

# Gumbel Softmax Estimation for Binary Random Variables

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In this script, we will show that given a normalized Bernoulli logit  $\alpha = [p, 1 - p]$ , the corresponding Gumbel softmax estimation is

$$\sigma \left\{ -\frac{1}{\tau} \left( \log \frac{p}{1-p} + \log \frac{u}{1-u} \right) \right\}, \quad (1)$$

where  $\sigma(x) = \frac{1}{1+e^{-x}}$  is a sigmoid function,  $\tau$  is a temperature coefficient,  $u \sim \text{Uniform}(0, 1)$ . It turns out that  $p(\lim_{\tau \rightarrow 0} x = 1) = p$ . Similarly, if given an unnormalized Bernoulli logit  $\alpha = [\alpha_1, \alpha_2]$ , the corresponding Gumbel softmax estimation is

$$\sigma \left\{ -\frac{1}{\tau} \left( \log \frac{\alpha_1}{\alpha_1 + \alpha_2} + \log \frac{u}{1-u} \right) \right\}. \quad (2)$$

It turns out that  $p(\lim_{\tau \rightarrow 0} x = 1) = \frac{\alpha_1}{\alpha_1 + \alpha_2}$ .

Let's first recall the Gumbel softmax [1] estimation for Category random variables. Given an unnormalized categorical logits  $\alpha = [\alpha_1, \dots, \alpha_n]$ , the Gumbel softmax estimation is

$$\frac{\exp \{(\log \alpha_k + g_k)/\tau\}}{\sum_i \exp \{(\log \alpha_i + g_i)/\tau\}}, \quad (3)$$

where  $g = [g_1, \dots, g_n]$  is standard Gumbel random variables. It turns out that  $p(\lim_{\tau \rightarrow 0} x_k = 1) = \frac{\alpha_k}{\sum_i \alpha_i}$ .

Next, it is helpful to know that the difference of two independent Gumbel random variables  $X_1 \sim \text{Gumbel}(a_1), X_2 \sim \text{Gumbel}(a_2)$  follows a Logistic distribution, *i.e.*,  $X_1 - X_2 \sim \text{Logistic}(\alpha_1 - \alpha_2)$  (see section 1.2 in [my another note](#) for details). Moreover, sampling from a standard logistic distribution can be done by  $x \sim \text{Logistic}(1) \Leftrightarrow x = \log \frac{u}{1-u}, u \sim \text{Uniform}(0, 1)$ . Now, we establish sufficient prerequisites to prove Gumbel softmax estimation for the binary case. Specifically,

$$\begin{aligned} \text{Binary Gumbel Softmax} &= \frac{\exp \{(\log p + g_1)/\tau\}}{\exp \exp \{(\log p + g_1)/\tau\} + \exp \exp \{(\log(1-p) + g_2)/\tau\}} \\ &= \frac{1}{1 + \exp \{(\log(1-p) - \log p + g_2 - g_1)/\tau\}} \\ &= \frac{1}{1 + \exp \{(\log(1-p) - \log p + \log u - \log(1-u))/\tau\}} \\ &= \sigma \left\{ -\frac{1}{\tau} \left( \log \frac{p}{1-p} + \log \frac{u}{1-u} \right) \right\}. \end{aligned}$$

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## References

- [1] Eric Jang, Shixiang Gu, and Ben Poole. Categorical reparameterization with gumbel-softmax. *arXiv preprint arXiv:1611.01144*, 2016.