

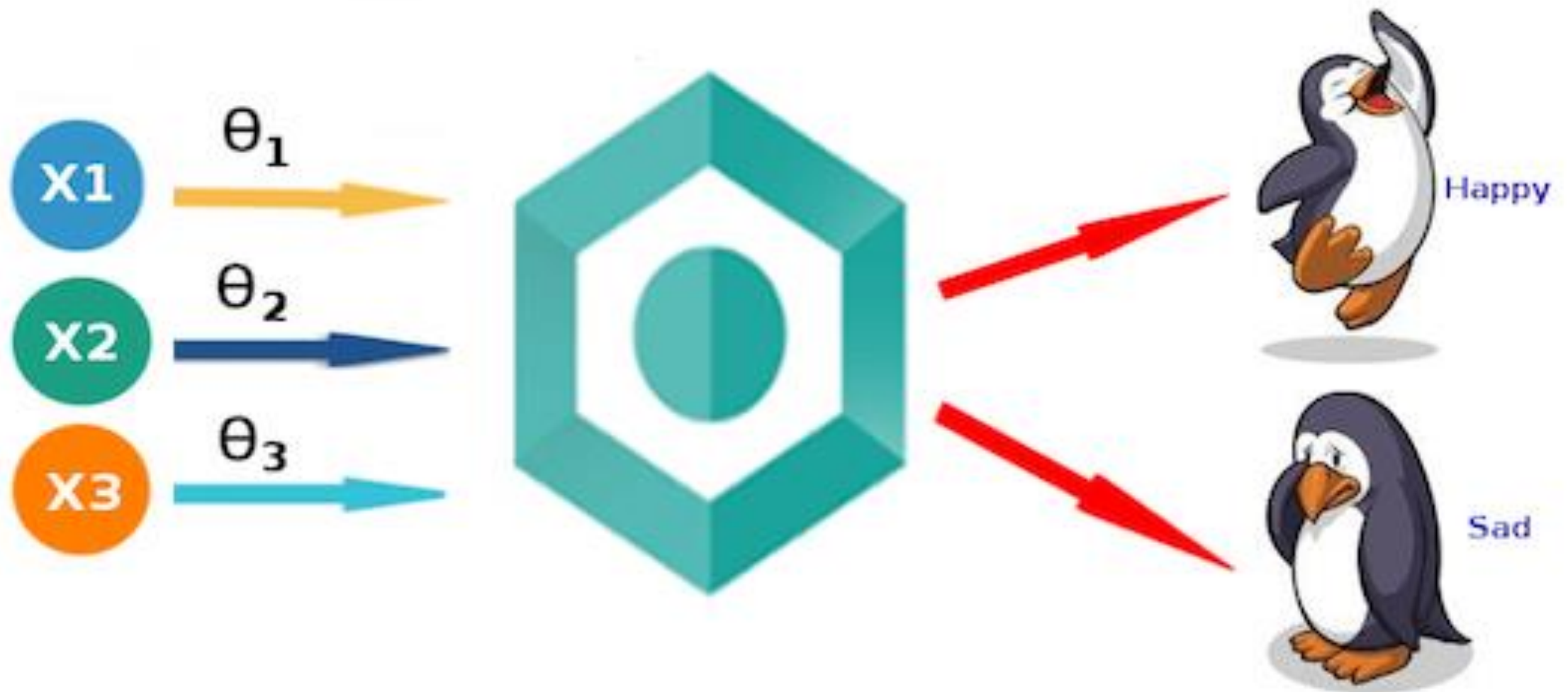
# Practical Machine Learning

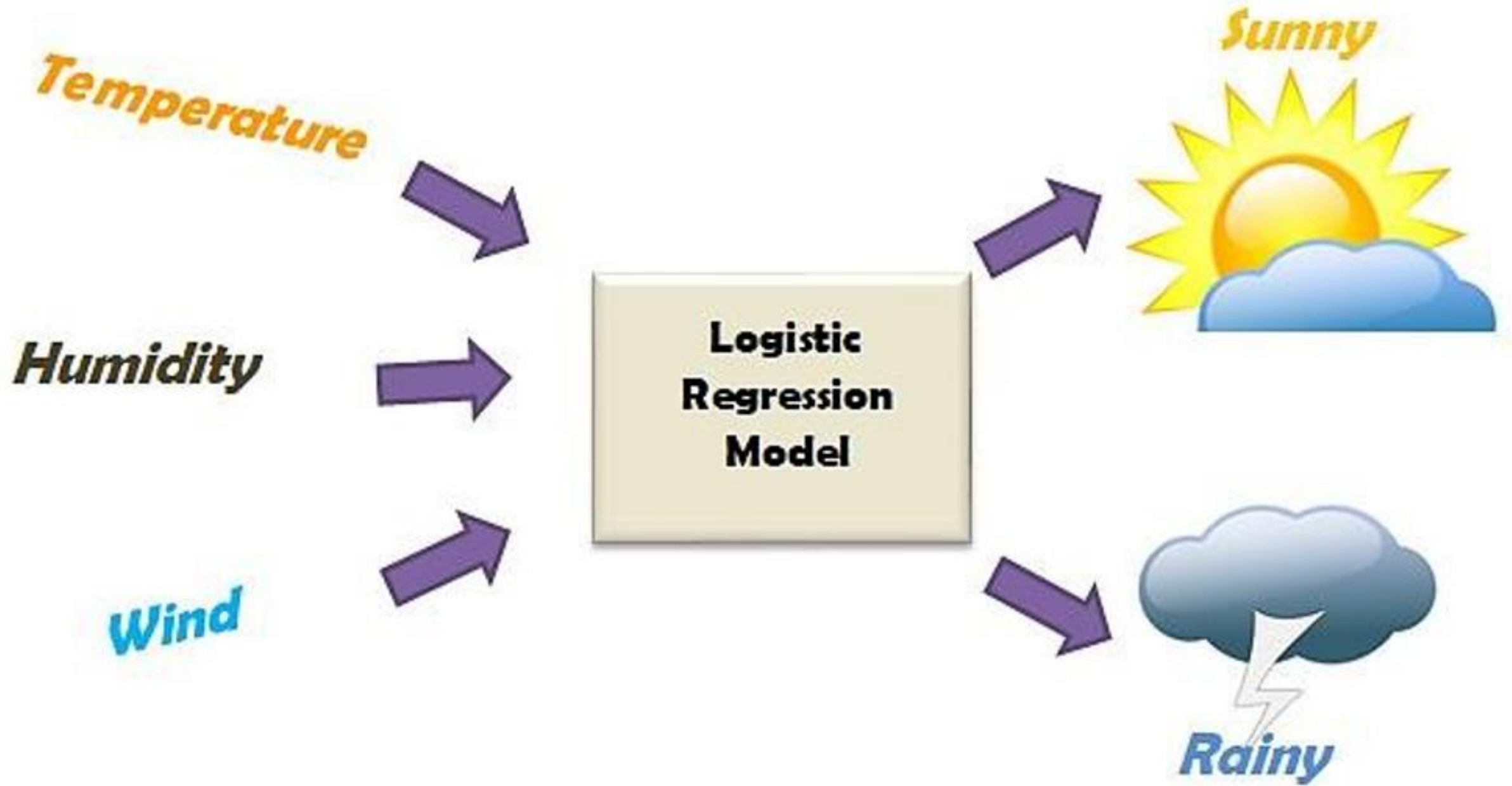
## Day 7: Sep22 DBDA

Kiran Waghmare

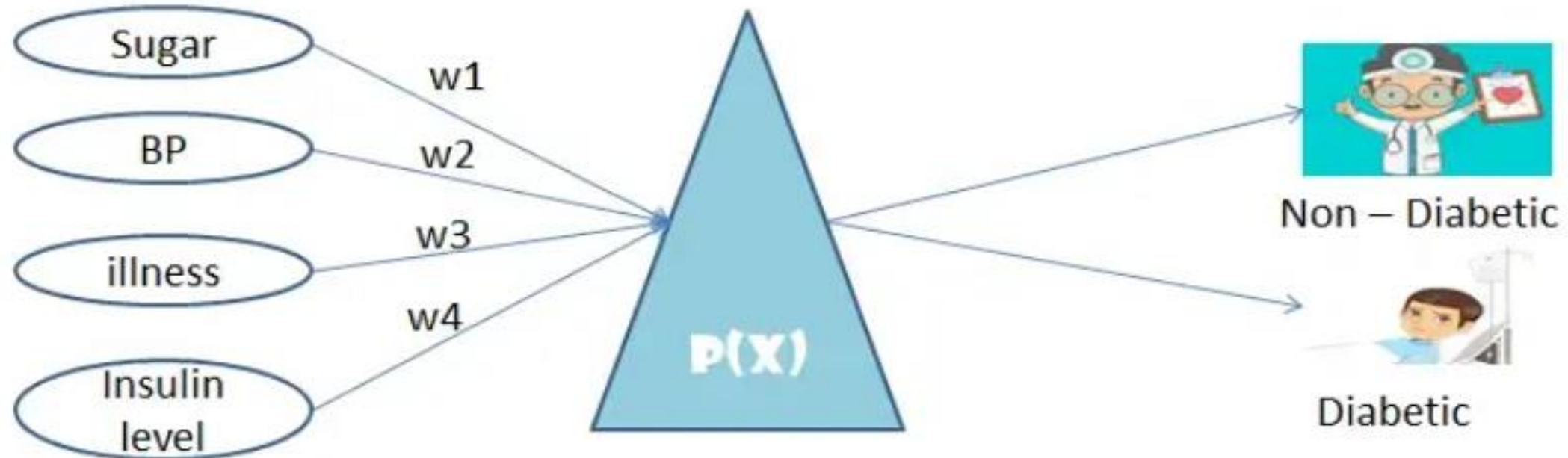
# Agenda

- Logistic Regression
- Classification
- Measures for classification
- KNN





