CIS 419 – Homework 2

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Problem 1

Using a constant α_k convergence will take longer because if α_k is too small it will move slowly towards convergence and if α_k is too large it will overstep the convergence point.

However, using a α_k as a function of k the step size will change as it approaches the converges point reaching the point faster and more accurately.

Problem 2

a) Write down a vector that is parallel to the optimal vector w.

$$\phi_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \phi_2 = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \phi_2 - \phi_1 = \begin{bmatrix} 0 \\ 2 \\ 2 \end{bmatrix}$$

b) What is the value of margin achieved by w?

$$\|\mathbf{w}\|_2 = \sqrt{8} = 2\sqrt{2}$$

c) Solve for w.

$$\frac{2}{\|\mathbf{w}\|_2} = \frac{2}{\sqrt{8}} = \frac{1}{\sqrt{2}} = \mathbf{w}$$

d) Solve for w_0 .

$$\begin{bmatrix} 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + w_0 \ge 1$$
$$\begin{bmatrix} 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} + w_0 \ge 1$$
$$w_0 \ge 1$$

 $w_0 = 1$

e) Write down the discriminant form $h(x) = w_0 + w^T \phi(x)$.

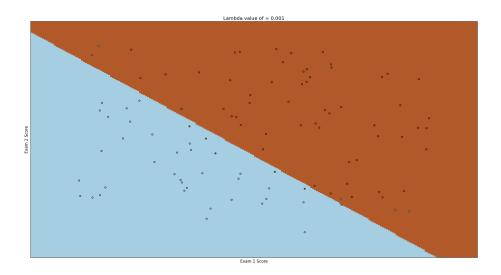
$$h(x) = 1 + \begin{bmatrix} 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ \sqrt{2}x \\ x^2 \end{bmatrix}$$

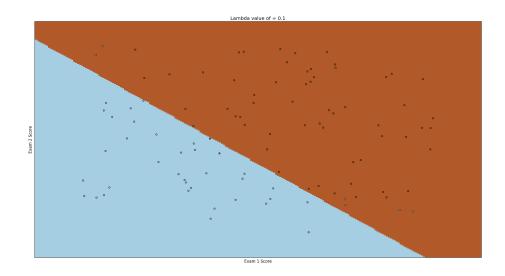
$$h(x) = 1 + 2\sqrt{2}x + 2x^2$$

Draw the complete (unpruned) decision tree, showing the class predictions at the leaves.

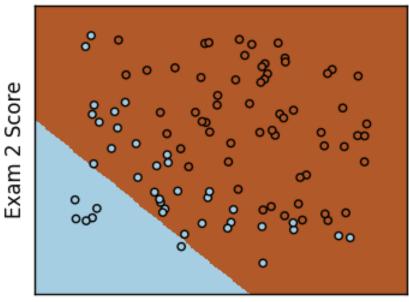
Problem 3

As I increased the value of lambda, I observed that the predicted line was shifted in the south west direction. Capturing more of the blue dots in the red region as the value increased. At a value of 1 about 3/4 of the figure was shaded red. The increase in lambda causes the gradient to increase more rapidly causing it to converge faster. See pictures below.





Lambda value of = 1



Exam 1 Score

Problem 4

As C increased, I observed that the algorithm overclassifies capturing as many points as possible even though they are extreme outliers. As sigma increases, the boundaries of the classification become more rigid with straight lined edges thus reducing the variances of the guassian. As d increases in the polynomial kernel, the polynomial curve of the boundaries increase accordingly. See pictures below.

