

# Communicating Economics:

## Data Visualisation in R

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Zaen de Souza

Azim Premji University

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## Overview

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## Designing a plot

Things to keep in mind while making a plot

## Scatterplots

Understanding how ggplot works using a simple scatterplot

## Themes

An example of a histogram, using a new theme

## Transitions & flows:

An example of transition matrices using an alluvial plot

## Gaps & Differences:

An example of a dumbbell plot!

There are two really nice animations on how the brain perceives shapes, colours, lines and so on. You can view them [here](#) and [here](#).

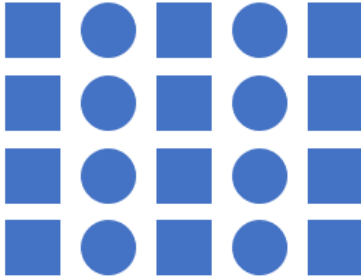
Here are 4 ways in which we organise/perceive information<sup>1</sup>

1. **Similarity**
2. **Proximity**
3. **Continuity**
4. **Enclosure**

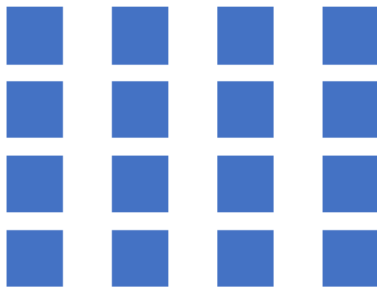
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<sup>1</sup>Taken from [here](#)

Items that are similar to each other tend to be seen as a group, rather than single objects. The squares and circles in the graphic below are not seen as single objects but are rather perceived as grouped columns.



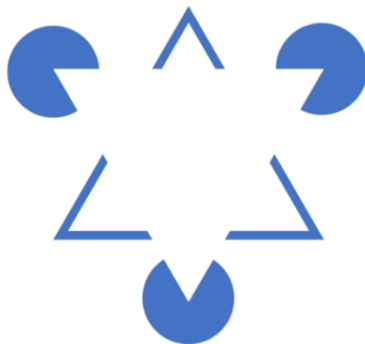
Objects that are closer together are seen as a group and not individually. The squares below are perceived as 4 columns.



The eye follows the smoothest path it can find. Did you notice how the eye follows the lighter points better?



Objects grouped together are seen as a whole. We ignore lines and fill in missing parts. Do you see the white triangle?





Load all the datasets from the google drive [folder](#).

```
ex_1 <- read.csv("ex_1.csv")  
ex_2 <- read.csv("ex_2.csv")  
ex_3 <- read.csv("ex_3.csv")  
map_data <- read.csv("map_data.csv")
```

```
library(pacman)
p_load(ggplot2, dplyr,
       ggforce, ggthemes,
       ggalluvial, ggtext)
```

## GGplot Review

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All the data visualisation that we will do here uses either the standard `ggplot2` package, or its extensions, such as `ggmaps`, `ggalluvial`, and so on. You can read more about the `ggplot2` package [here](#).

`ggplot` relies on **three** core arguments:

1. `data`

This is where your numbers (variables) come from.

2. `aes()`

This maps variables to different aesthetics—colours, shapes etc.

3. `geom_type()`

This takes the other arguments and builds up your graphic using geometries

**Note 1:** `geom_type()` is a placeholder; **type** will be replaced by the geometry you choose to use; for e.g: `geom_point()` .

**Note 2:** `aes()` can also contain functions. For instance:

```
aes(x = age_months/12, y = log(wages))
```

We need to map the following:

`data` → `aes()` → `geom_type()`

New layers/blocks are added ontop of each other, using the `+` sign

## Visualisation 1

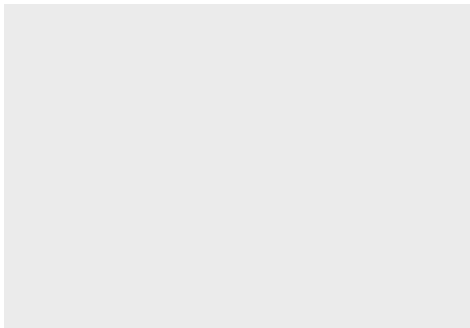
---

Let's use the dataset called `ex_1`, and use mapping the `aes()` using `x=`, `y=`, to three variables from our dataset.

1. `incgr` : Change in Income (%)
2. `mobility` : Change in Mobility (%)
3. `state_month` : State ID + Month



```
plot_1 <- ggplot()  
plot_1
```



**Figure 1:** An Empty Canvas

## Scatterplots Step 2: Mapping Variables

```
plot_1 <- ggplot(ex_1,  
# we mapping the x, and y to two variables from the data  
  aes(x = mobility,  
      y = incgr))  
plot_1
```

