

Exponential Family Examples

P.12. Bernoulli(ϕ)

$$\begin{aligned}
 p(y; \phi) &= \phi^y (1-\phi)^{1-y} \\
 &= e^{\log \phi^y (1-\phi)^{1-y}} \\
 &= e^{y \log \phi + (1-y) \log (1-\phi)} \\
 &= e^{y \log \phi + \log (1-\phi) - y \log (1-\phi)} \\
 &= e^{y \log \frac{\phi}{1-\phi} + \log (1-\phi)} \\
 &= \underbrace{1}_{b(y)} \cdot e^{\underbrace{y \log \frac{\phi}{1-\phi}}_{T(y)} - \underbrace{(-\log (1-\phi))}_{\text{log-partition function } a(\eta)}}
 \end{aligned}$$

Write $-\log(1-\phi)$ as a function of η :

First, solve for ϕ in $\eta = \log \frac{\phi}{1-\phi}$:

$$\eta = \log \frac{\phi}{1-\phi}$$

$$e^\eta = \frac{\phi}{1-\phi}$$

$$e^\eta - e^\eta \phi = \phi$$

$$e^\eta = \phi + e^\eta \phi = (1 + e^\eta) \phi$$

$$\phi = \frac{e^\eta}{1 + e^\eta} = \frac{1}{e^{-\eta} + 1} \quad \text{sigmoid function}$$

Substitute ϕ in $\log \frac{\phi}{1-\phi} - \log(1-\phi)$

$$a(\eta) = -\log\left(1 - \frac{e^\eta}{1 + e^\eta}\right)$$

$$= -\log\left(\frac{1}{1 + e^\eta}\right)$$

$$= -(-\log(1 + e^\eta))$$

$$= \log(1 + e^\eta) \leftarrow a(\eta)$$

P.13. Normal distribution $N(\mu, 1)$

$$\begin{aligned}
 p(y; \mu) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y-\mu)^2} = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y^2 + \mu^2 - 2y\mu)} \\
 &= \left[\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} \right] e^{-\frac{1}{2}(\mu^2 - 2y\mu)} \\
 &= \left[\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} \right] e^{\underbrace{\mu y}_{T(y)} - \underbrace{\frac{\mu^2}{2}}_{a(\eta)}} \quad a(\eta) = \frac{\mu^2}{2} = \frac{\eta^2}{2} \\
 &\quad \text{inverse: } \mu = \eta
 \end{aligned}$$

p16. Poisson(λ)

$$\begin{aligned}
 p(y; \lambda) &= \frac{\lambda^y e^{-\lambda}}{y!} = \frac{1}{y!} \lambda^y e^{-\lambda} = \frac{1}{y!} e^{\log \lambda^y} e^{-\lambda} \\
 &= \frac{1}{y!} e^{y \log \lambda - \lambda} \\
 &= \underbrace{\frac{1}{y!}}_{b(y)} e^{\underbrace{\log \lambda \cdot y - \lambda}_{\eta - T(\eta)}} \text{ partition function } a(\eta)
 \end{aligned}$$

since $\eta = \log \lambda$,
we have $\lambda = e^\eta$

Then, $a(\eta) = \lambda = \boxed{e^\eta} \leftarrow a(\eta)$

p26. Multinomial(ϕ_1, \dots, ϕ_k)

Since $\sum_{i=1}^k \phi_i = 1$, it suffices to ~~have~~ ^{use} $k-1$ parameters to define a multinomial distribution of k random variables.

Let $T(y) \in \mathbb{R}^{k-1}$ be defined as $T(y) = \begin{bmatrix} 1\{y=1\} \\ \vdots \\ 1\{y=k-1\} \end{bmatrix}$
where $1\{y=i\}$ is the indicator function $1\{y=i\} = \begin{cases} 1 & y=i \\ 0 & y \neq i \end{cases}$

Denote the i th element of $T(y)$ as $T(y)_i = 1\{y=i\}$

$$\begin{aligned}
 p(y; \phi) &= \prod_{i=1}^k \phi_i^{1\{y=i\}} \\
 &= \left(\prod_{i=1}^{k-1} \phi_i^{T(y)_i} \right) \phi_k^{1\{y=k\}} = \left(\prod_{i=1}^{k-1} \phi_i^{T(y)_i} \right) \phi_k^{1 - \sum_{i=1}^{k-1} T(y)_i} \\
 &= e^{T(y)_i \log \prod_{i=1}^{k-1} \phi_i + (1 - \sum_{i=1}^{k-1} T(y)_i) \log \phi_k} \\
 &= e^{T(y)_i \sum_{i=1}^{k-1} \log \phi_i + \log \phi_k - \sum_{i=1}^{k-1} T(y)_i \log \phi_k} \\
 &= e^{\sum_{i=1}^{k-1} (T(y)_i \log \phi_i - T(y)_i \log \phi_k) + \log \phi_k} \\
 &= e^{\sum_{i=1}^{k-1} (T(y)_i \log \frac{\phi_i}{\phi_k}) + \log \phi_k} \\
 &= e^{\underbrace{\begin{bmatrix} \log \frac{\phi_1}{\phi_k} \\ \vdots \\ \log \frac{\phi_{k-1}}{\phi_k} \end{bmatrix}^T}_{\eta} \underbrace{T(y)}_{\begin{bmatrix} 1\{y=1\} \\ \vdots \\ 1\{y=k-1\} \end{bmatrix}} + \underbrace{(\log \phi_k)}_{a(\eta)}}
 \end{aligned}$$

\rightarrow How to write it as a function of η ?
 $a(\eta) = -\log \phi_k$

$b(y) = 1$

Since $\eta = \begin{bmatrix} \log \frac{\phi_1}{\phi_k} \\ \vdots \\ \log \frac{\phi_{k-1}}{\phi_k} \end{bmatrix}$, we generalize the notation η_i to include $i=k$.
 i.e. $\eta_i = \log \frac{\phi_i}{\phi_k}$ ($1 \leq i \leq k$)
 and $\eta_k = \log \frac{\phi_k}{\phi_k} = 0$

Then we have $\eta_i = \log \frac{\phi_i}{\phi_k}$ ← canonical link function

$$e^{\eta_i} = \frac{\phi_i}{\phi_k}$$

$$\phi_i = \phi_k \cdot e^{\eta_i}$$

(1)

Since $\sum_{i=1}^k \phi_i = 1$, $1 = \sum_{i=1}^k \phi_i = \sum_{i=1}^k \phi_k \cdot e^{\eta_i}$

Then we can solve for ϕ_k :

$$1 = \phi_k \sum_{i=1}^k e^{\eta_i}$$

$$\phi_k = \frac{1}{\sum_{i=1}^k e^{\eta_i}}$$

(2)

Then $a(\eta) = -\log \phi_k = -\log \frac{1}{\sum_{i=1}^k e^{\eta_i}} = \log \sum_{i=1}^k e^{\eta_i}$ ← $a(\eta)$

Further, plug (2) into (1),

$$\phi_i = \frac{e^{\eta_i}}{\sum_{i=1}^k e^{\eta_i}}$$

← canonical response function