L.1: Intro to R and Review of Probability 1 Econometrics 1: ver. 2024 Fall Semester

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Introduction to R

What is R?

- **R** is a <u>free</u> and <u>open source</u> software for statistical analysis.
- R has no license limitations. You can install and run it anytime and anywhere.
- One of the greatest advantages of R over other softwares is that R users can freely distribute their own original packages through CRAN (Comprehensive R Archive Network, "KRAN" or "SEE-RAN").
- We can implement a wide variety of brand new statistical methods quite easily just by downloading them from CRAN.
- For now, R and Python are two of the most popular programming languages used in statistical analysis.¹





¹Python is a general purpose programming language, which is not specialized for statistical analysis but can be used for a variety of purposes.

How to install R on Windows

- Go to the website of R "The R Project for Statistical Computing": https://www.r-project.org.
- Click on the link download R.



The R Project for Statistical Computing

Getting Started

R is a free software environment for statistical computing and graphics. It compiles and runs on a wide variety of UNIX platforms, Windows and MacOS. Advantaged R. Dease choose your preferred CRAN mirror.

If you have questions about R like how to download and install the software, or what the license terms are, please read our answers to frequently asked questions before you send an email.

- Then, you will be asked to which server you want to connect. Choose "Japan The Institute of Statistical Mathematics, Tokyo"
- Click on the link Download R for Windows.

How to install R on Windows

Click on the link install R for the first time.



 Click on the link Download R X.X.X for Windows, where X.X.X gives the version of R.



Then, the installer will be downloaded as "R-X.X.X-win.exe".

How to install R on Windows

- Double-click the downloaded installer to launch the installer.
- Click "Next" several times.
- (Optional) At the page that says "Startup options", choose

"Yes (customized startup)"

The default choice is "No (accept defaults)".

• The next page says "Display Mode". Choose

"SDI (separate windows)"



• The other options can be left as defaults.

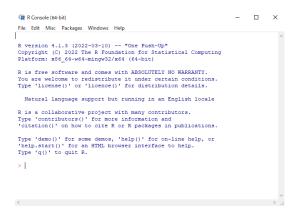
R-Studio

- R-Studio is a software that provides a more efficient and user-friendly programming environment for using R.
- It includes a code editor, debugging, visualization tools, and R markdown editor.²
- Although it is not mandatory, I would recommend using R-Studio.



 $^{^2}$ R markdown is a tool that allows you to integrate R codes and their outputs in a document file and presentation slides. For example, these lecture slides are created using R markdown.

- When you start R, the window that first appears is called the R console.
- You can type or paste commands here. The console window also displays the results of the commands and error reports (if any).



- Anything following a hash (sharp) sign # is ignored and it is not processed by R. This can be used to include comments.
- If you want to write more than one command in a single line, you can use a semicolon; as a command delimiter.

```
> 1 + 5 # addition
```

[1] 6

> 8 - 2 # subtraction

[1] 6

> 4 * 6 ; 3 / 7 # multiplication and division

[1] 24

[1] 0.4285714

• If you want to assign a number "a" to the variable "X", you can write

• The assign symbol consists of two separate characters < and -, "less than" and "minus" with no space between them.

```
> sqrt(5*(1 + exp(2)) + tan(0.5))
```

```
> X <- sqrt(5*(1 + exp(2)) + tan(0.5))
```

> X

[1] 6.518557

[1] 6.518557

> 2*X

[1] 13.03711

 Virtually any type of R objects (vector, matrix, data frames, functions, texts, etc) can be stored in a single object.

```
> A <- "Hoshino" # Texts must be enclosed in " ".
> A
```

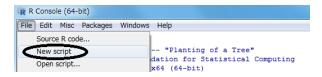
```
## [1] "Hoshino"
```

```
> A + 1 # You cannot add a number to a text.
```

Error in A + 1: non-numeric argument to binary operator

Script files

- When long and involved calculations are needed, typing each command directly into the console is inconvenient and error-prone.
- Besides, once the console window is closed all the commands you have executed will disappear
- A script file is a text file that contains a sequence of commands. You
 can execute the commands directly from the script file all at once.
- To create a new script file, click on "File" in the menu bar and then click "New Script".



