**Classifying Muffins and Cupcakes with SVM** Based on the quantity of ingrediants added into the recipes. We come up with two different types of snacks one is Muffin and second one is cupcake. Refrences: https://youtu.be/N1vOgolbjSc **Step 1:** Import Packages In [2]: # Packages for analysis import pandas as pd import numpy as np from sklearn import svm # Packages for visuals import matplotlib.pyplot as plt import seaborn as sns # Allows charts to appear in the notebook %matplotlib inline Step 2: Import Data In [3]: # Read in muffin and cupcake ingredient data recipes = pd.read\_csv('recipes\_muffins\_cupcakes.csv') Baking Powder Vanilla Salt Type Flour Milk Sugar Butter Egg Out[3]: Muffin 55 28 3 7 5 0 Muffin 47 24 12 9 1 0 0 6 1 1 0 2 Muffin 47 23 18 6 4 0 3 Muffin 45 11 17 17 8 0 0 2 12 5 0 4 Muffin 50 25 6 1 5 Muffin 55 27 3 7 5 1 0 7 5 2 0 0 6 Muffin 54 27 5 1 7 Muffin 47 26 10 10 0 0 0 8 Muffin 50 17 17 8 6 1 0 9 Muffin 50 17 0 0 17 11 4 **10** Cupcake 39 0 26 19 14 1 1 0 42 21 10 8 0 0 **11** Cupcake 16 2 5 1 0 **12** Cupcake 34 17 20 20 13 Cupcake 39 13 17 10 1 0 19 0 0 **14** Cupcake 38 15 23 15 8 1 1 0 **15** Cupcake 42 18 25 9 5 0 2 **16** Cupcake 36 14 21 14 11 1 0 38 31 1 0 **17** Cupcake 15 8 6 1 18 Cupcake 9 1 0 36 16 24 12 1 11 13 19 Cupcake 34 17 23 1 0 Step 3: Prepare the Data # Plot two ingredients sns.lmplot('Flour', 'Sugar', data=recipes, hue='Type', fit\_reg=False); C:\Users\Manav\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From versi on 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misi nterpretation. warnings.warn( 30 25 20 Type Muffin Cupcake 10 5 35 40 45 50 Flour In [5]: # Specify inputs for the model # ingredients = recipes[['Flour', 'Milk', 'Sugar', 'Butter', 'Egg', 'Baking Powder', 'Vanilla', 'Salt']].as\_matrix() ingredients = recipes[['Flour', 'Sugar']].to\_numpy() type\_label = np.where(recipes['Type']=='Muffin', 0, 1) Step 4: Fit the Model In [6]: # Fit the SVM model model = svm.SVC(kernel='linear') model.fit(ingredients, type\_label) SVC(kernel='linear') Out[6]: Step 5: Visualize Results In [7]: # Get the separating hyperplane  $w = model.coef_[0]$ a = -w[0] / w[1]xx = np.linspace(30, 60) $yy = a * xx - (model.intercept_[0]) / w[1]$ # Plot the parallels to the separating hyperplane that pass through the support vectors b = model.support\_vectors\_[0]  $yy_down = a * xx + (b[1] - a * b[0])$ b = model.support\_vectors\_[-1]  $yy_up = a * xx + (b[1] - a * b[0])$ # Plot the hyperplane sns.lmplot('Flour', 'Sugar', data=recipes, hue='Type', fit\_reg=False) plt.plot(xx, yy, linewidth=2, color='black'); C:\Users\Manav\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From versi on 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misi nterpretation. warnings.warn( le16 1.0 0.5 0.0 Type Muffin Cupcake -0.5-1.0-1.530 35 50 55 Flour In [9]: # Look at the margins and support vectors sns.lmplot('Flour', 'Sugar', data=recipes, hue='Type', palette='Set1', fit\_reg=False, scatter\_kws={"s": 70}) plt.plot(xx, yy, linewidth=2, color='black') plt.plot(xx, yy\_down, 'k--') plt.plot(xx, yy\_up, 'k--') #plt.scatter(model.support\_vectors\_[:, 0], model.support\_vectors\_[:, 1]); C:\Users\Manav\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From versi on 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misi nterpretation. warnings.warn( [<matplotlib.lines.Line2D at 0x203388938b0>] Out[9]: 1.0 0.5 0.0 Muffin Cupcake -0.5-1.0-1.530 50 55 35 45 Flour Step 6: Predict New Case In [10]: # Create a function to guess when a recipe is a muffin or a cupcake def muffin\_or\_cupcake(flour, sugar): if(model.predict([[flour, sugar]]))==0: print('You\'re looking at a muffin recipe!') print('You\'re looking at a cupcake recipe!') In [11]: # Predict if 50 parts flour and 20 parts sugar muffin\_or\_cupcake(50, 20) You're looking at a muffin recipe! In [12]: # Plot the point to visually see where the point lies sns.lmplot('Flour', 'Sugar', data=recipes, hue='Type', palette='Set1', fit\_reg=False, scatter\_kws={"s": 70}) plt.plot(xx, yy, linewidth=2, color='black') plt.plot(50, 20, 'yo', markersize='9'); C:\Users\Manav\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From versi on 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misi nterpretation. warnings.warn( le16 1.0 0.5 0.0 Туре Muffin Cupcake -0.5-1.0-1.560 30 35 40 45 50 55 Flour In [13]: # Predict if 40 parts flour and 20 parts sugar muffin\_or\_cupcake(40,20) You're looking at a cupcake recipe!

