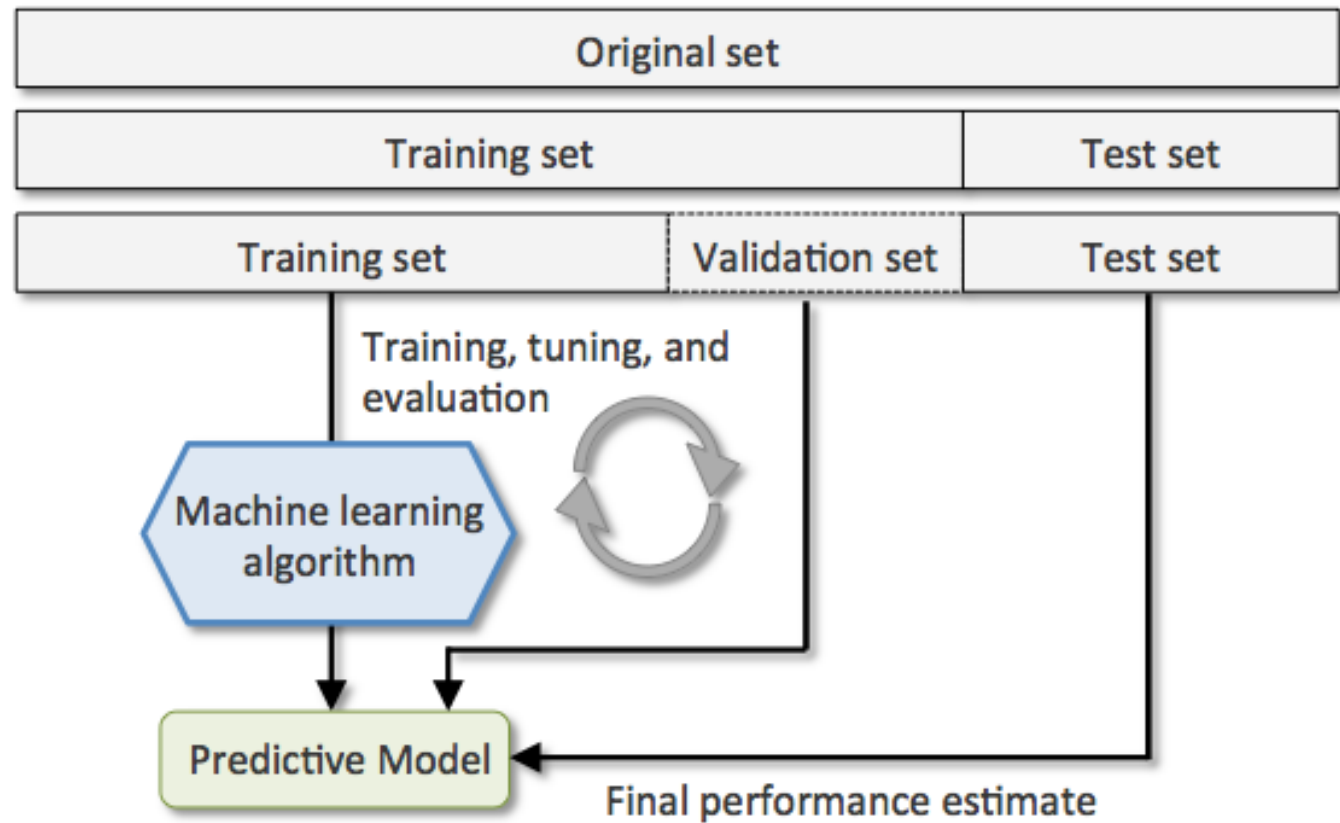


# Test Train split

## Test Train Split



# Random Seed

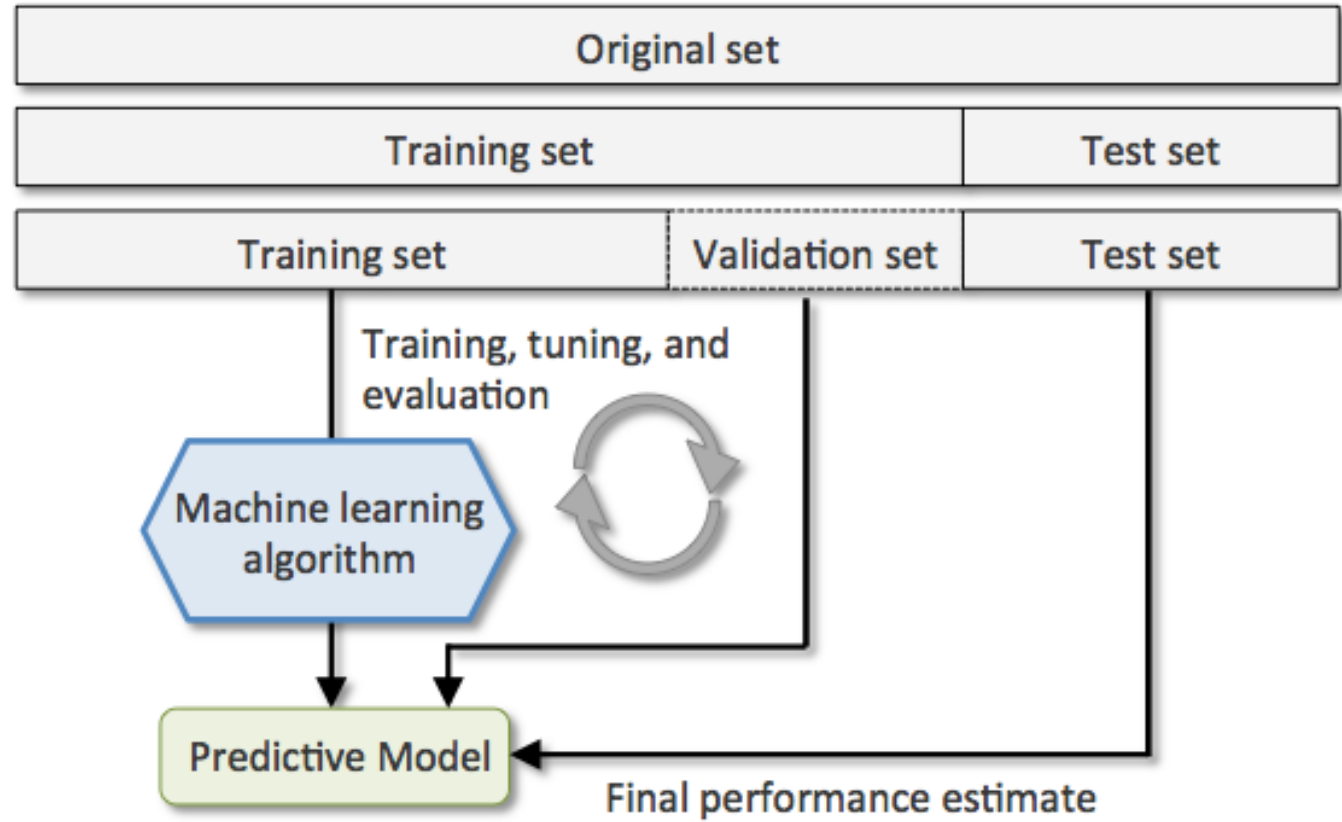
## WHY

Generate same set of Random numbers while

1. Test Train Split
