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Subject- Mathematical Foundation of Computer Science

Topics covered:

- Hypergeometric Distribution
- Geometric Distribution and Negative Binomial (Pascal) Distribution
- Gamma Distribution

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Subject :- Mathematical foundation of Comp. Science
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Topic :- Hypergeometric Distribution

Definition :- In probability theory and statistics, the hypergeometric distribution is a discrete probability distribution that describes the probability of successes in draws, without replacement.

Hypergeometric Distribution characteristics :-

- There are only 2 possible outcomes. — success or failure
- The probability of a success is not the same on each trial without replacement, thus events are not independent
- In which population is finite
- Trials are dependent.
- It is similar to the binomial distribution but the difference is the method of sampling.
- X is the no. of successes in the n trials
- It is discrete distribution, the no. of objects in the population, N , is finite & known.

Probability function

N = population size

n = sample size

A = no. of success in population

x = no. of success in sample

$$P(x) = \frac{(A C x)(N-A C n-x)}{N C n}$$

mean value =
$$\mu = \frac{A \cdot n}{N}$$

Variance & Standard Deviation

(2)

$$r^2 = \frac{A(N-A)n(N-n)}{N^2(N-1)}$$

$$r = \sqrt{r^2}$$

Ex 1

3 light bulbs were selected from 10. ~~one~~ of the 10, there were 4 defective. What is the probability that 2 of the 3 selected are defective?

Sol

$$P(x) = \frac{{}^A C_x {}^{N-A} C_{n-x}}{{}^N C_n} = P(2) = \frac{{}^4 C_2 {}^6 C_1}{{}^{10} C_3}$$

$$P(x=2) = .30$$

Ex 2

Suppose that a shipment contains 5 defective items and 10 non defective items. If 7 items are selected at random without replacement, what is the probability that at least 3 defective items will be obtained?

Sol $N=15$ (5=defective, 10=non defective) $n=7$

$$P(x \geq 3) = 1 - P(x \leq 2) = 1 - [P(0) + P(1) + P(2)] = 0.4267$$

$$P(0) = \frac{{}^5 C_0 \cdot {}^{10} C_7}{{}^{15} C_7} = 0.0186$$

$$P(1) = \frac{{}^5 C_1 \cdot {}^{10} C_6}{{}^{15} C_7} = 0.1631$$

$$P(2) = \frac{{}^5 C_2 \cdot {}^{10} C_5}{{}^{15} C_7} = 0.3916$$

Ex 3

If a random variable X has a hypergeometric Distribution with parameter $A=8$, $B=20$ & n , for what value of n will $\text{Var}(X)$ be maximum?

$$\text{Var}(x) = \frac{n \cdot A \cdot B}{(A+B)^2} \cdot \frac{A+B-n}{A+B-1} = \frac{n \cdot 8 \cdot 20}{(8+20)^2} \cdot \frac{8+20-n}{8+20-1}$$

$$= \frac{160n}{28^2} \cdot \frac{(28-n)}{27} = 0$$

$n=28$, or $n=0$ for variance to be maximum

use case of hypergeometric Distribution in real life

- case → considers the situation in a factory where around 150 parts are made everyday. you supply these parts in boxes of 500 parts every week (so, lot size is 500).
- It is known that 2% of parts produced are defective. Your customer inspects each box by choosing 25 parts randomly one by one from a box
- then he records the no. of defective parts in his sample of 25 parts. The no. of defective parts out of 25 inspected parts can be analyzed through a Hypergeometric distribution

This distribution is used in all such cases where objects are chosen out of a finite lot size with a known no. of defects and the sample parts chosen without replacement.

Topic:- Geometric and Negative Binomial Distribution (Pascal Distribution)

- The geometric distribution is a special case of the negative Binomial distribution
- A Negative Binomial Experiment has following properties
 - The experiment consists of x repeated trials
 - each trial results in 2 possible outcomes $\left\{ \begin{array}{l} \text{success} \\ \text{failure} \end{array} \right.$
 - The probability of success, denoted by P , is the same on every trial
 - The trials are independent, that is, the outcome on one trial does not affect the outcome on other trials
 - the experiment continues until r successes are observed, where r is specified in advance

consider the following statistical experiment (use case)

→ you flip a coin repeatedly and the count the no. of times the coin lands on heads. You continue flipping the coin until it has landed 5 times on heads. This is a ~~ge~~ negative binomial experiment because :-

- The experiment consists of repeated trials we flip a coin repeatedly until it has landed 5 times on heads
- each trial can result in just two possible outcomes $\begin{matrix} \text{Head} \\ \text{Tails} \end{matrix}$
- The probability of success is constant = 0.5 on every trial
- The trials are independent that is, getting heads on one trial does not affect whether we get heads on other trials
- The experiment continues until a fixed no. of success have occurred, in this case, 5 heads

Notation and formula

A negative binomial random variable is the no. X of repeated trials to produce r success in a negative binomial experiment. The probability distribution of a negative binomial random variable is called a negative binomial distribution, also known as pascal distribution.

suppose a negative binomial experiment consists of x trials, result in r success. If the probability of success on an individual trial is p , then the negative binomial probability

is

$$b^*(x; r, p) = \binom{x-1}{r-1} p^r (1-p)^{x-r}$$

x = no. of trials required to produce r success

r = no. of successes

p = Prob. of success on an individual trial

q = The prob. of failure " " = $(1-p)$

$b^*(x; r, p)$ = Negative Binomial Probability - The probability that an x -trial negative binomial experiment results in the r th success on the x th trial, when success prob is p .

nCr :- the no. of combinations of n things, taken r at a time

The mean of Negative Binomial Distribution :-

If we define the mean of the negative binomial distribution as the average no. of trials required to produce r successes, then the mean is equal to

$$\mu = r/p$$

μ = mean no. of trials

r = no. of successes

p = Prob. of a success on any given trial

→ if no. of successes before the binomial experiment results in k failures, the mean of R is :-

$$\mu_R = k p / q$$

→ the no. of failures before the binomial experiment results in r successes, the mean of k is

$$\mu_k = r q / p$$

→ Geometric Distribution

The geometric distribution is a special case of the negative binomial distribution. It deals with the no. of trials required for a single success. Thus the geometric distribution is negative binomial distribution where the no. of successes (r) is equal to 1.

Geometric Probability formula

suppose a negative binomial experiment consists of x trials and results in one success. If the probability of success on an individual trial is p , then the geometric probability is

$$g(x; p) = p * q^{x-1}$$

Ex

Bob is a high school basketball player. He is a 70% free throw shooter. During the season, what is the probability that Bob makes his third free throw on his 5th shot?

To solve this,
solution

— This is an example of a negative binomial experiment.

$$\text{Success } (P) = 0.7$$

$$\text{no of trials } (x) = 5$$

$$\text{no of success } (r) = 3$$

So, Acc to negative binomial formula

$$b^*(x; r, P) = {}^{x-1}C_{r-1} \times P^r \times Q^{x-r}$$

$$b^*(5; 3, 0.7) = {}^4C_2 \times (0.7)^3 \times (0.3)^2$$

$$b^*(5; 3, 0.7) = 6 \times 0.343 \times 0.09$$
$$= 0.18522$$

thus, the probability that Bob will make his third successful free throw on his fifth shot is $= 0.18522$

Q2. Let's consider question 1, slightly different question, what is the probability that Bob makes his first free throw on his 5th shot?

→ This is an example of a geometric distribution which is a special case of negative binomial distribution

$$\text{Prob. of success } (P) = 0.7$$

$$\text{no. of trials } (x) = 5$$

$$\text{no of success } (r) = 1$$

Acc to geometric formula,

$$g(x; P) = P \times Q^{x-1}$$

$$g(5; 0.7) = 0.7 \times (0.3)^{5-1}$$

$$= 0.7 \times (0.3)^4$$

$$= \underline{\underline{0.00567}}$$

Topic — Gamma Distribution

The gamma distribution is a family of right skewed, continuous probability distributions. These distributions are useful in real-life where something has a natural minimum of 0. For example, it is commonly used in finance for elapsed times or during poisson processes.

gamma Distribution PDF

If x is a continuous random variable, then the probability distribution function is

$$P(x) = \frac{x^{\alpha-1} e^{-x/\beta}}{\beta^{\alpha} \Gamma(\alpha)} \quad x > 0$$

where

$$\Gamma(n) = \int_0^{\infty} t^{n-1} e^{-t} dt \quad \Gamma(x) = \text{the gamma function,}$$

