

1. This distribution is not in the exponential family. We notice here that the support $\theta < y < \infty$ is dependent on the parameter of the distribution θ . Therefore, regardless of how we can write the density, this distribution is not in the exponential family.
2. (a) Let $Y \sim \text{Geom}(p)$. Then we can write its density as

$$\mathbb{P}(Y = k) = (1 - p)^k p = \exp \left\{ k \log(1 - p) + \log(p) \right\}$$

Let $a(\phi) = 1$, $c(y, \phi) = 0$, $\theta = \log(1 - p)$, and lastly $b(\theta) = -\log(1 - \exp(\theta))$.

(b)

$$\begin{aligned} \mu &= \frac{d}{dt} b(t) \Big|_{t=\theta} = \frac{\exp(t)}{1 - \exp(t)} \Big|_{t=\theta} = \frac{1 - p}{1 - 1 + p} = \frac{1 - p}{p} \\ V &= a(\phi) \frac{d^2}{dt^2} b(t) \Big|_{t=\theta} = \frac{d}{dt} \frac{\exp(t)}{1 - \exp(t)} \Big|_{t=\theta} = \frac{(1 - \exp(t)) \exp(t) + \exp(t)^2}{(1 - \exp(t))^2} \Big|_{t=\theta} \\ &= \frac{p(1 - p) + (1 - p)^2}{p^2} = \frac{p - p^2 + 1 - 2p + p^2}{p^2} = \frac{1 - p}{p^2} = \frac{\mu}{p} \end{aligned}$$

- (c) The canonical link function is given by the function $g(\mu) = \theta$. Identically, it is the inverse of $\frac{d}{dt} b(t) = \frac{\exp(t)}{1 - \exp(t)}$ which is satisfied by the function $g(t) = \log\left(\frac{t}{1+t}\right)$. To check this choice, notice that

$$g(\mu) = g\left(\frac{1 - p}{p}\right) = \log\left(\frac{1 - p}{p} / 1 + \frac{1 - p}{p}\right) = \log\left(\frac{1 - p}{p} / \frac{1}{p}\right) = \log(1 - p) = \theta$$

- (d) Here we compare a proposed model to the saturated model where $\tilde{\theta}_i = \log\left(\frac{y_i}{1 + y_i}\right)$ and $\hat{\theta}_i = \log\left(\frac{\hat{\mu}_i}{1 + \hat{\mu}_i}\right)$

$$\begin{aligned} D(y_i, \hat{\mu}_i) &= 2 \sum_{i=1}^n \left[y_i (\tilde{\theta}_i - \hat{\theta}_i) - (b(\tilde{\theta}_i) - b(\hat{\theta}_i)) \right] \\ &= 2 \sum_{i=1}^n \left[y_i \log\left(\frac{y_i(1 - \hat{\mu}_i)}{\hat{\mu}_i(1 - y_i)}\right) - \log\left(\frac{1 + y_i}{1 - \hat{\mu}_i}\right) \right] \end{aligned}$$

3. (a) Suppose that $Y \sim f(y|\mu) = \left(\frac{\lambda}{2\pi y^3}\right)^{1/2} \exp\left\{\frac{-\lambda(y-\mu)^2}{2\mu^2 y}\right\}$ where λ is known. Notice

that we can write this distribution in canonical form as follows.

$$\begin{aligned}
 f(y|\mu) &= \exp \left\{ \frac{1}{2} \log(\lambda) - \frac{1}{2} \log(2\pi y^3) - \frac{\lambda(y - \mu)^2}{2\mu^2 y} \right\} \\
 &= \exp \left\{ \frac{1}{2} \log(\lambda) - \frac{1}{2} \log(2\pi y^3) - \frac{\lambda(y^2 - 2\mu y + \mu^2)}{2\mu^2 y} \right\} \\
 &= \exp \left\{ \frac{1}{2} \log(\lambda) - \frac{1}{2} \log(2\pi y^3) - \frac{\lambda y}{2\mu^2} + \frac{\lambda}{\mu} - \frac{\lambda}{2y} \right\} \\
 &= \exp \left\{ \frac{y/\mu^2 - 2/\mu}{-2/\lambda} - \frac{\lambda}{2y} + \frac{1}{2} \log(\lambda) - \frac{1}{2} \log(2\pi y^3) \right\}
 \end{aligned}$$

Now setting $a(\phi) = a(\lambda) = -2/\lambda$, $\theta = 1/\mu^2$, $b(\theta) = 2\sqrt{\theta}$, and lastly $c(y, \lambda) = -\frac{\lambda}{2y} + \frac{1}{2} \log(\lambda) - \frac{1}{2} \log(2\pi y^3)$ we see that this is in the exponential family.

(b)

$$\begin{aligned}
 \mathbb{E}(Y) &= \frac{d}{dt} b(t) \Big|_{t=\theta} = \frac{1}{\sqrt{\theta}} = \frac{1}{\sqrt{1/\mu^2}} = \mu \\
 V &= a(\lambda) \frac{d^2}{dt^2} b(t) \Big|_{t=\theta} = \frac{-2}{\lambda} \left(\frac{-1}{2} \theta^{-3/2} \right) = \frac{1}{\lambda(\sqrt{\theta})^3} = \frac{\mu^3}{\lambda}
 \end{aligned}$$

(c) The canonical link is given by the function that satisfies $g(\mu) = \theta = 1/\mu^2$.
Therefore the link function is given by $g(t) = \frac{1}{t^2}$

(d) Here we compare a proposed model to the saturated model where $\tilde{\theta}_i = \frac{1}{y_i^2}$ and $\hat{\theta}_i = \frac{1}{\hat{\mu}_i^2}$.

$$\begin{aligned}
 D(y_i, \hat{\mu}_i) &= 2 \sum_{i=1}^n \left[y_i(\tilde{\theta}_i - \hat{\theta}_i) - (b(\tilde{\theta}_i) - b(\hat{\theta}_i)) \right] \\
 &= 2 \sum_{i=1}^n \left[y_i \left(\frac{1}{y_i^2} - \frac{1}{\hat{\mu}_i^2} \right) - \left(\frac{2}{y_i} - \frac{2}{\hat{\mu}_i} \right) \right] \\
 &= 2 \sum_{i=1}^n \left[\frac{2}{\hat{\mu}_i} - \frac{y_i}{\hat{\mu}_i^2} - \frac{1}{y_i} \right]
 \end{aligned}$$