

## Power Series Randomization

We say that the discrete random variable  $x$  follows the **Power Series Distribution** if

$$P(X = x) = \frac{a_x \theta^x}{\eta(\theta)} \quad x = 0, 1, 2, \dots$$

for any coefficients  $a_x \geq 0$  where

$$\eta(\theta) = \sum_{x=0}^{\infty} a_x \theta^x$$

Suppose that  $X_1, \dots, X_n$  are independent and identically distribution **Power Series** random variables. Then for  $t = 0, 1, \dots$

$$\begin{aligned} P(X_1 + \dots + X_n = t) &= \sum_{\substack{(x_1, \dots, x_n) \ni \\ x_1 + \dots + x_n = t}} \frac{a_{x_1} \cdots a_{x_n} \theta^t}{(\eta(\theta))^n} \\ &= \frac{\theta^t}{(\eta(\theta))^n} c(t, n) \end{aligned}$$

where

$$c(t, n) = \sum_{\substack{(x_1, \dots, x_n) \ni \\ x_1 + \dots + x_n = t}} a_{x_1} \cdots a_{x_n}$$

Furthermore,

$$\begin{aligned} 1 &= \sum_{t=0}^{\infty} \frac{\theta^t}{(\eta(\theta))^n} c(t, n) \\ &= \frac{1}{(\eta(\theta))^n} \sum_{t=0}^{\infty} c(t, n) \theta^n \end{aligned}$$

Therefore,

$$(\eta(\theta))^n = \sum_{t=0}^{\infty} c(t, n) \theta^n$$

Define  $\mathbb{S}^n$  to be the product space  $\{0,1,\dots\} \times \dots \times \{0,1,\dots\}$  and let  $\mathbb{S}_t^n$  be the set of all vectors  $(s_1, s_2, \dots, s_n)$  in  $\mathbb{S}^n$  such that  $s_1 + \dots + s_n = t$ .

Let  $\mathcal{A} \subseteq \mathbb{S}^n$  and define  $\mathcal{A}_t = \mathcal{A} \cap \mathbb{S}_t^n$ . It follows that

$$\begin{aligned} P((X_1, \dots, X_n) \in \mathcal{A}) &= \sum_{t=0}^{\infty} P\left((X_1, \dots, X_n) \in \mathcal{A} \mid \sum_{i=1}^{\infty} X_i = t\right) P\left(\sum_{i=1}^{\infty} X_i = t\right) \\ &= \sum_{t=0}^{\infty} \sum_{\mathcal{A}_t} \left( \frac{\frac{a_{x_1} \cdots a_{x_n} \theta^t}{(\eta(\theta))^n}}{\frac{\theta^t}{(\eta(\theta))^n} c(t, n)} \right) \frac{c(t, n)}{(\eta(\theta))^n} \theta^t \\ &= \sum_{t=0}^{\infty} \sum_{\mathcal{A}_t} \left( \frac{a_{x_1} \cdots a_{x_n}}{c(t, n)} \right) \frac{c(t, n)}{(\eta(\theta))^n} \theta^t \end{aligned}$$

Therefore,

$$(\eta(\theta))^n P((X_1, \dots, X_n) \in \mathcal{A}) = \sum_{t=0}^{\infty} P((Y_{1,t}, \dots, Y_{n,t}) \in \mathcal{A}_t) c(t, n) \theta^t$$

where

$$P((Y_{1,t}, \dots, Y_{n,t}) = (y_1, \dots, y_n)) = \begin{cases} \frac{a_{y_1} \cdots a_{y_n}}{c(t, n)} & y_1 + \dots + y_n = t, \quad y_j \geq 0 \quad \forall j \\ 0 & \text{otherwise} \end{cases}$$

and

$$c(t, n) = \sum_{\substack{(y_1, \dots, y_n) \ni \\ y_1 + \dots + y_n = t}} a_{y_1} \cdots a_{y_n}$$

Hence,

$$\frac{d^r}{d\theta^r} ((\eta(\theta))^n P((X_1, \dots, X_n) \in \mathcal{A})) \Big|_{\theta=0} = \sum_{t=0}^{\infty} P((Y_{1,t}, \dots, Y_{n,t}) \in \mathcal{A}_t) c(t, n) r! I_{\{r\}}(t)$$

Therefore,

$$P((Y_{1,r}, \dots, Y_{n,r}) \in \mathcal{A}_r) = \frac{1}{c(r, n) r!} \left( \frac{d^r}{d\theta^r} ((\eta(\theta))^n P((X_1, \dots, X_n) \in \mathcal{A})) \Big|_{\theta=0} \right)$$

We will refer to this result as the **Power Series Randomization Theorem**.