

Large Margin Classification

Kernels

SVMs in Practice

Review

Reading: Lecture Slides

10 min

Quiz: Support Vector Machines

5 questions

Programming Assignment: Support Vector Machines

3h

QUIZ • 10 MIN

Support Vector Machines

Submit your assignment

DUE Nov 4, 1:29 PM IST

ATTEMPTS 3 every 8 hours

Receive grade

TO PASS 80% or higher

✓ Congratulations! You passed!

TO PASS 80% or higher

Keep Learning

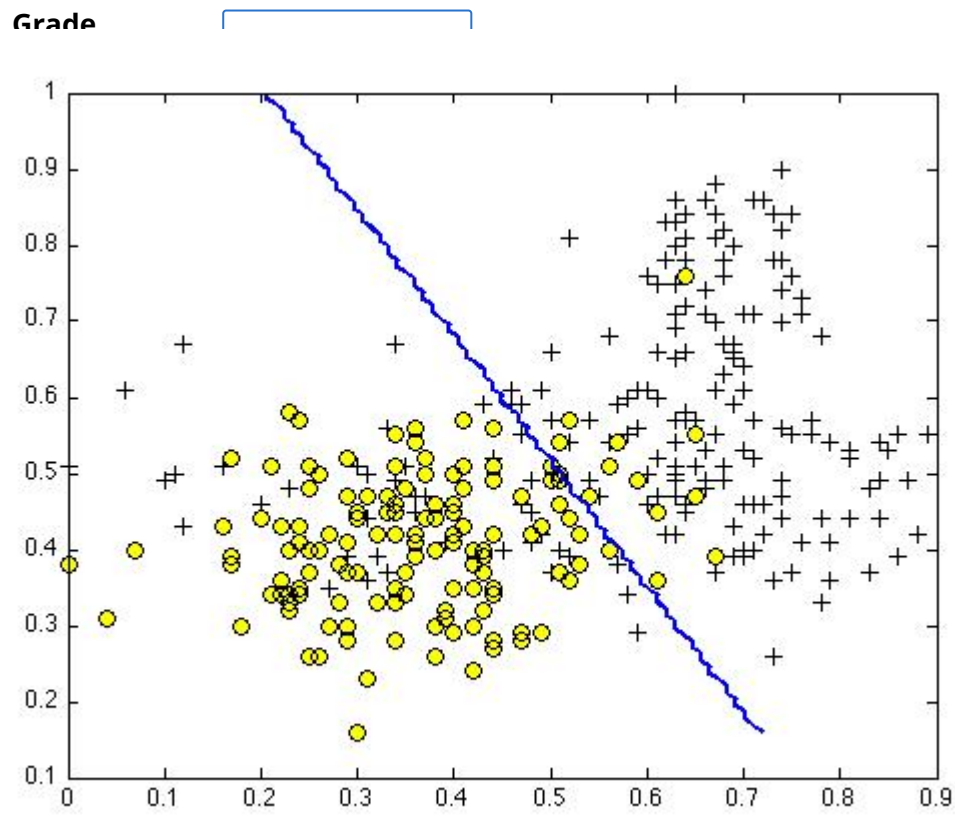
GRADE 100%

Support Vector Machines

LATEST SUBMISSION GRADE 100%

Try again

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

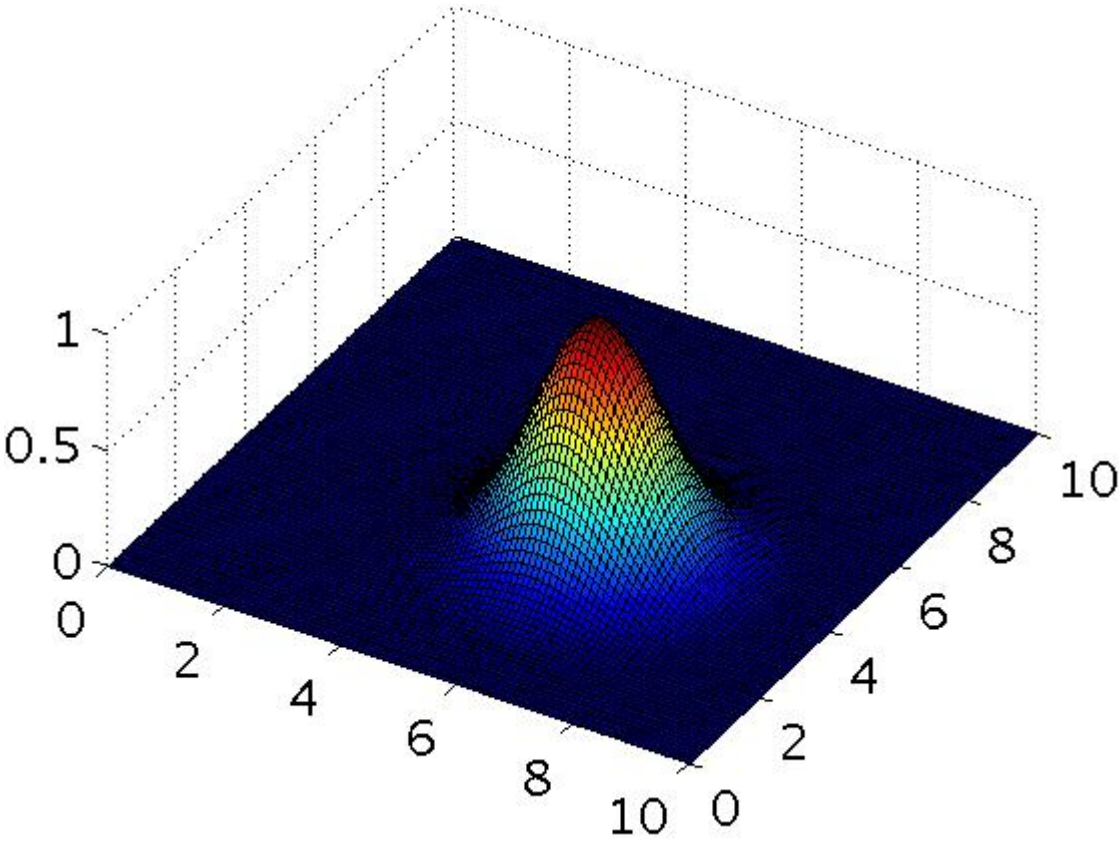


- You suspect that the SVM is underfitting your dataset. Should you try increasing or decreasing C ? Increasing or decreasing σ^2 ?
- ☐ It would be reasonable to try **increasing** C . It would also be reasonable to try **increasing** σ^2 .
 - ☒ It would be reasonable to try **increasing** C . It would also be reasonable to try **decreasing** σ^2 .
 - ☐ It would be reasonable to try **decreasing** C . It would also be reasonable to try **increasing** σ^2 .
 - ☐ It would be reasonable to try **decreasing** C . It would also be reasonable to try **decreasing** σ^2 .

