

Probability theory I

24/12/21

Fact: (Fundamental theorem of Calculus)

If $F: [a, b] \rightarrow \mathbb{R}$ is continuously differentiable, then

$$\int_a^b F'(x) dx = F(b) - F(a).$$

Note: The continuity of F' is not necessary. For example

$$\text{Let } F(x) = \begin{cases} x^2 \sin\left(\frac{1}{x}\right) & \text{if } x \in (0, 1] \\ 0 & \text{if } x = 0 \end{cases}$$

$$\lim_{x \rightarrow 0^+} \frac{F(x) - F(0)}{x - 0} \text{ exists.}$$

H.w.

Then F is differentiable on $[0,1]$.

$$F'(x) = \begin{cases} 2x \sin\left(\frac{1}{x}\right) - \cos\left(\frac{1}{x}\right) & \text{if } x \in (0,1] \\ 0 & \text{if } x=0 \end{cases}$$

Here F' is not continuous at $x=0$. (HW). But F' is integrable. We shall have

$$\int_0^1 F'(x) dx = F(1) - F(0). \quad \square$$

Definition (Probability generating function):

Let X be a discrete random variable taking values $0, 1, 2, \dots$. The probability generating function (PGF) of X is a function $f: [-1, 1] \rightarrow \mathbb{R}$ defined by

$$f(s) = \sum_{n=0}^{\infty} P(X=n) s^n, \quad s \in [-1, 1].$$

Note: The PGF of X can also be represented as

$$f(s) = E(s^X) = \sum_{n=0}^{\infty} s^n P(X=n) = \sum_{n=0}^{\infty} a_n s^n$$

$|f(s)|$

Theorem: The pgff of X is continuous on $[-1, 1]$ and infinitely differentiable on $(-1, 1)$. Furthermore,

$$\mathbb{P}(X=n) = \frac{1}{n!} D^n f(0) = \frac{1}{n!} \left. \frac{d^n}{ds^n} (f(s)) \right|_{s=0} \quad \text{for all } n \geq 0 \text{ with}$$

$$D^0 f \equiv f.$$

Proof: Denote $p_n := \mathbb{P}(X=n)$, for all $n \geq 0$.

$$\text{Since, } \limsup_{n \rightarrow \infty} p_n^{1/n} \leq 1,$$

It follows that the radius of convergence of the power series $\sum_{n=0}^{\infty} p_n s^n$

is at least 1. Hence it follows that f is infinitely differentiable in $(-1, 1)$. Furthermore for a fixed $n \geq 0$,

$$D^n f(s) = \sum_{k=n}^{\infty} k(k-1) \cdots (k-n+1) p_k s^{k-n}, \quad |s| < 1.$$

putting $s=0$, we get

$$D^n f(0) = n! p_n.$$

HW: Show that f is continuous on $s=+1, s=-1$.

$$f(1) = \lim_{s \rightarrow 1^-} f(s) = \lim_{s \rightarrow 1^-} \sum_{n=0}^{\infty} p_n s^n = \sum_{n=0}^{\infty} \lim_{s \rightarrow 1^-} (p_n s^n) = 1$$

if $\limsup a_n^{1/n} \leq R$
then $\sum a_n s^n$ is
convergent on
 $(-\frac{1}{R}, \frac{1}{R})$. (check)!

f is unif cont
on (a, b)

$$\& f(a) = \lim_{s \rightarrow a} f(s)$$

$$f(b) = \lim_{s \rightarrow b} f(s)$$

f unif. cont
on $[a, b]$.

Corollary: If X and Y are random variables taking values $0, 1, 2, \dots$ and having the same PGF, then

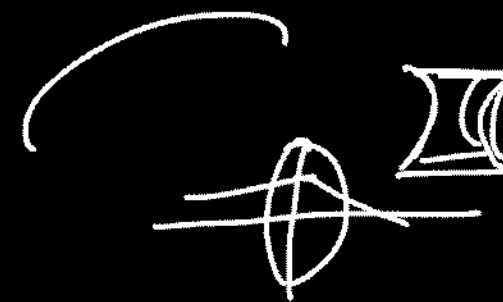
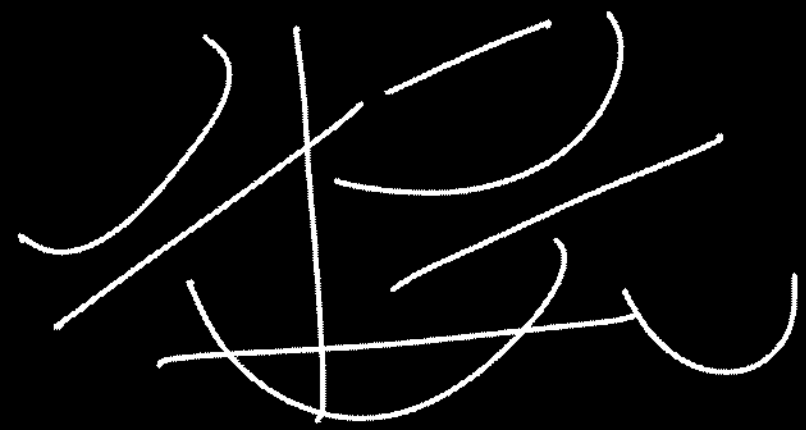
$$X \stackrel{d}{=} Y \quad (\Rightarrow) \quad P(X=n) = P(Y=n) \quad \forall n \geq 0.$$

pf: $P(X=n) = D^n f(0) \cdot \frac{1}{n!} = P(Y=n) \quad \text{for all } n \geq 0.$

Definition: (convex function):

A function f is convex on an interval $[a, b]$ if for any two points x_1 and x_2 in $[a, b]$ and any $0 < \lambda < 1$,

$$f(\lambda x_1 + (1-\lambda)x_2) \leq \lambda f(x_1) + (1-\lambda)f(x_2).$$



Theorem: If f is a PGF of X , then $f(1)=1$ and f is convex on $[0,1]$. The convexity is strict if

and only if $P(X \leq 1) < 1$. (or $P(X > 1) > 0$). [$\because P(X \leq 1) + P(X > 1) = 1$]

Proof: We have $f(s) = \sum_{n=0}^{\infty} p_n s^n$
 So, $f'(s) = \sum_{n=1}^{\infty} n p_n s^{n-1}$

$$\begin{aligned} 1 &= P(X \leq 1) + P(X > 1) \\ 1 &< 1 + P(X > 1) \\ 0 &< P(X > 1) \end{aligned}$$

and $f''(s) = \sum_{n=2}^{\infty} n(n-1) p_n s^{n-2} \geq 0 \quad \forall s \geq 0.$

$f''(s) > 0$ if $p_0 + p_1 < 1$.
 $\parallel$
 $P(X \leq 1)$ $\square$

$$\sum_{n=2}^{\infty} n(n-1) p_n s^{n-2} > 0 \iff p_0 + p_1 < 1$$

$$\iff$$

Let us compute p.f.'s of some random variables.

Example: ① Let $X \sim \text{Binomial}(n, p)$.

We know that if $X_1, X_2, \dots, X_n$ are iid $\text{Bern}(p)$. Then $X_1 + X_2 + \dots + X_n \stackrel{d}{=} X$. (H.W.).

The p.f. of X is

$$\begin{aligned} f_X(x) = E(s^X) &= E(s^{X_1 + X_2 + \dots + X_n}) = E(s^{X_1} s^{X_2} \dots s^{X_n}) \\ &= \left(E(s^{X_i})\right)^n \quad \left(\text{since } X_1, X_2, \dots, X_n \text{ are iid}\right) \\ &= (ps + (1-p))^n. \end{aligned}$$

② Let $X \sim \text{Poisson}(\lambda)$. Then

$$\begin{aligned} f_X(s) = E(s^X) &= \sum_{n=0}^{\infty} s^n \frac{\lambda^n}{n!} e^{-\lambda} = e^{-\lambda} \sum_{n=0}^{\infty} \frac{(\lambda s)^n}{n!} \\ &= e^{-\lambda + \lambda s} = e^{\lambda(s-1)} \end{aligned}$$

