

20211013_Supplier_Management_MBW7

October 13, 2021

```
[1]: # Machine Learning applied to Supply Chain Management
```

```
[2]: # Support Vector Machines
```

```
[3]: # This method is used to separate datasets by distance metrics
```

```
[4]: # Image you have data from two different processes.  
# You want to know which datapoints belong to which process,  
# depending on the values of the process.
```

```
[5]: # This method separates the data and puts those data together that are closer  
→ to each other.
```

```
[8]: !pip install numpy # Numpy. Numerical Python  
!pip install scipy # SciPy. Scientific Python  
!pip install matplotlib # Matplotlib. Graphical representations  
!pip install seaborn # Seaborn. Create datasets  
!pip install sklearn # Sklearn. Machine Learning (learn)
```

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Requirement already satisfied: numpy in  
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (1.19.5)  
Requirement already satisfied: scipy in  
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (1.6.2)  
Requirement already satisfied: numpy<1.23.0,>=1.16.5 in  
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Requirement already satisfied: pillow>=6.2.0 in  
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from matplotlib) (8.3.2)  
Requirement already satisfied: python-dateutil>=2.1 in  
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from matplotlib) (2.8.1)  
Requirement already satisfied: pyparsing!=2.0.4,!2.1.2,!2.1.6,>=2.0.3 in  
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from matplotlib) (2.4.7)  
Requirement already satisfied: cycycler>=0.10 in
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Requirement already satisfied: six in
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from
cyclar>=0.10->matplotlib) (1.15.0)
Requirement already satisfied: seaborn in
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(2021.1)
Collecting sklearn
  Using cached sklearn-0.0-py2.py3-none-any.whl
Requirement already satisfied: scikit-learn in
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from sklearn) (0.24.1)
Requirement already satisfied: numpy>=1.13.3 in
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from scikit-learn->sklearn)
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/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from scikit-learn->sklearn)
(2.1.0)
Requirement already satisfied: joblib>=0.11 in
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from scikit-learn->sklearn)
(1.0.1)

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Requirement already satisfied: scipy>=0.19.1 in
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (from scikit-learn->sklearn)
(1.6.2)

Installing collected packages: sklearn

Successfully installed sklearn-0.0

```
[9]: import numpy as np # everytime I call "np" the kernel understands "numpy"
import matplotlib.pyplot as plt # everytime I call "plt" the kernel understands
↳ "matplotlib.pyplot"
from scipy import stats
import seaborn as sns; sns.set()
import sklearn
```

```
[14]: # Generate some dataset using SKLEARN

from sklearn.datasets import make_blobs

# the sub package datasets within sklearn has a function in which
# we can create datasets made of points (blobs)
```

```
[26]: # now we generate the dataset

X , y = make_blobs (n_samples = 100, # first we define the number of points
                    centers = 2,      # number of classes
                    cluster_std = 0.8)# standard deviation of the data
```

```
[27]: print(X)
```

```
[[ -8.39297718  -0.33496523]
 [ -7.85711898   1.01005544]
 [ -8.48855104   1.72715275]
 [  1.29583396   1.56418481]
 [  0.88840294   1.77740796]
 [ -0.23741854   0.84981091]
 [ -8.49851301   0.09431891]
 [ -9.8795934    1.50279029]
 [  2.17126885   1.27992086]
 [  2.3496711    1.36364099]
 [ -8.79203411   0.84270945]
 [ -7.47174704   1.89673019]
 [ -8.9281964    0.96439594]
 [  2.56106708   1.80105966]
 [ -8.4982567    0.66157709]
 [ -7.53904084   0.41043409]
 [ -7.42685306   1.46719359]
 [  0.93092993   2.28938255]
 [ -8.301917     0.63102994]
 [  2.68069347   1.21564931]
```

