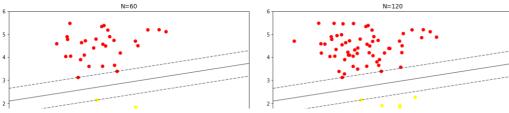
→ Part A: Generation of support vectors for a dataset

```
import numpy as np
import matplotlib.pyplot as plt
#Generate a dataset
from sklearn.datasets.samples_generator import make_blobs
X, y = make blobs(n samples=50, centers= 2, random state=0, cluster std=0.60)
plt.scatter(X[:,0],\ X[:,1],\ c=y,\ s=50,\ cmap='autumn')
 <matplotlib.collections.PathCollection at 0x7fa70c7e81d0>
      4
      3
      2
      0
         -0.5
               0.0
                    0.5
xfit= np.linspace(-1,3.5)
xfit
                        , -0.90816327, -0.81632653, -0.7244898 , -0.63265306,
     array([-1.
             -0.54081633, -0.44897959, -0.35714286, -0.26530612, -0.17346939,
                          0.01020408, 0.10204082, 0.46938776, 0.56122449,
            -0.08163265,
                                                     0.19387755,
                                                                  0.28571429,
             0.37755102,
                                                     0.65306122,
                                                                  0.74489796,
                                                     1.1122449 ,
             0.83673469,
                          0.92857143, 1.02040816,
                          1.3877551 ,
             1.29591837,
                                       1.47959184,
                                                     1.57142857,
                                                                  1.66326531.
                          1.84693878,
             1.75510204,
                                       1.93877551,
                                                     2.03061224,
                                                                  2.12244898,
                                                     2.48979592,
             2.21428571, 2.30612245,
                                       2.39795918,
                                                                  2.58163265,
             2.67346939,
                          2.76530612,
                                       2.85714286,
                                                     2.94897959,
             3.13265306, 3.2244898,
                                       3.31632653,
                                                    3.40816327,
                                                                 3.5
#Plotting lines
plt.scatter(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
for m,b in [(1,0.65), (0.5,1.6), (0.2, 2.9)]:
  #print(m,b)
  #print(xfit, m*xfit+b)
  plt.plot(xfit, m*xfit+b, '-k')
  plt.xlim(-1,3.5)
      -1.0 -0.5
                0.0
                     0.5
                          10
                               1.5
                                    2.0
                                              3.0
plt.scatter(X[:,0],\ X[:,1],\ c=y,\ s=50,\ cmap='autumn')
for m,b,d in [(1,0.65,0.33), (0.5,1.6,0.55), (0.2, 2.9, 0.2)]:
  #print(m,b)
  #print(xfit, m*xfit+b)
  yfit = m*xfit+b
  plt.plot(xfit,yfit,'-k')
  plt.fill_between(xfit, yfit-d, yfit+d, edgecolor= 'none', color='#AAAA', alpha=0.4)
  plt.xlim(-1,3.5)
```

```
#Create a SVM model
from sklearn.svm import SVC
model = SVC(kernel = 'linear', C=1E10)
model.fit(X,y)
     SVC(C=100000000000.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0, decision_function_shape='ovr', degree=3, gamma='scale',
         kernel='linear', max_iter=-1, probability=False, random_state=None,
         shrinking=True, tol=0.001, verbose=False)
        \verb|#define a function paralled plot_svc_decision_function|\\
{\tt def\ plot\_svc\_decision\_function(model,\ ax=None,\ plot\_support=True):}
  if ax is None:
                                              #get coordinate axis
    ax = plt.gca()
  xlim = ax.get_xlim()
  ylim = ax.get_ylim()
  #create grid to evaluate model
  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 = ['--', '-', '--'])
  #plot support vectors
  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)
\verb"plt.scatter"(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
plot_svc_decision_function(model)
      2
      1
         -0.5
                0.0
model.support_vectors_
     array([[0.44359863, 3.11530945],
             [2.33812285, 3.43116792],
[2.06156753, 1.96918596]])
def plot_svm(N=10, ax= None):
  X,y = make_blobs(n_samples=200, n_features=2, centers=2, cluster_std=0.60, random_state=0)
  X=X[:N]
  y=y[:N]
  model = SVC(kernel='linear', C=1E10)
  model.fit(X,y)
  ax=ax or plt.gca()
  \verb"ax.scatter"(X[:,0], X[:,1], c=y, cmap='autumn')
  ax.set xlim(-1,4)
  ax.set_ylim(-1,6)
  \verb"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,[60,120]):
  plot_svm(N,axi)
  axi.set_title('N={0}'.format(N))
```



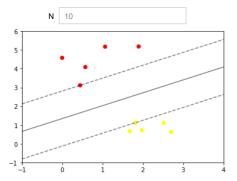
pip install ipywidgets

```
Requirement already satisfied: ipywidgets in /usr/local/lib/python3.7/dist-packages (7.6.3)
Requirement already satisfied: jupyterlab-widgets>=1.0.0 in /usr/local/lib/python3.7/dist-packages (from ipywidgets) (1.0.0)
Requirement already satisfied: hpykerlab=4.2.0 in /usr/local/lib/python3.7/dist-packages (from ipywidgets) (4.10.1) Requirement already satisfied: nbformat>=4.2.0 in /usr/local/lib/python3.7/dist-packages (from ipywidgets) (5.1.3)
Requirement already satisfied: traitlets>=4.3.1 in /usr/local/lib/python3.7/dist-packages (from ipywidgets) (5.0.5)
Requirement already satisfied: widgetsnbextension~=3.5.0 in /usr/local/lib/python3.7/dist-packages (from ipywidgets) (3.5.1) Requirement already satisfied: ipython>=4.0.0 in /usr/local/lib/python3.7/dist-packages (from ipywidgets) (5.5.0)
                                       jupyter-client in /usr/local/lib/python3.7/dist-packages (from ipykernel>=4.5.1->ipywidgets) (5.3.5)
Requirement already satisfied:
Requirement already satisfied: tornado>=4.0 in /usr/local/lib/python3.7/dist-packages (from ipykernel>=4.5.1->ipywidgets) (5.1.1) Requirement already satisfied: pickleshare in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (0.7.5)
Requirement already satisfied:
                                      pygments in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (2.6.1)
                                       prompt-toolkit<2.0.0,>=1.0.4 in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (1.0.18)
Requirement already satisfied:
Requirement already satisfied: simplegeneric>0.8 in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (0.8.1) Requirement already satisfied: setuptools>=18.5 in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (57.4.0)
                                      pexpect in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (4.8.0)
Requirement already satisfied:
Requirement already satisfied:
                                       decorator in /usr/local/lib/python3.7/dist-packages (from ipython>=4.0.0->ipywidgets) (4.4.2)
Requirement already satisfied: jsonschema!=2.5.0,>=2.4 in /usr/local/lib/python3.7/dist-packages (from nbformat>=4.2.0->ipywidgets) (2.6.0) Requirement already satisfied: ipython-genutils in /usr/local/lib/python3.7/dist-packages (from nbformat>=4.2.0->ipywidgets) (0.2.0)
Requirement already satisfied:
                                      jupyter-core in /usr/local/lib/python3.7/dist-packages (from nbformat>=4.2.0->ipywidgets) (4.7.1)
Requirement already satisfied:
                                      wcwidth \ in \ /usr/local/lib/python 3.7/dist-packages \ (from \ prompt-toolkit < 2.0.0, >= 1.0.4-) ipython >= 4.0.0-) ipywidgets) \ (0.2.5)
Requirement already satisfied: six>=1.9.0 in /usr/local/lib/python3.7/dist-packages (from prompt-toolkit<2.0.0,>=1.0.4->ipython>=4.0.0->ipywidgets) (1.15.0)
Requirement already satisfied:
                                      notebook>=4.4.1 in /usr/local/lib/python3.7/dist-packages (from widgetsnbextension~=3.5.0->ipywidgets) (5.3.1)
Requirement already satisfied: Send2Trash in /usr/local/lib/python3.7/dist-packages (from notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (1.8.0)
Requirement already satisfied: nbconvert in /usr/local/lib/python3.7/dist-packages (from notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (5.6.1)
Requirement already satisfied: jinja2 in /usr/local/lib/python3.7/dist-packages (from notebook>=4.4.1->widgetsnbextension==3.5.0->ipywidgets) (2.11.3)
                                       terminado>=0.8.1 in /usr/local/lib/python3.7/dist-packages (from notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (0.11.0)
Requirement already satisfied:
Requirement already satisfied: pyzmq>=13 in /usr/local/lib/python3.7/dist-packages (from jupyter-client->ipykernel>=4.5.1->ipywidgets) (22.2.1)

Requirement already satisfied: python-dateutil>=2.1 in /usr/local/lib/python3.7/dist-packages (from jupyter-client->ipykernel>=4.5.1->ipywidgets) (2.8.2)

Requirement already satisfied: ptyprocess in /usr/local/lib/python3.7/dist-packages (from terminado>=0.8.1->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets)
                                      MarkupSafe>=0.23 in /usr/local/lib/python3.7/dist-packages (from jinja2->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (2.
Requirement already satisfied:
Requirement already satisfied: pandocfilters>=1.4.1 in /usr/local/lib/python3.7/dist-packages (from nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidge
Requirement already satisfied: testpath in /usr/local/lib/python3.7/dist-packages (from nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (0.5.0)
Requirement already satisfied: defusedxml in /usr/local/lib/python3.7/dist-packages (from nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (0.7.1
Requirement already satisfied: mistune<2,>=0.8.1 in /usr/local/lib/python3.7/dist-packages (from nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets)
Requirement already satisfied: bleach in /usr/local/lib/python3.7/dist-packages (from nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets) (4.0.0)
Requirement already satisfied: entrypoints>=0.2.2 in /usr/local/lib/python3.7/dist-packages (from nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets
Requirement already satisfied: packaging in /usr/local/lib/python3.7/dist-packages (from bleach->nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidgets)
Requirement already satisfied: webencodings in /usr/local/lib/python3.7/dist-packages (from bleach->nbconvert->notebook>=4.4.1->widgetsnbextension~=3.5.0->ipywidge
Requirement already satisfied: pyparsing>=2.0.2 in /usr/local/lib/python3.7/dist-packages (from packaging->bleach->nbconvert->notebook>=4.4.1->widgetsnbextension~=
```