③ Let $X \sim \text{Geometric}(p)$. Then,

$$\begin{aligned} f_X(s) = E(s^X) &= \sum_{n=1}^{\infty} s^n q^{n-1} p = ps \sum_{n=1}^{\infty} (qs)^{n-1} \\ &= ps \frac{1}{1-qs} = \frac{ps}{1-qs} \end{aligned}$$

where $q = 1-p$.

$$\begin{aligned} &2 \sum_{i=1}^n \sum_{\substack{j=1 \\ j \neq i \\ j \neq i+1}}^n \text{Cov}(X_i, X_j) \\ &\quad \sum_{i=1}^n \text{Var}(X_i) \\ &2 \sum_{1 \leq i < j \leq n} \text{Cov}(X_i, X_j) \\ &\quad \sum_{i=1}^n \sum_{j=1}^n \text{Cov}(X_i, X_j) \end{aligned}$$

Problem: (already done): Suppose X and Y are independent, $X \sim \text{poi}(\lambda)$ and $Y \sim \text{poi}(\mu)$. Find the distribution of $X+Y$.

Solution: $f_{X+Y}(s) = E(s^{X+Y}) = E(s^X) E(s^Y) \quad (\because X \& Y \text{ are independent})$

$$= e^{\lambda(s-1)} e^{\mu(s-1)}$$

$$= e^{(\lambda+\mu)(s-1)} = f_{\text{poi}(\lambda+\mu)}(s)$$

Theorem! Let X be a discrete random variable with PGF $f_X(s) (\equiv f(s))$.

Then,

$$1) E(X) = f'(1)$$

$$2) E(\underbrace{X(X-1)\cdots(X-k+1)}_{k\text{th factorial moment}}) = f^{(k)}(1) = \left. \frac{d^k}{ds^k} f(s) \right|_{s=1}$$

k th factorial moment

$E(X^k)$ is
called the
 k th moment
of X

Proof:

$$1) f(s) = \sum_{h=0}^{\infty} s^h P(X=h)$$

diff. w.r.t. s .

$$f'(s) = \sum_{h=1}^{\infty} h s^{h-1} P(X=h)$$

$$\therefore f'(1) = \sum_{n=1}^{\infty} n P(X=n) = E(X).$$

$$2) f^{(k)}(s) = \sum_{n=k}^{\infty} n(n-1) \dots (n-k+1) s^{n-k} P(X=n)$$

$$\therefore f^{(k)}(1) = E(X(X-1) \dots (X-k+1)) \quad \square$$

Theorem: Suppose that $X_1, X_2, \dots, X_n$ are independent random variables and

$Y = X_1 + X_2 + \dots + X_n$. Then

$$f_Y(s) = \prod_{i=1}^n f_{X_i}(s) = f_{X_1}(s) f_{X_2}(s) \dots f_{X_n}(s)$$

Proof: $f_Y(s) = E(s^Y) = E(s^{X_1 + X_2 + \dots + X_n}) = E(s^{X_1} s^{X_2} \dots s^{X_n})$

$$= E(s^{X_1}) E(s^{X_2}) \dots E(s^{X_n}) \quad (\text{by independence})$$

$$= f_{X_1}(s) f_{X_2}(s) \dots f_{X_n}(s)$$

Proof:
Convergence: $\varepsilon > 0, \exists n \in \mathbb{N}$

$$|f_n(x) - f(x)| < \varepsilon \quad \forall n \geq n_0.$$

unif. conv.: $\varepsilon > 0, \exists n \in \mathbb{N}$

$$|f_n(x) - f(x)| < \varepsilon \quad \forall n \geq n_0 \text{ \& } x \in D.$$

$$f_n: D \rightarrow \mathbb{R}$$

Recall:

Fact: If two power series agree on an interval containing 0, however small, then all terms of the two series are equal.

Corollary: If two random variables have the same PGF in some interval containing 0, then the two random variables have the same distribution. So, every PGF corresponds to a

unique distribution.

$$\sum a_n x^n \rightarrow R_1$$
$$\sum b_n x^n \rightarrow R_2$$

~~$$\left(\frac{1}{x} \right)$$~~
$$-R^{-\epsilon} \leq R$$