

IS5 in R: Relationships Between Categorical Variables–Contingency Tables (Chapter 3)

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Introduction and background

This document is intended to help describe how to undertake analyses introduced as examples in the Fifth Edition of *Intro Stats* (2018) by De Veaux, Velleman, and Bock. More information about the book can be found at http://wps.aw.com/aw_deveaux_stats_series. This file as well as the associated R Markdown reproducible analysis source file used to create it can be found at <http://nhorton.people.amherst.edu/is5>.

This work leverages initiatives undertaken by Project MOSAIC (<http://www.mosaic-web.org>), an NSF-funded effort to improve the teaching of statistics, calculus, science and computing in the undergraduate curriculum. In particular, we utilize the `mosaic` package, which was written to simplify the use of R for introductory statistics courses. A short summary of the R needed to teach introductory statistics can be found in the `mosaic` package vignettes (<http://cran.r-project.org/web/packages/mosaic>). A paper describing the `mosaic` approach was published in the *R Journal*: <https://journal.r-project.org/archive/2017/RJ-2017-024>.

Chapter 3: Relationships Between Categorical Variables–Contingency Tables

Section 3.1: Contingency Tables

XX note: Code chunk was flowing out of chunk space in pdf so I fixed that

```
library(mosaic)
library(readr)
library(janitor)
OKCupid <-
  read_csv("http://nhorton.people.amherst.edu/is5/data/OKCupid_CatsDogs.csv", skip = 1) %>%
  clean_names()

## Parsed with column specification:
## cols(
##   CatsDogsBoth = col_character(),
##   Gender = col_character(),
##   `drugsY/N` = col_character(),
##   `smokesY/N` = col_character()
## )

names(OKCupid)

## [1] "cats_dogs_both" "gender"          "drugs_y_n"       "smokes_y_n"
```

By default, `read_csv()` prints the variable names. These messages can be suppressed using the `message=FALSE` code chunk option to save space and improve readability.

Here we use the `clean_names()` function from the `janitor` package to sanitize the names of the columns (which would otherwise contain special characters or whitespace). You can use the `names()` function to check the cleaned names.

We use `skip = 1` because the first line in the original data set is `Col1`, `Col2`, etc.

XX A reference saying how we are replicating Table 3.1-3.4 starting at pg. 65 would be helpful

```
## XX Consider using useNA = "no" (in the first tally command for example) to match the table in page65
```

```
tally(~ cats_dogs_both + gender, margin = TRUE, data = OKCupid)
```

```
##           gender
## cats_dogs_both  F      M Total
##   Has Both    897    577 1474
##   Has cats   3412   2388 5800
##   Has dogs   3431   3587 7018
##   <NA>       16377 29274 45651
##   Total      24117 35826 59943
```

```
tally(~ cats_dogs_both + gender, format = "percent", margin = TRUE, data = OKCupid)
```

```
##           gender
## cats_dogs_both  F      M      Total
##   Has Both    1.4964216  0.9625811  2.4590027
##   Has cats    5.6920741  3.9837846  9.6758587
##   Has dogs    5.7237709  5.9840182 11.7077891
##   <NA>        27.3209549 48.8363946 76.1573495
##   Total      40.2332216 59.7667784 100.0000000
```

```
tally(~ cats_dogs_both | gender, format = "percent", margin = TRUE, data = OKCupid)
```

```
##           gender
## cats_dogs_both  F      M
##   Has Both    3.719368  1.610562
##   Has cats   14.147697  6.665550
##   Has dogs   14.226479 10.012282
##   <NA>        67.906456 81.711606
##   Total     100.000000 100.000000
```

```
tally(~ gender | cats_dogs_both, format = "percent", margin = TRUE, data = OKCupid)
```

```
##           cats_dogs_both
## gender  Has Both  Has cats  Has dogs  <NA>
##   F      60.85482  58.82759  48.88857  35.87435
##   M      39.14518  41.17241  51.11143  64.12565
##   Total 100.00000 100.00000 100.00000 100.00000
```

Example 3.1: Exploring Marginal Distributions

```
SuperBowl <-
  read_csv("http://nhorton.people.amherst.edu/is5/data/Watch_the_Super_bowl.csv",
    skip = 1)
```

```
## Parsed with column specification:
## cols(
##   Plan = col_character(),
##   Sex = col_character()
## )
```

```
tally(~ Plan + Sex, data = SuperBowl)
```

```
##           Sex
## Plan      Female Male
##   Commercials    156    81
```

```
## Game          200  279
## Wont Watch    160  132
```

Example 3.2: Exploring Percentages: Children and First-Class Ticket Holders First?

