

# Data Normalization

## Attribute Normalization

- In the context of machine learning, it is termed as **feature normalization**
- An attribute is normalised by **scaling its value** so that they **fall within a small specified range** (for example 0.0 to 1.0)
- Normalization is particularly useful for classification algorithms involving distance measurements and clustering
- For distance based approaches, normalization **helps prevent attributes with large ranges from overweighting attributes with smaller ranges**

### Data Standardization (z-score Normalization)

- The process of rescaling one or more attributes so that the transformed data have 0 mean and unit variance i.e. standard deviation of 1
- Standardization assumes that data has a Gaussian distribution
  - This assumption does not strictly have to be true, but this technique is more effective if your attribute distribution is Gaussian
- In this process, values of an attribute, A, are normalised based on the mean and standard deviation of A
- A value,  $x$ , of attribute A is normalised to  $\hat{x}$  by computing

$$\hat{x} = \frac{x - \mu_A}{\sigma_A}$$

- $\mu_A$ : mean of attribute A
- $\sigma_A$ : standard deviation of attribute A

### Data Standardization (z-score Normalization)

- This method of normalization is useful
  - when the actual minimum and maximum of attribute A are unknown
  - when there are outliers that dominates the Min-Max normalization
  - when data has Gaussian distribution (symmetric distribution)
- This method of normalization is useful when the ML algorithms make any assumptions of Gaussian distribution

## Illustration of Data Standardization (z-score Normalization)

	Temperature	Humidity	Rain		Temperature	Humidity	Rain
1							
2	25.46875	82.1875	6.75		1.05444	-1.57673	-0.97166
3	26.19298	83.14912	1762		1.51216	-1.41995	2.62269
4	25.17021	85.34043	653		0.86576	-1.06268	0.35088
5	24.29851	87.68657	963		0.31484	-0.68016	0.98680
6	24.06923	87.64615	254		0.16993	-0.68675	-0.46476
7	21.20779	95.94805	340		-1.63853	0.66679	-0.28965
8	23.48571	96.17143	38.3		-0.19886	0.70321	-0.90714
9	21.79487	98.58974	29.3		-1.26749	1.09749	-0.92558
10	25.09346	88.3271	4.5		0.81726	-0.57573	-0.97627
11	25.39423	90.43269	113		1.00735	-0.23244	-0.75508
12	23.89076	94.53782	736		0.05714	0.43686	0.52138
13	22.5098	99	608		-0.81564	1.16438	0.25871
14	22.904	98	718		-0.56650	1.00134	0.48451
15	21.72464	99	513		-1.31187	1.16438	0.06517



$\mu$ : 23.80035 91.86 481  
 $\sigma$ : 1.58225 6.13 488

0.000 0.000 0.000  
 1 1 1