R Reference Card

Help for R studio Open the cheat sheet for RStudio by selecting the Help menu -> Cheatsheets -> RStudio IDE Cheat Sheet. Note the cheatsheet will usually be downloaded in a web browser as a .pdf.

Installing Packages Use the argument dependencies=TRUE to load all other packages required by the targeted package.

Loading Packages After installing a package, we can load it into the R environment using the library() or require() functions, which more or less do the same thing.

Use library(dplyr) The package dplyr contains several easy-to-use data management functions that we will learn to use to manage our data.

Starting R and RStudio Launch R
Studio, R starts automatically within RStudio

Quitting R Studio Use q() Type yes or no to confirm choice Loading data set (CSV File) my_data <- read.csv("path/to/your/data.csv") # Load data from a CSV fileViewing

Data: After loading your data, you can inspect it using the head() function to view the first few rows.

Building a Linear Regression Model: You can build a linear regression model using the lm() function. For example, let's predict miles per gallon (mpg) based on weight (wt) and horsepower (hp) using the mtcars dataset. model <- lm(mpg ~ wt + hp, data = mtcars) # Build a linear model

Model Interpretation The summary(model) function provides key statistics, such as coefficients, p-values, R-squared value, etc., which help assess the model's fit and significance.

Understanding data

str() Displays the structure of the dataset, including variable types. summary(): Provides summary statistics for each variable (e.g., min, max, mean, median).

head(): Displays the first few rows of the data.

Evaluating a model Residuals residuals <- resid(model) # Get

residuals from the model plot(residuals) # Plot residuals

R-squared: The R-squared value

from the model summary indicates how well the independent variables explain the variation in the dependent variable (mpg). Higher R-squared values (close to 1) indicate a better fit.

Interacting with RStudio Console:

This is where you enter R commands. It executes your commands and displays results. You can type code directly here for quick tests or experimentation.

Environment/History Pane: Displays the objects (variables, models, etc.) created in your R session and the history of previously executed commands.

Plots Pane: This is where graphical outputs (like plots and charts) are displayed. You can save plots using the "Export" button for further use.

Packages Pane: View and manage the libraries and packages installed in R. You can install new packages here and load them for use.

Interacting with R Basic Arithmetic

3 + 2 # Addition

10 / 2 # Division

4 * 5 # Multiplication

2³ # Exponentiation

Creating Variables: assign values to

variables using the <-

x < -10 # Assign the value 10 to x

y <- 20 # Assign the value 20 to y

Frequently used commands

summary(): Gives an overview of a

dataset or model.

str(): Displays the structure of an

object, such as a dataset or model.

mean() and sd(): Used to calculate the

mean and standard deviation of a variable.

1 0 Fig. 1

lm(): Fits a linear model to data.

Basic Plotting: Use plot()

- length(collection) returns the number of elements contained in the variable collection
- c(value1, value2, value3) creates a vector
- container[i] selects the i'th element from the variable container

https://swcarpentry.github.io/r-novice-inflammation/reference.html#glossary

References

Group, I. S. C. (n.d.-b). Introduction to R. https://stats.oarc.ucla.edu/stat/data/intro r/intro r interactive.html#(15)

FOR THE R CHEAT SHEET

Basic Operation

- # this is a comment in R
- Use x <- 3 to assign a value, 3, to a variable, x
- R counts from 1, unlike many other programming languages (e.g., Python)

Getting Started with R Gomments, Indinate, and semicidions # Anything prefaced by a pound sign (#) is a comment. # Comments are not executed by R. Instead, they explain what the code is doing, and indinated code (that is not a comment) will run in R as if it was on one line if a substant and dividently makes as a state of the semicolon marking the line break if the code was on separate lines, with the semicolon marking the line break if the code was on separate lines, with the semicolon marking the line break if the code was on separate lines, and the semicolon marking the line break if the code was on separate lines, with the semicolon marking the line break if the code was on separate lines, and the semicolon marking the line break if the code was on separate lines, and semicolon semicolon marking the line break if the code was on separate lines, and semicolon semicolon marking the line break if the code was on separate lines, and semicolon semicolon marking the line break if the code was on separate lines, and semicolon semicolon marking the line break if the code was on separate lines, and semicolon se

