

VE401 Probabilistic Methods in Eng.

Recitation Class (Mid Part2)

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Discrete Random Variable

1 Fundamentals

1.1 Probability Density Function

A discrete random variable is a map $X : S \rightarrow \Omega$ together with a function $f_X : \Omega \rightarrow \mathbb{R}$ having the properties that

1. $f_X(x) \geq 0$ for all $x \in \Omega$ and
2. $\sum_{x \in \Omega} f_X(x) = 1$

Notation:

For discrete random variables, we define the density function f_X in such a way that

$$f_X(x) = P[X = x]$$

1.2 Cumulative Distribution Function

The cumulative distribution function of a random variable, defined as follows,

$$F_X(x) := P[X \leq x]$$

For a discrete random variable

$$F_X(x) = \sum_{y \leq x} f_X(y)$$

1.3 Expectation

Let (X, f_X) be a discrete random variable. Then the expected value of X is

$$E[X] = \sum_{x \in \Omega} x \cdot f_X(x)$$

provided that the sum (possibly series if Ω is infinite) on the right **converges absolutely**.

- St. Petersburg Paradox – Expectation doesn't exist.

Properties of the Expectation:

1. Let X be a random variable, and for some $c \in \mathbb{R}$,

$$E[c] = c, \quad E[cX] = cE[X]$$

2. Let X and Y be random variables,

$$E[X + Y] = E[X] + E[Y]$$

3. Let (X, f_X) be a discrete random variable and $\varphi : \Omega \rightarrow \mathbb{R}$ some function. Then the expected value of $\varphi \circ X$ is

$$E[\varphi \circ X] = \sum_{x \in \Omega} \varphi(x) \cdot f_X(x)$$

1.4 Variance

The variance of X is defined as the mean square deviation from the mean,

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

- Standard deviation: $\sigma_X = \sqrt{\text{Var } X} = \sqrt{\sigma_X^2}$
- Using the properties of the mean, we can find that

$$\begin{aligned} \text{Var } X &= E[(X - E[X])^2] \\ &= E[X^2 - 2E[X] \cdot X + E[X]^2] \\ &= E[X^2] - E[X]^2 \end{aligned}$$

Properties of the Variance

Let X be a random variable, and for some $c \in \mathbb{R}$,

$$\text{Var } c = 0, \quad \text{Var } cX = c^2 \text{Var}[X]$$

1.5 Moments

n^{th} (Ordinary) Moments of X :

$$E[X^n], \quad n \in \mathbb{N}$$

n^{th} Central Moments of X :

$$E\left[\left(\frac{X - \mu}{\sigma}\right)^n\right], \quad n = 3, 4, 5, \dots$$

Moment-generating function of X :

For (X, f_X) such that the sequence of $E[X^k]$, $n \in \mathbb{N}$ exists. If the power series

$$m_X(t) := \sum_{k=0}^{\infty} \frac{E[X^k]}{k!} t^k$$

has radius of convergence $\varepsilon > 0$, then thereby the function

$$m_X(t) : (-\varepsilon, \varepsilon) \rightarrow \mathbb{R}$$

is called the m.g.f of X .

Theorem Let $\varepsilon > 0$ be given such that $E[e^{tX}]$ exists and has a power series expansion in t that converges for $|t| < \varepsilon$. Then the moment-generating function exists and

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

for $|t| < \varepsilon$ Furthermore,

$$E[X^k] = \left. \frac{d^k m_X(t)}{dt^k} \right|_{t=0}$$

2 Common Distributions

2.1 Bernoulli Distribution

Suppose that the probability of success is p , where $0 < p < 1$.

$$f_X(x) = \begin{cases} 1-p & \text{for } x = 0 \\ p & \text{for } x = 1 \end{cases}$$

Such an experiment is said to be a Bernoulli trial.

