Tables

Introduction

First, load {tidyverse} and {papaja} and import the data.

Now let's build a data frame that will be our table.

```
(condition_prepost_means <- hai_long |>
  group_by(condition, prepost) |>
  summarise(across(starts_with("panas"), ~ mean(.x, na.rm = TRUE))))
```

```
## # A tibble: 4 x 3
## # Groups: condition [2]
##
    condition prepost panas
    <fct>
##
              <fct>
                      <dbl>
## 1 Control
              Pre
                        2.99
## 2 Control Post
                       2.76
## 3 HAI
              Pre
                        2.97
## 4 HAI
              Post
                       3.23
```

Tables by {knitr}

The {knitr} package uses the kable() function to format tables.

library(knitr)

kable(condition_prepost_means)

condition	prepost	panas
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Column and row names

You can control column names and row names with col.names and row.names.

Condition	Prepost	Mean PANAS
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Column alignment

By default, character columns are left aligned and numeric columns are right aligned. You can set alignment manually with the align argument with l = left, c = center, and r = right. You can just pass a character string with a series of those letters.

Condition	Prepost	Mean PANAS
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Digit rounding

Round the digits for all numeric data columns with digits argument.

Condition	Prepost	Mean PANAS
Control	Pre	2.99
Control	Post	2.76
HAI	Pre	2.97
HAI	Post	3.23

If you want different digits for different columns, you can pass a vector to the \mathtt{digits} argument.

```
condition_prepost_means |>
  pivot_wider(id_cols = condition, names_from = prepost, values_from = panas) |>
  kable(digits = c(0, 2, 3))
```

condition	Pre	Post
Control	2.99	2.761
HAI	2.97	3.232

Table 1 PANAS scores by condition and prepost

Condition	Prepost	Mean PANAS
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Table titles

Add a title to the table with the caption argument. The good news is that we can cross-reference easily (Table 1). The bad news is that with captions, tables in PDFs are automatically placed at the top of the page. We'll see how to fix this later.

Supplementing kable with {kableExtra}

The kable() function is intentionally simple to use and therefore does not have a lot of additional functionality. The {kableExtra} package supplements the kable() functionality with additional formatting options by adding additional functions after the kable() function call using the |> pipe (a bit like how ggplot() works). Check out Create Awesome LaTeX Table with knitr::kable and kableExtra.

```
# install.packages("kableExtra")
library(kableExtra)
```

General styling

The kable_styling() function formats a number of things such as font size, table width, and table alignment. I'll also add latex_options = "hold_position" to keep the table in the text. Otherwise, it puts it at the top of the page.

Table 2
PANAS scores by condition and prepost

Condition	Prepost	Mean PANAS
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Labels spanning rows

If you want to label groups of rows, use pack_rows(). Let's get rid of the condition column and label the conditions explicitly.

```
condition_prepost_means2 <- condition_prepost_means |>
  ungroup(condition) |>
  select(-condition)
```

```
kable(condition_prepost_means2,
```

booktabs = TRUE)

prepost	panas
Pre	2.989873
Post	2.760759
Pre	2.971795
Post	3.232051

	Mean PANAS
Contro	1
Pre	2.989873
Post	2.760759
HAI	
Pre	2.971795
Post	3.232051

Notice that we removed the first column name with "".

Labels spanning columns

You can labels groups of columns with the add_header_above() function. Let's rearrange the data into wide format to illustrate this.

```
(wide_means <- condition_prepost_means |>
  unite(cond_prepost, condition:prepost) |>
  pivot_wider(names_from = cond_prepost, values_from = panas))
## # A tibble: 1 x 4
     Control_Pre Control_Post HAI_Pre HAI_Post
##
##
           <dbl>
                        <dbl>
                                 <dbl>
                                          <dbl>
## 1
            2.99
                         2.76
                                  2.97
                                           3.23
kable(wide_means, booktabs = TRUE)
      Control_Pre Control_Post
                                 HAI_Pre
                                           HAI_Post
         2.989873
                       2.760759
                                 2.971795
                                             3.232051
```

Now that the data are in wide format, we can add the column names by repeating *Pre* and *Post* then add the headers.

Cor	ntrol	Н	ΑI
Pre	Post	Pre	Post
2.99	2.76	2.97	3.23

Maybe we need a column stating this is Mean PANAS.

	Control		Н	AI
	Pre	Post	Pre	Post
Mean PANAS	2.99	2.76	2.97	3.23

Table footnotes

Add table notes with the footnote() function.

Table 3
PANAS scores by condition and prepost

Condition*	Prepost	Mean PANAS
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Note: Source: Thayer & Stevens (2021)

^{* 73} control participants, 72 HAI participants

Landscape

Rotate wide tables with landscape() function.

Table 4 PANAS scores by condition and prepost

Condition	$\operatorname{Prepost}$	Mean PANAS
Control	Pre	2.989873
Control	Post	2.760759
HAI	Pre	2.971795
HAI	Post	3.232051

Note: Source: Thayer & Stevens (2021)

Tables by {papaja}

The {papaja} package uses the kable() function to format tables in APA format with the apa_table() function. You can use many of the same arguments that are available in the kable() function. You can control where the table is placed (here, top, bottom) with the placement argument. You can add a general footnote with the note argument.

Table 5
PANAS scores by condition and prepost

Condition	Prepost	Mean PANAS
Control	Pre	2.99
Control	Post	2.76
$_{ m HAI}$	Pre	2.97
HAI	Post	3.23

Note. Source: Thayer & Stevens (2021)

Notice the alignment is different, with everything left aligned. Let's right align the means.

Table 6
PANAS scores by condition and prepost

Condition	Prepost	Mean PANAS
Control	Pre	2.99
Control	Post	2.76
$_{ m HAI}$	Pre	2.97
HAI	Post	3.23

Note. Source: Thayer & Stevens (2021)

You can rotate to landscape orientation with the landscape = TRUE argument.

