CSL003P1M: Probability and Statistics Lecture 25 (Joint Moment Generating Functions)

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An Important Property of MGF

The moment generating function uniquely determines the distribution. That is, if $M_X(t)$ exists and is finite in some region about t = 0, then the distribution of X is uniquely determined.

For instance, if

$$M_X(t) = \left(\frac{1}{2}\right)^{10} (e^t + 1)^{10}$$

then it follows that X is a binomial random variable with parameters 10 and $\frac{1}{2}$. Note that the mgf of binomial random variable X is

$$M_X(t) = (pe^t + 1 - p)^n.$$



Suppose that the moment generating function of a random variable X is given by $M(t) = e^{3(e^t - 1)}$. What is $P\{X = 0\}$?

The mgf of the Poisson random variable X with parameter λ is

$$M_X(t) = e^{-\lambda} e^{\lambda e^t}$$

It is clear from $M_X(t)$ that X is a Poisson random variable with parameter $\lambda=3$. Thus,

$$P{X = 0} = e^{-3} \frac{3^0}{0!} = e^{-3}.$$



Prove that the sum of independent random variables equals the product of the individual moment generating functions.

- Suppose that X and Y are independent and have moment generating functions $M_X(t)$ and $M_Y(t)$ respectively.
- Then $M_{X+Y}(t)$, the moment generating function of X+Y is given by

$$M_{X+Y}(t) = E[e^{t(X+Y)}]$$

= $E[e^{tX}e^{tY}]$
= $E[e^{tX}]E[e^{tY}]$ (since X and Y are independent)
= $M_X(t)M_Y(t)$

MGF of the sum of a random number of random variables

Let X_1, X_2, \ldots be a sequence of independent and identically distributed random variables, and let N be a nonnegative integer. We want to compute the moment generating function of

$$Y = \sum_{i=1}^{N} X_i$$

It follows easily from the previous result that

$$M_Y(t) = [M_{X_1}(t)]^N$$

Joint Moment Generating Functions

For any n random variables X_1, \ldots, X_n , the joint moment generating function, $M(t_1, \ldots, t_n)$, is defined, for all real values of t_1, \ldots, t_n by

$$M(t_1,\ldots,t_n)=E[e^{t_1X_1+\cdots+t_nX_n}]$$

The individual moment generating functions can be obtained from $M(t_1, \ldots, t_n)$ by letting all but one of the t_i 's be 0. That is,

$$M_{X_i}(t) = E[e^{tX_i}] = M(0, \dots, 0, t, 0, \dots, 0)$$

where t is in the ith place.



Joint Moment Generating Functions

- It can be proven that the joint moment generating function $M(t_1, t_2, ..., t_n)$ uniquely determines the joint distribution of $X_1, ..., X_n$.
- This result can then be used to prove that the n random variables X_1, \ldots, X_n are independent if and only if

$$M(t_1,\ldots,t_n)=M_{X_1}(t_1)\cdots M_{X_n}(t_n)$$

Proof: If the n random variables are independent, then

$$M(t_1, \dots, t_n) = E[e^{(t_1X_1 + \dots + t_nX_n)}]$$

$$= E[e^{t_1X_1} \dots e^{t_nX_n}]$$

$$= E[e^{t_1X_1}] \dots E[e^{t_nX_n}]$$
 (by independence)
$$= M_{X_1}(t_1) \dots M_{X_n}(t_n)$$

For the other direction, try yourself. (Hint: use the fact that the joint moment generating function uniquely determines the joint distribution.)

If X and Y are independent binomial random variables with parameters (n, p) and (m, p), respectively, what is the distribution of X + Y?

The moment generating function of X + Y is given by

$$M_{X+Y}(t) = M_X(t)M_Y(t) = (pe^t + 1 - p)^n(pe^t + 1 - p)^m$$

= $(pe^t + 1 - p)^{m+n}$

However, $(pe^t + 1 - p)^{m+n}$ is the moment generating function of a binomial random variable having parameters m + n and p. Thus, this must be the distribution of X + Y.



If X and Y are independent Poisson random variables with parameters λ_1 and λ_2 , what is the distribution of X + Y?

The moment generating function of X + Y is given by

$$M_{X+Y}(t) = M_X(t)M_Y(t)$$

$$= e^{-\lambda_1}e^{\lambda_1e^t}e^{-\lambda_2}e^{\lambda_2e^t}$$

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

However, $e^{-(\lambda_1+\lambda_2)}e^{(\lambda_1+\lambda_2)e^t}$ is the moment generating function of a Poisson random variable having parameter $\lambda_1+\lambda_2$. Thus, this must be the distribution of X+Y.



Let X be a negative binomial random variable with parameters r and p. Find the mgf of X given that the mgf of a geometric random variable with parameter p is

$$\frac{pe^t}{1-(1-p)e^t}$$

Solution:

- For $1 \le i \le r$, let X_i be independent geometric random variables each with parameter p.
- Then

$$X = X_1 + X_2 + \cdots + X_r$$

Thus,

$$M_X(t) = [M_{X_1}(t)]^r = \left[\frac{pe^t}{1 - (1 - p)e^t}\right]^r$$

Thank You