Generalized Linear Models

Son Nguyen

Generalized Linear Model

- The GLM models $\mu = E(y)$ as follows.

$$g(\mu) = \beta_0 + \beta_1 x_1 + \ldots + \beta_p x_p = x'\beta$$

  where y is assumed to follow an exponential distribution family.

- Exponential distribution family includes all the basic distribution such as normal distribution, binomial distribution, Poisson distribution...

- $g(\mu)$ is called the canonical link function

- For logistic regression, the link function is a logit function

$$g(x) = \ln\left(\frac{x}{1-x}\right)$$

Some GLMs

$$g(\mu) = \beta_0 + \beta_1 x_1 + \ldots + \beta_p x_p = x'\beta$$

Distribution	Canonical Link Function	Mathematical Form
Normal	Identity	$g(\mu) = \mu$
Binomial	Logit	$g(\pi) = \ln[\pi/(1-\pi)]$
Poisson	Natural log	$g(\mu) = \ln \mu$
Gamma	Inverse	$g(\mu) = 1/\mu$
Inverse Gaussian	Squared inverse	$g(\mu) = 1/\mu^2$

Two Elements of GLM

- Response Assumptions
- Link Function

Goodness of Fit: Deviance

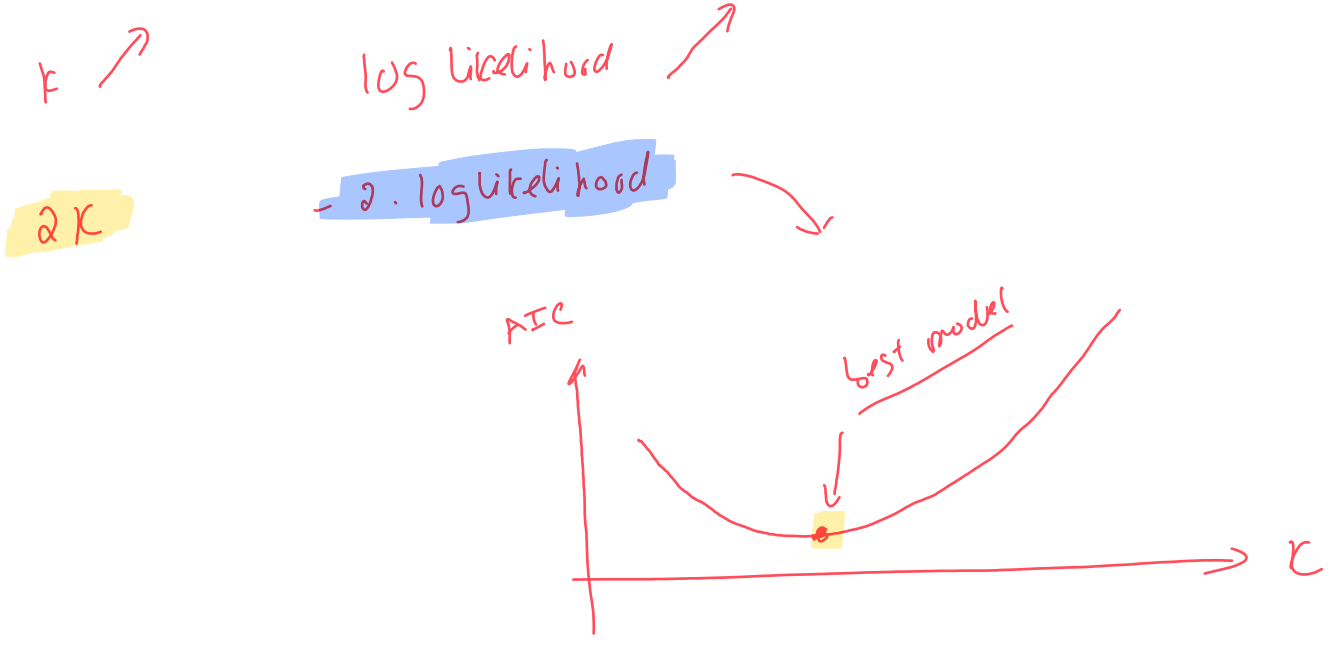
- The deviance generalizes the Residual Sum of Squares (RSS) of the linear model
- Compare three models
- Model 1: The Perfect Model (Saturated Model)
- Model 2: Your model
- Model 3: The worst model: does not use any predictors. Also called Null Model.
- The deviance can be considered the "distance" of the model to the perfect model.
- The smaller the deviance the better the model
- Deviance 0 means the model is perfect!
- The deviance of Model 3 is also called Null Deviance.
- $R^2 = 1 - \frac{\text{Deviance of your model}}{\text{Deviance of the perfect model}}$

Goodness of Fit: The Loglikelihood

- The loglikelihood of a model measures how likely the data is governed by the model.
- The higher the loglikelihood value, the better the model.
- Adding varibles to the models will increase the loglikelihood of the model.

Goodness of Fit: AIC

- AIC = Akaike information criterion
- AIC = 2k - 2 * loglikelihood of the model.
- Smaller AIC means larger loglikelihood, or better model



Goodness of Fit: BIC

- BIC = Bayesian information criterion
- BIC = k·ln(n) - 2 * loglikelihood of the model.
- Smaller BIC means larger loglikelihood, or better model