2.3 Moment Generating Function

$$E[X]$$
 1st moment
What is a Moment? $E[X^2]$ 2nd moment

The r^{th} moment of a random variable, X, is $E[X^r]$.

Also called: moment about the origin, raw moment

The r^{th} central moment of a random variable, X, is the expected value of the rth power of the deviation of a random variable from its mean: $E[(X - \mu_X)^r]$

Moments

Moment	Real World	Statistics	
0th	Total mass	Total Probability	7
1st	Center of Mass	Expected Value	×
2nd _	Rotational Inertia (torque required for desired angular acceleration)	Variance (2 nd central moment)	7
3rd		Skewness (3 rd standardized moment)	Ī (X
4th		Kurtosis (4 th standardized moment)	

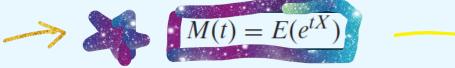


Definition 2.3-1

Let X be a random variable of the discrete type with pmf f(x) and space S. If there is a positive number h such that

$$E(e^{tX}) = \sum_{x \in S} e^{tx} f(x)$$

exists and is finite for -h < t < h, then the function defined by



is called the **moment-generating function of** X (or of the distribution of X). This function is often abbreviated as mgf.

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

$$e^{4x} = 1 + 4x + \frac{(4x)^{2}}{2!} + \frac{(4x)^{3}}{3!} + \dots$$

$$M(t) = \overline{E}[e^{4x}] + \frac{(4x)^{2}}{3!} + \frac{(4x)^{3}}{3!} + \dots$$

$$M(t) = \overline{E}[x']$$

$$M'(t) = \overline{E}[0 + x + \frac{2(4x)x}{2!} + \frac{3(4x)^{2}(x)}{3!} + \dots]$$

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Moment Generating Function

Suppose the sample space of X is $S = \{x_1, x_2, x_3, ...\}$

What is an expression for mgf?

$$M(t) = E[e^{tx}] = \sum_{\chi} e^{t\chi} \cdot f(\chi) \in$$

$$M(t) = e^{tx_1} f(x_1) + e^{tx_2} f(x_2) + e^{tx_3} f(x_3) + \dots$$
The coefficient of each e^{tx_i} is the probability, $f(x_i) = P(X = xi)$

Simple mgf example

Example 2.3-5

If X has the mgf



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$$M(t) = e^t \left(\frac{3}{6}\right) + e^{2t} \left(\frac{2}{6}\right) + e^{3t} \left(\frac{1}{6}\right), \qquad -\infty < t < \infty,$$

then the support of X is $S = \{1, 2, 3\}$ and the associated probabilities are

$$P(X=1) = \frac{3}{6}$$
, $P(X=2) = \frac{2}{6}$, $P(X=3) = \frac{1}{6}$.

Or we could write $f(x) = \frac{4-x}{6}$, x = 1,2,3.

$$M(t) = E[e^{tt}]$$

$$M_{x+y}(t) = E[e^{t(x+y)}]$$

More Properties of MGFs

- 1. If two random variables have the same MGF, then they have the same distribution. i.e. if X and Y are random variables that have the same MGF: $M_X(t) = MY(t)$, then X and Y have the exact same distribution (pmf, cdf, etc)
- 2. For two independent random variables, X and Y, the MGF of their sum is the product of their MGFs:

$$M_{X+Y}(t) = M_X(t)MY(t)$$
 (works for more than 2 as well)

More Properties of MGFs



3. The nth derivative of $M_X(t)$ evaluated at t=0 is equal to the nth moment, $E[X^n]$.



Examples

Moment Generating Function

Let $X \sim Binom(n, p)$. The mgf is:

$$M_{x}(t) = E[e^{tX}] = \sum_{x=0}^{n} e^{tx} \binom{n}{x} p^{x} (1-p)^{n-x}$$

$$= \sum_{x=0}^{n} \binom{n}{x} (pe^{t})^{x} (1-p)^{n-x} = [pe^{t} + (1-p)]^{n}$$
Sinomial expansion

Computing the first two moments using the mgf:

$$M'(t) = n[pe^{t} + (1-p)]^{n-1}pe^{t}.$$

$$M''(t) = n(n-1)[pe^{t} + (1-p)]^{n-2}p^{2}e^{2t} + n[pe^{t} + (1-p)]^{n-1}pe^{t}.$$

$$E[X] = M'(0) = np$$

$$E[X^{2}] = M''(0) = n(n-1)p^{2} + np$$

$$Var[X] = E[X^{2}] - (E[X])^{2} = n(n-1)p^{2} + np - (np)2$$

$$= np - np^{2} = np(1-p)$$

Example: known distributions

Suppose a random variable X has moment generating

function: M(t) =
$$(\frac{2}{3} + \frac{1}{3}e^{t})_{10}$$

Binomial
$$f(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x},$$

 $b(n,p)$
 $0 $M(t) = (1-p+pe^t)^n, -\infty$$

What is the pmf of X?

What is the pmf of X?

$$X \sim \mathcal{B}_{inom} (n = 10, p = 1/3)$$

$$f(x) = {10 \choose x} {1 \over 3}^x {2 \choose 3}^{10-x}$$

Say we have 3 random variables, W, X, Y \sim Bernoulli(p) Let Z be the sum of all three: Z = W + X + Y. $\sim Binom (n = 3, P)$ Show that $Z \sim Binom(n,p)$

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Note: The mgf of a Bernoulli random variable is

$$M(t) = (1 - p + pe^t)$$

M(t) =
$$(1 - p + pe^t)$$
 M_{w+x+y} $(+) = M_{w}$ $(+) M_{y}$ $(+) M_{y}$

Q: Can you do the same for $\sum Geometric = NB$ at home?