

# A-Z Machine Learning using Azure Machine Learning (AzureML)

Hands on AzureML: From Azure Machine Learning Introduction to Advance Machine Learning Algorithms. No Coding Required.

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Created by Jitesh Khurkhuriya Last updated 3/2018 English English

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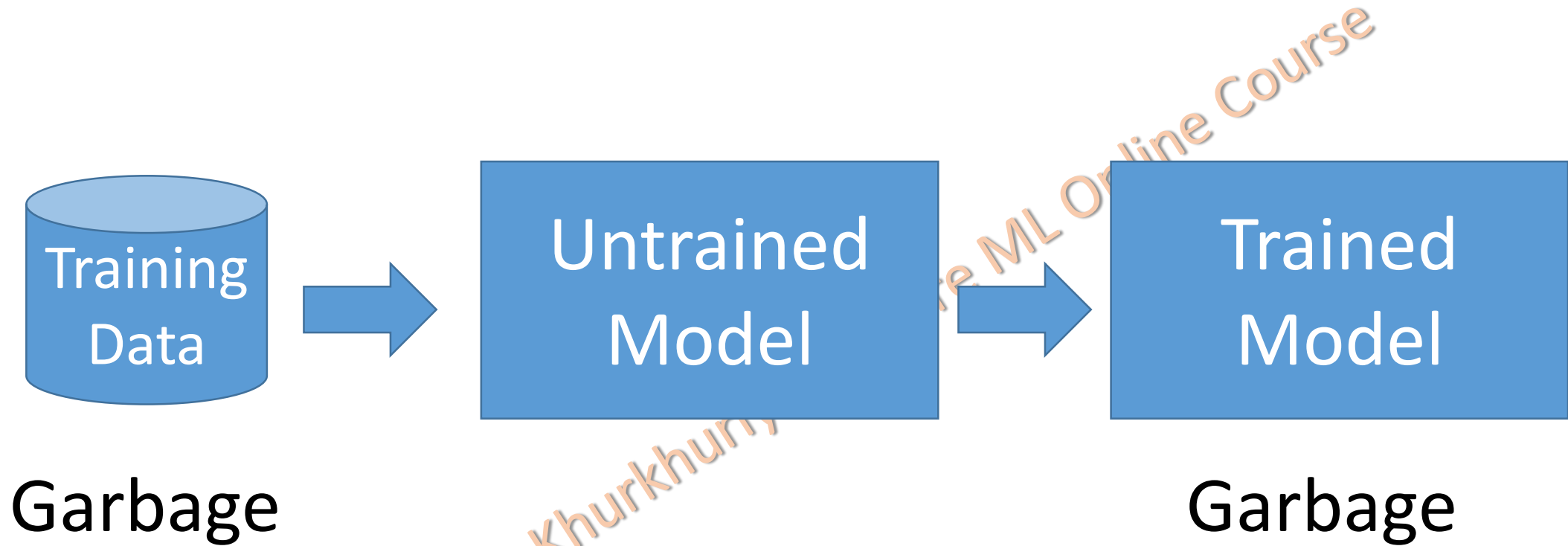
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# Feature Selection

# Loan Approval Data

- **Personal Information** – Name, Gender, Marital Status, Age etc
- **Financial Information** – Monthly Income, Additional Income, Credit History, Any other loan
- **Demographic Information** – Address, Race, Address type, Type of customer
- **Loan Provider Information** – Product type, existing customer, Group based on past behaviour
- **Application Details** – Date of Application, Advisor information, Campaign Information
- **Educational Details** – Level of education, Type of education, Institute
- **Employment Information** – Type of employment, employer category, Designation, Years of experience
- **Coapplicant Details** – all the above

# Importance of Feature Selection



# Why to Use Feature Selection?

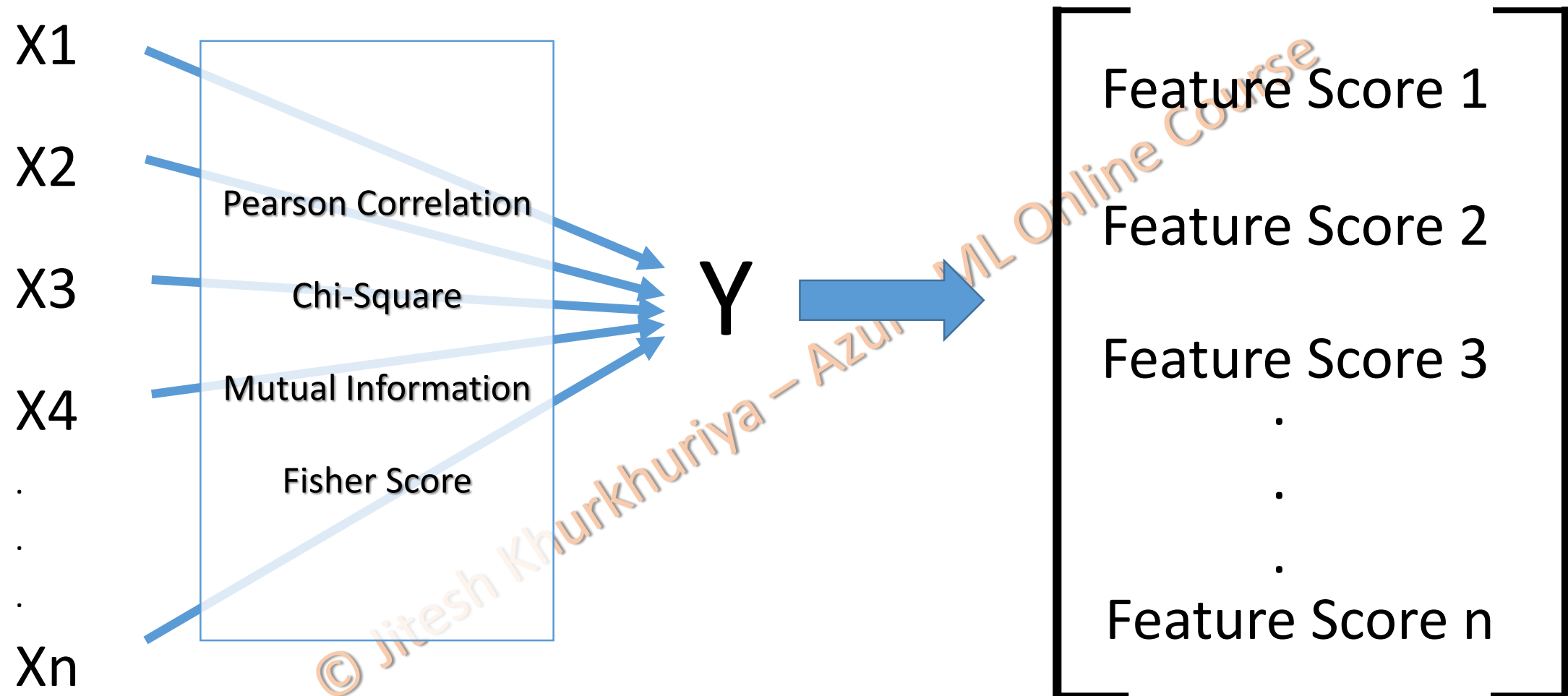
- Simplification of models to make them easier to interpret by researchers/users
- Shorter training times
- Improves Accuracy
- To avoid the curse of dimensionality,
- Enhanced generalization by reducing overfitting

# Methods of Feature Selection

- **Filter Based Methods** – Uses Correlation with the outcome variables
- **Wrapper Methods** – Uses a subset of features by selecting the best set
- **Embedded Methods**

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# Filter Based Feature Selection



# Wrapper Method

X1 X2 X3 X4 ..... Xn

Learning Algorithm

Classification  
Regression

Performance

Accuracy  
Precision  
Root-Mean-Squared

Subset  
X2 X4



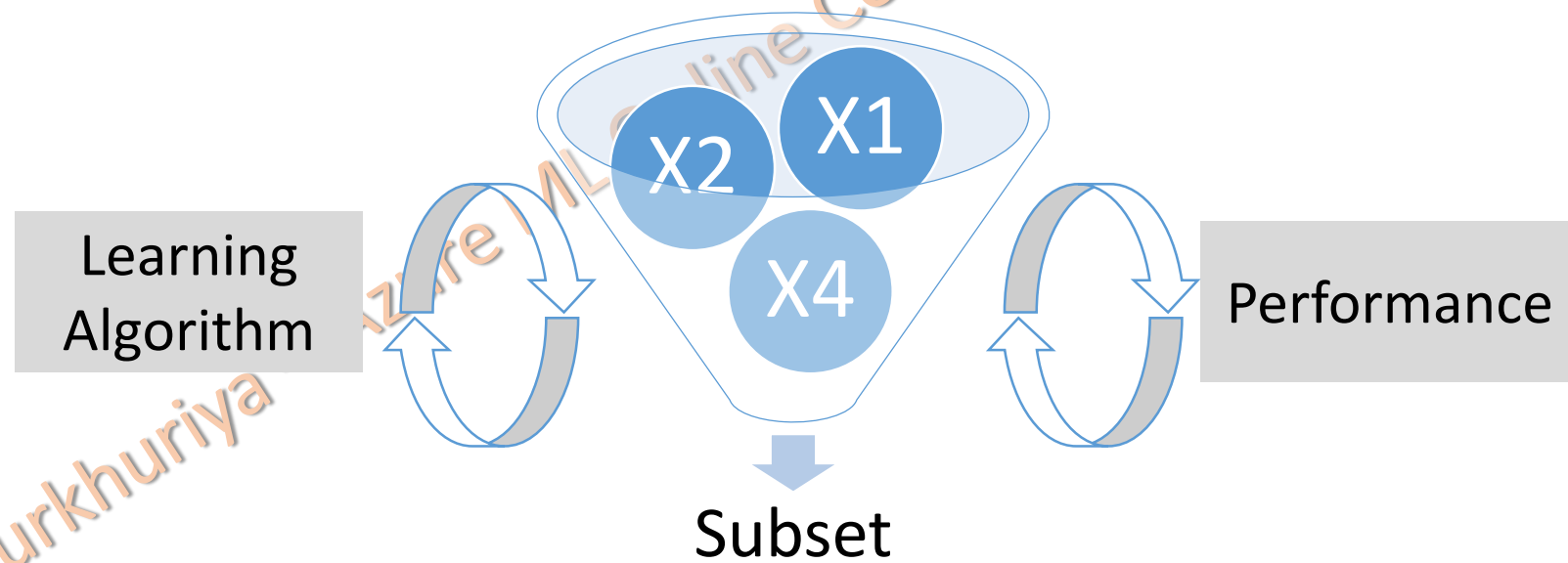
# Wrapper Method

X1 X2 X3 X4 ..... Xn

Forward Selection

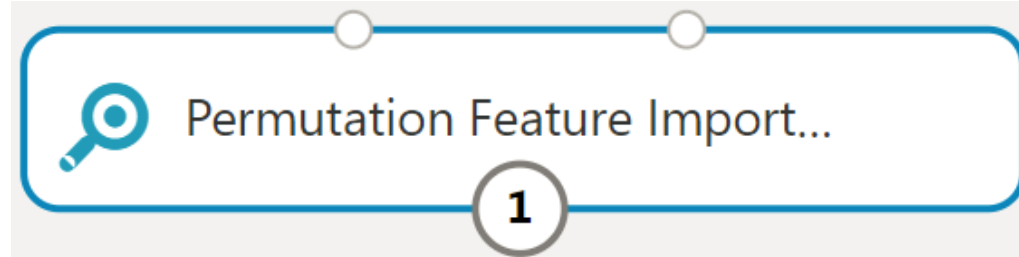
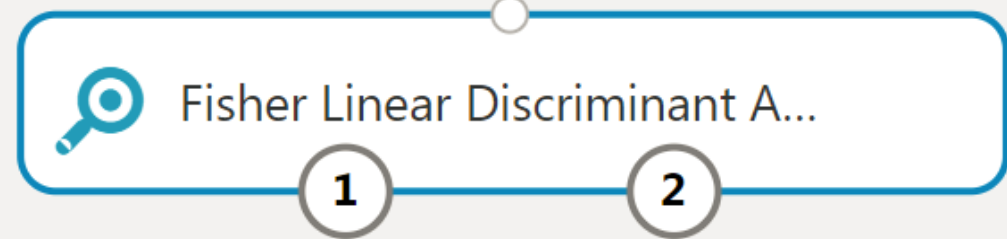
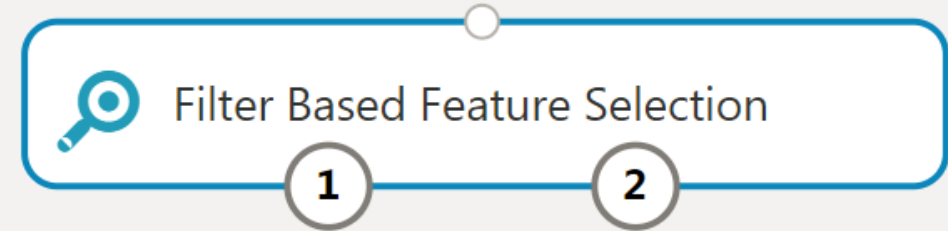
Backward Elimination

Recursive Feature Elimination



# AzureML Feature Selection Methods

- Filter Based Feature Selection
- Fisher Linear Discriminant Analysis
- Permutation Feature Importance



# Filter Based Feature Selection

# Types of Feature Scoring Methods

- Pearson Correlation
- Mutual Information
- Kendall Correlation
- Spearman Correlation
- Chi Squared
- Fisher Score
- Count Based



