

Week 3

Quiz 1-Logistic Regression

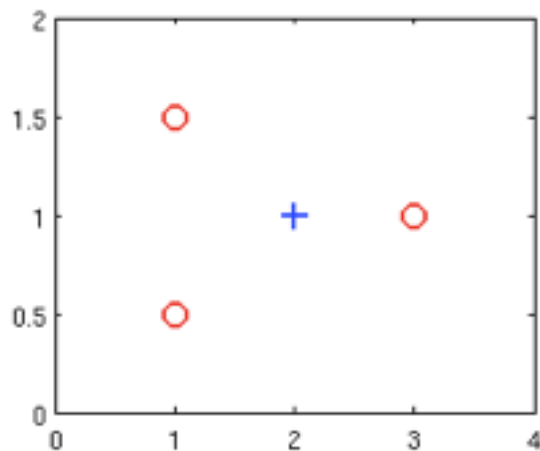
Ques1. Suppose that you have trained a logistic regression classifier, and it outputs on a new example x a prediction $h_{\theta}(x) = 0.7$. This means (check all that apply):

Answer:

- i) Our estimate for $P(y=1|x;\theta)$ is 0.7.
- ii) Our estimate for $P(y=0|x;\theta)$ is 0.3.

Ques2. Suppose you have the following training set, and fit a logistic regression classifier $h_{\theta}(x) = g(\theta_0 + \theta_1 x_1 + \theta_2 x_2)$.

x_1	x_2	y
1	0.5	0
1	1.5	0
2	1	1
3	1	0



Answer:

- i) Adding polynomial features could increase how well we can fit the training data.
- ii) At the optimal value of θ (e.g., found by `fminunc`), we will have $J(\theta) \geq 0$. (As a linear decision boundary could not perfectly fit the dataset)

Ques3.

For logistic regression, the gradient is given by $\frac{\partial}{\partial \theta_j} J(\theta) = \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) x_j^{(i)}$.

Which of these is a correct gradient descent update for logistic regression with a learning rate of α ? Check all that apply.

☒ $\theta := \theta - \alpha \frac{1}{m} \sum_{i=1}^m \left(\frac{1}{1+e^{-\theta^T x^{(i)}}} - y^{(i)} \right) x^{(i)}.$

☐ $\theta_j := \theta_j - \alpha \frac{1}{m} \sum_{i=1}^m (\theta^T x - y^{(i)}) x_j^{(i)}$ (simultaneously update for all j).

☒ $\theta := \theta - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) x^{(i)}.$

☐ $\theta := \theta - \alpha \frac{1}{m} \sum_{i=1}^m (\theta^T x - y^{(i)}) x^{(i)}.$

Ques4.

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

A) The cost function $J(\theta)$ for logistic regression trained with $m \geq 1$ examples is always greater than or equal to zero.

B) For logistic regression, sometimes gradient descent will converge to a local minimum (and fail to find the global minimum). This is the reason we prefer more advanced optimization algorithms such as fminunc (conjugate gradient/BFGS/L-BFGS/etc).

C) Since we train one classifier when there are two classes, we train two classifiers when there are three classes (and we do one-vs-all classification).

D) The one-vs-all technique allows you to use logistic regression for problems in which each $y^{(i)}$ comes from a fixed, discrete set of values.

Answer:

A) Correct.

B) Wrong. The reason of choosing advanced algorithms is because they don't need to choose alpha

C) Wrong. We need to train k classifier when there are k classes ($k > 1$).

D) Correct.

Ques5. Suppose you train a logistic classifier $h_{\theta}(x)=g(\theta_0+\theta_1x_1+\theta_2x_2)$. Suppose $\theta_0=6,\theta_1=-1,\theta_2=0$. Which of the following figures represents the decision boundary found by your classifier?

Answer:



Figure:

