

Moment Generating Function (MGF) of a random variable X is defined as :

$$M_X(s) = E[e^{xs}]$$

$$= \sum_{x \in \mathbb{Z}_0} e^{xs} P_X(x)$$

Similar to z transform of PDF by setting $z = e^s$.

for X discrete

Laplace transform of pdf.

$$\{ = \int_{x \in \mathbb{R}} e^{xs} f_X(x) dx. \text{ for } X \text{ continuous}$$

May not well-defined at all values of s .

Characteristic function:

$$\phi_X(t) = E[e^{jxt}] \quad \{ \text{similar to Fourier transform.}$$

Example.

① $X \sim \text{Bernoulli}(p)$

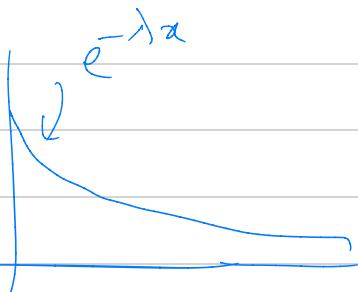
$$\begin{aligned} M_X(s) &= E[e^{xs}] \\ &= (1-p)e^{0 \times s} + p e^{1 \times s} \\ &= (1-p) + pe^s. \end{aligned}$$

② $X \sim \text{Poisson}(\lambda)$.

$$\begin{aligned} M_X(s) &= \sum_{k=0}^{\infty} e^{\lambda} \frac{\lambda^k}{k!} e^{ks} \\ &= \sum_{k=0}^{\infty} e^{\lambda} \frac{(\lambda e^s)^k}{k!} \\ &= e^{-\lambda} e^{\lambda e^s} = e^{\lambda(e^s - 1)} \end{aligned}$$

③ $X \sim \text{Exponential}(\lambda)$

$$M_X(s) = E[e^{Xs}] = \int_0^\infty e^{xs} \underbrace{\lambda e^{-\lambda x}}_{f_X(x)} dx.$$



$$= \int_0^\infty \lambda e^{-x(\lambda-s)} dx$$

$$= \begin{cases} \frac{\lambda}{\lambda-s} & s < \lambda \\ \text{not well defined} & \text{otherwise.} \end{cases}$$

④ $X \sim \text{Normal}(0, 1)$

X : standard normal R.V.

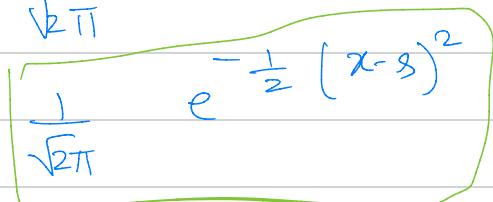
$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}.$$

$$M_X(s) = E[e^{Xs}] = \int_{-\infty}^\infty e^{xs} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx.$$

$$= \int_{-\infty}^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x^2 - 2xs + s^2)} e^{\frac{s^2}{2}} dx$$

$$= \int_{-\infty}^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-s)^2} e^{\frac{s^2}{2}} dx$$

$$= e^{s^2/2}$$



Properties

① Shift and Scaling : $Y = aX + b$.

$$M_Y(s) = E[e^{Ys}] = E[e^{(ax+b)s}]$$

$$= E[e^{bs} e^{asX}]$$

$$= e^{bs} E[e^{asX}] = e^{bs} M_X(as)$$

Example: $X \sim N(0, 1)$, $Y \sim N(\mu, \sigma^2)$. $b = \mu$, $a = 0$

$$Y = \sigma X + \mu.$$

$$\begin{aligned} M_Y(s) &= e^{bs} M_X(as) \\ &= e^{\mu s} M_X(\sigma s) = e^{\mu s + \sigma^2 s^2/2}. \end{aligned}$$

② Sum of Independent R.Vs:

$$M_Y(s) = E[e^{Ys}] = E[e^{(X_1 + X_2)s}]$$

$$\begin{aligned} &= E[e^{X_1 s} e^{X_2 s}] = E[e^{X_1 s}] E[e^{X_2 s}] \\ &= M_{X_1}(s) M_{X_2}(s). \end{aligned}$$

$$\text{If } Y = X_1 + X_2 - \dots + X_n, \quad M_Y(s) = \prod_{i=1}^n M_{X_i}(s)$$

Examples: Normal, Poisson, Binomial.

Normal case: $X_1 \sim N(\mu_1, \sigma_1^2)$, $X_2 \sim N(\mu_2, \sigma_2^2)$, $Y = X_1 + X_2$

$$\begin{aligned} M_Y(s) &= M_{X_1}(s) M_{X_2}(s) \\ &= e^{\mu_1 s + \sigma_1^2 s^2/2} e^{\mu_2 s + \sigma_2^2 s^2/2} \\ &= e^{(\mu_1 + \mu_2)s + (\sigma_1^2 + \sigma_2^2)s^2/2}. \end{aligned}$$

$\Rightarrow Y$ is a $N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$.

(We have used the inversion property of MGF).

Binomial: $X_1 \sim \text{Binomial}(n, p)$, $X_2 \sim \text{Binomial}(m, p)$.

$$Y = X_1 + X_2$$

$$X_1 = X_{11} + \dots + X_{1n}$$

where X_{11}, \dots, X_{1n} are Bernoulli(p) R.Vs.

Exercise:

$$M_{X_1}(s) = \prod_{i=1}^n M_{X_{1i}}(s)$$

$$= (1-p + pe^s)^n$$

Sum of two Poisson R.Vs with parameters λ_1, λ_2 respectively

is Poisson

with parameter $\lambda_1 + \lambda_2$

$$M_{X_2}(s) = (1-p + pe^s)^m$$

$$M_Y(s) = M_{X_1}(s) M_{X_2}(s) = (1-p + pe^s)^{m+n}$$

$\Rightarrow Y \sim \text{Binomial}(m+n, p)$.

③ Finding moments from MGF.

(Inversion property).

if X is
discrete
& nonnegative

$$\text{④ } P(X=0) = \lim_{s \rightarrow -\infty} M_X(s)$$

$$\text{⑤ } M_X(0) = 1 \quad (\text{Exercise})$$

$$\text{⑥ } M_X(s) = E[e^{xs}]$$

$$= E\left[1 + \sum_{n=1}^{\infty} \frac{(xs)^n}{n!}\right]$$

$$= 1 + \sum_{n=1}^{\infty} E[x^n] \left(\frac{s^n}{n!}\right).$$

$$\frac{\partial M_X(s)}{\partial s} =$$

$$E[X] + \frac{2xs}{2!} E[X^2]$$

Moments of RV

are $E[X], E[X^2], \dots, E[X^n]$

$E[X^n]$ is n -th moment
of R.V X .

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$$

$$f(x) = f(0) + f'(0)x + f''(0) \frac{x^2}{2!} + \dots \rightarrow \text{①}$$

$$\left. \frac{\partial M_X(s)}{\partial s} \right|_{s=0} = E[X]$$

$$\left. \frac{\partial^2 M_X(s)}{\partial s^2} \right|_{s=0} = E[X^2] + \frac{3x^2 + 3xs}{3!} E[X^3] + \dots$$

$$\left. \frac{\partial^n M_X(s)}{\partial s^n} \right|_{s=0} = E[X^n]. \quad \text{In general,}$$

$$\left. \frac{\partial^n M_X(s)}{\partial s^n} \right|_{s=0} = E[X^n].$$

④ MGF of sum of random # of R.Vs.

Let X_1, X_2, \dots be iid R.Vs with
MGF given by $M_X(s)$ and N be R.V that is
independent of X_i 's.

$$Y = X_1 + X_2 + \dots + X_N.$$

$$M_Y(s) = E[e^{Ys}]$$

$$= E[e^{(X_1 + \dots + X_N)s}]$$

$$\begin{aligned}
 & \text{total law of expectation} \leftarrow = \sum_n P(N=n) E[e^{(X_1 + \dots + X_N)s} \mid N=n] \\
 & = \sum_n P(N=n) E[e^{(X_1 + \dots + X_n)s} \mid N=n] \\
 & \text{N is independent of } X_1, \dots, X_n \dots = \sum_n P(N=n) E[e^{(X_1 + \dots + X_n)s}] \\
 & = \sum_n P(N=n) (M_X(s))^n = E[M_X(s)^N] \\
 & = E[e^{N \log M_X(s)}] \\
 & = M_N(\log M_X(s)).
 \end{aligned}$$

Example: $N \sim \text{Poisson}(\lambda)$, $X_i \sim \text{Bernoulli}(p)$. with MGF $M_X(s)$

$$M_X(s) = (1-p+pe^s)$$

$$Y = X_1 + X_2 + \dots + X_N$$

$$\begin{aligned}
 M_Y(s) &= e^{\lambda(e^s - 1)} M_X(s) = M_N(\log M_X(s)). \\
 &= e^{\lambda(e^{\log M_X(s)} - 1)}. \\
 &= e^{\lambda(M_X(s) - 1)}. \\
 &= e^{\lambda(\lambda - p + pe^s - 1)} = e^{\lambda p(e^s - 1)}
 \end{aligned}$$

Y is Poisson(λp) (by inversion property)

(5) MGF of "mixture" distributions : R.Vs X_1, \dots, X_n . and $Y = X_i$ with probability p_i .

$$f_Y(y) = \sum_{i=1}^n p_i f_{X_i}(y).$$

$$M_Y(s) = E[e^{Ts}]$$

$$= \int_{-\infty}^{\infty} e^{ys} f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} e^{ys} \sum_{i=1}^n p_i f_{X_i}(y) dy.$$

$$= \sum_{i=1}^n p_i \int_{-\infty}^{\infty} e^{ys} f_{X_i}(y) dy$$

$$M_Y(s) = \sum_{i=1}^n p_i M_{X_i}(s)$$

Example: IITH's SBI bank has 3 employees, two of them are too slow, one luckily fast. Time to assist customers is exponentially distributed with parameter at $\lambda = 6$ for fast ones and $\lambda = 2$ for slow ones. A person visits the bank and picks an employee at random with prob. $\frac{1}{3}$. What is the pdf of time it takes to assist this person & what is the MGF? (M_Y).

$$X_1 \sim \exp(6)$$

$$f_Y(y) = \frac{1}{3} f_{X_1}(y) + \frac{2}{3} f_{X_2}(y). \quad X_2 \sim \exp(2)$$

$$= \frac{1}{3} \times 6 \times e^{-6y} + \frac{2}{3} \times 2 \times e^{-2y} \quad P_1 = \frac{1}{3}, \quad P_2 = \frac{2}{3}$$

$$M_Y(s) = \frac{1}{3} \left(\frac{6}{6-s} \right) + \frac{2}{3} \left(\frac{2}{2-s} \right).$$

⑥

Inversion property

The transform $M_X(s)$ uniquely identifies the probability law of R.V X (PMF if X is discrete or p.d.f if X is continuous R.V) - In particular if $M_X(s)$ and $M_Y(s)$ are the same $\forall s$, then X and Y are identically distributed