2.22 The R Zone

Read in the Cars and Cars2 datasets

obs_cubicinches <-

head(cars.4var)

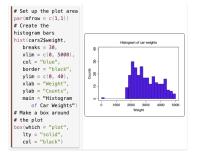
__sample(na.omit(cars.4var\$cubicinches), 1)
cars.4var[2,2] <- obs_cubicinches
cars.4var[4,4] <- obs_brand

cars <- read.csv("C:/ _/cars.txt",
stringsAsFactors = FALSE)
cars2 <- read.csv("C:/ _/cars2.txt",
stringsAsFactors = FALSE)</pre>

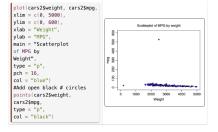
Missing data

Look at four variables from cars cars.4var <- cars[, c(1, 3, 4, 8)] head(cars.4var) # Make certain entries missing cars.4var[2,2] <- cars.4var[4,4] <- NA head(cars.4var) # Replace missing values with constants cars.4var[2,2] <- 0 cars.4var[4,4] <- "Missing" head(cars.4var) # Replace values with mean and mode cars.4var[2,2] <mean(na.omit(cars.4var\$cubicinches)) our_table <- table(cars.4var\$brand) our_mode <- names(our_table)[our_table == max(our_table)]
cars.4var[4,4] <- our_mode head(cars.4var) # Generate random observations obs_brand <sample(na.omit(cars.4var\$brand), 1)

Create a Histogram



Create a Scatterplot



Descriptive Statistics

```
mean(cars$weight) # Mean
median(cars$weight) # Median
length(cars$weight) # Number of observations
sd(cars$weight) # Standard deviation
summary(cars$weight) # Min, 01, Median, Mean, 03, Max
```

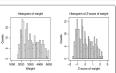
Transformations

```
# Min-max normalization
summary(carsweight)
ma <- main(carsweight)
ma <- main(carsweight)
minmax.weight <- (carsweight - mi)/(ma - mi)
minmax.weight <- (carsweight - mi)/
minmax.weight <- mi)/
minma
```

Decimal scaling
max(abs(cars\$weight)) # 4 digits
d.weight <- cars\$weight/(10^4); d.weight</pre>

Side-by-Side Histograms

```
par(mfrow = c(1,2))
# Create two histograms
hist(cars/weight, breaks = 20,
xlim = c(1000, 5000),
main = "Histogram of Weight",
xlab = "Meight",
ylab = "Counts")
box(which = "plot",
lty = "solid",
col = "black")
```



hist(z.weight, breaks = 20, xlim = c(-2, 3), main = "Histogram of Zscore of Meight", xlab = "Z-score of Weight", ylab = "Counts") how/white = "plat"

box(which = "plot", lty = "solid", col = "black")

Skewness

(3*(mean(cars\$weight) - median(cars\$weight)))/sd(cars\$weight) (3*(mean(z.weight) - median(z.weight)))/sd(z.weight)

Transformations for Normality

sqrt.weight - sqrt(cars\$weight) # Square root
sqrt.weight_skew - (3*(mean(sqrt.weight) - median(sqrt.weight)) / sd(sqrt.we:
ln.weight - (op)(cars\$weight) # Batural (ln.weight)) / sd(ln.weight)
ln.weight_skew - (3*(mean(ln.weight) - median(ln.weight))) / sd(ln.weight)
invsqrt.weight - 1 / sqrt(cars\$weight) # Inverse square root
invsqrt.weight_skew - (3*(mean(invsqrt.weight) - median(invsqrt.weight))) / sd