# LOGISTIC REGRESSION MODELLING



$w_1, w_2, w_3, w_4$  – Amount of each individual medical problem

$P(x)$  – Probability Calculation

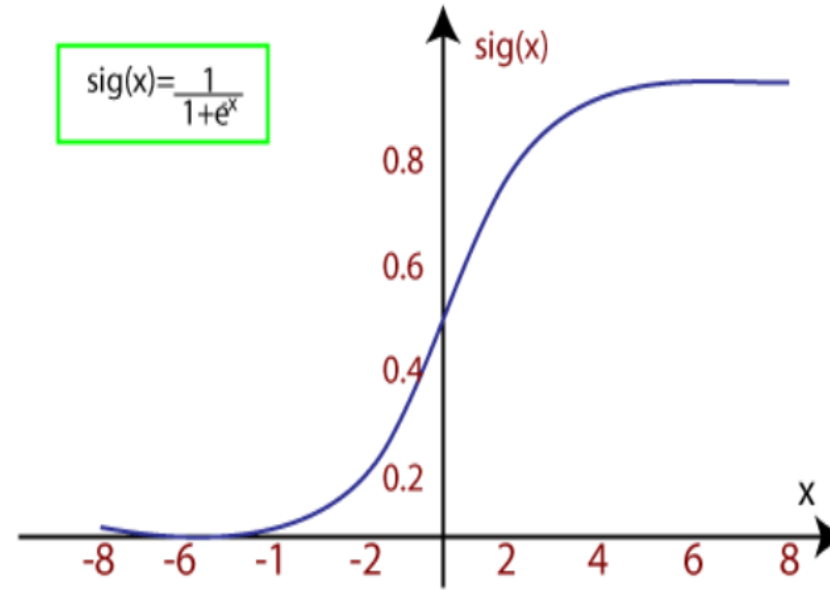
Logistic Regression



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

- $f(x)$  = Output between the 0 and 1 value.
- $x$  = input to the function
- $e$  = base of natural logarithm.