[-8.29235778	-1.28422]
[1.58680928	1.30600376]
[2.25551565	1.33556006]
[2.0066601	0.05869983]
[2.5811785	2.2646207]
[2.11215996	1.76221356]
[-8.32118632	0.9705696]
[1.3894115	1.62374328]
[-8.59667508	0.88226113]
[-9.17615997	0.12923971]
[1.54702384	2.68184511]
[-8.7166899	1.18713438]
[-7.20502537	1.62983032]
[-8.27236967	2.178396]
[0.04380205	1.24694589]
[1.74016323	2.09298349]
[0.85765643	1.71975366]
[2.41042389	1.44448195]
[-9.03042081	0.41087721]
[2.12023237	-0.07853032]
[-8.47418881	0.27568784]
[0.72525466	1.50768113]
[-0.01562176	2.15088837]
[-9.07316276	-0.23873019]
[-7.34750437	0.8183989]
[1.49629365	1.69402541]
[0.69299168	1.62722525]
[-8.09375048	2.03559795]
[1.74156237	1.45562521]
[-8.31000615	1.2331159]
[-7.64921106	1.22010546]
[0.53601981	0.35501926]
[0.66603062	-0.02076456]
[-8.71315176	-0.14195039]
[-8.10892591	0.97377296]
[1.09260064	1.1802545]
[-8.17952907	2.18112821]
[1.84469539	2.63850358]
[0.15870064	2.10687792]
[-8.03893475	1.3764203]
[0.5963779	0.04867026]
[2.57782064	1.06685514]
[3.95383742	1.15444756]
[2.26968518	0.55742431]
[-8.25905374	1.00170354]
[2.65086554	0.31928087]
[-8.22925878	2.06254716]
[-6.84929485	-0.08844469]

```

[ -8.13472006    1.08572591]
[  2.10212779    0.62256938]
[  1.59057737    1.53970799]
[ -8.21573665    1.78369623]
[ -9.42390872    1.01863893]
[ -8.42527562    0.22948207]
[  0.99420322    0.51250243]
[  0.90474053    1.28796154]
[ -8.29536433    1.78693483]
[ -8.07315117    0.3814427 ]
[ -9.55377076    0.9944138 ]
[  2.95745982    2.09886931]
[ -0.34884223   -0.23867254]
[  1.42381658    1.39863499]
[ -9.59454649    1.74020796]
[-11.32644585    2.46699745]
[ -7.8874499     1.88421551]
[ -0.1619124     2.17454672]
[  1.52366127    2.36693851]
[ -9.11008465    0.15712576]
[  1.21041666    1.45426962]
[  0.67277855    3.03340399]
[  3.14330866   -0.57315537]
[ -8.28140108    1.87301047]
[  1.33427973    1.2559591 ]
[-10.63801996    1.78811681]
[ -9.43634897    1.91276351]
[ -9.12290059    2.31346692]
[ -9.6139816     0.46762008]
[ -8.02447395    0.23918307]
[  1.73186966    1.15105054]
[  0.46400489    1.99953319]]

```

```
[28]: print(y)
```

```

[0 0 0 1 1 1 0 0 1 1 0 0 0 1 0 0 0 1 0 1 0 1 1 1 1 1 0 1 0 0 1 0 0 0 1 1 1
 1 0 1 0 1 1 0 0 1 1 0 1 0 0 1 1 0 0 1 0 1 1 0 1 1 1 1 0 1 0 0 0 1 1 0 0 0
 1 1 0 0 0 1 1 1 0 0 0 1 1 0 1 1 1 0 1 0 0 0 0 0 1 1]

