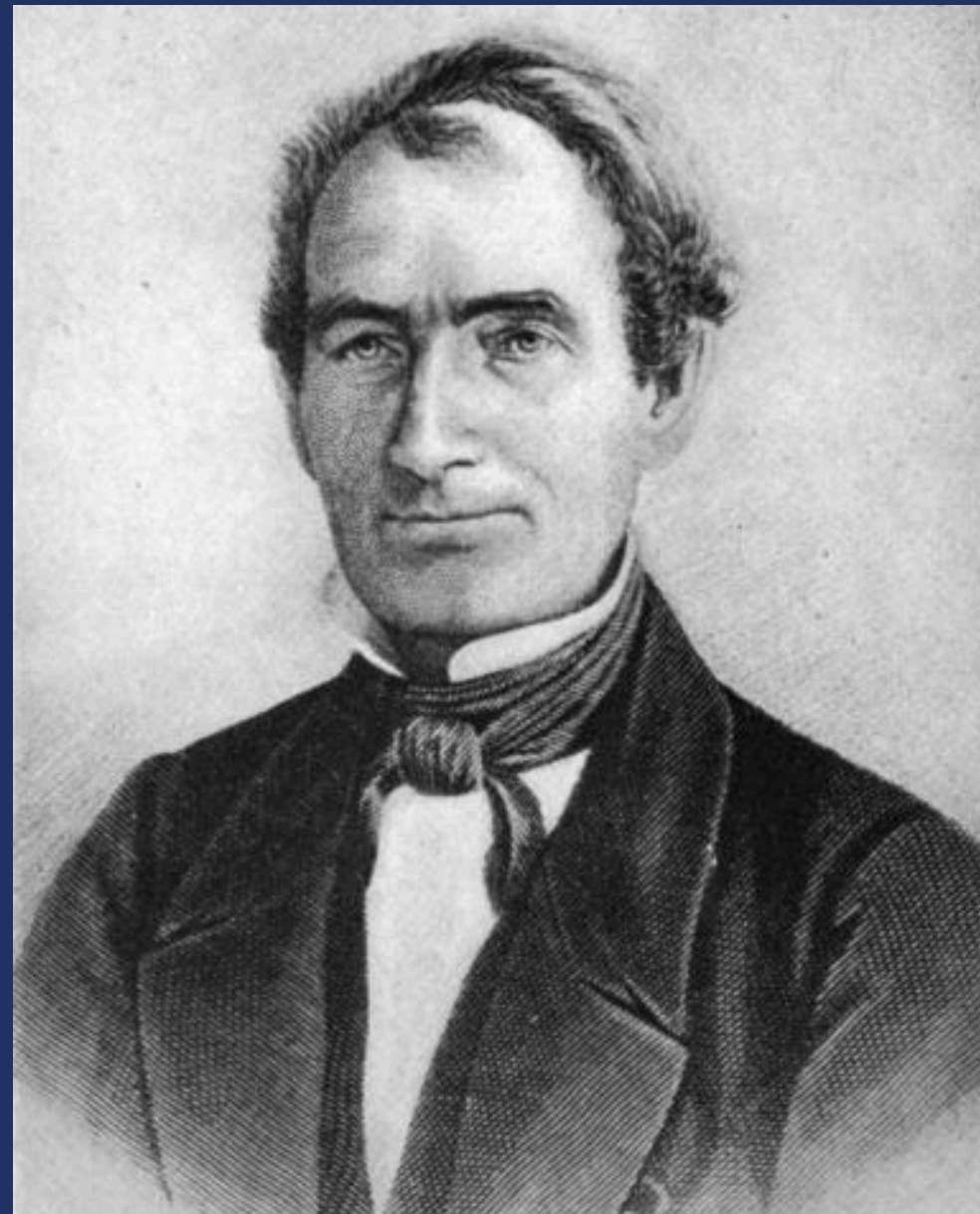
Used to determine relationships between variables and predict the value of one variable based on independent variables.

# History

## Logistic Regression Model

Found by Pierre-Francois Verhulst in 1838.

Rediscovered in 1920 as a model for population growth.



Predicts the probability of an event occurring, classifies data into different categories.

# History

## Time series Model

Data dates as far back as 800 BC China.

Analysis was created by Udny Yule in 1927.



Understand past performance and predict future outcomes in a relevant and actionable way.

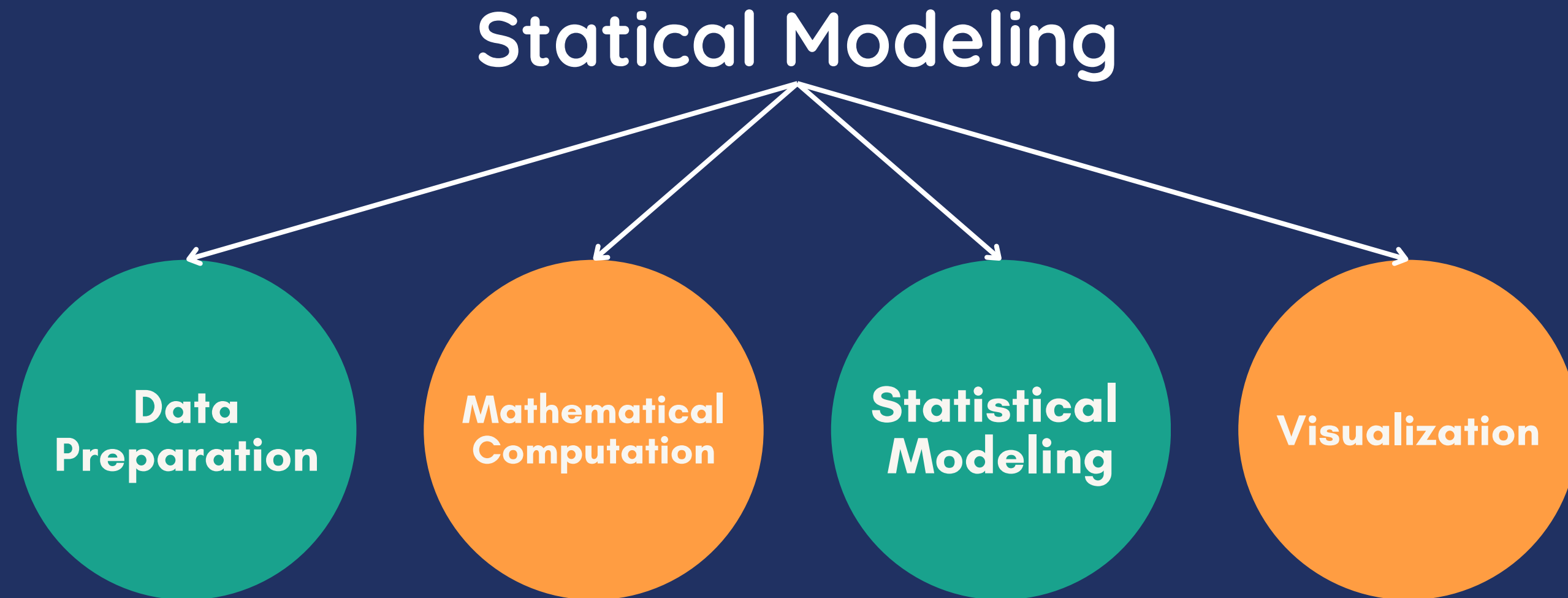
## Usecases

# Popularity and Community Support



Time Series	Logistic Regression	Linear Regression
Forecasting Capability	Binary Classification	Statistical Foundation
Widely Applicable	Widely Applicable	Widely Applicable
Temporal Dependency		Simplicity

# Popularity and Community Support



## Preprocessing



# Data Cleaning and Preparation

Pandas

```
#Check for missing values
```

```
print(df.isnull().sum())
```

```
#Drop rows with missing values a
```

```
df_cleaned = df.dropna()
```

```
#Fill missing values with mean fo
```

```
df_filled = df.fillna(df.mean())
```

# Uses of NumPy



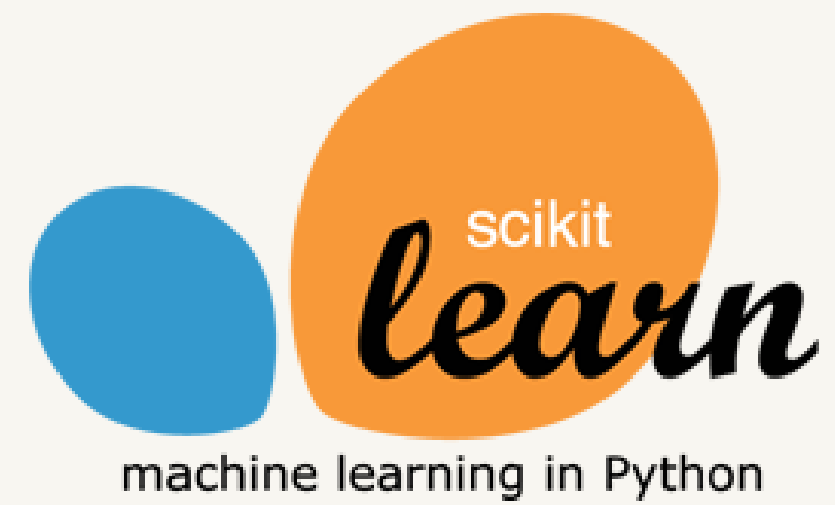
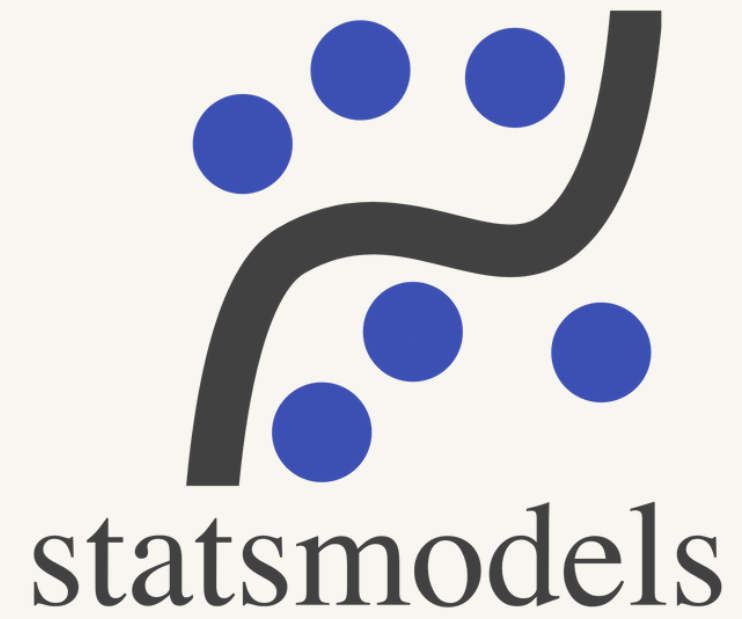
Doing the math



# Statical Computation

NumPy  
SciPy

## Statistical Modeling



Statsmodels  
Scikit-learn  
R







Matplotlib  
Seaborn  
Plotly  
ggplot2

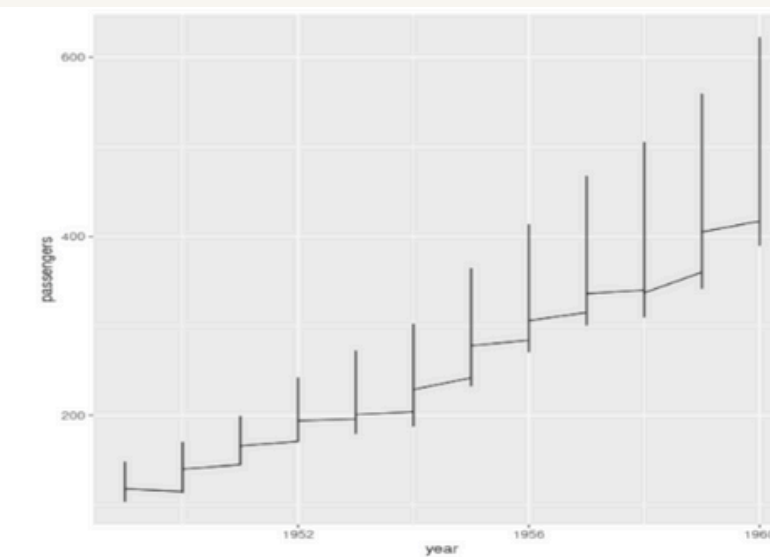


## Visualization

```
library(ggplot2)

flights <- read.csv("flights.csv")

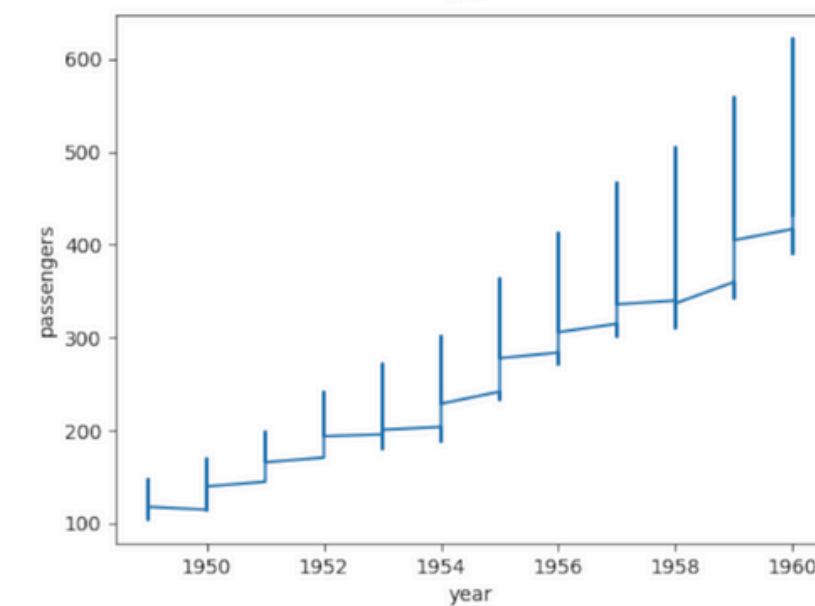
ggplot() + geom_line(aes(y = passengers, x = year),
                     data = flights)
```










```
import matplotlib.pyplot as plt
import pandas as pd

flights = pd.read_csv('flights.csv')

plt.plot(flights['month'], flights['passengers'])
plt.xlabel('year')
plt.ylabel('passengers')
plt.show()
```



# Tabular Comparison

	Predictive Analysis	Probability Analysis	Requires lots of data
Linear Regression			
Logistic Regression			
Time Series			



## Linear Regression



### Pros

best model for linear  
relationships  
fast & simple  
low training times



### Cons

susceptible to overfitting  
cannot determine complex  
relationships  
needs high sample size





# Logistic Regression



## Pros

- can extend to multiple classes (labels)
- fast & simple
- good accuracy for linear datasets**



## Cons

- assumes linear boundaries**
- easily outperformed in determining complex relationships
- assumes few outliers**





## Time Series



### Pros

- identifies historical trends
- identifies outliers in data
- shows effect over time



### Cons

- needs comprehensive data
- data must have linear relationship
- requires human interpretation



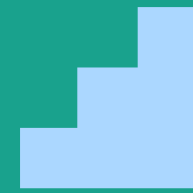
# Industry and Academic Relevance

## Linear Regression Model



### Academic

Biological,  
behavioral,  
environmental,  
social sciences



### Industry

Business,  
insurance



### Use Case

Continuous  
variable based  
on another  
variable

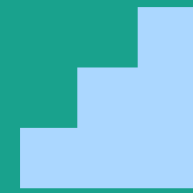
# Industry and Academic Relevance

## Logistic Regression Model



**Academic**

Machine  
learning



**Industry**

Marketing,  
healthcare



**Use Case**

Categorical  
dependent  
variable



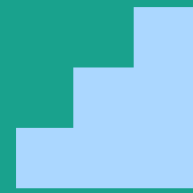
# Industry and Academic Relevance

## Time Series Model



### Academic

Economics,  
education



### Industry

Finance,  
retail



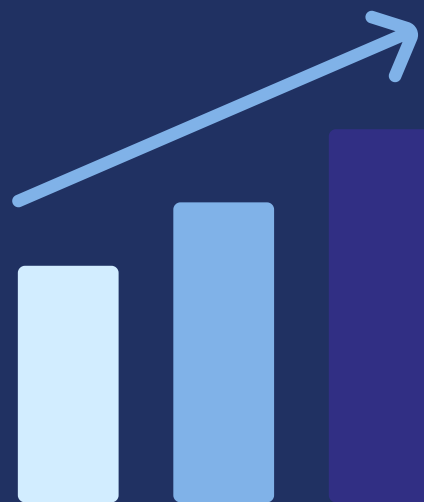
### Use Case

Data that are  
constantly  
fluctuating  
over time





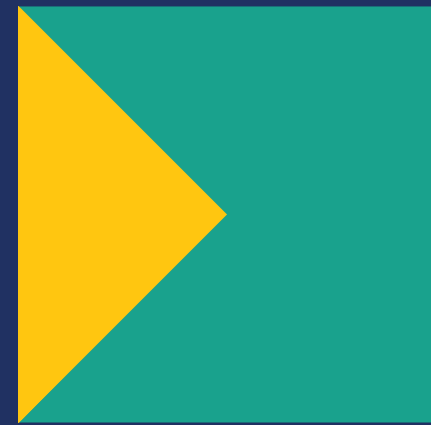
# Recommended Analysis Technique: Time Series



- **Forecasting**
  - predict future values
- **Classification**
  - identify and assign categories
- **Descriptive**
  - find trends, patterns, cycles
- **Curve fitting**
  - study relationship between variables
- Obtaining lots of data can be easier due to periodic data collection

