

Convolution: X, Y - independent

$$Z = X + Y.$$

Distribution of Z is called the "Convolution" of the distributions of X & Y .pmf of X : $p_X(x)$, $x \in D_X$ pmf of Y : $p_Y(y)$, $y \in D_Y$ (X, Y) - set of values from D_X and D_Y .

$$\& \ p(x, y) = p_X(x) \cdot p_Y(y). \quad [\because X, Y - \text{ind.}]$$

$$D_Z = \{x + y : x \in D_X, y \in D_Y\}.$$

pmf of Z :

$$p_Z(z) = P(Z = z)$$

$$= P(X + Y = z)$$

$$= \sum_{x \in D_X} P(X = x, Y = z - x) = \sum_{y \in D_Y} P(X = z - y, Y = y)$$

$$= \sum_{x \in D_X} P(X = x) \cdot P(Y = z - x) = \sum_{y \in D_Y} P(X = z - y) \cdot P(Y = y)$$

$$= \underbrace{\sum_{x \in D_X} p_X(x)}_{p_X * p_Y(z)} \cdot \underbrace{\sum_{y \in D_Y} p_Y(z - x)}_{p_Y * p_X(z)}$$

 D_1, D_2 - countable sets $\subset \mathbb{R}$ b, b_2 are pmfs supported on D_1 & D_2 respectively.

D_1, D_2 countable

p_1, p_2 are pmfs supported on D_1 & D_2 respectively.

$$D = \{x+y \mid x \in D_1, y \in D_2\}$$

Define: $p_1 * p_2(z) := \sum_{x \in D_1} p_1(x) \cdot p_2(z-x), \quad z \in D.$

Show that: ① $p_1 * p_2$ is a pmf.

② $p_1 * p_2 = p_2 * p_1$

③ $p_1 * e = p_1$, where $e(0) = 1$
 $e(x) = 0 \quad \forall x \neq 0.$

(i.e. $e \rightarrow$ degenerate r.v.
(degenerate at $x=0$).

Eg: $p_1 = \text{Bin}(n, p)$
 $p_2 = \text{Bin}(m, p).$

Intuitively,
 $p_1 \rightarrow n$ independent Bernoulli trials with $\text{prob}(H) = p.$

$p_2 \rightarrow m$ independent Bernoulli trials with $\text{prob}(H) = p.$

$\therefore p_1 + p_2 \rightarrow (m+n)$ independent Bernoulli trials with $\text{prob}(H) = p.$

$\therefore p_1 + p_2 = \text{Bin}(m+n, p).$

Eg: $p_1 = \text{Geo}(p).$
 $p_2 = \text{Geo}(p).$

$\Rightarrow p_1 * p_2 = \text{N.B}(2, p)$
 \downarrow
negative binomial.

(no. of trials for 2 H)
($\text{prob}(H) = p$)

$$\begin{aligned}
 \text{Ex. } p_1 &= \text{poi}(\lambda_1) \\
 p_2 &= \text{poi}(\lambda_2) \Rightarrow p_1 * p_2(z) = \sum p_1(x) \cdot p_2(z-x) \\
 &= \sum_{x=0}^z \frac{e^{-\lambda_1} \lambda_1^x}{x!} \cdot \frac{e^{-\lambda_2} \lambda_2^{z-x}}{(z-x)!} \\
 &= \frac{e^{-(\lambda_1+\lambda_2)}}{z!} \cdot \underbrace{\sum_{x=0}^z \binom{z}{x} \lambda_1^x \lambda_2^{z-x}}_{\text{Binomial expansion of } (\lambda_1+\lambda_2)^z} \\
 &= \frac{e^{-(\lambda_1+\lambda_2)}}{z!} \cdot (\lambda_1+\lambda_2)^z.
 \end{aligned}$$

$$\therefore p_1 * p_2 = \text{Poi}(\lambda_1 + \lambda_2).$$

Digression:

Power Series around 0

Let $\{a_n\}_{n \geq 0}$ be any real sequence.