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# Pearson Correlation

# Standard Deviation

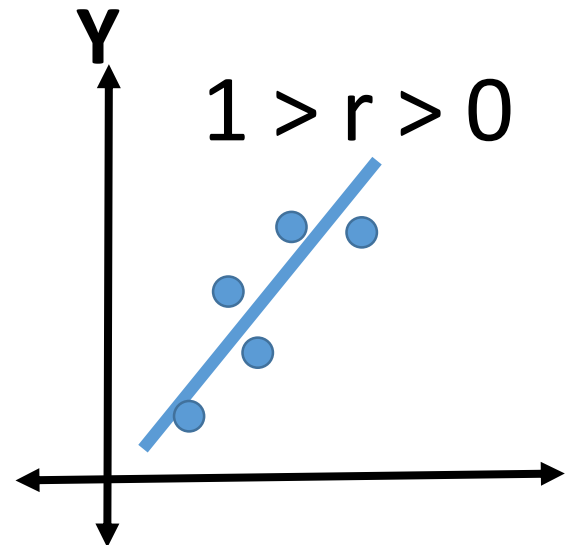
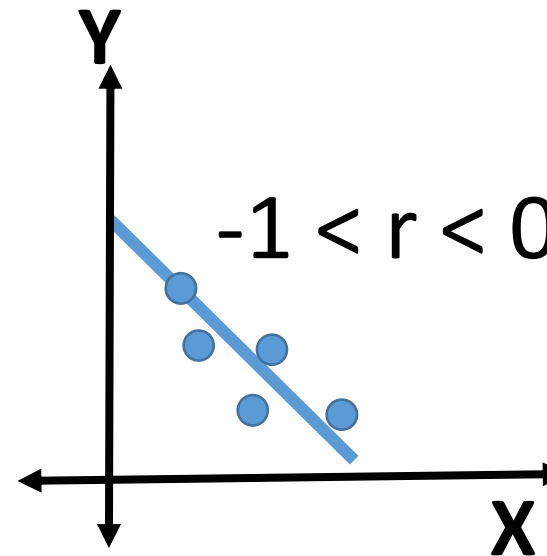
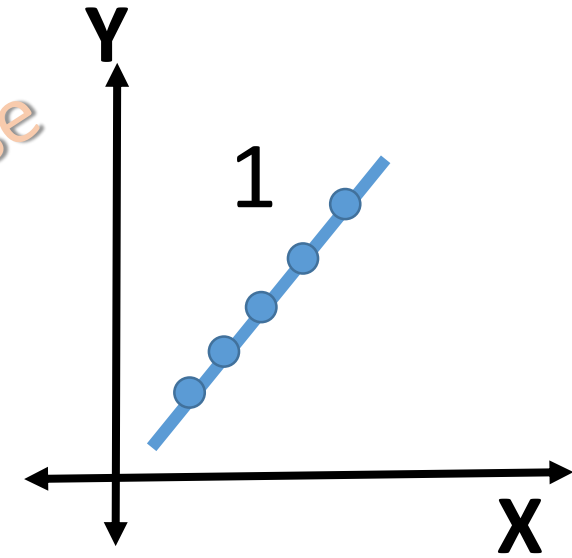
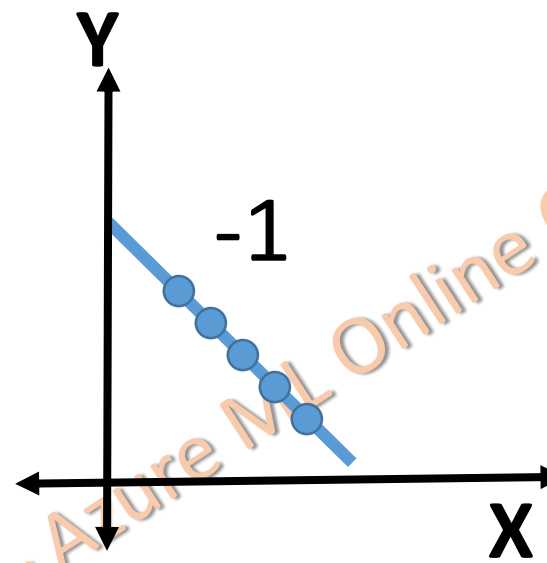
$$\sigma_x = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$$

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# Pearson Correlation Coefficient

- Measure of linear correlation between two variables

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(N - 1) \sigma_x \sigma_y}$$



# Pearson Correlation Coefficient

Engine-size	Horsepower	Price
130	111	13495
130	111	16500
152	154	16500
109	102	13950
136	115	17450
136	110	15250
136	110	17710
136	110	18920
131	140	23875
108	101	16430
108	101	16925
164	121	20970
164	121	21105
164	121	24565

$$\sigma_x = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$$

Engine Size

$$\sigma_x = 18.9$$

HP

$$\sigma_x = 14.39$$

Price

$$\sigma_y = 3267.78$$



# Engine Size – r

engine-size	horsepower	price
130	111	13495
130	111	16500
152	154	16500
109	102	13950
136	115	17450
136	110	15250
136	110	17710
136	110	18920
131	140	23875
108	101	16430
108	101	16925
164	121	20970
164	121	21105
164	121	24565

Engine Size,  $\sigma_x = 18.9$

Price,  $\sigma_y = 3267.78$

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(N - 1) \sigma_x \sigma_y}$$

$$= \frac{519,985}{13 * 18.9 * 3267.78} = 0.65$$

# Horsepower – r

engine-size	horsepower	price
130	111	13495
130	111	16500
152	154	16500
109	102	13950
136	115	17450
136	110	15250
136	110	17710
136	110	18920
131	140	23875
108	101	16430
108	101	16925
164	121	20970
164	121	21105
164	121	24565

HP,  $\sigma_x = 14.39$

Price,  $\sigma_y = 3267.78$

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(N - 1) \sigma_x \sigma_y}$$

$$= \frac{286,400}{13 * 14.39 * 3267.78} = 0.47$$

# Feature Selection

engine-size	horsepower	price
130	111	13495
130	111	16500
152	154	16500
109	102	13950
136	115	17450
136	110	15250
136	110	17710
136	110	18920
131	140	23875
108	101	16430
108	101	16925
164	121	20970
164	121	21105
164	121	24565

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(N-1) \sigma_x \sigma_y}$$

Horsepower  $r = 0.47$

Engine Size  $r = 0.65$



# Pearson Correlation Coefficient

- Advantages
  - Identifies the correlation as well as the extent
  - Identifies the positive or negative correlation
  - Easiest to apply and understand among all the other correlation techniques
- Disadvantages
  - Outliers affect the value of “r”
  - Not good for non-linear relationship
  - Might not be good to find out causal

Engine	HP	Price
130	111	<b>13495</b>
130	111	<b>16500</b>
152	154	<b>16500</b>
109	102	<b>13950</b>

# Chi-Square

# Chi Squared Test of Independence

- Developed by Karl Pearson
- Evaluates the relationship between two categorical variables
- Steps to Evaluate the Independence
  - Define Hypothesis – Null and Alternate
  - Define Alpha
  - Calculate the Degrees of Freedom
  - State Decision Rule
  - Calculate Test Statistics
  - Results
  - Conclusion

# Chi Square Test of Independence

Flight Status	Weather
Delayed	Rainy
Delayed	Rainy
Delayed	Rainy
Ontime	Rainy
Delayed	Rainy
Ontime	Sunny
Delayed	Rainy
Delayed	Rainy
Ontime	Sunny
Delayed	Rainy
Delayed	Overcast
Delayed	Overcast
Delayed	Overcast
Delayed	Overcast
Delayed	Overcast

## Step 1

**Null Hypothesis** – There is no relationship between Flight Status and Weather

**Alternate Hypothesis** – There is relationship between Flight Status and Weather

## Step 2

$\alpha = 0.05$

# Chi Square Test of Independence

## Step 3

	Rainy	Sunny	Overcast	
Delayed	36	16	13	65
On time	11	84	40	135
	47	100	53	

Calculate the degrees of freedom

$(\text{no of Rows} - 1) * (\text{no of Columns} - 1)$

$(2-1)*(3-1)$

2



# Chi Square Test of Independence

## Step 4

State Decision Rule using  
chi square degrees of  
freedom table

**Table 3-1** Critical Values of the  $\chi^2$  Distribution

df	P						
	0.995	0.975	0.9	0.5	0.1	0.05	0.025
1	.000	.000	0.016	0.455	2.706	3.841	5.024
2	0.010	0.051	0.211	1.386	4.605	5.991	7.378
3	0.072	0.216	0.584	2.366	6.251	7.815	9.348
4	0.207	0.484	1.064	3.357	7.779	9.488	11.143
5	0.412	0.831	1.610	4.351	9.236	11.070	12.832
6	0.676	1.237	2.204	5.348	10.645	12.592	14.449
7	0.989	1.690	2.833	6.346	12.017	14.067	16.013
8	1.344	2.180	3.490	7.344	13.362	15.507	17.535
9	1.735	2.700	4.168	8.343	14.684	16.919	19.023

