

# Comparing Groups: Tables and Visualizations

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- This presentation is based on (Chapman and Feit 2019, chap. 5)

- Use descriptive summaries by groups and visualize them to investigate differences between groups

- **age**: age of the consumer in years
- **gender**: if the consumer is male or female
- **income**: yearly disposable income of the consumer
- **kids**: number of children of the consumer
- **ownHome**: if the consumer owns a home
- **subscribe**: if the consumer is subscribed or not
- **Segment**: market segment assigned by a clustering algorithm (Chapman and Feit 2019, chap. 11), expert assignment or a segmentation typing tool

- **Segment:**

- **Moving up:** consumers experiencing upward mobility in terms of their socioeconomic status
- **Suburb mix:** consumers living in suburban areas
- **Travelers:** consumers who prioritize experiences and adventures
- **Urban Hip:** consumers interested in urban culture, artistic expression, and modern trends

## ● Import data

```
segmentation <- read_csv(file = "http://goo.gl/qw303p")
segmentation |> head(n = 5)
```

```
# A tibble: 5 x 7
```

	age	gender	income	kids	ownHome	subscribe	Segment
	<dbl>	<chr>	<dbl>	<dbl>	<chr>	<chr>	<chr>
1	47.3	Male	49483.	2	ownNo	subNo	Suburb mix
2	31.4	Male	35546.	1	ownYes	subNo	Suburb mix
3	43.2	Male	44169.	0	ownYes	subNo	Suburb mix
4	37.3	Female	81042.	1	ownNo	subNo	Suburb mix
5	41.0	Female	79353.	3	ownYes	subNo	Suburb mix

## ● Transform data

```
segmentation <- segmentation |>
  mutate(gender = factor(gender, ordered = FALSE),
         kids = as.integer(kids),
         ownHome = factor(ownHome, ordered = FALSE),
         subscribe = factor(subscribe, ordered = FALSE),
         Segment = factor(Segment, ordered = FALSE))

segmentation |> head(n = 5)
```

# A tibble: 5 x 7

	age	gender	income	kids	ownHome	subscribe	Segment
	<dbl>	<fct>	<dbl>	<int>	<fct>	<fct>	<fct>
1	47.3	Male	49483.	2	ownNo	subNo	Suburb mix
2	31.4	Male	35546.	1	ownYes	subNo	Suburb mix
3	43.2	Male	44169.	0	ownYes	subNo	Suburb mix
4	37.3	Female	81042.	1	ownNo	subNo	Suburb mix
5	41.0	Female	79353.	3	ownYes	subNo	Suburb mix



## • Basic Formula Syntax

- $\sim$  and  $+$ : operators
- $y$ : response variable
- $x, z$ : explanatory variables
- $y \sim x + z$ : a formula which means that  $y$  depends on  $x$  and  $z$ 
  - $+$  is used to indicate the addition of predictor variables to the right of the formula
  - Be careful not to confuse the arithmetic operator  $+$  with  $+$  within a formula

```
?`+` # Arithmetic Operators  
?formula # operators in a formula
```

## • Descriptives for n-Way Groups: the base R way

- Split data into  $n$  subsets and compute summary statistics

```
aggregate(x = income ~ Segment + ownHome,
          data = segmentation, FUN = mean)
```

	Segment	ownHome	income
1	Moving up	ownNo	54497.68
2	Suburb mix	ownNo	54932.83
3	Travelers	ownNo	63188.42
4	Urban hip	ownNo	21337.59
5	Moving up	ownYes	50216.37
6	Suburb mix	ownYes	55143.21
7	Travelers	ownYes	61889.12
8	Urban hip	ownYes	23059.27

## • Descriptives for n-Way Groups: the base R way

- Split data into  $n$  subsets and compute summary statistics

```
aggregate(x = kids ~ Segment + ownHome,
          data = segmentation, FUN = sum)
```

	Segment	ownHome	kids
1	Moving up	ownNo	82
2	Suburb mix	ownNo	90
3	Travelers	ownNo	0
4	Urban hip	ownNo	43
5	Moving up	ownYes	52
6	Suburb mix	ownYes	102
7	Travelers	ownYes	0
8	Urban hip	ownYes	12

## • Descriptives for n-Way Groups: the tidyverse way

- Split data into  $n$  subsets and compute summary statistics

```
segmentation |>
  group_by(Segment, ownHome) |>
  summarise(mean_income = mean(income))
```

```
# A tibble: 8 x 3
# Groups:   Segment [4]
  Segment    ownHome mean_income
  <fct>      <fct>      <dbl>
1 Moving up ownNo      54498.
2 Moving up ownYes     50216.
3 Suburb mix ownNo      54933.
4 Suburb mix ownYes     55143.
5 Travelers ownNo      63188.
6 Travelers ownYes     61889.
7 Urban hip  ownNo      21338.
8 Urban hip  ownYes     23059.
```

## • Descriptives for n-Way Groups: the tidyverse way

- Split data into  $n$  subsets and compute summary statistics

```
segmentation |>
  group_by(Segment, ownHome) |>
  summarise(sum_kids = sum(kids))
```

```
# A tibble: 8 x 3
# Groups:   Segment [4]
  Segment    ownHome sum_kids
  <fct>      <fct>    <int>
1 Moving up  ownNo       82
2 Moving up  ownYes      52
3 Suburb mix ownNo      90
4 Suburb mix ownYes    102
5 Travelers  ownNo        0
6 Travelers  ownYes        0
7 Urban hip  ownNo       43
8 Urban hip  ownYes      12
```

## • Basic Formula Syntax

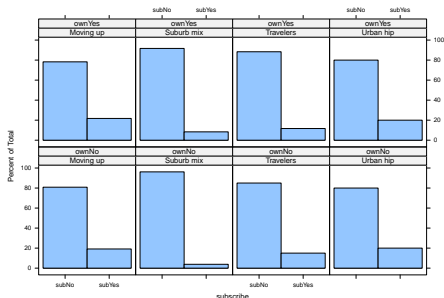
- $\sim$ ,  $+$  and  $|$ : operators
- $y$ : response variable
- $x$ : explanatory variable
- $z$ : grouping variable
- $y \sim x|z$ :  $y$  depends on  $x$  based on different groups defined by  $z$ 
  - $|$  is used to separate the grouping variable from the explanatory variable
  - Be careful not to confuse the logical operator  $|$  with  $|$  within a formula

```
?`|` # Logical Operators
```

```
?lattice::xyplot # operators in a formula (you need first to install the package lattice)
```

## • Visualization by group as frequencies: the lattice way

```
library(lattice)
histogram(~ subscribe | Segment + ownHome, data = segmentation)
```



## ● Visualization by group as frequencies: the tidyverse way

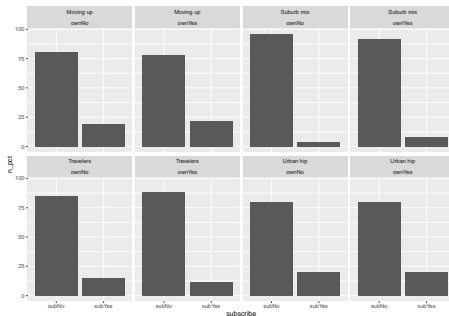
```
# Prepare data
subscriber_by_segment_home_ownership <- segmentation |>
  count(subscribe, Segment, ownHome) |>
  group_by(Segment, ownHome) |>
  mutate(n_pct = (n / sum(n)) * 100) |>
  ungroup()
subscriber_by_segment_home_ownership
```

```
# A tibble: 16 x 5
  subscribe Segment    ownHome     n n_pct
  <fct>      <fct>      <fct>   <int> <dbl>
1 subNo     Moving up   ownNo    38 80.9
2 subNo     Moving up   ownYes   18 78.3
3 subNo     Suburb mix  ownNo    50 96.2
4 subNo     Suburb mix  ownYes   44 91.7
5 subNo     Travelers  ownNo    17 85
6 subNo     Travelers  ownYes   53 88.3
7 subNo     Urban hip   ownNo    32 80
8 subNo     Urban hip   ownYes    8 80
9 subYes    Moving up   ownNo    9 19.1
10 subYes    Moving up   ownYes    5 21.7
11 subYes    Suburb mix  ownNo    2 3.85
12 subYes    Suburb mix  ownYes    4 8.33
13 subYes    Travelers  ownNo    3 15
14 subYes    Travelers  ownYes    7 11.7
15 subYes    Urban hip   ownNo    8 20
16 subYes    Urban hip   ownYes    2 20
```



## • Visualization by group as frequencies: the tidyverse way

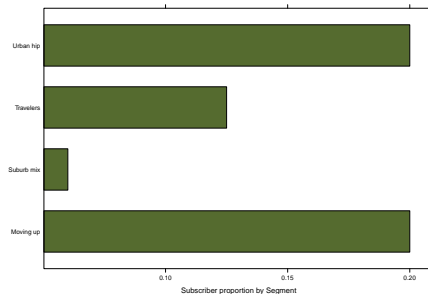
```
subscriber_by_segment_home_ownership |>
  ggplot() +
  geom_col(aes(x = subscribe, y=n_pct)) +
  facet_wrap(facets = vars(Segment,ownHome),
            nrow = 2, ncol = 4)
```



## • Visualization by group as proportions: the lattice way

```
# Prepare data
prop_table <- table(segmentation$subscribe,
                    segmentation$Segment) |>
  prop.table(margin = 2) |>
  _[2, ] # You can use _ as a placeholder. Check ?pipeOp

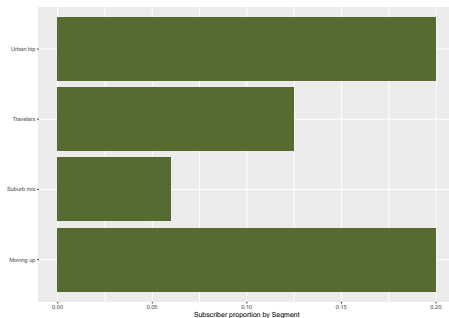
barchart(prop_table,
         xlab='Subscriber proportion by Segment', col='darkolivegreen')
```



## ● Visualization by group as proportions: the tidyverse way

```
# Prepare data
prop_table <- segmentation |>
  count(subscribe, Segment) |>
  group_by(Segment) |>
  mutate(n_pct = n / sum(n)) |>
  filter(subscribe == 'subYes')

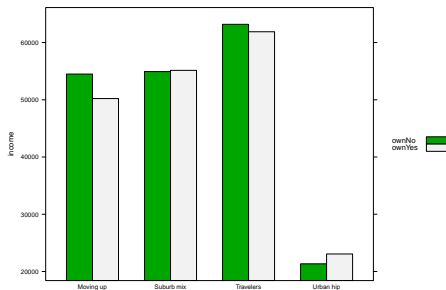
prop_table |> ggplot() +
  geom_col(aes(x=n_pct, y=Segment),
           fill='darkolivegreen') +
  labs(x='Subscriber proportion by Segment',
       y=NULL)
```



## • Visualization by group with continuous data: the lattice way

```
# Prepare data
seg_income_agg <- aggregate(income ~ Segment + ownHome,
                             data=segmentation, FUN = mean)

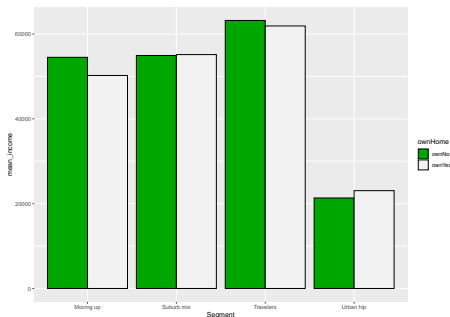
barchart(income ~ Segment, data = seg_income_agg,
          groups=ownHome, auto.key=TRUE, # Add groups
          par.settings=simpleTheme(col=terrain.colors(n = 2))) # Change default colors
```



## • Visualization by group with continuous data: the tidyverse way

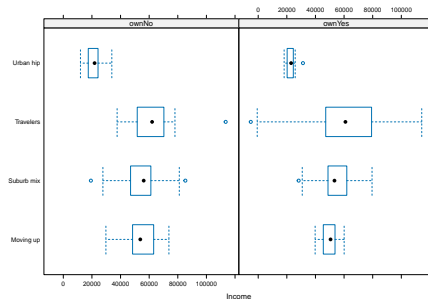
```
# Prepare data
seg_income_agg <- segmentation |>
  group_by(Segment, ownHome) |>
  summarise(mean_income = mean(income)) |>
  ungroup()

seg_income_agg |> ggplot() +
  geom_col(aes(x=Segment, y=mean_income, fill=ownHome),
           position = position_dodge(), color='black') +
  scale_fill_manual(values=terrain.colors(n = 2))
```



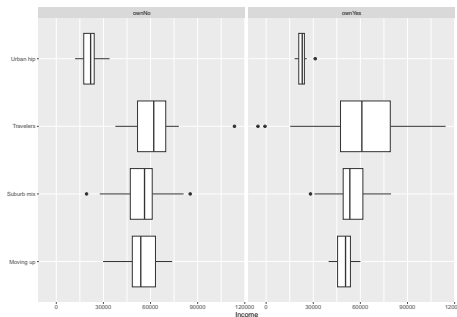
## • Visualization by group with continuous data: the lattice way

```
bwplot(Segment ~ income | ownHome,
      data = segmentation,
      xlab = 'Income')
```



## ● Visualization by group with continuous data: the lattice way

```
segmentation |> ggplot() +
  geom_boxplot(aes(x=income, y=Segment)) +
  facet_wrap(facets = vars(ownHome)) +
  labs(x='Income',
       y=NULL)
```



- To my family that supports me
- To the taxpayers of Colombia and the **UMNG students** who pay my salary
- To the **Business Science** and **R4DS Online Learning** communities where I learn **R** and  **$\pi$ -thon**
- To the **R Core Team**, the creators of **RStudio IDE**, **Quarto** and the authors and maintainers of the packages **tidyverse** and **tinytex** for allowing me to access these tools without paying for a license
- To the **Linux kernel community** for allowing me the possibility to use some **Linux distributions** as my main **OS** without paying for a license



# References I

Chapman, Chris, and Elea McDonnell Feit. 2019. *R For Marketing Research and Analytics*. 2nd ed. 2019. Use R! Cham: Springer International Publishing : Imprint: Springer.  
<https://doi-org.ezproxy.umng.edu.co/10.1007/978-3-030-14316-9>.