```

```

[31]: # graphical representation with "plt"

# plotting on the horizontal axis the first column of X "X[:,0]"
# and in the vertical axis the second column of X "X[:,1]"

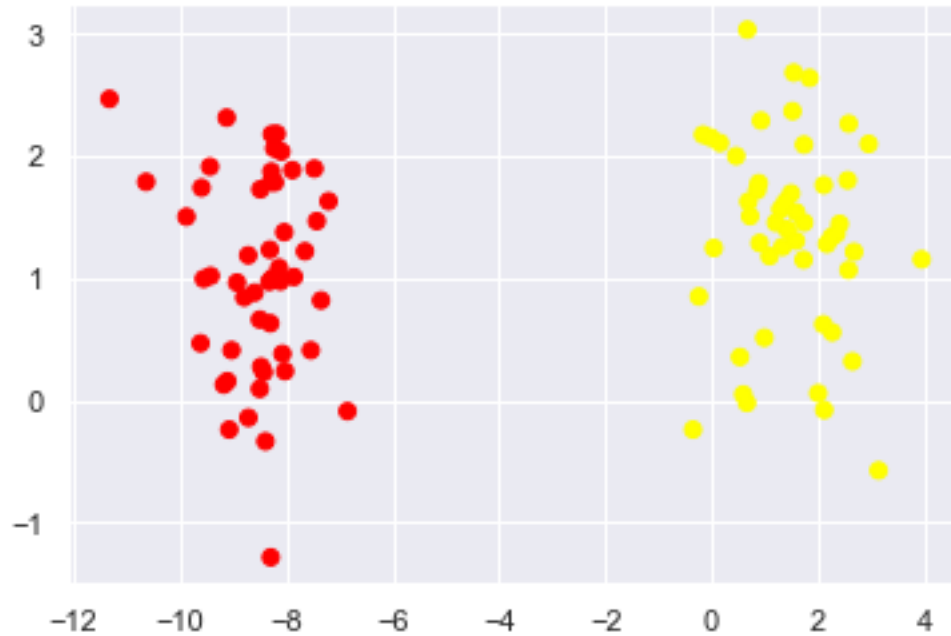
# the color "c" is given by the value of "y"

# we can also choose a color palette "autumn"

```

```
plt.scatter(X[:,0], X[:,1], c=y, cmap='autumn')
```

[31]: <matplotlib.collections.PathCollection at 0x7fd6eed92ac0>



[32]: *# in reality we do not know what are the colors yet*

```
plt.scatter(X[:,0], X[:,1])
```

[32]: <matplotlib.collections.PathCollection at 0x7fd6eef0a8b0>



```
[36]: # Linear Support Vector Machines
# This algorithm is going to try to draw a separation line
# between the two datasets which is optimal (maximal distance to all points)

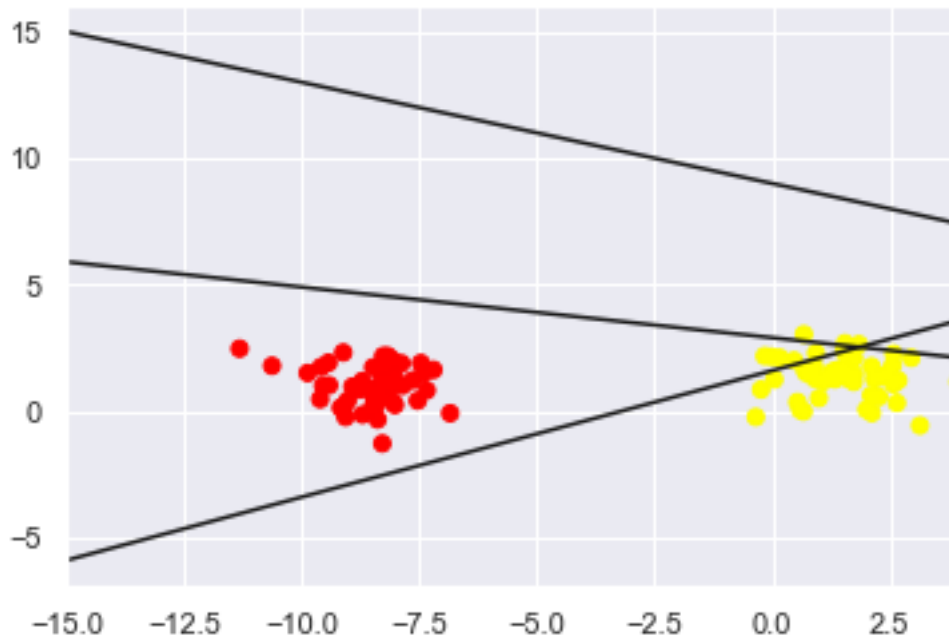
# for a "2D" dataset we can do it by hand, but for many dimensions, it is more
→difficult

xfit = np.linspace(-15,4)
plt.scatter(X[:,0], X[:,1], c=y, cmap='autumn')

for m, b in [(-0.4,9), (0.5, 1.6), (-0.2, 2.9)]:
    plt.plot(xfit, m*xfit+b, "-k")

plt.xlim(-15, 4)
```

