

MACHINE LEARNING PRACTICAL

LOGISTIC REGRESSION



AGENDA

- Data Collection || Reading Data
- Data Exploration
- Data Processing (Separating The Features From The Target)
- Splitting the data to training data & Test data
- Data Scaling (Data Standardization, Normalization)
- Build Model
- Model Evaluation
- Saving Model
- Build Test Method



✓ ABOUT DATA

Parkinson's Disease (PD) is a degenerative neurological disorder marked by decreased dopamine levels in the brain. It manifests itself through a deterioration of movement, including the presence of tremors and stiffness. There is commonly a marked effect on speech, including dysarthria (difficulty articulating sounds), hypophonia (lowered volume), and monotone (reduced pitch range). Additionally, cognitive impairments and changes in mood can occur

✓ SCALE DATA

Separating The Features From The Target

Define X → Input / Attributes of Data

Define Y → Output of Dataset (Result)

75%	25%
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75%	25%
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Train

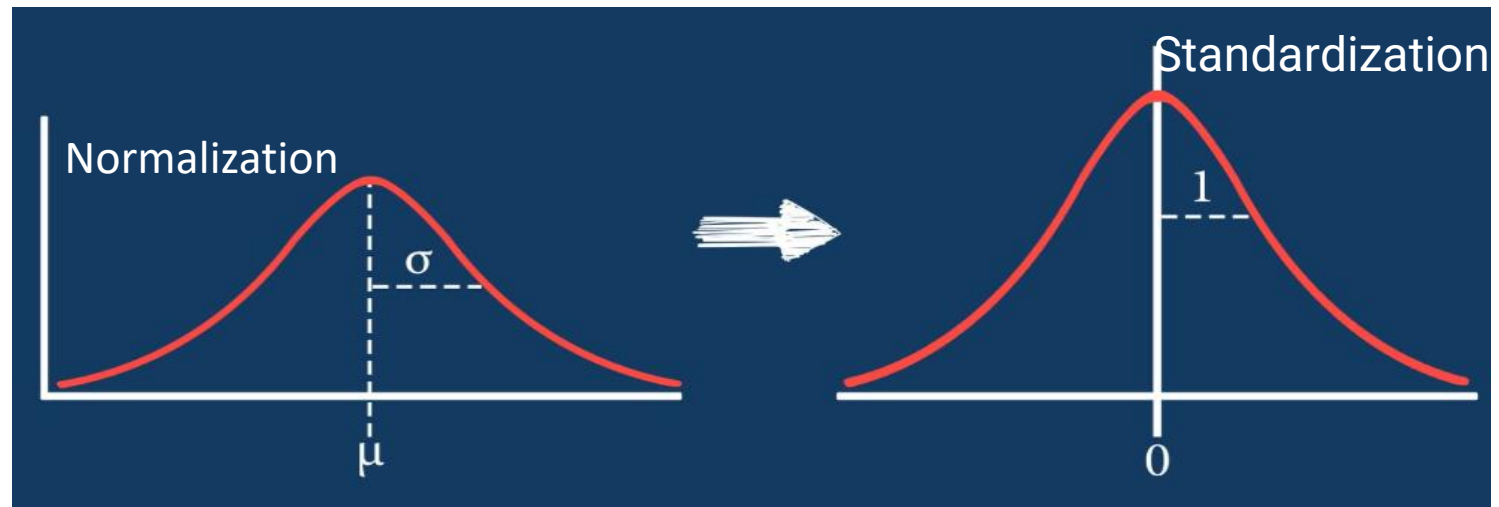
Test

DATA STANDARDIZATION

input (Gap Data) ➔ Run Slow

NORMALIZATION


Data Imbalance




Feature scaling



Normalization


$$X_{new} = \frac{X - X_{min}}{X_{max} - X_{min}}$$

Standardization


$$X' = \frac{X - \text{Mean}}{\text{Standard deviation}}$$

LABEL ENCODING

	name	gender	age	city
a	Abby	F	33	Berlin
b	Ben	M	16	Tokyo
c	Charlie	M	22	Sydney
d	Dave	M	65	York
e	Ella	F	18	Sydney

→

	name	gender	age	city
a	Abby	0	33	0
b	Ben	1	16	2
c	Charlie	1	22	1
d	Dave	1	65	3
e	Ella	0	18	1

