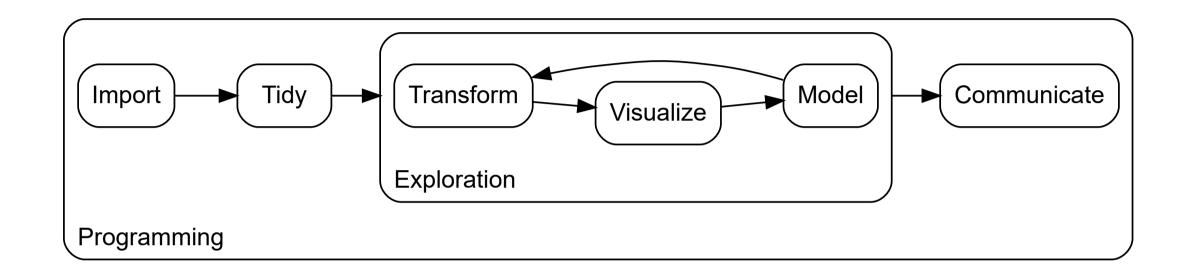
Data Analysis using R

Visualization

Sven Werenbeck-Ueding

16.09.2024



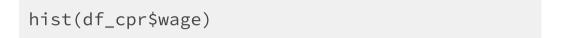
Source: Wickham and Grolemund (2016)

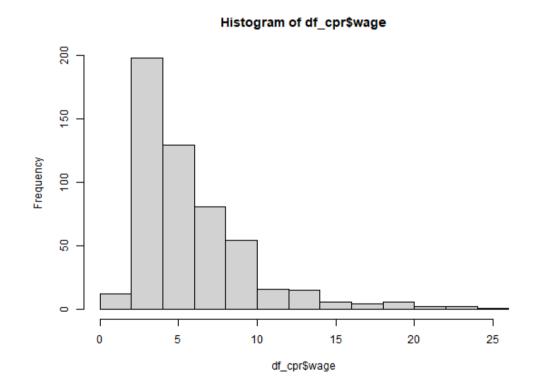
Prerequisites

```
# Load required packages
# install.packages("wooldridge")
library(tidyverse)
library(RColorBrewer)
# Load the CPR data set from the wooldridge package
df_cpr <- wooldridge::wage1 %>%
 mutate(
    sex = ifelse(female == 1, "Female", "Male"),
    region = case_when(
      west == 1 ~ "West",
      south == 1 ~ "South",
      northcen == 1 ~ "North Central",
     T ~ "East"
```

Graphics in base R

- Functions for graphics in base R are fast, but limited
- Useful for simple tasks such as quickly plotting histograms
- Graphics content can only be written on top of each other
- Elements of a plot cannot be modified or deleted







Graphics in base R



- base R is rather limited when it comes to creating beautiful, customized plots
- Layered approach: Adds plot components on top of each other
 - Opens up for more flexibility
 - Layers can overwrite each other to allow for post-hoc editing of components



ggplot2

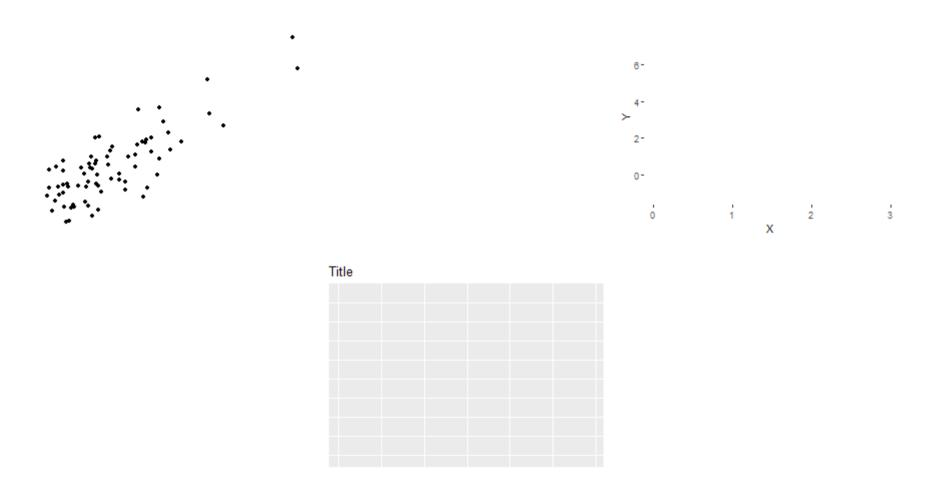
ggplot2 is a system for declaratively creating graphics, based on "The Grammar of Graphics" by Wilkinson (2005). You provide the data, tell ggplot2 how to map variables to aesthetics, what graphical primitives to use, and it takes care of the details.