[36]: (-15.0, 4.0)



```
[37]: # Support Vector Machines to calculate the line that best separates the dataset

from sklearn.svm import SVC # "SVC" means support vector classifier

model = SVC(kernel='linear')

model.fit(X,y)
```

```
[37]: SVC(kernel='linear')
```

```
[43]: # Plot the results of the model (not relevant for the exam)

def plot_svc_decision_function(model, ax=None, plot_support=True):
    if ax is None:
        ax=plt.gca()
    xlim=ax.get_xlim()
    ylim=ax.get_ylim()
    x=np.linspace(xlim[0],xlim[1],30)
    y=np.linspace(ylim[0],ylim[1],30)
    Y,X = np.meshgrid(y,x)
    xy=np.vstack([X.ravel(),Y.ravel()]).T
    P=model.decision_function(xy).reshape(X.shape)
    ax.contour(X, Y, P, colors='k',
               levels=[-1, 0, 1], alpha=0.5,
               linestyles=['--', '-', '--'])
```



```

if plot_support:
    ax.scatter(model.support_vectors_[0],
               model.support_vectors_[1],
               s=300, linewidth=1, facecolors='none');
ax.set_xlim(xlim)
ax.set_ylim(ylim)

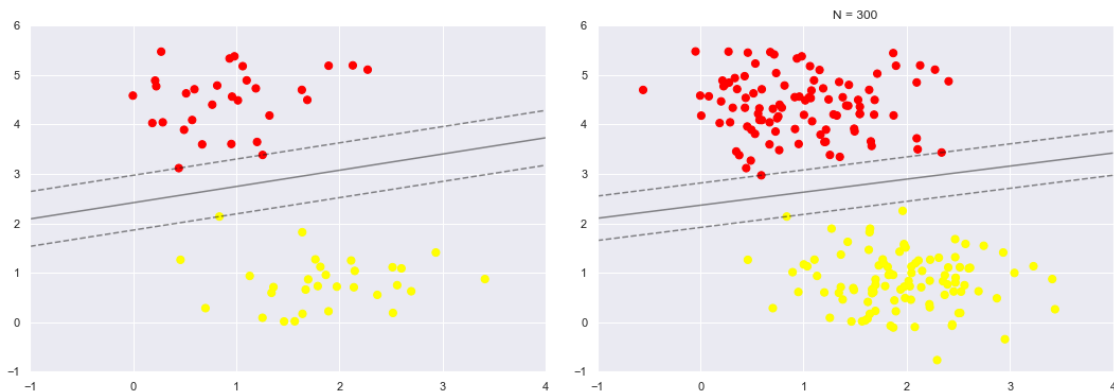
def plot_svm(N=10, ax=None):
    X,y =make_blobs(n_samples=200, centers=2,
                    random_state=0, cluster_std=0.6)

    X=X[:N]
    y=y[:N]
    model=SVC(kernel='linear', C=1E10)
    model.fit(X,y)
    ax=ax or plt.gca()
    ax.scatter(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
    ax.set_xlim(-1, 4)
    ax.set_ylim(-1, 6)
    plot_svc_decision_function(model,ax)

fig,ax=plt.subplots(1,2,figsize=(16,6))
fig.subplots_adjust(left=0.0625, right=0.95, wspace=0.1)
for axi, N in zip(ax, [60, 300]): plot_svm(N, axi)
axi.set_title('N = {0}'.format(N))

```

[43]: Text(0.5, 1.0, 'N = 300')



```

[44]: # what happens if the dataset cannot be separated with lines!?
      # what happens if the dataset has a circular (non linear) shape?!

      # then we need to use other kernel

