

## ✓ Worksheet 15

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### Topics

- Support Vector Machines

### Support Vector Machines

a) Follow along in class to implement the perceptron algorithm and create an animation of the algorithm.

```
1  import numpy as np
2  from PIL import Image as im
3  import matplotlib.pyplot as plt
4  import sklearn.datasets as datasets
5
6  TEMPFILE = "temp.png"
7  CENTERS = [[0, 1], [1, 0]]
8
9  # Dataset
10 X, labels = datasets.make_blobs(n_samples=10, centers=CENTERS, cluster_std=0.2, random_state=0)
11 Y = np.array(list(map(lambda x : -1 if x == 0 else 1, labels.tolist())))
12
13 # Initializing w and b
14 w = np.array([1, 1])
15 b = 0.1
16
17 # Perceptron Parameters
18 epochs = 100
19 alpha = .05
20 expanding_rate = .99
21 retracting_rate = 1.1
22
23 def snap(x, w, b, error):
24     """
25     Plot the street induced by w and b.
26     Circle the point x in red if it was
27     misclassified or in yellow if it was
28     classified correctly.
29     """
30
31     xplot = np.linspace(-3, 3)
32     cs = np.array([x for x in 'gb'])
33
34     svm = - (w[1]/w[0]) * xplot - b/w[0]
35     left_svm = - (1/w[0]) - (w[1]/w[0]) * xplot - b/w[0]
36     right_svm = (1/w[0]) - (w[1]/w[0]) * xplot - b/w[0]
37
38     fig, ax = plt.subplots()
39     ax.scatter(X[:,0],X[:,1],color=cs[labels].tolist(), s=50, alpha=0.8)
40     if error:
41         ax.add_patch(plt.Circle((x[0], x[1]), .2, color='r',fill=False))
42     else:
43         ax.add_patch(plt.Circle((x[0], x[1]), .2, color='y',fill=False))
44     ax.plot(xplot, left_svm, 'g--', lw=2)
45     ax.plot(xplot, svm, 'r-', lw=2)
46     ax.plot(xplot, right_svm, 'b--', lw=2)
47     ax.set_xlim(min(X[:, 0]) - 1, max(X[:,0]) + 1)
48     ax.set_ylim(min(X[:, 1]) - 1, max(X[:,1]) + 1)
49     fig.savefig(TEMPFILE)
50     plt.close()
51
52     return im.fromarray(np.asarray(im.open(TEMPFILE)))
```

```

52     return im:FormatRay(np.asarray(img).open('r').file)
53
54
55 images = []
56 for _ in range(epochs):
57     # pick a point from X at random
58     i = np.random.randint(0, len(X))
59     x, y = X[i], Y[i]
60     error = False
61
62     y_predict = np.dot(w, x) + b
63     if not((y < 0 and y_predict < 0) or (y > 0 and y_predict > 0)):
64         # misclassified
65         error = True
66
67         # move in the direction of the point
68         w = w + y * x * alpha
69         b = b + y * alpha
70
71         #expand
72         w = w * expanding_rate
73         b = b * expanding_rate
74     else:
75         # are you in the street
76         if (y_predict < 1) and (y_predict > -1):
77             # yes
78             w = w * retracting_rate
79             b = b * retracting_rate
80
81
82
83     images.append(snap(x, w, b, error))
84
85 images[0].save(
86     'svm.gif',
87     optimize=False,
88     save_all=True,
89     append_images=images[1:],
90     loop=0,
91     duration=100
92 )

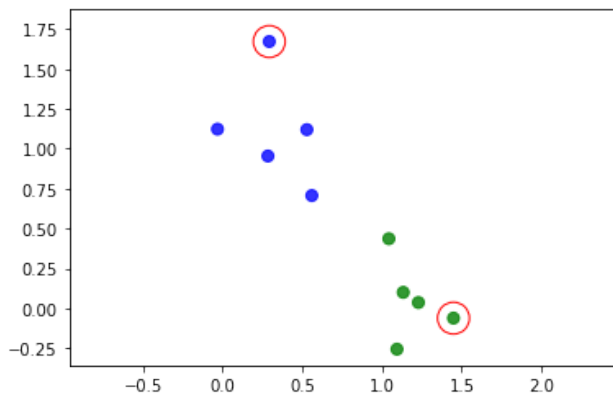
```

b) Consider the following dataset:

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 import sklearn.datasets as datasets
4
5 centers = [[0, 1], [1, 0]]
6 X, _ = datasets.make_blobs(n_samples=10, centers=centers, cluster_std=0.3, random_state=0)
7 Y = np.array([1 if x[0] - x[1] >= 0 else 0 for x in X])
8
9 cs = np.array([x for x in 'bg'])
10 _, ax = plt.subplots()
11 ax.scatter(X[:,0],X[:,1],color=cs[Y].tolist(), s=50, alpha=0.8)
12 ax.set_aspect('equal', adjustable='datalim')
13 ax.add_patch(plt.Circle((X[0][0], X[0][1]), .1, color='r',fill=False))
14 ax.add_patch(plt.Circle((X[1][0], X[1][1]), .1, color='r',fill=False))
15 plt.show()

```



if we fit an SVM to the above dataset, moved the points circled in red, and re-fit the SVM, describe how the fit would change depending on how the points are moved.

Since the two points are not support vectors, the fit would not cause much of a change unless they are moved to become the new support vectors of the graph.

c) If we were to fit an SVM to the above dataset, which points do you think would affect the decision boundary the most? Circle them in red.

The two points that are in the center (the support vectors - the blue and green points that are closest to each other), would affect the decision boundary the most. Below is a photo of the points.

```

1  import numpy as np
2  import matplotlib.pyplot as plt
3  import sklearn.datasets as datasets
4  import matplotlib.patches as patches
5
6  centers = [[0, 1], [1, 0]]
7  X, _ = datasets.make_blobs(n_samples=10, centers=centers, cluster_std=0.3, random_state=0)
8  Y = np.array([1 if x[0] - x[1] >= 0 else 0 for x in X])
9
10 cs = np.array([x for x in 'bg'])
11 _, ax = plt.subplots()
12 ax.scatter(X[:,0],X[:,1],color=cs[Y].tolist(), s=50, alpha=0.8)
13 ax.set_aspect('equal', adjustable='datalim')
14 circleOne = patches.Circle((1.05, 0.43), radius=0.05, edgecolor='r', facecolor='none')
15 ax.add_patch(circleOne)
16 circleTwo = patches.Circle((0.55, 0.7), radius=0.05, edgecolor='r', facecolor='none')
17 ax.add_patch(circleTwo)
18 plt.show()

```