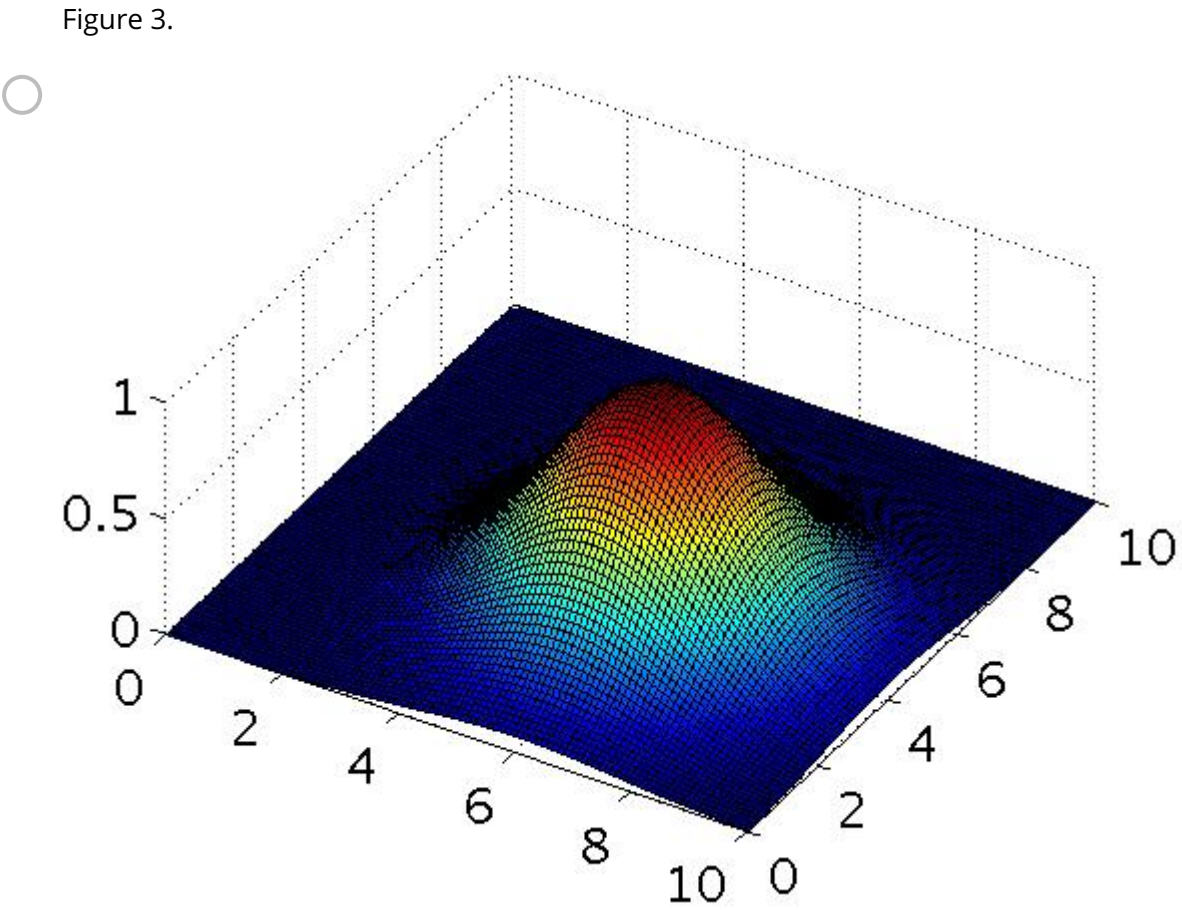
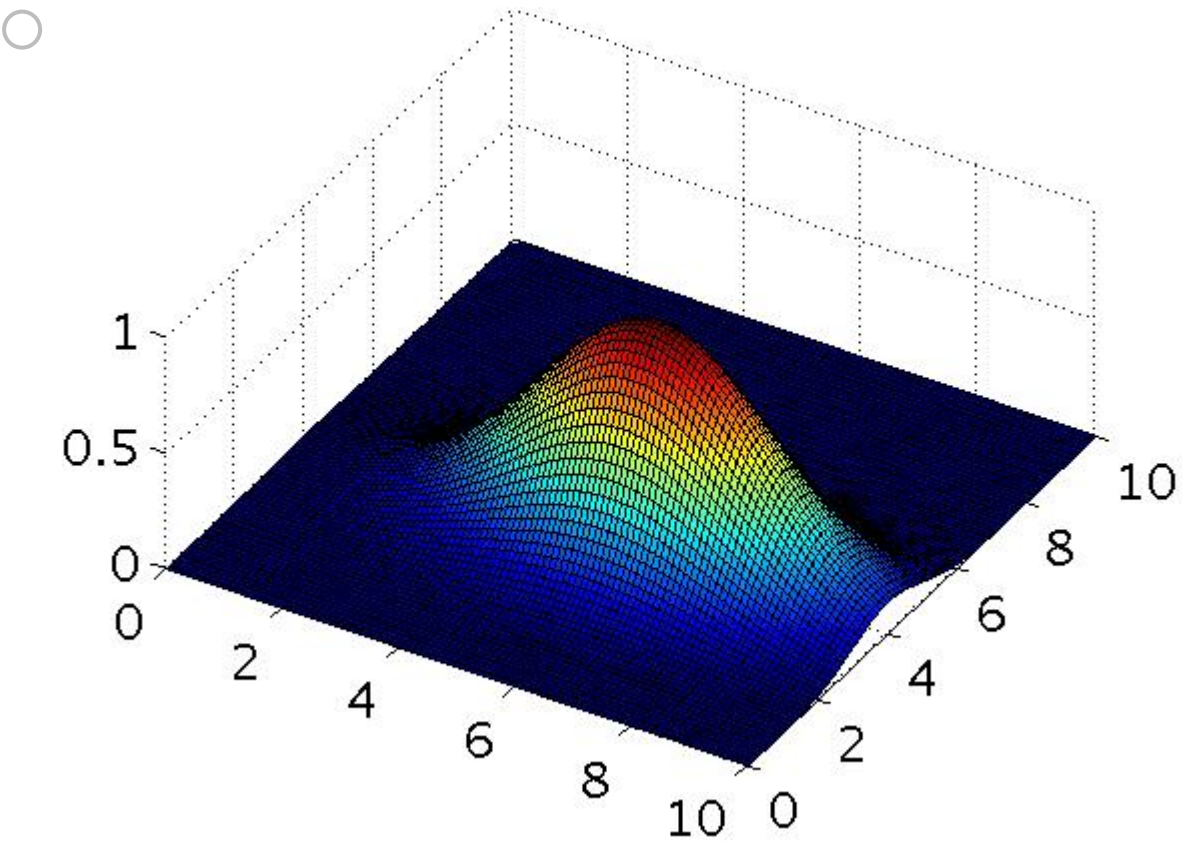
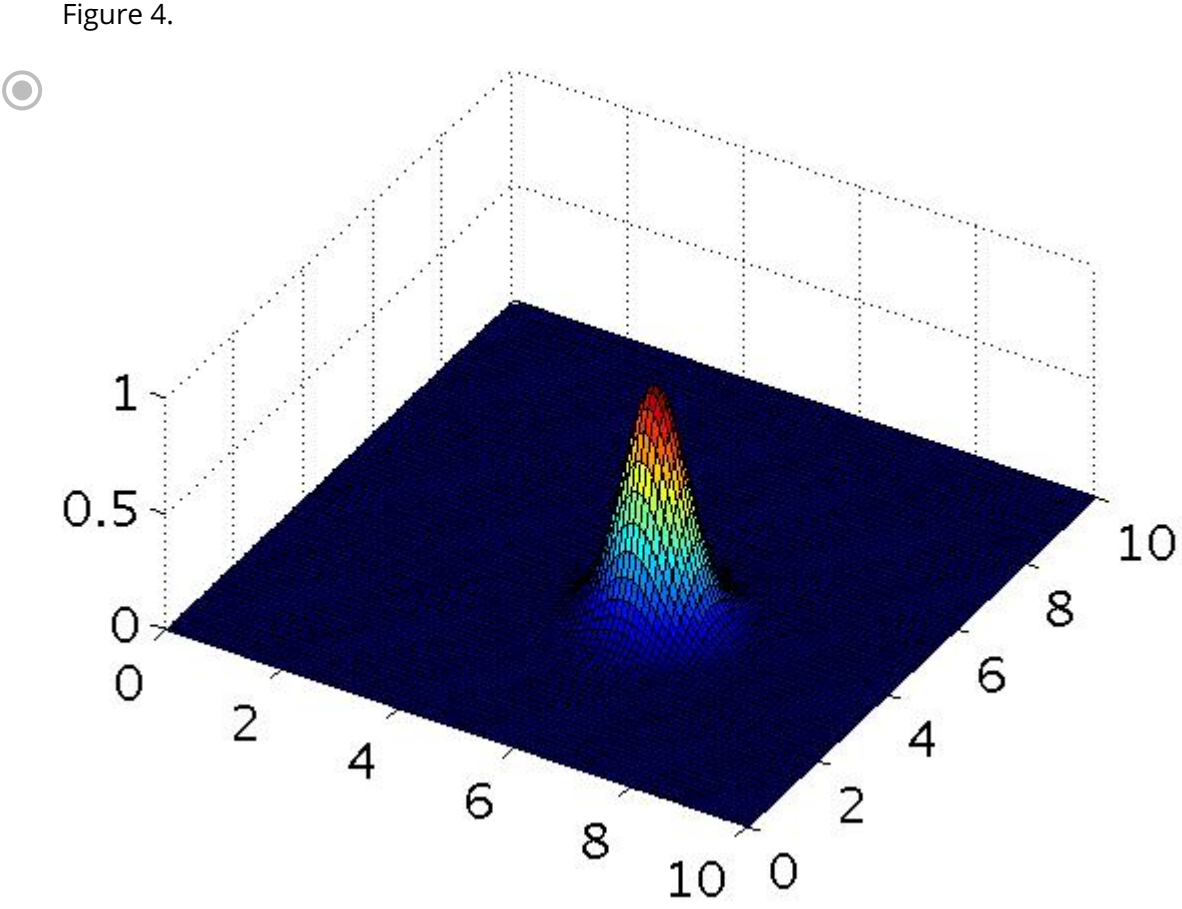
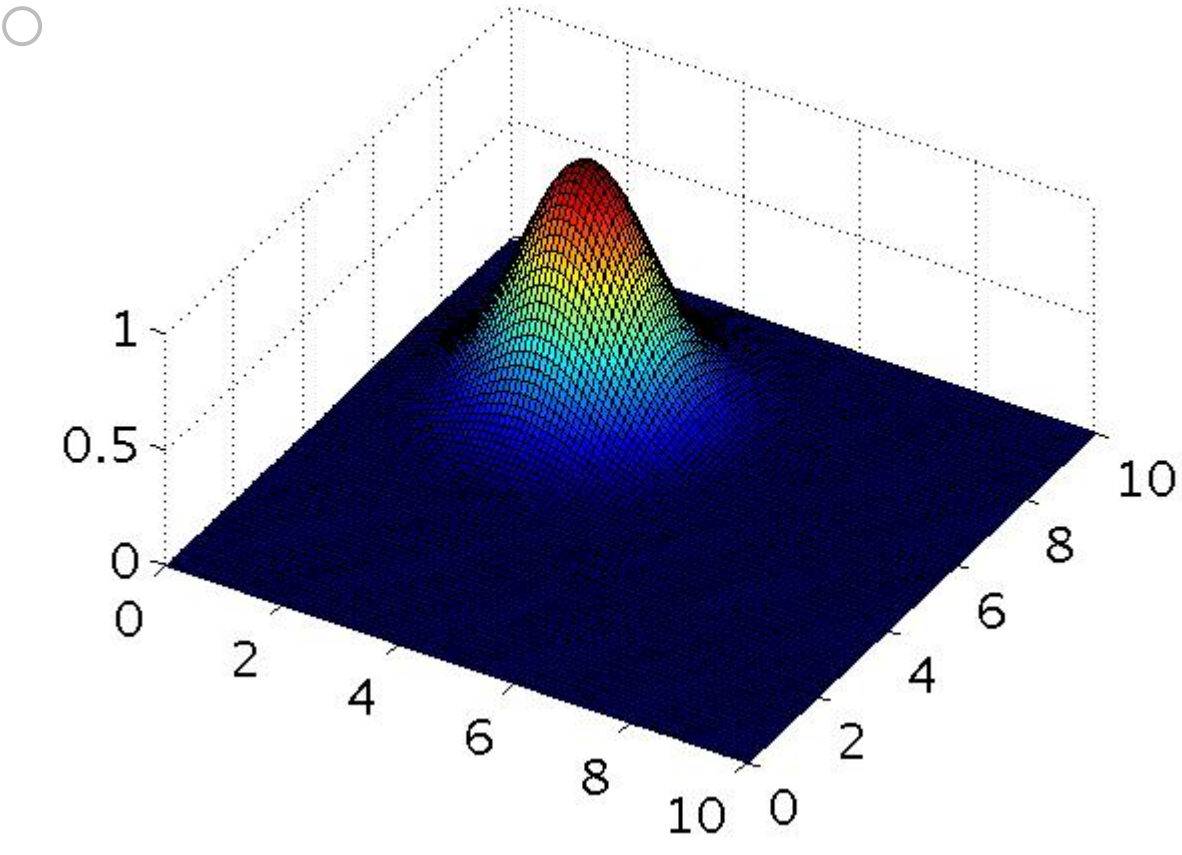
✓ Correct
The figure shows a decision boundary that is underfit to the training set, so we'd like to lower the bias / increase the variance of the SVM. We can do so by either increasing the parameter C or decreasing σ^2 .