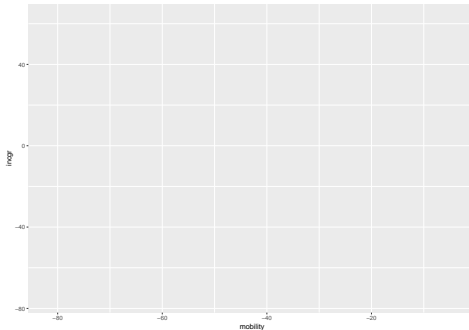


Figure 2: Canvas + Mapped Variables

## Scatterplots Step 3: Adding A Geom

```
plot_1 <- plot_1 + geom_point()  
plot_1
```

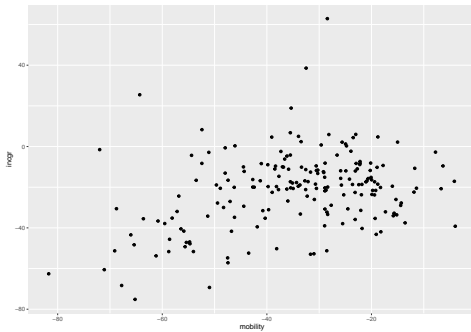


Figure 3: A Scatter Plot

## Scatterplots Step 4: Editing the aes()

```
plot_1 <- ggplot(ex_1,  
  aes(x = mobility,  
      y = incgr,  
      colour = state)) +  
  geom_point()  
plot_1
```

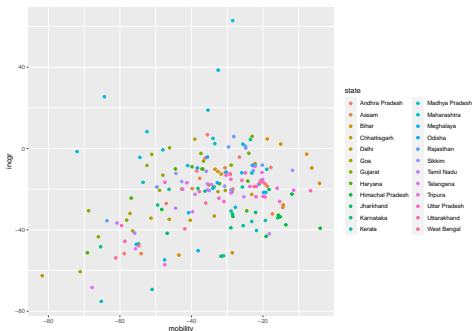


Figure 4: A Scatter Plot with Colours

## Scatterplots Step 5: Regression Line + Labels

```
plot_1 <- ggplot(ex_1,  
  aes(x = mobility,  
      y = incgr)) +  
  geom_text(aes(label = state_month)) +  
  geom_smooth(method = "lm")  
# try without method="lm", i.e geom_smooth()  
plot_1
```

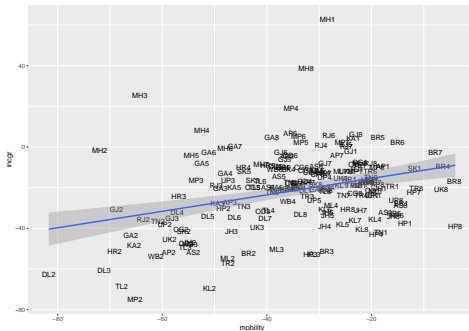


Figure 5: A Scatter Plot with Labels

What we did

1. Add data
2. Map variables
3. Customise the aesthetics
4. Add geoms ('geom\_point', 'geom\_text', 'geom\_smooth')

What you need to do:

1. Add axis labels
2. Add a title, subtitle, caption
3. Add (Make!) a theme
4. Add annotation to the plot—notes, reference lines etc

Hint: Use `annotate()` , `geom_hline()` , `geom_vline()`

1. `geom_histogram()` histogram
2. `geom_bar()` bar plot
3. `geom_col()` bar plot 2
4. `geom_density()` Kernel density plot
5. `geom_point()` Scatterplot
6. `geom_smooth(method=)` regression lines
  - `geom_smooth()`
  - `geom_smooth(method="lm")`

There is a massive list of geoms and as well, ggplot options [here](#). Make sure to check this out.

## Themes

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`ggplot` comes with a range of in-built themes that are ready to be used. These can be accessed using `library(ggthemes)` and then adding a theme object to your `ggplot`.

```
plot_1 +  
  theme_classic()
```

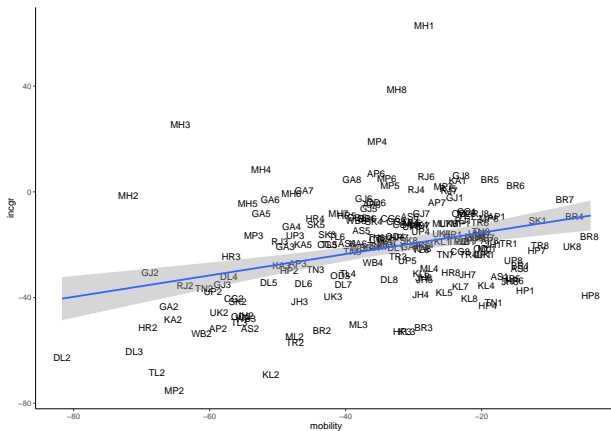


Figure 6: Classic

```
plot_1 +  
  theme_fivethirtyeight()
```

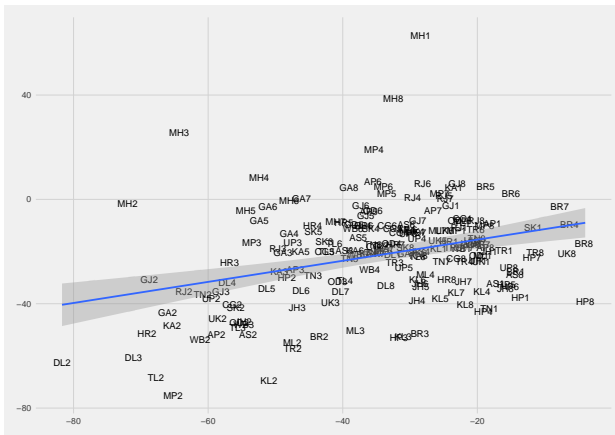
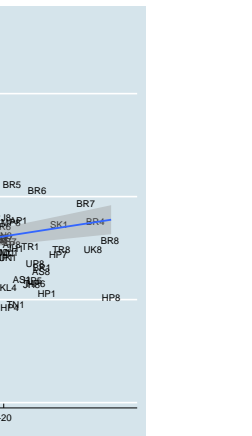


