

Probability Distributions & Simulation

Probability distributions

A **probability distribution** provides a comprehensive description of how probabilities are assigned to various outcomes of a random experiment with all the probabilities summing upto 1.

Two major types:

Discrete distributions: take countable values (e.g., Binomial, Poisson)

Continuous distributions: take values over an interval (e.g., Normal)

R provides a rich set of built-in functions for probability, simulation, and visualization.

Probability mass function and probability density function

Probability mass function is the probability distribution of a discrete random variable, and provides the possible values and their associated probabilities denoted by $p_X(x)=P(X=x)$.

Probability density function is the probability density function of a continuous random variable and is denoted by $f(x)$. The probability is given by the integral of the pdf over that range.

Binomial distribution

Binomial distribution is a discrete distribution with parameters n and p

n - no of independent Bernoulli trials

p -probability of success

Binomial distribution models no of successes and the probability mass function is

$$P(X = x) = nC_x p^x q^{n-x}, x = 0, 1, 2, \dots n$$

Examples

1. A fair coin is tossed **10 times**. Find the probability of getting **exactly 6 heads**.
2. A die is thrown 12 times. Find the probability of getting exactly 4 sixes.
3. If 40% of emails are spam, find the probability that exactly 5 out of 10 emails are spam.

Code with examples

$P(X = x)$

```
dbinom(x=3 , size=10 , prob=0.4)
```

x denotes no of successes, size=size of the sample, prob is the probability of success

$P(X \leq x)$: to find cumulative probability

```
sum(dbinom(0:3, 10 , 0.4))
```

```
pbinom(3, 10,0.4)
```

```
sum(dbinom(0:10, size = 10, prob = 0.4))
```

```
pbinom(10, 10,0.4) to check sum of prob is 1
```

To find the value of x when probability is given

```
qbinom(0.95, size = 10, prob = 0.4)
```

Generates random samples from Binomial distribution

`rbinom(n = 10, size = 10, prob = 0.4)` generates 10 random values with probability of success 0.4 in between 0 to 10(size of the distribution)

`x<-rbinom(n = 10, size = 10, prob = 0.4)` (assigning to x)

x

Plotting in Binomial distribution

```
x <- 0:10
```

```
y <- dbinom(x, size = 10, prob = 0.4)
```

```
barplot(y, names.arg = x, main = "Binomial Distribution  
(n=10, p=0.4)", xlab = "x", ylab = "P(X = x)")
```

```
plot(x,y,main="Binomial",xlab = "x", ylab = "P(X = x)")
```


Comparison between exact & simulated value

`dbinom(3, 10, 0.4)`----Probability of getting exactly 3 successes out of 10 trials.

Assuming `x` is generated using `rbinom`

`x==3` creates a logical vector

`mean(x == 3)` proportion of times the value 3 occurs in the simulation

First one is the exact probability and the second one is the approximate probability from simulation

Simulation using rbinom

```
n <- 1000
```

```
x <- rbinom(n, size=10, prob=0.4)
```

```
table(x)/n
```

Poisson distribution

The **Poisson distribution** is a discrete probability distribution that expresses the probability of a given number of events occurring in a fixed interval of time with constant mean.

The probability mass function is

$$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}, x = 0, 1, 2, 3, \dots$$

$\lambda = np$ is the mean of the distribution, $e = 2.71828\dots$

n is very large

p is very small

$\lambda = np$ is finite and a constant

Mean=Variance= $\lambda = np$

Examples

- Number of accidents at a traffic junction per day
- Number of defects in a roll of fabric
- Number of customers arriving at a bank per minute

Coding with examples

There are 4 standard functions in R for Poisson

`dpois(x,lambda)`---- for $P(X = x)$

`ppois(x,lambda)`----for $P(X \leq x)$

`dpois(3, lambda = 2)`

`ppois(3, lambda = 2)` or `sum(dpois(0:3,lambda=2))`

`qpois(0.95,2)`

Generating values

`rpois(5, lambda = 2)` – generates 5 values

`set.seed(123)` ---- This gives the same output everytime (We can use any integer inside seed)

```
rpois(5, lambda = 2)
x <- rpois(10, lambda = 2)
X
```

Note: `rpois` command is used for simulation (generating values)

Plotting in Poisson distribution

```
x <- 0:10
```

```
prob <- dpois(x, lambda = 3)
```

```
barplot(prob, names.arg = x, xlab = "Number of  
occurrences (x)", ylab = "Probability", main = "Poisson  
Distribution ( $\lambda = 3$ )") # creates a barplot
```

To create an usual plot

```
plot(x,prob, xlab = "Number of occurrences (x)", ylab =  
"Probability",main = "Poisson Distribution ( $\lambda = 3$ )")
```

To create a histogram

```
plot(x, prob, type = "h", lwd = 2, xlab = "x", ylab = "P(X = x)",  
     main = "Poisson Distribution ( $\lambda = 3$ )")
```

```
points(x, prob, pch=16)
```

To create a line plot

```
plot(x, prob, type = "l", lwd = 2, xlab = "x", ylab = "P(X = x)",  
     main = "Poisson Distribution ( $\lambda = 3$ )")
```

```
points(x, prob, pch = 16)
```

To plot only points

```
plot(x, prob, type = "p", pch = 16, xlab = "x", ylab = "P(X = x)",  
     main = "Poisson Distribution ( $\lambda = 3$ )")
```

#lwd is linewidth

Normal distribution

Normal distribution is a continuous distribution

with pdf $f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, -\infty < x < \infty$

μ is the mean of the distribution

σ is the standard deviation of the distribution

Key properties:

- Symmetric, bell-shaped curve
- Mean = Median = Mode
- Total area under curve = 1

Standard normal variate

Definition

If

$$X \sim N(\mu, \sigma^2)$$

then the **standard normal variate** Z is defined as

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

Here,

- Mean of $Z = 0$
- Standard deviation of $Z = 1$

So,

$$Z \sim N(0, 1)$$

Suppose the heights of students are normally distributed with $\mu = 170$ cm and $\sigma = 10$ cm.

