

## Week - 2



1. Get the data

Downloading - data =

import os

import tarfile

from six.moves import urllib

DOWNLOAD\_ROOT = "http://raw.githubusercontent.com/  
agron/handson-ml2/master/"

HOUSING\_PATH = os.path.join("data", "01")

HOUSING\_URL = download\_root + "datasets/  
housing/housing.tgz"

def fetch\_housing\_data(housing\_url=  
Housing\_URL, housing\_path=  
Housing\_path)

os.makedirs(name=housing\_path,  
exist\_ok=True)

tgz\_path = os.path.join(housing\_path,  
"housing.tgz")

with requests.get(url=housing\_path,  
"housing.tgz") as:

housing\_tgz.close()

→ Create test set

```
import numpy as np
```

```
def split_train_test(data, test_ratio=0.2):
```

```
    shuffled_indices = np.random.permutation(len(data))
```

```
    test_set_size = int(len(data) * test_ratio)
```

```
    test_indices = shuffled_indices[:test_set_size]
```

```
    train_indices = shuffled_indices[test_set_size:]
```

```
    return data.iloc[train_indices],  
           data.iloc[test_indices]
```

⇒ Discovers & visualize Data to gain insights

~~show strat\_train\_set.shape, strat\_test\_set.shape~~

```
Strat_test_set.reset_index().to_feather  
(fname='data/01/strat-test-set.f')
```

```
Housing = strat_train_set.copy();
```

```
Housing.shape,
```

Prepare Data for Machine Learning Model

```
housing = strat_train_set.drop(["median_house_value"], axis=1)
```

```
housing_labels = strat_train_set["median_house_value"].copy()
```

housing.shape, housing\_label.shape

Select & Train Model

```
from sklearn.linear_model import LinearRegression
```

```
lin-reg = linearRegression()
```

```
lin-reg.fit(X=housing_prepared,  
y=housing_labels),
```

Aug 11/24



## Week -3

### Multiple Linear Regression

```
import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as sns  
from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LinearRegression
```

⇒ Import data

```
df_start = pd.read_csv('content/  
-startups.csv')  
df_start.head()
```

⇒ Describe

```
df_start.describe()
```

⇒ Distribution

```
plt.title('Profit Distribution Plot')  
sns.distplot(df_start['Profit'])  
plt.show()
```

Relationship b/w Profit & R&D Spend

plt.scatter(df\_start['R&D Spend'], df\_start['Profit'], color='lightcoral')

plt.title('Profit vs R&D Spend')

plt.xlabel('R&D Spend')

plt.ylabel('Profit')

plt.show()

Train Model

regressor = LinearRegression()

regressor.fit(x\_train, y\_train)

Predict Results

y\_pred = regressor.predict(x\_test)

Compare predictions

np.set\_printoptions(precision=2)

result = np.concatenate((y\_pred,

reshape(len(y\_pred), 1), y\_test,

reshape(len(y\_test), 1)), 1)

result.

Mulu

## Lab - 5

```
import pandas as pd  
from matplotlib import pyplot as plt  
%matplotlib inline  
  
df = pd.read_csv("insurance_data.csv")  
df.head()  
  
plt.scatter(df.age, df.bought_insurance,  
            marker='+', color='r')  
  
from sklearn.model_selection import  
train_test_split  
  
X_train, X_test, y_train, y_test =  
train_test_split(df[['age']], df.bought_  
insurance, train_size=0.8)  
  
from sklearn.linear_model  
import LogisticRegression  
  
model = LogisticRegression()  
  
model.fit(X_train, y_train)  
  
y_predicted = model.predict(X_test)  
  
model.predict_proba(X_test)
```

model.score - proba(x-test)

\* y-predicted

x-test

model.coef -

model.intercept -

import math

def sigmoid(n):  
    return 1 / (1 + math.exp(-n))

def prediction\_function(age):

$$z = 0.042 * \text{age} - 1.53$$

y = sigmoid(z)

return y

prediction\_function(age)

age = 43

prediction\_function(age)

0.568565299077705

Day 25/11/24

## Lab - 4

### Program - 4)

#### Linear Regression

```
import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as sns  
df = start.head()  
df.describe()
```

Distribution

```
plt.title('profit Distribution plot')  
sns.distplot(df['Profit'])  
plt.show()
```

split Data

```
y = df['Profit'].values  
X = df[['Adress', 'Age', 'Area', 'Expenditure', 'Gender', 'Income', 'Marital Status', 'Occupation', 'Profession', 'Salary', 'Spouse Income']]  
X_train, X_test, Y_train, Y_test = train_test_split(X, y, test_size=0.2, random_state=0)
```

