



# Machine Learning & Deep Learning

635 | Vishal Baraiya |  
23010101014

Lab - 5

## SVR

### Importing the libraries

```
In [1]: import pandas as pd  
import numpy as np
```

### Read World bank CSV

```
In [2]: df = pd.read_csv("WorldBank.csv")
```

```
In [3]: df.head()
```

Out[3]:	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	1962
0	India	IND	Export value index (2000 = 100)	TX.VAL.MRCH.XD.WD	NaN	NaN	NaN
1	India	IND	Insurance and financial services (% of commerc...	TX.VAL.INSF.ZS.WT	NaN	NaN	NaN
2	India	IND	Merchandise imports by the reporting economy, ...	TM.VAL.MRCH.RS.ZS	4.983551	6.48805	10.124611
3	India	IND	Food imports (% of merchandise imports)	TM.VAL.FOOD.ZS.UN	NaN	NaN	17.080013
4	India	IND	Share of tariff lines with international peaks...	TM.TAX.MRCH.IP.ZS	NaN	NaN	NaN

5 rows × 65 columns



## Perform conditional selection to find - Population ages 15-64 (% of total population)

In [21]: `df1 = df[df['Indicator Name'] == 'Population ages 15-64 (% of total population)']`

In [22]: `df1`

Out[22]:	Country Name	Country Code	Indicator Name	Indicator Code	1960	1961	1962	1
9	India	IND	Population ages 15-64 (% of total population)	SP.POP.1564.TO.ZS	56.49748	56.177532	55.807455	55.461

1 rows × 65 columns



## Divide the data into input and output

```
In [23]: X = np.arange(1960,2020).reshape(-1,1)
```

```
In [40]: y = df1.values[0][4:-1:].reshape(-1,1)
# y = df1.values[0][4:-1:]
```

```
In [41]: y
```

```
Out[41]: array([[56.49748004],  
 [56.17753236],  
 [55.80745463],  
 [55.46166361],  
 [55.24893881],  
 [55.21135053],  
 [55.09090078],  
 [55.15534672],  
 [55.34507283],  
 [55.57014408],  
 [55.78194745],  
 [55.85676846],  
 [55.95268174],  
 [56.07247186],  
 [56.23447551],  
 [56.44405309],  
 [56.49722595],  
 [56.62068516],  
 [56.78900152],  
 [56.9691436],  
 [57.1425581],  
 [57.18105454],  
 [57.22630775],  
 [57.28875038],  
 [57.39054366],  
 [57.54142108],  
 [57.55350984],  
 [57.6545119],  
 [57.81875323],  
 [58.01501187],  
 [58.22990246],  
 [58.37403848],  
 [58.5472698],  
 [58.75605047],  
 [59.012126],  
 [59.31657719],  
 [59.56507329],  
 [59.8572303],  
 [60.18600058],  
 [60.53971518],  
 [60.90862046],  
 [61.18898716],  
 [61.4993847],  
 [61.83084479],  
 [62.173897],  
 [62.52276485],  
 [62.80842981],  
 [63.10261029],  
 [63.40924784],  
 [63.74196691],  
 [64.10821053],  
 [64.429404],  
 [64.80551944],  
 [65.20848906],  
 [65.5959799],  
 [65.94416405],
```

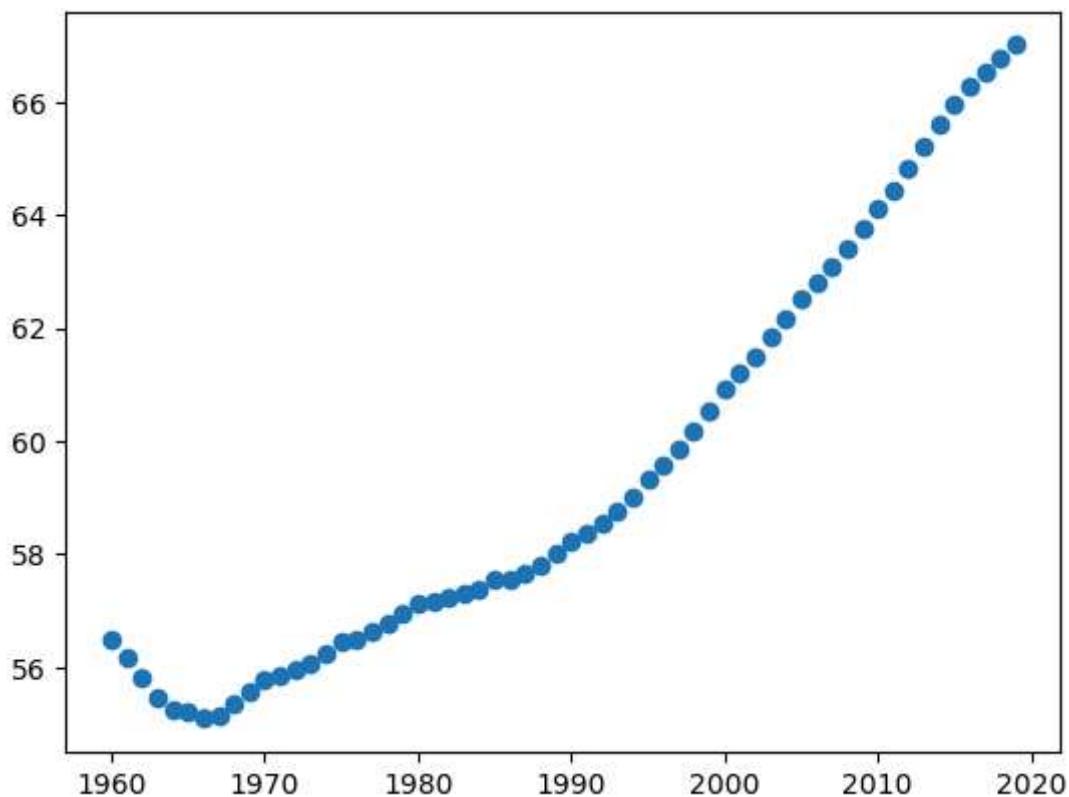
```
[66.27426247],  
[66.53818711],  
[66.7667425],  
[67.00381119]], dtype=object)
```

## Plot scatter plot of Population ages 15-64 (% of total population)

```
In [42]: import matplotlib.pyplot as plt
```

```
In [44]: plt.scatter(X,y)
```

```
Out[44]: <matplotlib.collections.PathCollection at 0x2359d1b6f60>
```



## Feature Scaling (Mandatory for SVR)\*\*

SVR is highly sensitive to the range of data points. If we don't scale (normalize) the data, the model will fail to find the correct hyperplane.

```
In [45]: from sklearn.preprocessing import StandardScaler
```

```
In [47]: scaler_x = StandardScaler()  
scaler_y = StandardScaler()  
  
x_scaled = scaler_x.fit_transform(X)
```

```
y_scaled = scaler_y.fit_transform(y)

print("Scaling Complete")
```

Scaling Complete

In [ ]:

## Splitting the dataset into the Training set and Test set

In [ ]:

In [ ]:

In [ ]:

## Fitting SVR on 3 Different Kernel on dataset

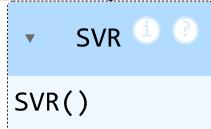
In [65]: `from sklearn.svm import SVR`

In [66]: `svr_linear = SVR()`

In [67]: `svr_linear.fit(x_scaled,y_scaled)`

```
D:\Anaconda\Lib\site-packages\sklearn\utils\validation.py:1339: DataConversionWarning
  g: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
    y = column_or_1d(y, warn=True)
```

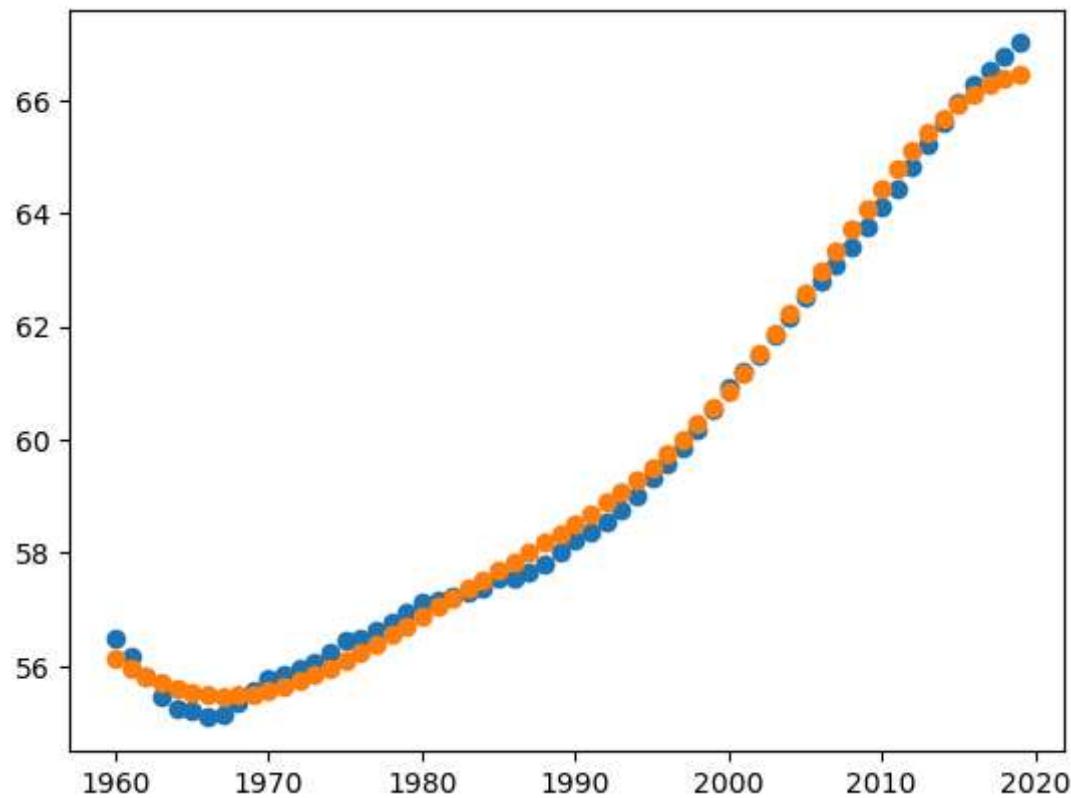
Out[67]:



In [68]: `y_predicted = scaler_y.inverse_transform(svr_linear.predict(x_scaled).reshape(-1,1))`

In [69]: `plt.scatter(X,y)
plt.scatter(X,y_predicted)`

Out[69]: <matplotlib.collections.PathCollection at 0x2359db66060>



In [ ]:

In [61]:

## Predict the x\_test using 3 Kernel

In [ ]: `model_rbf.score(x_train,y_train)`Out[ ]: `0.9947138713539011`In [ ]: `model_rbf.score(x_test,y_test)`Out[ ]: `0.9880623181593732`

In [ ]:

In [ ]:

