



# Introduction to Statistics

FinTech



# Outline

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Summary statistics, such as mean, median, mode, variance, and standard deviation



Plotting, characterizing, and quantifying a normally distributed dataset.



Qualitatively and quantitatively identifying potential outliers in a dataset.



Differentiating between a sample and a population in regard to a dataset.



Defining and quantifying correlation between two factors.



Calculating and plotting a linear regression.

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**What are the three measures  
of central tendency?**

# Measure of Central Tendency = Center of a Dataset

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The three most common measures are **mean**, **median**, and **mode**.

## Mean

Mean is the sum of all values divided by the number of elements in a dataset.

## Median

Median is the middle value in a sorted dataset.

## Mode

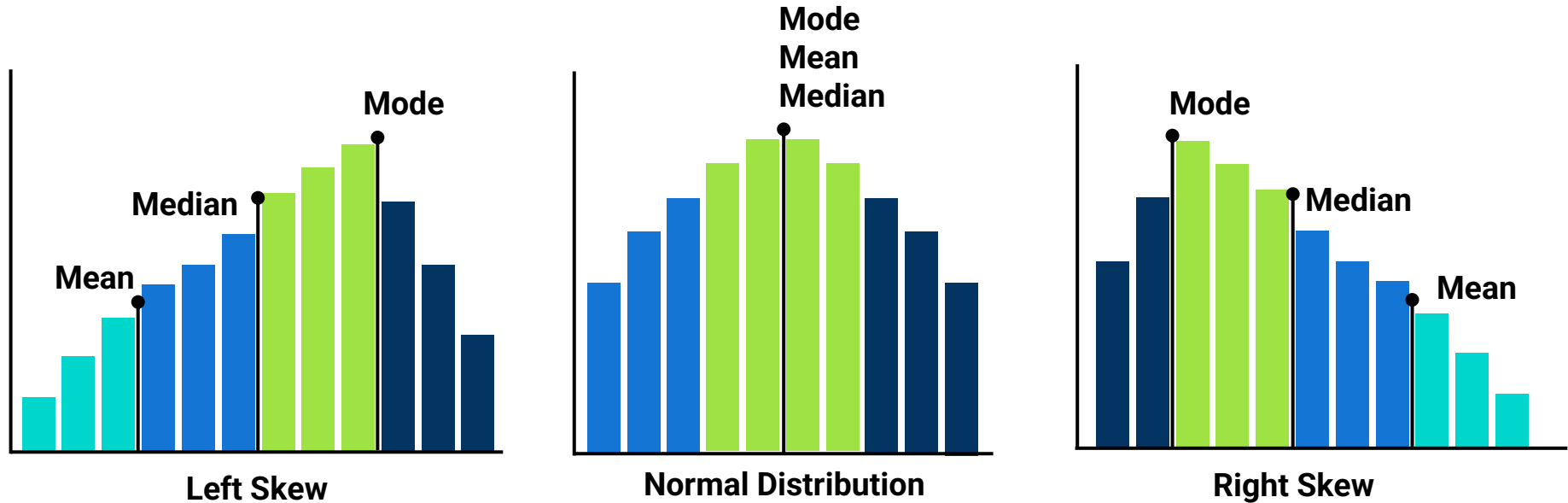
Mode is the most frequently occurring value(s) in a dataset.

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# The mean, median and mode.

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The mean, median and mode.





The **mean** can be used to determine the average value of a portfolio or stock over time.



**What are the measures of  
central tendency used for?**



Metrics used to describe  
the center of a data set.





**How do you describe  
the variability of a data set?**

# Variability of a Data Set

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Three summary statistics metrics for describing variability:

01

Variance

02

Standard Deviation

03

Z-Score

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**What are variance and  
standard deviation?**

# Variance and Standard Deviation

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Variance and standard deviation describe variability of data.



**Variance** is the measurement of how far each value is from the mean of the dataset.



**Standard deviation** is the square root of variance.

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# Variance

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Used to describe how far values in the data set are from the mean



Describes how much variation exists in the data



Variance considers the distance of each value in the data set from the center of the data

The value of the one observation

The mean value of all observations

Sample variance

$$S^2 = \frac{\sum (x_i - \bar{x})^2}{n - 1}$$

The number of observations

# Standard Deviation

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Describes how spread out the data is from the mean