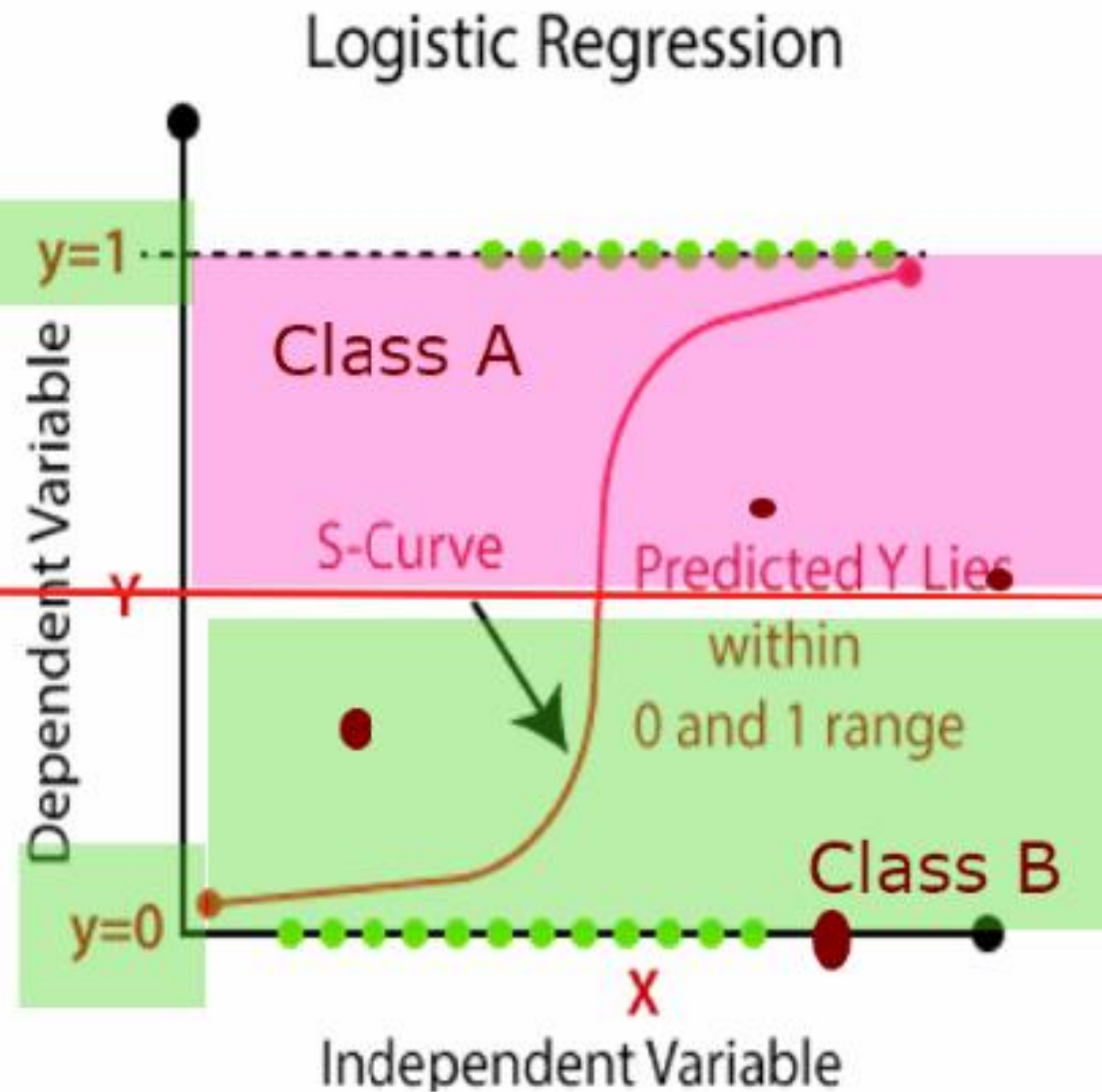
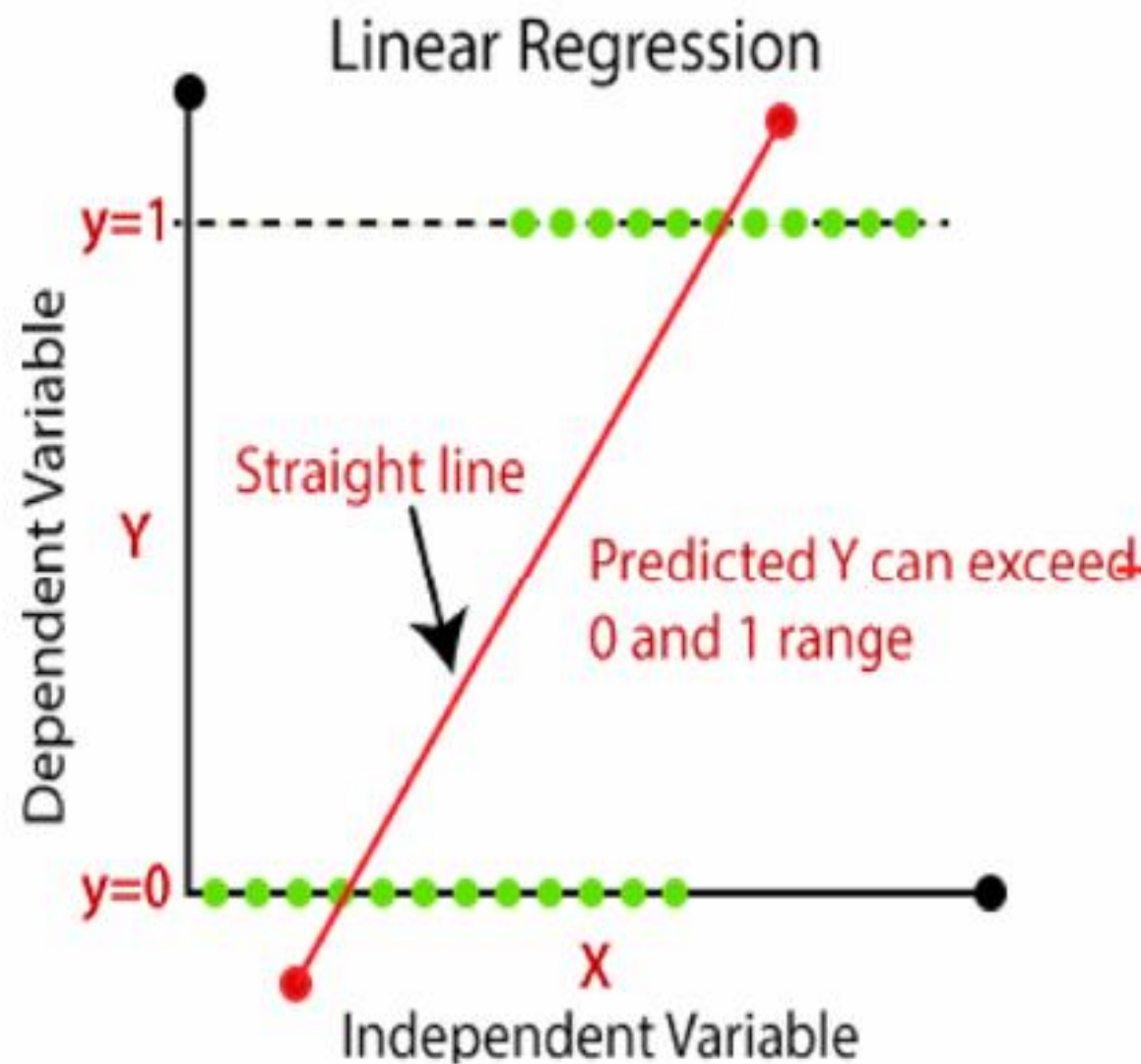
When we provide the input values (data) to the function, it gives the S-curve as follows