Table 7
PANAS scores by condition and prepost

Condition	Prepost	Mean PANAS
Control	Pre	2.99
Control	Post	2.76
HAI	Pre	2.97
HAI	Post	3.23

Note. Source: Thayer & Stevens (2021)

APA-formatted statistics by {papaja}

{papaja} also includes apa_print(), which extracts statistical values in APA format.

Linear regression

```
hai_lm <- lm(panas_post ~ condition * gender, data = hai)
summary(hai_lm)
##
## Call:
## lm(formula = panas_post ~ condition * gender, data = hai)
## Residuals:
##
       Min
                1Q
                     Median
                                  3Q
                                         Max
## -1.80909 -0.58525 0.01475 0.48333 1.88333
## Coefficients:
##
                         Estimate Std. Error t value Pr(>|t|)
                          ## (Intercept)
## conditionhai
                                              3.943 0.000122 ***
                          0.52385
                                    0.13287
## genderMale
                          0.33142
                                    0.20066 1.652 0.100662
## conditionhai:genderMale -0.18218
                                    0.30884 -0.590 0.556142
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7481 on 153 degrees of freedom
## Multiple R-squared: 0.109, Adjusted R-squared: 0.09149
## F-statistic: 6.236 on 3 and 153 DF, p-value: 0.0005053
apa_print(hai_lm)
## $estimate
## $estimate$Intercept
## [1] "b = 2.69, 95\\% CI [2.50, 2.87]$"
##
## $estimate$conditionhai
## [1] "b = 0.52, 95\\% CI [0.26, 0.79]$"
##
## $estimate$genderMale
## [1] "b = 0.33, 95\\% CI [-0.07, 0.73]"
##
## $estimate$conditionhai_genderMale
## [1] "b = -0.18, 95\\% CI -0.79, 0.43
##
## $estimate$modelfit
## $estimate$modelfit$r2
```

```
## [1] "R^2 = .11"
##
## $estimate$modelfit$r2 adj
## [1] "$R^2_{adj} = .09$"
##
## $estimate$modelfit$aic
## [1] "$\\mathrm{AIC} = 360.36$"
##
## $estimate$modelfit$bic
## [1] "$\\mathrm{BIC} = 375.64$"
##
##
##
## $statistic
## $statistic$Intercept
## [1] "$t(153) = 28.03$, $p < .001$"
##
## $statistic$conditionhai
## [1] "$t(153) = 3.94$, $p < .001$"
## $statistic$genderMale
## [1] "$t(153) = 1.65$, $p = .101$"
## $statistic$conditionhai_genderMale
## [1] "$t(153) = -0.59$, $p = .556$"
##
## $statistic$modelfit
## $statistic$modelfit$r2
## [1] "$F(3, 153) = 6.24$, $p = .001$"
##
##
##
## $full result
## $full result$Intercept
## [1] "b = 2.69, 95\% CI [2.50, 2.87], t(153) = 28.03, p < .001"
##
## $full_result$conditionhai
## [1] "b = 0.52$, 95\\% CI $[0.26, 0.79]$, t(153) = 3.94$, p < .001$"
##
## $full_result$genderMale
## [1] "b = 0.33, 95\% CI [-0.07, 0.73], t(153) = 1.65, p = .101"
##
## $full_result$conditionhai_genderMale
## [1] "b = -0.18, 95\% CI -0.79, 0.43, t(153) = -0.59, p = .556"
##
```

```
## $full_result$modelfit
## $full_result$modelfit$r2
## [1] "R^2 = .11, F(3, 153) = 6.24, p = .001"
##
##
##
## $table
## A data.frame with 6 labelled columns:
##
##
                                term estimate
                                                   conf.int statistic df p.value
## 1
                                         2.69 [2.50, 2.87]
                                                               28.03 153 < .001
                           Intercept
## 2
                         Conditionhai
                                         0.52 [0.26, 0.79]
                                                               3.94 153 < .001
                          GenderMale
                                       0.33 [-0.07, 0.73]
## 3
                                                                1.65 153
                                                                            .101
## 4 Conditionhai $\\times$ GenderMale -0.18 [-0.79, 0.43] -0.59 153
                                                                            .556
## term : Predictor
## estimate : $b$
## conf.int : 95\\% CI
## statistic: $t$
          : $\\mathit{df}$
## p.value : $p$
## attr(,"class")
## [1] "apa_results" "list"
apa_table(apa_print(hai_lm)$table,
         caption = "Linear regression results",
         placement = "h")
```

Table 8
Linear regression results

Predictor	b	95% CI	t	df	p
Intercept	2.69	[2.50, 2.87]	28.03	153	< .001
Conditionhai	0.52	[0.26, 0.79]	3.94	153	< .001
GenderMale	0.33	[-0.07, 0.73]	1.65	153	.101
Conditionhai \times GenderMale	-0.18	[-0.79, 0.43]	-0.59	153	.556

Let's clean up those predictor names.

 $\begin{array}{c} \text{Table 9} \\ \textit{Linear regression results} \end{array}$

term	b	95% CI	t	df	p
Intercept	2.69	[2.50, 2.87]	28.03	153	< .001
Condition	0.52	[0.26, 0.79]	3.94	153	< .001
Gender	0.33	[-0.07, 0.73]	1.65	153	.101
Condition \times Gender	-0.18	[-0.79, 0.43]	-0.59	153	.556

How could we name the first column *Predictor* instead of *term*?

Other table packages

- {gt} RStudio's grammar of tables (logically like ggplot2)
- {flextable} Good Word output but a bit tricky to work with
- {huxtable} Very flexible but tricky to work with