

# Tables

## Introduction

This file can be found at `dsvr_2021/docs/inclass_demos/23_tables.Rmd`. First, load `{tidyverse}` and `{papaja}` and import the data.

```
library(tidyverse)
library(papaja)
hai <- read_csv("https://decisionslab.unl.edu/data/thayer_stevens_2020_data1.csv") %>%
  select(experiment, participant, condition, gender,
         panas_pre = panas_pre_pos, panas_post = panas_post_pos)
hai_long <- hai %>%
  pivot_longer(contains("panas"), names_to = "prepost", values_to = "panas") %>%
  mutate(condition = fct_recode(condition, "Control" = "control", "HAI" = "hai"),
         prepost = fct_recode(prepost, "Pre" = "panas_pre", "Post" = "panas_post"),
         prepost = fct_relevel(prepost, c("Pre", "Post")))
```

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)))
condition_prepost_means
```

```
## # 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.

```
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`.

```
kable(condition_prepost_means, col.names = c("Condition", "Prepost", "Mean PANAS"))
```

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.

```
kable(condition_prepost_means, align = "rcl",  
      col.names = c("Condition", "Prepost", "Mean PANAS"))
```

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.

```
kable(condition_prepost_means, digits = 2,  
      col.names = c("Condition", "Prepost", "Mean PANAS"))
```

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

**How would we round the digits in the data itself?**

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.

```
kable(condition_prepost_means, caption = "PANAS scores by condition and prepost",  
      col.names = c("Condition", "Prepost", "Mean PANAS"))
```

### LaTeX format

You can also set the `format` to be `"latex"`, which will output the LaTeX code for the table. You only need to do this if you want to use some LaTeX options to format the table (e.g., turn the table to landscape orientation, split the table across pages, etc.). Here we used the LaTeX-specific `booktabs` argument to make the table look nicer.

```
kable(condition_prepost_means, format = "latex", booktabs = TRUE,
      col.names = c("Condition", "Prepost", "Mean PANAS"))
```

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

While using the LaTeX formatting allows for powerful control over your table, the downside is that it can only output to PDFs. Though some publishers allow you to submit LaTeX files as the editable version of a manuscript, some only allow Word documents. So you'll either need to find another way to export the table to Word or recreate the table in Word.

### 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.

```
kable(condition_prepost_means, booktabs = TRUE,
      caption = "PANAS scores by condition and prepost",
      col.names = c("Condition", "Prepost", "Mean PANAS")) %>%
  kable_styling(font_size = 6, latex_options = "hold_position")
```

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



Wrap table in text

Need to add wrap *wrapfig* package to YAML header.

```
kable(condition_prepost_means, booktabs = TRUE) %>%
  kable_styling(position = "float_right")
```

Lorem ipsum dolor sit amet, consectetur adipiscing elit. Cras sit amet mauris in ex ultricies elementum vel rutrum dolor. Phasellus tempor convallis dui, in hendrerit mauris placerat scelerisque. Maecenas a accumsan enim, a maximus velit. Pellentesque in risus eget est faucibus convallis nec at nulla. Phasellus nec lacinia justo. Morbi fermentum, orci id varius accumsan, nibh neque porttitor ipsum, consectetur luctus risus arcu ac ex. Aenean a luctus augue. Suspendisse et auctor nisl. Suspendisse cursus ultrices quam non vulputate. Phasellus et pharetra neque, vel feugiat erat. Sed feugiat elit at mauris commodo consequat. Sed congue lectus id mattis hendrerit. Mauris turpis nisl, congue eget velit sed, imperdiet convallis magna. Nam accumsan urna risus, non feugiat odio vehicula eget.

condition	prepost	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

```
kable(condition_prepost_means2, booktabs = TRUE,
  col.names = c("", "Mean PANAS")) %>%
  pack_rows("Control", 1, 2) %>%
  pack_rows("HAI", 3, 4)
```

	Mean PANAS
<b>Control</b>	
Pre	2.989873
Post	2.760759
<b>HAI</b>	
Pre	2.971795
Post	3.232051

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

### Labels spanning columns

You can label 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.

```
kable(wide_means, booktabs = TRUE, col.names = rep(c("Pre", "Post"), 2),
  digits = 2) %>%
  add_header_above(c("Control" = 2, "HAI" = 2))
```

Control		HAI	
Pre	Post	Pre	Post
2.99	2.76	2.97	3.23

Maybe we need a row name.

```
rownames(wide_means) <- "Mean PANAS"
```

```
## Warning: Setting row names on a tibble is deprecated.
```

```
kable(wide_means, booktabs = TRUE, col.names = rep(c("Pre", "Post"), 2),
  digits = 2) %>%
  add_header_above(c("", "Control" = 2, "HAI" = 2))
```

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

**Table footnotes**

Add table notes with the `footnote()` function.

```
kable(condition_prepost_means, booktabs = TRUE,
      caption = "PANAS scores by condition and prepost",
      col.names = c("Condition*", "Prepost", "Mean PANAS")) %>%
  kable_styling(latex_options = "hold_position") %>%
  footnote(general = "Source: Thayer & Stevens (2021)",
          symbol = "73 control participants, 72 HAI participants",
          footnote_as_chunk = TRUE)
```

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.

```
kable(condition_prepost_means, booktabs = TRUE,
      caption = "PANAS scores by condition and prepost",
      col.names = c("Condition", "Prepost", "Mean PANAS")) %>%
kable_styling(latex_options = "hold_position") %>%
footnote(general = "Source: Thayer & Stevens (2021)",
      footnote_as_chunk = TRUE) %>%
landscape()
```

Table 4  
*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
<i>Note:</i> 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.

```
apa_table(condition_prepost_means,
  caption = "PANAS scores by condition and prepost",
  col.names = c("Condition", "Prepost", "Mean PANAS"),
  placement = "h",
  note = "Source: Thayer & Stevens (2021)")
```

Table 5  
*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_table(condition_prepost_means, booktabs = TRUE,
  caption = "PANAS scores by condition and prepost",
  col.names = c("Condition", "Prepost", "Mean PANAS"),
  align = c("l", "l", "r"),
  placement = "h",
  note = "Source: Thayer & Stevens (2021)")
```

Table 6  
*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)

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

```
apa_table(condition_prepost_means, booktabs = TRUE,  
  caption = "PANAS scores by condition and prepost",  
  col.names = c("Condition", "Prepost", "Mean PANAS"),  
  align = c("l", "l", "r"),  
  placement = "h",  
  note = "Source: Thayer & Stevens (2021)",  
  landscape = TRUE)
```



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.

#### Correlation

```
hai_corr <- cor.test(hai$panas_pre, hai$panas_post)
hai_corr

##
## Pearson's product-moment correlation
##
## data: hai$panas_pre and hai$panas_post
## t = 14.494, df = 155, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.6830883 0.8180051
## sample estimates:
## cor
## 0.7585621

apa_print(hai_corr)

## $estimate
## [1] "$r = .76$, 95\\% CI $[.68, .82]$"
##
## $statistic
## [1] "$t(155) = 14.49$, $p < .001$"
##
## $full_result
## [1] "$r = .76$, 95\\% CI $[.68, .82]$, $t(155) = 14.49$, $p < .001$"
##
## $table
## A data.frame with 5 labelled columns:
##
## estimate conf.int statistic df p.value
## 1 .76 [.68, .82] 14.49 155 < .001
##
## estimate : $r$
## conf.int : 95\\% CI
## statistic: $t$
## df : $\\mathit{df}$
## p.value : $p$
## attr("class")
## [1] "apa_results" "list"
```

Pre and post PANAS are correlated:  $r = .76$ , 95% CI  $[.68, .82]$ ,  $t(155) = 14.49$ ,

$p < .001$ .

### T-test

```
hai_ttest <- t.test(hai$panas_pre, hai$panas_post, paired = TRUE)
hai_ttest

##
## Paired t-test
##
## data: hai$panas_pre and hai$panas_post
## t = -0.33057, df = 156, p-value = 0.7414
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.09774442 0.06971894
## sample estimates:
## mean of the differences
## -0.01401274

apa_print(hai_ttest)

## $estimate
## [1] "$M_d = -0.01$, 95\\% CI $[-0.10, 0.07]$"
##
## $statistic
## [1] "$t(156) = -0.33$, $p = .741$"
##
## $full_result
## [1] "$M_d = -0.01$, 95\\% CI $[-0.10, 0.07]$, $t(156) = -0.33$, $p = .741$"
##
## $table
## A data.frame with 5 labelled columns:
##
##   estimate      conf.int statistic  df p.value
## 1    -0.01 [-0.10, 0.07]    -0.33 156    .741
##
## estimate : $M_d$
## conf.int : 95\\% CI
## statistic: $t$
## df       : $\\mathit{df}$
## p.value  : $p$
## attr("class")
## [1] "apa_results" "list"
```

Pre scores do not differ from post scores:  $t(156) = -0.33$ ,  $p = .741$ .

### 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)      2.68525     0.09578  28.035 < 2e-16 ***
## conditionhai      0.52385     0.13287   3.943 0.000122 ***
## genderMale        0.33142     0.20066   1.652 0.100662
## conditionhai:genderMale -0.18218     0.30884  -0.590 0.556142
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 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] "$R^2_{adj} = .09$"
##
## $estimate$model$fit$aic
## [1] "$\\mathrm{AIC} = 360.36$"
##
## $estimate$model$fit$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] "$t(153) = -0.59$, $p = .556$"
##
## $statistic$model$fit
## $statistic$model$fit$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 $[-0.79, 0.43]$, $t(153) = -0.59$, $p = .556$"
##
## $full_result$model$fit
## $full_result$model$fit$r2
## [1] "$R^2 = .11$, $F(3, 153) = 6.24$, $p = .001$"

```

```
##
##
##
## $table
## A data.frame with 5 labelled columns:
##
##               term estimate      conf.int statistic p.value
## 1             Intercept      2.69 [2.50, 2.87]    28.03 < .001
## 2             Conditionhai      0.52 [0.26, 0.79]     3.94 < .001
## 3              GenderMale      0.33 [-0.07, 0.73]     1.65  .101
## 4 Conditionhai $\\times$ GenderMale -0.18 [-0.79, 0.43]   -0.59  .556
##
## term      : Predictor
## estimate  : $b$
## conf.int  : 95\\% CI
## statistic: $t(153)$
## p.value   : $p$
## attr(,"class")
## [1] "apa_results" "list"
```

Here's the model fit:  $R^2 = .11$ ,  $F(3, 153) = 6.24$ ,  $p = .001$ . There was an effect of condition ( $b = 0.52$ , 95% CI [0.26, 0.79],  $t(153) = 3.94$ ,  $p < .001$ ) but not gender ( $b = 0.33$ , 95% CI [-0.07, 0.73],  $t(153) = 1.65$ ,  $p = .101$ ) and no interaction ( $b = -0.18$ , 95% CI [-0.79, 0.43],  $t(153) = -0.59$ ,  $p = .556$ ) (Table 8).

```
apa_table(apa_print(hai_lm)$table,
  caption = "Linear regression results",
  placement = "h")
```

Table 8

*Linear regression results*

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

Let's clean up those predictor names.

```
hai_lm_table <- apa_print(hai_lm)$table
hai_lm_table <- hai_lm_table %>%
  mutate(term = str_replace(term, "Conditionhai", "Condition"),
         term = str_replace(term, "GenderMale", "Gender"))
apa_table(hai_lm_table,
  caption = "Linear regression results",
  placement = "h")
```

Table 9

*Linear regression results*

term	<i>b</i>	95% CI	<i>t</i> (153)	<i>p</i>
Intercept	2.69	[2.50, 2.87]	28.03	< .001
Condition	0.52	[0.26, 0.79]	3.94	< .001
Gender	0.33	[-0.07, 0.73]	1.65	.101
Condition × Gender	-0.18	[-0.79, 0.43]	-0.59	.556

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