# Methods of Applied Statistics

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# Contents

1	$\mathbf{Rev}$	riew	<b>2</b>
	1.1	Generalized Linear Model	2
	1.2	Likelihood Ratio Test	2

## 1 Review

#### 1.1 Generalized Linear Model

The GLM is

$$Y_i \sim G(\mu_i, \theta)$$
$$h(\mu_i) = X_i^T \beta$$

where G is the distribution of the response variable,  $\mu_i$  is a location parameter for observation i,  $\theta$  are additional/non-linear parameters for the density of G, h is a link function,  $X_i$  are covariates for observation i,  $\beta$  is a vector of regression coefficients/fixed effects.

**Example 1.1** (OLS). G is a Normal distribution,  $\theta$  is the variance parameter, denoted  $\sigma^2$ , h is the identity function, i.e.,

$$Y_i \sim \mathcal{N}(\mu_i, \sigma^2)$$
$$\mu_i = X_i^T \beta$$

**Example 1.2** (Binomial/Logistic Regression). G is a binomial distribution or a Bernoulli if  $N_i = 1$ , h is the logit link,  $X_i^T \beta$  can be negative,  $\mu_i$  is between 0 and 1, i.e.,

$$Y_i \sim \text{Binomial}(N_i, \mu_i)$$

$$\ln\left(\frac{\mu_i}{1 - \mu_i}\right) = X_i^T \beta$$

### 1.2 Likelihood Ratio Test

$$2[\ln L(\hat{\beta}; \mathbf{y}) - \ln L(\beta; \mathbf{y})] \sim \chi_P^2$$

where P is the number of parameters in  $\beta$ .