

Generalized linear models

11/05/2018

Exponential family: group/class of ~~any~~ distributions which can for which pmf / pdf can be written in the form of

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

$\eta \rightarrow$ natural parameter

$T(y) \rightarrow$ sufficient statistic (usually $T(y) = y$)

Fixing $b(y)$, $T(y)$, $a(\eta)$, then $p(y|\eta)$ defines a set of distributions as we vary η .

Exponential Family = \int Bernoulli, Normal, Poisson, Geometric, ... multinomial

Is Bernoulli distribution is Generalized model?

Logistic regression: $y|x;\theta \sim \text{Bernoulli}(\phi)$ Prob. of success
 $p(y=1|x;\theta) = \phi$, $p(y=0|x;\theta) = 1-\phi$

we define hypothesis; $h_\theta(x) = \phi$

$$p(y|x;\theta) = h_\theta(x)^y (1-h_\theta(x))^{1-y} = \phi^y (1-\phi)^{1-y}$$

fixed

$$p(y|x;\theta) = \exp(\log \phi^y (1-\phi)^{1-y}) = \exp(y \log \phi + (1-y) \log(1-\phi))$$

$$= \exp(y \log \frac{\phi}{1-\phi} + \log(1-\phi))$$

Comparing: $p(y|\eta) = b(y) \exp(\eta^T T(y) - a(\eta))$

$$b(y) = 1, \quad \eta = \log \frac{\phi}{1-\phi}, \quad a(\eta) = -\log(1-\phi) = y$$

in terms of η : $\phi = \frac{1}{1+e^{-\eta}} \quad a(\eta) = \log(1+e^\eta)$

Hence, Bernoulli distribution is in Exponential family

Generalized linear model:

Assumption-1: $y|x, \theta \sim \text{Exp}(\phi) ; \text{Bern}(\theta)$

Assumption-2: $h_\theta(x) = E[\tau(y)|x] = E[y|x] = \eta(y=1|x; \theta)$
 $= \phi = \frac{1}{1+e^{-\eta}}$

Assumption-3: $\eta = \theta^T x$

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

Prove Gaussian/Normal distribution is an Exponential family?

Random variable: $y|x, \theta \sim \text{Gaussian}(\mu, \sigma^2)$

$\sim \mathcal{N}(\mu, \sigma)$ σ does not matter in linear regression

Probability density function:

$$p(y|x, \theta) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y-\mu)^2\right)$$

$$p(y|x, \theta) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y^2 + \mu^2 - 2y\mu)\right)$$

$$= \frac{1}{\sqrt{2\pi}} \exp^{-\frac{1}{2}y^2} \exp\left(y\mu - \frac{\mu^2}{2}\right)$$

comparing: $p(y|\eta) = b(y) \exp(\eta^T \tau(y) - a(\eta))$

$$b(y) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}y^2\right) \quad \tau(y) = y, \quad \eta = \mu$$
$$a(\eta) = \frac{\mu^2}{2} = \frac{\eta^2}{2}$$

Generalized linear model:

Assumption-1: $y|x;\theta \sim \mathcal{N}(\mu, \sigma)$

Assumption-2: $h_{\theta}(x) = E[\tau(y)|x] = E[y|x] = \mu = \eta$

Assumption-3: $\eta = \theta^T x$

$$h_{\theta}(x) = \eta = \theta^T x$$