

# Fundamentals of Machine learning for and with engineering applications

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Sep 10, 2025



# Correlated variables

Let's make a few example of correlated variables:

- 1 Basketball playing skill and hight
- 2 Age and hair loss
- 3 Oil and gas production vs oil price
- 4 Wind and wind turbine efficiency
- 5 Wind and energy production
- 6 Rain and energy production
- 7 Solar irradiance and energy production

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# How to approach

There are a set of questions that one shall pose when relating two variables.

## 1 Statistical dependence

Two variables have their distribution and, even if very similar, are unrelated

## 1 Causal dependence

Two variables depend on each other.

### Discussion point

How does this relate to soft and hard modeling?

- Scatterplots (MatPLOtLib)

It is one of the simplest ways to graphically display their relationship (it can be 3D).

- Heathmaps (seabon)
- Correlation matrix plots (pandas)

# Correlation

The covariance or joint variance between two random variables is an extension of the concept of variance.

$$\text{Cov}[XY] = \sigma_{xy} = E[(X - \bar{X})(Y - \bar{Y})] = \frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{X})(y_i - \bar{Y}) = \frac{N}{N-1} (E[XY] - E[X]E[Y])$$

- Generalization of variance
- Consider the covariance of a variable with itself

Cov [XY]

Variance: always positive Covariance: positive or negative

# Correlation Analysis

- The correlation between two random variables is a measure of the strength of their linear relationship
- Parametric Correlation

Measures a linear (Pearson) dependence between two variables ( $x$  and  $y$ ) is known as parametric correlation test because it depends

xxx

# Pearson's $\rho$ Value

The correlation coefficient  $\rho$  between to

- It is closely linked to the concept of covariance:
- Assumes normal distribution
- $\rho$  ranges between -1 (per

xxxx

# Correlation examples

Strong negative, weak and Strong Positive -1 0 +1

# Example in python

python

note: the implementation gives the covariance matrix ( $\text{Cov}(x, y)$ )



# Correlation Coefficient - Interpretation

What do I do with this number

The variation of  $x$  *explains* a corr coef variation of  $y$

What does it mean to explain?

Correlation is not causation!!

# Regression Coefficient - interpretation

Plotting two variables on a scatter plot, a straight line means a  $\rho = +/ - 1$

# Regression Coefficient - examples

Python

# Regression Coefficient - limitation

Anscarbe's Quartet: Four different pairs of variables

4 distributions with the same means (7.5), standard deviation (4.12), correlation (0.81) and regression line ( $y=3 + 0.5x$ )

plots

# Spearman Rank Correlation

The Spearman correlation evaluates a monotonic relationship between two variables - Continuous or Ordinal and it is based on the *ranked* values for each variable rather than the raw data-

Plots Not monotonic fails (more advanced methods for this)

# Spearman Rank Correlation

- Rank correlation compares the ranks (ordering)

# Pearson vs Spearman

Figures

# Spearman Rank Correlation

- Calculated the same way as the Person correlation coefficient but using ranks instead of values

missing



# Spearman Rank Correlation example

python

# Kendall tau

quick, not so much used

# Correlation does NOT indicate Causation

Fancy slides here

# Graphing Bivariate Data

Scatterplots between two variables is one of the most  
R<sup>2</sup> coefficient of determination

# Scatterplots combined with Histograms

plots

# Uncorrelated and Independent Random Variables

The two random variables  $X$  and  $Y$  are said to be

**Uncorrelated** if :  $Cov(X, Y) = 0$  **Independent** if :  $f_{XY}(x, y) = f_X(x)f_Y(y)$

# Correlation versus Dependence

- Uncorrelated Random Variables

Random variables are uncorrelated if there is no linear relationship between them. Mathematically, two random variables  $X$  and  $Y$  are uncorrelated if their covariance is zero.  $\text{Cov}(X, Y) = 0$

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# Causation Implies Dependency

Statistical dependency: Variables are causally related must be statistically dependent

- Causal Dependency



# Dependency Does not Imply Causation

- Correlation without Causation

Two variables might be correlated (and hence dependent) due to a coincidence, a lurking variable, or confounding factor.

- Common cause

Two variables might be dependent because they are both influenced by a third variable