CHAPTER 4 SOME DISCRETE PROBABILITY DISTRIBUTIONS

1. Bernoulli distribution

Definition "Bernoulli trial"

It is a random experiment, whose outcomes is either:

(a) "A = success" with probability p, 0<p<1, or

(b) "**not A = failure**" with probability 1 - p = q

Examples :

(i) Tossing a coin once and looking for "Head"(ii) Rolling a die once and looking for "face 1"

Now let X denote the number of successes in one Bernoulli trial.

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The possible values of X are 0, 1.
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It is clear that X is a discrete r. v. with

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P(X = 1) = P(success) = p, 0
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P(X = 0) = P(failure) = q, 0 < q < 1
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Such that p + q = 1.

The probability mass function (PMF) of X is

$$f(x) = P(X = x) = p^{x}q^{1-x}, x = 0, 1, 0 < p, q < 1, p + q = 1$$

This is called the Bernoulli distribution with one parameter p.

This is referred to simply, as

$$X \sim Ber(p)$$

Characteristics of Bernoulli distribution

(a) Mean E(X) = p

(b) Variance Var(X) = p q

(c) Standard deviation $\sigma = \sqrt{pq}$

(d) The moment generating function

$$M_{X}(t) = E[e^{tX}] = q + pe^{t}$$

(e) The probability generating function is $G_X(t) = E[t^X] = q + pt$

2. Binomial distribution

Consider n Bernoulli trials such that:

- C1. The outcome of each trial is either "success" with probability p, or "failure" with probability q
- C2. The probability p of success remains constant, i.e. it doesn't change from trial to trial.
- C3. The trials are independent.

Let X be a r. v. denoting the number of successes in these n trials

The possible values of X are X = 0, 1, 2, ..., n

It is clear that X is a discrete r. v.

The probability mass function of X is

$$f(x) = P(X = x) = {\binom{n}{x}} p^{x} q^{n-x}, x = 0, 1, 2, ..., n$$

This is called binomial distribution with parameters n and p. We write $X \sim Bin(n,p)$

Characteristics of binomial distribution

(a) Mean = E(X) = n p
(b) Variance = Var(X) = n p q
(c) Standard deviation
$$\sigma = \sqrt{n p q}$$

(d) Moment generating function

$$M_{X}(t) = E[e^{tX}] = (q + pe^{t})^{n}$$

(e) Probability generating function $G(t) = (q + pt)^n$

Example (1):

A fair coin is tossed 6 times. Find the probability that the Head will appear:

- (a) Exactly 3 times
- (b) At least 2 times
- (c) At most 4 times

<u>Solution</u>

We have

$$f(x) = P(X = x) = {\binom{6}{x}} {\left(\frac{1}{2}\right)^6}, x = 0, 1, ..., 6$$

(a) P(X=3) = f(3)=0.313(b) $P(X \ge 2) = f(2) + f(3) + ... + f(6) = 1 - [f(0)+f(1)]$ = 1 - 0.109 = 0.891(c) $P(X \le 4) = 1 - [f(5) + f(6)] = 1 - 0.109 = 0.891$

Example (2) :

A person fires at a certain target in 6 independent successive trials with a constant probability 0.9 of hitting the target. Find the probability of hitting the target:

- (i) Exactly 2 times
- (ii) At least 4 times

(iii) At most 3 times

<u>Solution</u>:

Let X = number of times of hitting the target. X ~ Bin (6, 0.9). $f(x) = P(X = x) = {6 \choose x} (0.9)^x (0.1)^{6-x}, x = 0,1,2,...,6$

(i) Req. Prob. = P(X = 2) = f(2) = ...(ii) Req. Prob. = $P(X \ge 4) = f(4) + f(5) + f(6) = ...$ (iii) Req. Prob. = $P(X \le 3) = f(0)+f(1)+f(2)+f(3)=...$

• Example (3) :

Find the mean, the variance, and the standard deviation of the number of hitting in **Example** (2).

• Solution

Mean = E(X) = n p = 6 * 0.9 = 5.4

Variance = Var (X) = n p q = $6^* 0.9 * 0.1 = 0.54$

Standard deviation = σ = 0.73

Example (4): The moment generating function of the r. v. X is

$$M_X(t) = (0.25 + 0.75e^t)^6$$

Determine the probability distribution of X. Find the mean and variance.