Histogram with Normal Distribution Overlay

lar(mfrowec(1,1))
x <- rnorm(1000000,
mean = mean (invsqrt.weight),
sd = sd(invsqrt.weight),
hist(invsqrt.weight,
breaks = 30,
xtim = c(0.0125, 0.0275),
col = "lightblue",
prob = TRUE,
border = "black",
xlab = "Invsres Square</pre>

Root of Weight", ylab = "Counts", main = "Histogram of Inverse Square Root

of Weight")
box(which = "plot",
lty = "solid",

col="black")
Overlay with Normal density
lines(density(x), col = "red")

Normal Q-Q Plot

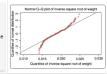
quoran(invsqrt.weight,
data= TRUE,
col = "red",
ylim = c(0.01, 0.03),
main = "Mormal 0-0 Plot of Inverse
Square Root of Meight")
qqline(invsqrt.weight,
col = "blue",
data= TRUE)



0.015 0.020 Inverse square root of weight

Normal Q-Q Plot

qqnorm(invsqrt.weight,
datax = TRUE,
col = "red",
ylim = c(0.01, 0.03),
main = "Normal O-O Plot of Inverse
Square Root of Weight")
qqline(invsqrt.weight,
col = "blue",
datax = TRUE)



De-transform data

```
# Transform x using y = 1 / sqrt(x) x \leftarrow \text{carsSweight}[1]; \ y \leftarrow 1 / \text{sqrt}(x) \\ \# \ \text{Detransform} \ x \ \text{using} \ x = 1 / (y)^2 \\ \text{detransformed} \ \leftarrow 1 / y^2 \\ x; \ y; \ \text{detransformed} \\ \text{detra
```

Create indicator variables

```
north_flag <- east_flag <- south_flag <-
c(rep(MA, 10))
region <- c(rep(c("north", "south", "east",
"west"), 2), "north", "south")

# Change the region variable to indicators
for (i in litengthiregion) {
i(fregioni] i = "north", north_flag[i] = 1
else north_flag[i] = 0
if(regioni] = "east", east_flag[i] = 1
else cast_flag[i] = 0
if(regioni] = "south", south_flag[i] = 1
else south_flag[i] = 0
}
north_flag[i] = south_flag[i] = 1
else south_flag[i] = 1
else south_flag[i] = 1
else south_flag[i] = 1</pre>
```

Index fields

Data frames have an index field;
the left-most column
cars
cars[order(carsSmpg),]
matrix < t(r
indexed m lord

For vectors or matrices, # add a column to act as an index field x < c.(11,13:1,1:4,3); y < c.(9,9:1) z < c.(2,1:9) matrix < t(rbind(x,y,z)); matrix indexed_m < cbind(c.(1:length(x)), matrix indexed_m [order(z),]

Duplicate records

er of duplicate records, use anyDuplicated ted(cars) ne each record, use Duplicated

(cars)
record is a duplicate,
record is not a duplicate

Let's duplicate the first record new.cars <- rbind(cars, cars[1,]) # Check for duplicates anyDuplicated(new.cars) # The 262nd record is a duplicated full(care(new.cars)

CHAPTER 3

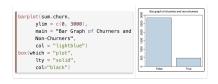
3.13 The R Zone

Read in the Churn data set

```
churn <- read.csv(file = "C://_churn.txt", stringsAsFactors=TRUE) # Show the first ten records churn[1:0.]

# Summarize the Churn variable sum.churn <- summary(churn$Churn) sum.churn <- summary(churn$Churn) prop.churn <- sum(churn$Churn = "True") / length(churn$Churn) prop.churn
```

Bar chart of variable Churn

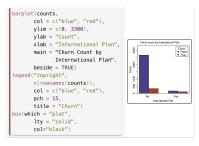


Create a table of proportions over rows

Create a table of proportions over columns

```
col.margin <- round(prop.table(counts, margin = 2), col.margin col.margin col.margin col.margin
```