2. Model Training

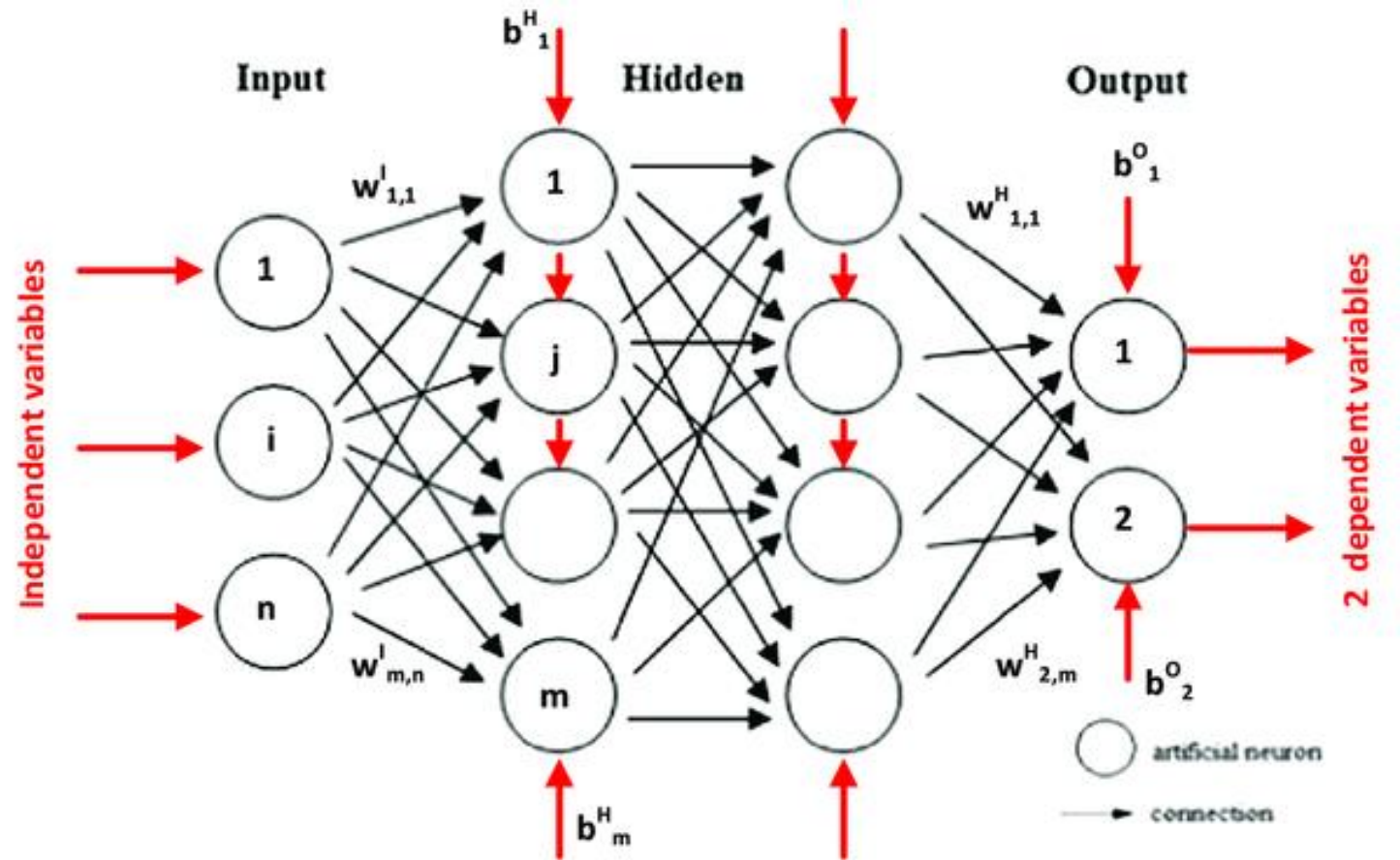
# Random Seed

## Test Train Split



# Random Seed

## Model Training



# Feature Scaling

## Types

Feature scaling is a method used to normalize