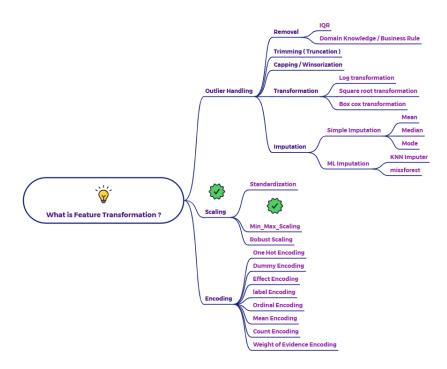
# Explain Min-max scaling

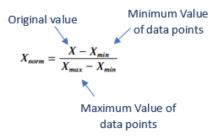


# Min-Max Scaling (Normalization)

## 1. Explanation of Min-Max Scaling

 Min-Max scaling is a technique that scales numerical data to fit within a specific range, typically between 0 and 1. It transforms values linearly, based on the minimum and maximum values of the original data.

# 2. How to Calculate Min-Max Scaling



The formula for Min-Max scaling is:

 $x'=(x-x\min)/(x\max-x\min)$ 

#### Where:

- o (x): Original value
- $\circ$  (x\_{min}): Minimum value of the variable
- $\circ$  (x\_{max}): Maximum value of the variable

#### Example:

Let's say we have the following data for a variable "Income" (in thousands of dollars): 30, 50, 20, 60, 40

Calculate the minimum value  $((x_{\min}))$ : 20

Calculate the maximum value  $((x_{max}))$ : 60

Scale each value:

- For 30: (30 20) / (60 20) = 0.25
- For 50: (50 20) / (60 20) = 0.75
- For 20: (20 20) / (60 20) = 0
- For 60: (60 20) / (60 20) = 1
- For 40: (40 20) / (60 20) = 0.5

So, the Min-Max scaled "Income" values are: 0.25, 0.75, 0, 1, 0.5

### 3. When to Use Min-Max Scaling

- When you need your data to be within a specific range, such as 0 to 1.
- When you are using algorithms that are sensitive to the scale of the input data,
  such as:
  - Neural Networks
  - K-Nearest Neighbors (KNN)
- When you do not assume any specific distribution for your data.

## 4. Strengths and Weaknesses of Min-Max Scaling

#### o Strengths:

- Simple to implement and easy to understand.
- Preserves the original distribution of the data.
- Useful when you need data in a bounded range.

#### O Weaknesses:

- Sensitive to outliers. Outliers can significantly affect the  $(x_{\min})$  and  $(x_{\max})$  values, leading to poor scaling for the rest of the data.
- Not robust to new data points that fall outside the original range used to calculate  $(x_{min})$  and  $(x_{max})$ .