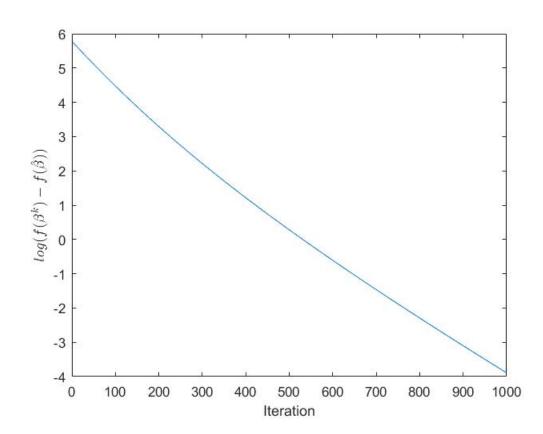
Homework 2

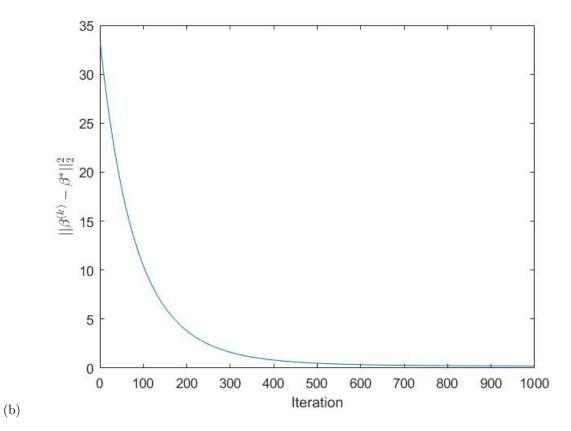
1. (a) The ordinary least squares estimator is implemented in the code as

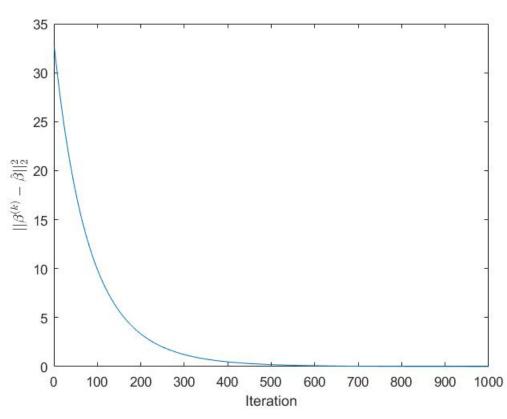
$$\widehat{\beta} = (X'X)^{-1}X'y.$$

(b) The squared error is 0.1991.



2. (a)

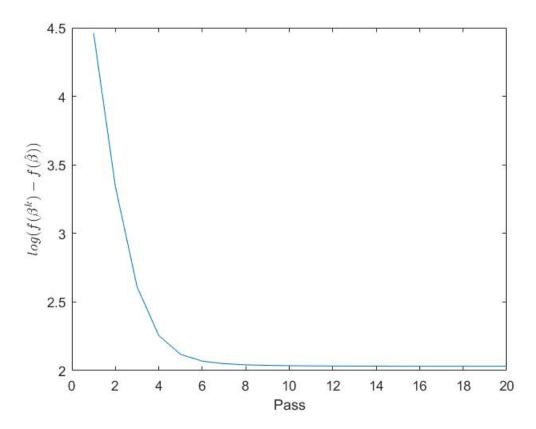


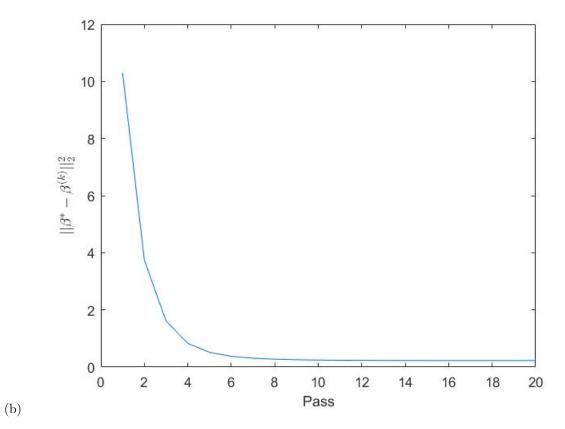


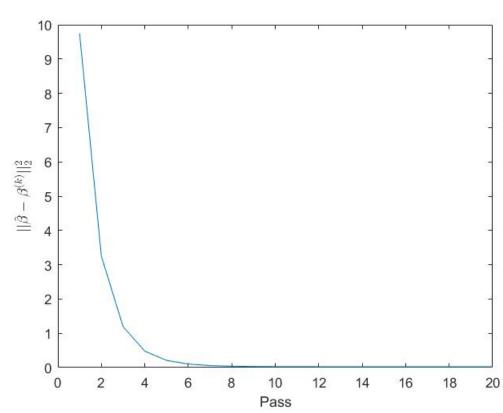
(c)

(d) We notice that β^k converges to $\widehat{\beta}$ faster than β^* .