the range of independent variables or features of data.

1. Normalization
2. Standardization

# Feature Scaling

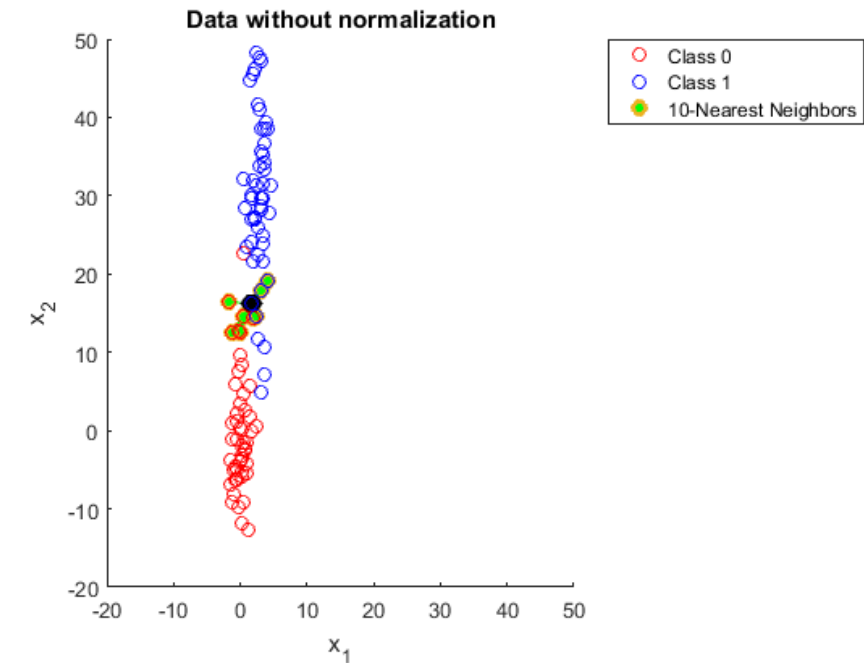
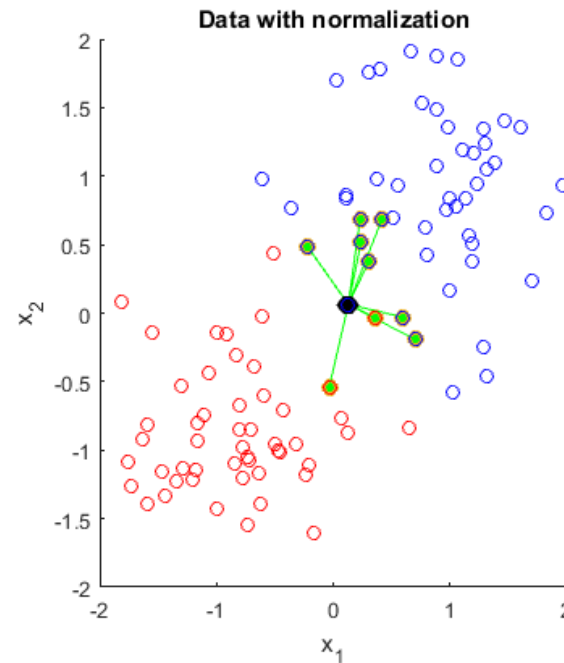
## Example