```

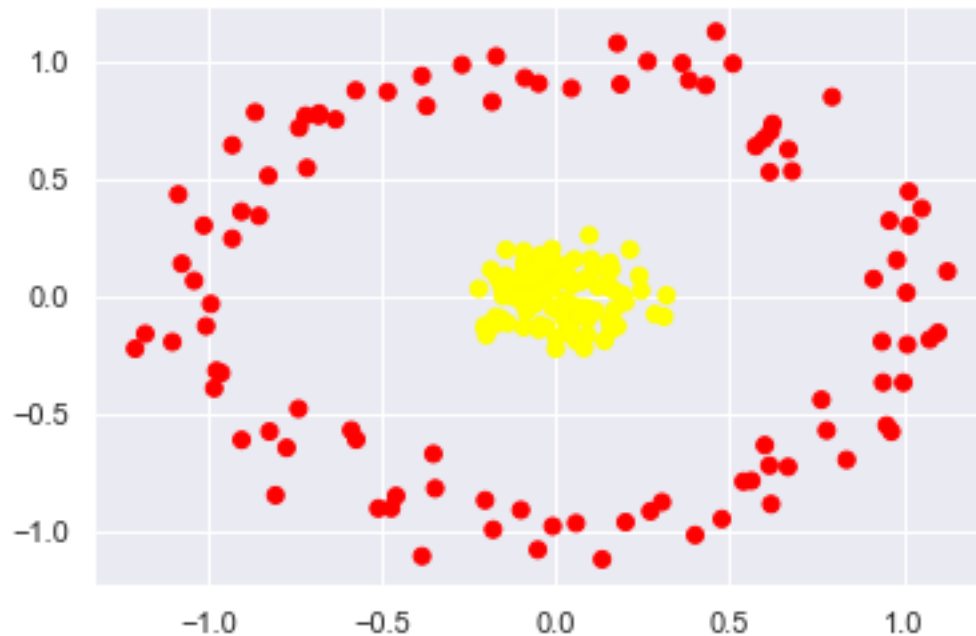
```
[75]: # lets create some non--linear dataset

from sklearn.datasets import make_circles

X , y = make_circles(200, factor =.1, noise=.1)

plt.scatter(X[:,0], X[:,1], c=y, cmap='autumn')
```

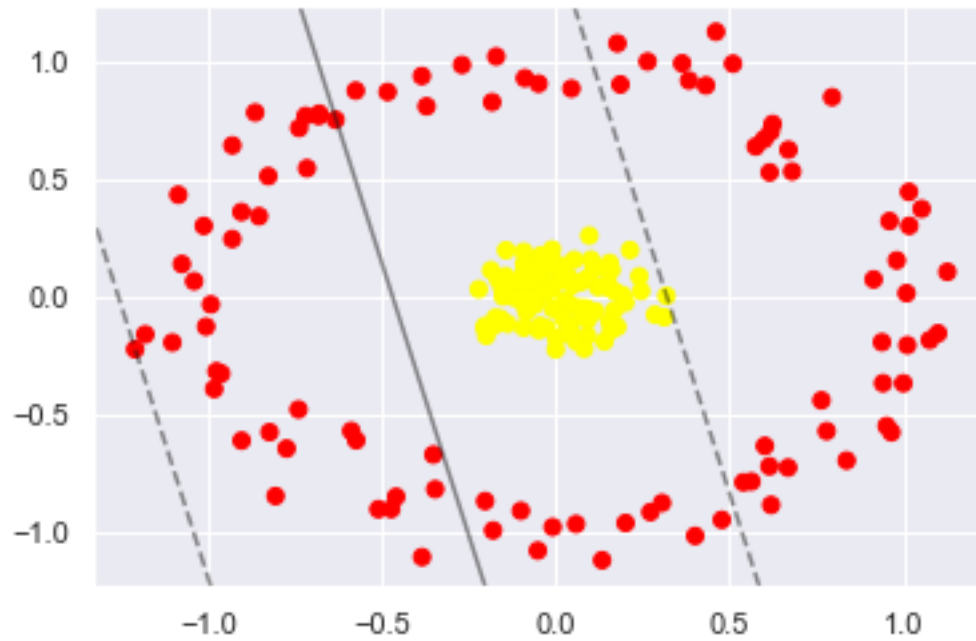
[75]: <matplotlib.collections.PathCollection at 0x7fd6f06c2a60>



```
[78]: # now we try to separate these dataset with a linear kernel

clf = SVC(kernel='linear').fit(X,y) # clf means "classifier linear function"

plt.scatter(X[:,0], X[:,1], c=y, cmap='autumn')
plot_svc_decision_function(clf, plot_support=False);
```



```
[72]: # our linear classifier is not succeeding in separating the data
```

```
# we need to define another kernel
```

```
# Radial basis function (circles)
```

```
r = np.exp(-(X**2).sum(1))
```

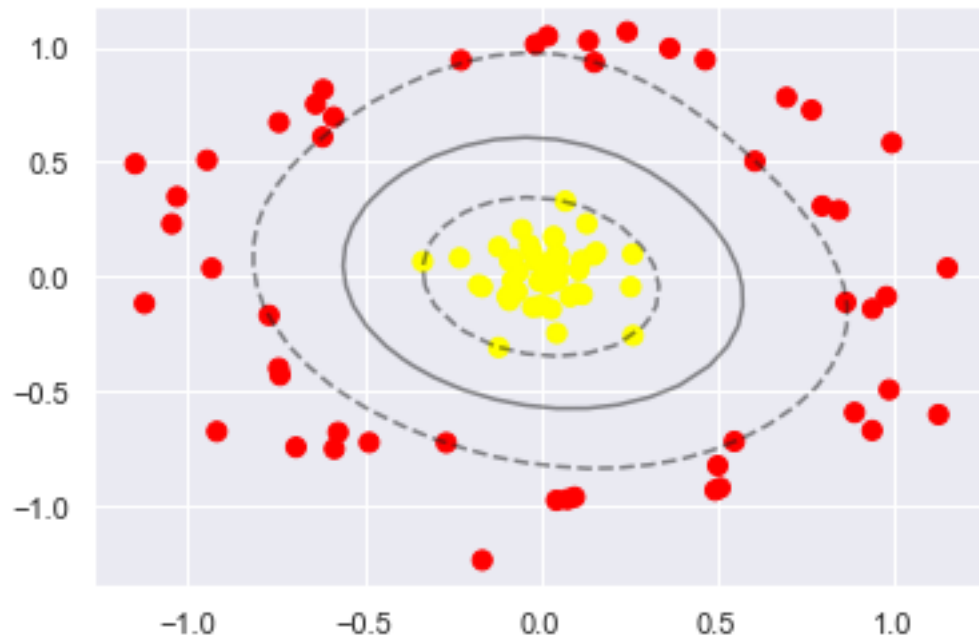
```
[73]: # now we create another kernel based on the RBF
```

```
rbf = SVC(kernel='rbf', C=1E6)
```

```
rbf.fit(X,y)
```

```
[73]: SVC(C=1000000.0)
```

```
[74]: plt.scatter(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
plot_svc_decision_function(rbf)
```



```
[79]: # https://scikit-learn.org/stable/modules/generated/sklearn.svm.SVC.html
```

```
[80]: # Decision Trees. Separation of many dimensional datasets
```

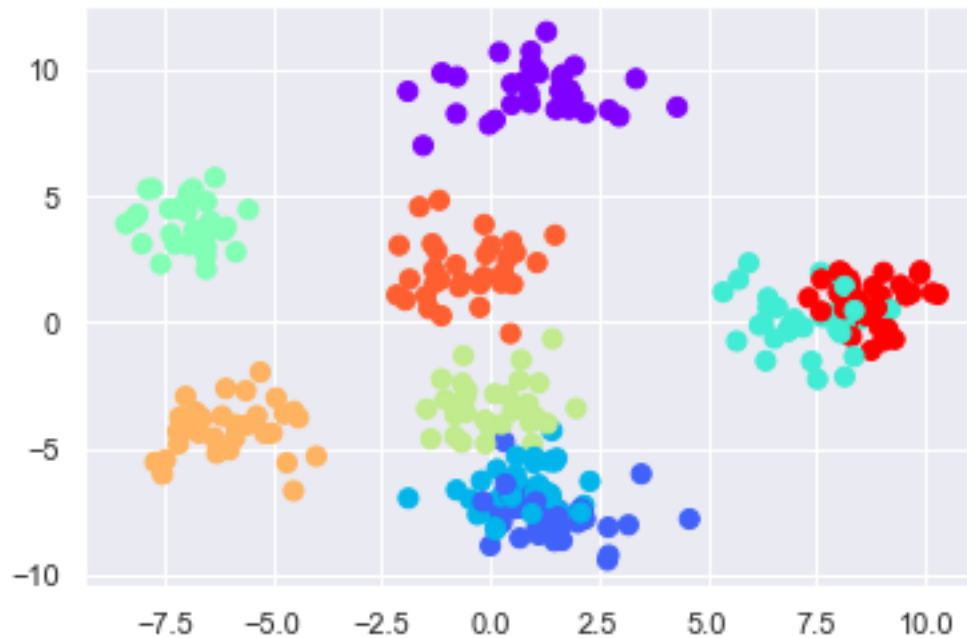
```
[81]: # dataset with more than 2 centers
```

```
[89]: from sklearn.datasets import make_blobs

X, y = make_blobs(n_samples=300,
                  centers=9,
                  cluster_std=1)

plt.scatter(X[:,0], X[:,1], c=y, s=50, cmap='rainbow')
```

```
[89]: <matplotlib.collections.PathCollection at 0x7fd6ef0fea90>
```



```
[90]: from sklearn.tree import DecisionTreeClassifier
tree = DecisionTreeClassifier().fit(X,y)
```