Calculated from the square root of the variance



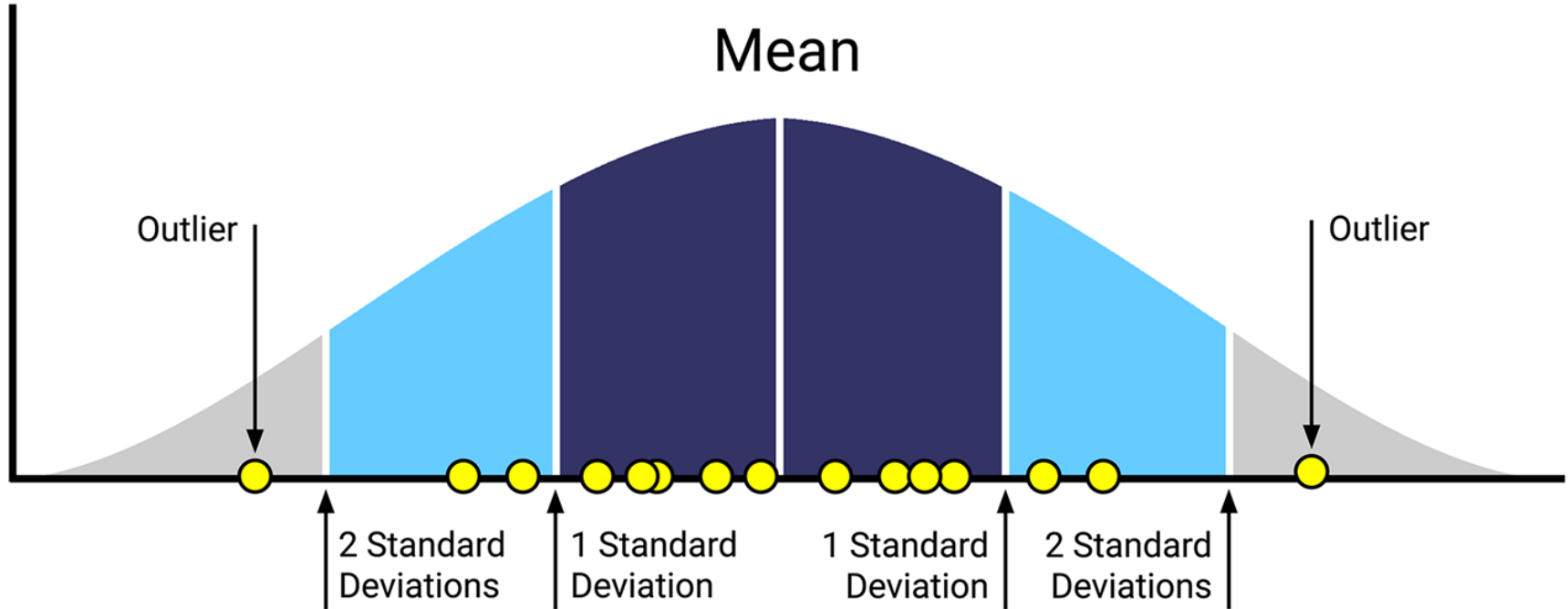
In the same units of measurement as the mean

$$\text{Standard deviation } \sigma = \sqrt{S^2} \text{ The variance}$$

# Standard Deviation

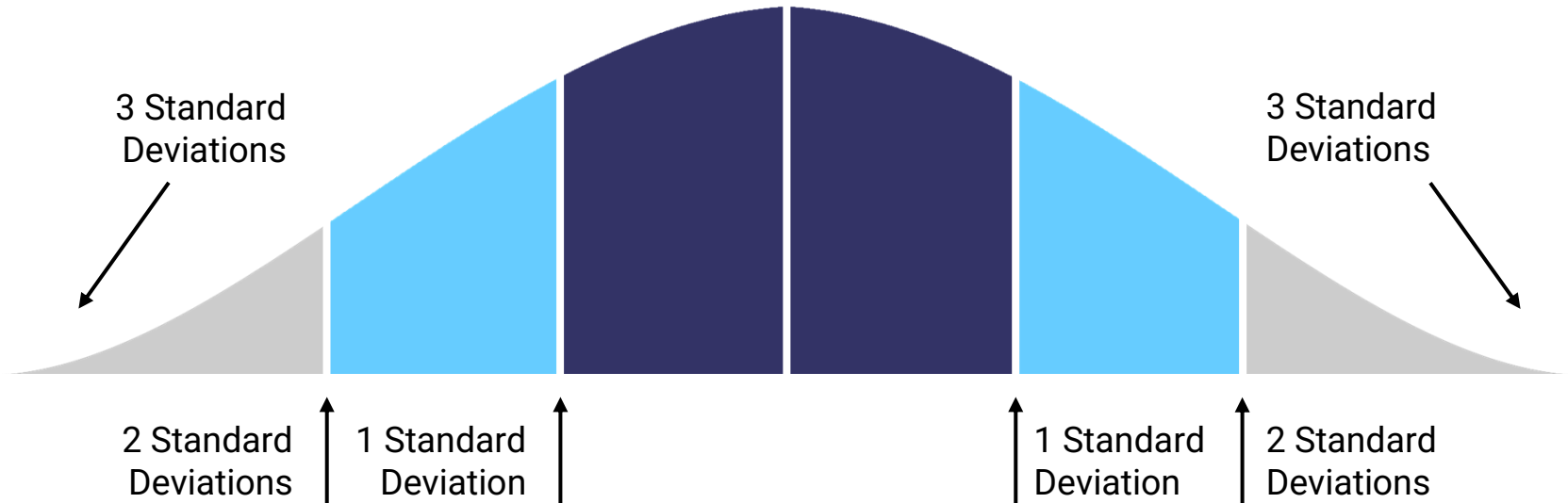
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Square root of the variance; a measure used to quantify the dispersion of a set of observations.



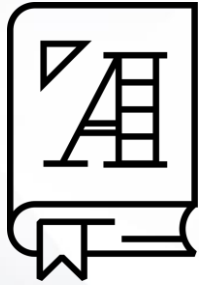
# Standard Deviation

The `std` Pandas function is used to calculate standard deviation for a DataFrame. Standard deviation can be used to determine the risk associated with an investment. Standard deviation is also used to calculate how much returns have been distributed from the average.





# Standard Deviation and Risk



**Standard deviation** identifies precisely how far away a value is from the average.



The greater the standard deviation, the greater the risk (and the potential) for a larger payout.

# Z-Score

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Z-Score describes a single value's distance from the mean of the data set  
The distance is in terms of standard deviations. Can be positive or negative:

**If negative**

the value is less than the mean

**If positive**

the value is greater than the mean.

**The smaller the z-score, the  
closer the value is to the mean**

$$Z = \frac{\text{A single value } X - \text{The mean of the dataset } \mu}{\text{The standard deviation of the dataset } \sigma}$$

# Real-World Data

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Be careful when describing real-world data:



Real world data can contain extreme values



Some summary statistics such as the mean take into account all values of a data set



Extreme values can skew these statistics!

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**But how can we summarize  
real-world data?**

# Quantiles: Used to Describe Segments of a Dataset

Quantiles separate a sorted dataset into equally sized fragments.

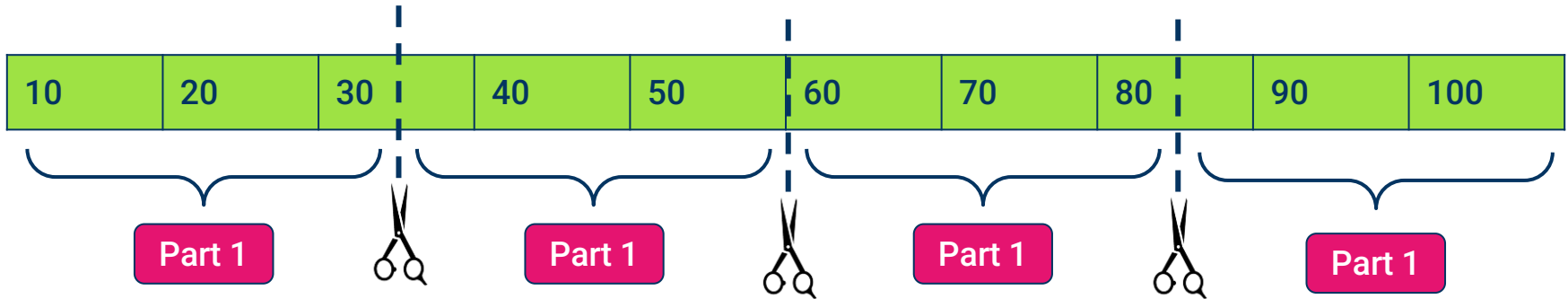
The two most popular types of quantiles are **quartiles** and **percentiles**.

01

**Quartiles** divide the dataset into four equally sized parts.

02

**Percentiles** divide the dataset into 100 equally sized parts.



# Quantiles, Quartiles, and Outliers

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Quantiles, quartiles, and outliers describe a dataset.

## Quantiles

Quantiles divide data into well-defined regions based on a sorted dataset.

## Quartiles

Quartiles are a specific type of quantile where a sorted dataset is split into four equal parts.

