#### Introduction to R

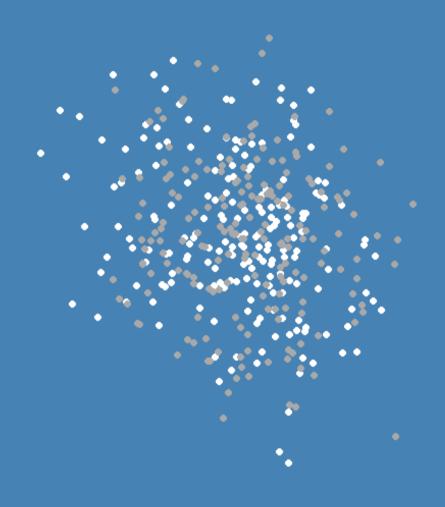
#### 4.2 Plotting Anything

**Example: Scatterplots** 

Lion Behrens, M.Sc.

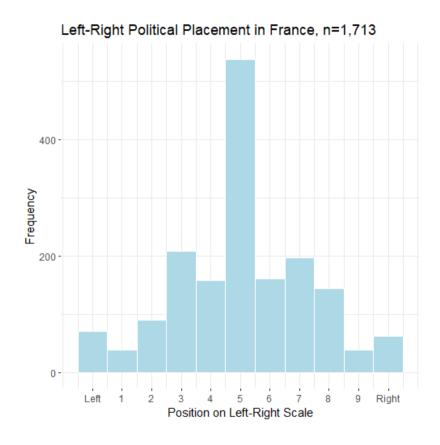


University of Mannheim Chair of Social Data Science and Methodology Chair of Quantitative Methods in the Social Sciences



What we did so far

# Our first plot: A Histogram



# Steps for Plotting: Summary

#### A. Specify necessary layers

- 1. Data.
- 2. Aesthetics (axis definitions).
- 3. Geometric objects to use (plot type).

```
ggplot(data = ess10,
    aes(x = left_right)) +
    geom_histogram()
```

#### B. Customize your plot

- 1. Shape, color and size of geometric objects.
- 2. Fine-tune your x-scale and y-scale (axis ticks, ticks labels).
- 3. Label your axes, assign a title to your plot.
- 4. Choose a ggtheme.
- 5. Customize your theme.

Now: Plotting anything!

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### ggplot2: Overview of Different Geoms

#### Histogram

geom\_histogram()

#### Densities

geom\_density(), stat\_density()

#### Straight lines

geom\_abline(), geom\_hline(), geom\_vline()

#### Curves

geom\_curve()

#### Bar charts

geom\_bar(), geom\_col(), stat\_count()

#### Boxplots

geom\_boxplot(), stat\_boxplot()

#### Dotplots

geom\_dotplot()

#### Scatterplots

geom\_point()

#### Violin plots

geom\_violin(), stat\_ydensity()

#### Prerequisite: Data Wrangling Pipeline (I/III)

```
library(tidyverse)
ess10 <- haven::read dta("./dat/ESS10.dta")
ess10 <- ess10 %>% # subset variables
  select(country = cntry, # sociodemographics
         gender = gndr.
         education years = eduyrs.
         trust social = ppltrst, # multidimensional trust
         trust parliament = trstprl,
         trust legalSys = trstlgl,
         trust police = trstplc,
         trust politicians = trstplt,
         trust parties = trstprt,
         trust EP = trstep.
         trust UN = trstun,
         left right = lrscale, # attitudes
         life satisfaction = stflife,
         pol interest = polintr,
         voted = vote, # turnout
         party choice = prtvtefr # party choice
         ) %>%
 mutate_at(c("country", "gender", "voted", "party_choice"), as.character) %>% # change types
 mutate at("pol interest", as.numeric) %>% # change types
  filter(country == "FR") # subset cases (only include France)
```

#### Prerequisite: Data Wrangling Pipeline (II/III)

```
ess10 <- ess10 %>%
 mutate(gender = recode factor(gender,
                                `1` = "Male".
                                `2` = "Female").
         voted = recode_factor(voted,
                               `1` = "Yes".
                               `2` = "No".
                               `3` = "Not eligible"),
         party_choice = recode_factor(party_choice,
                                      `1` = "Lutte Ouvriére",
                                       `2` = "Nouv. Parti Anti-Capitaliste",
                                      `3` = "Parti Communiste Français",
                                      `4` = "La France Insoumise",
                                      `5` = "Parti Socialiste",
                                      `6` = "Europe Ecologie Les Verts",
                                      `7` = "La République en Marche",
                                      `8` = "Mouvement Démocrate",
                                      `9` = "Les Républicains",
                                      `10` = "Debout la France",
                                      `11` = "Front National",
                                      `12` = "Other",
                                      `13` = "Blank",
                                      `14` = "Null")
```

