Proofs you probably weren't taught

Friday, October 21, 2011

Why is the error function minimized in logistic regression convex?

We want to prove that the error/objective function of logistic regression:

$$J(\theta) = \sum_{i=1}^{m} y^{i} \left[-\log \left(h_{\theta} \left(x^{i} \right) \right) \right] + \left(1 - y^{i} \right) \left[-\log \left(1 - h_{\theta} \left(x^{i} \right) \right) \right] \quad (1)$$

where
$$h_{\theta}(x) = \frac{1}{1 + e^{-\theta^T x}}$$

is convex.

Proof:

Before beginning the proof, i would first like to make you review/recollect a few definitions/rules/facts/results related to convex functions:

• **Definition of a convex function:** A function f(x) is said to be convex if the following inequality

$$f\left(\alpha x+\left(1-\alpha\right)y\right)\leq\alpha f\left(x\right)+\left(1-\alpha\right)f\left(y\right)\ \forall x,y\in Domain(f)\ and\ \alpha\in\left[0,1\right]$$

• First-order condition of convexity: A function f(x) which is differentiable is convex if the following inequality condition holds true:

$$f\left(y\right) \; \geq \; f\left(x\right) + \nabla_{x}^{T} f\left(x\right) \left(y-x\right); \quad \forall x,y \in Domain(f)$$

Intuitively, this condition says that the tangent/first-order-taylor-series approximation of f(x) is globally an under-estimator.

Second-order condition of convexity: A function f(x) which is twice-differentiable is convex if and only if its hessian matrix (matrix of second-order partial derivatives) is positive semi-definite,

$$\forall z: z^{T} \nabla_{x}^{2} f(x) z \geq 0 \text{ where } \nabla_{x}^{2} f(x) \text{ is the hessian}$$

• Sum/Linear-combination of two or more convex functions is also convex: Let f(x) and g(x) be two convex functions. Then any linear combination of these two functions

$$(\lambda_1 f + \lambda_2 g)(x) = \lambda_1 f(x) + \lambda_2 g(x)$$

is also a convex function (this can be easily proved using the definition of the convex function).

Now notice that if we can prove that the two functions

$$-\log\left(h_{\theta}\left(x^{i}\right)\right) \ and \ -\log\left(1-h_{\theta}\left(x^{i}\right)\right)$$

are convex, then our objective function

$$J(\theta) = \sum_{i=1}^{m} y^{i} \left[-\log \left(h_{\theta} \left(x^{i} \right) \right) \right] + \left(1 - y^{i} \right) \left[-\log \left(1 - h_{\theta} \left(x^{i} \right) \right) \right]$$

must also be convex since any linear combination of two or more convex functions is also convex.

Let us now try to prove that

$$-\log\left(h_{\theta}\left(x\right)\right) = -\log\left(\frac{1}{1 + e^{-\theta^{T}x}}\right) = \log\left(1 + e^{-\theta^{T}x}\right)$$

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Deepak Roy Chittajallı

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is a convex function of theta. In order to do this, we will use the <u>second-order condition of convexity</u> described above. Let us first compute the hessian matrix:

$$\begin{aligned} & \operatorname{grad}: \\ & \nabla_{\theta} \left[-\log \left(h_{\theta} \left(x \right) \right) \right] & = & \nabla_{\theta} \left[\log \left(1 + e^{-\theta^{T} x} \right) \right] \\ & = & \left(\frac{-e^{-\theta^{T} x}}{1 + e^{-\theta^{T} x}} \right) x \\ & = & \left(\frac{1}{1 + e^{-\theta^{T} x}} - 1 \right) x \\ & = & \left(h_{\theta} \left(x \right) - 1 \right) x \end{aligned} \end{aligned} \qquad \begin{aligned} & \operatorname{hessian}: \\ & \nabla_{\theta}^{2} \left[-\log \left(h_{\theta} \left(x \right) \right) \right] & = & \nabla_{\theta} \left(\nabla_{\theta} \left[-\log \left(h_{\theta} \left(x \right) \right) \right] \right) \\ & = & \nabla_{\theta} \left(\left(h_{\theta} \left(x \right) - 1 \right) x \right) \\ & = & h_{\theta} \left(x \right) \left(1 - h_{\theta} \left(x \right) \right) x x^{T} \end{aligned}$$

