EdX and its Members use cookies and other tracking technologies for performance, analytics, and marketing purposes. By using this website, you accept this use. Learn more about these technologies in the <a href="Privacy Policy">Privacy Policy</a>.





<u>Unit 2 Nonlinear Classification,</u> <u>Linear regression, Collaborative</u>

Course > Filtering (2 weeks)

> Homework 3 > 4. Kernels-II

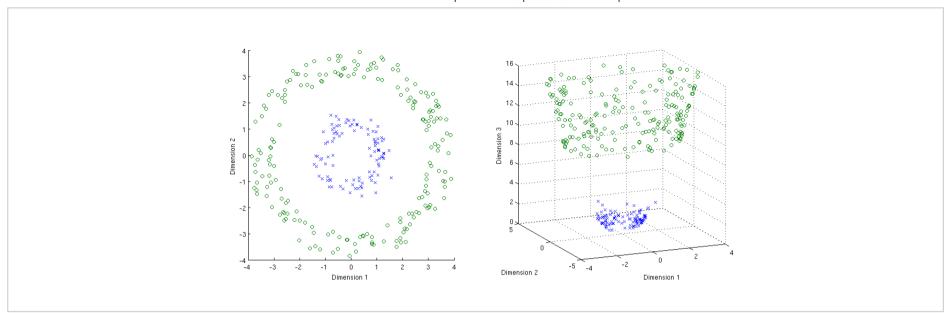
## 4. Kernels-II

In this question, we will practice some specific kernel methods.

## 4. (a)

2/2 points (graded)

In the figure below, a set of points in 2-D is shown on the left. On the right, the same points are shown mapped to a 3-D space via some transform  $\phi(x)$ , where x denotes a point in the 2-D space. Notice that  $\phi(x)_1 = x_1$  and  $\phi(x)_2 = x_2$ , or in other words, the first and second coordinates are unchanged by the transformation.



Which of the following functions could have been used to compute the value of the 3rd coordinate,  $\phi(x)_3$  for each point?

$$\circ \ \phi(x)_3 = x_1 + x_2$$

$$ullet \phi(x)_3 = x_1^2 + x_2^2 \, ullet$$

$$\circ \ \phi(x)_3 = x_1x_2$$

$$\circ \ \phi(x)_3 = x_1^2 - x_2^2$$

Think about how a linear decision boundary in the 3 dimensional space ( $\{\phi \in \mathbb{R}^3: \theta \cdot \phi + \theta_0 = 0\}$ ) might appear in the original 2 dimensional space.

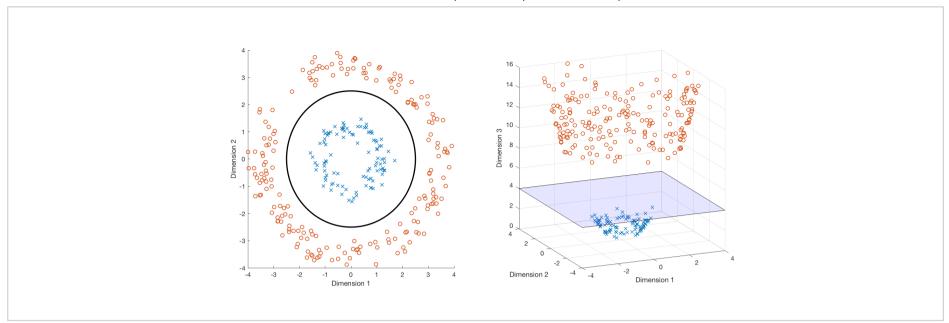
For example, suppose the decision boundary in the 3 dimensional space is z=4.