XX Consider using `y ~ x` format instead of using `y | x` in `tally()` calls.

```
Titanic <- read_csv("http://nhorton.people.amherst.edu/is5/data/Titanic.csv")
```

```
## Parsed with column specification:
## cols(
##   Name = col_character(),
##   Survived = col_character(),
##   Boarded = col_character(),
##   Class = col_character(),
##   MWC = col_character(),
##   Age = col_double(),
##   Adut_or_Chld = col_character(),
##   Sex = col_character(),
##   Paid = col_double(),
##   Ticket_No = col_character(),
##   Boat_or_Body = col_character(),
##   Job = col_character(),
##   Class_Dept = col_character(),
##   Class_Full = col_character()
## )
```

```
tally(~ Class + Survived, format = "percent", margin = TRUE, data = Titanic)
```

```
##           Survived
## Class      Alive      Dead      Total
## 1      9.103261  5.570652 14.673913
## 2      5.389493  7.518116 12.907609
## 3      8.152174 24.003623 32.155797
## Crew    9.601449 30.661232 40.262681
## Total 32.246377 67.753623 100.000000
```

```
tally(~ Survived | Class, format = "percent", margin = TRUE, data = Titanic)
```

```
##           Class
## Survived      1      2      3      Crew
## Alive 62.03704 41.75439 25.35211 23.84702
## Dead  37.96296 58.24561 74.64789 76.15298
## Total 100.00000 100.00000 100.00000 100.00000
```

```
tally(~ Class | Survived, format = "percent", margin = TRUE, data = Titanic)
```

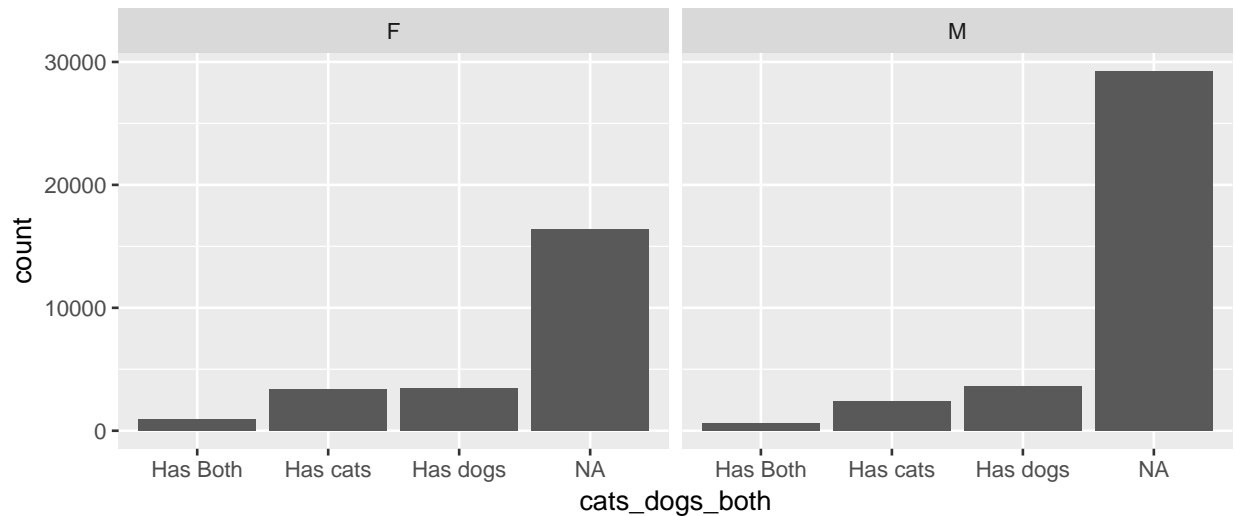
```
##           Survived
## Class      Alive      Dead
## 1      28.230337  8.221925
## 2      16.713483 11.096257
## 3      25.280899 35.427807
## Crew  29.775281 45.254011
## Total 100.000000 100.000000
```

Section 3.2: Conditional Distributions

See displays on 68-69.

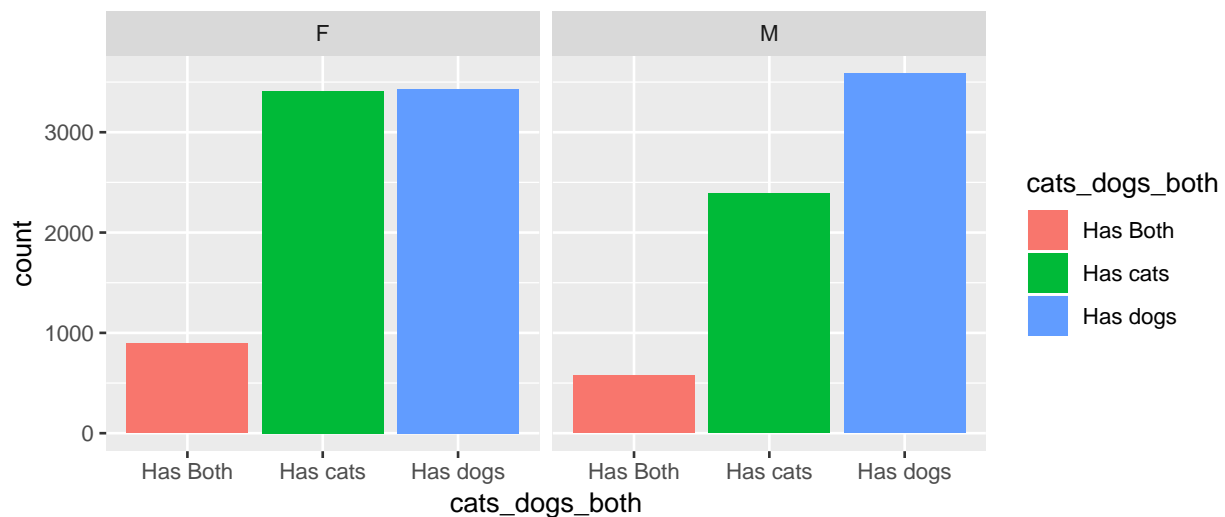
XX The colored bar graph is a bit confusing because the book's y axis is different so I don't know what can be done here.

```
gf_bar(~ cats_dogs_both | gender, data = OKCupid)
```



There are many who don't own either (Figure 3.2, page 69)

```
gf_bar(~ cats_dogs_both | gender, fill = ~ cats_dogs_both,  
data = filter(OKCupid, cats_dogs_both != "NA"))
```



Example 3.3: Finding Conditional Distributions: Watching the Super Bowl

```
tally(~ Plan + Sex, margin = TRUE, data = SuperBowl)
```

```
##           Sex  
## Plan      Female Male Total  
## Commercials  156   81  237
```

```
##   Game           200  279  479
##   Wont Watch     160  132  292
##   Total          516  492 1008
```

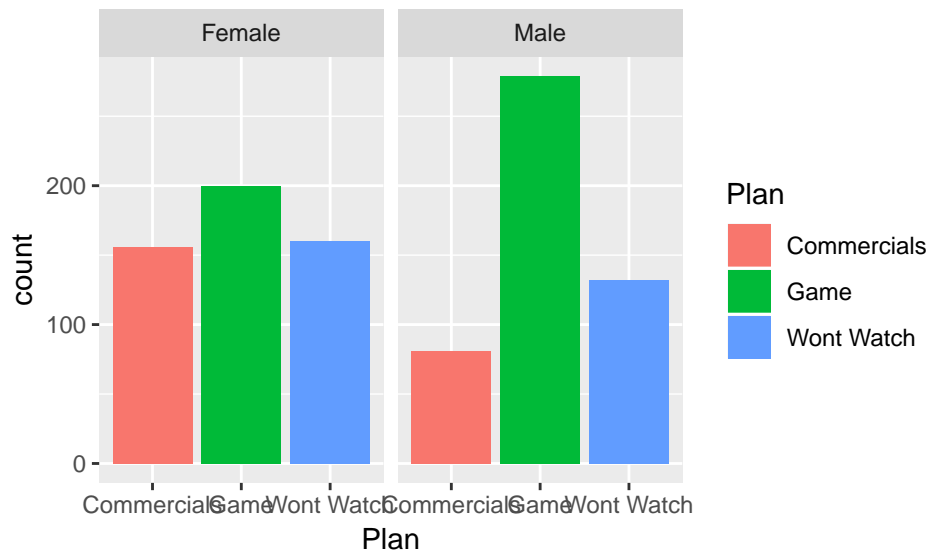
```
tally(~ Plan | Sex, format = "percent", data = SuperBowl)
```

```
##           Sex
## Plan      Female      Male
##   Commercials 30.23256 16.46341
##   Game        38.75969 56.70732
##   Wont Watch  31.00775 26.82927
```

Example 3.4: Looking for Associations Between Variables: Still Watching the Super Bowl

XX Consider this: `gf_bar(~ Plan, fill = ~ Sex, data = SuperBowl, position = "dodge")` XX The barplot still doesn't have percentages like the book(it has counts) but it is equally readable than the current version with the added bonus that it looks like the chart in the book.

```
gf_bar(~ Plan | Sex, fill = ~ Plan, format = "percent", data = SuperBowl)
```



Examining Contingency Tables

See displays on page 72.

XX Why not use `margins = TRUE` in this `tally()` call?

```
FishDiet <- read_csv("http://nhorton.people.amherst.edu/is5/data/Fish_diet.csv", skip = 1) %>%
  clean_names()
```

```
## Parsed with column specification:
## cols(
##   `Diet:Counts` = col_character(),
##   `Cancer:Counts` = col_character()
## )
```

```
tally(~ diet_counts + cancer_counts, data = FishDiet)
```

```
##           cancer_counts
```

```
## diet_counts    No  Yes
##      Large      507  42
##      Moderate 2769 209
##      Never     110  14
##      Small    2420 201
```

Random Matters

See display on page 74.

XX Consider changing the order in the `tally()` call into `~ Dream + Side` and adding `margins = TRUE`

```
Nightmares <- read_csv("http://nhorton.people.amherst.edu/is5/data/Nightmares.csv", skip = 1)
```

```
## Parsed with column specification:
## cols(
##   Side = col_character(),
##   Dream = col_character()
## )
```

```
Nightmares <- Nightmares %>%
  mutate(Dream = ifelse(Dream == "N", "Nightmare", "SweetDreams"))
tally(~ Side + Dream, data = Nightmares)
```

```
##      Dream
## Side Nightmare SweetDreams
##   L           9           13
##   R           6           35
```

Section 3.3: Displaying Contingency Tables

```
tally(~ Class + Survived, format = "count", data = Titanic)
```

```
##      Survived
## Class  Alive Dead
##   1      201 123
##   2      119 166
##   3      180 530
## Crew    212 677
```

```
tally(~ Class + Survived, format = "percent", data = Titanic)
```

```
##      Survived
## Class    Alive    Dead
##   1    9.103261 5.570652
##   2    5.389493 7.518116
##   3    8.152174 24.003623
## Crew  9.601449 30.661232
```

Figure 3.4, page 75

```
gf_percents(~ Class, fill = ~ Survived, position = position_dodge(), data = Titanic)
```

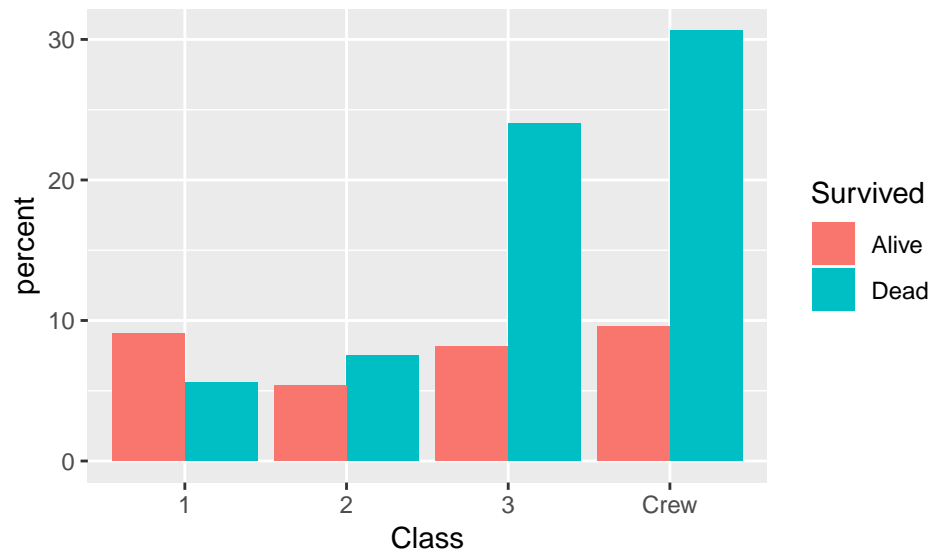


Figure 3.5

```
gf_percents(~ Survived, fill = ~ Class, position = "fill", data = Titanic)
```

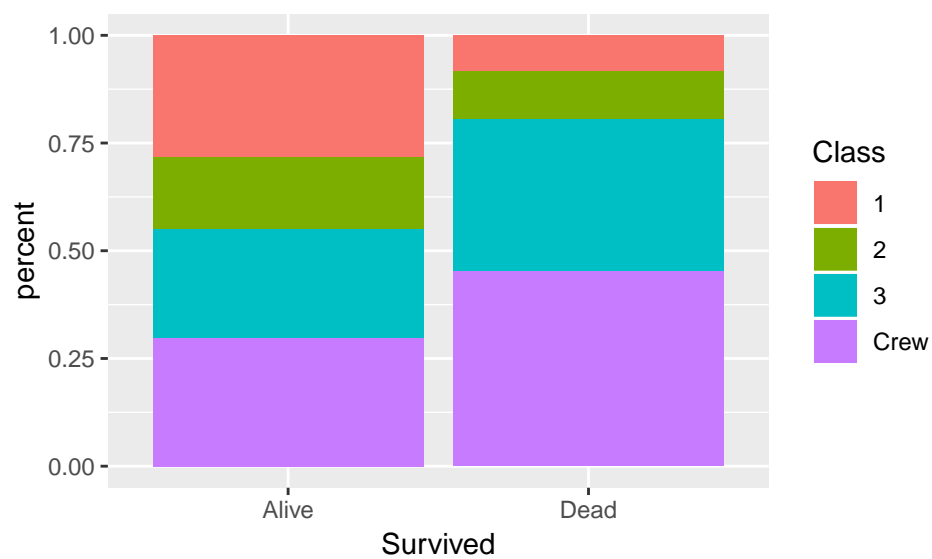
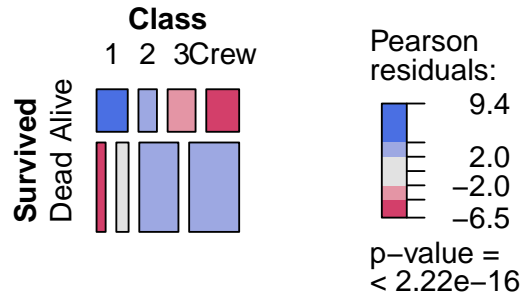


Figure 3.6, page 76

```
vcd::mosaic(tally(~ Survived + Class, data = Titanic),
  main = "Mosaic plot of Class by Survival",
  shade = TRUE)
```

Mosaic plot of Class by Survival



See the mosaic plots on page 77.

Section 3.4: Three Categorical Variables

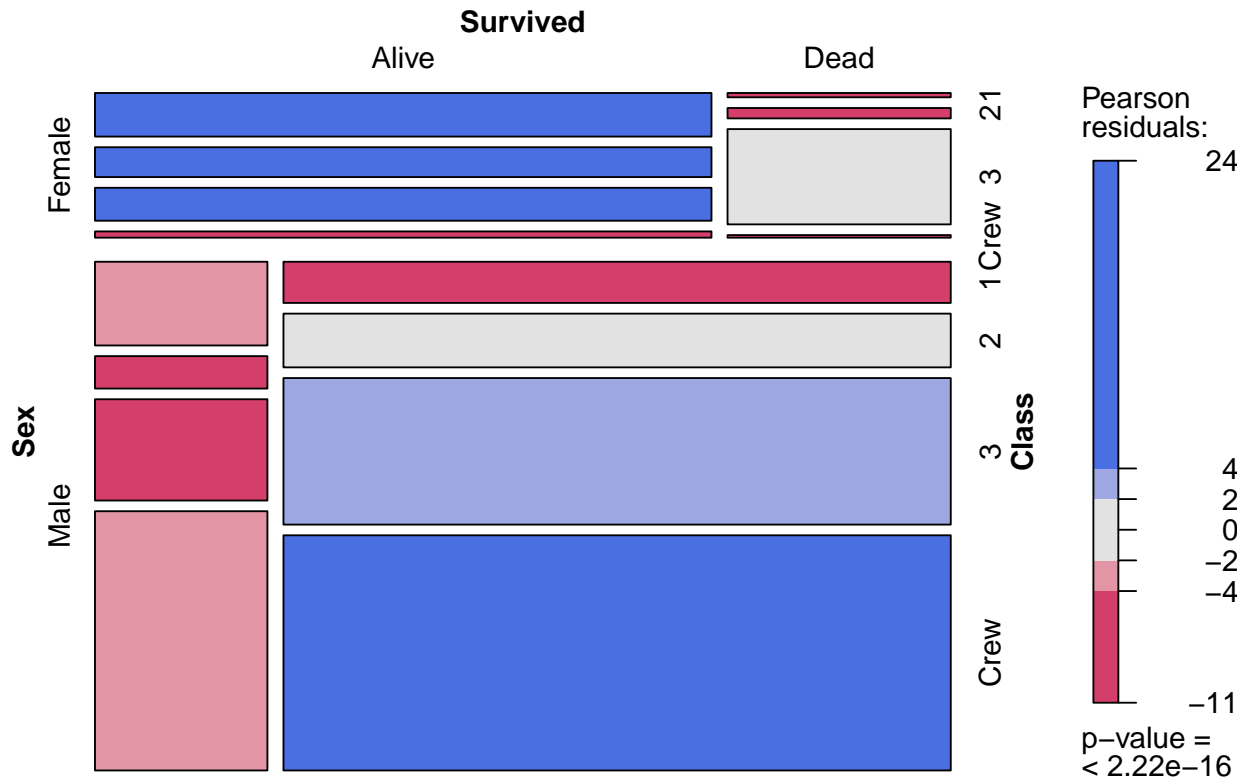
Consider using `useNA = "no"` option

```
tally(~ gender + cats_dogs_both + drugs_y_n, format = "percent", data = OKCupid)
```

```
## , , drugs_y_n = No
##
##      cats_dogs_both
## gender  Has Both  Has cats  Has dogs  <NA>
##      F  1.0243064  3.4199156  3.9437466 18.0187845
##      M  0.5922293  2.0819779  3.7769214 30.0719016
##
## , , drugs_y_n = Yes
##
##      cats_dogs_both
## gender  Has Both  Has cats  Has dogs  <NA>
##      F  0.2085314  0.8941828  0.6272626  2.9794972
##      M  0.1901807  0.8658225  0.9041923  6.9132342
##
## , , drugs_y_n = NA
##
##      cats_dogs_both
## gender  Has Both  Has cats  Has dogs  <NA>
##      F  0.2635837  1.3779757  1.1527618  6.3226732
##      M  0.1801712  1.0359842  1.3029044 11.8512587
```

Example 3.7: Looking for Associations Among Three Variables at Once

```
vcd::mosaic(tally(~ Sex + Survived + Class, data = Titanic), shade = TRUE)
```

Example 3.8: Simpson's Paradox: Gender Discrimination?

Here we demonstrate how to generate one of the tables on page 80.
XX Why do we have to have 825-512 can we just not put the difference?

```
# Create a dataframe from the counts
# http://mathemathinking.blogspot.com/2012/06/simpsons-paradox.html
Berk <- rbind(
  do(512) * data.frame(admit = TRUE, sex = "M", school = "A"),
  do(825 - 512) * data.frame(admit = FALSE, sex = "M", school = "A"),
  do(89) * data.frame(admit = TRUE, sex = "F", school = "A"),
  do(19) * data.frame(admit = FALSE, sex = "F", school = "A")
)
```

In this case, do(n) creates n observations with the specified values in data.frame(). The rbind() function can then be used to combine the data frames into one.

```
tally(~ sex + admit, data = Berk)

##      admit
## sex TRUE FALSE
##  M   512   313
##  F    89    19

tally(~ admit | sex, format = "percent", data = Berk)

##      sex
```

```
## admit      M      F
##   TRUE 62.06061 82.40741
##   FALSE 37.93939 17.59259
```