Script file

- Once the commands are typed in the script, select the part you want to execute.
- Right click and choose "Run Line or Selection" or press [Ctrl] and [R] at the same time.³



- ullet rnorm(100): draw 100 random numbers from the standard normal distribution, N(0,1).
- hist(X): create a histogram of X (a new window will pop up).

 $^{^{3}}$ R-studio users: [Ctrl] + [Enter]

- > X <- rnorm(100)
- > hist(X)



Script file (cont.)

- To save the script file, click on "File" in the menu bar in the script editor, and then click "Save as". (The file extension is "r".)
- Note that double-clicking the script file does NOT open the R console window. To open the saved script file, launch R first, and choose "Open script" in the "File" in the menu bar.
- If you want to open the script file only, you can use any text editor like Notepad

If the .r-extension is associated with R-studio, you can launch R-studio just by double-clicking the ${\bf r}$ file.

- A random variable is a variable whose values are determined probabilistically.
- Examples of random variables are:
 - outcomes of rolling dice;
 - number of car accidents in a day;
 - income of a randomly sampled person.
- A more formal definition of random variable is as follows:

Random Variable

A random variable is a "function" that maps the outcomes of a random experiment to a numerical value.

- A random variable is a "rule" (i.e., function) that associates a number with each outcome in the set of possible outcomes.
- The set of possible outcomes is called the sample space.
 - Sample space is not necessarily a set of numbers.
- In the following, we use a capital letter, say X, to denote a random variable. A realized value of X is denoted by a small letter x.
- The set of possible realizations of X, $\{x_1, x_2, \dots, \}$, is called the support, and denoted by \mathcal{X} .

Random variable X	Sample space	Support $\mathcal X$
Outcome of rolling a dice	$\{1, 2, 3, 4, 5, 6\}$	$\{1, 2, 3, 4, 5, 6\}$
No. of heads when tossing a coin twice	$\{HH, TH, HT, TT\}$	$\{0, 1, 2\}$

Discrete Random Variable

A discrete random variable is a random variable where its support is a discrete set.

Continuous Random Variable

A continuous random variable is a random variable where its support is an interval (or a collection of intervals).

- Intuitively, a discrete random variable is a random variable whose possible values can be written down in a list.
- ullet For example, letting X denote the height (in meters) of a randomly selected person, then X is a continuous random variable.
 - ullet When X is continuous, there are infinitely many possible values in $\mathcal{X}.$

A dice roll simulation

- \bullet Dice roll outcome = Uniform random variable on $\mathcal{X} = \{1, 2, \dots, 6\}$
 - Each element of $\mathcal X$ occurs with equal probability, 1/6.

```
> #install.packages("extraDistr")
> library(extraDistr)
> rdunif(2, 1, 6)
```

```
## [1] 3 5
```

- install.packages(): install a new package. You need to run this code only once.
- We can use extraDistr package to compute discrete uniform distributions.
- library(): load the specified package.
- rdunif(2, 1, 6): draw two X's from Uniform $\{1, 2, ..., 6\}$

Cumulative Distribution Function: CDF

The probability that the random variable X takes a value less than or equal to x, $\Pr(X \leq x)$, is called the cumulative distribution function (CDF) or simply the distribution function of X, and we denote

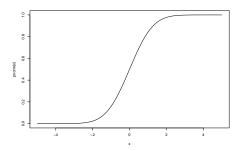
$$F(x) = \Pr(X \le x).$$

NOTE: x can be any value, even outside the support \mathcal{X} :

• E.g., X = dice roll, F(7) = 1, F(-1) = 0, F(3.1) = 0.5, etc...

CDF of the standard normal distribution (i.e., normal distribution with mean 0 and standard deviation 1)

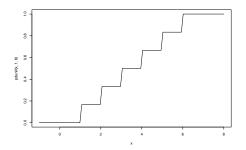
```
> curve(pnorm(x), xlim = c(-5,5))
```



- pnorm(x): standard normal CDF
- curve: depict the trajectory of the function given in the first argument.

CDF of dice roll $X \sim \mathsf{Uniform}\{1,2,\dots,6\}$

```
> curve(pdunif(x, 1, 6), xlim = c(-1,8))
```



- [p] $+ \sim =$ distribution function of \sim
- [d] $+ \sim =$ density function of \sim
- [r] $+ \sim =$ draw a number from \sim

Properties of CDF

• For any X, the probability that X takes a value smaller (larger) than negative (positive) infinity is zero (one):

$$\lim_{x\to -\infty} F(x) = 0, \qquad \lim_{x\to \infty} F(x) = 1$$

CDF is non-decreasing:

$$x_1 \leq x_2 \Rightarrow \Pr(X \leq x_1) \leq \Pr(X \leq x_2) \iff F(x_1) \leq F(x_2)$$

• In the following, we will mainly consider only continuous random variables (i.e., CDF is not a step function).

ullet The probability that X takes a value within an interval $\left[a,b\right]$ is

$$\begin{split} \Pr(X \in [a,b]) &= \Pr(a \leq X \leq b) = \Pr(X \leq b) - \Pr(X \leq a) \\ &= F(b) - F(a) \end{split}$$

- For a continuous random variable X, what is the probability that X takes a specific value, say $\Pr(X=1)$?
- Intuitively, it is expected that $\Pr(X=1)$ can be approximated by $\Pr(X \in [1,1+h])$ for a sufficiently small h>0:⁴ namely,

$$\Pr(X = 1) \approx \Pr(X \in [1, 1 + h])$$

= $F(1 + h) - F(1)$

 $^{^4}$ The probability that a person's height is 1.7 (m) should be approximately equal to the probability that his height is in between [1.7, 1.700001] (m).

ullet For example, let F be the standard normal CDF. Then,

```
> Fdiff <- function(h) pnorm(1 + h) - pnorm(1)
> Fdiff(2); Fdiff(0.1); Fdiff(0.001)
## [1] 0.1573054
```

```
## [1] 0.0002418497
```

[1] 0.02298919

(function(xxx): create an original function of xxx)

- As h approaches to zero, the probability gets smaller and smaller; in the limit the probability will be exactly 0.
- This is due to the continuity of $F(\cdot)$: $a \to b \Rightarrow F(a) \to F(b)$
- Hence, if X is continuous, Pr(X = x) = 0 holds for any x!

• On the other hand, if we divide F(x+h)-F(x) by h, the value does not degenerate to zero but converges to a specific value:

```
> Fdiff2 <- function(h) (pnorm(1 + h) - pnorm(1))/h
> Fdiff2(0.1); Fdiff2(0.01); Fdiff2(0.001)
```

```
## [1] 0.2298919
```

[1] 0.2407609

[1] 0.2418497

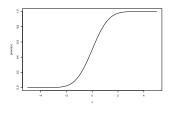
Probability Density Function: PDF

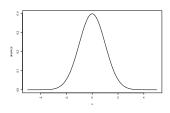
The probability density function (PDF) or simply the density function of X is defined as the "derivative" of the CDF of X:

$$f(x) = \lim_{h \to 0} \frac{F(x+h) - F(x)}{h} = \frac{\partial F(x)}{\partial x}$$

CDF and PDF of the standard normal distribution.

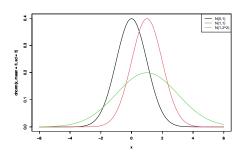
> curve(pnorm(x), xlim = c(-5,5)); curve(dnorm(x), xlim = c(-5,5))





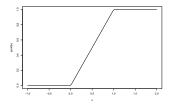
PDFs of normal distributions.

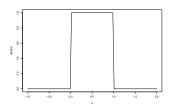
```
> curve(dnorm(x, mean = 0, sd = 1), xlim = c(-6,6), ylim = c(0, 0.4), col = 1)
> par(new = TRUE)
> curve(dnorm(x, mean = 1, sd = 1), xlim = c(-6,6), ylim = c(0, 0.4), col = 2)
> par(new = TRUE)
> curve(dnorm(x, mean = 1, sd = 2), xlim = c(-6,6), ylim = c(0, 0.4), col = 3)
> legend("topright", c("N(0,1)", "N(1,1)", "N(1,2^2)"),
+ lty = c(1,1,1), col = c(1,2,3))
```



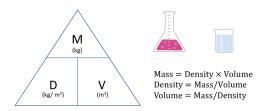
CDF and PDF of continuous $\mathsf{Uniform}[0,1].$

```
> curve(punif(x), xlim = c(-1,2)); curve(dunif(x), xlim = c(-1,2))
```





In what sense "density" function?



• The probability density of the event $\{X \in [x, x+h]\}$:

$$\frac{\Pr(X \in [x, x+h])}{h} = \frac{F(x+h) - F(x)}{h}$$

• Hence, f(x) is the probabilistic mass density of X at x.

- \bullet More intuitively, one can interpret the value of the density f(x) as
 - $f(x): \text{ How likely } X \text{ takes a value in the neighborhood of } x \in \mathcal{X}$
- Note that f(x) is NOT a probability, and thus its value can be larger than one but never be negative.

- When X is a discrete random variable, we can easily compute the probability $\Pr(X=x)$ for any $x\in\mathcal{X}.$
- The function $p(x) = \Pr(X = x)$ is called the probability mass function.

X	Measurement of the "likelihood" of the event $\{X=x\}$	
Discrete	probability mass function $p(x)$	
Continuous	probability density function $f(x)$	

Since PDF is the derivative of CDF, the following properties are straightforward.

Properties of PDF

- F(x) is non-decreasing $\Rightarrow f(x) \geq 0$ for all $x \in \mathcal{X}$
- $F(a) = \int_{-\infty}^{a} f(x) dx \Rightarrow$

$$\begin{split} \Pr(X \in [a,b]) &= F(b) - F(a) \\ &= \int_{-\infty}^b f(x) dx - \int_{-\infty}^a f(x) dx = \int_a^b f(x) dx \end{split}$$

• $F(\infty) = 1 \Rightarrow \int_{-\infty}^{\infty} f(x) dx = 1$

[1] 0.8413447

```
> pnorm(1) # Standard normal CDF
```

> integrate(dnorm, -Inf, 1)

```
## 0.8413448 with absolute error < 1.5e-05
```

> punif(0.2) # Uniform[0,1] CDF

[1] 0.2

> integrate(dunif, -Inf, 0.2)

- ## 0.2 with absolute error < 9.4e-05
 - integrate(h,a,b) = $\int_a^b h(x)dx$
 - Note: when $X \sim \mathsf{Uniform}[0,1]$, $\Pr(X \le a) = a$ for any $a \in [0,1]$.

Expectation and Variance

Expectation

Let X be a continuous random variable with density function f(x). The expectation of X, also called the mean of X, is defined as

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x f(x) dx.$$

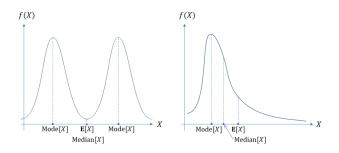
• NOTE $\int_{-\infty}^{\infty} x f(x) dx = \int_{\mathcal{X}} x f(x) dx$ (because f(x) = 0 for $x \notin \mathcal{X}$).

In the case of discrete random variable X, the expectation of X is given by

$$\mathbb{E}[X] = \sum_{i=1}^{\kappa} x_i p(x_i),$$

where $p(x_i) = \Pr(X = x_i)$, and $\{x_1, \dots, x_k\} = \mathcal{X}$.

 Note that the expectation is NOT the value that can be "expected" to occur.



- Median $[X]=x^*$ such that $F(x^*)=0.5$ (i.e., the 50-percentile point of X).
- $\mathsf{Mode}[X] = x^*$ such that $f(x^*) = \max_x f(x)$.

Example: If $X \sim \mathsf{Uniform}[a,b]$, then $\mathbb{E}[X] = (b+a)/2$.

Proof The PDF of uniform distribution on $\left[a,b\right]$ is

$$f(x) = \left\{ \begin{array}{cc} (b-a)^{-1} & \text{if } x \in [a,b] \\ 0 & \text{otherwise} \end{array} \right.$$

By the definition of expectation, we have

$$\begin{split} \mathbb{E}[X] &= \int_a^b x \cdot (b-a)^{-1} \cdot dx \\ &= [0.5 \cdot x^2]_a^b \cdot (b-a)^{-1} \\ &= 0.5 \cdot (b^2 - a^2) \cdot (b-a)^{-1} = (b+a)/2 \quad \blacksquare \end{split}$$

• In general, when a random variable X follows a symmetric distribution, the expectation $\mathbb{E}[X]$ coincides with the center of symmetry.

```
> h <- function(x) x * dunif(x, -2, 4) # X ~ Uniform[-2, 4]
> integrate(h, -Inf, Inf) # E(X)
```