Solution

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X ~ Bin (6, 0.75).
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E(X) = 6*0.75 = 4.5. Var (X) = 6*0.75*0.25 = 1.125
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3. Poisson distribution

Let X be a random variable denoting the number of successes in a sequence of n (n \ge 30) Bernoulli trials with probability p (p < 0.5) of successes, satisfying the conditions C1 – C3 of the binomial distribution such that n p = λ is finite.

It is clear that X is a discrete r. v. with possible values X = 0, 1, 2, ...

The probability mass function of X is

$$f(x) = P(X = x) = \frac{\lambda^{x}}{x!}e^{-\lambda}, x = 0, 1, 2, ..., 0 < \lambda < \infty$$

In this case, we say X has a Poisson Distribution with parameter λ and write

 $X \sim Poi(\lambda)$

Characteristics of Poisson distribution

- (a) Poisson distribution is a limiting distribution
 of the binomial distribution as n tends to ∞.
- (b) Poisson distribution is called the distribution of rare events. It is used when n is large and p is small, such that λ = n p < ∞.</p>

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(c) Mean = Variance = \lambda
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(d) Standard deviation = $\sqrt{\lambda}$

(e) The MGF is

$$M_X(t) = e^{\lambda(\exp t - 1)}$$

(f) The PGF is

$$G_X(t) = e^{\lambda(t-1)}$$

Example (5)

The number X of annual earthquakes in a certain country has a mean 4. What is the probability distribution of X.

Solution

Because of the earthquake is a rare event, then the distribution of X is Poisson distribution with parameter $\lambda = 4$.

Example (6)

Consider the case when X has a Poisson distribution with parameter 3.

In this case:

(i) The PMF of X is
$$f(x) = P(X = x) = \frac{3^{n}}{x!}e^{-3}, x = 0, 1, 2, ...\infty$$

(ii) E(X) = Var(X) = 3
(iii) P(X = 2) = f(2) = 4.5 e⁻³ = 0.224

(iv) P(X is at least 3) = P(X
$$\ge$$
 3) = f(3) + f(4) + ...
= 1 - [f(0) + f(1) + f(2)]
= 1 - 8.5 e⁻³ = 1 - 0.423 = 0.577

(v) P(X is at most 2) = P(X
$$\le$$
 2) = f(0) + f(1) + f(2)
= 8.5 e⁻³ = 0.423

4. Geometric distribution

Let X be a r. v. denoting the number of Bernoulli trials, required to obtain the first success.
The possible values of X are X = 1, 2, ...
It is clear that X is a discrete r. v. with
P(X = 1) = p

In general
$$f(x) = P(X = x) = pq^{x-1}, x = 1, 2, ...\infty$$

This is called the geometric distribution with parameter p.

We denote this by writing $X \sim Geom(p)$

$$E(X) = \frac{1}{p}, \quad Var(X) = \frac{q}{p^2},$$

$$M_{X}(t) = \frac{p e^{t}}{1 - q e^{t}}, \quad G_{X}(t) = \frac{p t}{1 - q t}$$

Example (7)

Consider the case $X \sim Geom(0.65)$

This means that X has a geometric distribution with parameter p = 0.65

Thus

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$$f(x) = P(X = x) = (0.65)(0.35)^{x-1}, x = 1, 2, ...\infty$$

/e have:

(i)
$$E(X) = 1/0.65 = 1.538$$

(ii) $Var(X) = 0.35 / (0.65)^2 = 0.828$
(iii) $P(X = 3) = f(3) = 0.080$
(iv) $P(X > 2) = 1 - f(1) - f(2) = 0.122$
(v) $P(X < 4) = f(1) + f(2) + f(3) = 0.957$

• <u>Example (8)</u>:

A person fires at a certain target in a sequence of independent successive trials with a constant probability 0.9 of hitting the target. Find the probability of hitting the target for the first time in:

- (i) Exactly 2 trials
- (ii) At least 4 trials
- (iii) At most 3 trials

<u>Solution</u>

Let X = number of trials, required to hit the target for the first time. Consequently,

X ~ Geo (0.9).

$$f(x) = (0.9)(0.1)^{x-1}, x = 1, 2, 3, \dots$$

(i) Req. Prob. = P(X = 2) = f(2) = ...

(ii) Req. Prob. = $P(X \ge 4) = 1 - [f(1)+f(2)+f(3)] = ...$ (iii) Req. Prob. = $P(X \le 3) = f(1) + f(2) + f(3) = ...$

• Example (9) :

Find the mean, the variance, and the standard deviation of the number of hitting in **Example** (8).

