

Review

- Conditional expectation $E[X|Y=y]$
- Calculate expectation by conditioning

$$E[X] = E[E[X|Y]]$$

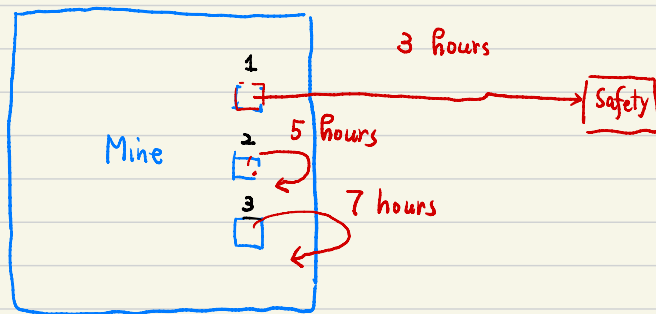
If Y is a discrete r.v., then we have

$$E[X] = \sum_y E[X|Y=y] P\{Y=y\}.$$

Below we give an example.

Example 1.

A miner is trapped in a mine containing 3 doors. The first door leads to a tunnel that will take him to safety after 3 hours of travel. The second door leads to a tunnel that will return him to the mine after 5 hours of travel. The third door leads to a tunnel that will return him to the mine after 7 hours. If we assume that the miner is at all times equally likely to choose any one of the doors, what is the expected length of time until he reaches safety?



Solution: Let X denote the length of time (in hours) until the miner reaches safety.

Let Y denote the door that he chooses in the first time.

By Prop 2,

$$\begin{aligned} E[X] &= E[E[X|Y]] \\ &= E[X|Y=1] \cdot P\{Y=1\} \\ &\quad + E[X|Y=2] \cdot P\{Y=2\} \end{aligned}$$

$$\begin{aligned}
 &+ E[X|Y=3] \cdot P\{Y=3\} \\
 &= \frac{1}{3} (E[X|Y=1] + E[X|Y=2] + E[X|Y=3]) \\
 &= \frac{1}{3} (3 + (5 + E[X]) + (7 + E[X])).
 \end{aligned}$$

Solving this equation, we obtain

$$E[X] = 3 + 5 + 7 = 15 \quad (\text{hours})$$

§ 7.7 Moment generating functions.

Def. Let X be a r.v. and $t \in \mathbb{R}$. Define

$$M_X(t) = E[e^{tX}].$$

For convenience, we also write $M(t) = M_X(t)$ and call it the moment generating function of X .

Remark:

$$\textcircled{1} \quad e^{tX} = \sum_{n=0}^{\infty} \frac{1}{n!} t^n \cdot X^n.$$

Hence

$$(i) \quad M_X(t) = \sum_{n=0}^{\infty} \frac{t^n}{n!} E[X^n]$$

↳ n-th moment of X .

② If $M_X(t)$ exists and is finite for all $-t_0 < t < t_0$ for some $t_0 > 0$,

then

$$(2) \quad E[X^n] = M_X^{(n)}(0), \quad n=1, 2, \dots$$

Example 2. Let X be a binomial r.v. with parameters (n, p) .

$$\begin{aligned} M_X(t) &= E[e^{tX}] = \sum_{k=0}^n e^{tk} P\{X=k\} \\ &= \sum_{k=0}^n e^{tk} \cdot \binom{n}{k} p^k (1-p)^{n-k} \\ &= \sum_{k=0}^n \binom{n}{k} (e^t p)^k (1-p)^{n-k} \end{aligned}$$

By Binomial Thm

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

Example 3. Let X be a Poisson r.v. with parameter λ .

$$\begin{aligned}M_X(t) &= E[e^{tX}] = \sum_{k=0}^{\infty} e^{tk} \cdot e^{-\lambda} \cdot \frac{\lambda^k}{k!} \\&= \sum_{k=0}^{\infty} e^{-\lambda} \cdot \frac{(e^t \cdot \lambda)^k}{k!} \\&= e^{-\lambda} \cdot e^{(e^t \lambda)} \\&= e^{\lambda(e^t - 1)}.\end{aligned}$$

Example 4. Let Z be a standard normal r.v.

$$\begin{aligned}M_Z(t) &= E[e^{tZ}] \\&= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{tz} e^{-\frac{z^2}{2}} dz \\&= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\frac{t^2}{2}} \cdot e^{-\frac{(z-t)^2}{2}} dz \\&\stackrel{\text{Letting } x = z-t}{=} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\frac{t^2}{2}} e^{-\frac{x^2}{2}} dx \\&= e^{t^2/2}\end{aligned}$$

Example 5. Let X be a normal r.v. with mean μ and variance σ^2 .

Notice that $Z := \frac{X - \mu}{\sigma}$ is a standard normal r.v.

$$\text{Hence } X = \mu + \sigma Z.$$

$$\begin{aligned} M_X(t) &= E[e^{t(\mu + \sigma Z)}] \\ &= E[e^{t\mu} \cdot e^{t\sigma Z}] \\ &= e^{t\mu} E[e^{t\sigma Z}] \\ &= e^{t\mu} \cdot M_Z(t\sigma) \\ &= e^{t\mu} \cdot e^{\frac{t^2 \sigma^2}{2}} = e^{\frac{\sigma^2 t^2}{2} + \mu \cdot t} \end{aligned}$$

Thm 6. Let X, Y be two r.v.'s.

If $\exists t_0 > 0$ such that

$$M_X(t) = M_Y(t) \quad \text{for } t \in (-t_0, t_0)$$

and both are finite.

Then X and Y have the same distribution

$$(\text{i.e. } F_X = F_Y)$$

Due to this result, we say that the moment generating function determines the distribution.

Prop 7. If X and Y are independent

$$\text{then } M_{X+Y}(t) = M_X(t) \cdot M_Y(t).$$

$$\begin{aligned}
 \text{Pf. } M_{X+Y}(t) &= E[e^{tx+ty}] \\
 &= E[e^{tx} \cdot e^{ty}] \\
 &= E[e^{tx}] \cdot E[e^{ty}] \quad (\text{here we use the} \\
 &= M_X(t) \cdot M_Y(t). \quad \text{independency of } X \text{ and } Y)
 \end{aligned}$$

□

Example 8. Let X, Y be independent Poisson r.v.'s with parameters λ_1, λ_2 , respectively.

Show that $X+Y$ is a Poisson r.v. with parameter $\lambda_1 + \lambda_2$.

Pf. Notice that by Example 3,

$$M_X(t) = e^{\lambda_1(e^t-1)}, \quad M_Y(t) = e^{\lambda_2(e^t-1)}.$$

Since X and Y are independent,

$$M_{X+Y}(t) = M_X(t) M_Y(t) = e^{(\lambda_1 + \lambda_2)(e^t-1)}.$$

By Example 3 and Thm 6, $X+Y$ has the Poisson distribution with parameter $\lambda_1 + \lambda_2$. □