

very useful fact. We will see examples of how we use it shortly. Right now let's state this fact more precisely as a theorem. We omit the proof here.

**Theorem 6.1** Consider two random variables  $X$  and  $Y$ . Suppose that there exists a positive constant  $c$  such that MGFs of  $X$  and  $Y$  are finite and identical for all values of  $s$  in  $[-c, c]$ . Then,

$$F_X(t) = F_Y(t), \text{ for all } t \in \mathbb{R}.$$

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### Example 6.7

For a random variable  $X$ , we know that

$$M_X(s) = \frac{2}{2-s}, \text{ for } s \in (-2, 2).$$

Find the distribution of  $X$ .

**Solution**

We note that the above MGF is the MGF of an exponential random variable with  $\lambda = 2$  (Example 6.5). Thus, we conclude that  $X \sim \text{Exponential}(2)$ .

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## Sum of Independent Random Variables:

Suppose  $X_1, X_2, \dots, X_n$  are  $n$  independent random variables, and the random variable  $Y$  is defined as

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

Then,

$$\begin{aligned} M_Y(s) &= E[e^{sY}] \\ &= E[e^{s(X_1 + X_2 + \dots + X_n)}] \\ &= E[e^{sX_1} e^{sX_2} \dots e^{sX_n}] \\ &= E[e^{sX_1}] E[e^{sX_2}] \dots E[e^{sX_n}] \quad (\text{since the } X_i\text{'s are independent}) \\ &= M_{X_1}(s) M_{X_2}(s) \dots M_{X_n}(s). \end{aligned}$$

If  $X_1, X_2, \dots, X_n$  are  $n$  independent random variables, then

$$M_{X_1+X_2+\dots+X_n}(s) = M_{X_1}(s)M_{X_2}(s)\cdots M_{X_n}(s).$$

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**Example 6.8**

If  $X \sim \text{Binomial}(n, p)$  find the MGF of  $X$ .

**Solution**

We can solve this question directly using the definition of MGF, but an easier way to solve it is to use the fact that a binomial random variable can be considered as the sum of  $n$  independent and identically distributed (i.i.d.) Bernoulli random variables. Thus, we can write

$$X = X_1 + X_2 + \cdots + X_n,$$

where  $X_i \sim \text{Bernoulli}(p)$ . Thus,

$$\begin{aligned} M_X(s) &= M_{X_1}(s)M_{X_2}(s)\cdots M_{X_n}(s) \\ &= (M_{X_1}(s))^n \quad (\text{since the } X_i\text{'s are i.i.d.}) \end{aligned}$$

Also,

$$M_{X_1}(s) = E[e^{sX_1}] = pe^s + 1 - p.$$

Thus, we conclude

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

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**Example 6.9**

Using MGFs prove that if  $X \sim \text{Binomial}(m, p)$  and  $Y \sim \text{Binomial}(n, p)$  are independent, then  $X + Y \sim \text{Binomial}(m + n, p)$ .

**Solution**

We have

$$\begin{aligned}M_X(s) &= (pe^s + 1 - p)^m, \\M_Y(s) &= (pe^s + 1 - p)^n.\end{aligned}$$

Since  $X$  and  $Y$  are independent, we conclude that

$$\begin{aligned}M_{X+Y}(s) &= M_X(s)M_Y(s) \\&= (pe^s + 1 - p)^{m+n},\end{aligned}$$

which is the MGF of a *Binomial*( $m + n, p$ ) random variable. Thus,

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

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