Provide an equation  $f(x_1, x_2) = 0$  in the 2 dimensional space such that all the points  $(x_1, x_2)$  with  $f(x_1, x_2) > 0$  correspond to z > 4 in the 3 dimensional space.

$$f(x_1,x_2) = 0 = oxed{x_1^2 + x_2^2 - 4}$$
 Answer: x\_1^2 + x\_2^2 - 4

#### **Solution:**

- With  $x=[x_1;x_2]$ , one mapping which could satisfy the mapping is  $\phi(x)_3=x_1^2+x_2^2$ . The decision boundary is shown below.
- ullet As a result, the decision boundary at z=4 corresponds to  $x_1^2+x_2^2=4$



Submit

You have used 1 of 3 attempts

**1** Answers are displayed within the problem

# 4. (b)

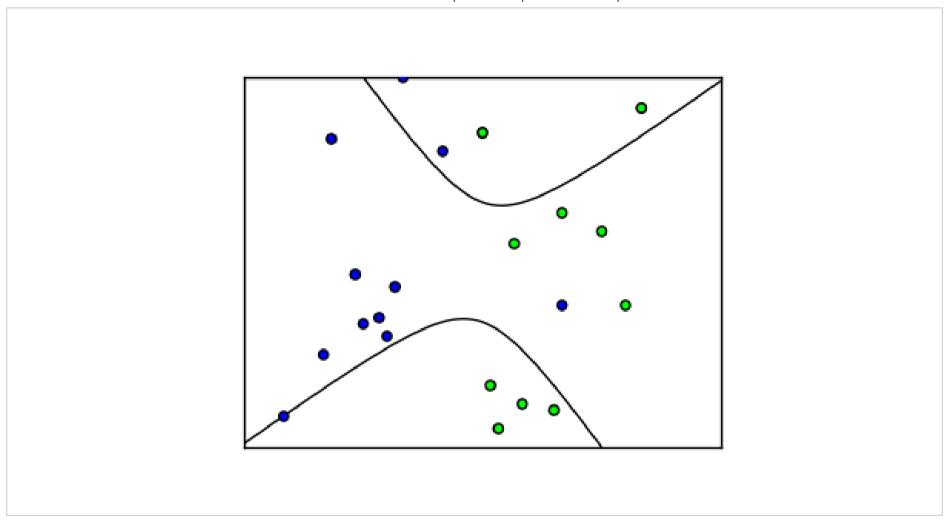
5/5 points (graded)

Consider fitting a kernelized SVM to a dataset  $(x^{(i)},y^{(i)})$  where  $x^{(i)}\in\mathbb{R}^2$  and  $y^{(i)}\in\{1,-1\}$  for all  $i=1,\ldots,n$ . To fit the parameters of this model, one computes  $\theta$  and  $\theta_0$  to minimize the following objective:

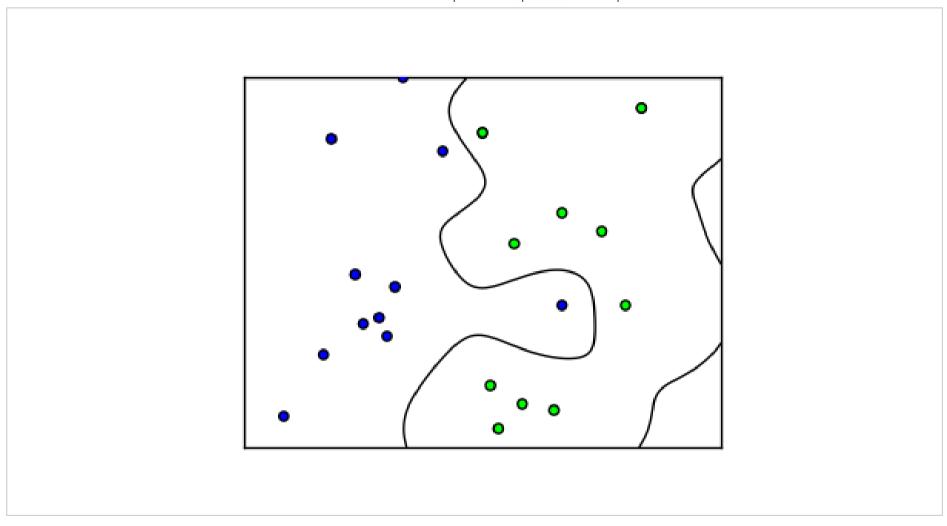
$$L\left( heta, heta_0
ight) = rac{1}{n} \sum_{i=1}^n \mathrm{Loss}_h\left(y^{(i)}\left( heta \cdot \phi\left(x^{(i)}
ight) + heta_0
ight)
ight) + rac{\lambda}{2} \| heta\|^2$$