#### Prerequisite: Data Wrangling Pipeline (III/III)

The European Social Survey Wave 10 asked people about their

- life satisfaction ranging from extremely dissatisfied to extremely satisfied
- trust in politicians ranging from no trust at all to complete trust

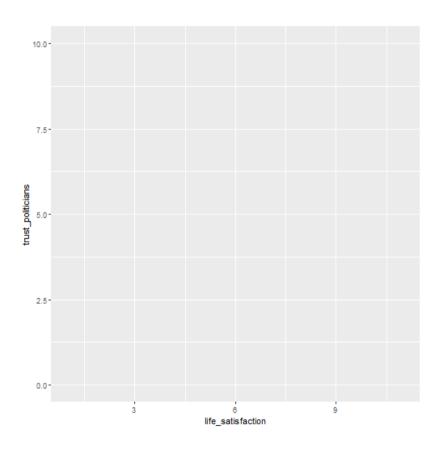
#### Measurement Life Satisfaction

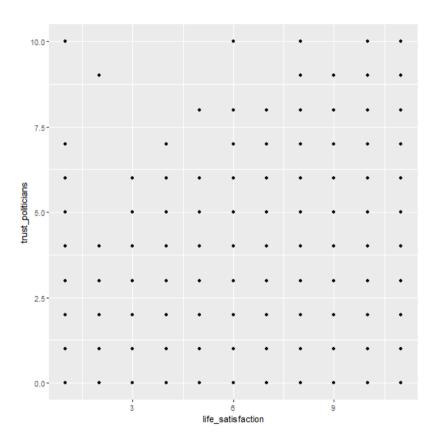
All things considered, how satisfied are you with your life as a whole nowadays? Please answer using this card, where 0 means mextremely dissatisfied and 10 means extremely satisfied.

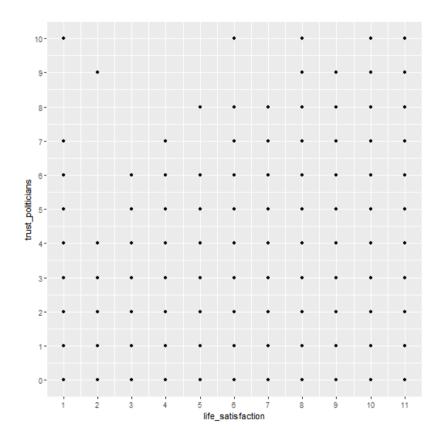
#### Measurement Trust in Politicians

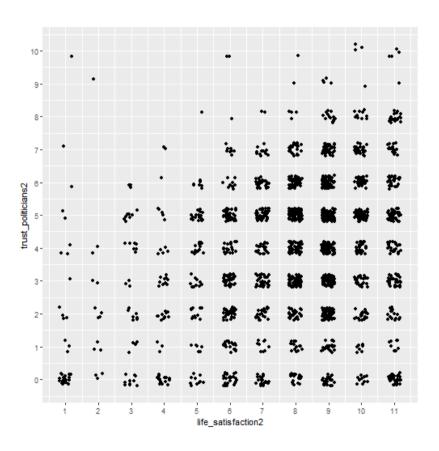
Using this card, please tell me on a score of 0-10 how much you personally trust each of the institutions I read out. 0 means you do not trust an institution at all, and 10 means you have complete trust.

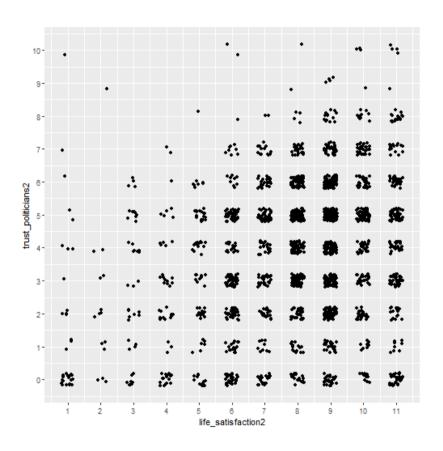
```
table(ess10$life_satisfaction)
##
##
       15 43 53 81 191 187 351 502 227 222
table(ess10$trust_politicians)
##
##
                                        9 10
## 188 100 220 280 255 428 240 126 51 8 11
cor(ess10$life_satisfaction, ess10$trust_politicians)
## [1] 0.26955
```

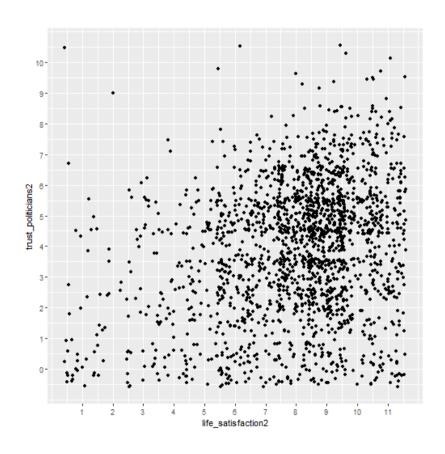


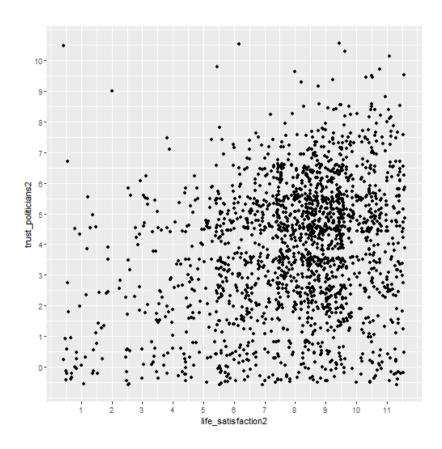


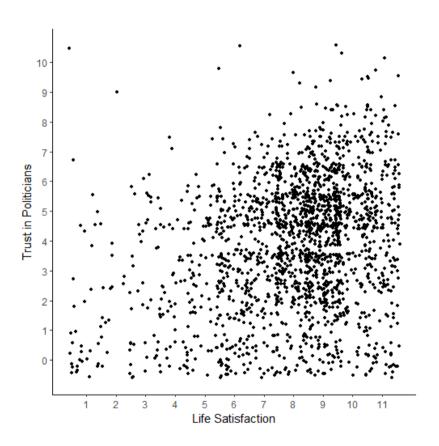


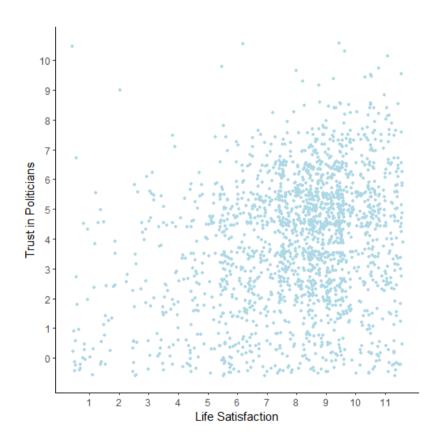


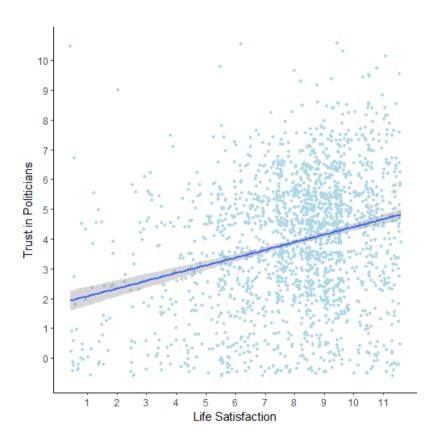


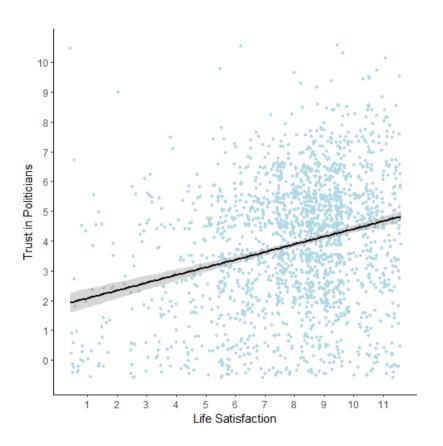










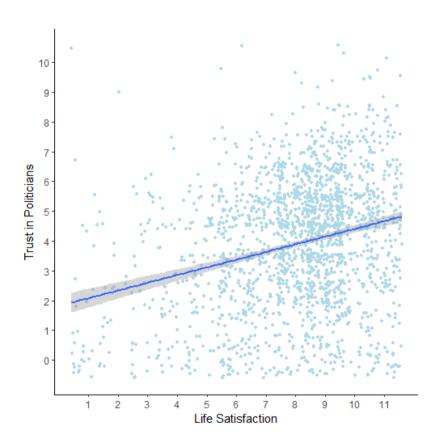


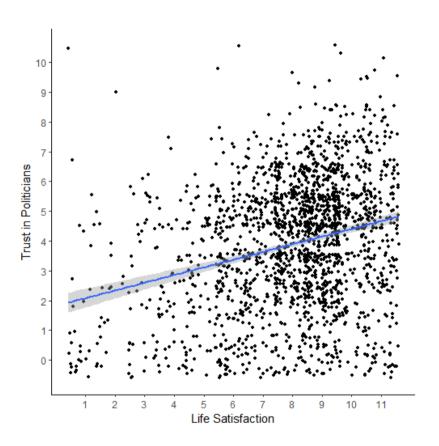
What if the relationship between life satisfaction and trust in politicians is moderated by political interest?