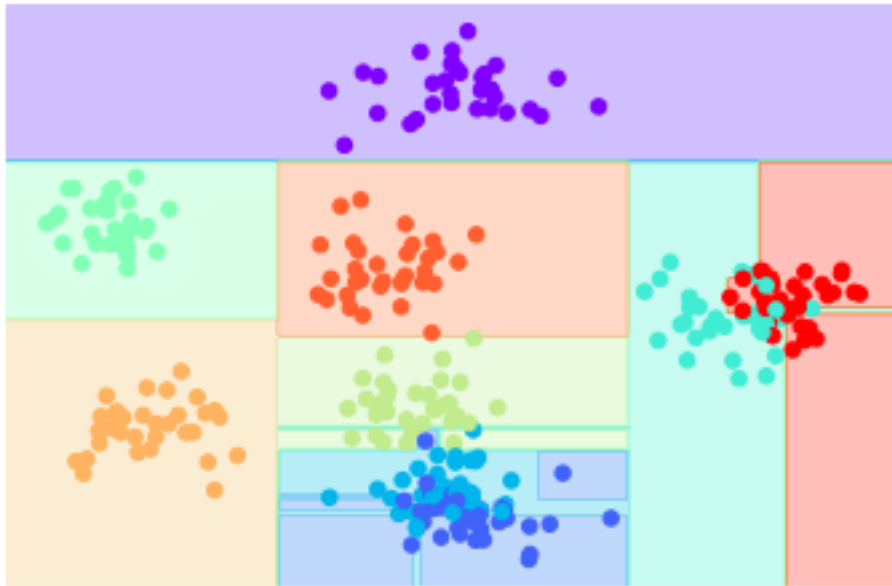
```
[91]: # visualization is not relevant for the exam
```

```
def visualize_classifier(model, X, y, ax=None, cmap='rainbow'):
    ax=ax or plt.gca()
    # Plot the training points
    ax.scatter(X[:, 0], X[:, 1], c=y, s=30, cmap=cmap,
               clim=(y.min(), y.max()), zorder=3)
    ax.axis('tight')
    ax.axis('off')
    xlim=ax.get_xlim()
    ylim=ax.get_ylim()
    # fit the estimator
    model.fit(X,y)
    xx, yy = np.meshgrid(np.linspace(*xlim, num=200),
                          np.linspace(*ylim, num=200))
    Z = model.predict(np.c_[xx.ravel(), yy.ravel()]).reshape(xx.shape)
    # create a color plot with the results
    n_classes = len(np.unique(y))
    contours=ax.contourf(xx,yy,Z, alpha=0.3,
                        levels=np.arange(n_classes+1)-0.5,
                        cmap=cmap, clim=(y.min(),y.max()),
                        zorder=1)
```

```
ax.set(xlim=xlim, ylim=ylim)
```

```
[92]: visualize_classifier(DecisionTreeClassifier(),X,y)
```

<ipython-input-91-da4776368e8b>:19: UserWarning: The following kwargs were not used by contour: 'clim'
contours=ax.contourf(xx,yy,Z, alpha=0.3,



```
[88]: # ABC - XYZ Analysis -- here we described several strategies depending on  
# the amount of value of products and the frequency of these products
```

```
[93]: # https://scikit-learn.org/stable/modules/generated/sklearn.tree.  
→DecisionTreeClassifier.html
```

```
[94]: # homework: find or create three dimensional dataset X (three columns) and y  
→(classes)  
# that is distributed linearly.  
# apply decision tree classifier to it and visualize the results
```

```
[ ]:
```