3. (a)

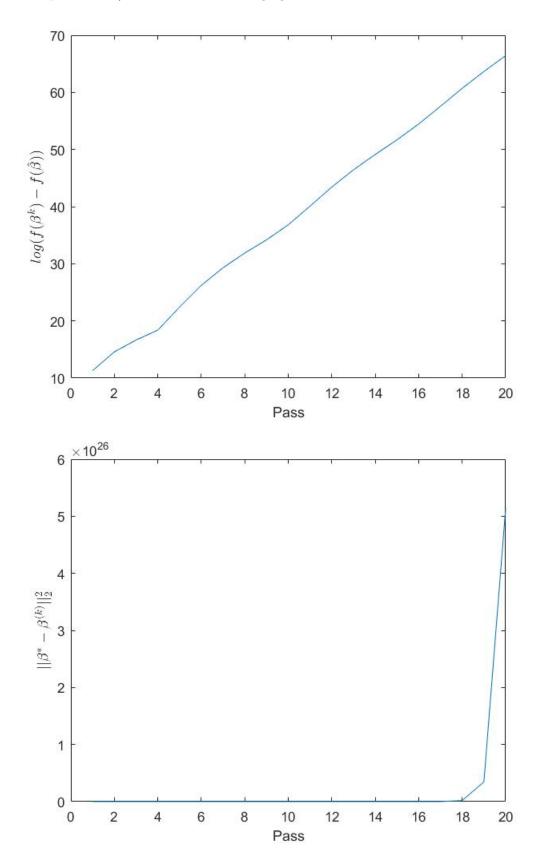


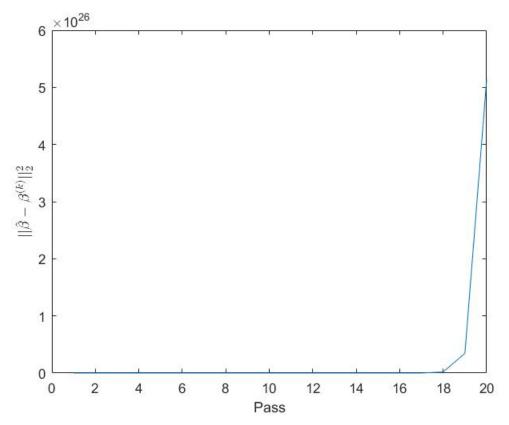




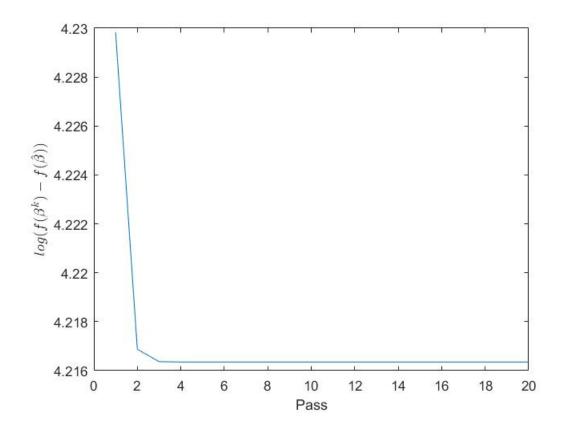
(c)

