

Possion_GLM

Wednesday, August 12, 2020 2:10 PM

3. [25 points] Poisson Regression

(a) [5 points] Consider the Poisson distribution parameterized by λ :

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

Show that the Poisson distribution is in the exponential family, and clearly state the values for $b(y)$, η , $T(y)$, and $a(\eta)$.

Answer:

Ans. (a)

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

$$p(y; \lambda) = \frac{e^{-\lambda} \lambda^y}{y!} = \frac{1}{y!} e^{-\lambda} e^{y \log \lambda} = \frac{1}{y!} e^{y \log \lambda - \lambda}$$

$$\text{So, } b(y) = 1/y!, \quad T(y) = y$$

$$\eta = \log \lambda, \quad a(\eta) = \lambda = e^\eta$$

(b) [3 points] Consider performing regression using a GLM model with a Poisson response variable. What is the canonical response function for the family? (You may use the fact that a Poisson random variable with parameter λ has mean λ .)

Answer:

(c) [7 points] For a training set $\{(x^{(i)}, y^{(i)}); i = 1, \dots, m\}$, let the log-likelihood of an example be $\log p(y^{(i)} | x^{(i)}; \theta)$. By taking the derivative of the log-likelihood with respect to θ_j , derive the stochastic gradient ascent update rule for learning using a GLM model with Poisson responses y and the canonical response function.

Answer:

(b)

$$h(\theta) = E(\eta | x) = \lambda = e^\eta = e^{\theta^T x} \quad \text{So, } h(\theta) = e^{\theta^T x} \quad \left[\begin{array}{l} g(\eta) = e^\eta \\ \text{canonical response function} \end{array} \right]$$

for learning GLM

$$l(\theta) = \sum_{i=1}^m \log p(y^{(i)} | x^{(i)}; \theta) = \sum_{i=1}^m \log \left(\frac{1}{y!} e^{y \log \lambda - \lambda} \right) = \sum_{i=1}^m \left(\log \left(\frac{1}{y!} \right) + y \log \lambda - \lambda \right)$$

$$= \sum_{i=1}^m \left(\log \left(\frac{1}{y!} \right) + y \log \lambda - e^{\theta^T x} \right)$$

$$\frac{\partial l(\theta)}{\partial \theta_j} = \sum_{i=1}^m \frac{\partial}{\partial \theta_j} \left(\log \left(\frac{1}{y!} \right) + y \log \lambda - e^{\theta^T x} \right) = \sum_{i=1}^m \left(\frac{y^{(i)}}{\lambda^{(i)}} - e^{\theta^T x^{(i)}} x_j^{(i)} \right)$$

So, stochastic gradient descent rule is:

$$\theta := \theta + \alpha (y^{(i)} - e^{\theta^T x^{(i)}}) x^{(i)}$$

- (d) [7 points] **Coding problem.** Consider a website that wants to predict its daily traffic. The website owners have collected a dataset of past traffic to their website, along with some features which they think are useful in predicting the number of visitors per day. The dataset is split into train/valid/test sets and follows the same format as Datasets 1-3:

`data/ds4_{train,valid}.csv`

We will apply Poisson regression to model the number of visitors per day. Note that applying Poisson regression in particular assumes that the data follows a Poisson distribution whose natural parameter is a linear combination of the input features (*i.e.*, $\eta = \theta^T x$). In `src/p03d_poisson.py`, implement Poisson regression for this dataset and use gradient ascent to maximize the log-likelihood of θ .

Answer: