

GLM

1. Definition of GLM

a. Exponential family with canonical form

GLM: Response variable Y must be a random variable coming from a exponential family with canonical form.

Def. A random variable Y has a distribution in the **Exponential Family** if its pdf (for continous random variables) or its pmf (for discrete random variables) has the following form:

$$\begin{aligned} f(y; \theta) &= s(y) \cdot t(\theta) \cdot \exp\{a(y) \cdot b(\theta)\} \\ &= \exp\{a(y) \cdot \underbrace{b(\theta)}_{\text{natural parameter}} + c(\theta) + d(y)\} \end{aligned}$$

* Sample space of y does not depend on θ

If $a(y) = y$, then the distribution of y is in **Canonical** form.

Common exponential family:

$$\begin{aligned} Y &\sim N(\mu, \sigma^2) \\ Y &\sim \text{Bernoulli}(\pi) \\ Y &\sim \text{Binomial}(n, \pi) \\ Y &\sim \text{Poisson}(\lambda) \\ Y &\sim \text{Exp}(\lambda) \end{aligned}$$

Ex. $y \sim N(\mu, \sigma^2)$

$$\begin{aligned} f(y; \mu) &= \frac{1}{\sqrt{2\pi} \cdot \sigma} \exp\left\{-\frac{(y - \mu)^2}{2\sigma^2}\right\} \\ &= \exp\left\{-\frac{y^2}{2\sigma^2} + \frac{\mu y}{\sigma^2} - \frac{\mu^2}{2\sigma^2} - \lg(\sqrt{2\pi} \cdot \sigma)\right\} \\ &= \exp\left\{\underbrace{y}_{a(y)} \underbrace{\left(\frac{\mu}{\sigma^2}\right)}_{b(\mu)} - \underbrace{\left(\frac{\mu^2}{2\sigma^2} + \lg(\sqrt{2\pi} \cdot \sigma)\right)}_{c(\mu)} - \underbrace{\frac{y^2}{2\sigma^2}}_{d(y)}\right\} \end{aligned}$$

$\because a(y) = y$,

$-\infty < y < \infty$: range of y does not depend on μ

$\therefore y \sim N(\mu, \sigma^2)$ is a distribution in the Exponential Family with Canonical form.

b. Link function

$$g(\mu) = \mathbf{X}\boldsymbol{\beta}$$

Ex.

$g(\mu) = \mu$	$y \sim Normal$
$g(\mu) = \lg \frac{\pi}{1-\pi}$	$y \sim Binomial$

c. Linear component

$\beta_0 + \beta_1 X_1 + \beta_2 X_2$	✓
$\beta_0 + \beta_1 X_1^2 + \beta_2 X_2$	✓
$\beta_0 + e^{\beta_1} X_1 + \beta_2 X_2$	✓
$\beta_0 + e^{\beta_1 X_1} + \beta_2 X_2$	✗

2. Multiple regression vs. Multivariate regression

Multiple regression	Multivariate regression
Univariate: one response variable Y	Multivariate: multiple response variables \mathbf{Y}
$Y_1 \text{ vs. } X_1, X_2$ $Y_2 \text{ vs. } X_1, X_2$	$\mathbf{Y} \text{ vs. } X_1, X_2$ \mathbf{Y} : a vector of response variables

3. Generalized linear model vs. General linear model

	Types of Y
Generalized linear model	$Y \sim N(\mu, \sigma^2) \Rightarrow Y \sim \text{Exp. Family with Canonical form}$
Generalized linear model	Univariate \Rightarrow Multivariate Independent \Rightarrow Dependent