Clustered Bar Chart, with legend



Make a table for counts of Churn and International Plan

```
counts <- table(churn$Churn, churn$Int.l.Plan, dnn=c("Churn", "International Plan"))
```

#Overlayed bar chart

```
barplot(counts,

legend = rownames(counts),
col = c("blue", "red"),
ylim = c(0, 3300),
ylab = "Count",
main = "Corporison Bar Chart:
Churn Proportions by
International Plan")
box(which = "plot",
lty = "solid",
cole"plack")
```

Create a table with sums for both variables

Clustered Bar Chart of Churn and International Plan with legend

```
barplott(counts),

col = c("blue", "green"),
ylin = (de, 3380),
ylab = "Counts",
xlab = "Churn",
main = "International Plan Count by
Churn",
beside = TRUE)
legend("tepright",
c(rownames(counts)),
col = c("blue", "green"),
pch = 15,
title = "Int' Plan")
box(mithe = "plot",
tity = "solim",
col="blue",
col="blue",
col="blue",
col="blue",
col="blue",
col="blue",
col="blue",
col="blue",
col="blue")
```

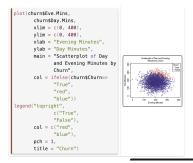
Histogram of non-overlayed Customer Service Calls

```
hist(churn$CustServ.Calls,
    xlim = (0,10),
    col = "LightNue",
    ylab = "Customer Service Calls",
    main = "Histogram of Customer Service
    Calls")
```

Two-sample T-Test for Int'l Calls

```
# Partition data
churn.false <- subset(churn,
churnSchurn ==
    "False")
churn.true <- subset(churn,
churnSchurn ==
    "True")
# Run the test
t.test(churn.falsesIntl.Calls,
churn.trueSIntl.Calls)
```

Scatterplot of Evening Minutes and Day Minutes, colored by Churn



Scatterplot of Day Minutes and Customer Service Calls, colored by Churn •

```
plot(churnSbay.Mins,
churnScutServ.Calls,
xlim = <0, 400),
xlab = "Oby Minutes",
ylub = "Gustemer Service Calls",
ylub = "Custemer Service Calls by Churn",
col = irelse(churnSchurn=="True",
    "blue"),
pch = irelse(churnSchurn=="True",
    "blue"),
col = c("red",
    "blue"),
col = c("red",
    "blue"),
pch = irelse(churnSchurn=="True",
col = c("red",
    "blue"),
pch = c(16, 20),
title = "Churn")
```

Scatterplot matrix



Correlation values and p-values in matrix form

```
# Collect variables of interest
 corrdata <-
        cbind(churn$Account.Length,
              churn$VMail.Message,
              churn$Day.Mins.
               churn$Day.Calls,
              churn$CustServ.Calls)
 # Declare the matrix
 corrpvalues <- matrix(rep(0, 25),
       ncol = 5)
 # Fill the matrix with correlations
  for (i in 1:4) {
       for (j in (i+1):5) {
              corrpvalues[i,j] <-
              corrpvalues[j,i] <-
                     round(cor.test(corrdata[,i],
                            corrdata[,j])$p.value,
  ound(cor(corrdata), 4)
 corrovalues
```

Scatterplot matrix



🗓 # Regression of Day Charge vs Day Minutes



Correlation values, with p-values

```
days <- cbind(churn$Day.Mins, churn$Day.Calls, churn$Day.Calls, churn$Day.Calls, churn$Day.Calls, churn$Day.Calls, MinsChargelest <- cor.test(churn$Day.Mins, churn$Day.Calls)

MinsChargelest <- cor.test(churn$Day.Mins, churn$Day.Calls, churn$Day.Charge)

CallsChargefest <- cor.test(churn$Day.Calls, churn$Day.Charge)

round(cor(days), d)

MinsChargefest$p.value

MinsChargefest$p.value

CallsChargefest$p.value
```