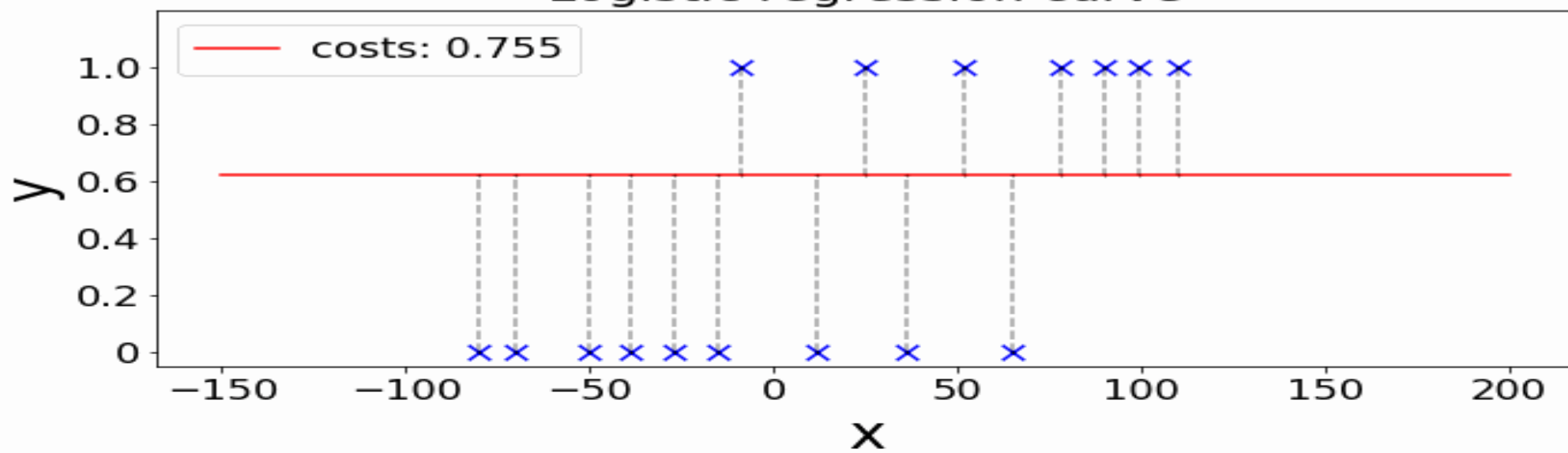
- It uses the concept of threshold levels, values above the threshold level are rounded up to 1, and values below the threshold level are rounded up to 0.

There are three types of logistic regression:

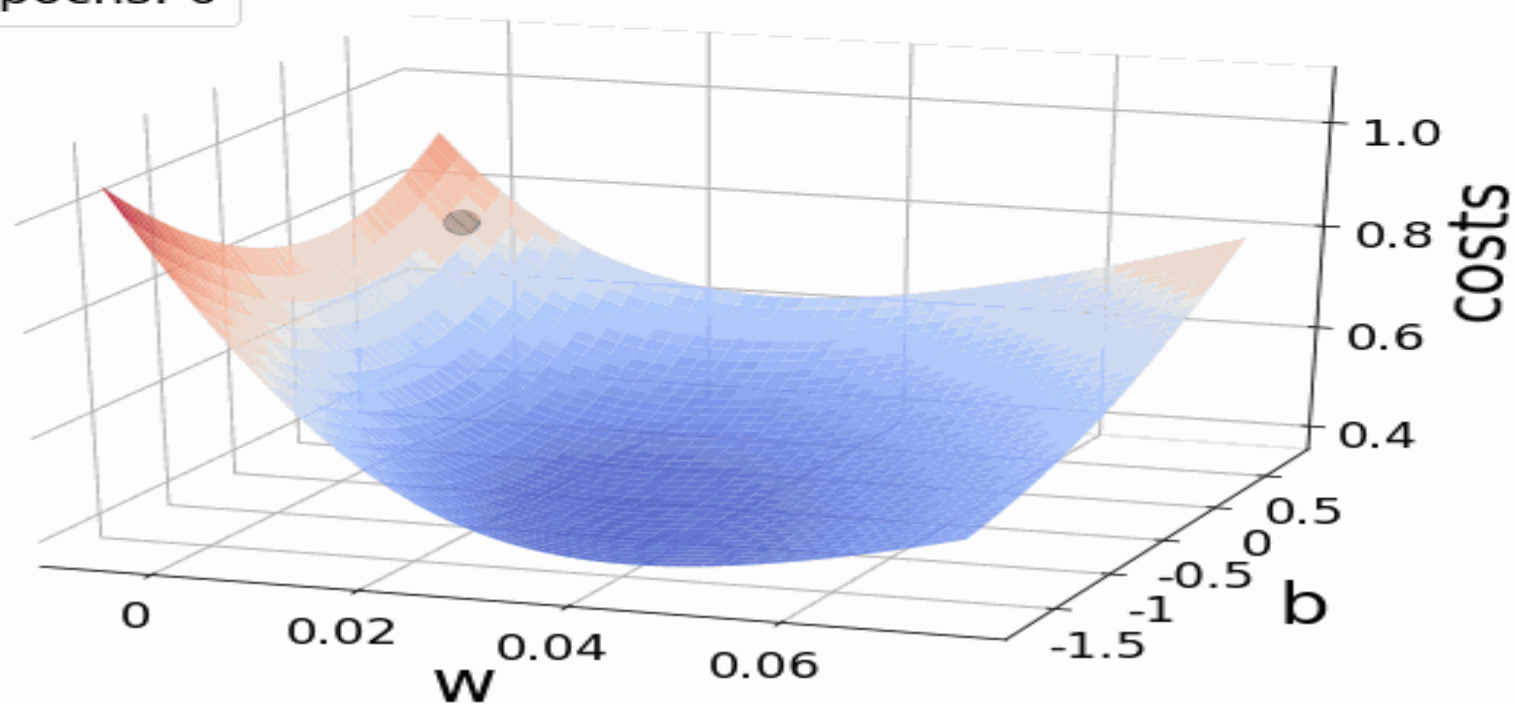
- **Binary(0/1, pass/fail)**
- **Multi(cats, dogs, lions)**
- **Ordinal(low, medium, high)**



