

PS5841

Data Science in Finance & Insurance

Poisson Regression

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Contingency Tables

- No constraints

$$E(Y_i) = \mu_i = n_i \theta_i$$

- μ_i Exposure dependent

- $\frac{\mu_i}{n_i}$ occurrence rate

- Constant exposure, $n_i = 1$

- Model with Poisson Distribution

$$f(\mathbf{y}; \boldsymbol{\mu}) = \prod_{i=1}^N \frac{\mu_i^{y_i} e^{-\mu_i}}{y_i!}, E(Y_i) = \mu_i$$


Poisson Distribution

- $Y_i \sim Po(\mu_i), i = 1, \dots, N$

$$f(y_i) = \frac{\mu_i^{y_i} e^{-\mu_i}}{y_i!}, y = 0, 1, 2, \dots$$

$$l(\boldsymbol{\mu}; \mathbf{y}) = \sum_{i=1}^N [y_i \log(\mu_i) - \mu_i - \log y_i!]$$

Poisson Regression

$$E(Y_i) = \mu_i = n_i \theta_i$$
$$\log(\mu_i) = \log(n_i) + \mathbf{x}_i^T \boldsymbol{\beta}$$


offset

Goodness of Fit (1)

- Pearson chi-squared statistic

$$X^2 = \sum_{i=1}^n \frac{(o_i - e_i)^2}{e_i} \sim \chi^2(N - p)$$

- Pearson residuals

$$r_i = \frac{o_i - e_i}{\sqrt{e_i}}$$

$$\sum_{i=1}^N r_i^2 = X^2$$

Goodness of Fit (2)

- Pseudo R^2

$$pseudo\ R^2 = \frac{l(\hat{\boldsymbol{\beta}}_{min}) - l(\hat{\boldsymbol{\beta}})}{l(\hat{\boldsymbol{\beta}}_{min})}$$

That was