# Conclusion

- Different problems require different techniques, especially if the problem is fine-grained
- Statistical analysis methods should be chosen based on:
  - **Data availability**
  - **Data quality**
  - **Focus of the problem at hand**

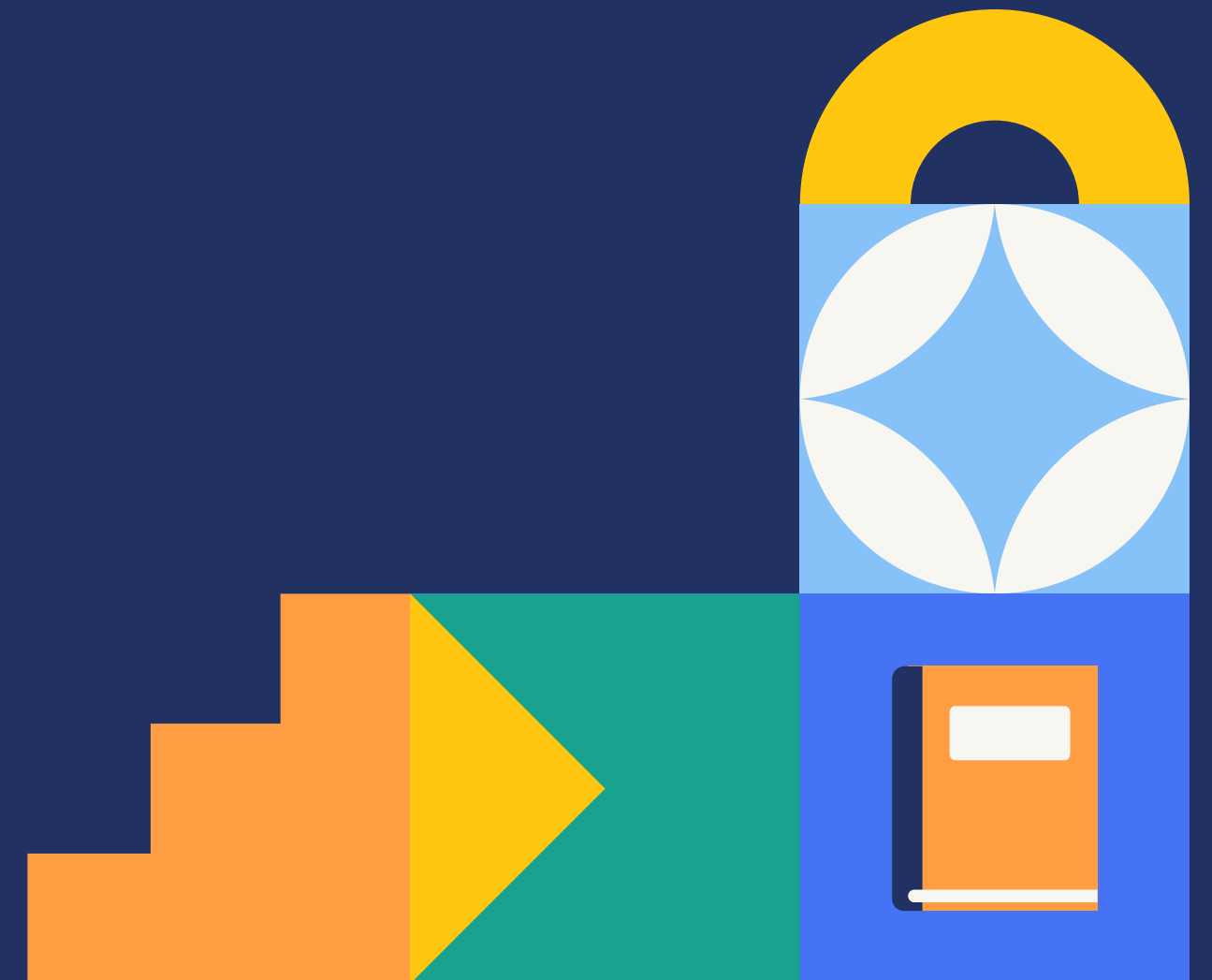


Thank You  
For  
Listening!



# References

- Mathspace. "Statistics."
- Institute of Mathematical Statistics (IMS)
- R Consortium: ocuses on supporting the R programming language and community, but it's also highly relevant for statisticians who use R for statistical models
- Cross Validated (StackExchange): A popular Q&A platform for statisticians, data scientists, and researchers



# Q&A

