

Programming Exercise 2: Logistic Regression

Throughout the exercise, you will be using the scripts `ex2.m` and `ex2_reg.m`.

Suppose that you are the administrator of a university department and you want to determine each applicant's chance of admission based on their results on two exams. You have historical data from previous applicants that you can use as a training set for logistic regression. For each training example, you have the applicant's scores on two exams and the admissions decision.

Your task is to build a classification model that estimates an applicant's probability of admission based the scores from those two exams. This outline and the framework code in `ex2.m` will guide you through the exercise.

1.1 Visualizing the data

Before starting to implement any learning algorithm, it is a good idea to visualize the data if possible. In the next part of the exercise, the code will load the data into a MATLAB variable named `X`.

you to look at the code in `plotDecisionBoundary.m` to see how to plot such a boundary using the `beta` values.

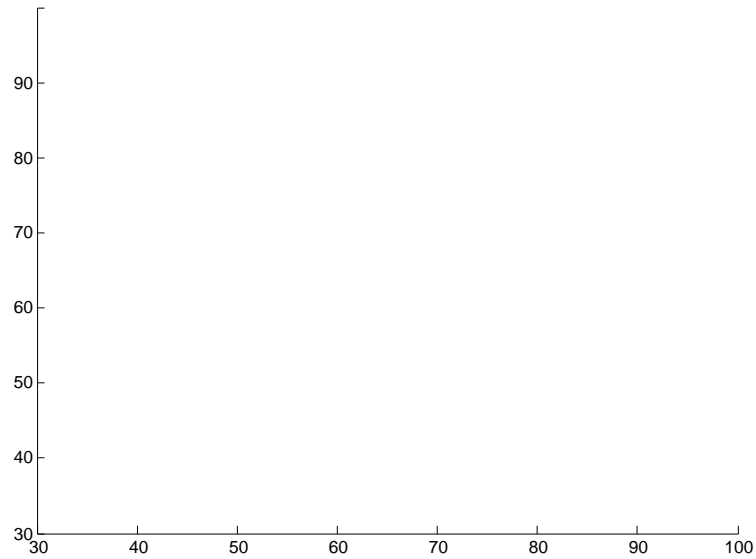


Figure 2: Training data with decision boundary

1.2.4 Evaluating logistic regression

After learning the parameters, you can use the model to predict whether a

2 Regularized logistic regression

negative examples by a straight-line through the plot. Therefore, a straight-

Note that you should not regularize the parameter θ_0 ; thus, the final summation above is for $j = 1$ to n , not $j = 0$ to n . The gradient of the cost function is a vector where the j^{th} element is defined as follows:

$$\frac{\partial J(\theta)}{\partial \theta_j} = \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i) x_i^j$$

2.5 Optional (ungraded) exercises

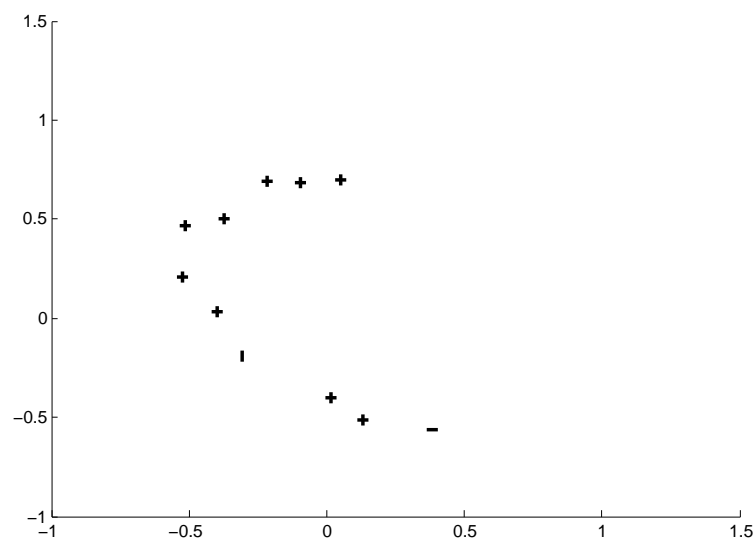


Figure 5: No regularization (Over fitting) ($\lambda = 0$)

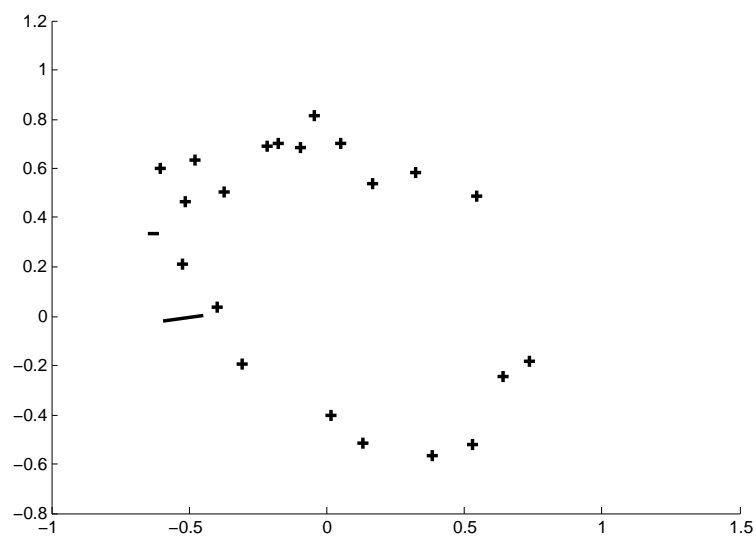


Figure 6: Too much regularization (Under fitting) ($\lambda = 100$)

Submission and Grading