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Reject the Null Hypothesis if the X square value is greater than 5.991

# Chi Squared Test of Independence

## Step 5 Calculate Test Statistics

$$f_e = \frac{f_c \cdot f_r}{n}$$

(Delayed, Rainy)  
= (47\*65)/200  
= 15.275

Actual

	Rainy	Sunny	Overcast	
Delayed	36	16	13	65
On time	11	84	40	135
	47	100	53	

Expected

	Rainy	Sunny	Overcast	
Delayed	15	33	17	65
On time	32	67	36	135
	47	100	53	

# Chi Squared Test of Independence

## Step 6 Calculate Results

$$\sum \frac{(f_o - f_e)^2}{f_e} = 55.6$$

o – Original/Actual  
e – Expected

	Rainy	Sunny	Overcast
Delayed	28.12	8.38	1.04
On time	13.54	4.03	0.5

## Step 7 Conclusion

Reject the null hypothesis as X-Squared value (55.6) is greater than 5.991

The weather and Flight Status are correlated

# Kendall Correlation

# Kendall Correlation Rank Correlation

- Named after Maurice Kendall
- Measure of Rank Correlation

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# Kendall Correlation Rank Correlation

engine-size	horsepower	price
130	111	13495
130	111	16500
152	154	16500
109	102	13950
136	115	17450
136	110	15250
136	110	17710
136	110	18920
131	140	23875
108	101	16430
108	101	16925
164	121	20970
164	121	21105
164	121	24565

# Kendall Correlation Rank Correlation

	engine-size	price
i	108	16430
	108	16925
	109	13950
k	130	13495
	130	16500
j	131	23875
	136	17450
	136	15250
	136	17710
	136	18920
	152	16500
	164	20970
	164	21105
	164	24565

$$x_i = 108$$

$$y_i = 16,430$$

$$x_j = 131$$

$$y_j = 23,875$$

$$x_i < x_j$$

$$y_i < y_j$$

Concordant Pair

$$x_i > x_k$$

$$y_i < y_k$$

Discordant Pair

$$x_i < x_k$$

$$y_i > y_k$$



# Kendall Correlation Rank Correlation

Kendall Coefficient Tau,  $T = \frac{\text{Number of Concordant Pairs} - \text{Number of Discordant pairs}}{n * (n - 1) / 2}$

1. T is always in the range of +1 to -1
2. Positive value indicates positive correlation
3. Negative value indicates negative correlation
4. If X and Y are independent, T will be zero

# Spearman's Rank Correlation

# Spearman's Rank Correlation

- Named after Charles Spearman
- Measure of Rank Correlation
- Relationship through monotonic function

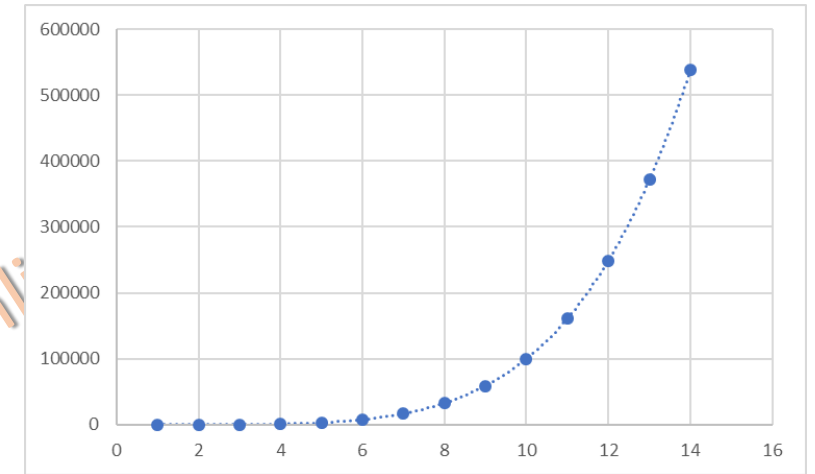
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# Spearman's Rank Correlation

X	Y (X^5)
1	1
2	32
3	243
4	1024
5	3125
6	7776
7	16807
8	32768
9	59049
10	100000
11	161051
12	248832
13	371293
14	537824

$$X1 < X2$$

$$f(X1) < f(X2)$$



Best used for non-linear correlation

$$\rho = 1 - \frac{6 \sum d^2}{n^3 - n}$$

# Calculate rho for Engine Size

engine-size	price
108	16430
108	16925
109	13950
130	13495
130	16500
131	23875
136	17450
136	15250
136	17710
136	18920
152	16500
164	20970
164	21105
164	24565

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# Calculate rho for EngineSize

engine-size	Rank-x	price
108	1	16430
108	1	16925
109	2	13950
130	3	13495
130	3	16500
131	4	23875
136	5	17450
136	5	15250
136	5	17710
136	5	18920
152	6	16500
164	7	20970
164	7	21105
164	7	24565

1. Calculate the rank of the feature w.r.t. other values in the column

## Calculate rho for EngineSize

engine-size	Rank-x		price	Rank-y
108	1		16430	4
108	1		16925	6
109	2		13950	2
130	3		13495	1
130	3		16500	5
131	4		23875	12
136	5		17450	7
136	5		15250	3
136	5		17710	8
136	5		18920	9
152	6		16500	5
164	7		20970	10
164	7		21105	11
164	7		24565	13

2. Calculate the rank of the predicted feature w.r.t. other values in the column

## Calculate rho for EngineSize

engine-size	Rank-x		price	Rank-y	D	D^2
108	1		16430	4	-3	9
108	1		16925	6	-5	25
109	2		13950	2	0	0
130	3		13495	1	2	4
130	3		16500	5	-2	4
131	4		23875	12	-8	64
136	5		17450	7	-2	4
136	5		15250	3	2	4
136	5		17710	8	-3	9
136	5		18920	9	-4	16
152	6		16500	5	1	1
164	7		20970	10	-3	9
164	7		21105	11	-4	16
164	7		24565	13	-4	16

3. Calculate the Rank-difference (RankX and RankY) between price and engine-size and its square



# Calculate rho for EngineSize

engine-size	Rank-x		price	Rank-y	D	D^2
108	1		16430	4	-3	9
108	1		16925	6	-5	25
109	2		13950	2	0	0
130	3		13495	1	2	4
130	3		16500	5	-2	4
131	4		23875	12	-8	64
136	5		17450	7	-2	4
136	5		15250	3	2	4
136	5		17710	8	-3	9
136	5		18920	9	-4	16
152	6		16500	5	1	1
164	7		20970	10	-3	9
164	7		21105	11	-4	16
164	7		24565	13	-4	16

$$\rho = 1 - \frac{\sum d^2}{n^3 - n}$$

$$= 0.6$$

## Calculate rho for HP

HP	Rank-x		price	Rank-y	D	D^2
101	1		16430	4	-3	9
101	1		16925	6	-5	25
102	2		13950	2	0	0
110	3		15250	3	0	0
110	3		17710	8	-5	25
110	3		18920	9	-6	36
111	4		13495	1	3	9
111	4		16500	5	-1	1
115	5		17450	7	-2	4
121	6		20970	10	-4	16
121	6		21105	11	-5	25
121	6		24565	13	-7	49
140	7		23875	12	-5	25
154	8		16500	5	3	9

$$\rho = 1 - \frac{6 \sum d^2}{n^3 - n}$$

$$= 0.49 < 0.6$$

Select Engine Size Over HP

# Correlation Coefficients

## Pearson Correlation

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(N - 1) \sigma_x \sigma_y}$$

