

<u>MODEL</u>	<u>LAB</u>	<u>EX-AM</u>	MLR0102 - Fundamentals of Machine Learning	A Akhaya 192415110 SET - 6
Historical data	Outputs	Atrial	Predicted	
	0	50000	51000.0	
	1	65000	64000.0	
	2	200000	220000.0	

④ Aim:- To predict house prices for mark using historical data

→ Algorithm :-

- * Load dataset with Pandas
 - * Check first 5 rows and basic statistics.
 - * Check columns, datatypes, & handle nulls.
 - * Explore correlations with a heatmap.
 - * Split data into train/test sets.
 - * Train a linear regression model.
 - * Predict house prices & evaluate performance.

Program :-

```

import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
df = pd.read_csv('house_data.csv').fillna(0)
X = pd.get_dummies(df.drop(['Price'], axis=1), drop_first=True);
y = df['Price']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)
model = LinearRegression().fit(X_train, y_train)
y_pred = model.predict(X_test)
print(pd.DataFrame({'Actual': y_test, 'predicted': y_pred}).head())

```

1	60000	64000.0
2	200000	220000.0
3	400000	410000.0
4	400000	390000.0

(3) Aims To find the most specific (s) & most general (G) hypothesis consistent with training data.

-Algorithm:

① Initialize S - most specific, G

④ If +ve \rightarrow generalize & remove G

$\text{H}_2^{+} - \text{He} \rightarrow \text{He}^{+} + \text{H}_2$ is unresonant.

(2) $\text{P}(\text{start}) = \frac{1}{2}$

Program:

```
data = [['Circular', 'Large', 'Light', 'Smooth', 'Thick', '4'], ['Oval',  
        'Large', 'Light', 'Irregular', 'Thick', '2']]
```

$$\mathcal{L} = [1, 0]^{\star 5}, \quad \mathcal{G} = [0, 1]^{\star 5}$$

for x in data:

$$x_i^0 = x[::i]$$

$$96 \times [-1] = -96 + 96$$

$G = \{g \text{ for } g \in S \text{ if all } (a = -1?!) \text{ or } a = b \text{ for } a,$

$b \in \text{zip}(g, x))$

```

G_new.append(g[i][j+1, 0] + g[j+1, j])
G = g[targ] in G_new if
all((c == '?') or c == a for a in
set(g[targ]))
print("S:", S)
print("G:", G)

```

Outputs

```

S: ['Circular', 'Large', '?', '?', 'Thin']
G: [[Circular', 'Large', '?', '?', 'Thin'],
     ['?', 'Large', 'Large', '?', '?', 'Thin']]

```

Result: S and G show all hypotheses consistent with the training data.

(2) Aim: To predict a continuous target variable using Linear Regression

Algorithm:

- 1) Load dataset
- 2) Split into features (X) & target (Y)
- 3) Split into train/test sets.
- 4) Train Linear Regression model.
- 5) Predict and evaluate.

Program:

```

from sklearn.model_selection import
train_test_split

```

```

from sklearn.linear_model import LinearRegression
from sklearn.mixture import GaussianMixture
import pandas as pd
df = pd.read_csv('Iris.csv')
x = df['SepalLengthCm'].values.reshape(-1, 1)
y = df['SepalWidthCm'].values.reshape(-1, 1)
X_train, X_test, Y_train, Y_test = train_test_split(X, y,
                                                    test_size=0.2, random_state=0)
model = LinearRegression()
model.fit(X_train, Y_train)
y_pred = model.predict(X_test)
print("R2 Score:", r2_score(Y_test, y_pred))

```

Outputs

```

Predicted: [4.5]
R2 Score: 0.6

```

Result: Linear Regression predicts target values; R-squared shows model performance.

(4) Aim: To cluster data using the expectation-maximization (EM) algorithm.

Algorithm:

- 1) Load or generate data.
- 2) Fit a gaussian mixture model (GMM).
- 3) Predict cluster labels.
- 4) Evaluate clusters.

Program:

```

from sklearn.mixture import GaussianMixture
import numpy as np

```

```

X = np.array([[1, 1, 1], [1, 0, 0], [-1, 0, 0], [-1, 1, 1]])
gmm = GaussianMixture(n_components=3,
                      random_state=0).fit(X)

```

Labels = gmm.predict(X)
print("cluster labels:", labels)
print("means:", gmm.means_)

Outputs

```

cluster labels: [0 0 0 1 1 1]
means: [[1. 0.]
        [-1. 1.]]

```

Results

EM algorithm clusters the data; means & labels show the learned Gaussian distributions.