Find the standard normal variate for a student of height **185 cm**.

$$Z = \frac{185 - 170}{10} = 1.5$$

This means the height is **1.5 standard deviations above the mean**.

Examples

Problem 1

The mean and standard deviation of a normal distribution are 50 and 10 respectively.

Find $P(X < 60)$.

Problem 2

If $X \sim N(100, 15^2)$, find the value of $P(X > 130)$.

Problem 3

The heights of students are normally distributed with mean 165 cm and standard deviation 8 cm.

Find the probability that a student's height lies between 160 cm and 175 cm.

Coding with examples

There are 4 standard functions in R for Normal distribution

`dnorm(55, mean = 50, sd = 10)`—it gives the value of $f(55)$

`pnorm(60, mean = 50, sd = 10)`—gives the cumulative prob till $x=60$ (ie) $P(X \leq x)$ left tail probability

`qnorm(0.8413447, mean = 50, sd = 10)`— it gives the value of x when cumulative prob is known

Right tail probability (ie) $P(X > x)$

`1-pnorm(60, mean = 50, sd = 10)`

(or)

`pnorm(60, mean = 50, sd = 10, lower.tail = FALSE)`

To find $P(40 < X < 60)$

`pnorm(60, mean = 50, sd = 10) - pnorm(40, mean = 50, sd = 10)`

Generating values

`rnorm(5,mean=0,sd=1)`—generates 5 random values in $-\infty < x < \infty$.

`set.seed(123)` will generate same set of values

`rnorm(5,mean=0,sd=1)`

Plotting in normal distribution

Basic normal curve

```
x <- seq(-4, 4, length = 100)
```

```
y <- dnorm(x, mean = 0, sd = 1)
```

```
plot(x, y, type = "p", pch = 16, xlab = "x", ylab = "f(x)",  
     main = "Normal Distribution (Points)")
```

Plot using points

```
plot(x, y, type = "p", pch = 16, xlab = "x", ylab = "f(x)",  
     main = "Normal Distribution (Points)")
```

Plot using both

```
plot(x, y, type = "b", pch = 16, lwd = 2, xlab = "x", ylab =  
"f(x)", main = "Normal Distribution (Line + Points)")
```

Plotting Normal distribution from simulated data

To plot histogram

```
set.seed(123)
```

```
data <- rnorm(1000, mean = 50, sd = 10)
```

```
hist(data, probability = TRUE, xlab = "x", main =  
"Histogram of Normal Data")
```

```
curve(dnorm(x, mean = mean(data), sd = sd(data)),  
      add = TRUE, lwd = 2)
```



```
x <- seq(-10, 10, length = 200)
plot(x, dnorm(x, 0, 1), type = "l", lwd = 2,
      xlab = "x", ylab = "Density",
      main = "Normal Distribution for Different
Parameters")
```

Code to draw a normal curve and to shade $P(40 < X < 60)$

```
mu <- 50;sigma <- 10;a <- 40;b <- 60
x <- seq(mu - 4*sigma, mu + 4*sigma, length = 1000)
y <- dnorm(x, mean = mu, sd = sigma)
plot(x, y, type = "l", lwd = 2, main = "Normal Distribution with
Shaded Region", xlab = "x", ylab = "Density")
x_shade <- seq(a, b, length = 500)
y_shade <- dnorm(x_shade, mean = mu, sd = sigma)
polygon(c(a, x_shade, b),c(0, y_shade, 0),col = "lightblue",
border = NA)
abline(v = c(a, b), col = "red", lwd = 2, lty = 2)
prob <- pnorm(b, mu, sigma) - pnorm(a, mu, sigma)
prob
```

Questions for practice- Binomial and Poisson distributions

1. A coin is tossed 8 times. Let X be the number of heads.
Find a) $P(X = 5)$ b) $P(X \geq 6)$
2. The probability that a student passes an exam is 0.7. If 5 students appear, find $P(\text{exactly 3 pass})$ and $P(\text{at least 4 pass})$
3. Write an R program to find and plot the binomial distribution for $n=10$, $p=0.5$.
4. The number of typing errors per page follows a Poisson distribution with mean 2. Find the probability that a page contains a) exactly 1 error b) at least 3 errors.
5. Write an R program to find $P(2 < X \leq 4)$ for a Poisson distribution with mean 4.

Questions for practice-Normal distribution

1. The heights of students are normally distributed with mean 165 cm and standard deviation 6 cm.
Find the probability that a randomly chosen student has height
 - (a) less than 160 cm
 - (b) between 160 cm and 175 cm.
2. Let $X \sim N(60, 8^2)$. Find $P(52 < X < 68)$.
3. For a standard normal variable Z , find
 - a) $P(-1.5 < Z < 2.0)$
 - b) $P(Z > 1.96)$
4. Write an R program to generate normal dat, compute probabilities and visualize shaded regions for different values of a and b.