$$\text{For } t \in \mathbb{R}, \varphi(t) = \sum_{n=0}^{\infty} a_n t^n.$$

Suppose, radius of convergence $r > 0$

Facts: ① $\varphi(t)$ converges for $|t| < r$, i.e., for $t \in (-r, r)$

② For any $0 < s < r$,

$\varphi(t)$ converges uniformly on $[-s, s]$

$$\left[\begin{array}{l} \text{Given } \varepsilon > 0, \exists N_\varepsilon \in \mathbb{N} \text{ s.t. } \forall t_1, t_2 \in [-s, s], \\ | \varphi(t_1) - \varphi(t_2) | < \varepsilon \quad \forall n > N_\varepsilon \end{array} \right]$$

$$\left[\begin{array}{l} \text{Given } \varepsilon > 0, \exists N_\varepsilon \in \mathbb{N}. \\ |\varphi(t_1) - \varphi(t_2)| < \varepsilon \quad \forall n > N_\varepsilon \end{array} \right]$$

③ $\varphi: (-r, r) \longrightarrow \mathbb{R}$ is a continuous function.

φ is differentiable on $(-r, r)$, and

$$\varphi'(t) = \sum_{n=1}^{\infty} n \cdot a_n \cdot t^{n-1}$$

φ' is continuously differentiable.
ie, φ' - cont. on $(-r, r)$.

$\therefore \varphi''$ is continuously diff.

Carrying on: φ is infinitely differentiable.

Note: if $\tilde{\varphi}(t) = \varphi(t) \quad \forall t \in (-r_0, r_0)$ ($r_0 = \min.$ of the radius of convergence of φ & $\tilde{\varphi}$.)

$$\sum_{n=0}^{\infty} b_n \cdot t^n = \sum_{n=0}^{\infty} a_n \cdot t^n$$

then, $b_n = a_n$.

ie, 2 functions "agreeing" can't give rise to 2 different power series.
The 2 power series must be the same.

Proof:
If $\varphi(t) = \sum_{n=0}^{\infty} a_n t^n$ has a

pos. radius of convergence, then.

$$a_n = \frac{\varphi^{(n)}(0)}{n!}$$

$\therefore \varphi$ & $\tilde{\varphi}$ "agree" on a non-degenerate open interval $(-r_0, r_0)$ around 0, their n^{th} derivatives must "agree" as well.

Hence, $\varphi = \tilde{\varphi}$ on $(-r_0, r_0)$

$$\Rightarrow \varphi^{(n)}(0) = \tilde{\varphi}^{(n)}(0)$$

$$\Rightarrow \frac{\varphi^{(n)}(0)}{n!} = \frac{\tilde{\varphi}^{(n)}(0)}{n!}$$

$$\Rightarrow a_n = b_n. \quad \square$$

X is a random variable taking values in $\{0, 1, 2, \dots\}$

Defⁿ: The probability generating function X is defined to be

(PGF)

$$\varphi_X(t) = \sum_{n=0}^{\infty} p_n \cdot t^n, \quad \text{where}$$

$$p_n = P(X=n)$$

$$|\varphi_X(t)| \leq \sum_{n=0}^{\infty} p_n |t|^n \leq 1 \quad \text{for } |t| \leq 1$$

• $\varphi_X(\cdot)$ determines the pmf $\{p_n\}$ of X .

$$\text{Indeed, } p_n = \frac{\varphi_X^{(n)}(0)}{n!}$$

Note: $\varphi_X(t) = \sum_{n=0}^{\infty} P(X=n) \cdot t^n = E(t^X)$.

Eg. $X \sim \text{Poi}(\lambda)$

$$\varphi_X(t) = \sum_{n=0}^{\infty} e^{-\lambda} \cdot \frac{\lambda^n}{n!} \cdot t^n = e^{-\lambda} \sum_{n=0}^{\infty} \frac{(\lambda t)^n}{n!} = e^{-\lambda} \cdot e^{\lambda t}$$

(Taylor series)

$$\Rightarrow \varphi_X(t) = e^{-\lambda(1-t)}$$

\rightarrow PGF determines distribution.

(i.e., 2 different distributions cannot have the same PGF.)