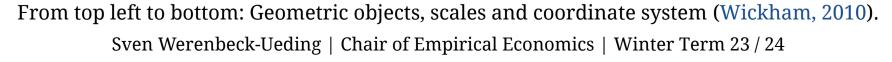
Wickham (2016)

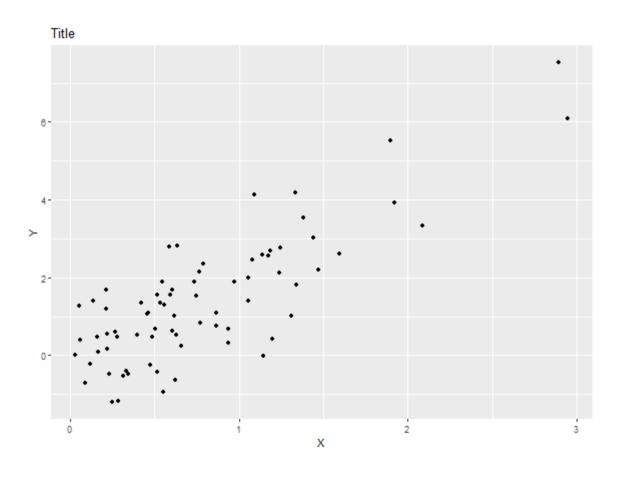


Component	Description
Layer	Geometric elements (geoms, e. g. points and lines) and statistical transformations (stats that summarize the data)
Scale	Mapping of data to aesthetics (e. g. color, shape or size) of geoms and draws legends and axes
Coord	Mapping of data to the plane of the graphic and provides axes/gridlines (usually the Cartesian coordinate system)
Facet	Plots subsets of the data
Theme	Styling of the plot such as fonts and background









Source: Wickham (2010)



- Graphics template consists of seven parameters (in angle brackets)
- *Any* plot can be generated from this template
- ggplot2 provides defaults for everything except for data, mappings and geoms
- Syntax is similar to using the pipe operator

Source: Wickham and Grolemund (2016)

Initialize a Plot

?ggplot

ggplot() initializes a ggplot object. It can be used to declare the input data frame for a graphic and to specify the set of plot aesthetics intended to be common throughout all subsequent layers unless specifically overridden.

Initialize a Plot – Function Call and Arguments

```
ggplot(
  data = NULL,
  mapping = aes(),
    ...
)
```



Initialize a Plot – Function Call and Arguments

```
ggplot(
  data = NULL,
  mapping = aes(),
   ...
)
```

data

- Default data set to use for the plot
- If necessary will be converted to a data.frame
- If not specified, must be supplied in each layer added to the plot

Initialize a Plot – Function Call and Arguments

```
ggplot(
  data = NULL,
  mapping = aes(),
    ...
)
```

mapping

- Default list of aesthetic mappings to use for plotting
- If not specified, must be supplied in each layer added to the plot

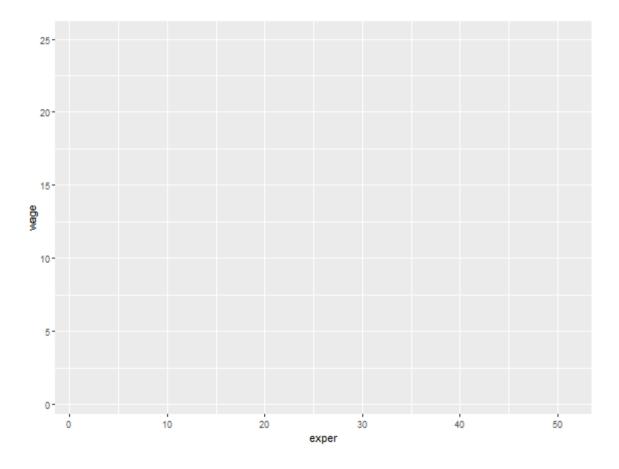
Initialize a Plot

Code Plot

Initialize a ggplot object using the CPR data set and map educ to the x-axis and wage to the y-axis.

Initialize a Plot

Code Plot



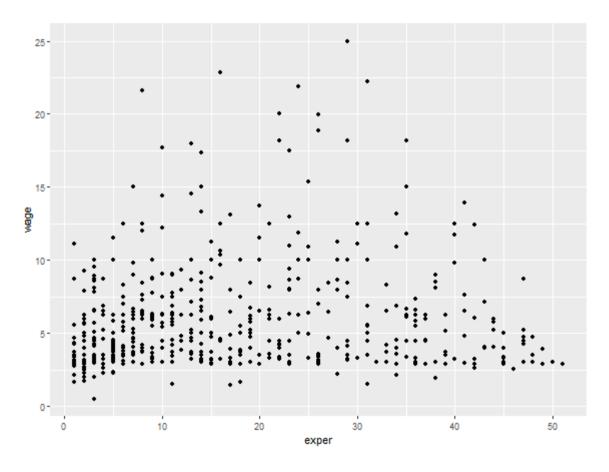
Add Geometric Objects

Code Plot

Create a scatterplot of wage vs. educ by adding points as geoms.

Add Geometric Objects

Code Plot



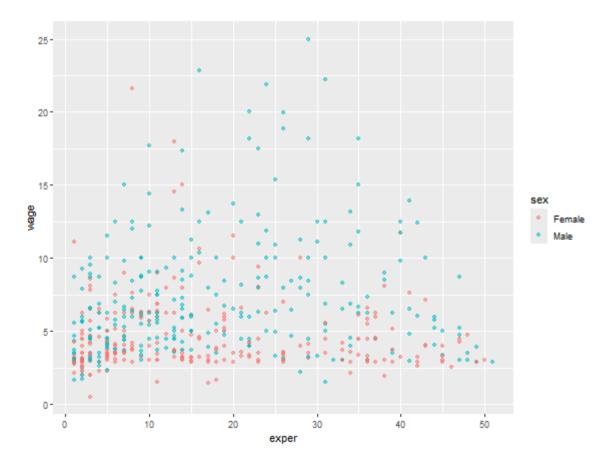
Map Data to Geom Aesthetics

Code Plot

Map sector to the color aesthetics and change the transparency of points to handle overlapping of points.