1 with absolute error < 8.3e-05

```
> h <- function(x) x * dnorm(x, mean = 2, sd = 1) # X ~ Normal(2, 1
> integrate(h, -Inf, Inf) # E(X)
```

2 with absolute error < 1.2e-05

Expectation of a function of a random variable

For a random variable X, the expectation of g(X), $\mathbb{E}[g(X)]$, is given by

$$\mathbb{E}[g(X)] = \int_{-\infty}^{\infty} g(x)f(x)dx$$

The proof is complicated and is omitted.

ullet An important special case is when g(X) is an indicator function:

$$g(X) = \mathbf{1}\{X \leq a\} = \left\{ \begin{array}{ll} 1 & \text{if } X \leq a \\ 0 & \text{if } X > a \end{array} \right.$$

• In this case, we have $\mathbb{E}(\mathbf{1}\{X\leq a\})=F(a)$:

$$\begin{split} \mathbb{E}(\mathbf{1}\{X \leq a\}) &= \int_{-\infty}^{\infty} \mathbf{1}\{x \leq a\} f(x) dx \\ &= \int_{-\infty}^{a} 1 \cdot f(x) dx + \int_{a}^{\infty} 0 \cdot f(x) dx = F(a). \end{split}$$

• In general, for an event A, $\mathbb{E}[\mathbf{1}\{A\}] = \Pr(A)$.

```
> g <- function(x) ifelse(x <= 1, 1, 0) # 1{x <= 1}
> f <- function(x) g(x)*dnorm(x)
> integrate(f, -Inf, Inf) # E(g(x))
```

0.8413447 with absolute error < 4.7e-05

```
> pnorm(1)
```

```
## [1] 0.8413447
```

• ifelse(A, x, y): if the condition A is true, return x, otherwise return y.

Linearity of expectation

For a random variable X and constants a and b,

$$\mathbb{E}[aX + b] = a\mathbb{E}[X] + b$$

This property is called the linearity of expectation.

Proof Since aX + b is a function of X,

$$\begin{split} \mathbb{E}[aX+b] &= \int_{-\infty}^{\infty} (ax+b)f(x)dx \\ &= a\underbrace{\int_{-\infty}^{\infty} xf(x)dx}_{\mathbb{E}[X]} + b\underbrace{\int_{-\infty}^{\infty} f(x)dx}_{1} \\ &= a\mathbb{E}[X] + b \quad \blacksquare \end{split}$$

Variance

Variance

ullet For a random variable X, the (population) variance of X is defined by

$$\mathbb{V}(X) = \mathbb{E}[\{X - \mathbb{E}(X)\}^2] = \mathbb{E}(X^2) - \mathbb{E}(X)^2$$

• In particular, if $\mathbb{E}(X) = 0$, $\mathbb{V}(X) = \mathbb{E}(X^2)$.

When a sample $\{X_1,\dots,X_n\}$ of n observations is available, the sample variance is given by

$$\mathsf{Sample}\mathbb{V}(X_1,\dots,X_n) = \frac{1}{n}\sum_{i=1}^n \left(X_i - \frac{1}{n}\sum_{i=1}^n X_i\right)^2$$

The definition of the variance of a random variable is obtained by replacing the sample average $n^{-1} \sum_{i=1}^{n}$ with the expectation \mathbb{E} .

Standard Deviation

Standard deviation

The square root of the variance of X, $\sqrt{\mathbb{V}(X)}$, is called the standard deviation of X.

It is often convenient to standardize X by subtracting its expectation $\mathbb{E}(X)$ and dividing by its standard deviation $\sqrt{\mathbb{V}(X)}$. The standardized random variable has mean zero and variance one:

$$\begin{split} Z &= \frac{X - \mathbb{E}(X)}{\sqrt{\mathbb{V}(X)}} : \text{standardization of } X \\ \mathbb{E}(Z) &= \mathbb{E}\left[\frac{X - \mathbb{E}(X)}{\sqrt{\mathbb{V}(X)}}\right] = \frac{\mathbb{E}(X) - \mathbb{E}(X)}{\sqrt{\mathbb{V}(X)}} = 0 \\ \mathbb{V}(Z) &= \frac{\mathbb{E}[\{X - \mathbb{E}(X)\}^2]}{\mathbb{V}(X)} = \frac{\mathbb{V}(X)}{\mathbb{V}(X)} = 1 \end{split}$$