• Solution

Mean = E(X) = 1/p = 1/0.9 = 1.111Variance = Var (X) = $q/p^2 = 0.1 / (0.9)^2 = 0.123$ Standard deviation = $\sigma = 0.351$

• <u>Example (10)</u>:

The MGF of the random variable X is

$$M_{X}(t) = \frac{0.65e^{t}}{1 - 0.35e^{t}}$$

 Find the probability distribution of X. Find the mean, the variance, and the standard deviation of X.

<u>Solution</u>

It is clear that X has a geometric distribution with parameter p = 0.65. Therefore

$$f(x) = P(X = x) = (0.65)(0.35)^{x-1}$$
, $x = 1, 2, ...$

Mean = E(X) = 1/0.65 = 1.538

Variance = Var (X) = 0.35 / 0.65 = 0.828

Standard deviation = σ = 0.91

5. Negative binomial distribution

Let X be a r. v. denoting the number of Bernoulli trials required to obtain the first k successes.The possible values of X are k, k+1, k+2, ...It is clear that X is a discrete random variable.We can prove that the PMF of X has the form:

$$f(x) = P(X = x) = {}^{x-1}C_{k-1}p^{k}q^{x-k}, x = k, k+1, ...\infty$$

In this case, we say that X has a negative binomial distribution with parameters k, p.
This is referred to by writing X ~NB(k,p)
<u>Note that</u>

The negative binomial distribution reduces to the geometric distribution when k = 1.

Thus, the geometric distribution is a special case of the negative binomial distribution.

Characteristics of negative binomial distribution

(a) E(X) = k/p(b) $Var(X) = k q/p^2$ (c) The MGF is $M_X(\cdot) = \left(\frac{p e^t}{1 - q e^t}\right)^k$

(d) The probability generating function is

$$G_X(t) = \left(\frac{p t}{1 - q t}\right)^k$$

Example (11)

Consider the case when $X \sim NB(5,0.8)$. This means that X has negative binomial distribution with

In this case

(a)
$$E(X) = 5/0.8 = 6.25$$

(b) $Var(X) = 5*0.2 / (0.8)^2 = 1.563$
(c) $P(X = 7) = {6 \choose 4} (0.8)^5 (0.2) = 0.197$

6. Hyper-geometric distribution

Consider a collection of k of objects of a certain Type and N – k of another Type.

A random sample of size n is drawn without replacement.

Let X be a random variable denoting the number of objects of the first type in the selected sample.

The PMF of X is

$$f(x) = P(X = x) = \frac{\binom{k}{x}\binom{N-k}{n-x}}{\binom{N}{n}}, \ x = \max(0, n+k-N), \dots, \min(k, n)$$

In this case, we say that X has a hyper-geometric distribution with parameters N, k, n and write $X \sim HG(N,k,n)$

 $\frac{Characteristics \ of \ hyper-geometric \ distribution}{X \sim HG \ (N, k, n)}$

(i) Mean

$$E(X) = n p = n \left(\frac{k}{N}\right)$$

(ii) Variance

$$Var(X) = npq = n\left(\frac{k}{N}\right)\left(1 - \frac{k}{N}\right)\left(\frac{N-n}{N-1}\right)$$

<u>Example (12)</u>

From a group of 6 men and 4 women, a random sample of size 5 persons is selected without replacement. Let X denotes the number of men in the sample.

It is clear that X has HG(10, 6, 5). The PMF of X is

$$f(x) = P(X = x) = \frac{\binom{6}{x}\binom{4}{5-x}}{\binom{10}{5}}, x = 1,...,5$$

We have

(i) P(2 men are selected) = P(X=2) = f(2) = 0.238(ii) P(selecting 2 women) = P(X=3) = f(3) = 0.476(iii) P(less than 3 men) = f(1)+f(2)= 0.024 + 0.238= 0.262

• Example (13) :

Calculate the mean, variance, and standard deviation of X in Example (12).

Solution

$$E(X) = n p = n \left(\frac{k}{N}\right) = 3$$

$$Var(X) = npq = n\left(\frac{k}{N}\right)\left(1 - \frac{k}{N}\right)\left(\frac{N-n}{N-1}\right) = 0.667$$

Standard deviation = σ = 0.816