Map Data to Geom Aesthetics

Code Plot



Add Titles and Labels

Code Plot

Add a title to the plot and modify axis labels and legend title.

```
ggplot(df_cpr,
    aes(x = exper,
        y = wage,
        color = sex)) +
geom_point(alpha = .5) +
labs(title = "Hourly Wage ~ Years of Experience",
    x = "Years of Experience",
    y = "Wage [USD/hour]",
    color = "Sex")
```

Add Titles and Labels

Code

Plot



Change the Styling

Code Plot

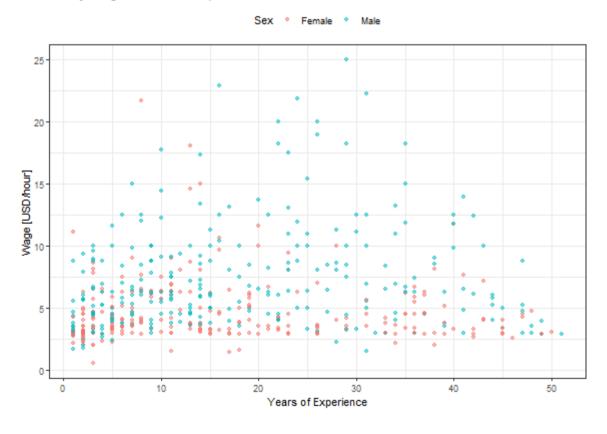
Change the theme to theme_bw() and adjust the position of the legend.

Change the Styling

Code

Plot

Hourly Wage ~ Years of Experience



Create Facets of Plots

Code Plot

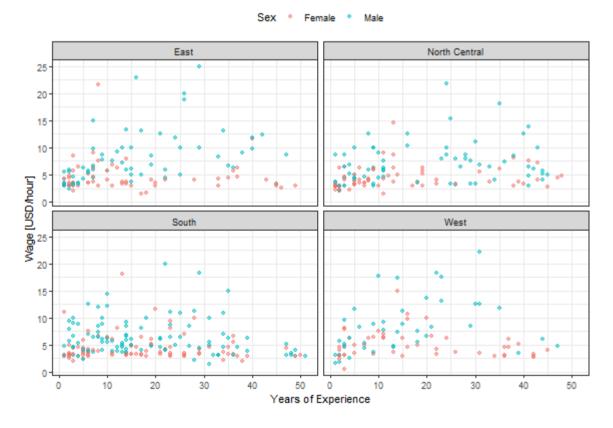
Create facets to display subplots for region.

Create Facets of Plots

Code

Plot

Hourly Wage ~ Years of Experience



i

For brevity, we will use the same ggplot object (short-hand gg) over the next slides. This way, we need not specify every time which data to use for plotting and only have to add geoms.

```
gg <- df_cpr %>%
  ggplot()
```

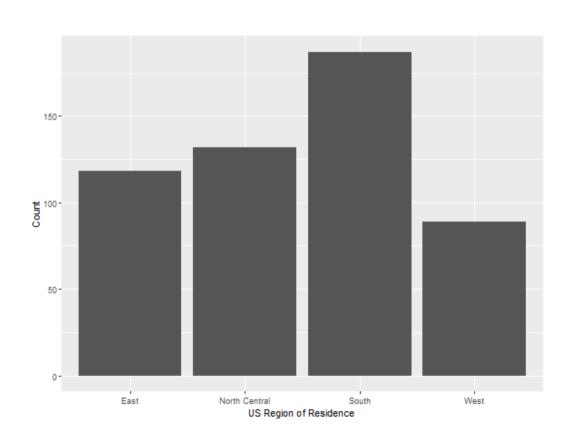
Univariate Graphs



Categorical Data

Count Bars

Frequency Bars

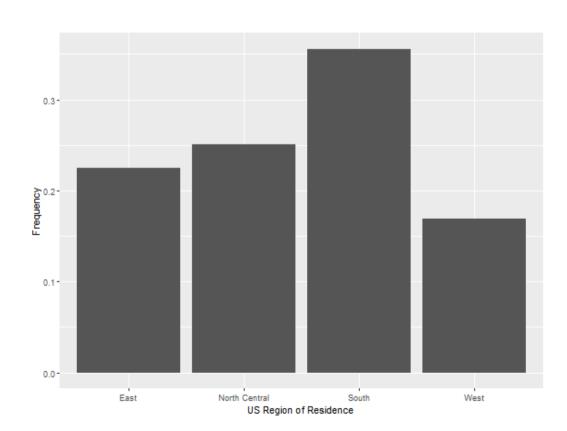


Bar Plot

Categorical Data

Count Bars

Frequency Bars



Bar Plot with Frequencies

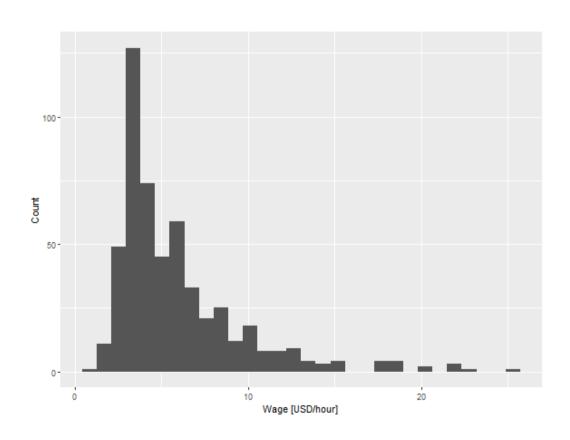
By mapping ...count.../sum(...count...) to the y-axis, the scale of the y-axis is changed to display frequencies:

```
freq_bar_plot <- gg +
  geom_bar(
    aes(x = region,
        y = ..count../sum(..count..))
) +
labs(x = "US Region of Residence",
    y = "Frequency")</pre>
```

Numerical Data

Histogram

Density

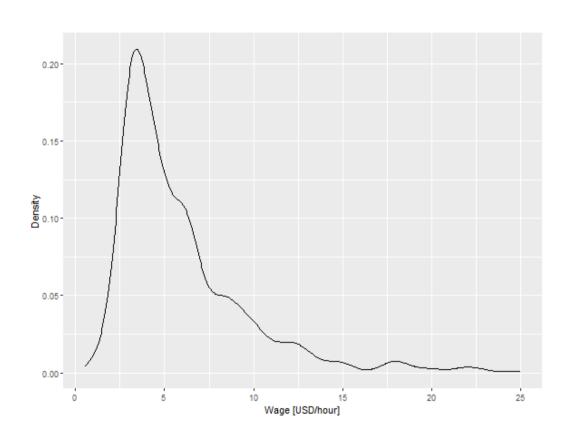


Histogram

Numerical Data

Histogram

Density



Kernel Density Plot

Bivariate Graphs

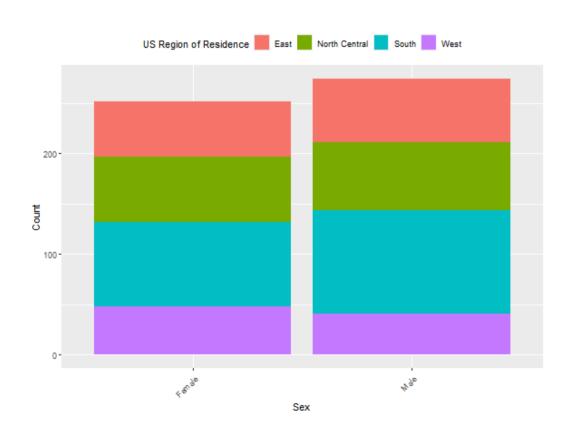


Categorical vs. Categorical

Stacked Bars

Grouped Bars

Segmented Bars



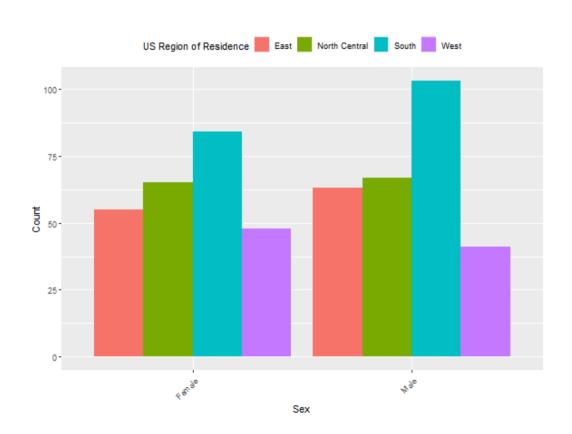
Stacked Bar Plot

Categorical vs. Categorical

Stacked Bars

Grouped Bars

Segmented Bars



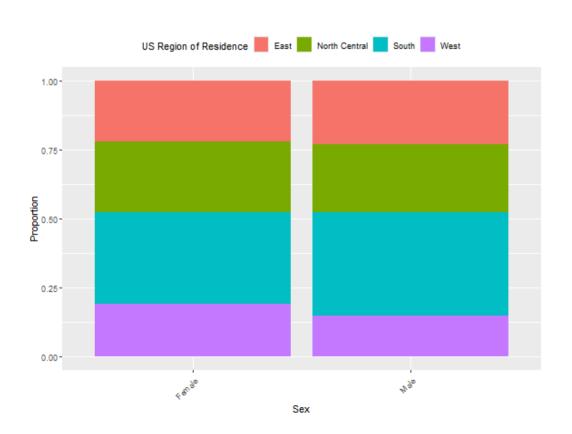
Grouped Bar Plot

Categorical vs. Categorical

Stacked Bars

Grouped Bars

Segmented Bars

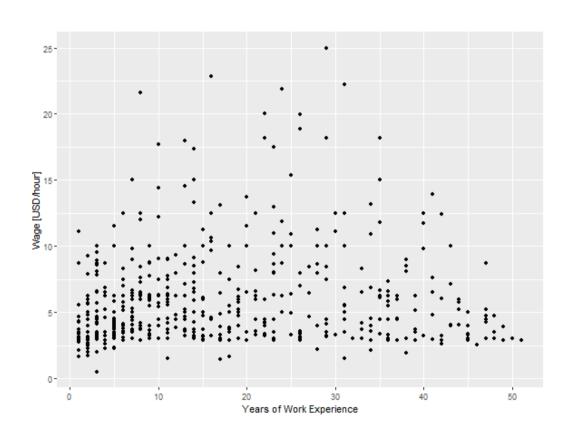


Segmented Bar Plot

Numerical vs. Numerical

Scatter

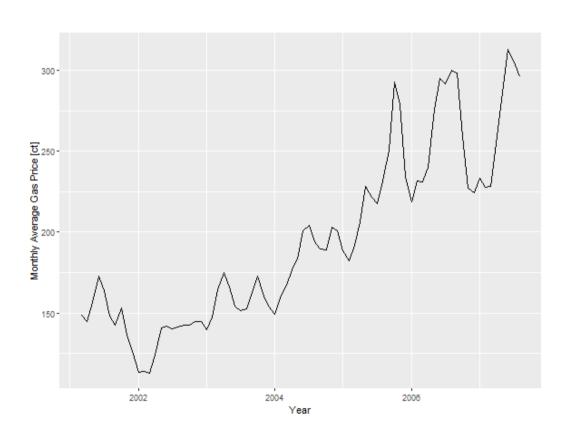
Line



Scatterplot

Numerical vs. Numerical

Scatter Line



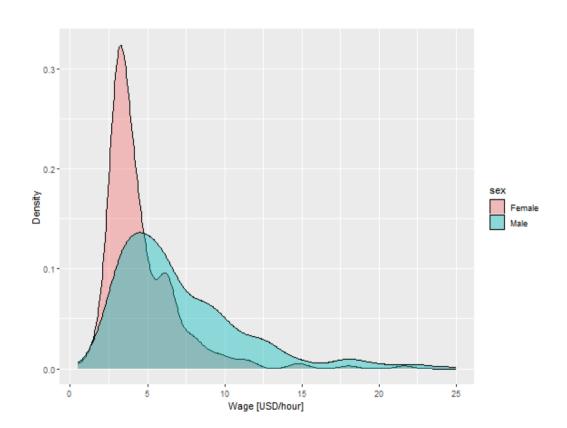
Line Plot

Categorical vs. Numerical

Grouped Density

Error Bars

Boxplot



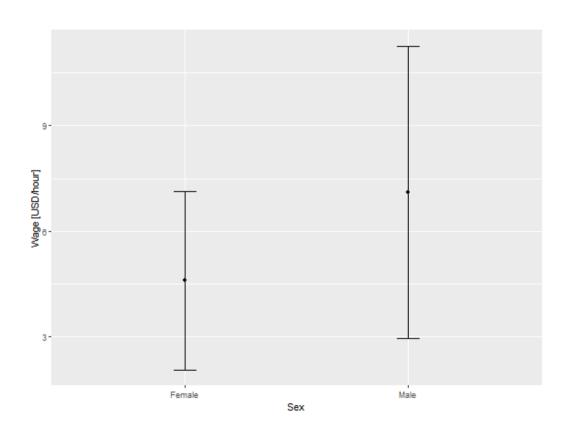
Grouped Kernel Density Plot

Categorical vs. Numerical

Grouped Density

Error Bars

Boxplot



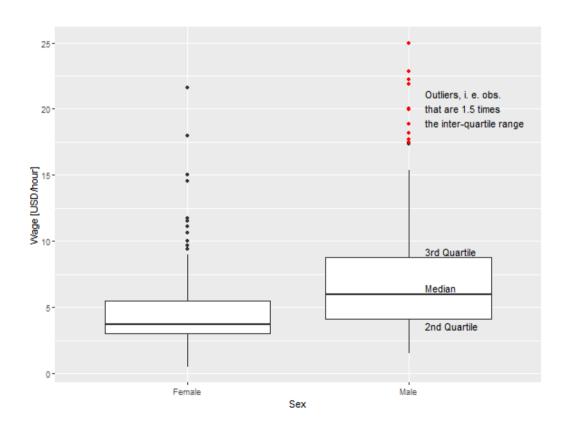
Mean with Standard Error Bars

Categorical vs. Numerical

Grouped Density

Error Bars

Boxplot



Boxplot

```
boxplot <- gg +
  geom_boxplot(aes(x = sex, y = wage)) +
  labs(x = "Sex", y = "Wage [USD/hour]")</pre>
```

Multivariate Graphs



Creating Multivariate Graphs

Including multiple variables into one graph can be achieved by combining the geometric objects seen on the previous slides, through mapping of different aesthetics and facetting.

