

# Feature Scaling

Feature Scaling is a very important step to **pre-process** the data set before apply the machine learning algorithm

## Context

The motivation before **Feature Scaling** is when we have number multiple Features or Variables  $p$

Feature A	Feature B
5	20000
10	30000
15	25000
20	40000
25	50000

- As you can notice both Features are a huge difference in the range of numbers
- In many machine learning algorithms they will interpret that the **Feature B** in this example have much more impact and will result in a very off results and predictions
- That's why we **Feature scale** our variable as a way to normalize the values which will results in a much better **approximations** and **convergence** in the case of [Gradient Descent](#)

## Feature Scaling Methods

There is mainly two major methods that are used in feature scaling

1. **Normalization**
2. **Standardization**

### Normalization

Also Known as *min-max scaling* Same as in **vector normalization** it shifts and scale the values within a range of  $[0, 1]$  or  $[-1, 1]$

Following this Formula :

$$X' = \frac{X - X_{min}}{X_{max} - X_{min}}$$

- Both  $X_{max}$  and  $X_{min}$  are the upper and lower bound of the normalization interval
- $X'$  is the new calculated value of  $X \rightarrow$  The normalized value of  $X$

### Standardization

Also Known as *z-score normalization* which is derived from the **normal distribution** it transforms the data to have a mean  $\mu$  of 0 and **Standard Deviation** of 1 Given by the z-score formula :

$$X' = \frac{X - \mu}{\sigma}$$

## They Key Differences

Normalization	Standardization
Re-scale values to range between 0 and 1	Centers data around the <b>mean</b> and scales to a standard deviation of 1
Useful when the distribution of the data is <b>unknown</b> or <b>not Gaussian</b>	Useful when the distribution of the data is <b>Gaussian</b> (Normal) or <b>unknown</b>
Sensitive to <b>outliers</b>	Less sensitive to <b>outliers</b>
Retains the shape of the <b>original distribution</b>	Changes the shape of the original distribution
<b>May not preserve</b> the relationships between the data points	Preserves the relationships between the data points