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In [1]: 1 # Handling Linearly Inseparable Classes Using Kernels
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In [2]: 1 # Importing Libraries
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In [3]: 1 from sklearn.svm import SVC
2 from sklearn import datasets
3 from sklearn.preprocessing import StandardScaler
4 import numpy as np
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In [4]: 1 # Set randomization seed
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In [4]: 1 np.random.seed(0)
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In [6]: 1 # Generate two features
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In [5]: 1 features = np.random.randn(200, 2)
```

```
In [8]: 1 # Use a XOR gate to generate linearly inseparable classes
```

```
In [6]: 1 target_xor = np.logical_xor(features[:, 0] > 0, features[:, 1] > 0)
2 target = np.where(target_xor, 0, 1)
```

```
In [7]: 1 # Create a support vector machine with a radial basis function kernel
2 svc = SVC(kernel="rbf", random_state=0, gamma=1, C=1)
```

```
In [8]: 1 # Train the classifier
2 model = svc.fit(features, target)
```

```
In [13]: 1 #Plot observations and decision boundary hyperplane
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```
In [9]: 1 from matplotlib.colors import ListedColormap
2 import matplotlib.pyplot as plt
```

```
In [11]: 1 def plot_decision_regions(X, y, classifier):
2         cmap = ListedColormap(("red", "blue"))
3         xx1, xx2 = np.meshgrid(np.arange(-3, 3, 0.02), np.arange(-3, 3, 0.02))
4         Z = classifier.predict(np.array([xx1.ravel(), xx2.ravel()]).T)
5         Z = Z.reshape(xx1.shape)
6         plt.contourf(xx1, xx2, Z, alpha=0.1, cmap=cmap)
7         for idx, cl in enumerate(np.unique(y)):
8             plt.scatter(x=X[y == cl, 0], y=X[y == cl, 1],
9                         alpha=0.8, c=cmap(idx),
10                        marker="+", label=cl)
```

```
In [12]: 1 # Create support vector classifier with a linear kernel
2         svc_linear = SVC(kernel="linear", random_state=0, C=1)
```

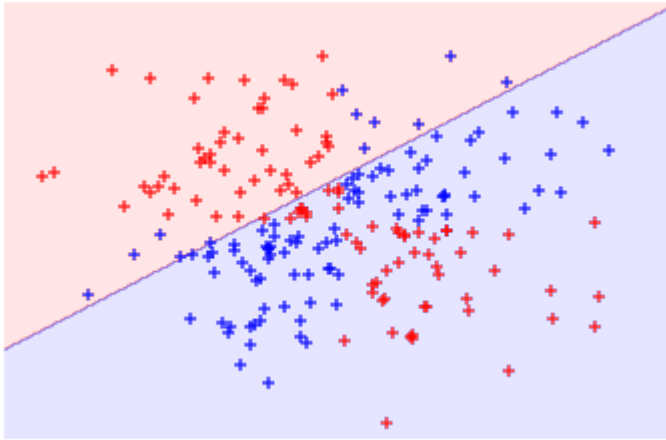
```
In [13]: 1 # Train model
2         svc_linear.fit(features, target)
```

```
Out[13]: SVC(C=1, kernel='linear', random_state=0)
```

```
In [20]: 1 # Plot observations and hyperplane
2 plot_decision_regions(features, target, classifier=svc_linear)
3 plt.axis("off"); plt.show();
```

\*c\* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with \*x\* & \*y\*. Please use the \*color\* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points.

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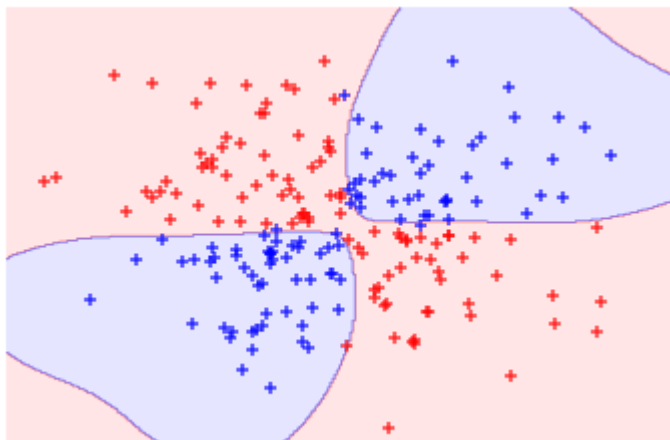
```
In [14]: 1 # Create a support vector machine with a radial basis function kernel
2 svc = SVC(kernel="rbf", random_state=0, gamma=1, C=1)
```

```
In [15]: 1 # Train the classifier
2 model = svc.fit(features, target)
```

```
In [23]: 1 plot_decision_regions(features, target, classifier=svc)
2 plt.axis("off"), plt.show();
```

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In [ ]: 1
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