The moment generating function method_(Thm 6.1)

Recall that the moment generating function (mgf) of a random variable X is
$$m_X(t) = Ee^{Xt}.$$

$$= \begin{cases} Z e^{xt} f(x) & \text{if } X \text{ discrete} \\ Y & \text{if } X \end{cases}$$

Mgf's can be used to identify distributions as follows:

If the mgf of a rv X is the same as that of another rv U, we may conclude that X has the same distribution as U.

(Ie, if
$$m_X(t) = m_U(t)$$
, then $F_X(k) = F_U(k)$ and $f_X(k) = f_U(k)$ for all k .)

Let us now tackle the problem in Example 8. $(Z \sim N(0.1))$. Find the dsn of $X = Z^2$.)

$$m_{X}(t) = Ee^{Xt} = Ee^{Z^{2}t} = \int_{-\infty}^{\infty} e^{z^{2}t} \sqrt{\frac{1}{2\pi}} e^{-\frac{1}{2}z^{2}} dz$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^{2}(1-2t)} dz$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^{2}} dz, \quad \text{where } c^{2} = \frac{1}{1-2t}$$

$$= c. \quad \text{(The integral must equal 1.)}$$

Thus
$$m_X(t) = (1-2t)^{-1/2}$$
.

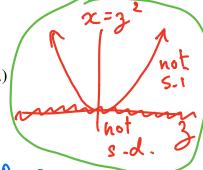
But
$$(1-2t)^{-1/2}$$
 is the mgf of $U \sim \text{Gam}(1/2,2)$.

(Recall that if $W \sim Gam(a,b)$ then $m(t) = (1-bt)^{-a}$.)

It follows that $X \sim \text{Gam}(1/2,2)$.

Equivalently, $X \sim \chi^2(1)$. (Recall that if $R \sim \text{Gam}(k/2,2)$ then $R \sim \chi^2(k)$.)

Therefore the pdf of *X* is $f(x) = \frac{x^{\frac{1}{2}-1}e^{-x/2}}{2^{1/2}\Gamma(1/2)} = \frac{1}{\sqrt{2\pi x}e^x}, x > 0$.



Another solution:

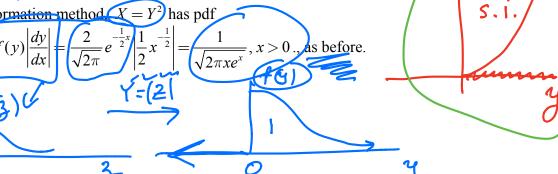
ile

Let
$$Y = |Z|$$
. Then $f(y) = \frac{(2)}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2}$, $y > 0$

(This follows by symmetry about z = 0. It can also be proved using the cdf method.) Now $x = y^2$ is a strictly increasing function, since y can't be negative.

So by the transformation method $X = Y^2$ has pdf

$$f(x) = f(y) \left| \frac{dy}{dx} \right| = \frac{2}{\sqrt{2\pi}} e^{-\frac{1}{2}x} \left| \frac{1}{2} x^{-\frac{1}{2}} \right| = \frac{1}{\sqrt{2\pi x} e^x}, x > 0$$
., as before.



Two useful results when applying the mgf technique

- 1. If X = a + bY, then $m_X(t) = e^{at} m_Y(bt)$. (Prove this as an exercise.)
- 2. If $Y_1, ..., Y_n$ are independent random variables and $X = Y_1 + ... + Y_n$, then $m_X(t) = m_{Y_1}(t) ... m_{Y_n}(t)$. (This is Thm 6.2.)

Example 9 $Y \sim N(0,1)$. Find the dsn of X = a + bY. (This is an earlier exercise.)

 $m_Y(t) = e^{\frac{1}{2}t^2}.$ (This is proved in Tutorial 7.)

Therefore $m_X(t) = e^{at}m_Y(bt) = e^{at}e^{\frac{1}{2}(bt)^2} = e^{at + \frac{1}{2}b^2t^2},$ which is the mgf of $U \sim N(a, b^2)$.

It follows that $X \sim N(a, b^2)$.

Example 10 Suppose that $Y_1, ..., Y_n$ are independent gamma rv's, such that the *i*th one has parameters a_i and b.

Find the distribution of $X = Y_1 + ... + Y_n$.

$$m_X(t) = m_{Y_1}(t) ... (m_{Y_n}(t))$$

= $(1-bt)^{-a_1} ... (1-bt)^{-a_n}$
= $(1-bt)^{-\dot{a}}$, where $\dot{a} = a_1 + ... + a_n$.

Hence $X \sim \text{Gam}(\dot{a}, b)$.

Corollary: If $Y_1, ..., Y_n \sim \text{iid } \chi^2(1)$, then $Y_1 + ... + Y_n \sim \chi^2(n)$. (NB: $\chi^2(r) = Gam(r/2,2)$.)

Exercise Suppose that $Y_1, ..., Y_n$ are independent normally distributed rv's such that the *i*th one has mear (a_i) and variance $(b_i)^2$

Let
$$X = \sum_{i=1}^{n} k_{i}Y_{i}$$
. Show that $X \sim N\left(\sum_{i=1}^{n} k_{i}a_{i}, \sum_{i=1}^{n} k_{i}^{2}b_{i}^{2}\right)$.
$$m_{X}(t) = Ee^{\left(\sum_{i=1}^{n} k_{i}Y_{i}\right)t} = E\prod_{i=1}^{n} e^{k_{i}Y_{i}t} = \prod_{i=1}^{n} Ee^{Y_{i}(k_{i}t)} = \prod_{i=1}^{n} m_{Y_{i}}(k_{i}t)$$

$$= \prod_{i=1}^{n} e^{a_{i}(k_{i}t) + \frac{1}{2}b_{i}^{2}(k_{i}t)^{2}} = e^{\left(\sum_{i=1}^{n} k_{i}a_{i}\right)t + \frac{1}{2}\left(\sum_{i=1}^{n} k_{i}^{2}b_{i}^{2}\right)t^{2}} \quad \text{(see Thm 6.3)}.$$

$$\times = 2Y + 1$$
, $\times = Y$, $\times = \frac{\sin Y}{1 - Y}$ etc.

