



# CS109A Introduction to Data Science:

## Homework 6 AC 209 : GLMs

Harvard University

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Names of people you have worked with goes here:

In [ ]:

In [1]:

```
# RUN THIS CELL FOR FORMAT
import requests
from IPython.core.display import import HTML
styles = requests.get("https://raw.githubusercontent.com/Harvard-IACS/2018-CS109A/master/
HTML(styles)
```

Out[1]:

In [2]:

```
# Imports
import numpy as np
import matplotlib.pyplot as plt
from sklearn.datasets import make_regression
from scipy.optimize import minimize

%matplotlib inline
```

## Question 5 [25 pts]

The probability density function of Poisson distribution is given by:

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

**5.1** Show that Poisson distribution belongs to the general exponential distribution family with probability density:

$$f(y|\theta) = \exp\left(\frac{y\theta - b(\theta)}{\phi} + c(y, \phi)\right).$$

**5.2** For the Poisson distribution calculate the canonical parameter  $\theta$  in terms of  $\lambda$ , the cumulant function  $b(\theta)$ , the dispersion parameter  $\phi$ , and the normalization function  $c(y, \phi)$ .

**5.3** Show that the canonical link  $g(\cdot)$  for the Poisson regression model is given by:

$$\eta = g(\lambda) = \log(\lambda).$$

**5.4** Derive the *normal equations* that maximizes the likelihood in the Poisson regression model.

## Answers

**5.1** Show that Poisson distribution belongs to the general exponential distribution family with probability density:

$$f(y|\theta) = \exp\left(\frac{y\theta - b(\theta)}{\phi} + c(y, \phi)\right).$$

*your answer here* We can rewrite the poisson pdf as

$$\begin{aligned} p(y|\lambda) &= \frac{\lambda^y}{y!} e^{-\lambda} \\ &= \frac{e^{y \ln(\lambda)}}{e^{\ln(y!)}} e^{-\lambda} \\ &= \exp(y \ln(\lambda) - \lambda - \ln(y!)) \end{aligned}$$

If we let

$$\begin{aligned} \theta &= \ln(\lambda) \\ \phi &= 1 \\ b(\theta) &= e^\theta \\ c(y, \phi) &= -\ln(y!) \end{aligned}$$

Then

$$\exp\left(\frac{y\theta - b(\theta)}{\phi} + c(y, \phi)\right) = \exp(y \ln(\lambda) - \lambda - \ln(y!))$$

as required.

**5.2** For the Poisson distribution calculate the canonical parameter  $\theta$  in terms of  $\lambda$ , the cumulant function  $b(\theta)$ , the dispersion parameter  $\phi$ , and the normalization function  $c(y, \phi)$ .

*your answer here*

In terms of  $\lambda$ :

$$\theta = \ln(\lambda)$$

In terms of  $b(\theta)$ :

$$\theta = \ln(b(\theta))$$

$\theta$  is not a function of  $\phi$  or  $c(y, \phi)$ .

**5.3** Show that the canonical link  $g(\cdot)$  for the Poisson regression model is given by:

$$\eta = g(\lambda) = \log(\lambda).$$

*your answer here*

The canonical link is defined such that  $g(\lambda) = \theta$ . Hence, since  $\theta = \ln(\lambda)$ , we know that  $g(\lambda) = \ln(\lambda)$

**5.4** Derive to the *normal equations* that maximizes the likelihood in the Poisson regression model.

*your answer here*

The normal equations for the exponential family with a canonical link is

$$\sum_{i=1}^n \omega_i (y_i - \mu_i) x_i^T = 0$$

where  $\omega_i$  is the weight of the dispersion parameters, which we assume to be 1, and  $g(\mu_i) = x_i^T \beta$  (only for canonical links).

For the Poisson regression with the canonical link,  $\mu_i = e^{x_i^T \beta}$  so the normal equation is

$$\sum_{i=1}^n (y_i - e^{x_i^T \beta}) x_i^T = 0$$

In [ ]: