

# PROBABILISTIC METHODS IN ENGINEERING VE401

## Assignment II

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Team number: 20

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## Exercise 2.1

1) Since  $\operatorname{ran} X \subset \mathbb{N}$ , then

$$E[X] = \sum_{x=0}^{\infty} x P[X = x] = \sum_{x=0}^{\infty} \sum_{y=1}^{x} P[X = x].$$

We change the order of summation,

$$E[X] = \sum_{x=0}^{\infty} \sum_{y=1}^{x} P[X = x] = \sum_{x=0}^{\infty} \sum_{y=x+1}^{\infty} P[X = y] = \sum_{x=0}^{\infty} P[X > x].$$

2) According to the result we obtain in 1), we perform the similar calculation to  $(X, f_X)$  where X = 1, 2, ..., N - r + 1.

$$E[X] = \sum_{x=0}^{N-r} P[X > x].$$

In fact, P[X > x] is equal to the probability that the first x balls are all black.

$$\begin{split} P[X > x] &= P[\text{The first x balls are all black}] \\ &= \frac{N-r}{N} \frac{N-r-1}{N-1} \dots \frac{N-r-x+1}{N-x+1} \\ &= \frac{(N-r)!/(N-r-x)!}{N!/(N-x)!} = \frac{\binom{N-x}{r}}{\binom{N}{r}} \\ E[X] &= \frac{1}{\binom{N}{r}} \sum_{x=0}^{r'} \binom{N-x}{N-r'} = \frac{1}{\binom{N}{r}} \sum_{x=0}^{r'} \binom{N-r'+x}{N-r'} \\ &= \frac{\binom{N+1}{N-r'+1}}{\binom{N}{r}} = \frac{\binom{N+1}{r+1}}{\binom{N}{r}} = \frac{N+1}{r+1}. \end{split}$$

#### Exercise 2.2

For  $p_0(t)$ , when t = 0 the probability of 0 failure is 1. For  $p_x(t)$   $(x \ge 1)$ , when t = 0 the probability of x failure is 0. Hence, the initial conditions are as the follows.

$$p_0(0) = 1$$
  
 $p_x(0) = 0, \quad x \ge 1.$ 

Since  $p_0' = -\lambda p_0$ , then  $p_0(t) = c \cdot e^{-\lambda t}$ . Also we know that  $p_0(0) = 1$ , thus c=1. Hence,  $p_0(t) = e^{-\lambda t}$ .

Assume that  $p_x(t) = (\lambda t)^x e^{-\lambda t}/x!$  for all  $x \in \mathbb{N}$ . For x = 0,  $p_0(t) = e^{-lambdat} = (\lambda t)^0 e^{-\lambda t}/0!$ , which follows our assumption. For  $x = n \ge 0$ , we assume the statement is true. When x = n + 1,

$$p'_{n+1} + \lambda p_{n+1} = \lambda p_n,$$

$$(e^{\lambda t} p_x)' = \lambda e^{\lambda t} p_n,$$

$$p_{n+1} = \frac{\int \lambda e^{\lambda t} p_n dt}{e^{\lambda t}}.$$

$$p_{n+1} = \frac{\lambda e^{\lambda t} p_n dt}{e^{\lambda t}} = \frac{\int \lambda e^{\lambda t} (\lambda t)^n e^{-\lambda t} / n! dt}{e^{\lambda t}}$$

$$= \frac{\lambda^{n+1} / n! \int t^n dt}{e^{\lambda t}} = \lambda^{n+1} / n! / (n+1)(t^{n+1} + C)e^{-\lambda t}$$

$$= (\frac{(\lambda t)^{n+1} + (\lambda)^{n+1} C}{(n+1)!})e^{-\lambda t}.$$

To satisfy that  $p_{n+1}(0) = 0$ , C has to be 0. Thus,

$$p_{n+1}(t) = \frac{(\lambda t)^{n+1}}{(n+1)!} e^{-\lambda t}.$$

Therefore,  $p_x(t) = (\lambda t)^x e^{-\lambda t} / x!$  for all  $x \in \mathbb{N}$ .

