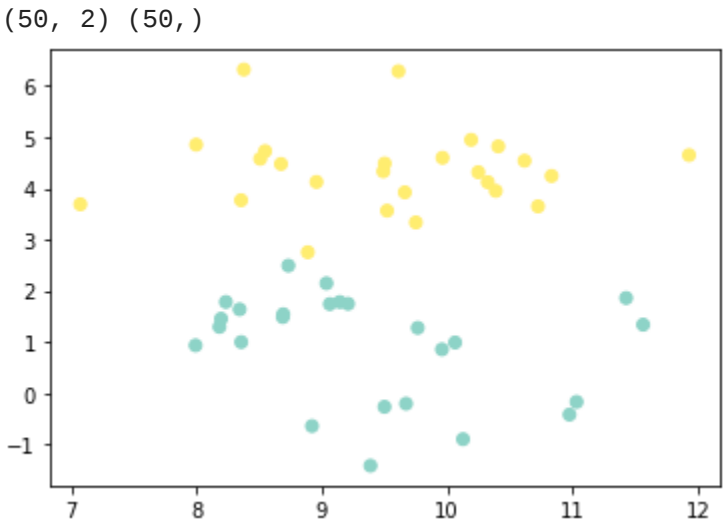


WEEK09_SVM_synthetic_dataset,_iris_dataset

```
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)
```

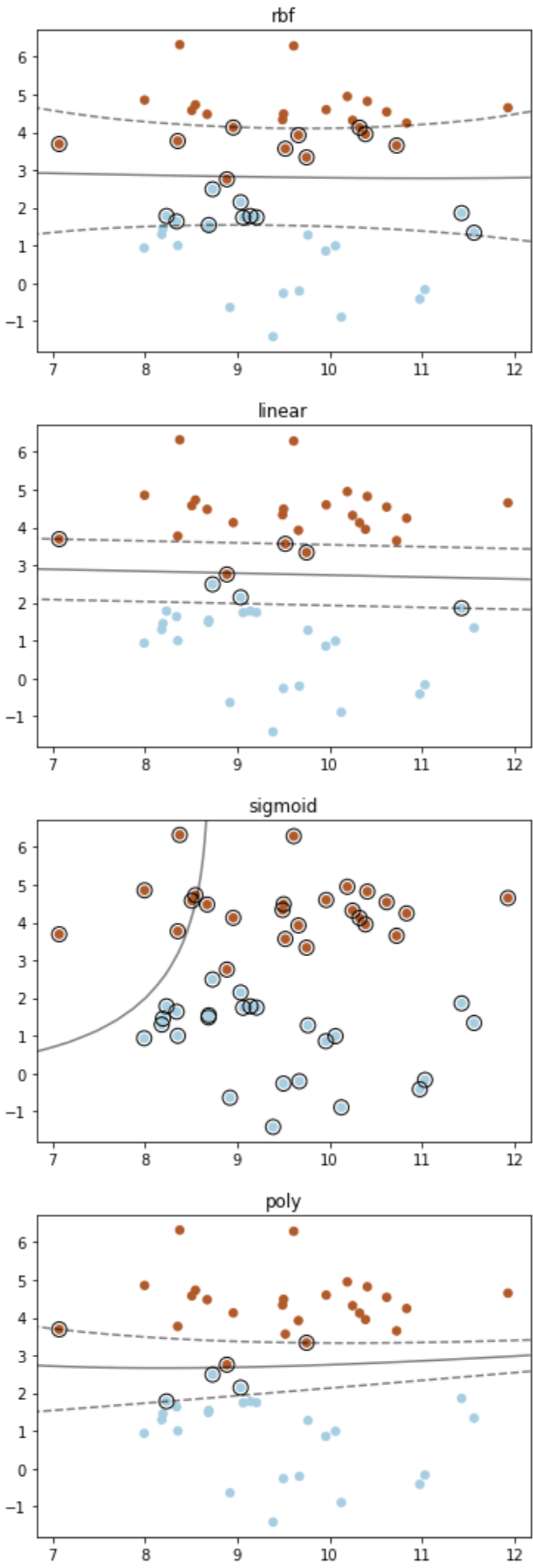


```
In [28]: for x in {'linear', 'poly', 'rbf', 'sigmoid'}:
# fit the model, don't regularize for illustration purposes
clf = SVC(kernel=x, C=0.70)
clf.fit(X, y)

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()
```



```
In [4]: # import some data to play with
iris = datasets.load_iris()
#iris
```

```
In [5]: iris.data.shape
```

Out[5]: (150, 4)

```
In [6]: # 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_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