In [1]:

# 19IT016: Manav Butani

WEEK09: SVM

## WEEK09\_SVM\_synthetic\_dataset,\_iris\_dataset

```
In [1]:
          import numpy as np
          import matplotlib.pyplot as plt
          from sklearn.svm import SVC
          from sklearn.datasets import make_blobs
          from sklearn import datasets
In [27]:
          X, y = make_blobs(n_samples=50, centers=2, random_state=4)
          plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Set3)
          print(X.shape, y.shape)
         (50, 2) (50,)
           6
           5
           3
           2
           1
          0
          ^{-1}
                                              11
In [28]:
          for x in {'linear', 'poly', 'rbf', 'sigmoid'}:
            # fit the model, don't regularize for illustration purposes
            clf = SVC(kernel=x, C=0.70)
            clf.fit(X, y)
            plt.title(x)
            plt.scatter(X[:, 0], X[:, 1], c=y, s=30, cmap=plt.cm.Paired)
            # plot the decision function
            ax = plt.gca()
            xlim = ax.get_xlim()
            ylim = ax.get_ylim()
            # create grid to evaluate model
            xx = np.linspace(xlim[0], xlim[1], 30)
            yy = np.linspace(ylim[0], ylim[1], 30)
            YY, XX = np.meshgrid(yy, xx)
            xy = np.vstack([XX.ravel(), YY.ravel()]).T
            Z = clf.decision_function(xy).reshape(XX.shape)
            # plot decision boundary and margins
            ax.contour(
                XX, YY, Z, colors="k", levels=[-1, 0, 1], alpha=0.5, linestyles=["--", "-", "--"]
            # plot support vectors
            ax.scatter(
                clf.support_vectors_[:, 0],
                clf.support_vectors_[:, 1],
                s=100,
                linewidth=1,
                facecolors="none",
                edgecolors="k",
            plt.show()
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           6
           5
           2
           0
          -1
 In [4]:
          # import some data to play with
          iris = datasets.load_iris()
 In [5]:
          iris.data.shape
          # Take the first two features. We could avoid this by using a two-dim dataset
          X_{train} = iris.data[:120, :2]
          y_train = iris.target[:120]
          X_{\text{test}} = iris.data[120:, :2]
          y_test = iris.target[120:]
          print(X_train.shape, y_train.shape, X_test.shape, y_test.shape)
         (120, 2) (120,) (30, 2) (30,)
In [8]:
          from sklearn.metrics import accuracy_score
          C = 1.0 # SVM regularization parameter
          models = (
              SVC(kernel="linear", C=C),
              SVC(C=C, max_iter=10000),
              SVC(kernel="rbf", gamma=0.7, C=C),
              SVC(kernel="poly", degree=3, gamma="auto", C=C),
          models = (clf.fit(X_train, y_train) for clf in models)
          for model in models:
            print(accuracy_score(model.predict(X_test),y_test))
         0.2
```

0.2 0.2 0.2