

Weekly presentation SCM research

4th presentation: Special distributions



Summary

I/Introduction

II/Poisson distribution

- 1) *Definition*
- 2) *Properties*
- 3) *Binomial approximation*
- 4) *Poisson process*

III/ Normal distribution

- 1) *Definition*
- 2) *Properties*
- 3) *Central limit theorem*



I/ Introduction

- An incredible variety of special distributions have been studied over the years, and new ones are constantly being added to the literature
- two general classes of distributions: location-scale families and exponential families
- Discrete/ Continuous families
- Special properties



II/ Poisson distribution

1) Definition

- Discrete probability distribution that expresses the probability of a given number of events occurring in a fixed interval of time and/or space if these events occur with a known average rate and independently of the time since the last even
- Can also be used for the number of events in other specified intervals such as distance, area or volume
- Useful approximation to binomial distributions with very small success probabilities



I/ Poisson distribution

1) Definition

- Probability function

Random variable with discrete distribution, X is said to have a Poisson distribution with mean λ ($\lambda > 0$) if

$$f(x/\lambda) = \begin{cases} \frac{e^{-\lambda} \lambda^x}{x!} & \text{for } x = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$



II/ Poisson distribution

1) Definition

- Mean and Variance

λ is both the mean and the variance of the Poisson distribution

- Proof

$$\begin{aligned} E(X) &= \sum_{x=1}^{\infty} x f(x/\lambda) = \sum_{x=1}^{\infty} x \frac{e^{-\lambda} \lambda^x}{x!} = \lambda \sum_{x=1}^{\infty} x \frac{e^{-\lambda} \lambda^{x-1}}{x-1!} \\ &= \lambda \sum_{y=0}^{\infty} \frac{e^{-\lambda} \lambda^y}{y!} = \lambda \end{aligned}$$



II/ Poisson distribution

1) Definition

$$E(X(X-1)) = \sum_{x=2}^{\infty} x(x-1)f(x/\lambda) = \lambda^2$$

$$E(X^2) = E(X(X-1)) + E(X) = \lambda^2 + \lambda$$

$$Var(X) = E(X^2) - E(X)^2 = \lambda$$

•Moment Generating function

$$\psi(t) = E(e^{tX}) = e^{-\lambda} \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x}{x!} = e^{\lambda(e^t-1)}$$



II/ Poisson distribution

2) *Properties*

- If the random variables $X_1 \dots X_k$ are independent and if H_i has a Poisson distribution with mean λ_i then the sum $X_1 + \dots + X_k$ has a Poisson distribution with mean $\lambda_1 + \dots + \lambda_k$

Proof using the moment generating function

• **Example (customer arrivals)**

- The number of customer who arrive in disjoint hours can be represented by independent random variable having a poisson distribution with mean 4.5
- *Probability that at least 12 customers arrive in a two-hour period?*
- Using the property the total number $X = X_1 + X_2$ has a Poisson distribution with mean 9
- **$P(X \geq 12) = 0; 1970$**



II/ Poisson distribution

3) *Binomial approximation*

•If the value of n is large and the value of p is close to 0, the binomial distribution with parameters n and p can be approximated by a Poisson distribution with mean np

•Proof

$$f(x/n, p) = \frac{n(n-1) \dots (n-x+1)}{x!} p^x (1-p)^{n-x}$$

$$f(x/n, p) = \frac{\lambda^x}{x!} \frac{n}{n} \frac{n-1}{n} \dots \frac{n-x+1}{n} \left(1 - \frac{\lambda}{n}\right)^{-x} \left(1 - \frac{\lambda}{n}\right)^n$$

$$f(x/n, p) \rightarrow \frac{e^{-\lambda} \lambda^x}{x!}$$



II/ Poisson distribution

3) *Binomial approximation*

•Example

- In a large population the proportion of people that have a certain disease is 0.01
- Probability that in a group of 200 at least 4 of them have the disease?*
- Binomial distribution with parameters $n=200$ and $p = 0.01$
- Approximated by a Poisson distribution with mean 2
- Reading on the Poisson tables, **$P(X \geq 4) = 0.1428$**
- The actual value is 0.1420



II/ Poisson distribution

4) *Poisson process*

- A poisson process with rate λ per unit time is a process that satisfies:

- *The number of arrivals in every fixed interval of time of length t has a Poisson distribution for which the mean is λt*
- *The numbers of arrivals in every two disjoint time intervals are independent*

- Poisson processes are more general (particles emitted from a radioactive source, defects on the surface of a manufactured product...)

