SVM_Kernelss_Implementation

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0.1 SVM (Support Vector Machine) Kernels Indepth Intuition And Practical Explanation

SVM (Support Vector Machine) Kernal: Implementation Using Sklearn- Machine Learning SVM algorithms use a set of mathematical functions that are defined as the kernel. The function of kernel is to take data as input and transform it into the required form. Different SVM algorithms use different types of kernel functions. These functions can be different types. For example linear, nonlinear, polynomial, radial basis function (RBF), and sigmoid.

```
[1]: #importing required libreries
     import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
[2]: # Creating data
     x = np.linspace(-5.0, 5.0, 100)
     y = np.sqrt(10**2 - x**2)
     y=np.hstack([y,-y])
     x=np.hstack([x,-x])
[3]: x1 = np.linspace(-5.0, 5.0, 100)
     y1 = np.sqrt(5**2 - x1**2)
     y1=np.hstack([y1,-y1])
     x1=np.hstack([x1,-x1])
[4]: # Visualization
     plt.scatter(y,x)
     plt.scatter(y1,x1)
```

[4]: <matplotlib.collections.PathCollection at 0x26bf80c4bb0>

```
4 - 2 - 0 - -2 - -4 - -10.0 -7.5 -5.0 -2.5 0.0 2.5 5.0 7.5 10.0
```

```
[5]: df1 =pd.DataFrame(np.vstack([y,x]).T,columns=['X1','X2'])
     df1['Y']=0
     df2 =pd.DataFrame(np.vstack([y1,x1]).T,columns=['X1','X2'])
     df2['Y']=1
     df = df1.append(df2)
     df.head(5)
[5]:
                       X2 Y
              Х1
     0 8.660254 -5.00000 0
     1 8.717792 -4.89899
     2 8.773790 -4.79798
     3 8.828277 -4.69697 0
     4 8.881281 -4.59596 0
[6]: ### Independent and Dependent features
     X = df.iloc[:, :2]
     y = df.Y
[7]: y
[7]: 0
           0
     1
           0
     2
           0
     3
           0
     4
           0
```

```
195
             1
      196
             1
      197
             1
      198
             1
      199
             1
      Name: Y, Length: 400, dtype: int64
 [8]: ## Split the dataset into train and test
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.
       \hookrightarrow25, random_state=0)
 [9]: y_train
 [9]: 50
             1
      63
             0
      112
             1
      159
             0
      83
             1
      123
             1
      192
             0
      117
             0
      47
             0
      172
             0
      Name: Y, Length: 300, dtype: int64
[10]: from sklearn.svm import SVC
      classifier=SVC(kernel="rbf")
      classifier.fit(X_train,y_train)
[10]: SVC()
[11]: from sklearn.metrics import accuracy_score
      y_pred = classifier.predict(X_test)
      accuracy_score(y_test, y_pred)
[11]: 1.0
[12]: df.head()
[12]:
               Х1
                        X2 Y
      0 8.660254 -5.00000 0
      1 8.717792 -4.89899 0
      2 8.773790 -4.79798 0
      3 8.828277 -4.69697 0
      4 8.881281 -4.59596 0
```

0.1.1 Polynomial Kernel

```
0.2 K(x, y) = (x^T y + c)^d
```

```
[13]: # We need to find components for the Polynomical Kernel
      #X1,X2,X1_square,X2_square,X1*X2
      df['X1_Square'] = df['X1']**2
      df['X2_Square'] = df['X2']**2
      df['X1*X2'] = (df['X1'] *df['X2'])
      df.head()
[13]:
                       X2 Y X1_Square X2_Square
               Х1
                                                         X1*X2
      0 8.660254 -5.00000 0 75.000000 25.000000 -43.301270
      1 8.717792 -4.89899 0 75.999898 24.000102 -42.708375
      2 8.773790 -4.79798 0 76.979390 23.020610 -42.096467
      3 8.828277 -4.69697 0 77.938476 22.061524 -41.466150
      4 8.881281 -4.59596 0 78.877155 21.122845 -40.818009
[14]: ### Independent and Dependent features
      X = df[['X1','X2','X1_Square','X2_Square','X1*X2']]
      y = df['Y']
[15]: y
[15]: 0
             0
      1
             0
      2
             0
      3
             0
             0
            . .
      195
            1
      196
      197
             1
      198
      199
             1
      Name: Y, Length: 400, dtype: int64
[16]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25,__
      \rightarrowrandom state = 0)
[17]: X_train
[17]:
                X1
                           X2 X1_Square X2_Square
                                                         X1*X2
      50
           4.999745 0.050505 24.997449
                                           0.002551
                                                      0.252512
      63
           9.906589
                    1.363636 98.140496
                                           1.859504
                                                    13.508984
      112 -3.263736 3.787879 10.651974 14.348026 -12.362637
      159 -9.953852 -0.959596 99.079176
                                           0.920824
                                                      9.551676
      83
           3.680983 3.383838 13.549638 11.450362 12.455852
```

```
123 -4.223140 2.676768 17.834915
                                          7.165085 -11.304366
      192 -9.031653 -4.292929 81.570758 18.429242 38.772248
      117 -9.445795 3.282828
                               89.223038 10.776962 -31.008922
           9.996811 -0.252525
      47
                               99.936231
                                           0.063769
                                                     -2.524447
      172 -9.738311 -2.272727
                                           5.165289 22.132526
                               94.834711
      [300 rows x 5 columns]
[18]: import plotly.express as px
      fig = px.scatter_3d(df, x='X1', y='X2', z='X1*X2', color='Y')
      fig.show()
                                                                                          0.8
         40
          20
                                                                                          0.6
            0
            -20
                                                                                          0.4
                                                                                          0.2
                 X2
                                                X1
  [19]: fig = px.scatter_3d(df, x='X1_Square', y='X1_Square', z='X1*X2', color='Y')
         fig.show()
                                                                                          0.8
            40
             20
                                                                                          0.6
              0
               _20
                                                                                          0.4
                                                                                          0.2
    [20]: classifier = SVC(kernel="linear")
          classifier.fit(X_train, y_train)
          y_pred = classifier.predict(X_test)
          accuracy_score(y_test, y_pred)
```

[20]: 1.0