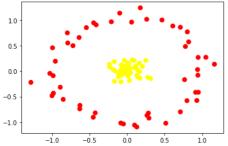
from ipywidgets import interact, fixed
interact(plot_svm, N=[10,50,100,150,200], ax=fixed(None));



from sklearn.datasets.samples_generator import make_circles X, y= make_circles(100, factor=.1, noise=.1)

 $\verb"plt.scatter"(X[:,0], X[:,1], c=y, s=50, cmap='autumn')$

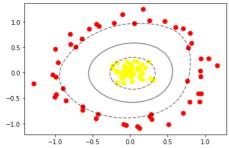
<matplotlib.collections.PathCollection at 0x7fa70b9e6050>



clf= SVC(kernel='linear').fit(X,y)
plt.scatter(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
plot svc decision function(clf,plot support = False)

```
1.0
       0.5
       0.0
      -0.5
r= np.exp(-(X**2).sum(1))
                -1.0
from \ mpl\_toolkits \ import \ mplot3d
def plot_3D(elev=30, azim=30, X=X, y=y):
  ax=plt.subplot(projection='3d')
  ax.scatter3D(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
  ax.view_init(elev=elev, azim=azim)
  ax.set_xlabel('X')
  ax.set_ylabel('Y')
  ax.set_zlabel('Z')
interact(plot_3D, elev=[-90,90], azip=(-180,180), X=fixed(X), y=fixed(y));
             elev -90
                                        30
            azim
clf=SVC(kernel='rbf', C=1E6)
clf.fit(X,y)
         tol=0.001, verbose=False)
```

```
SVC(C=1000000.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
plt.scatter(X[:,0], X[:,1], c=y, s=50, cmap='autumn')
plot_svc_decision_function(clf)
plt.scatter(clf.support_vectors_[:,0], clf.support_vectors_[:,1], lw=1, facecolors='none');
```



#PART B: Apllication of Svm for face recognition

 $from \ sklearn.datasets \ import \ fetch_lfw_people$ faces = fetch_lfw_people(min_faces_per_person=80) Downloading LFW metadata: https://ndownloader.figshare.com/files/5976012 Downloading LFW metadata: https://ndownloader.figshare.com/files/5976000
Downloading LFW metadata: https://ndownloader.figshare.com/files/5976006 Downloading LFW data (~200MB): https://ndownloader.figshare.com/files/5976015 print(faces.target_names) ['Colin Powell' 'Donald Rumsfeld' 'George W Bush' 'Gerhard Schroeder' 'Tony Blair'] print(faces.images.shape) (1140, 62, 47)

 ${\tt from \ sklearn.decomposition \ import \ PCA \ as \ Randomized PCA}$ from sklearn.pipeline import make_pipeline

```
model = make_pipeline(pca, svc)
from sklearn.model_selection import train_test_split
Xtrain,Xtest,ytrain,ytest = train_test_split(faces.data,faces.target,random_state=1)
model.fit(Xtrain,ytrain)
      Pipeline(memory=None,
                  steps=[('pca'
                           PCA(copy=True, iterated_power='auto', n_components=100, random_state=1, svd_solver='auto', tol=0.0, whiten=True)),
                          ('svc'
                            SVC(C=1.0, break_ties=False, cache_size=200,
class_weight='balanced', coef0=0.0,
decision_function_shape='ovr', degree=3, gamma='scale',
                                 kernel='rbf', max_iter=-1, probability=False,
                                random_state=None, shrinking=True, tol=0.001,
verbose=False))],
                  verbose=False)
y_pred =model.predict(Xtest)
from sklearn.metrics import classification_report
\verb|print(classification_report(ytest, y_pred, target_names=faces.target_names))| \\
```

	precision	recall	f1-score	support
Colin Powell	0.88	0.92	0.90	53
Donald Rumsfeld	0.76	0.76	0.76	21
George W Bush	0.88	0.94	0.91	139
Gerhard Schroeder	0.93	0.74	0.83	35
Tony Blair	0.88	0.76	0.81	37
accuracy			0.87	285
macro avg	0.86	0.82	0.84	285
weighted avg	0.87	0.87	0.87	285

```
target_names = faces.target_names
 _, h, w = faces.images.shape
#Visualization
import matplotlib.pyplot as plt
{\tt def\ plot\_gallery(images,\ titles,\ h,\ w,\ rows=3,\ cols=4):}
  plt.figure(figsize=(10,10))
  for i in range(rows*cols):
    plt.subplot(rows,cols,i+1)
     \verb|plt.imshow(images[i].reshape((h,w)),cmap=plt.cm.gray)|\\
     plt.title(titles[i])
     plt.xticks(())
     plt.yticks(())
\  \  \, \text{def titles(y\_pred, ytest, target\_names):}
  for i in range(y_pred.shape[0]):
    pred_name = target_names[y_pred[i]].split(' ')[-1]
true_name = target_names[ytest[i]].split(' ')[-1]
     yield 'predicted: {0}\ntrue: {1}'.format(pred_name, true_name)
prediction_titles = list(titles(y_pred, ytest, target_names))
```



plot_gallery(Xtest, prediction_titles, h, w)

predicted: Powell









true: Bush









