# Review (3.4)

- Uniform distribution, its mean, variance, m.g.f.
- Exponential distribution and its mean, variance.
- Memoryless properties of exponential distribution.

Today's Lecture (3.7, 3.5)

- Comparison between discrete and continuous random variables
- Cauchy distribution, GOT A MOMENT?
- Gamma distribution and its connection to Poisson Process
- Mean, variance and m.g.f. of Gamma distribution.
- $\chi^2$  distribution.

#### random variables

### discrete

#### continuous

probability mass function

$$p.m.f.$$

$$p(x) = P(X = x)$$

 $\forall x \quad 0 \le p(x) \le 1$ 

$$\sum_{\text{all } x} p(x) = 1$$

probability density function

$$p.d.f.$$
  $f(x)$ 

$$\forall x \quad f(x) \ge 0$$

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

cumulative distribution function

$$F(x) = P(X \le x)$$

$$F(x) = \sum_{y \le x} p(y)$$

$$F(x) = \int_{-\infty}^{x} f(y) dy$$

expected value

$$E(X) = \mu_X$$

discrete

continuous

If 
$$\sum_{\text{all } x} |x| \cdot p(x) < \infty$$
,

$$E(X) = \sum_{\text{all } x} x \cdot p(x)$$

If 
$$\int_{-\infty}^{\infty} |x| \cdot f(x) dx < \infty$$
,

$$E(X) = \int_{-\infty}^{\infty} x \cdot f(x) \, dx$$

discrete

If 
$$\sum_{\text{all } x} |g(x)| \cdot p(x) < \infty$$
,

$$E(g(X)) = \sum_{\text{all } x} g(x) \cdot p(x)$$

$$\text{If } \int\limits_{-\infty}^{\infty} \left| g(x) \right| \cdot f(x) \, dx < \infty,$$

$$E(g(X)) = \int_{-\infty}^{\infty} g(x) \cdot f(x) \, dx$$

#### variance

$$\mathrm{Var}(\, \mathrm{X} \,) \; = \; \sigma_{\mathrm{X}}^2 \; = \; \mathrm{E}(\, [\, \mathrm{X} - \mu_{\mathrm{X}} \,]^{\, 2}) \; = \; \mathrm{E}(\, \mathrm{X}^{\, 2}) \, - \, [\, \mathrm{E}(\, \mathrm{X} \,) \,]^{\, 2}$$

discrete

#### continuous

$$\operatorname{Var}(\mathbf{X}) = \sum_{\text{all } x} (x - \mu_{\mathbf{X}})^2 \cdot p(x) \qquad \operatorname{Var}(\mathbf{X}) = \int_{-\infty}^{\infty} (x - \mu_{\mathbf{X}})^2 \cdot f(x) dx$$
$$= \sum_{\text{all } x} x^2 \cdot p(x) - \left[ \mathbf{E}(\mathbf{X}) \right]^2 \qquad = \left[ \int_{-\infty}^{\infty} x^2 \cdot f(x) dx \right] - \left[ \mathbf{E}(\mathbf{X}) \right]^2$$

## moment-generating function

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

discrete

#### continuous

$$M_X(t) = \sum_{\text{all } x} e^{tx} \cdot p(x)$$
  $M_X(t) = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx$ 

#### Example

(Standard) Cauchy distribution: 
$$f_{X}(x) = \frac{1}{\pi(1+x^{2})}, -\infty < x < \infty.$$

Even though  $f_{\mathbf{X}}(x)$  is symmetric about zero,  $\mathbf{E}(\mathbf{X})$  is undefined since

$$\int_{-\infty}^{\infty} |x| \cdot \frac{1}{\pi (1+x^2)} dx = \infty.$$

$$F_X(x) = \int_{-\infty}^{x} \frac{1}{\pi(1+y^2)} dy = \frac{1}{\pi} \arctan(x) + \frac{1}{2}, \quad -\infty < x < \infty.$$

$$M_X(0) = 1.$$
  $M_X(t)$  is undefined for all  $t \neq 0$ .

# Volcano eruption (again)

Let  $N_t$  be the number of volcano eruptions to have occurred by time t, starting from now. Suppose that the volcano eruption forms a Poisson process with rate  $\lambda$ . Then  $N_t \sim$  \_\_\_\_\_\_.

Let X be the waiting time until the 4-th volcano eruption occurs and find the distribution of X.

$$F_X(x) = \underline{\hspace{1cm}}$$

In general, if X is the waiting time until the  $\alpha$ -th volcano eruption, then

$$F_X(x) = 1 - \sum_{k=0}^{\alpha - 1} \frac{(\lambda x)^k e^{-\lambda x}}{k!}, \ x > 0$$

and

$$f_X(x) = F'_X(x) = \frac{\lambda^{\alpha} x^{\alpha - 1}}{(\alpha - 1)!} e^{-\lambda x}, \quad x > 0$$

## Gamma function

$$\Gamma(t) = \int_0^\infty y^{t-1} e^{-y} dy, \quad t > 0$$

Properties of  $\Gamma(t)$ :

- $\Gamma(t) = (t-1)\Gamma(t-1), t > 1.$
- $\bullet \ \Gamma(1) = \int_0^\infty e^{-y} dy = 1.$
- When t = n, a positive integer,

$$\Gamma(n) = \Gamma(n-1)(n-1) = \dots = (n-1)!$$

### Gamma distribution

Definition: The random variable X has a gamma distribution if its p.d.f. is defined by

$$f(x) = \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x^{\alpha - 1} e^{-x/\theta}, \quad 0 \le x < \infty$$

Write Gamma $(\alpha, \theta)$ , where  $\theta = 1/\lambda$ ,  $\alpha > 0$  (not necessarily an integer).

$$M(t) = \frac{1}{(1 - \theta t)^{\alpha}}, \ t < 1/\theta$$

$$\mu = M'(0) = \alpha \theta$$

$$\sigma^{2} = M''(0) - [M'(0)]^{2} = \alpha(\alpha + 1)\theta^{2} - \alpha^{2}\theta^{2} = \alpha\theta^{2}$$

## Example

Telephone calls enter a college switchboard at a mean rate of 1/2 call per minute according to a Poisson process. Let X denote the waiting time until the second call arrives.

- What is the distribution of X?
- What is the average waiting time?
- What is the probability that the waiting time is longer than 3 minutes?

## Chi-square distribution

Gamma distribution:

$$f(x) = \frac{1}{\Gamma(\alpha)\theta^{\alpha}} x^{\alpha - 1} e^{-x/\theta}, \quad 0 \le x < \infty$$

Chi-square distribution:  $\theta=2$ ,  $\alpha=r/2$ , r is a positive integer.

$$f(x) = \frac{1}{\Gamma(r/2)2^{r/2}} x^{r/2 - 1} e^{-x/2}, \quad 0 \le x < \infty$$

r: degree of freedom. Write  $X \sim \chi^2(r)$ .

- Mean:  $\mu = \alpha\theta = (r/2)2 = r$
- Variance:  $\sigma^2 = \alpha \theta^2 = (r/2)2^2 = 2r$ .
- m.g.f.  $M(t) = (1-2t)^{-r/2}$ , t < 1/2.

Plot: