

Lecture Notes 11

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0.1 Convergence of m.g.f

Example: Consider $f_{X_n}(x) = \binom{n}{x} P_n^x (1 - P_n)^{n-x} \mathbb{1}_{\{0,1,2,\dots,n\}}(x)$, the corresponding m.g.f $M_{X_n}(t) = (P_n e^t + (1 - P_n))^n$. As $n \rightarrow \infty$, $nP_n \rightarrow \lambda$, we have $P_n = \frac{\lambda}{n}(1 + O(1))$ and

$$M_{X_n}(t) \rightarrow M_X(t) = e^{\lambda(e^t - 1)}$$

which is the m.g.f of Poisson distribution.

Proof: Let $y = (P_n e^t + (1 - P_n))^n$ we have $\ln y = \frac{\ln(P_n e^t + (1 - P_n))}{\frac{1}{n}} = \frac{\ln(\frac{\lambda}{n} e^t (1 + O(1)) + (1 - \frac{\lambda}{n} (1 + O(1))))}{\frac{1}{n}}$ applying *L'hôpital's* rule we get

$$\lim_{n \rightarrow \infty} \ln y = \lambda(e^t - 1)$$

so

$$M_{X_n}(t) \rightarrow M_X(t) = e^{\lambda(e^t - 1)} \text{ as } n \rightarrow \infty \text{ } nP_n \rightarrow \lambda$$

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Intuition (From the basic view)

The example construct a relationship between Poisson distribution and Binomial distribution. And it is quite reasonable from the definition of Poisson distribution.

Recall the definition of Poisson:

- It's a **counting process**. That is, $N(t)$ that counts the number of appearances before time t .
- **(Boundary condition)** $N(0) = 0$
- **(Stationary)** $\forall t_1 < t_2$, $N(t_2) - N(t_1) \sim N(t_2 - t_1)$
- **(Independence)** $\forall t_1 < t_2 < t_3 < t_4$, $N(t_4) - N(t_3) \sim N(t_2) - N(t_1)$
- **(Fixed frequency)** $\lim_{\Delta \rightarrow 0^+} \frac{Pr[N(\Delta) - N(0) = 1]}{\Delta} = \lambda$, and $\lim_{\Delta \rightarrow 0^+} \frac{Pr[N(\Delta) - N(0) > 1]}{\Delta} = 0$
- **(Density function)** $f_\lambda(t, k) = \frac{(\lambda t)^k e^{-\lambda t}}{k!} \mathbf{1}_{\{k=0,1,2,\dots\}}$

Since $\lim_{\Delta \rightarrow 0^+} \frac{Pr[N(\Delta) = 1]}{\Delta} = \lambda$ simply let $\Delta\lambda$ be the success probability of P_n .

0.2 Basic property of characteristic function

Property 1 (relation to moment) *Let X be a random variable. If $E[|X^n|] < \infty$, then $\frac{d^n}{(dt)^n}\phi_X(t)$ exists for all t and*

$$\frac{d^n}{(dt)^n}\phi_X(t) = E[e^{itX}(iX)^n]$$

so the lower moments are

$$E[X^n] = (-i)^n \frac{d^n}{(dt)^n}\phi_X(0)$$

Property 2 (Basic) *Let X and Y be random variables.*

1. $\phi_X(0) = 1$ and $|\phi_X(t)| \leq 1 \ \forall t$.
2. $\phi_{-X}(t) = \overline{\phi_X(t)}$ where bar denotes complex conjugation.
3. $\phi_{aX+b}(t) = e^{itb}\phi_X(at)$.
4. If X and Y are independent, $\phi_{X+Y}(t) = \phi_X(t) \times \phi_Y(t)$.