## Kandell's Correlation

$$T = \frac{C - D}{C + D}$$

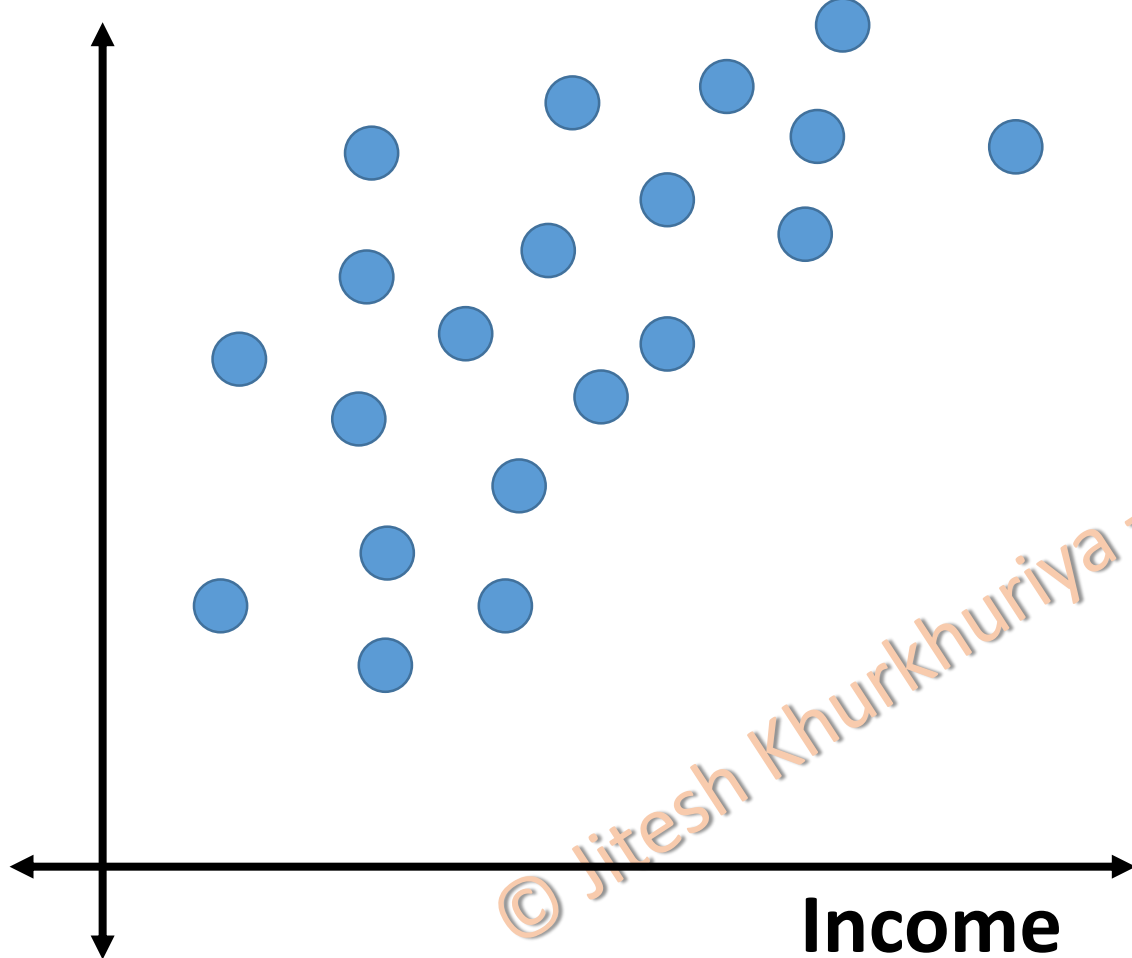
## Spearman's Correlation

$$\text{rho} = 1 - \frac{6 \sum d^2}{n^3 - n}$$

# Fisher Linear Discriminant Analysis

# Loan Approval Dataset

Loan Amt



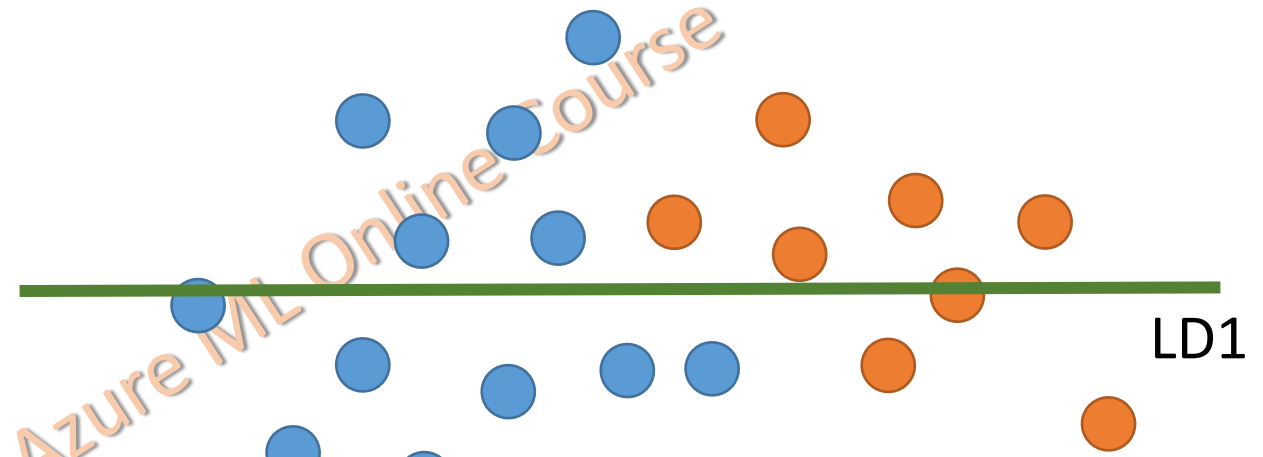
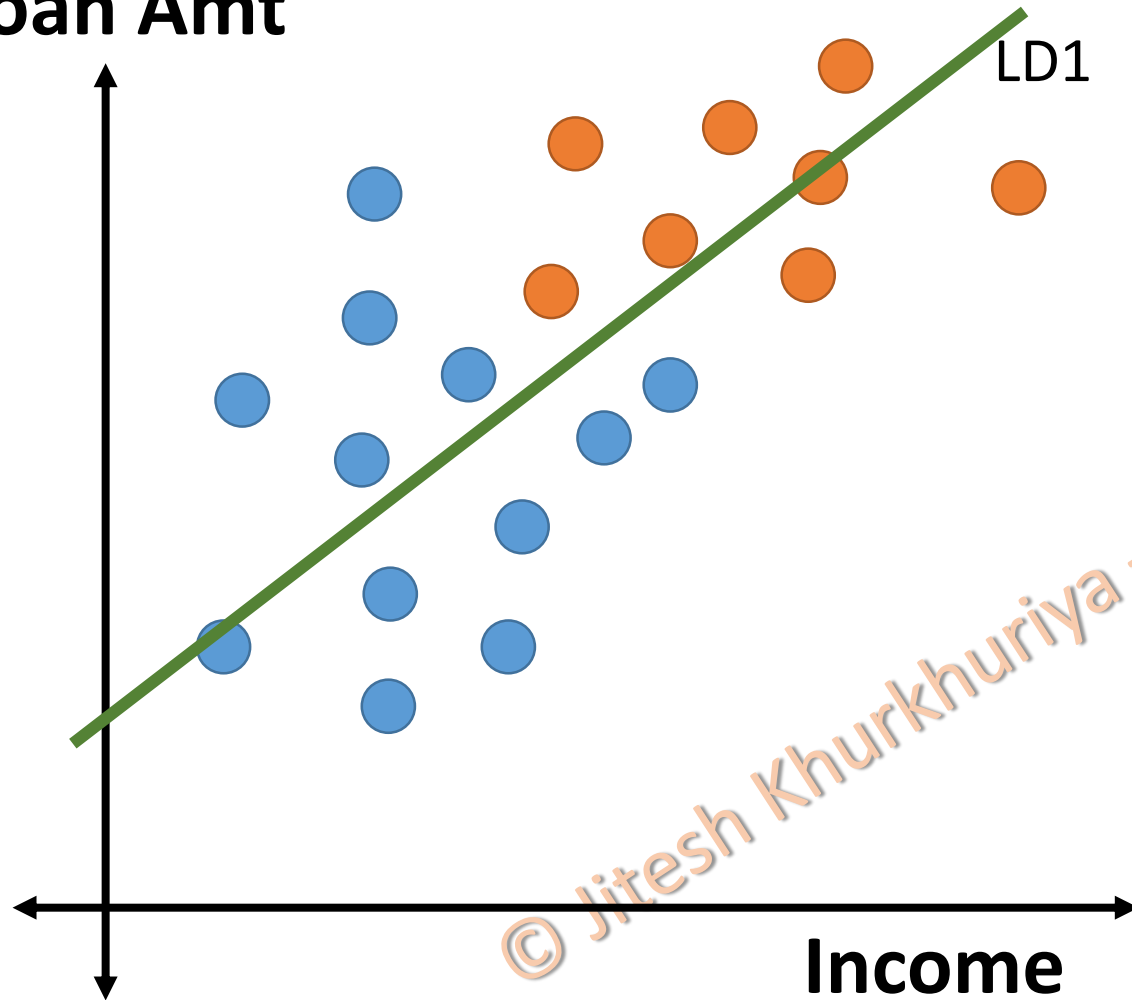
Loan Approved?

Binary Classification

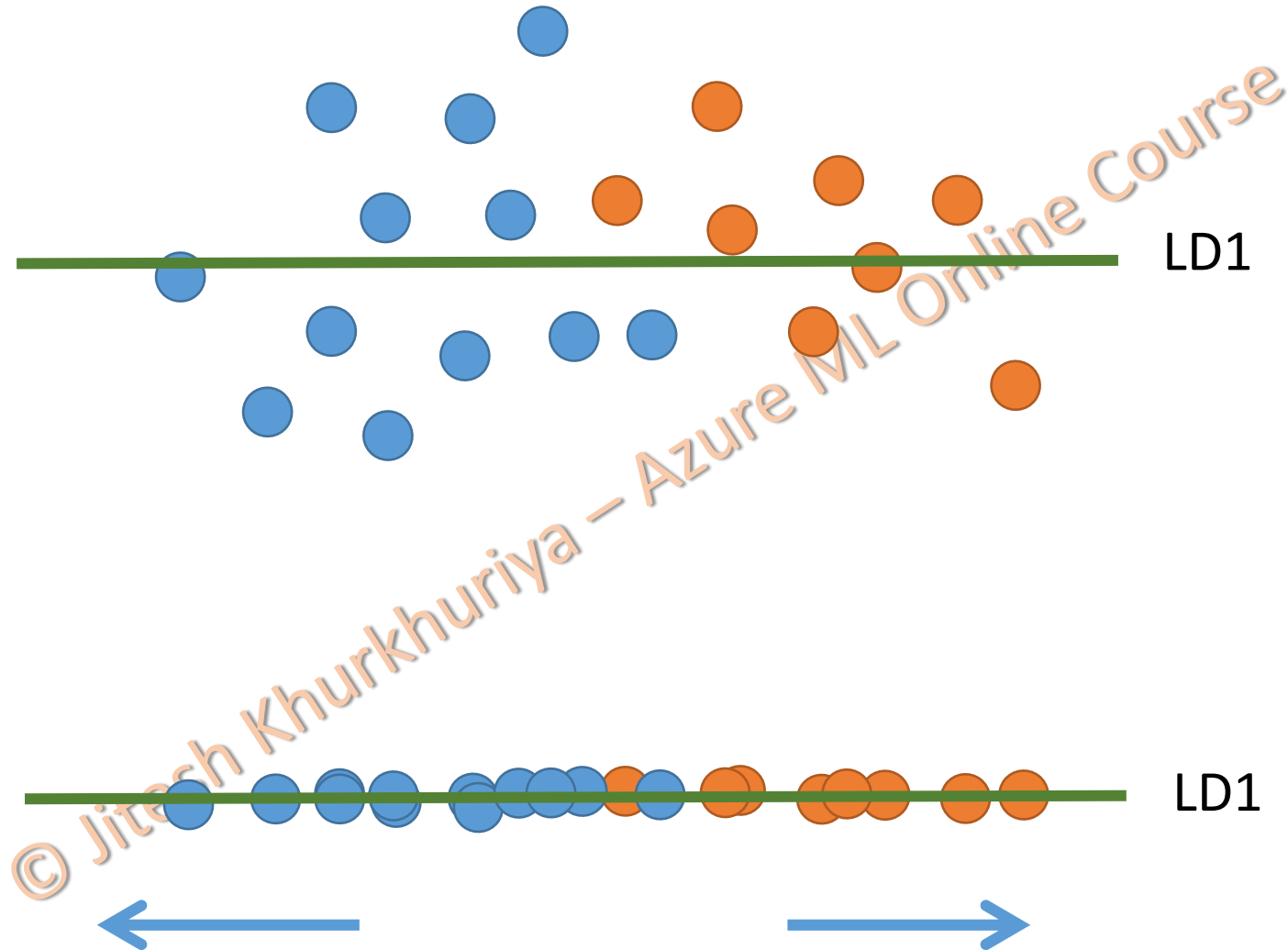
Yes or No

# Linear Discriminant Analysis

Loan Amt



# Linear Discriminant Analysis



# Sir Ronald Fisher

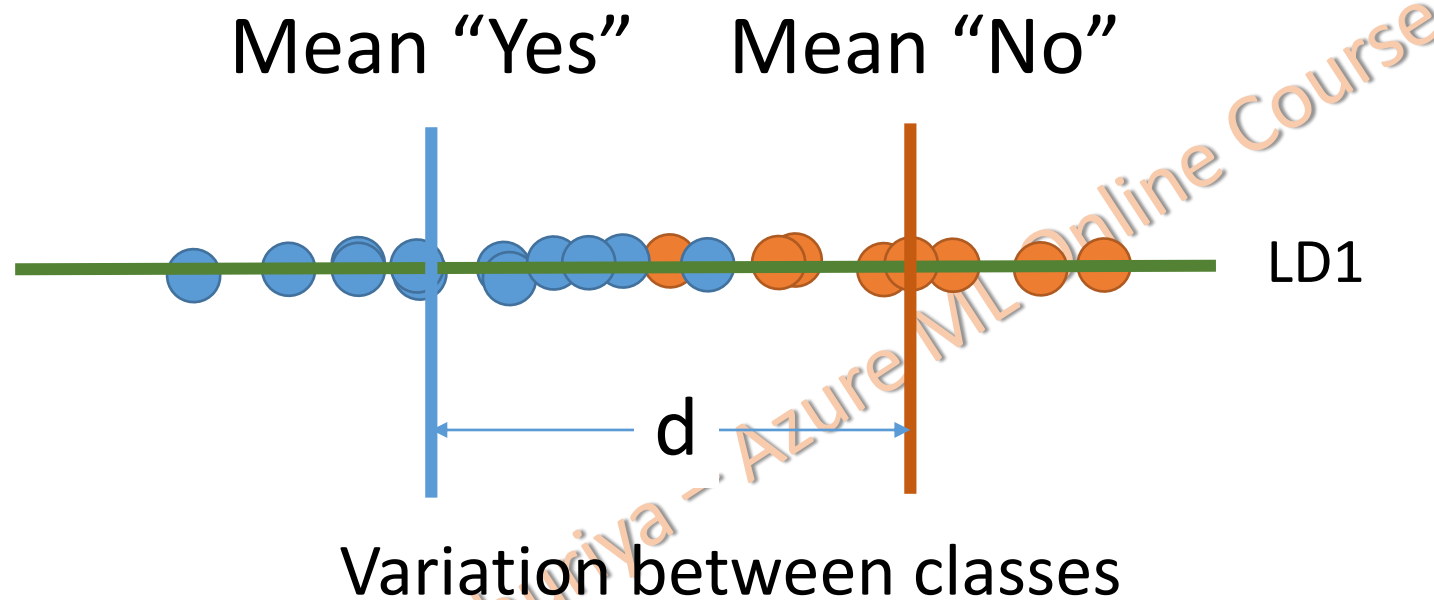
- British statistician and biologist who used mathematics to combine Mendelian genetics and natural selection.