Now below is the proof that this hessian matrix is positive semi-definite:

$$\forall z: \ z^{T} \nabla_{x}^{2} \left(-\log \left(h_{\theta}(x)\right)\right) = \ z^{T} \left[h_{\theta}(x) \left(1 - h_{\theta}(x)\right) x x^{T}\right] z \\ = \ h_{\theta}(x) \left(1 - h_{\theta}(x)\right) \left(x^{T} z\right)^{2} \ge 0 \quad (2)$$

Let us now try to prove that

$$-\log(1 - h_{\theta}(x)) = -\log\left(1 - \frac{1}{1 + e^{-\theta^T x}}\right)$$
$$= -\log\left(\frac{e^{-\theta^T x}}{1 + e^{-\theta^T x}}\right)$$
$$= \theta^T x + \log\left(1 + e^{-\theta^T x}\right)$$

is a convex function of theta. In order to do this, we will again use the <u>second-order condition of convexity</u> described above. Let us first compute its hessian matrix:

$$\begin{array}{ll} \operatorname{grad}: & \\ \nabla_{\theta} \left[-\log \left(1 - h_{\theta} \left(x \right) \right) \right] & = & \nabla_{\theta} \left[\theta^{T} x + \log \left(1 + e^{-\theta^{T} x} \right) \right] \\ & = & x + \nabla_{\theta} \left[\log \left(1 + e^{-\theta^{T} x} \right) \right] \end{array}$$

hessian:

$$\nabla_{\theta}^{2} \left[-\log\left(1 - h_{\theta}\left(x\right)\right) \right] = \nabla_{\theta} \left(\nabla_{\theta} \left[-\log\left(1 - h_{\theta}\left(x\right)\right) \right] \right)$$

$$= \nabla_{\theta} \left(x + \nabla_{\theta} \left[\log\left(1 + e^{-\theta^{T}x}\right) \right] \right)$$

$$= \nabla_{\theta}^{2} \left[-\log\left(h_{\theta}\left(x\right)\right) \right]$$
(we have proved in Eq. (2) above that this is positive semi – definite)

Above, we have proved that both

$$-\log(h_{\theta}(x^{i}))$$
 and $-\log(1-h_{\theta}(x^{i}))$

are convex functions. And, the error/objective function of logistic regression

$$J(\theta) = \sum_{i=1}^{m} y^{i} \left[-\log \left(h_{\theta} \left(x^{i} \right) \right) \right] + \left(1 - y^{i} \right) \left[-\log \left(1 - h_{\theta} \left(x^{i} \right) \right) \right]$$

is essentially a linear-combination of several such convex functions. Now, since a linear combination of two or more convex functions is convex, we conclude that the objective function of logistic regression is convex.

Hence proved ...

Following the same line of approach/argument it can be easily proven that the objective function of logistic regression is convex even if regularization is used.

Posted by Deepak Roy Chittajallu at $\underline{9:38\ PM}$

Labels: Logistic Regression, Machine Learning

12 comments:



Unknown November 2, 2015 at 3:42 PM

Just to correct a little mistake: only a positive linear combination of convex functions is guaranteed to be convex again. However, in the logistic regression case y_i are positive, so it works indeed.

Reply



for. October 17, 2020 at 8:54 AM

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Unknown February 13, 2016 at 10:44 AM

This comment has been removed by the author.

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Unknown February 13, 2016 at 11:09 AM

I think for equation(2) you lost a z in the left part.

Reply



Unknown August 14, 2017 at 11:42 PM

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Reply



Unknown March 9, 2018 at 9:26 AM

Awesome post, thank you.

Agree with above that (2) missing a z on L.H.S. $\,$

Reply



Effesian March 12, 2018 at 5:22 AM

I think in Eq.(2) you should write $\operatorname{grad}_{\hat{x}}$ and not w.r.t. x. This is of course just a typo. Nice post btw.!

Reply



Unknown September 24, 2018 at 4:30 AM

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