RAdelaide 2015: Writing Reports Using R Markdown

Steve Pederson Bioinformatics Hub University of Adelaide

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Welcome To Day 2

Today's Topics

- 1. Writing Reports Using rmarkdown
- 2. Data Types in R
- 3. Writing Functions In R
- 4. Parallel Processing in R

Writing Reports Using rmarkdown

Now that we know how to load & tidy data:

- rmarkdown is a cohesive way to
 - Load & tidy data
 - Analyse data, including figures & tables
 - Publish everything in a complete report/analysis

We can do this all in one document, with our analysis code embedded!

Writing Reports Using rmarkdown

Using rmarkdown we can output our analysis directly as

- ► HTML
- MS Word Documents
- ▶ PDF Documents (like this one)
- Slidy or ioslides presentations

We never need to use MS Word or Excel ever again!!!

Writing Reports Using rmarkdown

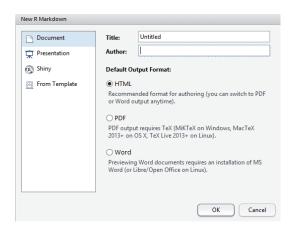
- ▶ The file format ends with .Rmd and allows us to include normal text alongside embedded *R* code.
- ▶ We can create all of our figures & tables directly from the raw data, as well as
 - Data or experimental descriptions
 - Descriptions of every stage of the analysis
 - Analytic conclusions
 - Any other information
- We can easily format text and include complex equations along with our code.

Creating an R Markdown document

Let's create our first rmarkdown document

- 1. Ensure you have rmarkdown installed
- 2. Go to the File drop-down menu in RStudio
- 3. New File -> R Markdown...

The New R Markdown Form



The New R Markdown Form

- 1. Change the Title to: My First Report
- 2. Change the Author to your own name
- 3. Leave everything else as it is & hit OK
- 4. Save the file as RMarkdownTutorial.Rmd

Looking at the file:

- ► The header section is contained within the two --- delineation lines
- ▶ Lines 8 − 10 are plain text
 - ► Bold is indicated by **Knit**
 - Italics can be indicated using a single *
- ▶ Lines 12 14 are an R code chunk
 - Chunks always begin with ""{r}
 - Chunks always end with ""
 - The code goes between these two delineators

- ► The default format is an html_document & we can change this later.
- ▶ Generate the document by clicking Knit HTML



A preview window will appear with the compiled report

- Note the hyperlink to the RMarkdown website & the bold typeface for the word Knit
- ► The *R* code and the results are printed for the line summary(cars)
- ▶ The plot of speed Vs. 'distance has been embedded
- ▶ The code generating the plot was hidden using echo = FALSE

We could also save this as an MS Word document (i.e. .docx format)



By default, this will be Read-Only

Saving as a .PDF may require an installation of LATEX.

Making our own report

Now we can modify the code to create our own analysis.

- ▶ Delete everything in your R Markdown file EXCEPT the header
- We'll analyse a dataset called ToothGrowth contained in the R package datasets
- ▶ First we'll need to describe the data

?ToothGrowth

Rename the report

First we should change the title of the report to something suitable, e.g. *The Effects of Vitamin C Delivery Methods on Tooth Growth*

Create a "Data Description" Section

Now let's add a section header for our analysis to start the report

- 1. Type # Data Description after the header and leaving a line
- 2. Use your own words to describe the data

My text was:

We are interested in the effects of Vitamin C on tooth growth, and for this experiment we used Guinea Pigs as the model organism. We gave guinea pigs Vitamin C supplements at three doses (0.5, 1 and 2mg) using two delivery methods. Delivery method 1 was orange juice, whilst delivery method 2 was ascorbic acid. Each treatment combination was given to 10 guinea pigs, giving a total sample size of 60.

The measured response was the length of odontoblasts.

Hopefully you mentioned that there were 10 guinea pigs in each group, with a total of 60.

Question

Can we get that information from the data itself?

Hopefully you mentioned that there were 10 guinea pigs in each group, with a total of 60.

Question

Can we get that information from the data itself? Yes

Solution

nrow(ToothGrowth) would give the total number of samples

We can actually embed this in our data description!

1. Instead of the number 60 in your description, enter:

```
'r nrow(ToothGrowth)'
```

2. Recompile the HTML document.

Adding some *R* code

After our description, we could have a look at the data in a summary

```
ToothGrowth %>%
  group_by(supp, dose) %>%
  summarise(Count = n())
```

However, we'll need to add a code chunk before this to load the package dplyr

Loading R packages

- Before the Data Description header, add a new header called Required Packages
- Create a code chunk with the contents library(dplyr)
- 3. Recompile the HTML

Adding this has loaded the package dplyr so that any code chunks following will be able to use the functions in the package

Tidying our output

Notice that loading dplyr gave us an overly informative message.

We can turn this off to make the report look nicer

- After the r inside the brackets at the start of the code chunk, add a comma
- Start typing the word message and use the auto-complete feature to set message = FALSE
- 3. Recompile

Add the data summary

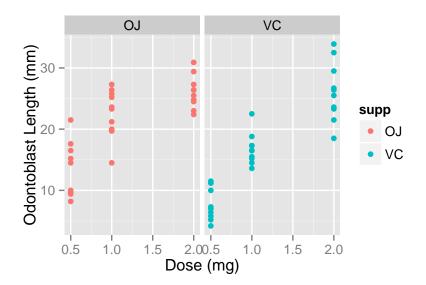
```
ToothGrowth %>%
 group_by(supp, dose) %>%
 summarise(Count = n())
## Source: local data frame [6 x 3]
## Groups: supp
##
##
    supp dose Count
## 1
      0.5
               10
## 2 OJ 1.0 10
## 3 OJ 2.0 10
## 4 VC 0.5 10
## 5 VC 1.0 10
    VC 2.0
## 6
             10
```

Add a plot of the data

We can use ggplot2 for this

- 1. Load this package back in the Required Packages section
- Create a plot using geom_point()
- 3. Colour the points based on the supp variable
- 4. Facet the plot based on the supp variable

Add a plot of the data



Analyse the data

One way to analyse this data would be to conduct a *t*-test at each dosage level

- 1. Create a new section called Analysis
- Create a lower level heading called Dosage == 0.5.
 (This can be done using 2 or 3 hashes)
- 3. Create a code chunk with the following code

```
(lowDosage <- t.test(len~supp, subset(ToothGrowth, dose==0.5)))</pre>
```

Analyse the data

Questions

- 1. Why would we store the results as an *R* object?
- 2. What do the round brackets at the start & end of the code do?

Analyse the data

- 1. Repeat the above for the medium and high dosage levels
- 2. Write some kind of conclusion based on the above results
- 3. To embed the p-values directly in our text, we can call the R objects we've created via the embedded code procedure 'r embedded code'

Hints

- 1. The *p*-values can be accessed via lowDosage\$p.value
- 2. We may also like to restrict the number of decimal places using the command round()

If we have time

We could provide an even more informative output by plotting confidence intervals

- 1. Create a new subsection called Confidence Intervals
- Create a data.frame with variables dose, diff, lower & upper filling with values from the R objects we've already created
- Plot these intervals usinggeom_point and geom_errorbar

Hints

- 1. The estimates of the mean tooth growth will be in the *R* objects in a component called \$estimate
- 2. The lower and upper bounds for the 95% confidence interval will be in the component called \$conf.int



Finishing the analysis

After you're happy with the way your analysis looks

- ► A good habit is to finish with a section called Session Info
- ▶ Add a code chunk which calls the R command sessionInfo()

Finishing the analysis

So far we've been compiling everything as HTML, but let's switch to an MS Word document

We could email this to our supervisors, or upload to Google docs for collaborators. . .

Summary

This basic process is incredibly useful

- We never need to cut & paste anything between document formats
- ▶ Every piece of information comes directly from our *R* analysis
- We can very easily incorporate new data
- Analyses can be simply changed as new information comes to hand
- ► Creates reproducible research