Mapping

Variables can be mapped to various aesthetics of geometric objects, such as

- shape (only categorical)
- size (only numerical)
- color (both)
- ..



Creating Multivariate Graphs

Including multiple variables into one graph can be achieved by combining the geometric objects seen on the previous slides, through mapping of different aesthetics and facetting.

Facets

Creating subplots for groups of variables, using facet_wrap() for one-dimensional and facet_grid() for two-dimensional sequences of panels.

Variables to create facets for can be passed to facet_*() using R formulas, e. g. ~ sector (one-dimensional) or female ~ sector (two-dimensional). Formulas of type ~ sector + region can be used to create groups for combinations of variables.



Correlation Matrix

Plot

Code



Correlation Matrix

Plot

Code

Customizing Plots



Color Schemes



Choosing an appropriate color scheme for your graph is an important step in conveying the information!



Make sure that your plot is readable. Avoid light on light and dark on dark colors!



Avoid red and green in the same color palette. Colorblind people will thank you!



Be consistent: Choose a color palette and stick to it!

RColorBrewer

?RColorBrewer

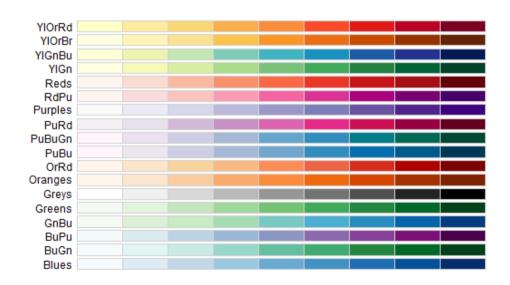
The RColorBrewer package provides color palettes from ColorBrewer 2.0 for creating thematic graphs. Run the display.brewer.all() function for a display of available color palettes.