Figure 7: fivethirtyeight



```
plot_1 +  
  theme_stata()
```

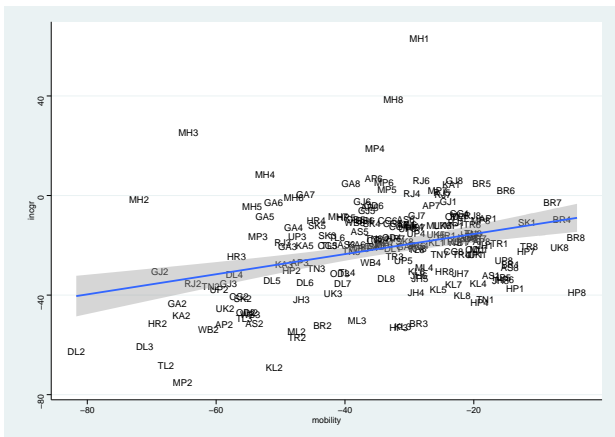


Figure 9: Stata

**ggplot** is very flexible. You can customize every single part of a graph. This is done using the `theme()` function, and selecting different options.

Each plot, consists of a range of different parts—axis lines, gridlines, labels, backgrounds, titles, shapes and so on.

Type `?theme` to see the full list of options. You should see something like this in your `help` tab.

```
theme(  
  line,  
  rect,  
  text,  
  title,  
  aspect.ratio,  
  axis.title,  
  axis.title.x,  
  ....  
)
```

You could choose to edit any of these—or something else. There are 3000+ permutations that you could experiment with!

1. `axis.line =`
2. `axis.ticks =`
3. `axis.ticks.length =`
4. `legend.position =`
5. `panel.background =`
6. `panel.grid.major =`
7. `panel.grid.minor =`

There is a complete, exhaustive list [here](#).



Each graph component, maps to a particular **element type** . The 4 main types are:

1. `element_blank(arguments)`
2. `element_line(arguments)`
3. `element_text(arguments)`
4. `element_rect(arguments)`

1. `theme(axis.title = )` → `element_blank()` → remove axis titles
2. `theme(axis.line = )` → `element_line(colour="black")` → axis line colour
3. `theme(axis.title = )` → `element_text(colour="red")` → axis title colour
4. `theme(plot.background = )` → `element_rect(fill="green")` → background colour

```
my_theme <- function(base_size = 12,  
                      base_family = "") {  
  # we will use theme_minimal as a base  
  theme_minimal(base_size =  
                base_size,  
                base_family =  
  # this tells R that we are replacing some elements in theme_minimal  
                base_family) %+replace%
```

```
# starting the modification using theme(.....)
theme(
  panel.grid.minor = element_blank(),
  panel.grid.major.x = element_blank(),
  panel.grid.major.y = element_line(
# you can add any common colour here - "red", "green" or a hexcode
    colour = "#c9c0b7",
# dotted, dashed, dotdash, longdash, solid, blank
    linetype = "dotted",
# controls line width 0 -> inf
    size = .5),
```

```
axis.ticks = element_line(  
  size = .5,  
  colour = "#c9c0b7"),  
axis.ticks.x = element_blank(),  
axis.line = element_line(  
  size = .5,  
  colour = "#c9c0b7",  
  linetype = "solid"),  
axis.line.y = element_blank(),
```

```
axis.text.y = element_text(  
  colour = "black",  
  # margin(r,l,t,b)  
  margin = margin(r = 2),  
  hjust = 1),  
axis.text.x = element_text(  
  colour = "black"),  
# we are done!  
complete = TRUE)  
# remember to close the curly bracket which we opened!!  
}
```

## Visualisation 2

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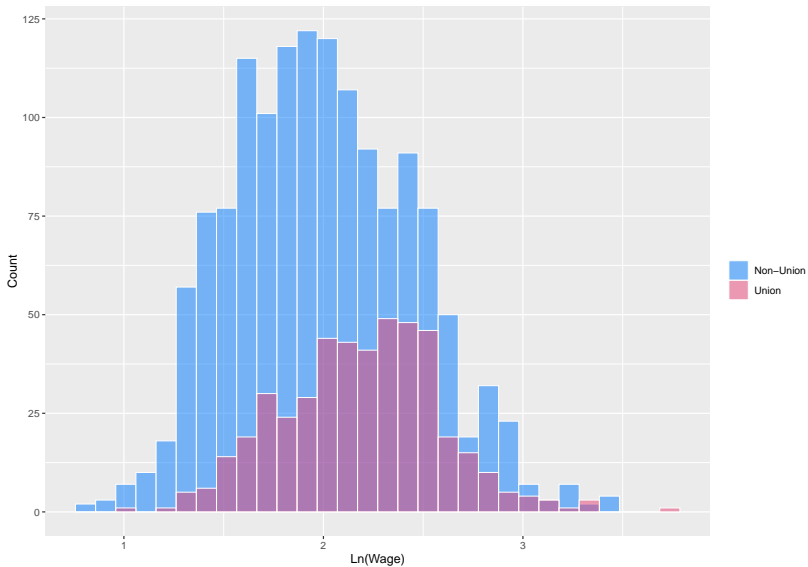
We will use the dataset called `ex_2`, to test our customised theme

1. `ln_wage` :  $\ln(\text{hourly wage})$
2. `union` : Union Status



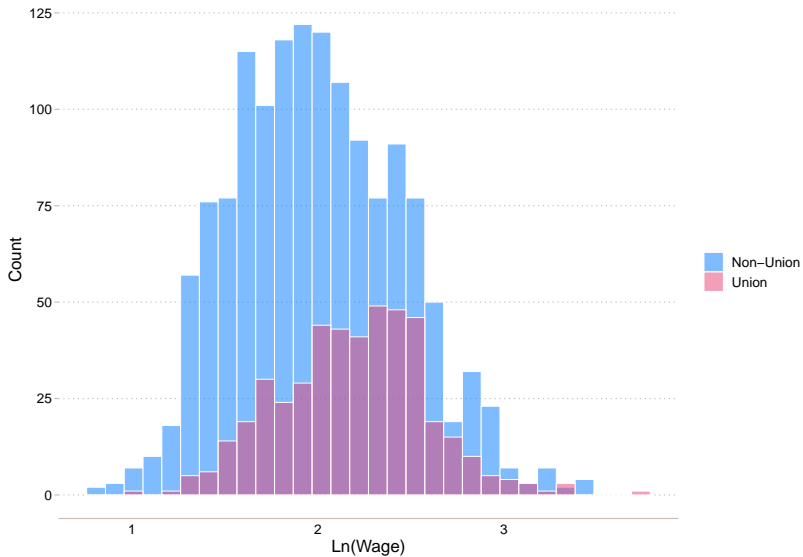
```
plot_2 <- ggplot(ex_2,  
  aes(x = ln_wage,  
      fill = union)) +  
  geom_histogram(position = "identity",  
    bins = 30,  
    size = 0.05,  
    alpha = .5,  
    colour = "white") +  
  scale_fill_manual(values = c("#007bff", "#E64173"),  
    labels = c("Non-Union", "Union")) +  
  ylab("Count") +  
  xlab("Ln(Wage)") +  
  labs(fill = "")  
plot_2
```

## Modifying a theme: ggplot default

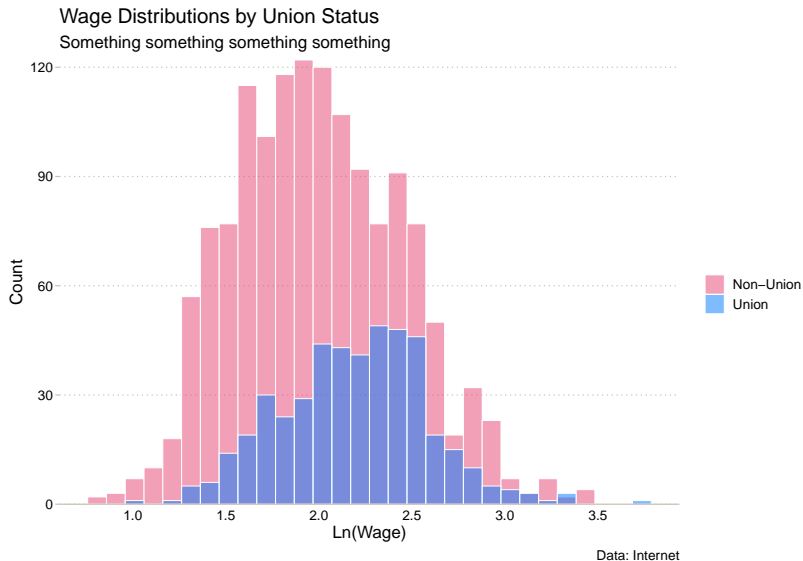


```
plot_2 <- plot_2 + my_theme(base_size = 15)  
plot_2
```

## Modifying a theme: Our theme!



## Modifying out theme a bit more—try to do this yourself!



## Visualisation 3

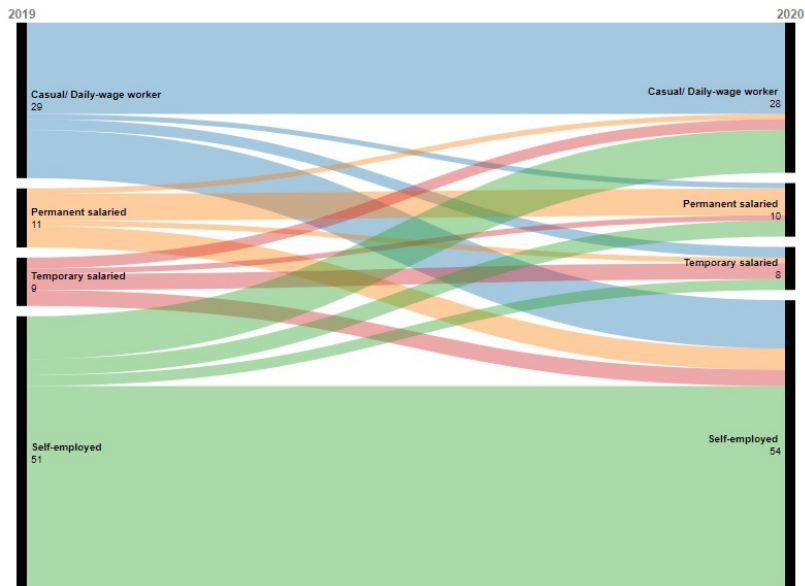
---

We will be using the file called `ex_3`, it is the data that was used to create this matrix:

	Casual/daily wage worker	Self-employed	Temporary salaried	Permanent salaried
Casual/daily wage worker	57.9	33.1	5.6	3.4
Self-employed	15.0	75.5	4.1	5.5
Temporary salaried	22.5	31.3	35.6	10.7
Permanent salaried	9.8	34.1	8.5	47.6

# The graph

And to create this graph:





This chart visualises **transitions, flows, networks** between different **categorical variables**. Each ribbon is **weighted** by some numerical value. When using dis-aggregated data, this could from **frequencies, percentages** etc.

Note that in this version of the data, the percentages in each category of **pre** add up to 100%, so our graph will look slightly different, as compared to the original!

```
library(ggforce)
data_1 <- read.csv("F:/Surbhi (Data Viz)/ex_3.csv")

data_alluvial <- data_1 %>%
  gather_set_data(2:1)

data_alluvial$x <-
  factor(data_alluvial$x,
    levels = c("pre", "post"),
    labels = c(1, 2))
```

## Alluvial Plot Step 1: Data

The setup

```
plot_3 <- data_alluvial %>%  
  ggplot(aes(x, id = id,  
             split = y,  
             value = percentage))  
plot_3
```



Figure 10: Step 1

## Alluvial Plot Step 2: Mapping

```
plot_3 <- plot_3 +  
  geom_parallel_sets(aes(fill = pre,  
                        alpha=0.5),  
                    strength = .4,  
                    show.legend = F)  
plot_3
```

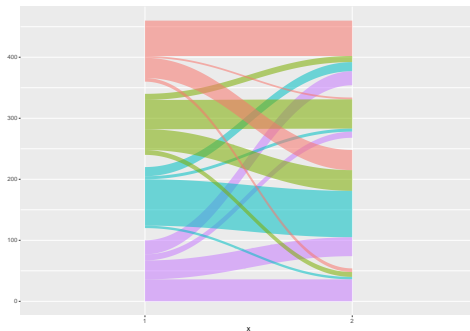


Figure 11: Step 2

## Alluvial Plot Step 3: More mapping

```
plot_3 <- plot_3 +  
  geom_parallel_sets_labels(angle = 0) +  
  geom_parallel_sets_axes(axis.width = 0.01)  
plot_3
```

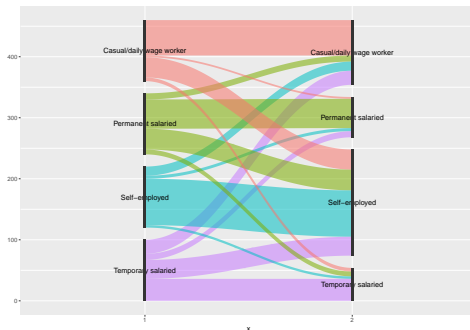


Figure 12: Step 3

## Alluvial Plot Step 4: More mapping

```
plot_3 <- plot_3 +  
  scale_x_discrete(labels = c('2019', '2020'),  
    position = "top") + theme_void()  
plot_3
```

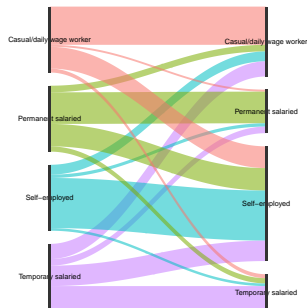


Figure 13: Step 4

```
plot_3 <- plot_3 +  
  theme(axis.text.x =  
    element_text(  
      size = 15,  
      face = "bold"),  
  plot.margin =  
    unit(c(1,1,1,1), "cm"))
```

The following extensions are remaining are:

- Adding **percentage labels**
- **Aligning** the labels

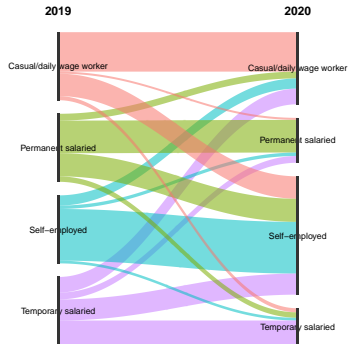


Figure 14: Step 6



## Visualisation 4

---

We will use the dataset called `map_data`, to make a dumbbell plot. This is a subset, we will share the full data, later on, if needed.

1. `recovery_male` : % of male workers who lost, and recovered jobs
2. `noeffect_male` : % of male workers who didn't change job status
1. `recovery_female` : % of female workers who lost, and recovered jobs
3. `noeffect_female` : % of female workers who didn't change job status
4. `recovery_total` : % of workers who lost, and recovered jobs
5. `noeffect_total` : % of workers who didn't change job status
6. `ST_NM` : State name

```
plot_data <- arrange(map_data,  
                      recovery_male) %>%  
  mutate(state.fact = factor(ST_NM,  
                             levels = unique(ST_NM)))
```

You can just type in some colours - 'red', 'seagreen', 'purple', or add a hexcode. Remember, these are going in as strings, so you need to add the quotation marks!

