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CHỦ ĐỀ 4

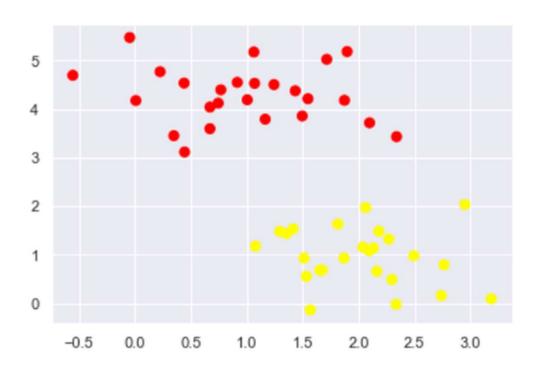
PHÂN LỚP (Thực hành SVM)

- Multi-class SVM?
 - Bản chất của SVM là chỉ thực hiện Binary classification
 - Mở rộng thành Multi-class SVM theo hai kiểu:
 - One-to-One
 - One-to-Rest
- Tìm hiểu các bước thực hiện SVM binary classification

```
%matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
from scipy import stats

# use seaborn plotting defaults
import seaborn as sns; sns.set()
```

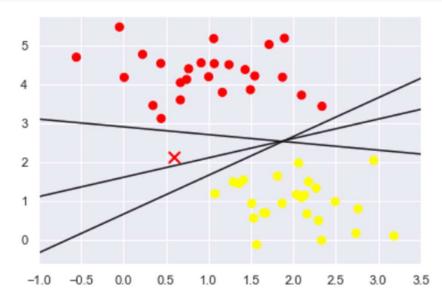
```
from sklearn.datasets import make_blobs
X, y = make_blobs(n_samples=50, centers=2, random_state=0, cluster_std=0.60)
plt.scatter(X[:, 0], X[:, 1], c=y, s=50, cmap='autumn');
```

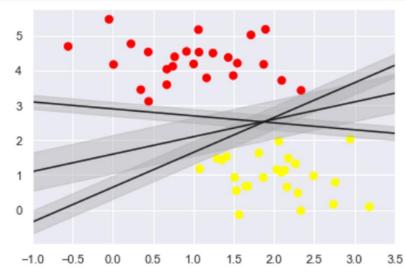


```
xfit = np.linspace(-1, 3.5)
plt.scatter(X[:, 0], X[:, 1], c=y, s=50, cmap='autumn')
plt.plot([0.6], [2.1], 'x', color='red', markeredgewidth=2, markersize=10)

for m, b in [(1, 0.65), (0.5, 1.6), (-0.2, 2.9)]:
    plt.plot(xfit, m * xfit + b, '-k')

plt.xlim(-1, 3.5);
```





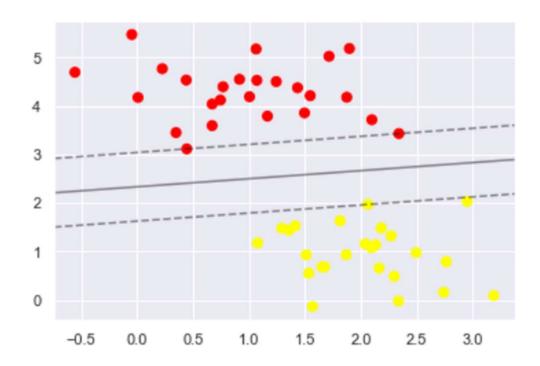
Tìm hiểu các bước thực hiện SVM binary classification

```
from sklearn.svm import SVC # "Support vector classifier"
model = SVC(kernel='linear', C=1E10)
model.fit(X, y)
```

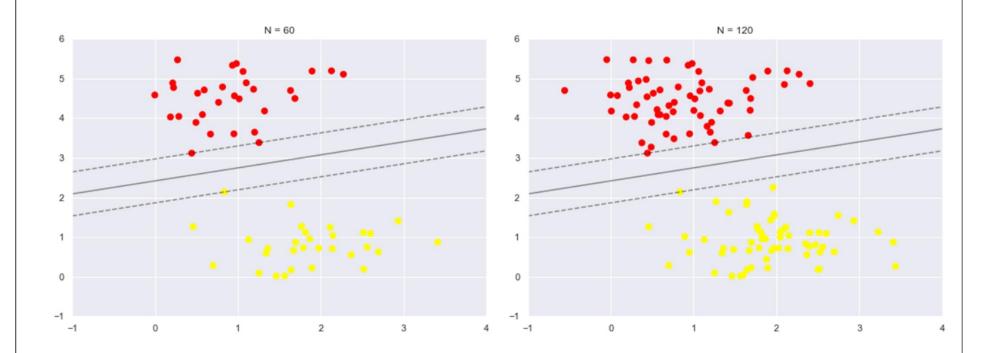
SVC(C=100000000000.0, kernel='linear')

```
def plot svc decision function(model, ax=None, plot support=True):
    """Plot the decision function for a 2D SVC"""
    if ax is None:
        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)
   # plot decision boundary and margins
   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)
```

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



```
def plot svm(N=10, ax=None):
   X, y = make_blobs(n_samples=200, centers=2,
                      random state=0, cluster std=0.60)
   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 vlim(-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, 120]):
    plot svm(N, axi)
    axi.set title('N = {0}'.format(N))
```



Úng dụng thực tế của SVM binary classification

```
#Import scikit-learn dataset library
from sklearn import datasets
#Load dataset
cancer = datasets.load_breast_cancer()

# print the names of the 13 features
print('Features: ', cancer.feature_names)
# print the label type of cancer('malignant' 'benign')
print('Labels: ', cancer.target_names)
```

```
Features: ['mean radius' 'mean texture' 'mean perimeter' 'mean area' 'mean smoothness' 'mean compactness' 'mean concavity' 'mean concave points' 'mean symmetry' 'mean fractal dimension' 'radius error' 'texture error' 'perimeter error' 'area error' 'smoothness error' 'compactness error' 'concavity error' 'concave points error' 'symmetry error' 'fractal dimension error' 'worst radius' 'worst texture' 'worst perimeter' 'worst area' 'worst smoothness' 'worst compactness' 'worst concavity' 'worst concave points' 'worst symmetry' 'worst fractal dimension'] Labels: ['malignant' 'benign']
```

• Ứng dụng thực tế của SVM binary classification

```
cancer.data.shape (569, 30)
```

```
# Import train_test_split function
from sklearn.model_selection import train_test_split# Split dataset into training set and test set
X_train, X_test, y_train, y_test = train_test_split(cancer.data, cancer.target,
test_size=0.3,random_state=109) # 70% training and 30% test
```

```
#Import svm model
from sklearn import svm
#Create a svm Classifier
clf = svm.SVC(kernel='linear') # Linear Kernel
#Train the model using the training sets
clf.fit(X_train, y_train)
#Predict the response for test dataset
y_pred = clf.predict(X_test)
```

Úng dụng thực tế của SVM binary classification

```
#Import scikit-learn metrics module for accuracy calculation
from sklearn import metrics
# Model Accuracy: how often is the classifier correct?
print('Accuracy:',metrics.accuracy_score(y_test, y_pred))
```

Accuracy: 0.9649122807017544

```
# Model Precision: what percentage of positive tuples are labeled as such?
print('Precision:',metrics.precision_score(y_test, y_pred))
# Model Recall: what percentage of positive tuples are labelled as such?
print('Recall:',metrics.recall_score(y_test, y_pred))
```

Precision: 0.9811320754716981 Recall: 0.9629629629629

Q/A