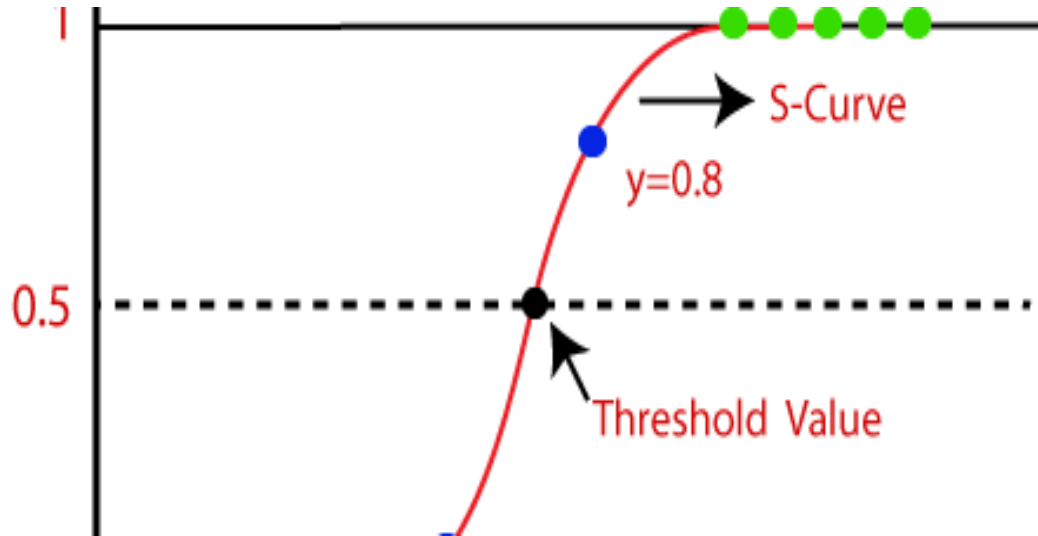
```
df['Gender'].map({'M': 1, 'F': 0})
```

HOT ENCODING

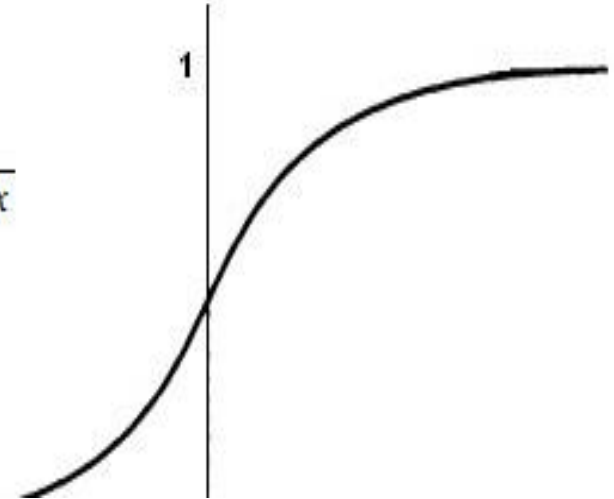
Color	One-hot encoding		
Red	1	0	0
Green	0	1	0
Blue	0	0	1

Distribution Attributes (Nationality(4) 4 Cols)

LOGISTIC REGRESSION



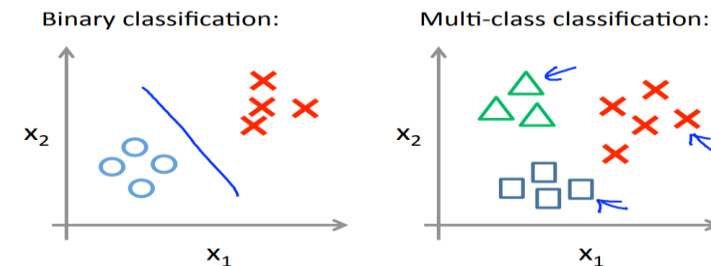
$$f(x) = \frac{1}{1 + e^{-\beta x}}$$



Logistic Function (Sigmoid)

- Logistic regression is used to handle the classification problems.
- Logistic Model Build on Probability , This type of statistical model
- Recall The Sigmoid Function

- Linear Regression (Squashing)
- Classification of (Binary Classifier & Multi Classifier)





SUMMARY

→ Threshold classifier output $h_{\theta}(x)$ at 0.5:

→ If $h_{\theta}(x) \geq 0.5$, predict " $y = 1$ "

If $h_{\theta}(x) < 0.5$, predict " $y = 0$ "

Classification: $y = 0$ or 1

$h_{\theta}(x)$ can be > 1 or < 0

Logistic Regression: $0 \leq h_{\theta}(x) \leq 1$

Classification

THANK YOU !

