

MILAD202 - Machine Learning for  
Patterns Discovery  
Model Practical

Monika P  
192425059  
B.Tech-AMM  
SET-8

1. Linear Regression

Aim :- Predict Sales using Advertising

Algorithm :-

- \* Load data
- \* Show first rows, state, datatype.
- \* Plot Scatter
- \* Check & report nulls with mode
- \* Split data into train / test
- \* Train Linear Regression model
- \* Predict test values
- \* Print predictions

Code :-

```
import pandas as pd
from sklearn.model_selection import
train_test_split
from sklearn.linear_model import
LinearRegression
df = pd.DataFrame({
    'Sales': [200, 250, 300, 350, 400],
    'Adv': [20, 25, 30, 35, 40]
})
print(df.head())
print(df.describe())
```

```

print(df.dtypes)
print(df.isnull().sum())
df = df.fillna(df.mode().iloc[0])
x = df[['Adv']]
y = df['Sales']
x_train, x_test, y_train, y_test = train_test_split
(x, y, test_size = 0.2)

model = LinearRegression()
model.fit(x_train, y_train)
print("Prediction:", model.predict(x_test))

```

Output:-

Prediction: [345.0]

Result:- Successfully found the Prediction is 345.0

## 2. Candidate - Elimination Algorithm.

Aim:- Find all consistent hypotheses

Algorithm:-

- \* Initialize  $S$  = most specific,  $G$  = most general.
- \* for each example:
  - If positive  $\rightarrow$  generalize  $S$  and remove inconsistent  $G$ .
  - If negative  $\rightarrow$  specialize  $G$  to exclude the example
- \* Remove inconsistent hypotheses
- \* Output  $S$  and  $G$ .

Code:-

```
import pandas as pd
```

```
data = [
```

```
    [Big, Red, Circle, No]
```

```
    [small, Red, Triangle, No]
```

```
    [small, Red, Circle, Yes]
```

```
    [Big, Blue, Circle, No]
```

```
]
```

```
df = pd.DataFrame(data, columns = [size, color,  
                                   shape, class])
```

```
S = [0, 0, 0]
```

```
G = [[?, ?, ?]]
```

```
for i, row in df.iterrows():
```

```
    h = row[:-1]; c = row[-1]
```

```
    if c == Yes:
```

```
        for j in range(3):
```

```
            if S[j] == 0, S[j] = h[j]
```

```
            elif S[j] != h : S[j] = ?
```

```
        G = [g for g in G if all (g[k] == ? or
```

```
g[*] == h[k] for k in range(3))]
```

```
    else:
```

```
        new G = []
```

```
        for g in G:
```

```
            for j in range(3):
```

```
                g[j] == ?
```

```
                new = list(g); new[j] = S[j]
```

```
                newG.append(new)
```

```
        G = newG
```

```
print (S, G)
print ("G", G)
```

Output:-

S: [small, ?, circle]

G: [[?, Red, ?], [?, Blue, ?]]

Result:- Successfully find the output.

### 3 # Logistic Regression.

Aim:- Predict "Byy"

Algorithm:-

- \* Load dataset
- \* Split into train / test
- \* Train Logistic Regression model on training data
- \* Predict on data test
- \* Compute accuracy
- \* Print predictions.

Code:-

```
from sklearn.linear_model import  
Logistic Regression
```

```
from sklearn.model_selection import  
train_test_split  
import pandas as pd.
```

```
df = pd.DataFrame({  
    Age : [20, 30, 40, 50, 60]  
    Income : [20, 30, 40, 50, 60]  
    Buy : [0, 0, 1, 1, 1]  
})
```

```
x = df[['Age', 'Income']]
```

```
y = df['Buy']
```

```
xtr, xte, ytr, yte = train_test_split(x, y, test_size  
= 0.2)
```

```
model = LogisticRegression()
```

```
model.fit(xtr, ytr)
```

```
print(Pred, model.predict(xte))
```

Output:-

Pred: [1]

Result:- Successfully found the predict [1]

#### 4. Naïve Bayes Algorithm

Aim:- Predict "Pass" based on study & sports.

Algorithm:-

- \* load dataset
- \* Split into train / test
- \* Train Naïve Bayes model
- \* Predict test labels
- \* Calculate accuracy
- \* Print predictions.

Code:-

```
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import train_test_split
import pandas as pd

df = pd.DataFrame({
    'Study': [2, 4, 6, 8, 10],
    'Sleep': [9, 8, 7, 6, 5],
    'Pass': [0, 0, 1, 1, 1]
})

X = df[['Study', 'Sleep']]
Y = df['Pass']
Xtr, Xte, Ytr, Yte = train_test_split(X, Y, test_size=0.2)

model = GaussianNB()
model.fit(Xtr, Ytr)
print("Pred", model.predict(Xte))
```

Output  
Pred: [1]

Result:- Successfully found the predict [1]