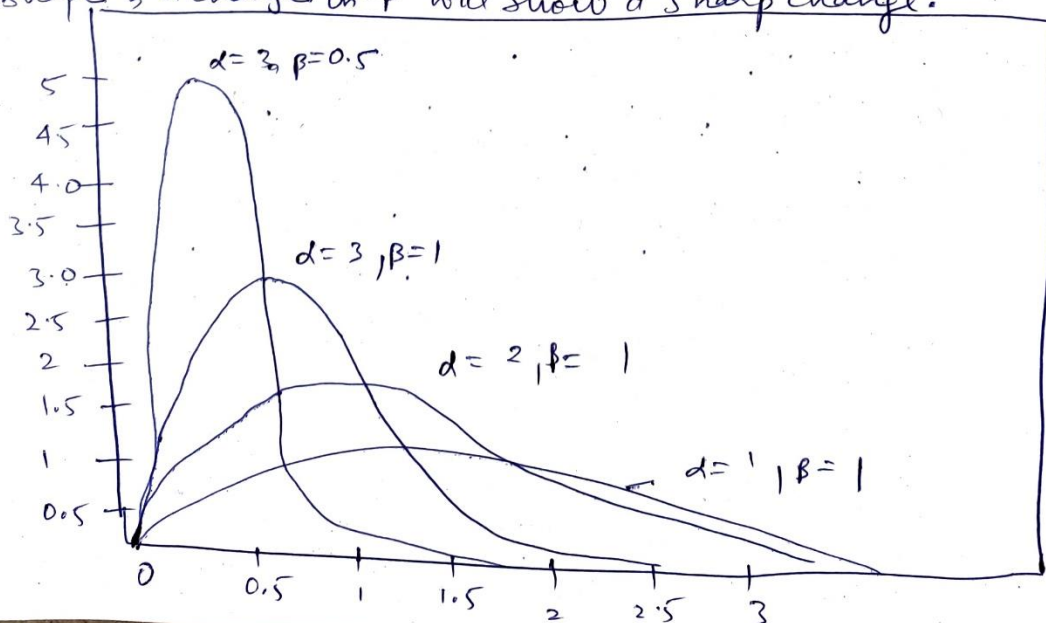
α = the shape parameter

β (sometimes θ is used instead) = The rate parameter (the reciprocal of the scale parameter)

When, α & β are both greater than 1

When $\alpha = 1$, this becomes the exponential distribution
When $\beta = 1$, this becomes the standard gamma distribution

α & β define the shape of graph, although they both have an effect on the shape, a change in β will show a sharp change.



$$\text{mean} = E(X) = \alpha\beta$$

$$\text{variance} = \text{var}(X) = \alpha\beta^2$$

$$\text{moment generating function} = M_X(t) = \frac{1}{(1-\beta t)^\alpha}$$

• Important properties of gamma function

- from definition of the integral, $\Gamma(1) = 1$
- for any $\alpha > 0$, $\Gamma(\alpha) = (\alpha-1)\Gamma(\alpha-1)$
- for any positive integers n , $\Gamma(n) = (n-1)!$ for $n=1, 2, 3, \dots$
- $\Gamma(\frac{1}{2}) = \sqrt{\pi}$
- $\Gamma(\alpha+1) = \alpha\Gamma(\alpha)$
- $\int_0^\infty x^{\alpha-1} e^{-\lambda x} dx = \frac{\Gamma(\alpha)}{\lambda^\alpha}$, for $\lambda > 0$;

Q. claim amounts on a certain type of policy are modelled as following a gamma distribution with parameters $\alpha=120$ & $\lambda=1.2$. calc. an approximate value for the prob. that an individual claim amt exceeds 120, giving a reason for the approach you use?

$$E(X) = \frac{\alpha}{\lambda} = \frac{120}{1.2} = 100 \rightarrow \text{variance} = \text{var}(X) = \frac{\alpha}{\lambda^2} = \frac{120}{(1.2)^2} = 83.33$$

$$P(X > 120) = P\left(Z > \left[\frac{120 - 100}{\sqrt{83.33}}\right]\right) \left\{ X \sim N(100, 83.33) \right\}$$

$$= P(Z > 2.191)$$

$$= 1 - 0.986 = 0.014$$

Example — Imagine you are solving difficult maths theorems, & you expect to solve one every $\frac{1}{2}$ hr. compute the probability that you will have to wait b/w 2 to 4 hrs before you solve four of them.

Sol. one theorem every $\frac{1}{2}$ hr means we would suppose to get $\theta = \frac{1}{0.5} = 2$ theorems every hour on any day. $\theta = 2$ & $K = 4$.

$$P(2 \leq X \leq 4) = \sum_{x=2}^4 \frac{x^{\theta-1} e^{-\theta x/2}}{\Gamma(4) 2^4} = 0.12388$$