Predicting whether or not you want to give loan depending on.

1. Age – Range 0-100
2. Income – 10,000 to 10,000,000

# Feature Scaling

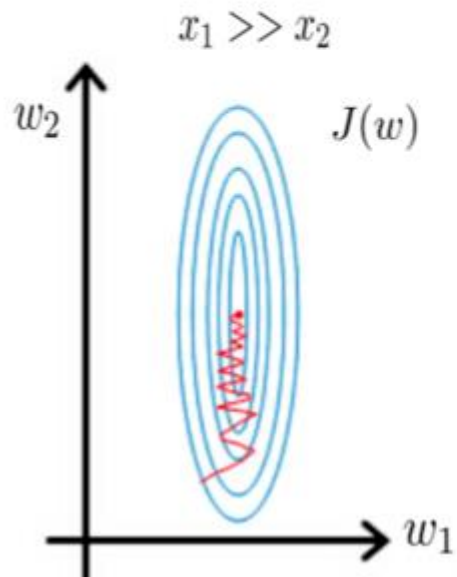
Age – Range  
0-100  
Income – 10,000  
to 10,000,000



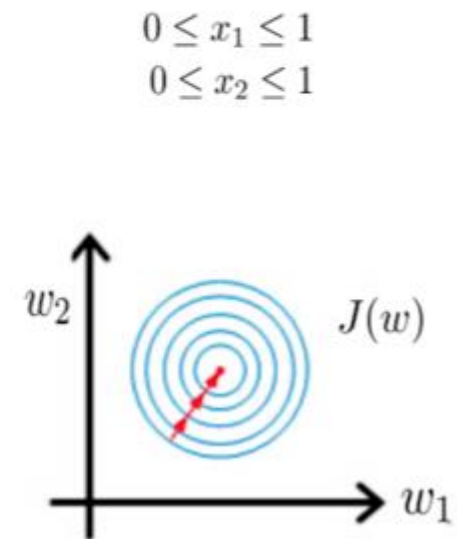
# Feature Scaling

WHY

Gradient descent  
without scaling



Gradient descent  
after scaling variables





# Feature Scaling

## Normalization

$$x_{\text{norm}} = \frac{x - \min(x)}{\max(x) - \min(x)}$$

Values 2,4,6

2 will become =  $(2-2)/(6-2) = 0$

4 will become =  $(4-2)/(6-2) = 0.5$

6 will become =  $(6-2)/(6-2) = 1$

# Feature Scaling

## Standardization

$$x_{\text{stand}} = \frac{x - \text{mean}(x)}{\text{standard deviation}(x)}$$

Values 2,4,6 => mean =4 & SD is 2

2 will become =  $(2-4)/(2) = -1$

4 will become =  $(4-4)/(2) = 0$

6 will become =  $(6-4)/(2) = 1$

# Feature Scaling

## Example

Age = 20, 40, 60      Income = 40,000 , 80,000 & 120,000

Age 20 will become = 0 in norm. and -1 stand.

Age 20 will become = 0.5 in norm. and 0 stand.

Age 60 will become = 1 in norm. and 1 stand.

Income 40,000 will become = 0 in norm. and -1 stand.

Income 80,000 will become = 0.5 in norm. and 0 stand.

Income 120,000 will become = 1 in norm. and 1 stand.