## Exercise 2.3

$$f(x) = \binom{n}{x} p^x (1-p)^{n-x} = \binom{n}{x} (\frac{k}{n})^x (1-\frac{k}{n})^{n-x} = k^x \frac{\binom{n}{x}}{n^x} ((1-\frac{1}{n/k})^{n/k})^k (1-\frac{k}{n})^{-x}$$

Since

$$\lim_{n \to \infty} \frac{\binom{n}{x}}{n^x} = \lim_{n \to \infty} \frac{n!}{x!(n-x)!n^x} = \frac{1}{x!},$$

$$\lim_{n \to \infty} ((1 - \frac{1}{n/k})^{-n/k})^{-k} = e^{-k},$$

$$\lim_{n \to \infty} (1 - \frac{k}{n})^{-x} = 1.$$

then

$$\lim_{n \to \infty} f(x) = \frac{k^x}{x!} e^{-k}.$$

## Exercise 2.4

1)

$$\begin{split} E[V] &= \int_0^\infty (\frac{2}{\pi})^{1/2} (m/kT)^{3/2} v^3 e^{-\frac{m}{kT} v^2/2} dv \\ &= (\frac{2}{\pi})^{1/2} (m/kT)^{3/2} \int_0^\infty v^3 e^{-\frac{m}{kT} v^2/2} dv \quad (w = v^2) \\ &= (\frac{2}{\pi})^{1/2} (m/kT)^{3/2} \int_0^\infty w^{3/2} e^{-\frac{m}{kT} w/2} \frac{1}{2} w^{-1/2} dw \\ &= (\frac{1}{2\pi})^{1/2} (m/kT)^{3/2} \int_0^\infty w e^{-\frac{m}{kT} w/2} dw \\ &= (\frac{m}{2kT\pi})^{1/2} \int_0^\infty \frac{m}{kT} w e^{-\frac{m}{kT} w/2} dw \quad (z = -\frac{m}{kT} w/2) \end{split}$$

$$E[V] = (\frac{8kT}{m\pi})^{1/2} \int_{-\infty}^{0} ze^{z} dz = (\frac{8kT}{m\pi})^{1/2}.$$

Since  $Var[V] = E[V^2] - E[V]^2$ , then

$$\begin{split} E[V^2] &= (\frac{2}{\pi})^{1/2} (m/kT)^{3/2} \int_0^\infty v^4 e^{-\frac{m}{kT}v^2/2} dv \\ &= -(\frac{2m}{k\pi T})^{1/2} \int_1^0 v^3 de^{-\frac{m}{kT}v^2/2} \\ &= -(\frac{2m}{k\pi T})^{1/2} (v^3 e^{-\frac{m}{kT}v^2/2} \Big|_0^\infty - \int_0^\infty 3v^2 e^{-\frac{m}{kT}v^2/2} dv) \\ &= 3(\frac{2m}{k\pi T})^{1/2} \int_0^\infty v^2 e^{-\frac{m}{kT}v^2/2} dv \\ &= 3(\frac{2kT}{m\pi})^{1/2} \int_0^\infty e^{-\frac{m}{kT}v^2/2} dv \\ &= \frac{3kT}{m}. \end{split}$$

$$Var[V] = E[V^2] - E[V]^2 = \frac{3kT}{m} - \frac{8kT}{m\pi} = \frac{kT}{m}(3 - \frac{8}{\pi}).$$

2)

$$E[E] = E[mV^2/2] = mE[V^2]/2 = \frac{3kT}{2}.$$

3) Since

$$E = \varphi(v) = \frac{mv^2}{2},$$

$$V = \varphi^{-1}(\varepsilon) = (\frac{2\varepsilon}{m})^{1/2},$$

then

$$f_E(\varepsilon) = f_V(\varphi^{-1}(\varepsilon)) \left| \frac{d\varphi^{-1}(\varepsilon)}{d\varepsilon} \right| = \left(\frac{1}{2m\varepsilon}\right)^{1/2} \left(\frac{2}{\pi}\right)^{1/2} \left(\frac{m}{kT}\right)^{3/2} \frac{2\varepsilon}{m} e^{-\frac{\varepsilon}{kT}}$$
$$= 2\left(\frac{\varepsilon}{\pi k^3 T^3}\right)^{1/2} e^{-\frac{\varepsilon}{kT}}.$$

## Exercise 2.5

$$\begin{split} \Gamma(\frac{2n+1}{2}) &= \int_0^\infty t^{(2n-1)/2} e^{-t} dt = (t^{(2n-1)/2} e^{-t}) \bigg|_0^\infty + \frac{2n-1}{2} \int_0^\infty t^{(2n-3)/2} e^{-t} dt \\ &= (\frac{2n-1}{2}) \Gamma(\frac{2n-1}{2}). \end{split}$$