2. The formula for the Gaussian kernel is given by $\text{similarity}(x, l^{(1)}) = \exp\left(-\frac{\|x - l^{(1)}\|^2}{2\sigma^2}\right)$. 1 / 1 point

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



Which of the following is a plot of f_1 when $\sigma^2 = 0.25$?

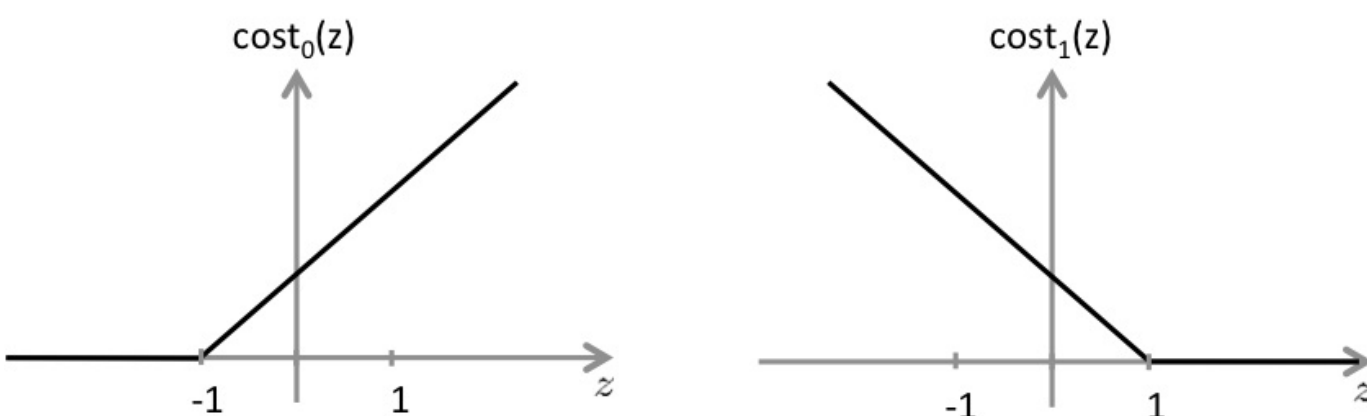


✓ Correct
This figure shows a "narrower" Gaussian kernel centered at the same location which is the effect of decreasing σ^2 .

3. The SVM solves 1 / 1 point

$$\min_{\theta} C \sum_{i=1}^m y^{(i)} \text{cost}_1(\theta^T x^{(i)}) + (1 - y^{(i)}) \text{cost}_0(\theta^T x^{(i)}) + \sum_{j=1}^n \theta_j^2$$

where the functions $\text{cost}_0(z)$ and $\text{cost}_1(z)$ look like this:



The first term in the objective is:

$$C \sum_{i=1}^m y^{(i)} \text{cost}_1(\theta^T x^{(i)}) + (1 - y^{(i)}) \text{cost}_0(\theta^T x^{(i)}).$$

This first term will be zero if two of the following four conditions hold true. Which are the two