Color Palettes for Numerical Data

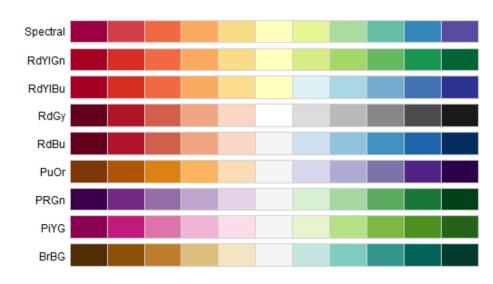
Sequential

Sequential color palettes assign light colors to low values of the data and dark colors to high values, e. g.



Diverging

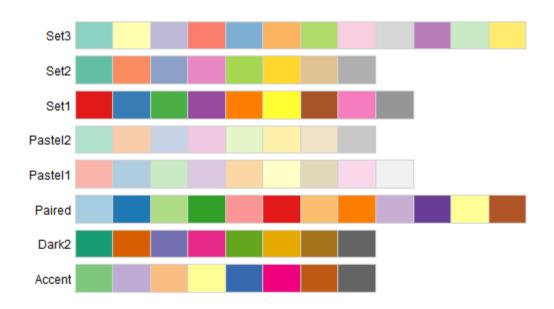
Diverging color palettes emphasize differences between extreme values by assigning contrasting dark colors to low and high values while assigning light colors to mid-range values, e. g.



Color Palettes for Categorical Data

Qualitative

Qualitative color palettes are used, when there is no difference in magnitude between values, i. e. for categorical data. The primary concern is to visualize differences between classes, e. g.





Change Discrete Color Palettes

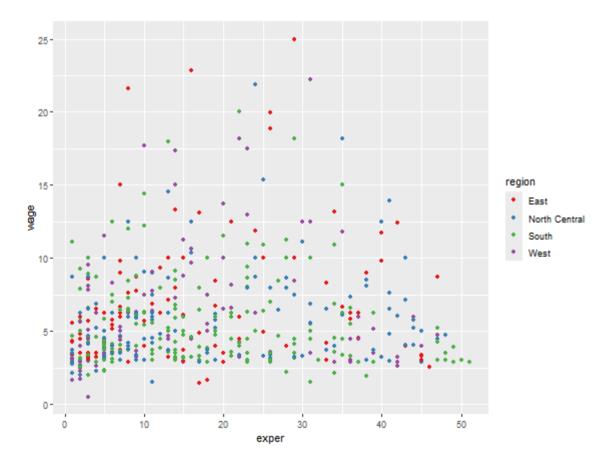
Code Plot

- Discrete color palettes for ggplot aesthetics, such as color or fill, can be changed to RColorBrewer palettes using the scale_*_brewer() functions, where the name of the palette can be passed as a string to the argument palette
- RColorBrewer supports only discrete color palettes

```
ggplot(
  data = df_cpr,
  mapping = aes(
    x = exper,
    y = wage,
    color = region
)
) +
  geom_point() +
  scale_color_brewer(palette = "Set1")
```

Change Discrete Color Palettes

Code Plot



Change Continuous Color Palettes

Code

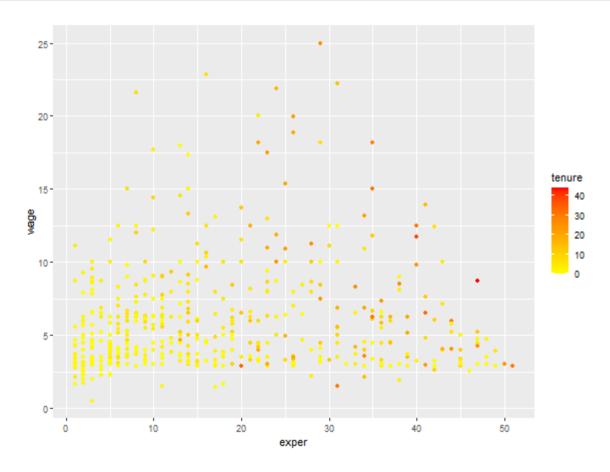
Plot

For continuous color palettes, it is best to use ggplot2's built-in functions:

- scale_*_gradient(): Creates a two color gradient based on strings passed to arguments low and high
- scale_*_gradient2(): Creates a diverging color gradient based on strings passed to arguments low, mid and high. The midpoint defaults to 0 and, in most cases, has to be set manually to ensure a meaningful color gradient.

Change Continuous Color Palettes

Code Plot



?theme

Themes are used to modify non-data components of plots, such as titles, labels, ticks, fonts for text elements, background etc. ggplot2 provides out-of-the-box themes for quick and easy adjustments of the overall look:

theme_gray()

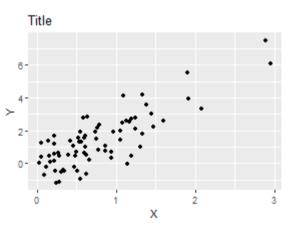
theme bw()

theme_light()

theme_dark()

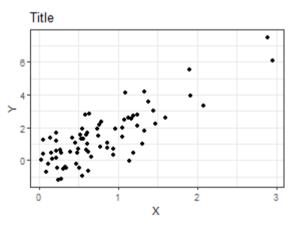
theme_minimal()

. . .



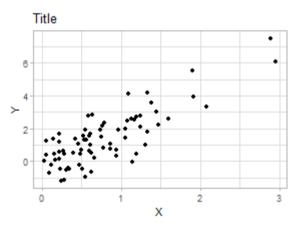
?theme

Themes are used to modify non-data components of plots, such as titles, labels, ticks, fonts for text elements, background etc. ggplot2 provides out-of-the-box themes for quick and easy adjustments of the overall look:



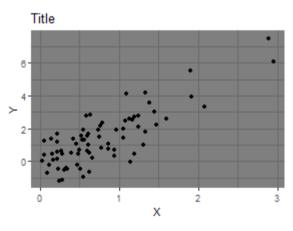
?theme

Themes are used to modify non-data components of plots, such as titles, labels, ticks, fonts for text elements, background etc. ggplot2 provides out-of-the-box themes for quick and easy adjustments of the overall look:



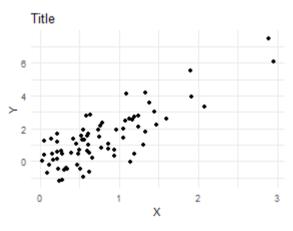
?theme

Themes are used to modify non-data components of plots, such as titles, labels, ticks, fonts for text elements, background etc. ggplot2 provides out-of-the-box themes for quick and easy adjustments of the overall look:



?theme

Themes are used to modify non-data components of plots, such as titles, labels, ticks, fonts for text elements, background etc. ggplot2 provides out-of-the-box themes for quick and easy adjustments of the overall look:



?theme

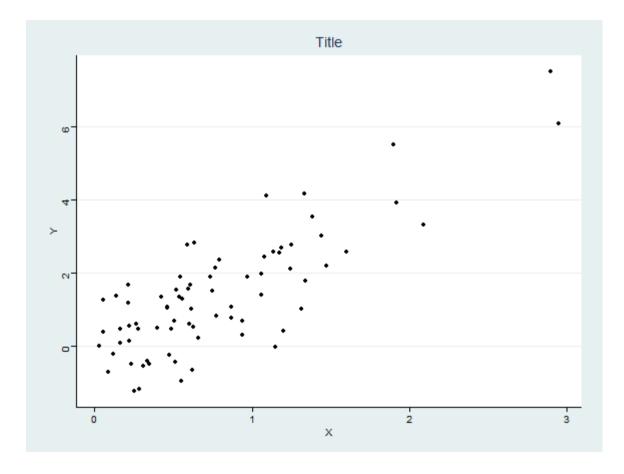
Themes are used to modify non-data components of plots, such as titles, labels, ticks, fonts for text elements, background etc. ggplot2 provides out-of-the-box themes for quick and easy adjustments of the overall look:

theme_gray() theme_bw() theme_light() theme_dark() theme_minimal() ...

... and more!

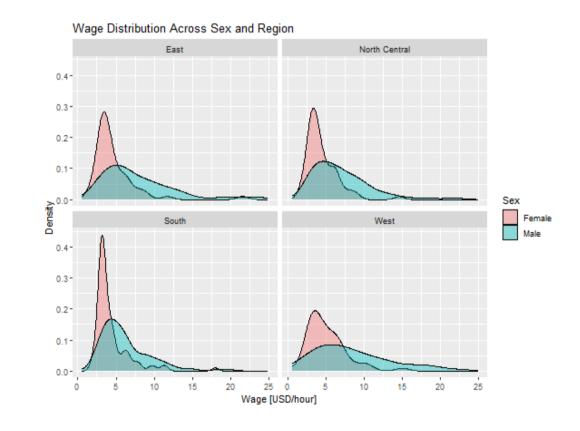
More Themes

The ggthemes package offers more out-of-the-box themes. Have a look at the Stata theme, e.g.:

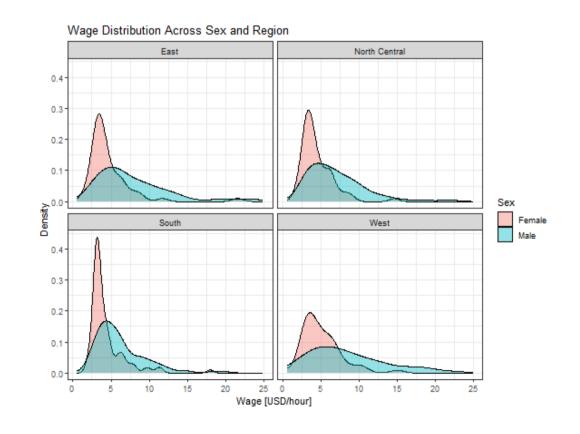




- The appearance of presentations and seminar papers greatly improves by using a consistent design throughout your work
- When outputting several plots, managing the design of each plot individually becomes very cumbersome
- Instead, you may want to think about writing your own custom theme function that is applied to each plot
- Changing your theme function then automatically changes the appearance of all your plots



- A good starting point for a custom theme is ggplot's theme_classic(): A black-and-white theme with minimal styling
- Take some time to think about how the components of your plot should look like
 - Font family and size
 - Legend position and orientation
 - Background and border colors
 - o ...
- Play around with the options until you decide on your design
- Run ?ggplot2::theme for a full list of theme arguments



Plot

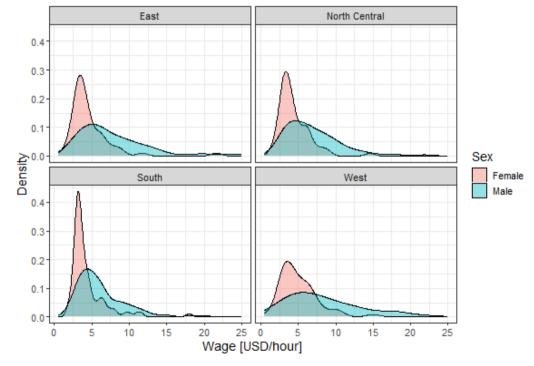
Code

```
theme_custom <- function() {
  theme_bw()
}
grouped_density_plot +
  theme_custom()</pre>
```

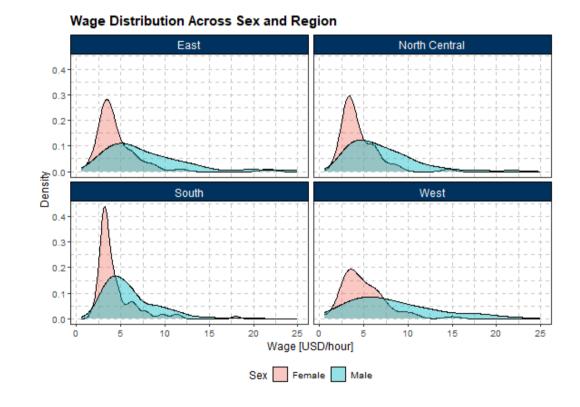
Plot Code

- Once you decided on what your design should look like, start implementing the options in your function
- The theme arguments of theme_classic() can be overwritten using %+replace%
- We may want to consider making the following changes to our design:
 - Change the default text font to "Helvetica"
 - Increase the base font size to 14.
 - Left-align the title, set its font to bold and its size to 150% of the base font size
 - Increase the top and bottom margin around the title

Wage Distribution Across Sex and Region



- Adjust the grid: Darker and thicker dashed lines to increase the grid's visibility
- Change the legend position and orientation
- Set the facet background to dark blue, its text color to white and increase the padding around the text

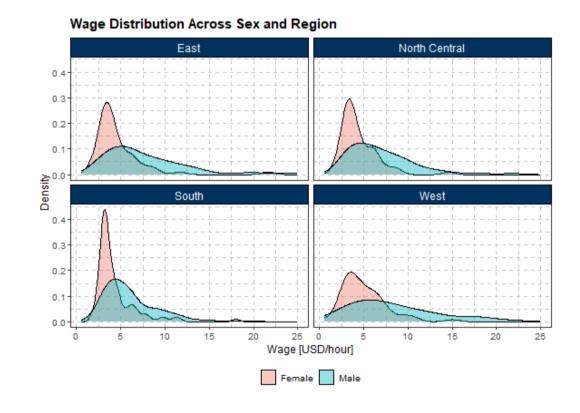


Plot Code

```
theme_custom <- function(base_family = "Helvetica",</pre>
                         base_size = 12,
                         padding = base_size*.75,
                         title size = base size*1.25) {
  theme_bw() %+replace%
    theme(
      # Add the code below to theme() in the theme_custom() function
      panel.grid = element_line(color = "#C8C8C8",
                                size = .5,
                                linetype = "dashed"),
      legend.position = "bottom",
      legend.direction = "horizontal",
      strip.text = element_text(color = "#ffffff",
                                margin = margin(t = 0.5*padding, b = 0.5*padding)),
      strip.background = element_rect(fill = "#003560")
```

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- Including control flows into your custom theme function is useful for easily toggling on or off specific components of your theme
- Perhaps you want to have a short-hand argument for removing the legend title
- In the theme shown on the right side, you can set show_legend_title to FALSE to remove the legend title



```
theme_custom <- function(base_family = "Helvetica",</pre>
                          base_size = 12,
                          padding = base_size*.75,
                          title_size = base_size*1.25,
                          show_legend_title = TRUE) {
  out <- theme_bw() %+replace%</pre>
    theme() # Include the previous theme arguments here
  if(!show_legend_title) {
    out <- out %+replace%</pre>
      theme(
        legend.title = element_blank()
 out
```

Exporting Plots



ggsave()

```
?ggplot2::ggsave
```

- By default, ggsave() saves your last plot to the path given by filename and path
 - Specify which plot in the environment to save by setting plot accordingly
 - Instead of giving both, filename and path, the directory and file name can be specified via filename alone
- The device to use, can be set implicitly by filename or explicitly by setting device to e.g. "png" or "jpeg".
- Set units to "px" and change width and height according to the desired resolution

References

Wickham, H. (2010). "A Layered Grammar of Graphics". In: *Journal of Computational and Graphical Statistics* 19.1, pp. 3-28. DOI: 10.1198/jcgs.2009.07098. eprint: https://doi.org/10.1198/jcgs.2009.07098. URL: https://doi.org/10.1198/jcgs.2009.07098.

Wickham, H. (2016). *ggplot2: Elegant Graphics for Data Analysis*. Springer-Verlag New York. ISBN: 978-3-319-24277-4. URL: https://ggplot2.tidyverse.org.

Wickham, H. and G. Grolemund (2016). *R for data science. import, tidy, transform, visualize, and model data.* O'Reilly. URL: https://r4ds.had.co.nz/.

Wilkinson, L. (2005). *The Grammar of Graphics*. 2nd ed. Springer New York, NY. ISBN: 978-0-387-28695-2. DOI: https://doi.org/10.1007/0-387-28695-0.