- Developed Analysis of Variance or ANOVA



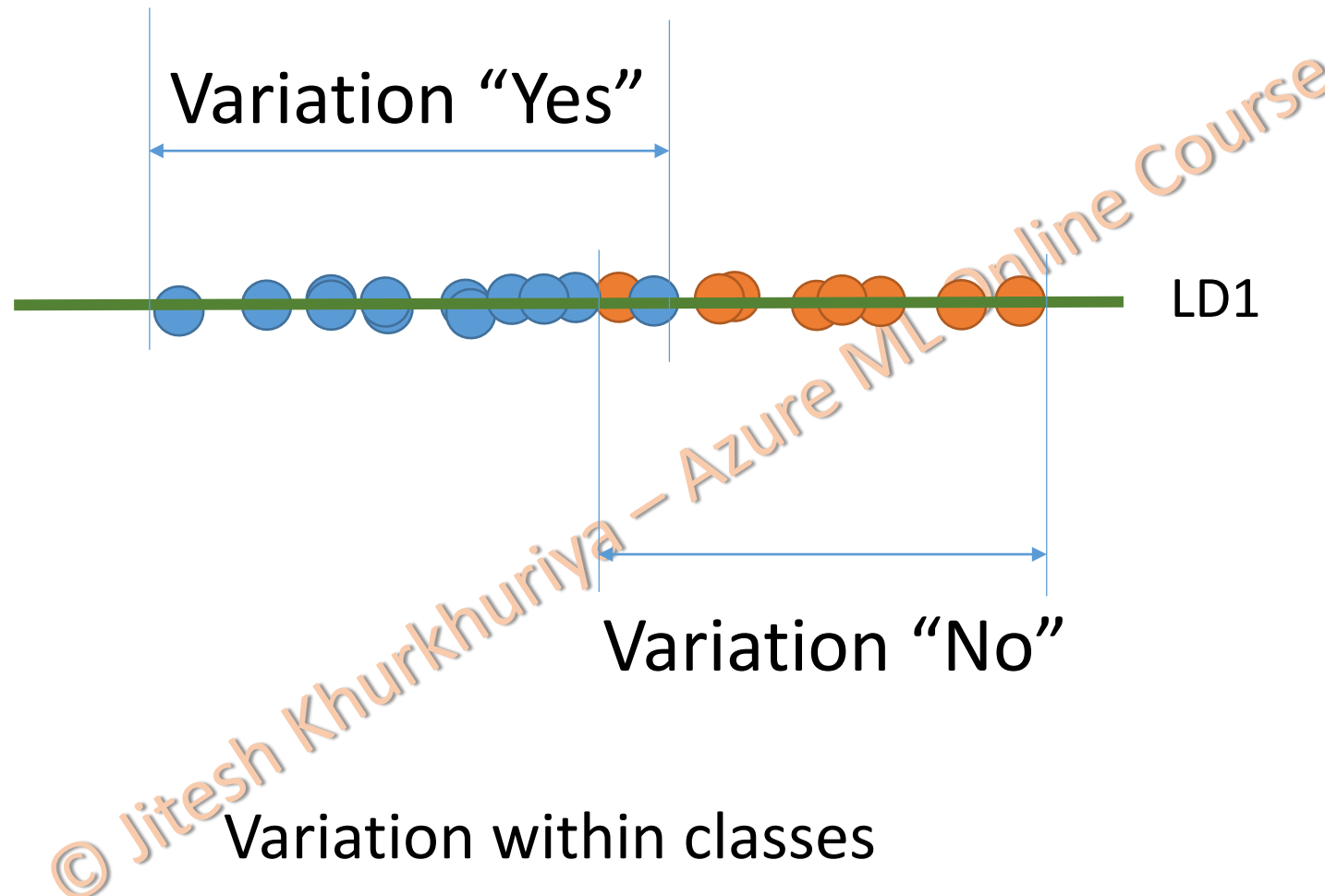
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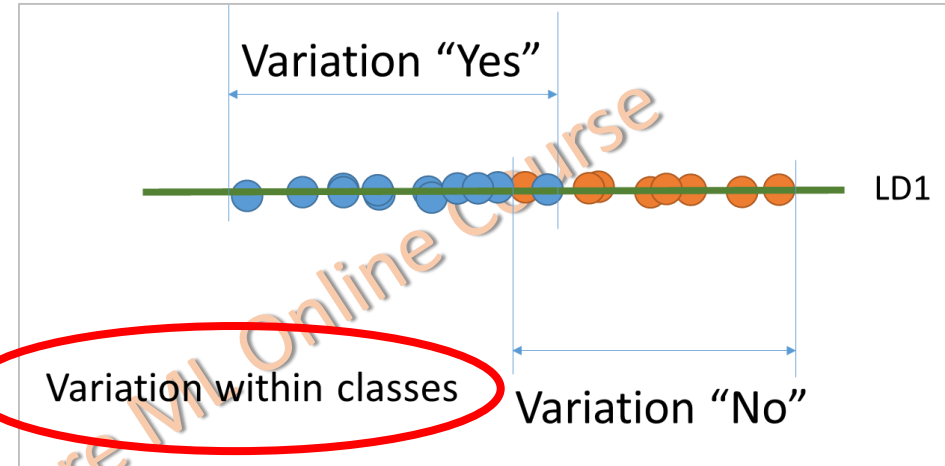
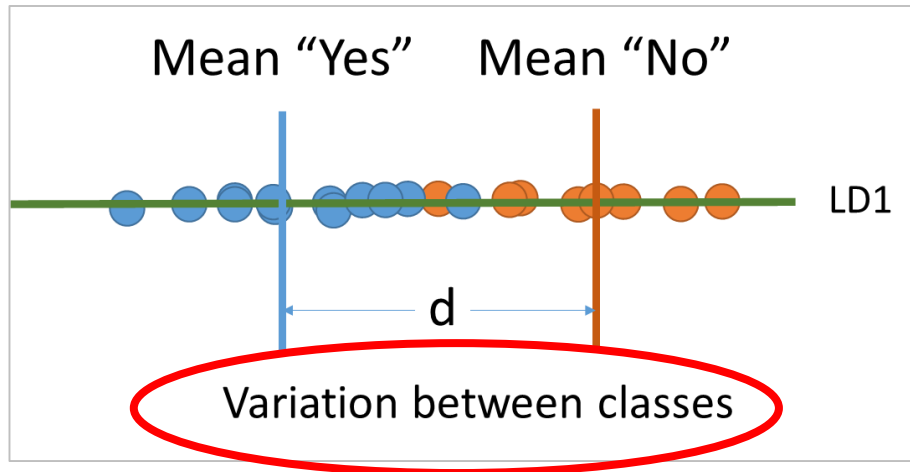
# How LDA is Created?



# How LDA is Created?



# Fisher LDA



$$S - \text{Separation} = \frac{\text{Variation Between Classes}}{\text{Total Variation Within Classes}}$$

# Permutation Feature Importance

# Wrapper Method

X1 X2 X3 X4 ..... Xn

Learning Algorithm

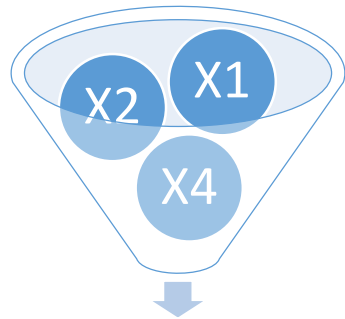
Classification  
Regression

Performance

Accuracy  
Precision  
Root-Mean-Squared

Subset  
X2 X4

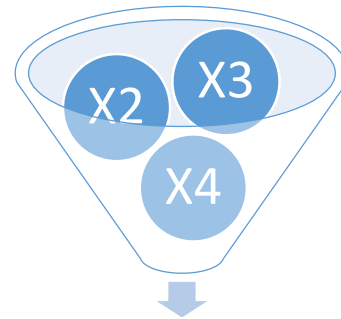
X1 X2 X3 X4 ..... Xn



Subset 1

Training 1

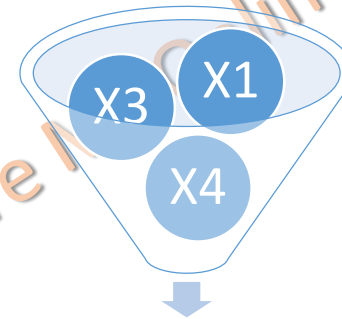
Performance 1



Subset 2

Training 2

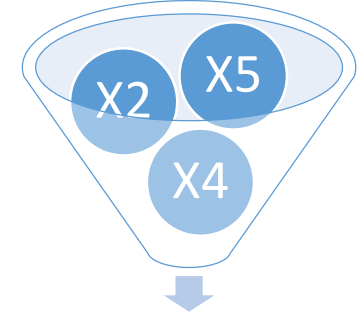
Performance 2



Subset 3

Training 3

Performance 3



Subset 4

Training 4

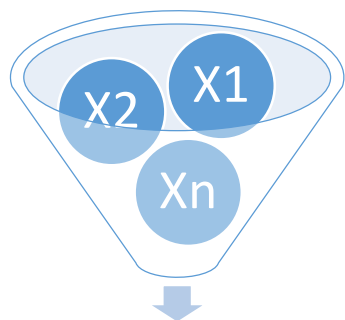
Performance 4

# Permutation Feature Importance

X1 X2 X3 X4 ..... Xn



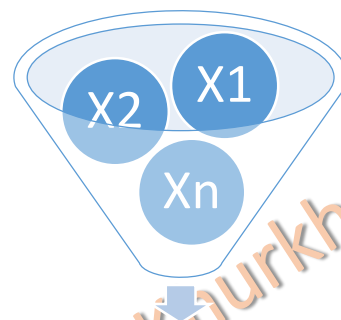
Shuffle



Test Data 1

Performance 1

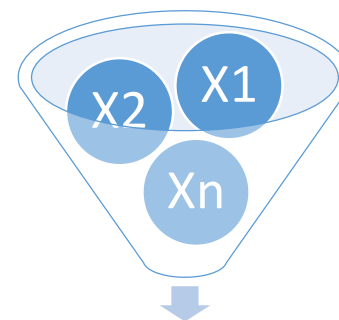
Shuffle



Test Data 2

Performance 2

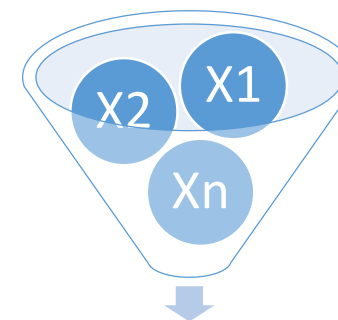
Shuffle



Test Data 3

Performance 3

Shuffle



Test Data 4

Performance 4



# Example using a dataset

Engine Size	HP	Colour	Price
130	111	Black	13495
130	111	Red	16500
152	154	Red	16500
109	102	Black	13950
136	115	Grey	17450
136	110	Black	15250
136	110	Black	17710
136	110	Grey	18920
131	140	Red	23875
108	101	Grey	16430
108	101	Red	16925
164	121	Black	20970
164	121	Red	21105
164	121	Red	24565

Training

Trained Model

# Shuffle and Compare

Trained  
Model

Engine Size	HP	Colour	Price
130	111	Black	13495
130	111	Red	16500
152	154	Red	16500
136	110	Black	17710
136	110	Grey	18920
131	140	Red	23875
108	101	Grey	16430
164	121	Red	24565

# Shuffle and Compare

Trained  
Model

Engine Size	HP	Colour	Price
130	111	Black	13495
130	154	Red	16500
152	111	Red	16500
136	110	Black	17710
136	110	Grey	18920
131	101	Red	23875
108	140	Grey	16430
164	121	Red	24565

Engine Size	HP	Colour	Price	Price Predicted
130	111	Black	13495	
130	154	Red	16500	
152	111	Red	16500	
136	110	Black	17710	
136	110	Grey	18920	
131	101	Red	23875	
108	140	Grey	16430	
164	121	Red	24565	

# Performance Measure

## Classification

- Accuracy
- Precision
- Recall
- Average Log-Loss

## Regression

- Mean Absolute Error
- Root Mean Squared Error
- Relative Absolute Error
- Co-Efficient of Determination
- Relative Squared Error

Thank You and have a Great  
Time..!!