

# Data Preprocessing and Scaling

⚠ Feature scaling is critical: Distance-based algorithms are sensitive to scale

## Scaling Methods



### StandardScaler

$$z = (x - \mu) / \sigma$$

- ✓ Zero mean
- ✓ Unit variance
- ✓ Assumes Gaussian



### MinMaxScaler

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

- ✓ Scale to [0, 1]
- ✓ Preserves zero
- ✓ Bounded range



### RobustScaler

$$x' = (x - \text{median}) / \text{IQR}$$

- ✓ Uses median
- ✓ Uses IQR
- ✓ Robust to outliers

## Preprocessing Pipeline

1



### Missing Values

Imputation before clustering

2



### Outlier Treatment

Remove or cap extremes

3



### Feature Selection

Remove irrelevant features

4



### Dimensionality

Handle curse of dimensions

## Calculation Examples

### StandardScaler Example

Data: [10, 20, 30, 40, 50]

```
 $\mu = 30$   
 $\sigma = 14.14$   
For  $x = 10$ :  
 $z = (10 - 30) / 14.14$   
 $z = -1.41$ 
```

Result: [-1.41, -0.71, 0, 0.71, 1.41]

### MinMaxScaler Example

Data: [10, 20, 30, 40, 50]

```
 $\text{min} = 10, \text{max} = 50$   
 $\text{range} = 50 - 10 = 40$   
For  $x = 20$ :  
 $x' = (20 - 10) / 40$   
 $x' = 0.25$ 
```

Result: [0, 0.25, 0.5, 0.75, 1.0]

### RobustScaler Example

Data: [10, 20, 30, 40, 100]

```
 $\text{median} = 30$   
 $Q1 = 20, Q3 = 40$   
 $\text{IQR} = 40 - 20 = 20$   
For  $x = 100$ :  
 $x' = (100 - 30) / 20 = 3.5$ 
```

Result: [-1.0, -0.5, 0, 0.5, 3.5]