

1. Suppose $\theta \sim \text{Beta}(a, b)$ s.t.

$$P(\theta | a, b) = \frac{1}{B(a, b)} \theta^{a-1} (1-\theta)^{b-1}$$

$$= \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \theta^{a-1} (1-\theta)^{b-1}$$

where $B(a, b) = \Gamma(a)\Gamma(b)/\Gamma(a+b)$

& $\Gamma(x)$ is the gamma function.

Derive the mean, mode, & variance of

$$B(a, b) = \int_0^1 \theta^{a-1} (1-\theta)^{b-1} d\theta = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$$

$$\Gamma(x+y) = x\Gamma(x)$$

$$E[\theta] = \int_0^1 \theta P(\theta | a, b) d\theta = \int_0^1 \theta \left(\frac{1}{B(a, b)} \theta^{a-1} (1-\theta)^{b-1} \right) d\theta$$

$$= \frac{1}{B(a, b)} \int_0^1 \theta^a (1-\theta)^{b-1} d\theta$$

$$= \frac{B(a+1, b)}{B(a, b)}$$

$$= \left[\frac{\Gamma(a+1)\Gamma(b)}{\Gamma(a+b+1)} \right] \left[\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \right]$$

$$= \left[\frac{a\Gamma(a)\Gamma(b)}{(a+b)\Gamma(a+b)} \right] \left[\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \right]$$

$$= \boxed{\frac{a}{a+b}} \text{ Mean}$$

Now def:

$$\text{Var}[\theta] = E[\theta^2] - E[\theta]^2$$

$$E[\theta^2] = \int_0^1 \theta^2 \left(\frac{1}{B(a, b)} \theta^{a-1} (1-\theta)^{b-1} \right) d\theta$$

$$= \frac{1}{B(a, b)} \int_0^1 \theta^{a+1} (1-\theta)^{b-1} d\theta = \frac{B(a+2, b)}{B(a, b)}$$

$$= \left[\frac{a(a+1)\Gamma(a)\Gamma(b)}{(a+b)(a+b+1)\Gamma(a+b)} \right] \left[\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \right]$$

$$= \frac{a(a+1)}{(a+b)(a+b+1)}$$

$$E[\theta]^2 - E[\theta]^2$$

$$= \frac{a(a+1)}{(a+b)(a+b+1)} - \frac{a^2}{(a+b)^2}$$

$$= \frac{a^2 + a^2b + a^2 + ab - a^3 - a^2b}{(a+b)^2(a+b+1)}$$

$$= \boxed{\frac{ab}{(a+b)^2(a+b+1)}} \text{ Variance}$$

$$\nabla_{\theta} P(\theta | a, b) = 0 \text{ on interval } [0, 1]$$

$$= \nabla_{\theta} [\theta^{a-1} (1-\theta)^{b-1}] = 0$$

$$(a-1)\theta^{a-2}(1-\theta)^{b-1} - (b-1)\theta^{a-1}(1-\theta)^{b-2} = 0$$

$$(a-1)\theta^{a-2}(1-\theta)^{b-1} = (b-1)\theta^{a-1}(1-\theta)^{b-2}$$

$$= (b-1)\theta$$

$$= a-1$$

$$\theta = \boxed{\frac{a-1}{a+b-1}}$$

I was given
very similar
question

Like this

$$2 \quad \text{Cat}(x|u) = \prod_{i=1}^k \mu_i^{x_i}$$

$$p(y; \eta) = b(y) \exp(\eta^T T(y) - a(\eta))$$

$$\begin{aligned} \text{Cat}(x|u) &= \prod_{i=1}^k \mu_i^{x_i} = \exp \left[\log \left(\prod_{i=1}^k \mu_i^{x_i} \right) \right] \\ &= \exp \left(\sum_{i=1}^k x_i \log(\mu_i) \right) \\ &= \exp \left(\sum_{i=1}^k x_i \log(\mu_i) \right) \end{aligned}$$

$$= \exp \left(\sum_{i=1}^k x_i \log(\mu_i) \right) = \exp \left(\sum_{i=1}^{k-1} x_i \log(\mu_i) + x_k \log(\mu_k) \right)$$

$$= \exp \left(\sum_{i=1}^{k-1} x_i \log(\mu_i) + \left(1 - \sum_{i=1}^{k-1} x_i \right) \log(\mu_k) \right)$$

$$= \exp \left(\sum_{i=1}^{k-1} x_i (\log(\mu_i) - \log(\mu_k)) + \log(\mu_k) \right)$$

$$= \exp \left[\sum_{i=1}^{k-1} x_i \log \left(\frac{\mu_i}{\mu_k} \right) + \log(\mu_k) \right]$$

$$\eta = \begin{bmatrix} \log \left(\frac{\mu_1}{\mu_k} \right) \\ \vdots \\ \log \left(\frac{\mu_{k-1}}{\mu_k} \right) \end{bmatrix}$$

$$\mu_k = 1 - \sum_{i=1}^{k-1} \mu_i = \left(1 - \sum_{i=1}^{k-1} \mu_k e^{\eta_i} \right)$$

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

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

$$\mu_i = \mu_k e^{\eta_i} = \frac{e^{\eta_i}}{1 + \sum_{i=1}^{k-1} e^{\eta_i}}$$

$$b(\eta) = 1$$

$$T(x) = x$$

$$a(\eta) = -\log(\mu_k) = \log \left(1 + \sum_{i=1}^{k-1} e^{\eta_i} \right) \quad \therefore \text{GLM of this dist.}$$

$$\mu = S(\eta) \text{ where } S(\eta) \text{ is softmax func.}$$

Same as softmax regression