<b>Q1</b>	25% of the data
<b>Q2</b>	50% of the data
<b>Q3</b>	75% of the data

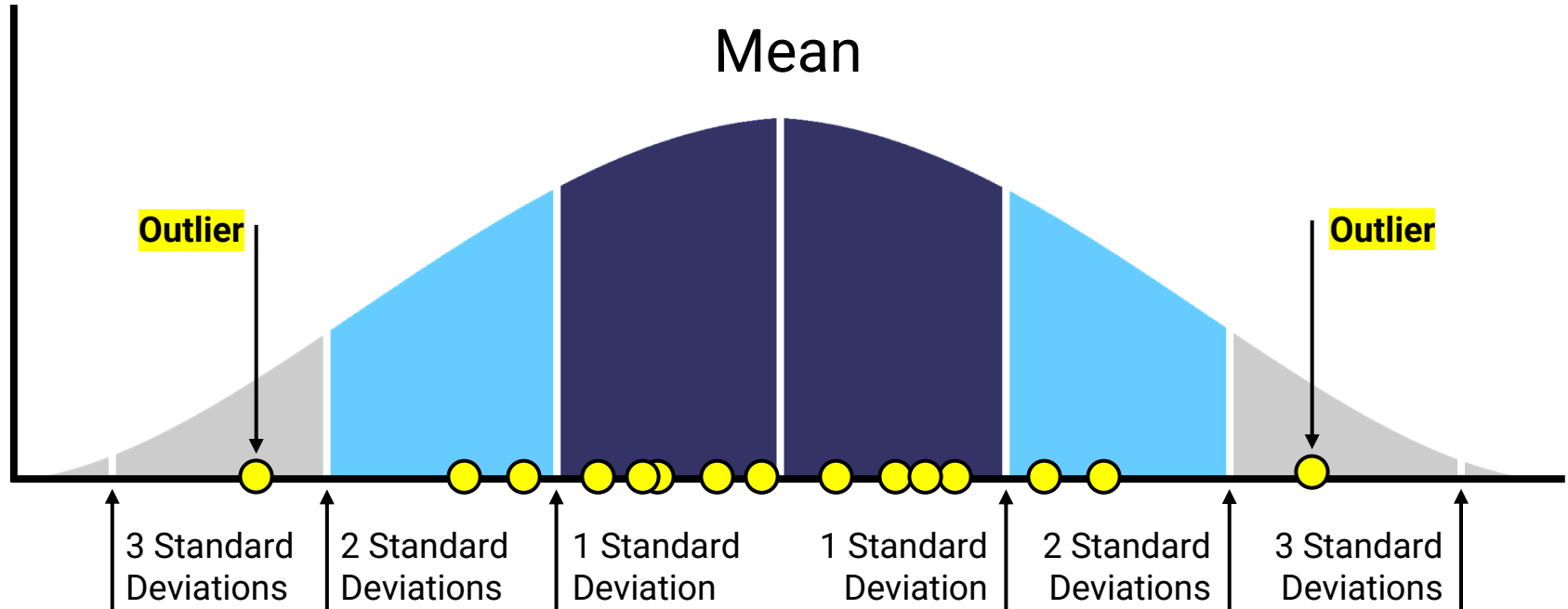
## Quartiles

Outliers are extreme values in a dataset that can skew calculations and results.



# Outliers

Suspicious values are called potential outliers. An outlier is a data point that differs from the rest of a data set. Outliers can inaccurately skew a data set.

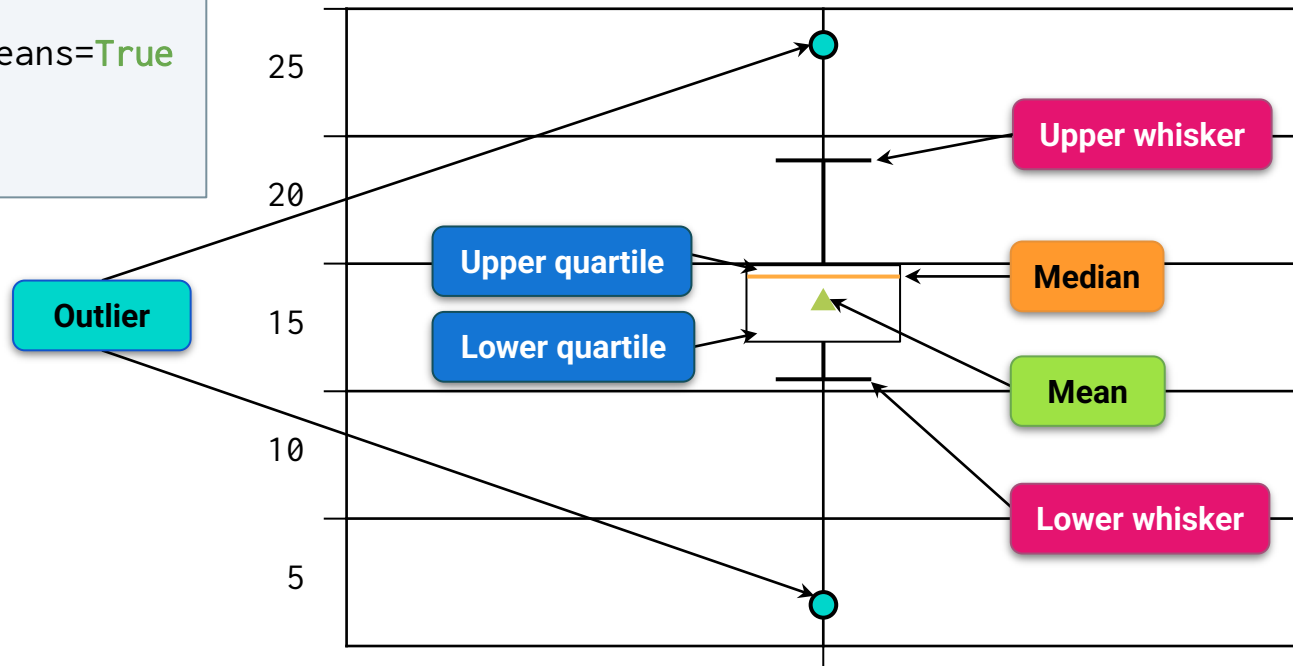


# How to Identify Potential Outliers: Qualitatively

Use box-and-whisker plots to visually identify potential outlier data points.

```
# Create box plot
```

```
plt.boxplot(arr, showmeans=True)  
plt.grid()  
plt.show()
```



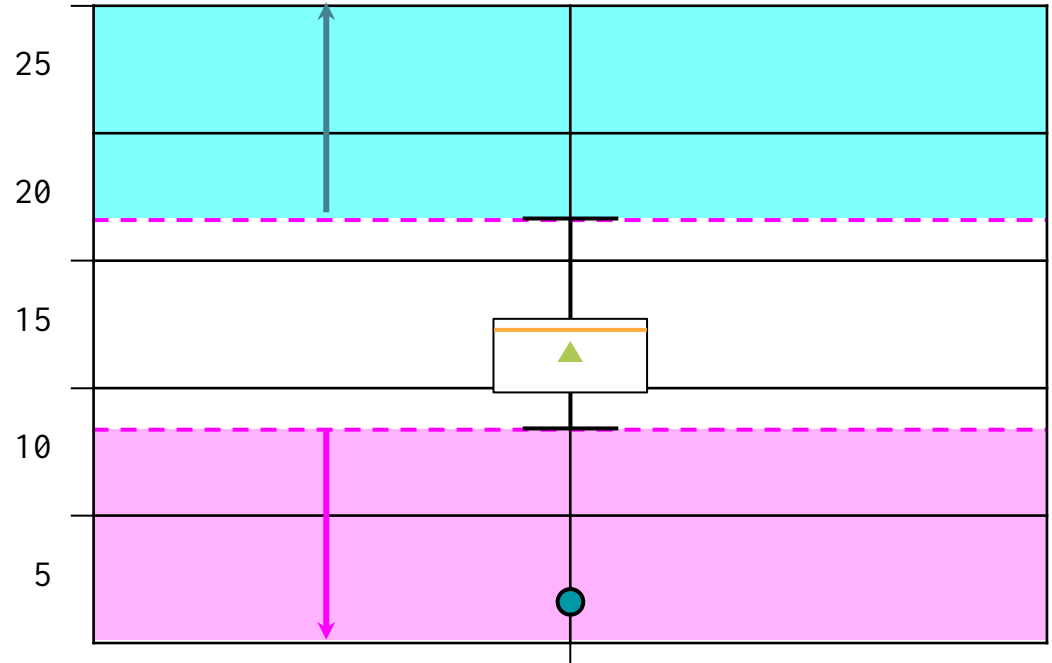
# How to Identify Potential Outliers: Quantitatively

Determine the outlier boundaries in a dataset by using the  **$1.5 \times \text{IQR}$  rule**.

The IQR is the range between the first and the third quartile.

Anything **less than, or below,** Quartile 1 –  $(1.5 \times \text{IQR})$  might be an outlier.

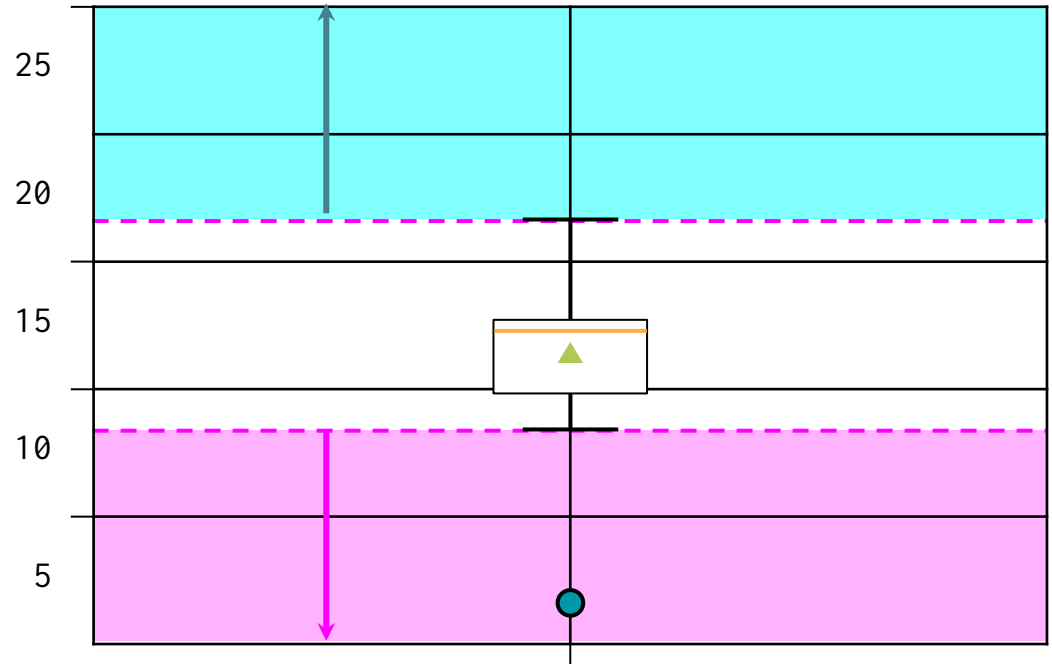
Anything **greater than, or above,** Quartile 3 +  $(1.5 \times \text{IQR})$  might be an outlier.



# How to Identify Potential Outliers in Python: Quantitatively

Use Pandas' `series.quantile` function to calculate the quantile.

Calculate the outlier boundaries.





**When new data comes along,  
you must plot it!**

# Why Plot Data?

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01

To determine if the data is normally distributed.

02

To determine if the data is multimodal.

03

To characterize clusters in the dataset.

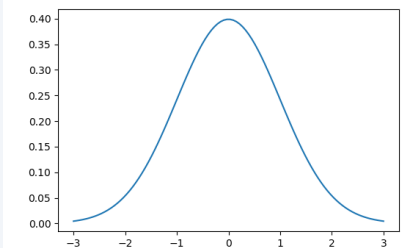
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# What Is Normally Distributed Data?

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01

The distribution of data follows a bell-curve shape.



02

We can quantitatively test if a dataset is normal using SciPy.

```
stats.normaltest()
```

03

Some statistical tests assume normally distributed data.

A stylized teal lightning bolt graphic with a 3D effect, pointing towards the text.

# Sample and Population Datasets

Let's think about  
the following scenario...



# Predicting the City Election

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Weeks before Election Day, a local newspaper wants to predict the winner of the mayoral election. The newspaper will poll voters for their intended candidate. Consider the following:



It would be prohibitively expensive to poll all voters.



It is logistically impossible to know who will actually go out to vote on Election Day.



Therefore, the newspaper must predict the outcome of the election using data from a subset of the population.



This calls for the use of a sample dataset in place of a population dataset.



# Population Dataset vs. Sample Dataset

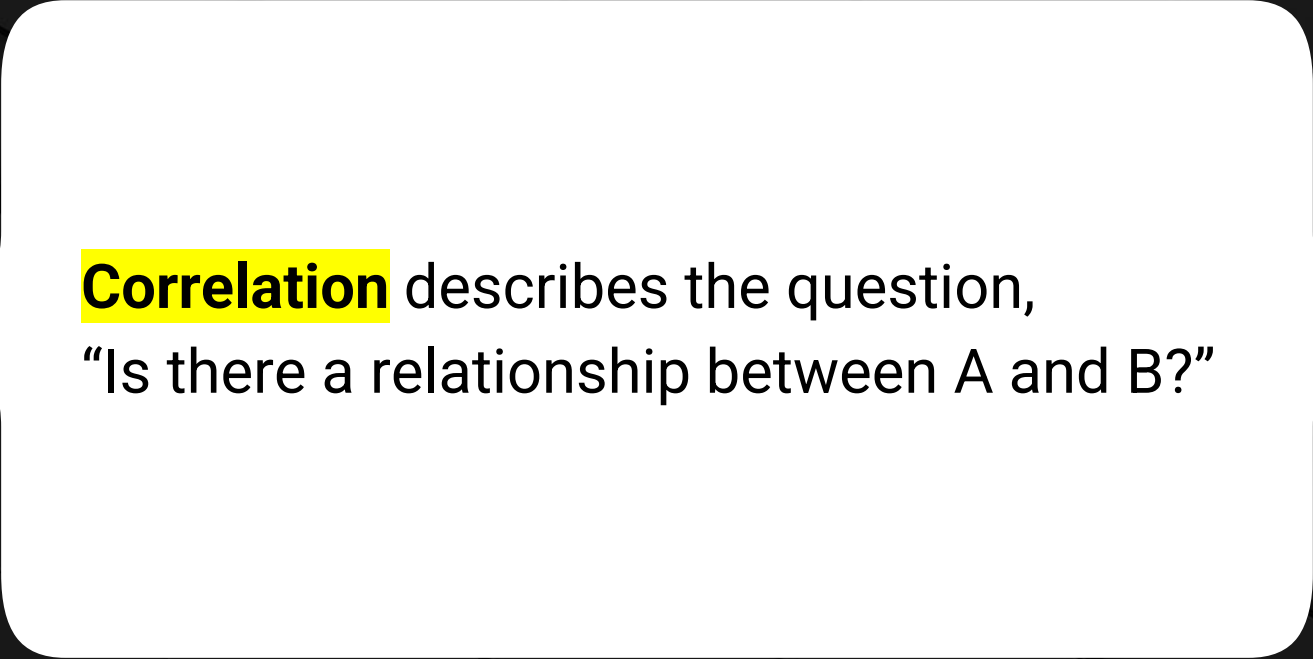
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## Population Dataset

- Dataset containing all possible elements of an experiment or study.
- In statistics, “population” does not mean “people.”
- Any complete set of data is a population dataset.

## Sample Dataset

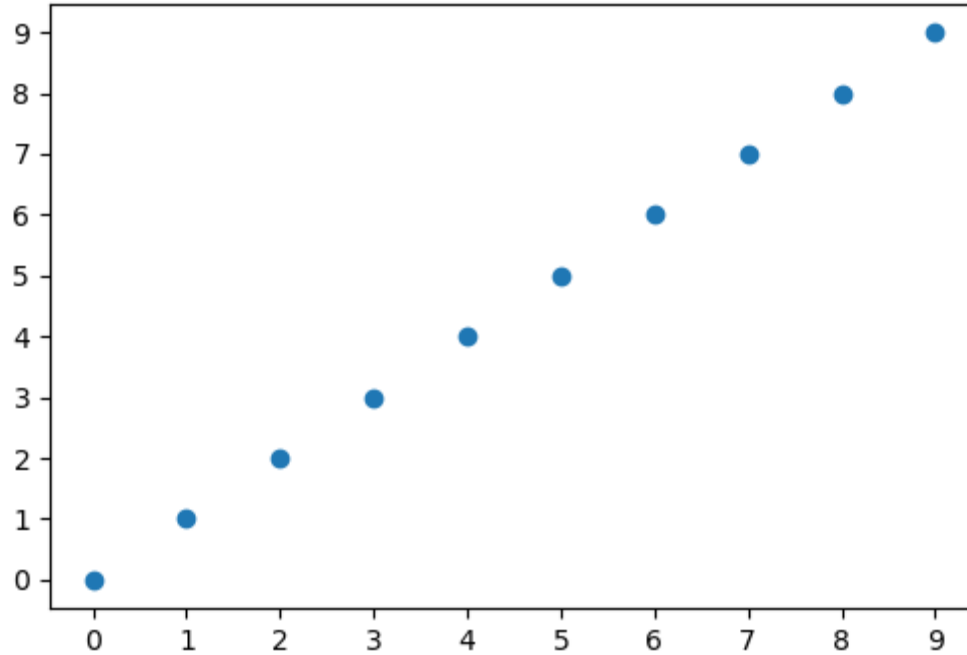
- A subset of population data.
- A sample dataset can be selected randomly from the population or selected with bias.



**Correlation** describes the question,  
“Is there a relationship between A and B?”

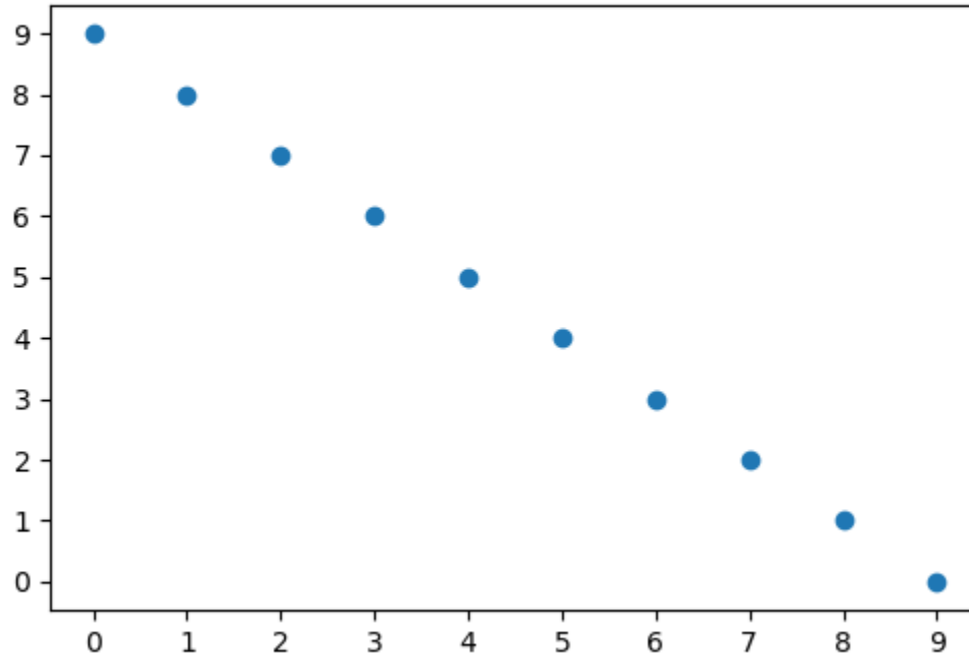
# Positive Correlation

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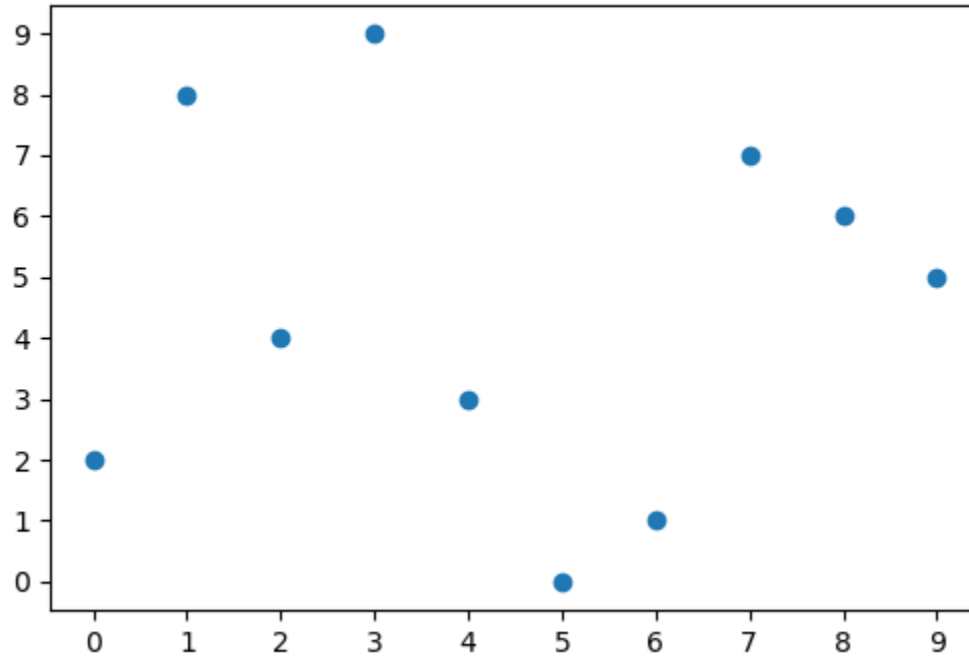
# Negative Correlation

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# No Correlation

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# Scatter plots are a powerful visualization tool!

Visualizes the comparison between two variables:

One variable	is located on the x-axis
Another variable	is plotted on the y-axis

- Each data point represents a pair of measurements
- Measurements on a scatter plot are independent
- Scatter plots can help to identify positive or negative relationships between two variables
- Adding a trend line to a scatterplot can visualize this relationship even easier!





**What is the equation of a line?**



The equation of a line is:

$$y = mx + b$$



Dependent variable



Slope



Independent variable



y-intercept

# The Equation of a Line Determines $y$ Values Given $x$

In this example:



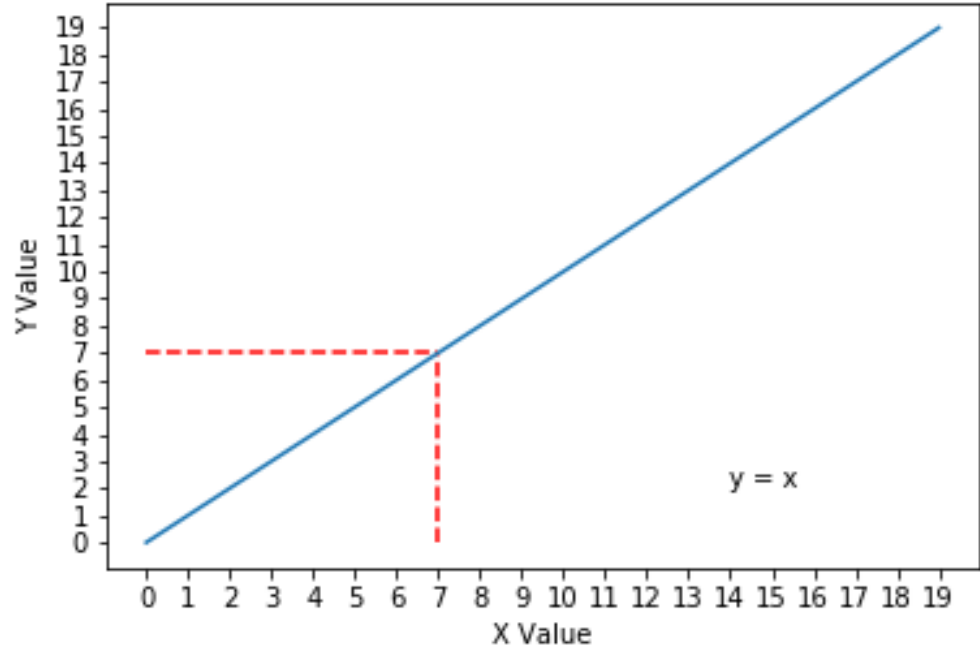
Slope = 1



$y$  intercept = 0



Whatever  $x$  is, the value of  $y$  is the same.



# The Equation of a Line Determines y Values Given x

In this example:



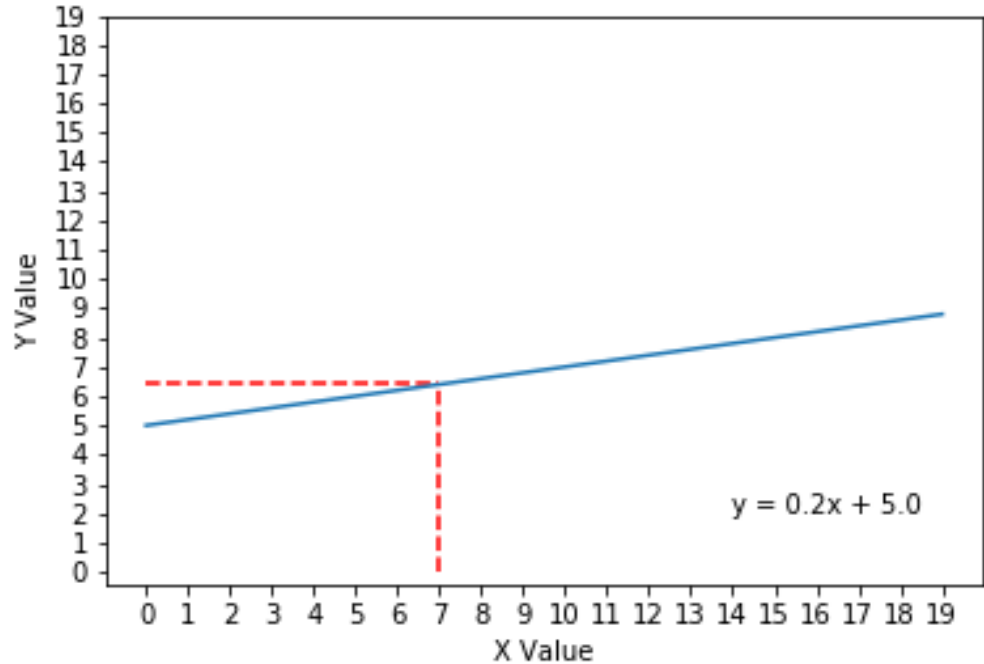
Slope = 0.2



y intercept = +5



If  $x = 7$ , then  $y = 6.4$



*The  
End*