STAT2001 CH06B Page 3 of 4

Order statistics

Suppose that $Y_1, ..., Y_n$ are iid rv's.

Let: (U_1) be the smallest of these U_2 be the second smallest

(ie, $U_1 = \min(Y_1, ..., Y_n)$)

 U_n be the largest

(Recall Problem 1 in Tutorial 6.)

Example 11 Suppose that $Y_1, Y_2 \sim \text{iid } Expo(b)$

Find the pdf of the second order statistic, $U_2 = \max(Y_1, Y_2)$.

 $F_{U_2}(u) = P(U_2 \le u) = P\{\max(Y_1, Y_2) \le u\} = P(Y_1 \le u, Y_2 \le u)$

 $= P(Y_1 < u)P(Y_2 < u)$ (by independence) = $P(Y_1 < u)^2$ = $(1 - e^{-u/b})^2$, u > 0.

So $f_{U_2}(u) = F'_{U_2}(u) = (2(1 - e^{-u/b})^{1}(-e^{-u/b})(-1/b)$ $= 2(1 - e^{-u/b}) \frac{1}{b} e^{-u/b}, \quad u > 0.$

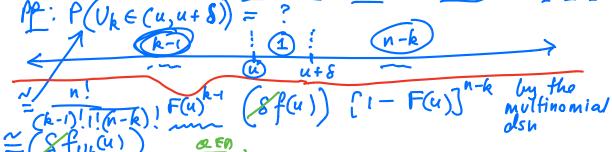
Exercise: Show that $EU_2 = 3b/2$ (NB: $EU_2 > EY_i = b$, as one would expect.)

 $(EU_2) = 2 \int_{a}^{\infty} u \frac{1}{b} e^{-u/b} du - \int_{a}^{\infty} u \frac{1}{b/2} e^{-u/(b/2)} du = 2b - b/2 = 3b/2.$

If $Y_1, ..., Y_n$ are continuous and iid, then the pdf of the kth order statistics U_k $f_{U_k}(u) = \underbrace{n!}_{(k-1)!(n-k)!} F(u)^{k-1} [1 - F(u)]^{n-k} f(u),$

where f(y) and F(y) are the pdf and cdf of Y_1 , respectively. (See Thm 6.5.)

Note that this formula is in agreement with $f_{U_2}(u)$ in Example 11, where n = k = 2.



Range restricted distributions

Example 12 Suppose that the number of accidents which occur each year at a certain intersection follows a Poisson distribution with mean λ .

Find the pdf of the number of accidents at this intersection last year if it is known that at least one accident occurred there during that year.

Let Y be the number of accidents at the intersection last year.

Then
$$X = (Y|Y>0)$$
 has pdf
$$\begin{aligned}
P(X) &= P(X=x) \\
&= P(Y=x|Y>0) \\
&= \frac{P(Y=x,Y>0)}{P(Y>0)} \\
&= \frac{P(Y=x)}{1-P(Y=0)} \quad \text{for } x>0 \\
&= \frac{e^{-\lambda}\lambda^x/x!}{1-e^{-\lambda}\lambda^2} \quad x = 1,2,3,...
\end{aligned}$$

For example, if $\lambda = 3.2$ then $p_X(4) = \frac{e^{-3.2} \cdot 3.2^4 / 4!}{1 - e^{-3.2}} = 0.186$, which we note is slightly higher than $p_Y(4) = e^{-3.2} \cdot 3.2^4 / 4! = 0.178$.

What is the expected number of accidents last year?

$$E(Y|Y>0) = EX = \sum_{x=1}^{\infty} x \frac{e^{-\lambda} \lambda^x / x!}{1 - e^{-\lambda}}$$

$$= \frac{1}{1 - e^{-\lambda}} \sum_{x=0}^{\infty} x \frac{e^{-\lambda} \lambda^x}{x!}$$
 (where the first term in the sum is zero)
$$= \frac{1}{1 - e^{-\lambda}},$$

which we note is higher that $EY = \lambda$.

For example, if $\lambda = 3.2$ then EX = 3.336 > 3.2 = EY.

Exis Find the poly of
$$(x+y)$$
 $x = (y+y)$

$$f(x) = (x+y)$$

$$f(y) = (y+y)$$

$$= \frac{e^{-\lambda} \times / x!}{|y-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|y-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|y-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|y-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{\lambda} - \lambda e^{-\lambda}|}, x = 2,3,...$$

$$= \frac{e^{-\lambda} \times / x!}{|x-e^{-\lambda} - \lambda e$$

$$F(x) = \begin{cases} 0, x < 0 \\ \overline{\Phi}(x), x \ge 0 \end{cases} - \underbrace{1}_{|L|} F(x)$$

$$\times \text{ has a } \text{ dsn}$$

$$f(x) = \begin{cases} 1/2, & x = 0 \text{ (discrete)} \\ \phi(x), & x > 0 \text{ (ets)} \end{cases}$$

$$0, & x < 0 \end{cases}$$

$$x < 0$$

Note:
$$\mathcal{L}(x) + \int f(x) dx$$

 $x \text{ discrete} \qquad x \text{ is cls}$
 $= \frac{1}{2} + \int \mathcal{B}(x) dx$
 $= \frac{1}{2} + \frac{1}{2} - 1$