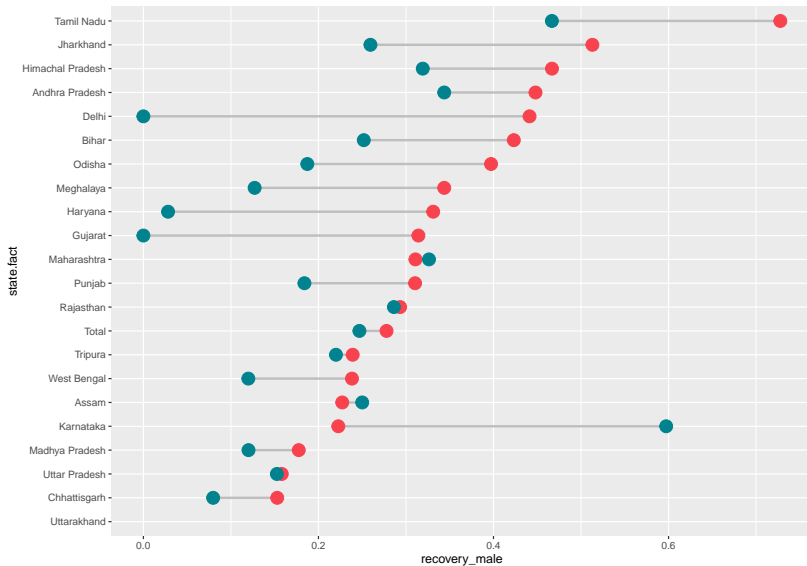
There's a really nice colour tool that you can access [here](#).

```
female <- "#00838f"
```

```
male <- "#f8434f"
```

```
plot_4 <- ggplot() +  
  geom_dumbbell(  
    data = plot_data,  
    aes(y = state.fact,  
        x = recovery_male,  
        xend = recovery_female),  
    size = 1,  
    color = "grey",  
    alpha = 1,  
    size_x = 5,  
    size_xend = 5,  
    colour_x = male,  
    colour_xend = female)
```

## Dumbbell Plot Step 3: Mapping



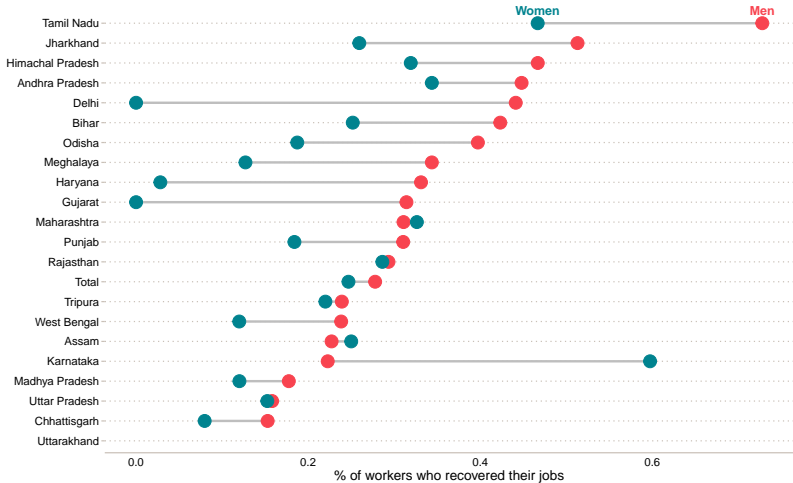
## Dumbbell Plot Step 4: Add some labels

```
plot_4 <- plot_4 + geom_text(  
  # use this to filter the y axis; we are only adding a label on top,  
  # in line with Tamil Nadu - so that it isn't cluttered  
  data = filter(plot_data,  
                 ST_NM == "Tamil Nadu"),  
  aes(x = recovery_male,  
       y = ST_NM,  
       label = "Men"),  
  color = male,  
  vjust = -.9,  
  fontface = "bold") +  
  # Don't forget to use your theme!  
  my_theme()
```

## Try to add labels wherever needed, by yourself

### Gender Differences in Job Loss & Recovery

Fewer women were able to recover their jobs, after losing them



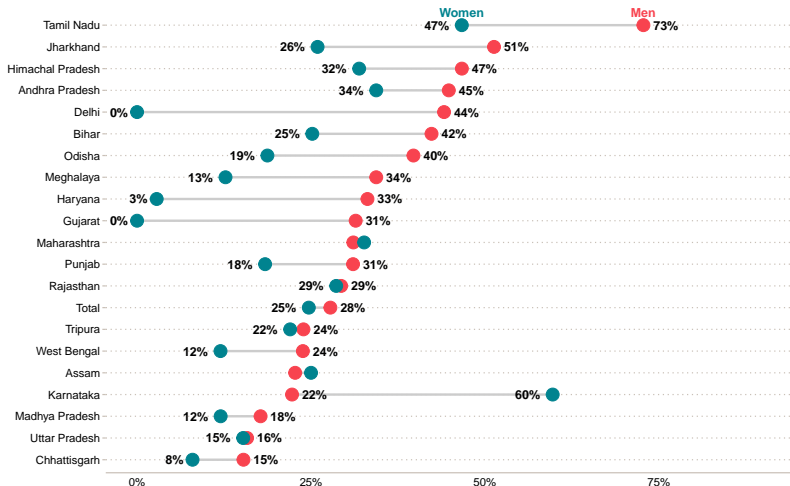
Data: State of Working India, 2021



# This is super cluttered—how do we make it better?!

## Gender Differences in Job Loss & Recovery

Fewer women were able to recover their jobs, after losing them



Data: State of Working India, 2021

The Karnataka point is super high—maybe we can add some explanation there, direct attention to it?