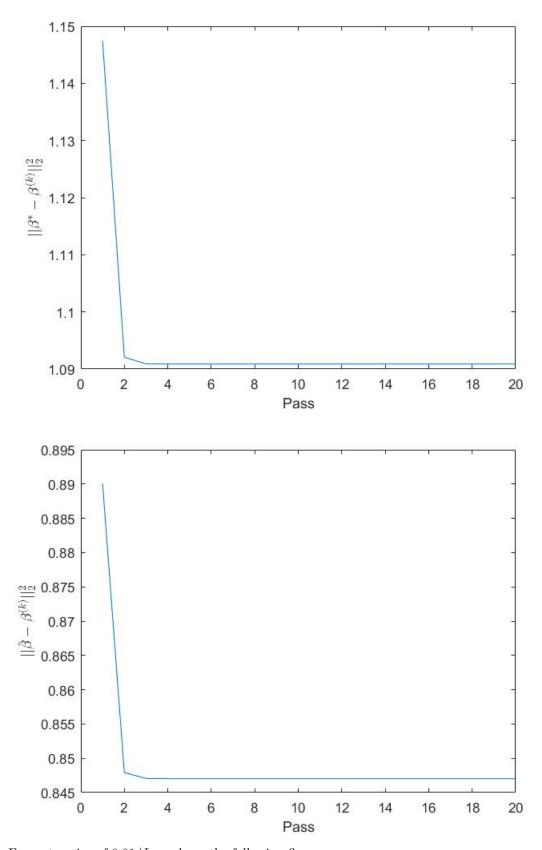
(d) For a step size of 1.7/L, we have the following figures:



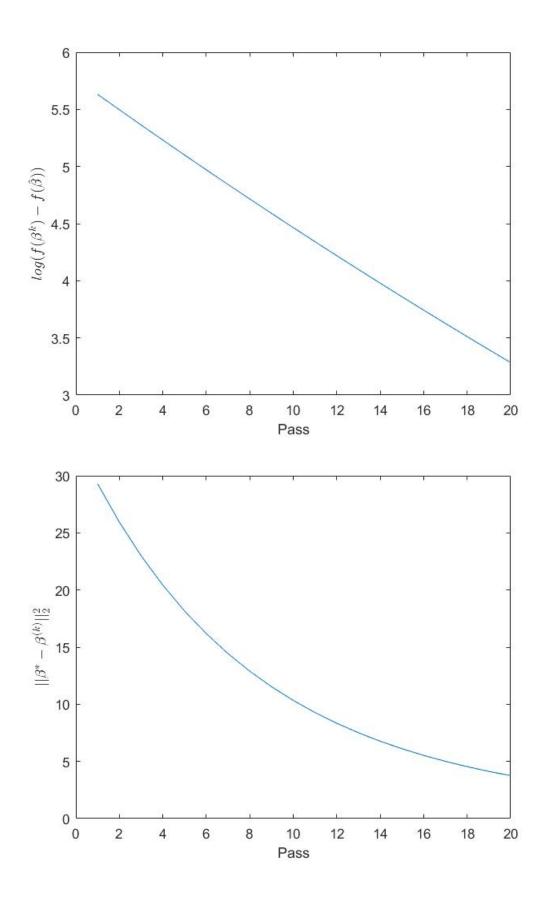


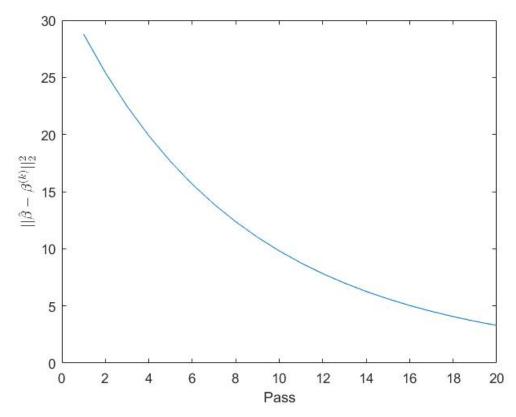
For a step size of 1/L, we have the following figures:





For a step size of 0.01/L, we have the following figures:





- (e) We notice that β^k converges to $\widehat{\beta}$ and β^* only for a step size of 0.1/L and appears to diverge for a step size of 1.7/L. In addition, we notice that β^k converges to some offset values of $\widehat{\beta}$ and β^* for a step size of 1/L. Finally, it appears that β^k converges for a step size of 0.01/L, but its convergence is very slow. Ideally, we should run more iterations and passes to confirm that β^k for this step size. These results show the importance of choosing a good step size when using the Stochastic Gradient Descent Algorithm.
- 4. (a) For GDA, the training error is 0.1233, and the testing error is 0.14.
 - (b) For NB-GDA, the training error is 0.12219, and the testing error is 0.135.
 - (c) For NB-BDA, the training error is 0.12191, and the testing error is 0.127.
 - (d) For QDA, the training error is 0.19217, and the testing error is 0.204.