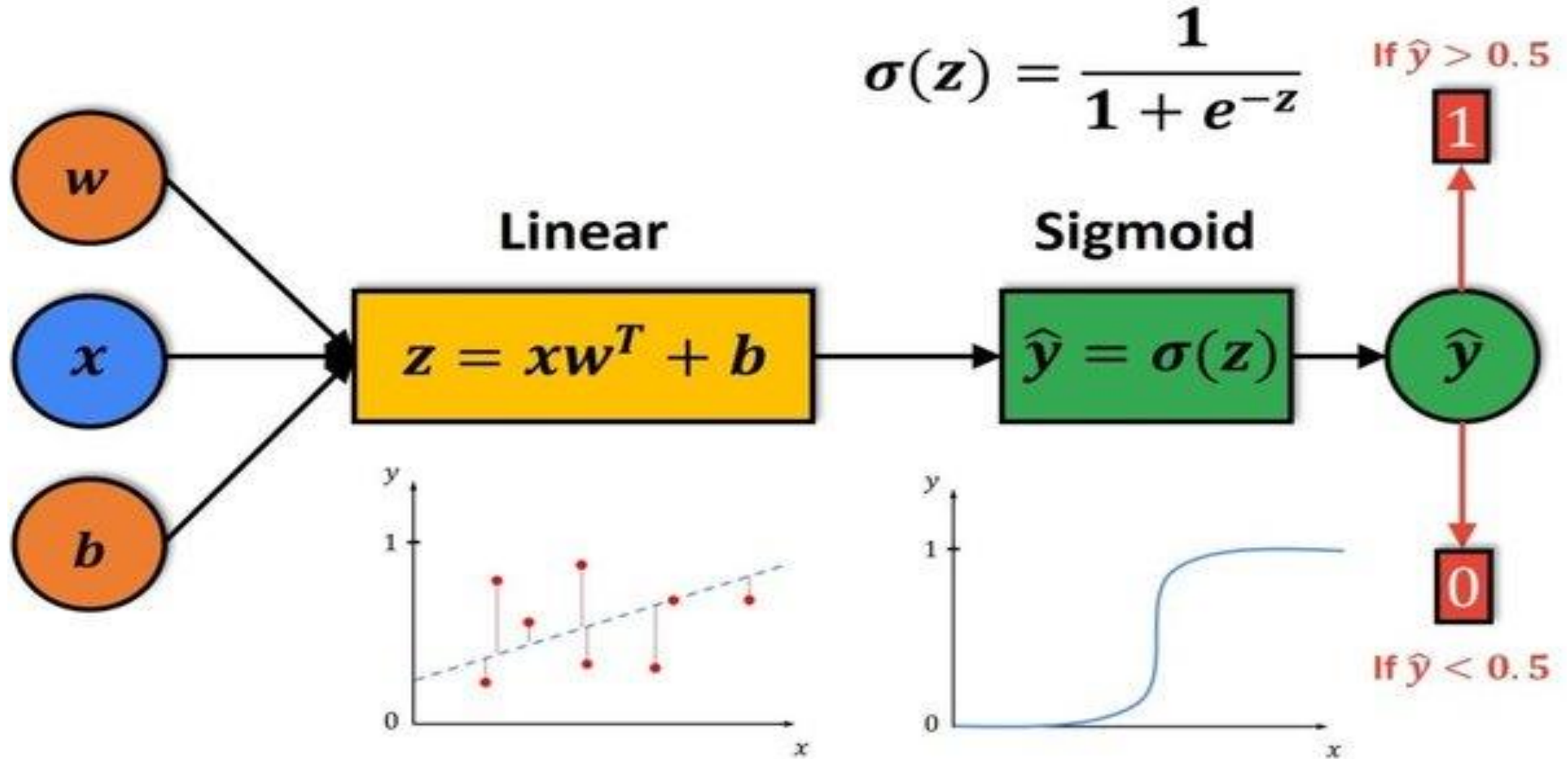
# Logistic regression curve



## epochs: 0









Run



Code



Commit



```
X_test=sc.fit_transform(X_test)
```

```
In [23]: #X_train
```

```
In [13]: #Logistic Model
```

