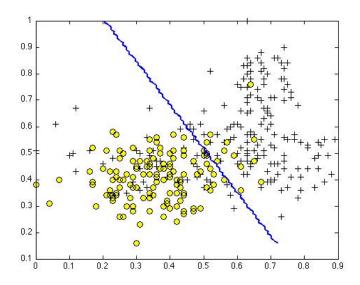
### Week 7 Support Vector Machines

### Question 1

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  $\sigma^2$ ?

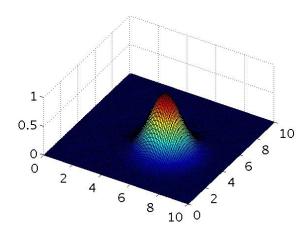
- It would be reasonable to try decreasing C. It would also be reasonable to try decreasing  $\sigma^2$ .
- CORRECT It would be reasonable to try increasing C. It would also be reasonable to try decreasing  $\sigma^2$ .
- It would be reasonable to try increasing C. It would also be reasonable to try increasing  $\sigma^2$ .
- It would be reasonable to try decreasing C. It would also be reasonable to try increasing  $\sigma^2$ .

# Question 2

The formula for the Gaussian kernel is given by

$$ext{similarity}(x,l^{(1)}) = \exp{(-rac{||x-l^{(1)}||^2}{2\sigma^2})}$$
 .

The figure below shows a plot of  $f_1$ =similarity(x,l(1)) when  $\sigma^2 = 1$ .



# Options

Which of the following is a plot of f1 when  $\sigma^2 = 0.25$ ?

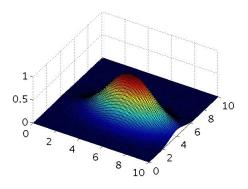


Figure 1: Option 1

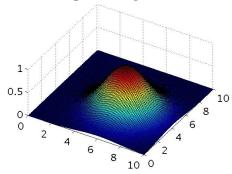


Figure 2: Option 2

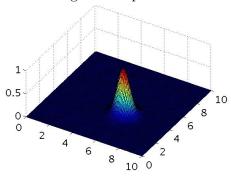
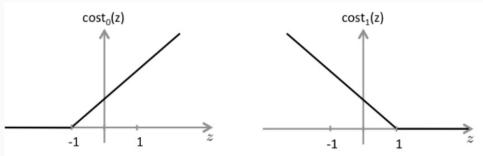


Figure 3: Option 3

#### Question 3

The SVM solves  $\min_{\theta} C \sum_{i=1}^{m} y^{(i)} \text{cost}_1(\theta^T x^{(i)}) + (1 - y^{(i)}) \text{cost}_0(\theta^T x^{(i)}) + \sum_{j=1}^{n} \theta_j^2$  where the functions  $\text{cost}_0(z)$  and  $\text{cost}_1(z)$  look like this:



The first term in the objective is:  $C\sum_{i=1}^m y^{(i)} \mathrm{cost}_1(\theta^T x^{(i)}) + (1-y^{(i)}) \mathrm{cost}_0(\theta^T x^{(i)})$ . 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?

- For every example with y(i)=1, we have that  $\theta^T x(i) \geq 0$ .
- For every example with y(i)=0, we have that  $\theta^T x(i) \leq 0$ .
- CORRECT For every example with y(i)=1, we have that  $\theta^T x(i) \geq 1$ .
- CORRECT For every example with y(i)=0, we have that  $\theta^T x(i) \leq -1$ .

#### Question 4

Suppose you have a dataset with n = 10 features and m = 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.

- CORRECT Try using a neural network with a large number of hidden units.
- WRONG Use a different optimization method since using gradient descent to train logistic regression might result in a local minimum.
- Reduce the number of examples in the training set.
- CORRECT Create / add new polynomial features.

### Question 5

Which of the following statements are true? Check all that apply.

- If the data are linearly separable, an SVM using a linear kernel will return the same parameters  $\theta$  regardless of the chosen value of C (i.e., the resulting value of  $\theta$  does not depend on C).
- CORRECT It is important to perform feature normalization before using the Gaussian kernel.
- $\bullet$  Suppose you are using SVMs to do multi-class classification and would like to use the one-vs-all approach. If you have K different classes, you will train K 1 different SVMs.
- CORRECT The maximum value of the Gaussian kernel (i.e., sim(x,l(1))) is 1.