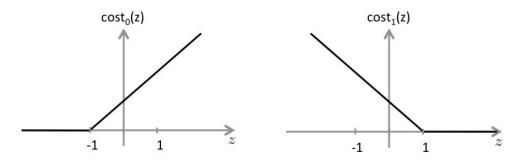
## HMC CS 158

## Quiz 5: SVMs, Kernels

1. We can think of the SVM as solving

$$\min_{\boldsymbol{\theta}} C \sum_{i=1}^{n} \left( y^{(i)} \text{cost}_{1} \left( \boldsymbol{\theta}^{T} \boldsymbol{x}^{(i)} \right) + \left( 1 - y^{(i)} \right) \text{cost}_{0} \left( \boldsymbol{\theta}^{T} \boldsymbol{x}^{(i)} \right) \right) + \sum_{j=1}^{d} \theta_{j}^{2},$$

where the cost functions  $cost_0(z)$  and  $cost_1(z)$  look like this:

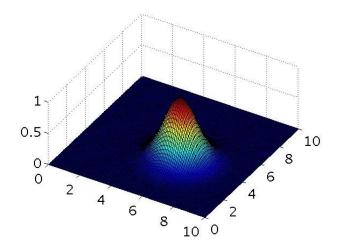


The first term in the objective is  $C\sum_{i=1}^{n} (y^{(i)} \cot_1(\boldsymbol{\theta}^T \boldsymbol{x}^{(i)}) + (1 - y^{(i)}) \cot_0(\boldsymbol{\theta}^T \boldsymbol{x}^{(i)}))$ . This first term will be zero if two of the following four conditions hold true. Which are the two conditions that would guarantee that this term equals zero?

- (a) For every example with  $y^{(i)} = 0$ , we have that  $\boldsymbol{\theta}^T \boldsymbol{x}^{(i)} \leq 0$ .
- (b) For every example with  $y^{(i)} = 1$ , we have that  $\boldsymbol{\theta}^T \boldsymbol{x}^{(i)} \geq 1$ .
- (c) For every example with  $y^{(i)} = 1$ , we have that  $\boldsymbol{\theta}^T \boldsymbol{x}^{(i)} \geq 0$ .
- (d) For every example with  $y^{(i)} = 0$ , we have that  $\boldsymbol{\theta}^T \boldsymbol{x}^{(i)} \leq -1$ .

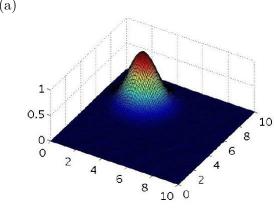
This quiz is adapted from course material by Andrew Ng (Stanford).

2. The formula for the Gaussian kernel is given by the similarity  $(x, z) = \exp\left(-\frac{||x-z||^2}{2\sigma^2}\right)$ . The figure below shows a plot of  $f = \text{similarity}(\boldsymbol{x}, \boldsymbol{z})$  when  $\sigma^2 = 1$ .

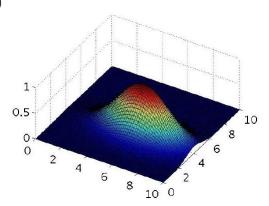


Which of the following is a plot of f when  $\sigma^2 = 0.25$ ?

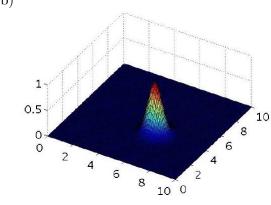
(a)



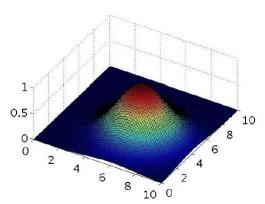
(c)



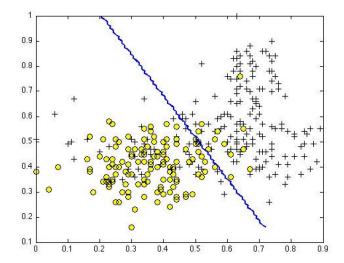
(b)



(d)



3. Suppose you have trained an SVM classifier with a Gaussian kernel, and it learned the following decision boundary on the training set:



You suspect that the SVM is underfitting your dataset. Should you try increasing or decreasing C? Increasing or decreasing  $\sigma^2$ ?

- 4. Suppose you have a dataset with d = 10 features and n = 5000 examples. After training your logistic regression classifier with gradient descent, you find that it has underfit the training set and does not achieve the desired performance on the training or cross-validation sets. Which of the following might be promising steps to take? Check all that apply.
  - (a) Create / add new polynomial features.
  - (b) Use an SVM with a Gaussian kernel.
  - (c) Use a different optimization method since using gradient descent to train logistic regression might result in a local minimum.
  - (d) Increase the regularization parameter  $\lambda$ .