

## Topic Lists (Tentative):

Week 7 : Expectation

Week 8: Moment Generating Functions

Week 9: Normal Approximation and Transformation of R.V.s

Week 10: Joint Distribution

### Expectation

For a random variable  $X$ ,

$$\mathbb{E}[X] \stackrel{\text{def}}{=} \begin{cases} \sum_{k \in S_X} k \mathbb{P}(X=k) & X: \text{discrete} \\ \int_{-\infty}^{\infty} x \underbrace{f(x)}_{\text{P.D.F}} dx & X: \text{continuous} \end{cases}$$

$\mathbb{E}[X]$ : the expected value  $X$ .

More generally, let  $g: \mathbb{R} \rightarrow \mathbb{R}$ , then  $g(X)$  is also a R.V.

with

$$\mathbb{E}[g(X)] \stackrel{\text{def}}{=} \begin{cases} \sum_{k \in S_X} g(k) \mathbb{P}(X=k) & X: \text{discrete} \\ \int_{-\infty}^{\infty} g(x) f(x) dx & X: \text{continuous} \end{cases}$$

Moreover,  $\text{Var}(X) \stackrel{\text{def}}{=} \mathbb{E}[(X - \mathbb{E}[X])^2]$

$$= \underbrace{\mathbb{E}[X^2] - (\mathbb{E}[X])^2}_{\text{easier way to calculate}}$$

easier way to calculate

$$\text{e.g. } f(t) = \begin{cases} 2t & t \in [0, 1] \\ 0 & \text{o.w.} \end{cases}$$

Claim 1:  $f$  is a valid density function.

Proof: ①  $f(t) \geq 0 \quad \forall t \in \mathbb{R}$

$$\begin{aligned} \text{② } \int_{-\infty}^{\infty} f(t) dt &= \int_{-\infty}^0 0 dt + \int_0^1 2t dt + \int_1^{\infty} 0 dt \\ &= \int_0^1 2t dt = t^2 \Big|_0^1 = 1. \end{aligned}$$

Claim 2: Let  $X$  be a R.V. with density  $f$ , calculate  $\mathbb{E}[X]$  and  $\text{Var}(X)$ .

$$\begin{aligned} \text{Soln: } \mathbb{E}[X] &= \int_{-\infty}^{\infty} t f(t) dt \\ &= \int_0^1 t \cdot 2t dt = 2 \int_0^1 t^2 dt = \frac{2}{3} \end{aligned}$$

$$\text{Var}(X) = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$$

$$\begin{aligned} \mathbb{E}[X^2] &= \int_{-\infty}^{\infty} t^2 f(t) dt = \int_0^1 t^2 \cdot 2t dt \\ &= 2 \int_0^1 t^3 dt = \frac{1}{2} \end{aligned}$$

$$\text{Var}(X) = \frac{1}{2} - \left(\frac{2}{3}\right)^2 = 5/18$$

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Normal Distribution

$$X \sim N(\mu, \sigma^2)$$

$$\text{density: } f(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\} \quad x \in \mathbb{R}$$

$$\mathbb{E}[X] = \mu, \quad \text{Var}(X) = \sigma^2$$

Facts:

① If  $X \sim N(\mu, \sigma^2)$ , then

$$Z = \frac{X - \mu}{\sigma} \sim N(0, 1) \text{ — standard Normal Dist}$$

② For  $Z \sim N(0, 1)$ , its CDF:

$$\Phi(z) = \mathbb{P}(Z \leq z) \text{ obtained from Normal Table.}$$

$$\text{e.g. } X \sim N(68, 16) \Rightarrow Z = \frac{X - 68}{\sqrt{16}} = \frac{X - 68}{4} \sim N(0, 1)$$

Type I Problem: Calculate the probability.

$$\begin{aligned} \mathbb{P}(X \geq 72) &= \mathbb{P}\left(\frac{X - 68}{4} \geq \frac{72 - 68}{4}\right) \\ &= \mathbb{P}(Z \geq 1) \quad Z \sim N(0, 1) \\ &= 1 - \Phi(1) = 1 - .8413 = .1587 \\ &\quad \uparrow \\ &\quad \text{Normal Table} \end{aligned}$$

Type II Problem: Given the probability, obtain the threshold

Example: given  $\mathbb{P}(X \geq x) = 28.1\%$ , calculate  $x$ ?

$$\begin{aligned} \text{Soln: } \mathbb{P}(X \geq x) &= \mathbb{P}\left(\frac{X - 68}{4} \geq \frac{x - 68}{4}\right) \\ &= \mathbb{P}\left(Z \geq \frac{x - 68}{4}\right) = 28.1\% \end{aligned}$$

$$\Phi\left(\frac{x-68}{4}\right) = 1 - 28.1\% = 71.9\%$$

Figure out  $\frac{x-68}{4}$  by Normal Table.

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## Gamma Distribution

Gamma Function  $\Gamma(\alpha) \stackrel{\text{def}}{=} \int_0^{\infty} x^{\alpha-1} e^{-x} dx \quad \alpha > 0$

Facts: ①  $\Gamma(1) = \int_0^{\infty} e^{-x} dx = 1$

②  $\Gamma(\alpha) = (\alpha-1)\Gamma(\alpha-1) \quad \alpha > 1$

(Integration by parts)

③  $\forall n \in \mathbb{N}, \Gamma(n) = (n-1)! \quad (\text{Mathematical Induction})$

④  $\alpha, \beta > 0 : \int_0^{\infty} x^{\alpha-1} e^{-\beta x} dx = \frac{\Gamma(\alpha)}{\beta^{\alpha}}$

$X \sim \text{Gamma}(r, \lambda)$  if its density

$$f(x) = \begin{cases} \frac{\lambda^r}{\Gamma(r)} x^{r-1} e^{-\lambda x} & x > 0 \\ 0 & \text{o.w.} \end{cases}$$

e.g. (From the Past Exam)  $X$  has a density

$$f(x) = \begin{cases} 2x e^{-2x} & x > 0 \\ 0 & \text{o.w.} \end{cases}$$

(a) Figure out  $c$  that makes  $f$  a valid p.d.f.

Soln:  $f$  is a form of Gamma distribution with

$$r = \lambda = 2, \text{ so } c = \frac{\lambda^r}{\Gamma(r)} = \frac{2^2}{\Gamma(2)} = 4$$

$$\Gamma(2) = (2-1)! = 1. \quad //$$

(b) Calculate  $\mathbb{E}[X]$  and  $\text{Var}(X)$ .

Soln: By Distribution sheet,  $X \sim \text{Gamma}(2, 2)$

$$\text{implies } \mathbb{E}[X] = \frac{r}{\lambda} = 1, \text{ Var}(X) = \frac{r}{\lambda^2} = \frac{1}{2}. \quad //$$

(c) More generally, calculate the  $n$ -th moment of  $X$ , which is  $\mathbb{E}[X^n]$ ,  $n \in \mathbb{N}$ .

$$\text{Soln: } \mathbb{E}[X^n] = \int_0^\infty x^n \cdot 4x e^{-2x} dx$$

$$= 4 \int_0^\infty x^{n+1} e^{-2x} dx$$

$$= 4 \cdot \frac{\Gamma(n+2)}{2^{n+2}} \quad (\text{use Fact } \textcircled{4}, \text{ with } \begin{cases} \alpha = n+2 \\ \beta = 2 \end{cases})$$

$$= \frac{4 (n+1)!}{2^{n+2}} = \frac{(n+1)!}{2^n} \quad //$$