Example: an experiment that can result in two possible outcomes, like tossing a coin. independent and identical trials

2.2 Binomial Distribution

Meaning: the probability of getting x successes in n Bernoulli trials.

$$f_X(x) = P[x \text{ successes in } n \text{ trials}] = \binom{n}{x} p^x (1-p)^{n-x}$$

Example: getting 3 ones when rolling a dice for 10 times.

Features:

- $m_X(t) = (q + pe^t)^n$, $m_X : \mathbb{R} \rightarrow \mathbb{R}$, $q = 1 - p$
- $E[X] = np$
- $\text{Var}[X] = npq$

2.3 Geometric Distribution

Meaning: the probability of requiring x trials to obtain the first success.

$$f_X(x) = P[\text{not succeed until } x\text{th trial}] = (1-p)^{x-1}p$$

Example: tossing a coin for 6 times to get the first head.

Features:

- $m_X(t) = \frac{pe^t}{1-qe^t}$, $m_X : (-\infty, -\ln q) \rightarrow \mathbb{R}$, where $q = 1 - p$
- $E[X] = \frac{1}{p}$
- $\text{Var}[X] = \frac{q}{p^2}$

2.4 Pascal Distribution

Meaning: the probability that $x \geq r$ trials are needed to obtain r successes.

Let $r \in \mathbb{N} \setminus \{0\}$, $x \in \{r, r+1, r+2, \dots\}$, the distribution function f_X given by

$$f_X(x) = \binom{x-1}{r-1} p^r (1-p)^{x-r}, \quad 0 < p < 1$$

is said to follow a Pascal distribution with parameters p and r .

Example: find the probability of using n tossing to get the third head.

Features:

- $m_X : (-\infty, -\ln q) \rightarrow \mathbb{R}, \quad m_X(t) = \frac{(pe^t)^r}{(1-qe^t)^r}, \quad q = 1 - p$
- $E[X] = r/p$
- $\text{Var } X = rq/p^2$

2.5 Negative Binomial Distribution

Meaning: the probability that x failures are needed to obtain r successes.

Let $r \in \mathbb{N} \setminus \{0\}$, $x \in \mathbb{N}$, and distribution function f_X is given by

$$f_X(x) = \binom{x+r-1}{r-1} p^r (1-p)^x = \binom{-r}{x} (-1)^x p^r (1-p)^x, \quad 0 < p < 1$$

is said to follow a negative binomial distribution with parameters p and r .

2.6 Hypergeometric Distribution

meaning: The probability of having exactly x red balls after drawing n balls from the urn that has r red balls and $N - r$ black balls.

Assume that

$$r > n \quad \text{and} \quad N - r > n$$

Then we get

$$\begin{aligned} f_X(x) &= P[\text{exactly } x \text{ red balls out of } n \text{ selected}] \\ &= \frac{(\# \text{ ways to select } x \text{ out of } r \text{ balls}) \cdot (\# \text{ ways to select } n - x \text{ out of } N - r \text{ balls})}{\# \text{ ways to select } n \text{ out of } N \text{ balls}} \\ &= \frac{\binom{r}{x} \binom{N-r}{n-x}}{\binom{N}{n}} \end{aligned}$$

Example: A production lot of 200 units has 8 defectives. Find the probability of selecting exactly one defective when we choose a random sample of 10 units.

Features:

- $E[X] = n \frac{r}{N}$
- $\text{Var } X = n \frac{r}{N} \frac{N-r}{N} \frac{N-n}{N-1}$

Relation with Binomial distribution:

- Binomial distribution: a series of i.i.d Bernoulli trials.
- Hypergeometric distribution: a series of identical but not independent Bernoulli trials.
- For Binomial distribution, $E[X] = np$, $\text{Var } X = npq$, where

$$p = \frac{r}{N}, \quad q = \frac{N-r}{N}$$

so only the variance differs by $\frac{N-n}{N-1}$

- the binomial distribution may be used to approximate the hypergeometric distribution if the sampling fraction n/N is small (< 0.05)

**For Assignment3

Exercise 3.4 A Tricky Question involving the Binomial Distribution

A mathematics textbook has 200 pages on which typographical errors in the equations could occur. Suppose there are in fact five errors randomly dispersed among these 200 pages.