X-train, X-test, Ytrain, Y-test

```
train_test_split(X, y, test_size=0.2,  
random_state=0)
```

date 6-7

Program 6-7

```
import pandas as pd  
import matplotlib.pyplot as plt  
from sklearn.model_selection import  
train_test_split  
from sklearn.preprocessing import Label  
Encoder  
from sklearn.neighbors import KNeighborsClassifier  
from sklearn.svm import SVC  
data = pd.read_csv('iris.csv')  
print(data.head())  
plt.figure(figsize=(10, 6))  
plt.title('Scatter plot of Iris Data')  
plt.show()  
X_train, X_test, y_train, y_test =
```

train\_test\_split(x, y, test\_size=0.2)

random\_state=52)

obj 30/6/24

## Program-8

### ANN

```
import numpy as np
```

```
x = np.array([2, 9], [1, 5], [3, 6])  
dtype = float)
```

```
y = np.array([92], [86], [89]) dtype = float)
```

```
x = x / np.array(x - axis = 0)
```

```
y = y / 100
```

```
epoch = 5000
```

```
lr = 0.1
```

input layer - neurons = 2

hidden layer - neurons = 3

output neurons = 1

```
wh = np.random.uniform(size =
```

(input layer - neurons,  
hidden layer - neurons))

```
bh = np.random.uniform(size = (1, hidden  
layer, neurons))
```

```
bout = np.random.uniform(size =
```

(1, output, neurons))

```
bout = np.random.uniform(size = (1, output  
- neurons))
```

```
print ("Input \n"+ str(x))  
print ("Actual Output \n", str(y))  
print ("predicted Output \n", output)
```

Input:

[0.666667 1.  
0.3333 0.55556]  
P1. 0.666777]

## Program - 99

### Random Forest

```
import pandas as pd
```

```
from sklearn.model_selection import  
train_test_split
```

```
from sklearn.ensemble import  
RandomForestClassifier
```

```
from sklearn.datasets import load_iris
```

```
from sklearn.model_selection  
import train_test_split
```

```
iris_data = pd.read_csv('Iris.csv')  
iris_data.head()
```

X\_train, X\_test, Y\_train, Y\_test  
= train\_test\_split (X, Y, test\_size = 0.4,  
random\_state = 42)

```
rf_classifier = RandomForestClassifier  
(n_estimators =  
100, random_state = 42)
```

```
clf_classifier = rf_classifier.fit (X_train, Y_train)
```

date 30/5/24

## Program - 9 b

Date \_\_\_\_\_  
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```
iris · data = pd.read_csv('Iris.csv')  
iris · data · drop ('species' - axis'=1)
```

```
iris · data · head()
```

```
n = iris · data · drop ('species', axis=1)
```

```
y = iris · data · info()
```

```
X-train, X-test, y-train, y-test,  
train-test - split (x, y, test-size=0.4  
random-state=42)
```

```
myLogregModel = LogisticRegression()
```

```
at adaboost = AdaBoat Classifiers(  
X-train, y-train).
```

~~accuracy = accuracy-score (y-test,  
y-pred).~~

(a) digit dataset

100% accuracy

## Program 10-a

(Breast Cancer - PCA).

```
import seaborn as sns  
import matplotlib.pyplot as plt  
import pandas as pd  
import numpy as np  
%matplotlib inline
```

```
from sklearn.datasets import  
load_breast_cancer
```

```
cancer = load_breast_cancer()
```

```
cancer.keys()
```

```
print(cancer['DESCR'])
```

```
df = pd.DataFrame(cancer['data'],  
columns=cancer['feature_names'])
```

```
df.head()
```

```
cancer['target_names']
```

```
scaler = StandardScaler()
```

```
scaler.fit(df)
```

```
n_pca = pca.transform(scaled_data)
```

```
n_pca.shape
```

## Program Lab 11

K means - Clustering of Iriess

```
import os  
import sys  
import tempfile import NamedTemporaryFile  
import zipfile import ZipFile  
import tarfile  
import shutil
```

CHUNK\_SIZE = 40960  
DATA\_SOURCE\_MAPPINGS = 'iris - flower - dataset'

try:  
 os.symlink(KAGGLE\_INPUT\_PATH, os.path.  
 join("../", "input"), target\_is\_  
 directory = True)

except FileExistsError:  
 pass

try:  
 os.symlink(KAGGLE\_Working\_Path,  
 os.path.join("../", "working"),  
 target\_is\_directory = True)

except FileExistsError:  
 pass

dday 30/6/24

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