

Assignment 1

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1) For each of the situations below, identify the response and explanatory variables, variable types, and the generalized linear model that is well-suited to model the data. Make sure to justify your answer. Write down the linear predictor as well as the link function.

- **The effect of age, sex, height, daily food intake and daily exercise on a person's weight.**

Response variable: W = Weight (Continuous, Normally distributed)

Explanatory variables: A = Age (Discrete), S = Sex (Nominal), H = Height (Continuous), F = Daily food intake (Continuous), E = Daily exercise (Continuous)

Because response variable is continuous and explanatory variables contain numeric and nominal variable, we should choose multiple regression.

Which is like: $E(W) = \beta_0 + \beta_A x_A + \beta_S x_S + \beta_H x_H + \beta_F x_F + \beta_E x_E$

Link function: $\eta = \mu$

- **The percentage of full-time graduate students that find employment upon graduation. For each student, sex, age, grades, major, prior years of work experience, and prior income levels are available.**

Response variable: P = Percentage of full-time graduate students that find employment upon graduation (Count data as proportion)

Explanatory variables: S = Sex (Nominal), A = Age (Discrete), G = Grades (Ordinal), M = Major (Nominal), W = Prior years of work experience (Discrete), I = Prior income levels (Ordinal)

Because response variable is count data as proportion and there are different types of explanatory variables, we should use Binomial regression.

Which is like: $P = \frac{\exp(\beta_0 + \beta_S x_S + \beta_A x_A + \beta_G x_G + \beta_M x_M + \beta_W x_W + \beta_I x_I)}{1 + \exp(\beta_0 + \beta_S x_S + \beta_A x_A + \beta_G x_G + \beta_M x_M + \beta_W x_W + \beta_I x_I)}$;

Link function: $\eta = \ln\left(\frac{\mu}{1 - \mu}\right)$

- **The number of mortgage loan defaults in a given year by different counties across the United States. For each household/borrower information on income, loan interest rate, age, debt, loan to value at origination are available.**

Response variable: D = Number of defaults. (Count data)

Explanatory variables: I = Income (Continuous), L = Loan interest rate (Continuous), A = Age (Discrete), B = Debt (Continuous), O = loan to value at origination (Continuous)

Because response variable is count data, we should use Poisson regression

Which is like: $\ln \lambda = (\beta_0 + \beta_I x_I + \beta_L x_L + \beta_A x_A + \beta_B x_B + \beta_O x_O); D \sim \text{Poisson}(\lambda)$

Link function: $\eta = \ln(\mu)$

2) Show that the probability distributions below belong to the exponential family. Determine the functions $a()$, $b()$, $c()$, and $d()$ from the general exponential distribution for each of the probability distributions below.

- Pareto distribution:

$$\begin{aligned} f(y; \theta) &= \theta y^{-\theta-1} \\ &= \exp\{\ln[\theta y^{-(\theta+1)}]\} \\ &= \exp[\ln \theta - (\theta + 1) \ln y] \\ &= \exp(\ln \theta - \theta \ln y - \ln y) \end{aligned}$$

$$\text{So, } a(y) = \ln y, b(\theta) = -\theta, c(\theta) = \ln \theta, d(y) = -\ln y$$

- Exponential distribution:

$$\begin{aligned} f(y; \theta) &= \theta e^{-y\theta} \\ &= \exp(\ln \theta - y\theta) \end{aligned}$$

$$\text{So, } a(y) = y, b(\theta) = -\theta, c(\theta) = \ln \theta, d(y) = 0$$

- Negative binomial:

$$\begin{aligned} f(y; \theta) &= \binom{y+r-1}{r-1} \theta^r (1-\theta)^y \\ &= \exp\{\ln[\binom{y+r-1}{r-1} \theta^r (1-\theta)^y]\} \\ &= \exp\left[\ln\binom{y+r-1}{r-1} + r \ln \theta + y \ln(1-\theta)\right] \end{aligned}$$

$$\text{So, } a(y) = y, b(\theta) = \ln(1-\theta), c(\theta) = r \ln \theta, d(y) = \ln\binom{y+r-1}{r-1}$$

3) Show that the gamma distribution with a scale parameter θ and nuisance parameter ϕ belongs to the exponential family of distributions. Using the properties of the exponential distributions, find $E[Y]$ and $VAR[Y]$.

$$\begin{aligned}
 f(y; \theta) &= \frac{y^{\phi-1} \theta^{\phi} e^{-y\theta}}{\Gamma(\phi)} \\
 &= \exp[\ln y^{(\phi-1)} + \ln \theta^{\phi} + \ln e^{-y\theta} - \ln \Gamma(\phi)] \\
 &= \exp[(\phi - 1) \ln y + \phi \ln \theta - y\theta - \ln \Gamma(\phi)]
 \end{aligned}$$

$$\text{So, } a(y) = y, b(\theta) = -\theta, c(\theta) = \phi \ln \theta - \ln \Gamma(\phi), d(y) = (\phi - 1) \ln y$$

$$\begin{aligned}
 E[a(Y)] &= -\frac{c'(\theta)}{b'(\theta)} \\
 &= -\frac{(\phi \ln \theta - \ln \Gamma(\phi))'}{-\theta'} \\
 &= \frac{\phi}{\theta}
 \end{aligned}$$

$$\begin{aligned}
 VAR[a(Y)] &= \frac{b''(\theta)c'(\theta) - c''(\theta)b'(\theta)}{[b'(\theta)]^3} \\
 &= \frac{(-\theta)''(\phi \ln \theta - \ln \Gamma(\phi))' - (\phi \ln \theta - \ln \Gamma(\phi))''(-\theta)'}{(-\theta')^3} \\
 &= \frac{0 - (-\frac{\phi}{\theta^2})(-1)}{(-1)^3} \\
 &= \frac{\phi}{\theta^2}
 \end{aligned}$$