Communicating with the Tidyverse with ISO data

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In this RMarkdown, we will explore: 1. The relationship between weekly working hours and monetary compensation in European countries, according to the International Labour Organization (ILO).

2. Communicating with the tidyverse. They say that a picture is worth a thousand words. Successfully promoting your data analysis is not only a matter of accurate and effective graphics, but also of aesthetics and uniqueness.

More specifically, it will show how to leverage the power of ggplot2 themes for producing publication-quality graphics that stick out from the mass of boilerplate plots out there. And, how to tweak and get the most out of ggplot2 in order to produce unconventional plots that draw attention on social media.

In the end, combining this knowledge will produce a slick and custom-styled report with RMarkdown and CSS – all of that within the powerful tidyverse.

```
if(!require(tidyverse)) install.packages("tidyverse", repos = "http://cran.us.r-project.org")
## Loading required package: tidyverse
## -- Attaching packages ------ tidyverse 1.3.0 --
## v ggplot2 3.2.1
                    v purrr
                              0.3.3
## v tibble 2.1.3
                     v dplyr
                              0.8.4
                    v stringr 1.4.0
           1.0.0
## v tidyr
## v readr
           1.3.1
                    v forcats 0.4.0
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
if(!require(forcats)) install.packages("forcats", repos = "http://cran.us.r-project.org")
if(!require(ggplot2)) install.packages("ggplot2", repos = "http://cran.us.r-project.org")
if(!require(dplyr)) install.packages("dplyr")
library(dplyr)
library(forcats)
library(ggplot2)
library(tidyverse)
load(url("http://s3.amazonaws.com/assets.datacamp.com/production/course 5807/datasets/ilo data.RData"))
# Count the resulting rows and examine
ilo_data %>% count()
```

A tibble: 1 x 1

```
##
         n
##
     <int>
## 1
       380
ilo_data
## # A tibble: 380 x 4
##
      country
                      year hourly_compensation working_hours
##
      <fct>
                      <fct>
                                           <dbl>
## 1 Finland
                      1980
                                            8.61
                                                           35.6
## 2 France
                      1980
                                            8.90
                                                           35.4
## 3 Italy
                      1980
                                            8.09
                                                           35.7
                                                           30.4
## 4 Norway
                      1980
                                           11.8
## 5 Spain
                      1980
                                            5.86
                                                           36.8
## 6 Sweden
                      1980
                                           12.4
                                                           29.2
## 7 Switzerland
                      1980
                                           11.0
                                                           34.7
                                           7.52
## 8 United Kingdom 1980
                                                           34.1
## 9 Finland
                      1981
                                            8.33
                                                           35.7
                                            7.98
## 10 France
                      1981
                                                           34.8
## # ... with 370 more rows
The loaded data contains 380 rows.
Can visualizations reveal a correlation between the relationship between weekly working hours and monetary
compensation in European countries within the dataset?
# Turn year and country into a factor
ilo_data_corrected <- ilo_data %>% mutate(year = as.factor(as.numeric(year)),country = as.factor(countr
ilo_data_corrected # See the results
## # A tibble: 380 x 4
##
      country
                      year hourly_compensation working_hours
      <fct>
                                           <dbl>
##
                      <fct>
## 1 Finland
                                            8.61
                                                          35.6
                      1
## 2 France
                                            8.90
                      1
                                                           35.4
## 3 Italy
                                            8.09
                                                           35.7
                      1
## 4 Norway
                      1
                                           11.8
                                                           30.4
## 5 Spain
                                            5.86
                                                           36.8
                      1
## 6 Sweden
                      1
                                           12.4
                                                           29.2
## 7 Switzerland
                                           11.0
                                                           34.7
## 8 United Kingdom 1
                                            7.52
                                                           34.1
## 9 Finland
                      2
                                            8.33
                                                           35.7
## 10 France
                      2
                                            7.98
                                                           34.8
## # ... with 370 more rows
#Filter for European countries, that also have sufficient data:
european_countries <- c("Finland", "France", "Italy", "Norway", "Spain", "Sweden", "Switzerland", "United King
european_countries # Examine the European countries vector
  [1] "Finland"
                          "France"
                                            "Italy"
                                                              "Norway"
## [5] "Spain"
                          "Sweden"
                                            "Switzerland"
                                                              "United Kingdom"
```

"Ireland"

"Germany"

[9] "Belgium"

[13] "Netherlands"

[17] "Czech Rep."

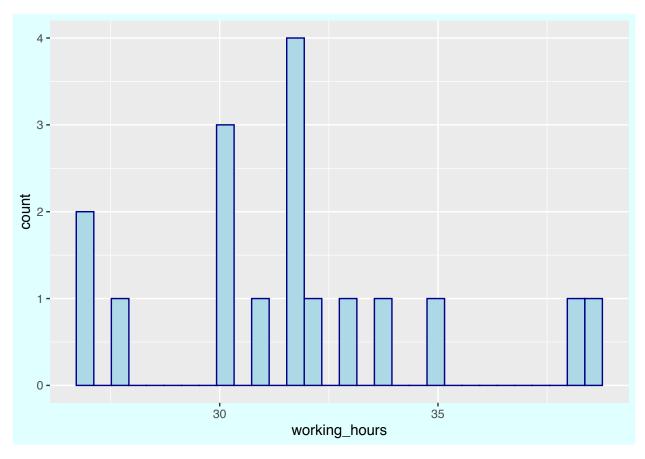
"Luxembourg"

"Hungary"

"Portugal"

"Austria"

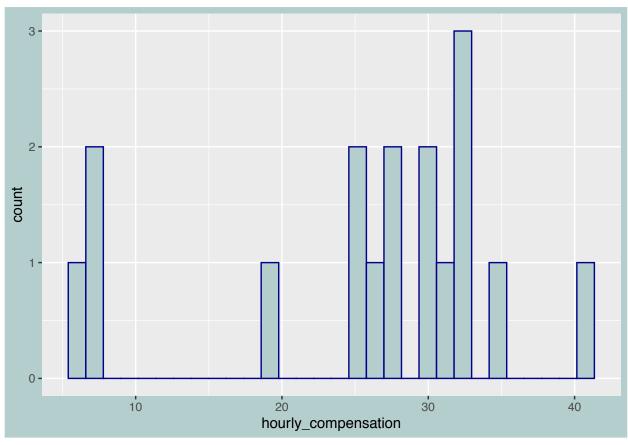
```
# Only retain European countries
ilo_data <- ilo_data %>% filter(country %in% european_countries)
str(ilo_data) # Examine the structure of ilo_data
## Classes 'tbl df', 'tbl' and 'data.frame':
                                                380 obs. of 4 variables:
## $ country
                         : Factor w/ 30 levels "Australia", "Austria", ...: 8 9 15 22 25 27 28 29 8 9 ...
## $ year
                         : Factor w/ 27 levels "1980", "1981", ...: 1 1 1 1 1 1 1 2 2 ....
## $ hourly_compensation: num 8.61 8.9 8.09 11.8 5.86 ...
## $ working_hours
                        : num 35.6 35.4 35.7 30.4 36.8 ...
# Group and summarize the data
ilo_data %>%
  group_by(year) %>%
  summarize(mean_hourly_compensation = mean(hourly_compensation),
            mean_working_hours = mean(working_hours))
## # A tibble: 27 x 3
##
     year mean_hourly_compensation mean_working_hours
##
      <fct>
                               <dbl>
## 1 1980
                                9.27
                                                   34.0
                                                   33.6
## 2 1981
                                8.69
## 3 1982
                                8.36
                                                   33.5
## 4 1983
                                7.81
                                                   33.9
                                                   33.7
## 5 1984
                                7.54
## 6 1985
                                7.79
                                                   33.7
                                                   34.0
## 7 1986
                                9.70
## 8 1987
                                                   33.6
                               12.1
## 9 1988
                               13.2
                                                   33.7
## 10 1989
                                                   33.5
                               13.1
## # ... with 17 more rows
Let's take a look at what the distribution for working hours looks like:
plot_data <- ilo_data %>%
filter(year == 2006)
ggplot(plot_data) + geom_histogram(
aes(x = working_hours), fill="lightblue", color="darkblue") +
   plot.background = element_rect(fill = "lightcyan1"))
```



Let's take a look at what the distribution for hourly_compensation looks like:

```
plot_data <- ilo_data %>%
filter(year == 2006)
ggplot(plot_data) + geom_histogram(
aes(x = hourly_compensation), fill="lightcyan3", color="darkblue") +
    theme(
    plot.background = element_rect(fill = "lightcyan3"))
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

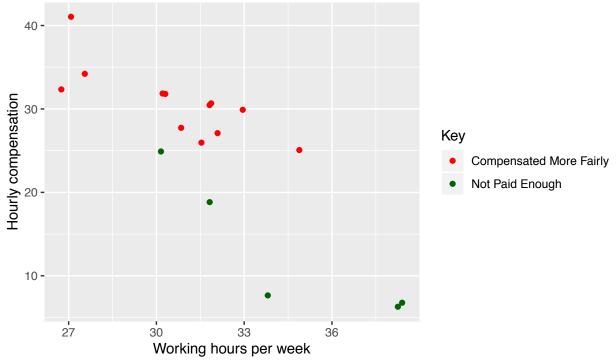


```
# Filter for 2006
plot_data <- ilo_data %>% filter(year %in% 2006)

# Add labels and save the scatter plot into a variable: ilo_plot
ilo_plot <- ggplot(plot_data) +
    geom_point(aes(x = working_hours, y = hourly_compensation, color = ifelse(hourly_compensation < 25, ":
    labs(
        x = "Working hours per week",
        y = "Hourly compensation",
        title = "The more people work, the less compensation they seem to receive",
        subtitle = "Working hours and hourly compensation in European countries, 2006",
        caption = "Data source: ILO, 2017"
    ) +
        scale_color_manual(name="Key", values = c("red", "darkgreen"))

ilo_plot</pre>
```

The more people work, the less compensation they seem to receive Working hours and hourly compensation in European countries, 2006

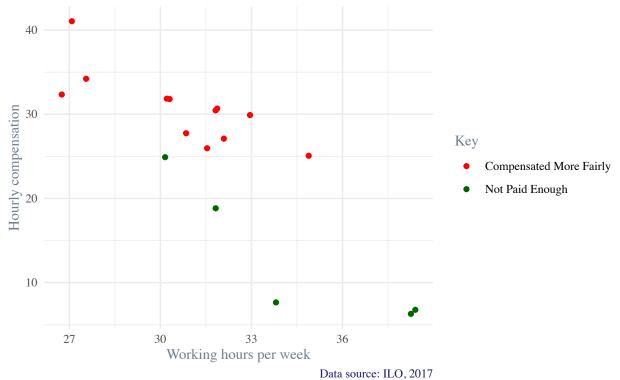


Data source: ILO, 2017

```
ilo_plot <- ilo_plot + theme_minimal() +
    # Customize the "minimal" theme with another custom "theme" call
theme(
    text = element_text(family = "serif"),
    title = element_text(color = "slategrey"),
    plot.caption = element_text(color = "midnightblue"),
    plot.subtitle = element_text(size = 12)
        )

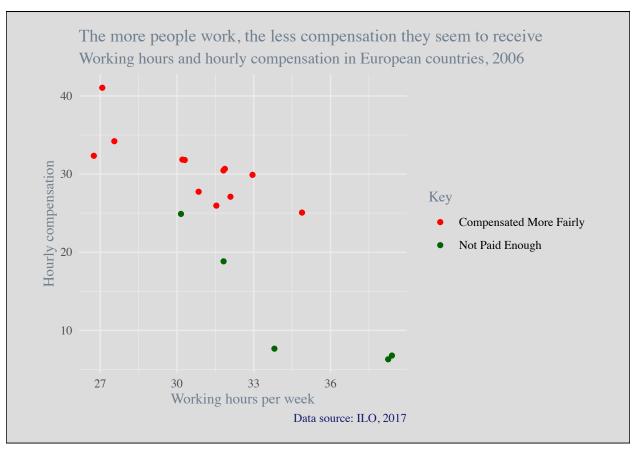
ilo_plot # Render the plot object</pre>
```

The more people work, the less compensation they seem to receive Working hours and hourly compensation in European countries, 2006



"theme" calls can be stacked upon each other, so this is already the third call of "theme".

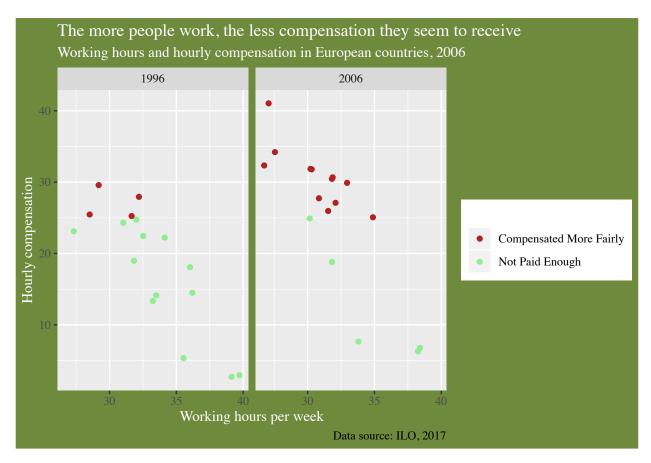
```
ilo_plot +
  theme(
    plot.background = element_rect(fill = "gainsboro"),
    plot.margin = unit(c(5, 10, 5, 10), units = "mm"))
```



```
# Filter ilo_data to retain the years 1996 and 2006
ilo_data <- ilo_data %>% filter(year=="1996" | year=="2006")
```

The data is now filtered so it only contains the years 1996 and 2006 – a good time range for comparison.

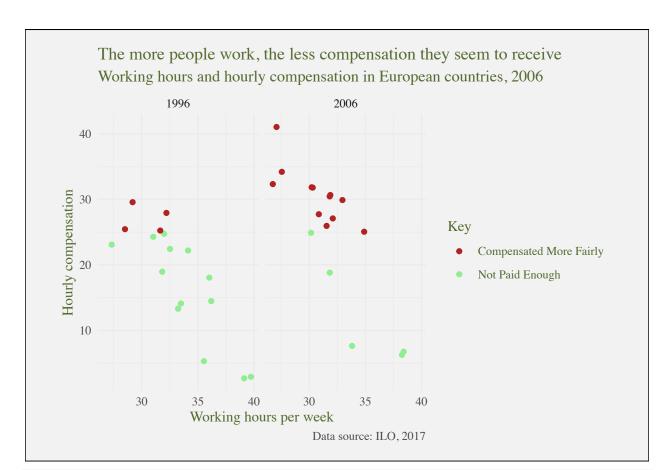
```
# Again, save the plot object into a variable so you can save typing later on
ilo_plot <- ggplot(ilo_data, aes(x = working_hours, y = hourly_compensation)) +</pre>
  geom_point(aes(color = ifelse(hourly_compensation < 25, "Not Paid Enough", "Compensated More Fairly")</pre>
   labs(
   x = "Working hours per week",
   y = "Hourly compensation",
   title = "The more people work, the less compensation they seem to receive",
   subtitle = "Working hours and hourly compensation in European countries, 2006",
    caption = "Data source: ILO, 2017"
  ) + theme(
        text = element_text(family = "serif"),
   title = element_text(color = "white"),
   plot.caption = element_text(color = "black"), plot.background = element_rect(fill = "darkolivegreen")
            )+
  # Add facets here
facet_grid(facets = . ~ year) +
  scale_color_manual(name="Key", values = c("firebrick","lightgreen"))
ilo_plot
```



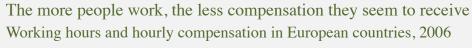
Another possible design, using a theme function:

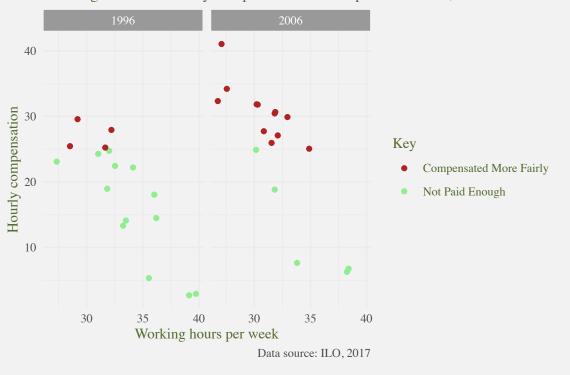
```
# Define theme function below
theme_ilo <- function() {
    theme_minimal() +
    theme(
    text = element_text(family = "serif", color = "darkolivegreen"),
    plot.subtitle = element_text(size = 12),
    plot.caption = element_text(color = "gray30"),
    plot.background = element_rect(fill = "gray95"),
    plot.margin = unit(c(5, 10, 5, 10), units = "mm")
)
}

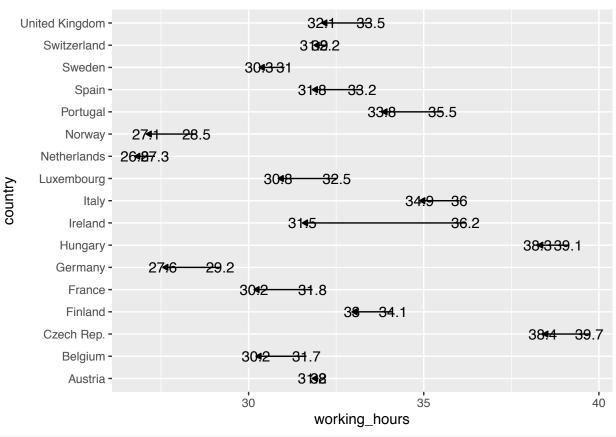
# Apply the theme function
ilo_plot <- ilo_plot + theme_ilo()
ilo_plot # Examine ilo_plot</pre>
```



```
ilo_plot +
    # Add another theme call
theme(
    # Change the background fill and color
    strip.background = element_rect(fill= "gray60", color="gray95"),
    # Change the color of the text
    strip.text = element_text(color="white")
)
```

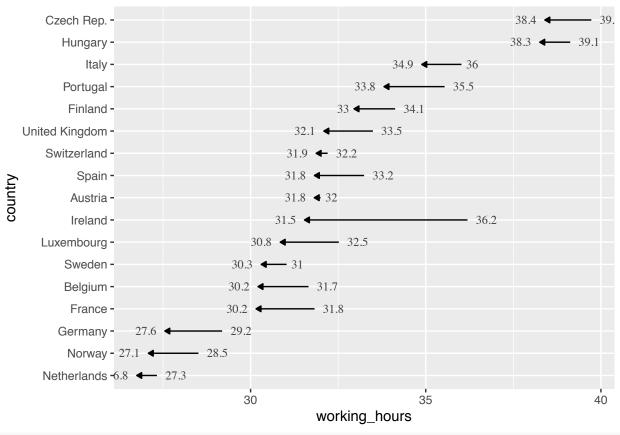




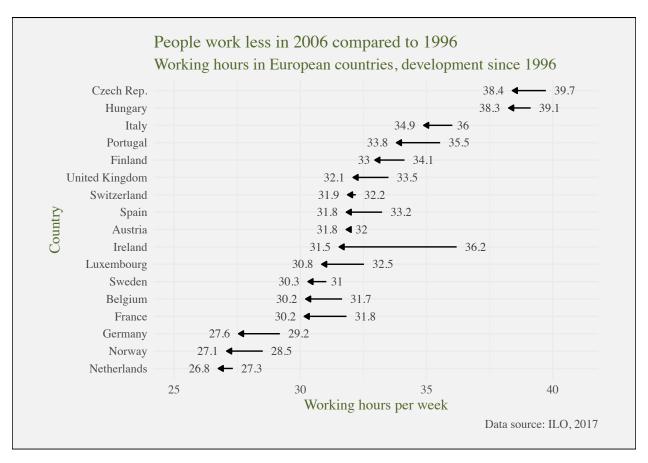


```
# Reorder country factor levels
ilo_data <- ilo_data %>%
  # Arrange data frame
arrange(year) %>%
  # Reorder countries by working hours in 2006
mutate(country = fct_reorder(country, working_hours, last))
```

```
# plot again, and save plot into an object for reuse
ilo_dot_plot <- ggplot(ilo_data) +</pre>
  geom_path(aes(x = working_hours, y = country),
            arrow = arrow(length = unit(1.5, "mm"), type = "closed")) +
    # Specify the hjust aesthetic with a conditional value
    geom_text(
          aes(x = working_hours,
              y = country,
              label = round(working_hours, 1),
              hjust = ifelse(year == "2006", 1.4, -0.4)
            ),
          # Change the appearance of the text
          size = 3,
          family = "serif",
          color = "gray25"
ilo_dot_plot
```



```
# Reuse ilo_dot_plot
ilo_dot_plot <- ilo_dot_plot +</pre>
  # Add labels to the plot
 labs(
   x = "Working hours per week",
   y = "Country",
   title = "People work less in 2006 compared to 1996",
   subtitle = "Working hours in European countries, development since 1996",
   caption = "Data source: ILO, 2017"
 ) +
  # Apply your theme
 theme_ilo() +
  # Change the viewport
  coord_cartesian(xlim = c(25, 41))
# View the plot
ilo_dot_plot
```



The Reduction in Weekly Working Hours in Europe.

Results... In the following, a plot that shows the reduction of weekly working hours from 1996 to 2006 in each country is produced.

```
# Compute temporary data set for optimal label placement
median_working_hours <- ilo_data %>%
  group by(country) %>%
  summarize(median_working_hours_per_country = median(working_hours)) %>%
  ungroup()
# Have a look at the structure of this data set
str(median_working_hours)
## Classes 'tbl_df', 'tbl' and 'data.frame':
                                                17 obs. of 2 variables:
   $ country
                                      : Factor w/ 30 levels "Netherlands",..: 1 2 3 4 5 6 7 8 9 10 ...
## $ median_working_hours_per_country: num 27 27.8 28.4 31 30.9 ...
ilo dot plot +
  # Add label for country
  geom_text(data = median_working_hours,
            aes(y = country, x = median_working_hours_per_country, label =country),
            vjust = 2,
            family = "Times",
            color = "gray25") +
  # Remove axes and grids
  theme(
    axis.ticks = element_blank(),
```

```
axis.title = element_blank(),
axis.text = element_blank(),
panel.grid = element_blank(),
# Also, let's reduce the font size of the subtitle
plot.subtitle = element_text(size = 7)
)
```

People work less in 2006 compared to 1996 Working hours in European countries, development since 1996 38.4 ← 39.7 38.3Czech Rep. Hungary 34.9 36 33.8 **■** Italy_{35.5} 33 Postugal 32.1 ← Finland₅ 31United3Kingdom Switzerland 33.2 31.8 **Spain** 31.5Austria **-** 36.2 30.8 ← 32.Igeland 30.3 Luxembourg 30.2Sweden 31.7 30.2 Belgium 31.8 27.6 ← 29.2 France 27.1 Germany 5 26.8 Norway Data source: ILO, 2017