```
# import libraries
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.metrics import r2_score, mean_squared_error
from math import sqrt
# this allows plots to appear directly in the notebook
%matplotlib inline
# read data into a DataFrame
data = pd.read_csv('Advertising.csv', index_col=0)
data.head()
data.columns = ['TV','Sales']
# print the shape of the DataFrame
data.shape
# visualize the relationship between the features and the response using scatterplots
#data.plot(kind='scatter', x='TV', y='Sales')
plt.scatter(data['TV'], data['Sales'])
# create X and y
#taking only one variable for now
X = data[['TV']]
Χ
y = data.Sales
# follow the usual sklearn pattern: import, instantiate, fit
from sklearn.linear model import LinearRegression
lm = LinearRegression()
Im.fit(X, y)
# print intercept and coefficients
print(lm.intercept_)
print(lm.coef_)
# manually calculate the prediction using above slope and intercept in b0+b1*x
7.032594 + 0.047537*50
# you have to create a DataFrame since the Statsmodels formula interface expects it
X new = pd.DataFrame({'TV': [230.1]})
```

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X_new.head()
# use the model to make predictions on a new value
lm.predict(X_new)
data['TV'].min()
# create a DataFrame with the minimum and maximum values of TV
X new = pd.DataFrame({'TV': [data['TV'].min(), data['TV'].max()]})
X new.head()
# make predictions for those x values and store them
preds = Im.predict(X_new)
preds
# first, plot the observed data
data.plot(kind='scatter', x='TV', y='Sales')
# then, plot the least squares line
plt.plot(X_new, preds, c='red', linewidth=2)
predictions = Im.predict(X)
print(sqrt(mean_squared_error(y, predictions)))
r2 = r2 score(y, predictions)
r2
```

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```
import numpy
from sklearn import linear_model
X = numpy.array([3.78, 2.44, 2.09, 0.14, 1.72, 1.65, 4.92, 4.37, 4.96, 4.52, 3.69,
5.88]).reshape(-1,1)
y = numpy.array([0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1])
logr = linear_model.LogisticRegression()
logr.fit(X,y)

def logit2prob(logr, X):
    log_odds = logr.coef_ * X + logr.intercept_
    odds = numpy.exp(log_odds)
    probability = odds / (1 + odds)
    return(probability)

print(logit2prob(logr, X))
```

```
import numpy as np
import matplotlib.pyplot as plot
%matplotlib inline
import sklearn
from sklearn.datasets import load digits
digits = load_digits()
X = digits.data
y = digits.target
print("Shape of X is {}".format(X.shape))
print("Shape of y is {}".format(y.shape))
X[0]
def plot digit(x,index):
  plot.imshow(x.reshape(8,8))
  print(index)
plot_digit(X[104],y[104])
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=1)
print(X_train.shape)
from sklearn.linear_model import LogisticRegression
Ir = LogisticRegression()
Ir.fit(X_train,y_train)
y_predict1 = lr.predict(X_train)
from sklearn.metrics import accuracy_score
accuracy = accuracy_score(y_predict1,y_train)
print(accuracy)
y_predict = Ir.predict(X_test)
accuracy = accuracy_score(y_predict,y_test)
print(accuracy)
Ir.predict([X[100], X[152]])
```

```
import pandas
from sklearn import linear model
df = pandas.read csv("cars.csv")
X = df[['Weight', 'Volume']].values
y = df['CO2'].values
regr = linear_model.LinearRegression()
regr.fit(X, y)
#predict the CO2 emission of a car where the weight is 1150kg, and the volume is 1600cm3:
predictedCO2=regr.predict([[1150, 1600]])
print(predictedCO2)
#predict the CO2 emission of a car where the weight is 2300kg, and the volume is 1300cm3:
predictedCO2 = regr.predict([[2300, 1300]])
print(predictedCO2)
Χ
print(regr.intercept_)
print(regr.coef )
# Basic packages
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
# Sklearn modules & classes
from sklearn.linear_model import Perceptron, LogisticRegression
from sklearn.svm import SVC
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn import datasets
from sklearn import metrics
# Load the data set; In this example, the breast cancer dataset is loaded.
bc = datasets.load breast cancer()
```

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X = bc.data
y = bc.target
print(X.shape)
print(y.shape)
# Create training and test split
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=1,
stratify=y)
sc = StandardScaler()
sc.fit(X_train)
X train std = sc.transform(X train)
X_test_std = sc.transform(X_test)
# Instantiate the Support Vector Classifier (SVC)
svc = SVC(C=1.0, random state=1, kernel='linear')
# Fit the model
svc.fit(X train std, y train)
# Make the predictions
y predict = svc.predict(X test std)
# Measure the performance
print("Accuracy score %.3f" %metrics.accuracy_score(y_test, y_predict))
                                              6
# -*- coding: utf-8 -*-
Created on Thu Mar 9 16:46:38 2023
@author: tiver
def hebbian learning(samples):
   print(f'{"INPUT":^8} {"TARGET":^16}{"WEIGHT CHANGES":^15}{"WEIGHTS":^25}')
  w1, w2, b = 1, 1, 1
  print(' ' * 45, f'({w1:2}, {w2:2}, {b:2})')
  for x1, x2, y in samples:
     w1 = w1 + x1 * y
     w2 = w2 + x2 * y
     b = b + v
     print(f'({x1:2}, {x2:2}) {y:2} ({x1*y:2}, {x2*y:2}, {y:2}) ({w1:2}, {w2:2}, {b:2})')
AND samples = {
```

```
'binary_input_binary_output': [
     [1, 1, 1],
     [1, 0, 0],
     [0, 1, 0],
     [0, 0, 0]
  ],
  'binary_input_bipolar_output': [
     [1, 1, 1],
    [1, 0, -1],
     [0, 1, -1],
     [0, 0, -1]
  ],
  'bipolar_input_bipolar_output': [
     [1, 1, 1],
    [1,-1,-1],
    [-1, 1, -1],
    [-1, -1, -1]
  ]
OR samples = {
  'binary_input_binary_output': [
    [1, 1, 1],
     [1, 0, 1],
     [0, 1, 1],
     [0, 0, 0]
  ],
  'binary_input_bipolar_output': [
    [1, 1, 1],
    [1, 0, 1],
     [0, 1, 1],
     [0, 0, -1]
  ],
  'bipolar_input_bipolar_output': [
    [1, 1, 1],
    [1,-1,1],
    [-1, 1, 1],
    [-1, -1, -1]
  ]
XOR samples = {
  'binary_input_binary_output': [
     [1, 1, 0],
     [1, 0, 1],
     [0, 1, 1],
```

```
[0, 0, 0]
  1,
  'binary_input_bipolar_output': [
    [1, 1, -1],
    [1, 0, 1],
    [0, 1, 1],
    [0, 0, -1]
  ],
  'bipolar input bipolar output': [
    [1, 1, -1],
    [1, -1, 1],
    [-1, 1, 1],
    [-1, -1, -1]
  ]
#For AND gate
print('-'*20, 'HEBBIAN LEARNING', '-'*20)
print('AND with Binary Input and Binary Output')
hebbian learning(AND samples['binary input binary output'])
print('AND with Binary Input and Bipolar Output')
hebbian learning(AND samples['binary input bipolar output'])
print('AND with Bipolar Input and Bipolar Output')
hebbian learning(AND samples['bipolar input bipolar output'])
# #OR Gate
# print('-'*20, 'HEBBIAN LEARNING', '-'*20)
# print('OR with binary input and binary output')
# hebbian learning(OR samples['binary input binary output'])
# print('OR with binary input and bipolar output')
# hebbian learning(OR samples['binary input bipolar output'])
# print('OR with bipolar input and bipolar output')
# hebbian learning(OR samples['bipolar input bipolar output'])
# #XOR Gate
# print('-'*20, 'HEBBIAN LEARNING', '-'*20)
# print('XOR with binary input and binary output')
# hebbian learning(XOR samples['binary input binary output'])
# print('XOR with binary input and bipolar output')
# hebbian learning(XOR samples['binary input bipolar output'])
# print('XOR with bipolar input and bipolar output')
# hebbian learning(XOR samples['bipolar input bipolar output'])
```

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7
```

```
# -*- coding: utf-8 -*-
Created on Thu Mar 16 13:18:59 2023
@author: tiver
.....
import numpy as np
# np.random.seed(seed=2)
I = np.random.choice([0,1], 3)# generate random vector I, sampling from {0,1}
# W = np.random.choice([-1,1], 3) # generate random vector W, sampling from {-1,1}
W = np.array([1,1,1])
print(f'Input vector:{I}, Weight vector:{W}')
dot = I @ W
print(f'Dot product: {dot}')
def linear threshold gate(dot: int, T: float) -> int:
  "Returns the binary threshold output"
  if dot >= T:
    return 1
  else:
    return 0
T = 3
activation = linear_threshold_gate(dot, T)
print(f'When Threshold =3, Activation: {activation}')
                                              8
# -*- coding: utf-8 -*-
Created on Thu Mar 23 12:29:55 2023
@author: tiver
111111
import numpy as np
class Perceptron:
```

```
def __init__(self, input_size, lr=1, epochs=100):
    self.W = np.zeros(input_size+1)
    self.epochs = epochs
    self.lr = Ir
  def activation fn(self, x):
    return 1 if x \ge 0 else 0
  def predict(self, x):
    z = self.W.T.dot(x)
    a = self.activation fn(z)
    return a
  def fit(self, X, d):
    for epoch in range(self.epochs):
       for i in range(d.shape[0]):
         x = np.insert(X[i], 2, 1)
         y = self.predict(x)
         e = d[i] - y
         self.W = self.W + self.lr * e * x
#For AND Gate
X1 = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
d1 = np.array([0, 0, 0, 1])
#For OR Gate
X2 = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
d2 = np.array([0, 1, 1, 1])
perceptron1 = Perceptron(input_size=2)
perceptron1.fit(X1, d1)
perceptron2 = Perceptron(input_size=2)
perceptron2.fit(X2, d2)
print(perceptron1.W)
# Output: [2. 1. -3.]
print(perceptron2.W)
# Output: [1. 1. -1.]
test_in=np.array([0, 1, 1])
AND_prediction=perceptron1.predict(test_in)
print(AND prediction)
```

```
OR_prediction=perceptron2.predict(test_in)
print(OR_prediction)
                                             9
# -*- coding: utf-8 -*-
Created on Wed Apr 5 14:24:40 2023
@author: tiver
111111
# the code for importing and splitting the dataset
import sklearn
from sklearn.datasets import load digits
digits = load digits()
X = digits.data
y = digits.target
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random state=1)
print(X_train.shape)
# Transforming the train and test sets such that they explain 95% of variance
from sklearn.decomposition import PCA
sklearn_pca = PCA(n_components=0.95)
sklearn pca.fit(X train)
X_train_transformed = sklearn_pca.transform(X_train)
print(X_train_transformed.shape)
print(X_test.shape)
X test transformed =sklearn pca.transform(X test)
```

print(X_test_transformed.shape)