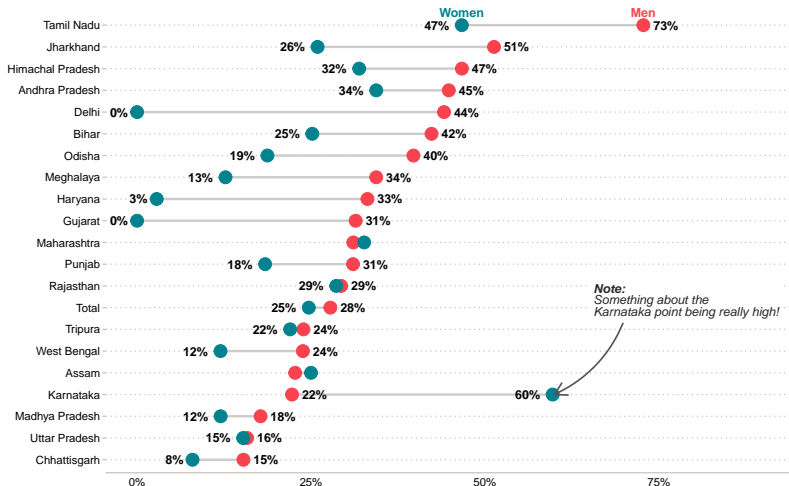
Again, this is done by geoms. For instance, this adds a curved arrow

```
plot_4 <- plot_4 + geom_curve(  
  aes(  
    x = .7,  
    y = 7.4,  
    xend = .6,  
    yend = 4),  
  colour = "#555555",  
  curvature = -0.2,  
  size = .7,  
  # NPC stands for Normalised Parent Coordinates  
  arrow = arrow(length = unit(0.03, "npc")))
```

```
plot_4 <- plot_4 +  
  geom_richtext(  
    aes(x = .65, y = 7.3,  
# this is markdown syntax! ** word** for bold  
# *word* for italic  
# <br> is common to markdown, HTML etc - it adds a linebreak  
label =  
"***Note:***<br>*Something about the <br>  
Karnataka point being really high!*"),  
    lineheight = 0.8,  
    colour = "#2b2b2b",  
    fill = "white",  
    vjust = 0,  
    hjust = 0,  
    label.size = NA,  
    size = 4)
```

## Gender Differences in Job Loss & Recovery

Fewer women were able to recover their jobs, after losing them



Data: State of Working India, 2021

Thank You!

## Data Visualisation Hall of Shame

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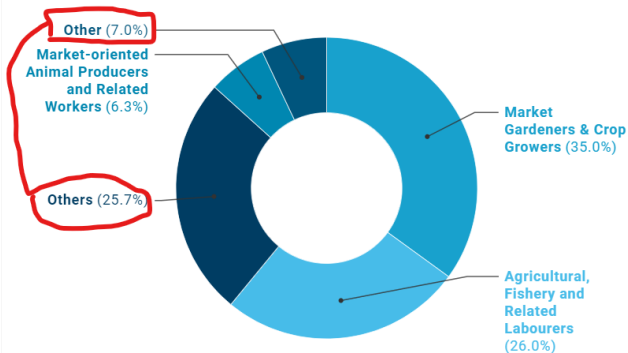


Source: Times now

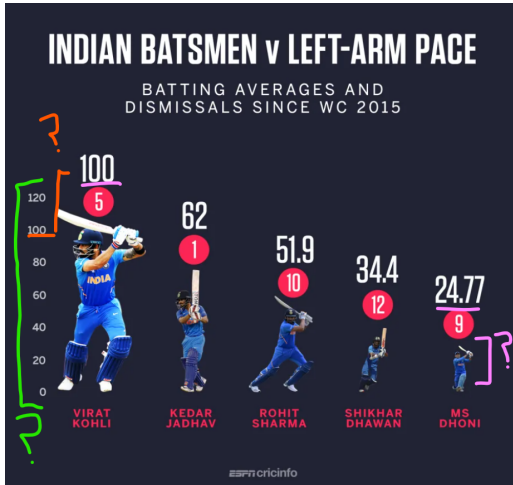


## Most rural women work in agriculture-related jobs

Share of working women in rural India (%)



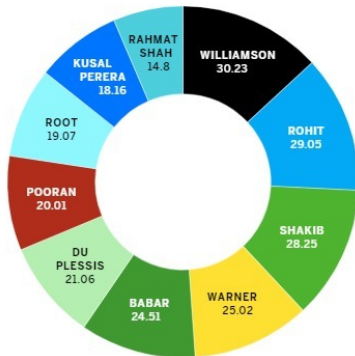
Source: [Rukmini S](#)



Source: [CricInfo](#)

## THE WORLD CUP'S BIG GUNS

% OF TEAM'S RUNS SCORED BY TOP SCORER

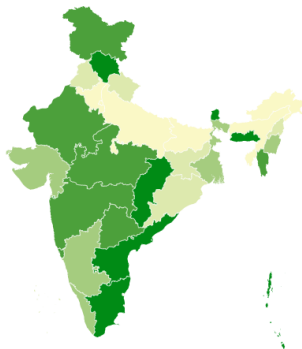
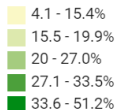


ESPNcricinfo

Source: [CricInfo](#)

## The Hindi belt scores low, while the south does better

Female labour force participation rate (%)



Source: [Rukmini S](#)

## Extra Slides

---

These are specified using `linetype="name"`

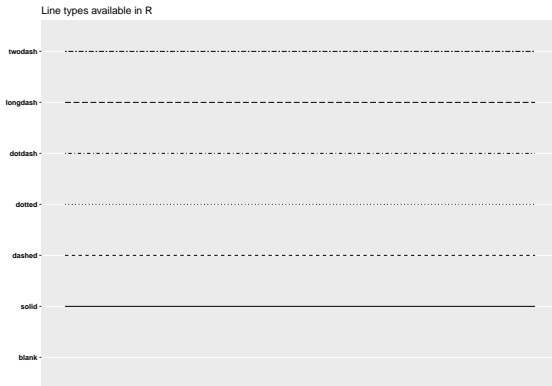


Figure 15: Line Types: Each name is an option added to `linetype=`

These are specified using `shape=number`

Point shapes available in R

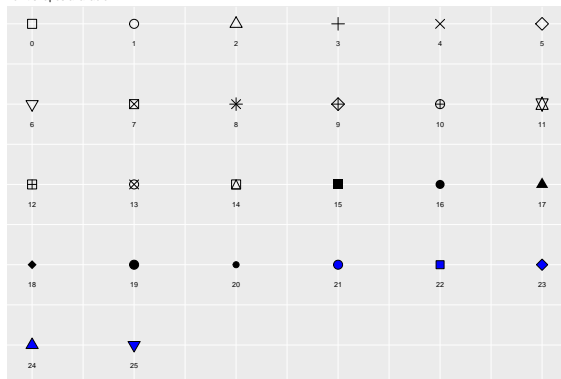
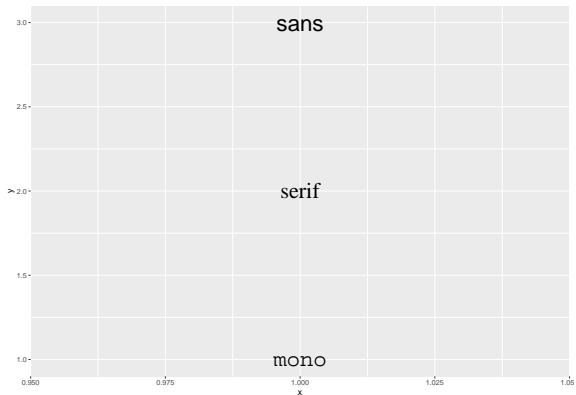


Figure 16: Shape Types: Each number is an option added to `shape=`

These are specified using `family="name"`



**Figure 17:** Font Types: Each nname is an option added to family=



These are specified using `face="name"`

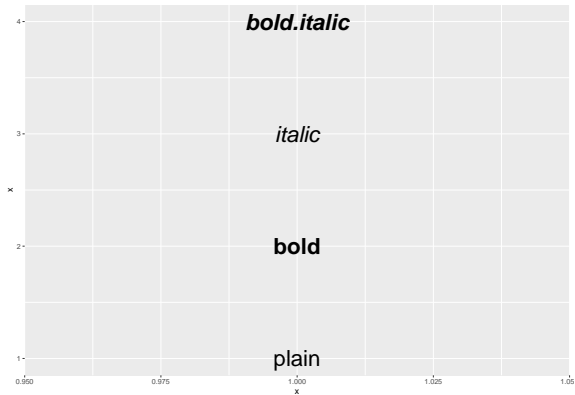


Figure 18: Font Types: Each name is an option added to face=

## Vertical adjustment (vjust) & Horizontal Adjustment (hjust)

These are specified using `vjust=c(x, y)` .  $x, y = [0, 1]$

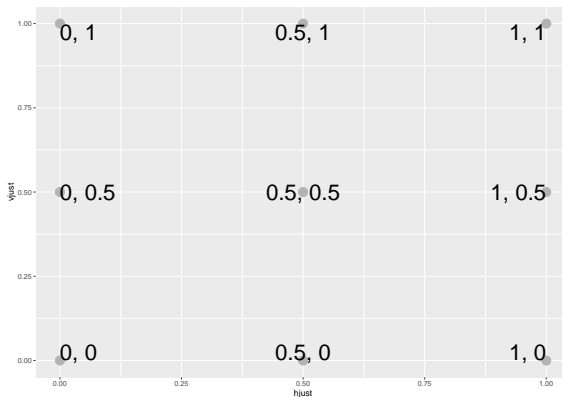


Figure 19: These are the coordinates added to vjust and hjust

Extra Extra Slides (For Nerds)

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## Some great resources to get great at ggplot

1. [Under the hood of ggplot2 graphics in R](#)
2. [Infographics In R](#)
3. [GGplot Cookbook](#)
4. [Beautiful plotting in R: A ggplot2 cheatsheet](#)
5. [Ggplot Gallery](#)
6. [Links to cool stories](#)
7. [Best of #TidyTuesday](#)
8. [xkcd Plots With R](#)

1. [Jasmine Mithani](#)
2. [Mona Chalabi](#)
3. [Allison Horst](#)
4. [Jim Vallandingham](#)
5. [Nadiyah Bremer](#)
6. [BBC's Data Journalism Team](#)
7. [Dear Data](#)