where  $\phi$  is the feature vector associated with the kernel function. Note that, in a kernel method, the optimization problem for training would be typically expressed solely in terms of the kernel function K(x,x') (dual) rather than using the associated feature vectors  $\phi(x)$  (primal). We use the primal only to highlight the classification problem solved.

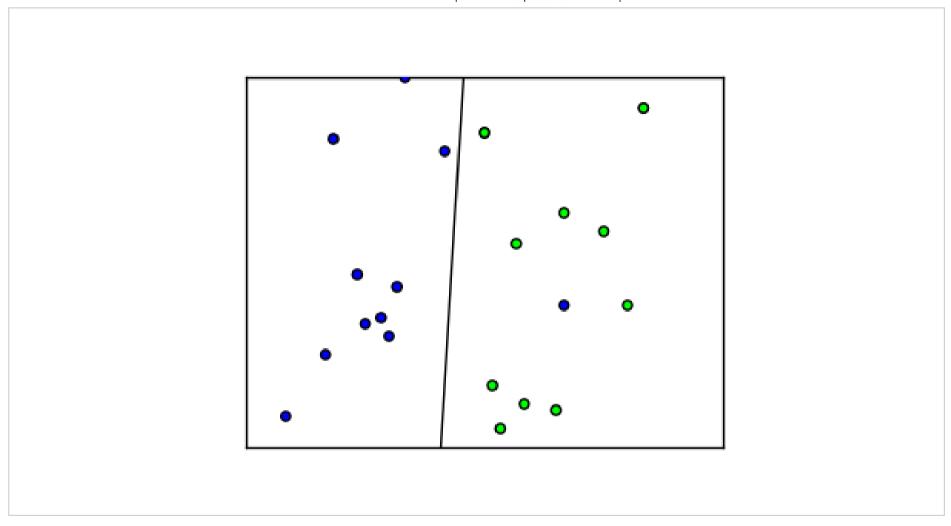
The plots below show 4 different kernelized SVM models estimated from the same 11 data points. We used a different kernel to obtain each plot but got confused about which plot corresponds to which kernel. Help us out by assigning each plot to one of the following models: linear kernel, quadratic kernel, order 3 kernel, and RBF kernel.



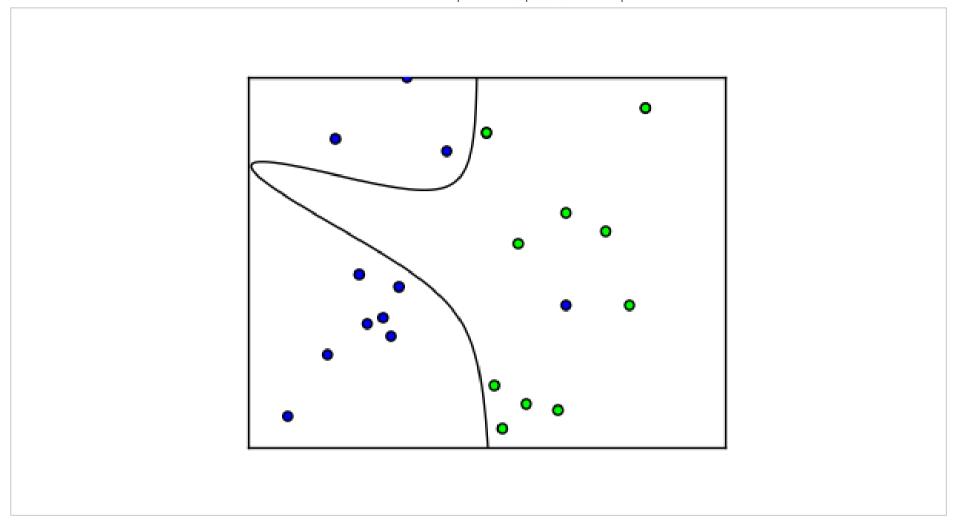
• quadratic kernel		
order 3 kernel		
RBF kernel		



