← Logistic Regression

Quiz, 5 questions

1 point

1

Suppose that you have trained a logistic regression classifier, and it outputs on a new example x a prediction $h_{\theta}(x)$ = 0.4. This means (check all that apply):

Our estimate for $P(y=0|x;\theta)$ is 0.4.

Our estimate for $P(y=1|x;\theta)$ is 0.6.

Our estimate for $P(y=0|x;\theta)$ is 0.6.

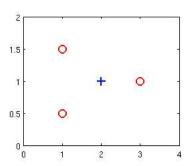
Our estimate for $P(y=1|x;\theta)$ is 0.4.

1 point

2

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	у		
1	0.5	0		
1	1.5	0		
2	1	1		
3	1	0		



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

 $J(\theta)$ will be a convex function, so gradient descent should converge to the global minimum.

Adding polynomial features (e.g., instead using $h_{\theta}(x)=g(\theta_0+\theta_1x_1+\theta_2x_2+\theta_3x_1^2+\theta_4x_1x_2+\theta_5x_2^2)$) could increase how well we can fit the training data.

The positive and negative examples cannot be separated using a straight line. So, gradient descent will fail to converge.

Because the positive and negative examples cannot be separated using a straight line, linear regression will perform as well as logistic regression on this data.

1 point

For logistic regression, the gradient is given by $\frac{\partial}{\partial \theta_j} J(\theta) = \frac{1}{m} \sum_{i=1}^m \frac{h_{\theta}(x^{(i)}) - y^{(i)} x_j^{(i)}}{1}$. Which of these is a correct gradient Logistic Regression with a learning rate of α ? Check all that apply.

$\theta := \theta := \epsilon$	$1 \sum^{m}$	(ρT_{α})	a(i)	$\alpha^{(i)}$	(simultaneously update for all j).
$v_j := v_j - v_j$	$m \angle i=1$	(0 a	9	r_i	(Similaricously aparate for all J).

$$egin{aligned} heta &:= heta - lpha rac{1}{m} \sum_{i=1}^m \left(rac{1}{1+e^{- heta^T x^{(i)}}} - y^{(i)}
ight) x^{(i)}. \ heta &:= heta - lpha rac{1}{m} \sum_{i=1}^m \left(h_ heta(x^{(i)}) - y^{(i)}
ight) x^{(i)}. \ heta &:= heta - lpha rac{1}{m} \sum_{i=1}^m \left(heta^T x - y^{(i)}
ight) x^{(i)}. \end{aligned}$$

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

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

point

4.

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

The sigmoid function $g(z) = \frac{1}{1+e^{-z}}$ is never greater than one (> 1).

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

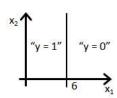
Linear regression always works well for classification if you classify by using a threshold on the prediction made by linear regression.

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).

1 point

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

Figure:



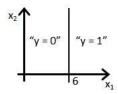


Figure:

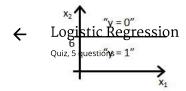
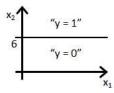


Figure:



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