- i) What is the probability that a random sample of 50 pages will contain at least one error?
(2 Marks)
- ii) How large must the random sample be to assure that at least three errors will be found with 90% probability? (You may use a normal approximation to the binomial distribution.)
(3 Marks)

Solution:

- This follows binomial distribution, as the appearance of typos are i.i.d.
- Each typo chooses a page from the book. The page can be the same.

i)

$$P[N \geq 1] = 1 - P[N = 0] = 1 - \left(\frac{150}{200}\right)^5 \approx 0.76$$

ii)

$$P[X \leq y] = \sum_{x=0}^y \binom{n}{x} p^x (1-p)^{n-x} \approx \Phi\left(\frac{y + 1/2 - np}{\sqrt{np(1-p)}}\right)$$

Denote the sample size by x , we have $n = 5, p = \frac{x}{200}$

$$\begin{aligned} P[N \geq 3] &= 1 - P[N \leq 2] = 0.1 = \Phi\left(\frac{2 + 1/2 - \frac{x}{40}}{\sqrt{\frac{x}{40} \cdot (1 - \frac{x}{200})}}\right) \\ &\Rightarrow \frac{2 + 1/2 - \frac{x}{40}}{\sqrt{\frac{x}{40} \cdot (1 - \frac{x}{200})}} \approx -1.28155 \\ &\Rightarrow x \geq 150 \end{aligned}$$

2.7 Poisson Distribution

Meaning: the probability of x occurrences of random events at a constant rate in a continuous environment.

Assumptions:

- (i) Independence: If the intervals $T_1, T_2 \subset [0, t]$ do not overlap, then the numbers of arrivals in these intervals are independent of each other.
- (ii) Constant rate of arrivals.

Mathematical Assumptions: There exists a number $\lambda > 0$ (arrival rate) such that for any small time interval of size Δt the following postulates are satisfied:

1. The probability that exactly one arrival will occur in an interval of width Δt is $\lambda \cdot \Delta t + o(\Delta t)$
2. The probability that exactly zero arrivals will occur in the interval is $1 - \lambda \cdot \Delta t + o(\Delta t)$
3. The probability that two or more arrivals occur in the interval is $o(\Delta t)$

After calculation, we find the probability x arrivals in the time interval $[0, t]$ is given by

$$p_x(t) = \frac{(\lambda t)^x}{x!} e^{-\lambda t}, \quad x \in \mathbb{N}$$

Or let $k = \lambda t \in \mathbb{R}, x \in \mathbb{N}$. the density function $f_X : \mathbb{N} \rightarrow \mathbb{R}$ given by

$$f_X(x) = \frac{k^x e^{-k}}{x!}$$

is said to follow a Poisson distribution with parameter k

Examples: A healthy individual may have an average white blood cell count of as low as 4500/mm³ of blood. To detect a white-cell deficiency, a 0.001mm³ drop of blood is taken and the number X of white blood cells is found ($k = \lambda t = 4500 \times 0.001 = 4.5$).

Features:

- $m_X : \mathbb{R} \rightarrow \mathbb{R}, \quad m_X(t) = e^{k(e^t - 1)}$
- $E[X] = k$
- $\text{Var } X = k$

** By checking the variance and expectation, we can know whether the case follows a Poisson distribution.

Relation with Binomial Distribution:

For a Binomial distribution, when n is large and p is small, we can approximate the binomial distribution by a Poisson distribution with parameter

$$k = pn$$

We can prove it mathematically by showing

$$\binom{n}{m} p^m (1-p)^{n-m} \xrightarrow[n \cdot p = k]{n \rightarrow \infty} \frac{k^m}{m!} e^{-k}$$