quadratic kernel		
order 3 kernel		
<ul><li></li></ul>		



quadratic kernel		
order 3 kernel		
RBF kernel		



o quadratic kernel		
order 3 kernel   ✓		
RBF kernel		

How would you describe qualitatively how the resulting classifiers vary with the value of  $\lambda$ ? If the value of  $\lambda$  is increased, the fitting of model would be

- better fit on training data (sharper decision boundary)
- worse fit on training data (flatter decision boundary)

### **Solution:**

- From examining the number of bends in the decision boundaries:
- 3rd plot corresponds to the linear kernel.
- 1st plot corresponds to the quadratic kernel.
- 4th plot corresponds to the 3rd-order kernel.

- 2nd plot corresponds to the Gaussian RBF kernel.
- Large  $\lambda$  penalty on  $\theta$  results in flatter/ less "squiggly" lines.

Submit

You have used 1 of 2 attempts

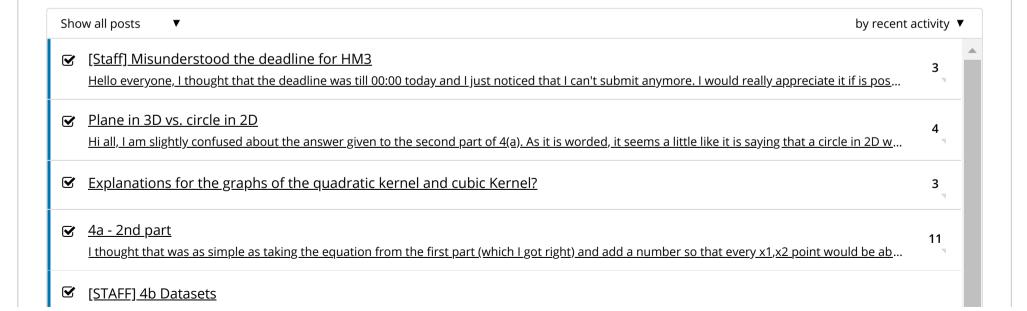
• Answers are displayed within the problem

### Discussion

**Topic:** Unit 2 Nonlinear Classification, Linear regression, Collaborative Filtering (2 weeks):Homework 3 / 4. Kernels-II

Add a Post

**Hide Discussion** 



	I found the right answers by guessing but I have no understanding of how it works. Can you provide us the numerical datasets of these example	2	
2	<u>Understanding RBF</u>	2	
Q	Very disappointing that people are trying to cheat at this graduate level class from MITx how will they do in their job even if they successfully cheat here? very very immature thinking	2	
€	4b: Matching the graphs. I guessed them correctly. But I don't really understand why they are right.  I'm mostly confused about the quadratic kernel, since it doesn't look like a "normal" quadratic graph. I still don't understand why the graph matc	4	
€	How can I understand order 3 kernel  I checked the property of quadratic, RBF and linear but I cannot understand how to identify the order 3. Is it because the boundary is pushed aw	3	
€	further explanation needed on $f(x_1, x_2)=0$	7	
2	[STAFF] [URGENT] It's rather disappointing that someone tried to cheat their way through:/ So as I search for the word "order 3" and then go to google images to have a visual image of what order 3 is. I stumbled upon an image of the ex	9	
₹	Finished Question but No Green Check  Title says it all - I answered everything but still don't have a green check at the top for this question. Could someone from the staff please confir	2	
2	[staff] Some wording / expression cleanup for 4b  Community TA	2	
<b>▽</b>	4(h)		•

© All Rights Reserved