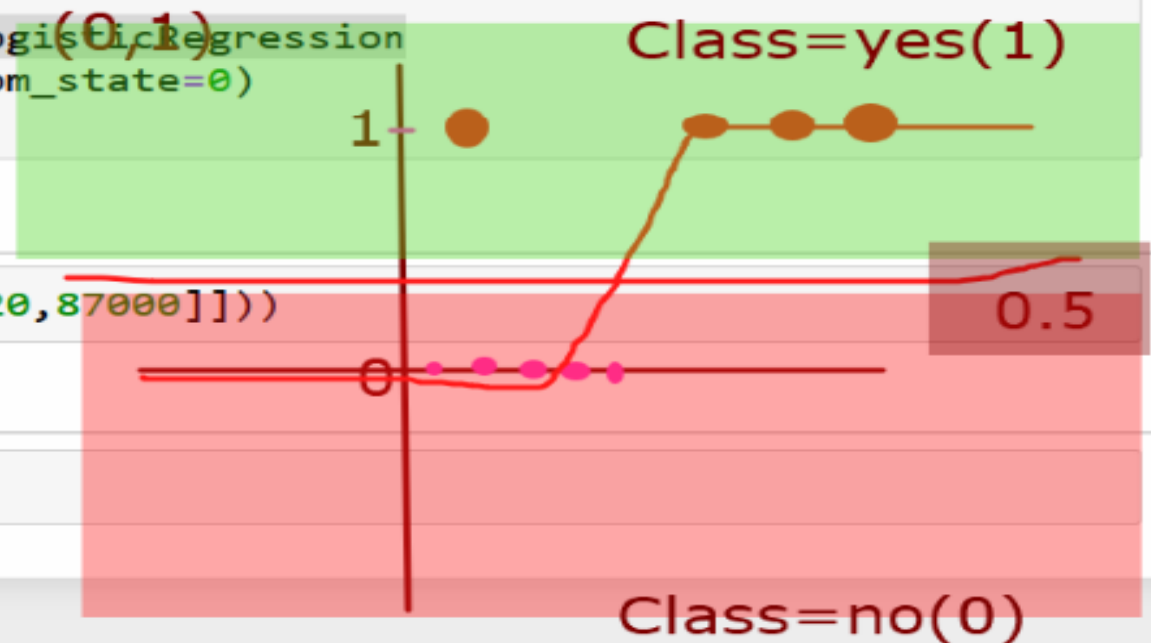
```
from sklearn.linear_model import LogisticRegression  
classifier=LogisticRegression(random_state=0)  
classifier.fit(X_train,y_train)
```

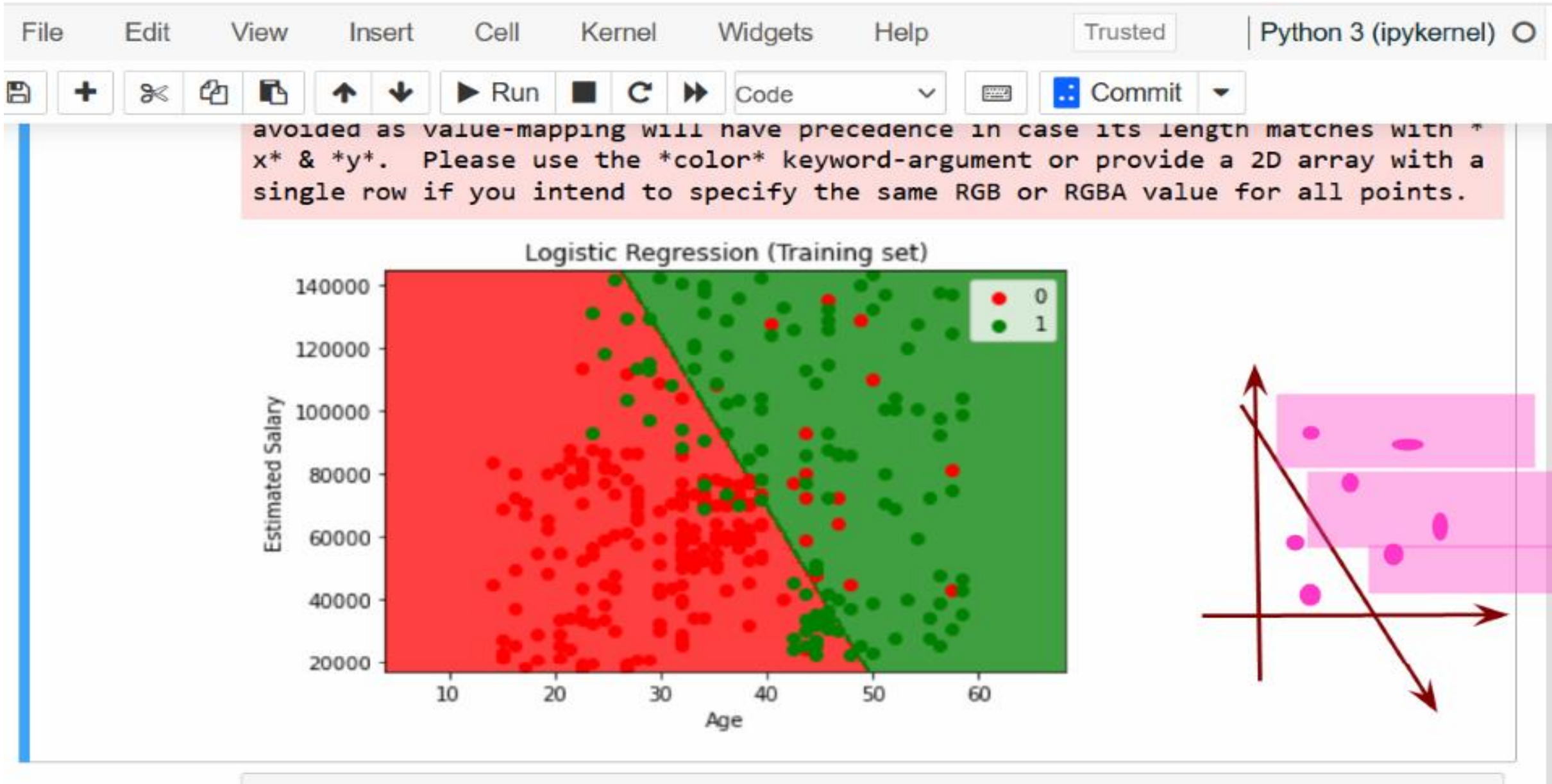
```
Out[13]: LogisticRegression(random_state=0)
```

```
In [24]: classifier.predict(sc.transform([[20,87000]]))
```

```
Out[24]: array([1], dtype=int64)
```

```
In [ ]:
```





```
In [17]: #Confusion Matrix
from sklearn.metrics import confusion_matrix, accuracy_score
cm=confusion_matrix(y_test,y_pred)
cm
```

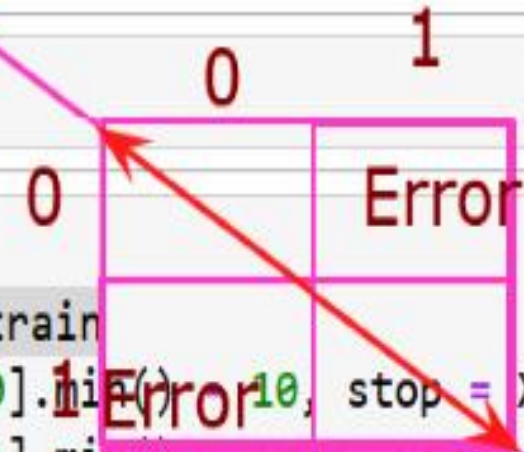
```
Out[17]: array([[52,  6],
               [ 3, 19]], dtype=int64)
```

min=Error

max=Accuracy

In [ ]:

```
In [14]: from matplotlib.colors import ListedColormap
X_set, y_set = sc.inverse_transform(X_train), y_train
X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min(), stop = X_set[:, 0].max() + 1, step = 10),
                     np.arange(start = X_set[:, 1].min() - 1000, stop = X_set[:, 1].max() + 1000, step = 1000))
plt.contourf(X1, X2, classifier.predict(sc.transform(np.array([X1.ravel(), X2.ravel()])).reshape((-1, 2))),
             alpha = 0.75, cmap = ListedColormap(('red', 'green')))
plt.xlim(X1.min(), X1.max())
```





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## Day 7: Classification

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Date: 09/01/2023

Topics:

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- Logistic Regression
- Classification concept

A,B,C,D

A={A+,A,A-}

Actual values

Predicted values

	N	P
N	TN	FP
P	FN	TP

Population = P + N

TP: True Positive  
TN: True Negative  
FP: False Positive  
FN: False Negative

P: positive case  
N: negative case

TP: correct predicted true values  
TN: correctly predicted false values  
FP: wrongly predicted positive values  
FN: wrongly predicted negative values



# Problem Statement

- **Titanic dataset**
- **Explore:** How does each feature relate to whether a person survives/alive?
- Do the EDA in more detail than usual and explain the results!
  - Splitting: 80-20, stratify: y, random\_state = 0
- **Preprocessing:**
  - \* Drop decks
  - \* Fill in the missing value using a simple imputer
  - \* One hot encoding: sex, alone
  - \* Ordinal encoding: class
  - \* Binary encoding: embark town
- **Model selection:**
  - \* Evaluation metrics used: F1\_score
  - Logistic Regression