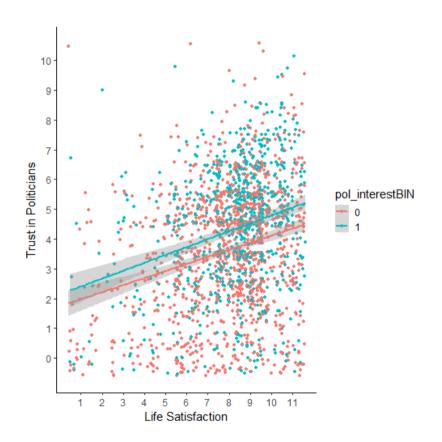
- Those who are politically interested might trust politicians either way
- For those that are not politically interested, life satisfaction might matter more

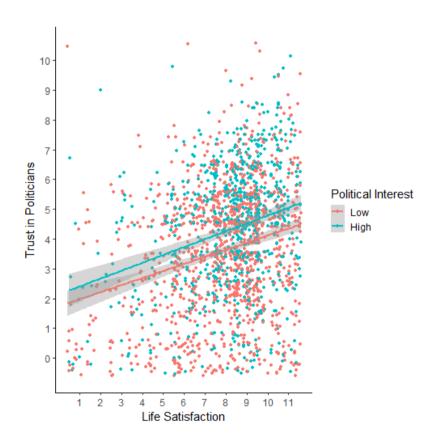
table(ess10\$pol\_interest, ess10\$pol\_interestBIN)

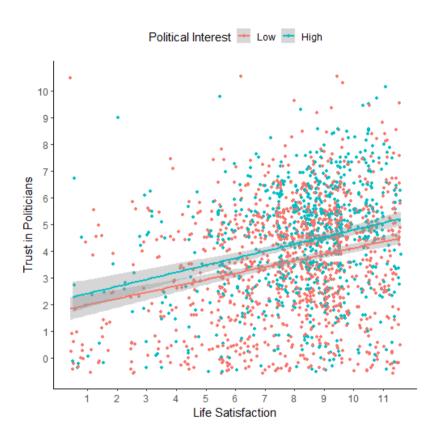
```
## 0 1
## 1 373 0
## 2 770 0
## 3 0 469
## 4 0 295
```



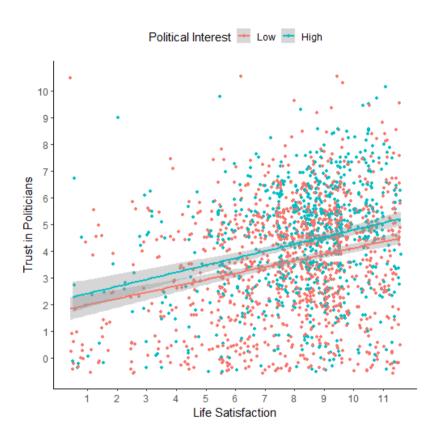








```
ggplot(data = ess10,
       aes(x = life satisfaction2,
           y = trust politicians2,
           color = pol interestBIN)) +
  geom point() +
  geom smooth(method = "lm") +
  scale x continuous(breaks = seq(1,11,1)) +
  scale_y_continuous(breaks = seq(0,10,1)) +
  labs(x = "Life Satisfaction",
       y = "Trust in Politicians",
       color="Political Interest") +
  theme classic(base size = 15) +
  scale color discrete(
    labels = c("Low", "High")) +
  theme(legend.position = "top")
```



```
ggplot(data = ess10,
       aes(x = life satisfaction2,
           y = trust politicians2,
           color = pol interestBIN)) +
  geom point() +
  geom smooth(method = "lm") +
  scale x continuous(breaks = seq(1,11,1)) +
  scale_y_continuous(breaks = seq(0,10,1)) +
  labs(x = "Life Satisfaction",
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       color="Political Interest") +
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```

#### References

Parts of this course are inspired by the following resources:

- Wickham, Hadley and Garrett Grolemund, 2017. R for Data Science Import, Tidy, Transform, Visualize, and Model Data. O'Reilly.
- Bahnsen, Oke and Guido Ropers, 2022. *Introduction to R for Quantitative Social Science*. Course held as part of the GESIS Workshop Series.
- Breuer, Johannes and Stefan Jünger, 2021. *Introduction to R for Data Analysis*. Course held as part of the GESIS Summer School in Survey Methodology.
- Teaching material developed by Verena Kunz, David Weyrauch, Oliver Rittmann and Viktoriia Semenova.