- Popular process

- *computationnaly convenient*
- *Mathematical justification*



III/ Normal distribution

1) Definition

- Most widely used model for random variables with continuous distribution and single most important distribution in statistics
 - *Mathematical convenience*
 - *Various physical experiments often have distributions that are approximately normal*
 - *For a large random sample from any distribution that has a finite variance, the distribution of the sample will be approximately normal (consequence of the central limit theorem)*
- It is said that a random variable X has a normal distribution with mean μ and variance σ^2 if:

$$f(x/\mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

- The p.d.f. cannot be integrated in closed form (hence d.f. tables are used in computer programs)



III/ Normal distribution

1) Definition

- Moment generating function

$$\psi(t) = E(e^{tX}) = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{\left[tx - \frac{(x-\mu)^2}{2\sigma^2}\right]} dx$$

$$\psi(t) = e^{\left(\mu t + \frac{1}{2}\sigma^2 t^2\right)}$$

- Mean and Variance

$$E(X) = \psi'(0) = \mu$$

$$\text{Var}(X) = \psi''(0) - [\psi'(0)]^2 = \sigma^2$$

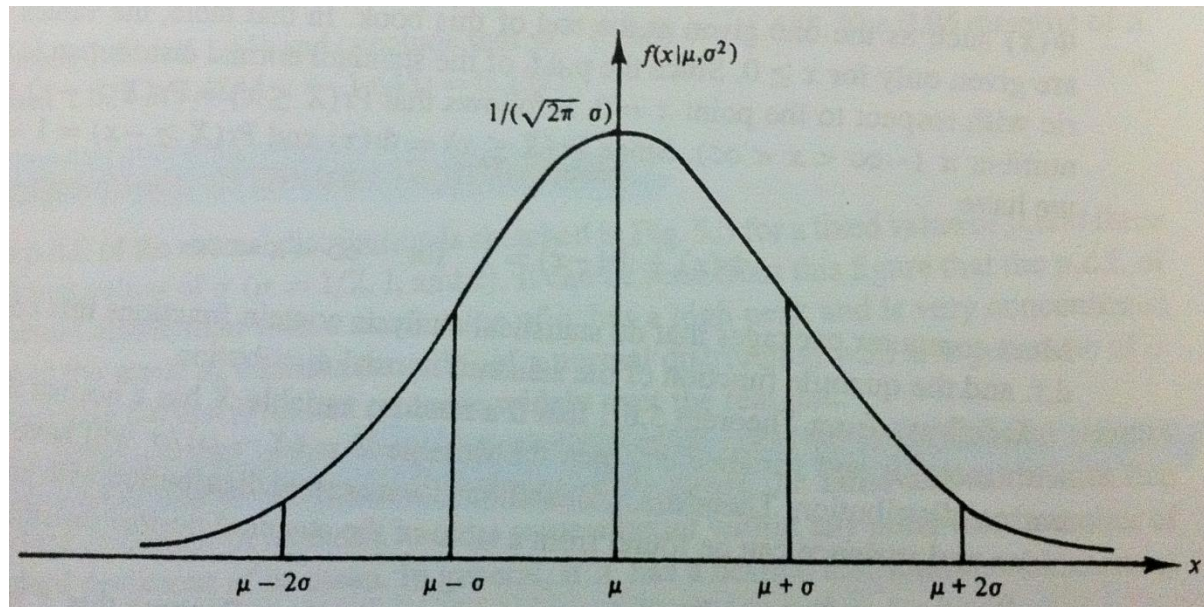


III/ Normal distribution

2) *Properties*

- Shape

- μ is the mode of the distribution



III/ Normal distribution

2) Properties

- Linear transformation

If X has a normal distribution with mean μ and variance σ^2 then $Y=aX+b$ (with $a \neq 0$) has a normal distribution with mean $a\mu+b$ and variance $a^2\sigma^2$ + generalization

- The normal distribution with mean 0 and variance 1 is called the standard normal distribution

$$\phi(x) = f\left(\frac{x}{0,1}\right) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2}$$

$$\Phi(x) = \int_{-\infty}^x \phi(u) du$$



III/ Normal distribution

2) Properties

Example (heights of men and women)

- women heights follow a normal distribution with $\mu=65$ and $\sigma=1$
- men heights follow a normal distribution with $\mu=68$ and $\sigma=2$

Probability that a randomly selected woman will be taller than a randomly selected man?

W, height of the woman

M, height of the man

The difference $W-M$ follows a normal distribution with mean $65-68=-3$ and variance $1^2+2^2=5$

We define
$$Z = \frac{W - M + 3}{\sqrt{5}}$$

It follows from the properties that Z has a standard normal distribution

$$P(W > M) = P(W - M > 0) = P(Z > 1.342) = 1 - \Phi(1.342) = 0.090$$

III/ Normal distribution

3) Central Limit Theorem

•Theorem

Whenever a random sample of size n is taken from ANY distribution with mean μ and variance σ^2 , the sample mean \bar{X}_n will have a distribution that is approximately normal with mean μ and variance σ^2/n

If the random variables X_1, \dots, X_n from a random sample of size n from a given distribution with mean μ and variance σ^2 , then for each fixed number x :

$$\lim_{n \rightarrow \infty} P \left[\frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma} \leq x \right] = \Phi(x)$$

$$\frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma}$$

will be approximately a standard normal distribution



III/ Normal distribution

3) Central Limit Theorem

- Theorem effects

- *Provides a plausible explanation for the fact that the distributions of many random variables studied in physical experiments are approximatively normal*

- *The distribution of the sum of many random variables can be approximatively normal even though the distribution of each random variable in the sum differs from the normal*



***Thank you for your
attention!***