Eg.

$$X \sim \text{Poi}(\lambda_1)$$

X, Y - ind.

$$Y \sim \text{Poi}(\lambda_2)$$

$$\therefore \varphi_{X+Y}(t) = E(t^{X+Y})$$

$$= E(t^X \cdot t^Y)$$

i.e., PGF is a "characteristic" of any distⁿ.)

$$= E(t^X) \cdot E(t^Y)$$

$$= e^{-(\lambda_1 + \lambda_2) \cdot (1-t)}$$

$\therefore X, Y$ - ind.

$$E(t^X \cdot t^Y) = E(t^X) \cdot E(t^Y)$$

\hookrightarrow PGF of $\text{Poi}(\lambda_1 + \lambda_2)$.

$$\left. \begin{array}{l} X \sim \text{Poi}(\lambda_1) \\ Y \sim \text{Poi}(\lambda_2) \end{array} \right\} \Rightarrow X+Y \sim \text{Poi}(\lambda_1 + \lambda_2).$$

$\varphi_X(t)$ has radius of convergence ≥ 1

$\varphi_X(t)$ is differentiable of any order on $(-1, 1)$.

$$\varphi_X(t) = \sum_{n=0}^{\infty} p_n \cdot t^n.$$

$$\varphi_X'(t) = \sum_{n=1}^{\infty} n \cdot p_n \cdot t^{n-1}, \quad t \in (-1, 1).$$

$$\lim_{t \rightarrow 1} \varphi_X'(t) \text{ exists, \& } \lim_{t \rightarrow 1} \varphi_X'(t) = \sum n p_n = E(X).$$

$$\therefore \varphi_X''(t) = \sum_{n=2}^{\infty} n(n-1) \cdot p_n \cdot t^{n-2}$$

$$\lim_{t \rightarrow 1} \varphi_X''(t) = \sum n(n-1) p_n = E(X(X-1)).$$

Offspring distribution: $\{p_0, p_1, \dots\} = \pi$

$X_0 \equiv 1 \leftarrow 1^{\text{st}} \text{ parent.}$

$X_1 = \# \text{ offspring produced by the ancestor} \sim \pi$

If $X_1 \equiv 0$, then $X_2 \equiv 0$. (extinct).

If $X_1 = k > 0$, then, $X_2 \sim Y_1 + \dots + Y_k$,

If $X_1 = k > 0$, then, $X_2 \sim Y_1 + \dots + Y_k$,
 where $Y_1, \dots, Y_k \sim \pi$.

If $X_n = 0$, then $X_{n+1} = 0$ (extinct).

If $X_n = k > 0$, then $X_{n+1} = Y_1 + \dots + Y_k$,
 $Y_1, \dots, Y_k \sim \pi$.

probability of extinction:

$$q = P\left(\bigcup_n \{X_n = 0\}\right)$$

$$= \lim_{n \rightarrow \infty} P(X_n = 0)$$

(Working with just the distributions seems difficult)
 let's work with the PGFs.

$\therefore \varphi_n = \text{PGF of } X_n$

$$\varphi_1(t) = \varphi(t)$$

$$\varphi_2(t) = E(t^{X_2})$$

$$= \sum_k E(t^{X_2} | X_1 = k) \cdot P(X_1 = k)$$

$$= \sum_k (\varphi(t))^k \cdot P(X_1 = k) \leftarrow \text{expression for expectation}$$

$$= \varphi(\varphi(t))$$

$$\varphi_n(t) = E(t^{X_n}) = \varphi(\underbrace{\varphi \dots \varphi(t)}_{n\text{-fold composition}})$$

$$\begin{aligned} X_1 = k &\Rightarrow \\ X_2 &= Y_1 + \dots + Y_k \\ \therefore E(t^{X_2} | X_1 = k) &= E(t^{Y_1 + \dots + Y_k} | X_1 = k) \\ &= \prod_{i=1}^k E(t^{Y_i} | X_1 = k) \end{aligned}$$

$$\therefore \text{extinction probability, } q = P\left(\bigcup_n \{X_n = 0\}\right) = \lim_{n \rightarrow \infty} P(X_n = 0) = \lim_{n \rightarrow \infty} \varphi_n(0)$$

$$\varphi_n(t) = \varphi(\varphi_{n-1}(t))$$

$$\varphi_n(t) = \varphi(\varphi_{n-1}(t))$$

$$t=0 : \varphi_n(0) = \varphi(\varphi_{n-1}(0))$$

$$\therefore \lim_{n \rightarrow \infty} \varphi_n(0) = \lim_{n \rightarrow \infty} \varphi(\varphi_{n-1}(0))$$

$$\Rightarrow \boxed{q = \varphi(q)}$$

Food for thought: can there be other solutions?