

## "Introduction to pandas"

### • Measures of central Tendency

#### 1) mean (Arithmetic Average)

- The sum of all values divided by the total number of values
- It's the common measure but very sensitive to Outliers

المتوسط: المجموع على عدد.

- It's used when the data is normal distribution

- there is no "outliers"

• معايير جيدة للفهم الشاذة

In Pandas: df['column'].mean()

#### 2) median (middle value)

- The middle point of the data when it is arranged in ascending order.

- It's not affected by outliers

- If you have a few people with massive salaries, the median gives a more "honest" picture of the average person's income.

الوسيط: المعتبر الوسيط بعد ترتيب المعايير.

لا يتأثر بالفهيم الشاذة

In Pandas: df['Column'].median()

#### 3) mode (most Frequent)

- The value that appears most often in the dataset.

- Mainly used for Categorical Data (words / labels)

الأكثر تكراراً

يُفضل استخدامه مع البيانات الاقتباسية

In Pandas: df['Column'].mode()

Note: strategy: skewness, Distribution:

before you choose which one to use for Imputation (filling missing values)

You look at the distribution of your data:

عندما تكون البيانات "كثيراً ما يدور حول قيم مشارقة" "أو ليس كذلك" "أو ليس كذلك" "أو ليس كذلك"

1- symmetrical (Normal distribution)

use the Mean

2- skewed (Asymmetrical Distribution)

The mean is pulled toward the "tail" Use the Median.

بـ "مقدمة" "أو مقدمة عاليه في"

### 3- Data Categorical:

Use the Mode

- `df['Column'].describe()`

↳ calculate all of them.

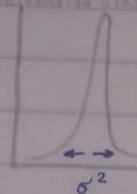
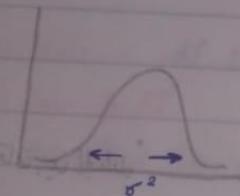
### 4) Variance ( $\sigma^2$ )

• It measures how far each number of the set from the mean.

• It is average of the squared difference from the mean.

• High variance means the data is spread out.

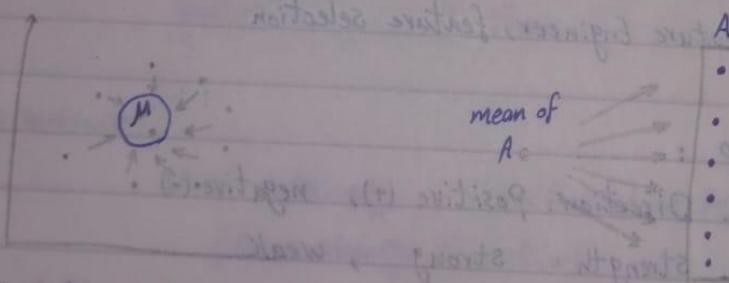
• Low variance means the data points are clustered closely around the average.



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

$$\sum_{i=1}^n \frac{(x_i - M)^2}{n}$$

In AI High Variance often refers to overfitting  
(where the model is too sensitive to small fluctuation).



If  $M$  is large, it means there is variance.

If  $M$  is small, it means there is no variance.  $\sigma^2$  means here the difference between elements of  $A$ .

In pandas: `df['Column'].var()` and there mean ( $M$ )

## 5) standard Deviation ( $\sigma$ )

- The square root of the variance.  $\sigma = \sqrt{\text{Variance}}$
  - We use it to identify Outliers.
- The difference : Variance ( $\sigma^2$ ) standard deviation ( $\sigma$ )
- The average of squared difference from the mean.
  - squared units (it makes it hard to visualize)
  - Used in mathematical optimization and loss functions.
  - The square root of the variance. original Units (This is intuitive)
  - use to detect Outliers and for data scaling.

They are used for calculate the distribution for data.

In pandas: `df['Column'].std()`

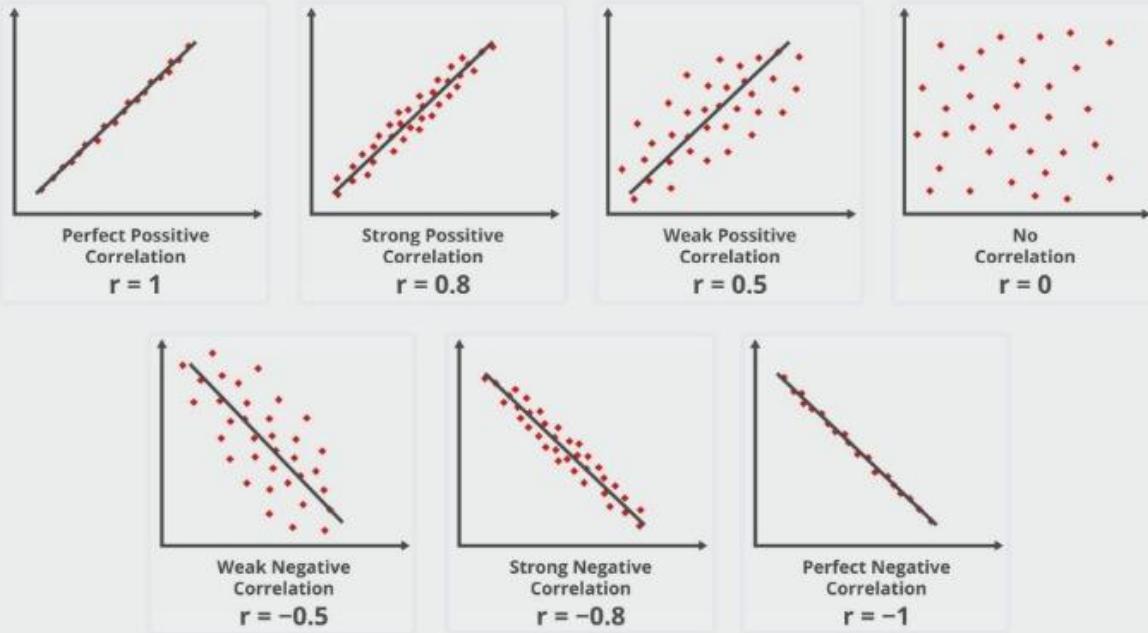
## 6) Covariance

- It means how two variables move together.
  - positive Covariance: Both variables move together
  - negative Covariance: One increases, the other decreases.
- Used for feature Engineering
- "PCA"
- Relationship:
  - Direction: Positive (+), negative (-)
  - strength: strong, weak

### Four Quadrants:

- strong positive: when Column A goes up, Column B goes up significantly.
- strong negative: when Column A goes up, Column B goes down significantly.
- weak positive: There is a slight upward trend, but it is 'noisy'
- weak negative: There is a slight downward trend,

## Correlation



## 7) Correlation

- A standardized version of Covariance . It ranges from -1 to +1
- +1 perfect positive relationship.
- -1 perfect negative relationship.
- 0 No relationship at all.
- It's vital for "feature selection". if two columns are 99% correlated you can drop one of them because they provide the same information for AI model.
- used in
  - Confusion matrix.
  - Correlation matrix
  - heatmap

In Pandas : df. corr()

Use Case In AI:

standard Deviation :

Data scaling and Outlier detection.

a	a
s	s

"main diagonal" means relationship with itself.

Variance :

Understanding model error

Covariance :

Used internally in algorithms like PCA  
(Dimensionality Reduction)

Correlation :

Selecting which features (columns) are most important for the model.

Feature Engineering: Creating new information for existing columns . If we have "length", "width", you might create a new feature like 'Area'.

Feature selection: The process of picking the most important variable . if two columns are (highly related), we can drop one of them to make model faster and more accurate.