

CS 229 Homework

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345.2016

These are solutions to the most recent problems posted for Stanford's CS 229 course, as of June 2016. I'm not sure if this course re-uses old problems, but please don't copy the answers if so. This document is also available as a [pdf](#).

1 Problem set 1

1.1 Logistic regression

1.1.1 Part (a)

The problem is to compute the Hessian matrix H for the function

$$J(\theta) = -\frac{1}{m} \sum_{i=1}^m \log(g(y^{(i)}x^{(i)})),$$

where $g(z)$ is the logistic function, and to show that H is positive semi-definite; specifically, that $z^T H z \geq 0$ for any vector z .

We'll use the fact that $g'(z) = g(z)(1 - g(z))$. We'll also note that since all relevant operations are linear, it will suffice to ignore the summation over i in the definition of J . I'll use the notation ∂_j for $\frac{\partial}{\partial \theta_j}$, and introduce t for $y\theta^T x$. Then

$$-\partial_j(mJ) = \frac{g(t)(1 - g(t))}{g(t)} x_j y = x_j y (1 - g(t)).$$

Next

$$-\partial_k \partial_j(mJ) = x_j y (-g(t)(1 - g(t))) x_k y,$$

so that

$$\partial_{jk}(mJ) = x_j x_k y^2 \alpha,$$

where $\alpha = g(t)(1 - g(t)) > 0$.

Thus we can use repeated-index summation notation to arrive at

$$z^T H z = z_i h_{ij} z_j = (\alpha y^2)(z_i x_i x_j z_j) = (\alpha y^2)(x^T z)^2 \geq 0.$$

This completes this part of the problem.

1.2 Poisson regression and the exponential family

1.2.1 Part (a)

Write the Poisson distribution as an exponential family:

$$p(y; \eta) = b(y) \exp(\eta^T T(y) - a(\eta)),$$

where

$$p(y; \lambda) = \frac{e^{-\lambda} \lambda^y}{y!}.$$

This can be done via

$$\begin{aligned} \eta &= \log(\lambda), \\ a(\eta) &= e^\eta = \lambda, \\ b(y) &= 1/y!, \text{ and} \\ T(y) &= y. \end{aligned}$$