

## PRACTICAL 5 (Continuous probability distribution)

This practical introduces the fundamental concepts of probability density functions (PDFs) and the normal distribution. It begins with verifying whether a function qualifies as a valid PDF, followed by practical applications for calculating probabilities. It also explores the properties of the normal distribution.

### Part 1 : Checking if f(x) is a pdf

Check if the following function is a pdf,

$$f(x) = \begin{cases} \frac{1}{2}e^{-x/2} & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

Steps:

1. Define the function
2. Check if the function is non negative on the domain
3. Check if the integral of the function over the domain is unity.

# Define the custom PDF

```
pdf_custom <- function(x) {  
  if (x >= 0) {  
    return((1/2) * exp(-x/2))  
  } else {  
    return(0)  
  }  
}
```

# Vectorize the PDF function, this is used to convert a scalar function into a vectorized  
# function so that it can accept and process vectors as inputs.

```
pdf_custom <- Vectorize(pdf_custom)
```

# Generate a sequence of values for x

# You can choose an appropriate range based on the domain of your PDF

```
x_values <- seq(-10, 10, length.out = 1000)
```

# Evaluate the PDF over the sequence of x values

```
pdf_values <- pdf_custom(x_values)
```

# Check for non-negativity

# The all() function in R is used to check whether all elements of a logical vector are TRUE.

```
if (all(pdf_values >= 0)) {
```

```
  print("The function is non-negative over the specified domain.")
```

## PRACTICAL 5 (Continuous probability distribution)

```
} else {  
  print("The function is negative at some points in the specified domain.")  
}  
# Integrate the PDF from 0 to infinity  
result <- integrate(pdf_custom, lower = 0, upper = Inf)  
  
# Print the result  
print(paste("The integral of the PDF from 0 to infinity is:", result$value))
```

### Part 2 : finding probability

Suppose that in a certain region the daily rainfall (in inches) is a continuous RV  $X$  with p.d.f  $f(x)$  given by

$$f(x) = \frac{3}{4}(2x - x^2), 0 < x < 2 \text{ \& } f(x) = 0 \text{ elsewhere}$$

Find the probability that on a given day in this region, the rainfall is:

- i. not more than 1 inch
- ii. greater than 1.5 inches
- iii. between 0.5 and 1.5 inches

```
# Define the PDF
```

```
pdf_rainfall <- function(x) {  
  if (x >= 0 && x <= 2) {  
    return(3/4 * (2*x - x^2))  
  } else {  
    return(0)  
  }  
}
```

```
# Vectorize the PDF function
```

```
pdf_rainfall <- Vectorize(pdf_rainfall)
```

```
# Integrate the PDF from 0 to 1
```

```
prob_not_more_than_1 <- integrate(pdf_rainfall, lower = 0, upper = 1)$value  
print(paste("Probability that rainfall is not more than 1 inch:", prob_not_more_than_1))
```

```
# Integrate the PDF from 1.5 to 2
```

```
prob_greater_than_1_5 <- integrate(pdf_rainfall, lower = 1.5, upper = 2)$value  
print(paste("Probability that rainfall is greater than 1.5 inches:", prob_greater_than_1_5))
```

```
# Integrate the PDF from 0.5 to 1.5
```

```
prob_between_0_5_and_1_5 <- integrate(pdf_rainfall, lower = 0.5, upper = 1.5)$value
```

## PRACTICAL 5 (Continuous probability distribution)

```
print(paste("Probability that rainfall is between 0.5 and 1.5 inches:",  
prob_between_0_5_and_1_5))
```

### Part 3: Normal Distribution

#### Normal Distribution

The normal distribution, also known as the Gaussian distribution, is one of the most important probability distributions in statistics. It is symmetric, bell-shaped, and is defined by its mean ( $\mu$ ) and standard deviation ( $\sigma$ ). The normal distribution describes many natural phenomena, such as heights, test scores, and measurement errors.

- ✓ Mean ( $\mu$ ): The average or central value of the distribution.
- ✓ Standard Deviation ( $\sigma$ ): A measure of the spread or dispersion of the distribution.
- ✓ Z-score: A standardized value that indicates how many standard deviations a data point is from the mean.
- ✓ Standard Normal Distribution: A special case of the normal distribution where the mean is 0 and the standard deviation is 1.

#### R Functions for Normal Distribution

##### 1. `dnorm()` - Probability Density Function

The `dnorm()` function returns the height of the probability density function for a given Z-score, mean, and standard deviation, using the syntax **`dnorm(Z, mean, sd)`**.

##### 2. `pnorm()` - Cumulative Distribution Function

The `pnorm()` function gives the cumulative probability up to a given Z-score, mean, and standard deviation, using the syntax **`pnorm(Z, mean, sd)`**.

##### 3. `qnorm()` - Quantile Function

The `qnorm()` function returns the Z-score (quantile) for a given cumulative probability, mean, and standard deviation, using the syntax **`qnorm(Z, mean, sd)`**.

#### Examples Using R

##### Example 1: Calculating Cumulative Probability

- ✓ To calculate the standard normal cumulative probability for  $Z = 1$ , use `pnorm(1)`.
- ✓ To calculate the cumulative probability for a normal distribution with a mean of 45 and a standard deviation of 10 at  $Z = 140$ , use `pnorm(140, 45, 10)`.

##### Example 2: Z-transformation with Real Data

Let's use the women dataset, which contains the height and weight of women.

## PRACTICAL 5 (Continuous probability distribution)

Start by loading the data and viewing the first few rows with

```
data("women")
```

```
head(women).
```

Calculate the mean and standard deviation of the height with

```
height.mean <- mean(women$height)
```

```
height.sd <- sd(women$height).
```

Find the probability of having height upto 68 cm

```
x <- 68
```

calculate the cumulative probability by specify the mean and standard deviation using

```
pnorm(x, mean = height.mean, sd = height.sd).
```

### Visualization: Plotting the Normal Distribution

Define the sequence of x values with

```
x <- seq(-10, 10, by = .1).
```

Calculate the corresponding y values for a normal distribution

```
y <- dnorm(x, mean = 2.5, sd = 0.5).
```

Plot the normal distribution

```
plot(x, y, type="l", col="blue").
```

### Standard Normal Distribution

Define the sequence of x values

```
x = seq(-4, 4, length = 1000).
```

Calculate the corresponding y values for a standard normal distribution

```
y = dnorm(x).
```

Plot the standard normal distribution

```
plot(x, y, type="l", col="red").
```

## Excercise

1. If  $f(x)$  is a density function then find i)  $c$ , ii)  $p(1 < x < 2)$

$$f(x) = \begin{cases} cx^2, & 0 < x < 3 \\ 0, & \text{otherwise} \end{cases}$$

2. For a continuous random variable  $x$ , following is the pdf. Find  $k$ , mean and variance.

$$f(x) = kx^2 e^{-x}, x \geq 0.$$

**PRACTICAL 5 (Continuous probability distribution)**

3. For a continuous random variable  $x$ , following is the pdf. find  $P(1/4 < x < 3/4)$

$$f(x) = \begin{cases} x^3, & 0 \leq x \leq 1 \\ 3(2-x)^3, & 1 \leq x \leq 2 \end{cases}$$

4. In a sample of 1000 cases, the mean of a certain test is 14 and standard deviation is 2.5. Assuming the distribution to be normal, find
- i. How many students score between 12 and 15?
  - ii. How many score above 18?
  - iii. How many score below 8?
- Write a R program for the above problem. Also plot the graph for each case.
5. In a male population of 1000, the mean height is 68.16 inches and standard deviation is 3.2 inches. How many men may be more than 6 feet (72 inches)? Write a R program for the above problem. Also plot the graph.