Hence,

$$\Gamma(\frac{2n+1}{2}) = \frac{(2n-1)(2n-3)\dots(1)}{2\cdot 2\dots 2}\Gamma(\frac{1}{2}) = \frac{(2n-1)!!}{2^n}\Gamma(\frac{1}{2}).$$

Also we obtain

$$\Gamma(\frac{1}{2}) = \int_0^\infty t^{-1/2} e^{-t} dt \quad (t = w^2)$$
$$= \int_0^\infty w^{-1} e^{-w^2} 2w dw = 2 \int_0^\infty e^{-w^2} dw = \sqrt{\pi}$$

Therefore,

$$\Gamma(\frac{2n+1}{2}) = \frac{(2n-1)!!}{2^n} \sqrt{\pi}.$$

## Exercise 2.6

1) Denote that

$$Z = \frac{X - \mu}{\sigma} = \frac{X - 6000}{100}$$

so that  $Z \sim N(0,1)$ . Then,

$$P[A \text{ sample strength is less than } 6250 \text{ kg/cm}^2]$$
  
=  $P[X < 6250] = P[Z < 2.50] = \Phi(2.50) = 0.9938.$ 

2)

$$P[5800 \le X \le 5900] = P[-2.00 \le Z \le -1.00]$$
$$= \Phi(-1.00) - \Phi(-2.00)$$
$$= 0.1587 - 0.0228$$
$$= 0.1359.$$

3) According to the standard normal distribution table, we know that

$$\Phi(-1.64) = 0.0505, \quad \Phi(-1.65) = 0.0495.$$

When 
$$Z = -1.65$$
,  $X = -1.65 \times 100 + 6000 = 5835 \text{kg/cm}^2$ .

Hence, the strength of  $5835 \text{ kg/cm}^2$  exceeds 5% samples, saying that is exceeded by 95% samples.

## Exercise 2.7

1) The random variable is a map  $X: S \to \mathbb{R}$  together with a function  $f_X: \mathbb{R} \to \mathbb{R}$ . Then, the two properties of a continuous random variable will be showed.

For a < x < b where b > a, f(x) = 1/(b-a) > 0. For other conditions, f(x) = 0 > 0. Hence,  $f(x) \ge 0$ .

Also,

$$\int_{-\infty}^{\infty} f(x)dx = \int_{a}^{b} 1/(b-a)dx = (b-a)/(b-a) = 1.$$

Hence, this is a density for a continuous random variable.

2) The graph is as the following.

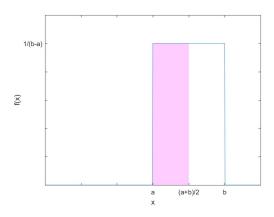


Figure 1: Graph for Exercise 2.7 (2)

3) 
$$P[X \le (a+b)/2] = \int_{-\infty}^{(a+b)/2} f(x)dx = \int_{a}^{(a+b)/2} 1/(b-a)dx = (b-a)/2/(b-a) = \frac{1}{2}.$$

4) Since [c,d] and [e,f] are the subintervals of [a,b], then

$$P[c \le X \le d] = \int_{c}^{d} 1/(b-a)dx = (d-c)/(b-a).$$
$$P[e \le X \le f] = \int_{e}^{f} 1/(b-a)dx = (f-e)/(b-a).$$

Since d-c=f-e, then  $P[c \le X \le d] = P[e \le X \le f]$ .

5)  $F(x) = \int_{-\infty}^{x} f(t)dt = \begin{cases} 0, & x < a \\ \frac{x-a}{b-a}, & a \le x \le b \\ 1, & x > b \end{cases}$ 

6)
$$E[X] = \int_{-\infty}^{\infty} x f(x) dx = \int_{a}^{b} \frac{x}{b-a} dx$$

$$= \frac{x^2}{2(b-a)} \Big|_{a}^{b} = \frac{b^2 - a^2}{2(b-a)} = \frac{a+b}{2}.$$

$$E[X^2] = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_{a}^{b} \frac{x^2}{b-a} dx$$

$$= \frac{x^3}{3(b-a)} \Big|_{a}^{b} = \frac{b^3 - a^3}{3(b-a)} = \frac{a^2 + ab + b^2}{3}.$$

Hence,

$$Var[X] = E[X^2] - E[X]^2 = \frac{a^2 + ab + b^2}{3} - (\frac{a+b}{2})^2$$
$$= \frac{(b-a)^2}{12}.$$