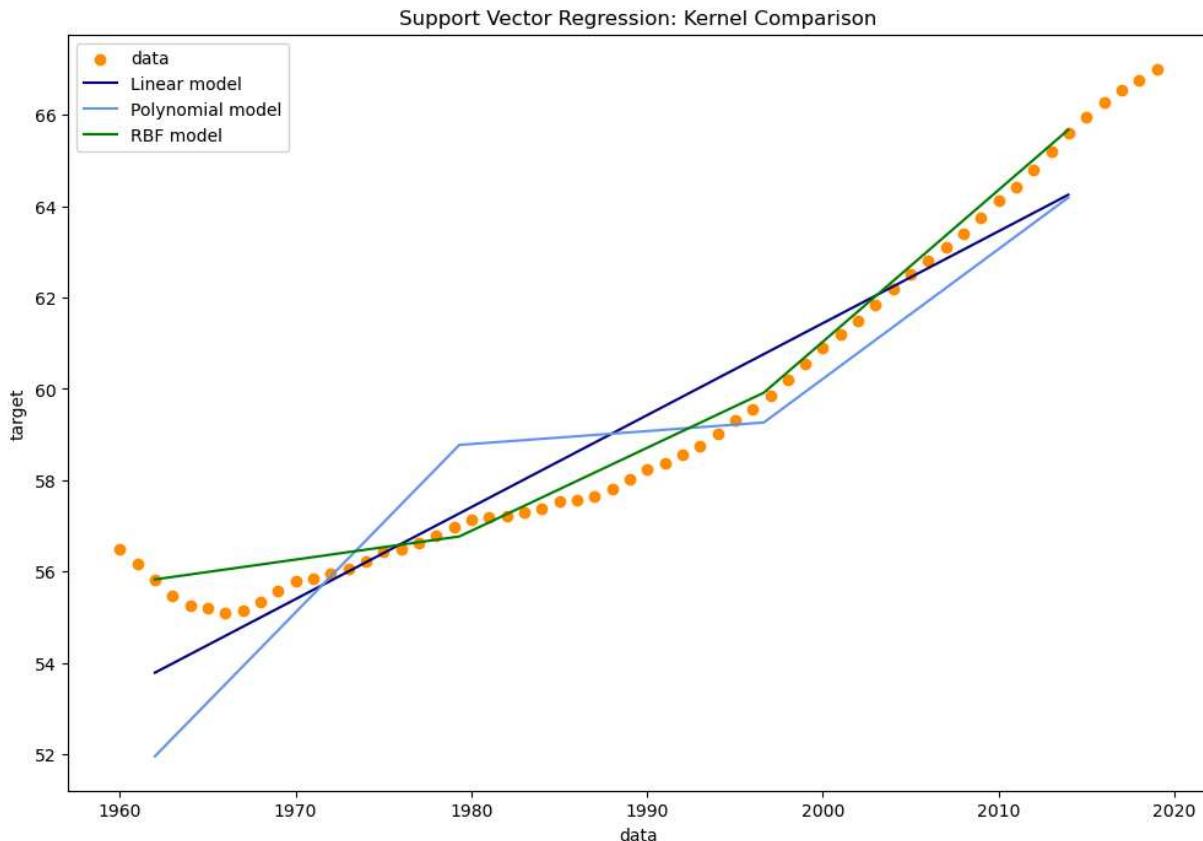
Out[ ]: `array([57.55350984, 59.31657719, 67.00381119, 57.81875323, 55.85676846, 55.80745463, 59.012126, 66.7667425, 60.90862046, 57.22630775, 55.24893881, 55.78194745, 58.22990246, 61.18898716, 58.75605047, 61.83084479, 63.74196691, 55.15534672], dtype=object)`

In [ ]:

```
Out[ ]: array([57.73445326, 59.30206118, 64.8451716 , 57.99812952, 55.96479236,
   56.12587167, 59.06485884, 64.91876466, 60.76906052, 57.24445561,
   55.93360816, 55.89821768, 58.29641078, 61.1088798 , 58.84682535,
   61.81299688, 63.81278461, 55.81527994])
```

## Visualising the results

In [ ]:



## Student Activity : Prediction

**Task:** Predict the value for the Year **2025** (or value 6.5 in the demo data). Remember: You must transform the input before predicting, and inverse transform the output.

```
In [75]: newData = [[2025]]
```

```
In [76]: predicted = svr_linear.predict(scaler_x.transform(newData))
```

```
In [77]: scaler_y.inverse_transform(predicted.reshape(-1,1))
```

```
Out[77]: array([[65.91120533]])
```

In [ ]: