### Happy International Women’s Day!

In the spirit of the day, I wanted to do a post **celebrating the many wonderful women of this world**. In keeping with my political analytics topics, I thought it would be interesting to look at the **proportion of women are in government by country**, and **how that has changed over the past few decades**. Luckily The World bank has such a dataset, measuring the proportion of government seat’s held by women in every country throughout the world. Let’s get to it and visualize this data!

### In total, I will walk you through 5 visualizations:

1. Graph #1 – Female Representation in Government by World Region
2. Graph #2 – Female Representation in Government by Country Income Level
3. Graph #3 – Female Representation in Government from Selected Countries
4. Graph #4 – 20 Countries with the Highest Proportion of Women in Government as of 2020
5. Graph #5 – The 20 Countries with the Largest Increases of Women in Government from 1997 to 2020

# Step 1 – Download & Clean the Data

First, let’s clean the data. To clean the data, there are three steps. First, change the column names so they are easier to read using substring() to cut the year names by a few characters and the make\_clean\_names() function from the janitor package to clean up the other names. Then change all the missing data (“..” in the original dataset) to NAs using the na\_if() function, while converting all the year columns to numeric with sapply(). There you have it, a relatively clean master dataframe!

# Load the necessary packages

# Tidyverse is our rock in data analysis (includes ggplot2) if(!require("tidyverse")) install.packages("tidyverse")

# Janitor cleans data like the best janitor out there! if(!require("janitor")) install.packages("janitor")

# ggsci has my favourite color palettes if(!require("ggsci")) install.packages("ggsci") # ggalt is necessary for the dumbbell chart if(!require("ggalt")) install.packages("ggalt")

# ggtext allows for great text to include for charts if(!require("ggtext")) install.packages("ggtext")

# Load the csv data file

df <- read\_csv("WomenInGovernment\_Data.csv")

# Change the column names to account for each year of the data 1997-2020

names <- colnames(df)[4:27] names <- substring(names, 7,12)

colnames(df)[4:27] <- names

colnames(df) <- janitor::make\_clean\_names(colnames(df))

# Change all the ..s in the data to NAs and then convert the columns to numeric

df[,4:27] <- na\_if(df[,4:27], "..")

df[,4:27] <- sapply(df[,4:27],as.numeric) %>% round(digits = 1)

After this, I want to make it simple for the reader so I **divide the large dataset into 5 separate ones, one for each plot that we are creating**. To do this we use the filter() function, which allows us to cut it whatever way we want. I also use the pivot\_longer() function to recreate the dataframes and make them more *tidy* for our analysis. This is a good practice I almost always neglect to do…

Lastly, for the dumbbell charts we create a **column that calculates the change between 1997 and 2020**. We then rank order the countries by either the proportion of females in government (4th chart) or the change from 1997-2020 (5th chart).

# Create a dataset for each chart, just to make it .

# For this we filter the data for the region, income or country, then make the graph longer using the pivot\_longer() function

# Regional chart dataset df.region <- df %>%

filter(category == "Region" | category == "World") %>% pivot\_longer(cols = starts\_with("yr"), names\_to = "year",

names\_prefix = "yr",

values\_to = "value", values\_drop\_na = TRUE)

# Income chart dataset df.income <- df %>%

filter(category == "Income" | category == "World") %>% pivot\_longer(cols = starts\_with("yr"), names\_to = "year",

names\_prefix = "yr",

values\_to = "value", values\_drop\_na = TRUE)

# Selected Countries chart (Chose countries that are either near and dear to me or where my readers tend to be)

selected\_countries <- c("World", "Canada", "China", "Denmark", "France", "Germany", "India",

"Italy", "Norway", "Sweden", "United Kingdom",

"United States") df.country <- df %>%

filter(country\_name %in% selected\_countries| category == "World") %>% pivot\_longer(cols = starts\_with("yr"), names\_to = "year",

names\_prefix = "yr",

values\_to = "value", values\_drop\_na = TRUE)

# Top Female representation by country dumbbell chart df.top.dumbbell <- df %>%

filter(category == "Country") df.top.dumbbell$change <- df.top.dumbbell$yr2020 -

df.top.dumbbell$yr1997

df.top.dumbbell <- head(arrange(df.top.dumbbell, desc(yr2020)), n = 20)

%>%

droplevels()

df.top.dumbbell$country\_name <- factor(df.top.dumbbell$country\_name, levels = as.character(df.top.dumbbell$country\_name)) # to retain the order in plot.

# Top Female Increase in representation dumbbell chart df.change.dumbbell <- df %>%

filter(category == "Country") df.change.dumbbell$change <- df.change.dumbbell$yr2020 - df.change.dumbbell$yr1997

df.change.dumbbell <- head(arrange(df.change.dumbbell, desc(change)), n

= 20) %>%

droplevels()

df.change.dumbbell$country\_name <- factor(df.change.dumbbell$ country\_name, levels = as.character(df.change.dumbbell$country\_name)) # to retain the order in plot.

# Step 2 - Time to Graph

Okay, now we have created five separate dataframes, each one will tell a different story about how female representation in national government has changed over the past 24 years.

## Graph #1 - Female Representation in Government by World Region

Before starting, we create set the basic theme with theme\_set() and create a folder for all the plots using dir.create(). I explain how I create each plot pretty detailed in the code, so no need to bore you with more text. The first graph concerns the **regions of the world**.

# Let's set the theme and create a folder to store your output theme\_set(theme\_bw())

dir.create("plots")

# First create a plot of all the regional representation of women in government

region.plot <- ggplot(df.region, aes(x=year,y=value,group= country\_name,color=country\_name)) +

geom\_line(size = 1.5) +

# My top color palette scale\_color\_simpsons() +

# Set the x-axis scale scale\_x\_discrete(breaks=c(1997,2002,2007,2012,2017)) +

# Update the labels of the axes labs(x = "Year",

y = "Percentage of Women in Government",

title = "Female Representation in Government by World Region", subtitle = "Unsurprisingly Europe leads the way in female

representation, followed by Latin America \n& the Caribbean and then North America. Since 1997, the world's average has increased \nfrom 11.7% to 25.2%",

color = "Region") +

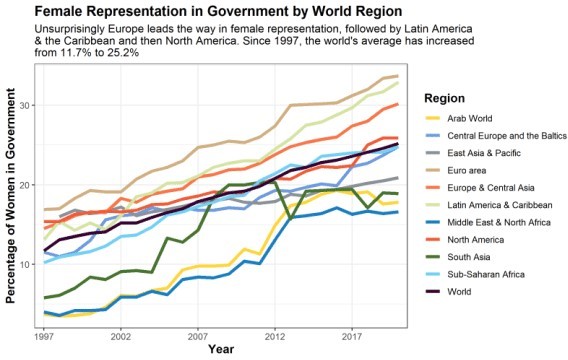
#Adjust the axes

theme(plot.title = element\_text(face="bold", size =14), axis.title.x = element\_text(face="bold", size = 12), axis.title.y = element\_text(face="bold", size = 12), legend.title = element\_text(face="bold", size = 12))

region.plot

# Save the plot!

ggsave("plots/RegionalPlot.png", region.plot, height = 5, width = 8)



The first plot is interesting. As we can see, female representation is highest in Europe and the Americas, while the Middle East, North Africa, and South East Asia are lagging behind. The biggest positive we can pull from this is that **the world average of women representation has more than doubled in the past 25 years to about a quarter.** However, that is still very low and hopefully we will see that number continue to rise. **At the current slope, we won’t see a 50/50 mix until around 2065, which is way too late for equal representation…**

## Graph #2 - Female Representation in Government by Country Income Level

Time to graph the second chart, this time **by income level of countries.**

# Next comes all the countries by income level

income.plot <- ggplot(df.income, aes(x=year,y=value,group= country\_name,colour=country\_name)) +

geom\_line(size = 1.5) + scale\_color\_simpsons() +

scale\_x\_discrete(breaks=c(1997,2002,2007,2012,2017)) + labs(x = "Year",

y = "Percentage of Women in Government",

title = "Female Representation in Government by Country Income Level",

subtitle = "Looking at it by income basically replicates the regional charts, with OECD and high income \ncountries leading the way with around 30% female representation. Interestingly heavily indebted

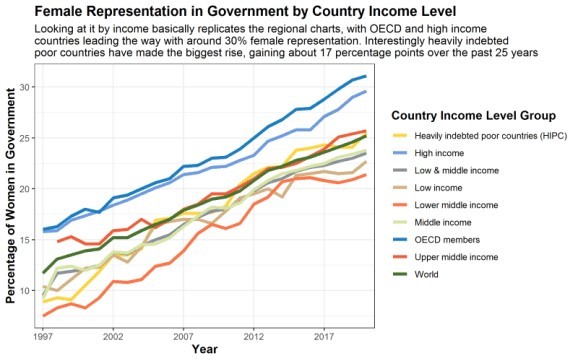
\npoor countries have made the biggest rise, gaining about 17 percentage points over the past 25 years",

color = "Country Income Level Group") + theme(plot.title = element\_text(face="bold", size =14),

axis.title.x = element\_text(face="bold", size = 12), axis.title.y = element\_text(face="bold", size = 12), legend.title = element\_text(face="bold", size = 12))

income.plot

ggsave("plots/IncomePlot.png", income.plot, height = 5, width = 8)



Much like the first plot, richer countries, that tend to be in Europe or the Americas, have a higher proportion of female representation in government. **The rankings break down neatly by income with the top three lines all being from upper middle to richer countries**, so there would definitely be a correlation between the two if you were to do further analysis.

## Graph #3 - Female Representation in Government by Country Income Level

Time to do some analysis on a few well-known countries (and some that are closer to my heart).

# Third comes my selected countries

country.plot <- ggplot(df.country, aes(x=year,y=value,group= country\_name,colour=country\_name)) +

geom\_line(size = 1.5) + scale\_color\_simpsons() +

scale\_x\_discrete(breaks=c(1997,2002,2007,2012,2017)) + labs(x = "Year",

y = "Percentage of Women in Government",

title = "Female Representation in Government from Selected

Countries",

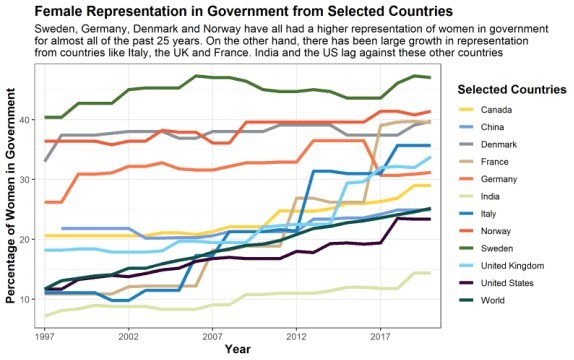
subtitle = "Sweden, Germany, Denmark and Norway have all had a higher representation of women in government \nfor almost all of the past 25 years. On the other hand, there has been large growth in representation \nfrom countries like Italy, the UK and France. India and the US lag against these other countries",

color = "Selected Countries") +

theme(plot.title = element\_text(face="bold", size =14), axis.title.x = element\_text(face="bold", size = 12), axis.title.y = element\_text(face="bold", size = 12), legend.title = element\_text(face="bold", size = 12))

country.plot

ggsave("plots/CountryPlot.png", country.plot, height = 5, width = 8)



The biggest takeaway from this plot is seeing **the plateau around 40% of female representation** in Germany, Denmark, Norway and Sweden. It will be interesting to see whether they ever reach that 50%. Meanwhile some **countries like Italy, the UK and France have shot up over the past ten years**. You can really see the jump in the UK in 2015 when the Labour Party (one of the top two) made a huge effort to field more female MPs, which brought up female representation in Parliament by about 7 percentage points! Definitely a positive for better representation and diversity.

## Graph #4 - Female Representation in Government by Country Income Level

**Now I wanted to plot some dumbbell charts,** a way to really compare data from two different points in history, or between two different groups. To plot these charts we use the geom\_dumbbell() function from the ggalt package, which allows you to add dumbbell-like icons to plot the data you initially called in the ggplot() function. This takes a lot more formatting, as you can see I use functions like geom\_text(), geom\_rect() and annotate() to add in some cool side borders and language. I do want to shout out to tessaeagle from GitHub for her dumbbell graph code from a recent TidyTuesday that I found on Twitter, which really helped me get the base code to make these next two charts!

So for the **first dumbbell chart I wanted to look at the 20 countries with the highest proportion of women in government**. I also show the percent increase that each country had from 1997 to 2020 on the right hand side.

# Okay now for the pretty dumbbell charts # The first dumbbell chart shows

top.dumbbell.plot <- ggplot(df.top.dumbbell, aes(y = country\_name, x = yr1997, xend = yr2020, group=country\_name)) +

geom\_dumbbell(size= 1, size\_x = 3, size\_xend = 3, colour\_x = "darkred", colour\_xend = "darkgreen", colour = "grey") +

labs(

y = "Country",

x = "Percentage of Women in Government",

title = "20 Countries with the Highest Proportion of Women \nin Government as of 2020",

subtitle = "Despite the higher income countries showing larger proportions of women in government, only 1 of

the top 10 countries in **2020** is from Europe and the top three countries are Rwanda, Cuba and Bolivia.

Definitely different than most people would expect!"

)+

scale\_x\_continuous(breaks = seq(from = 0, to = 60, by = 10)) + scale\_y\_discrete(limits = rev(levels(df.top.dumbbell$country\_name)))

+

theme(

panel.background = element\_rect(fill = "#FFEFCB", color = NA), plot.background = element\_rect(fill = "#FFEFCB", color = NA), panel.grid = element\_blank(),

axis.ticks = element\_blank(),

axis.text = element\_text(size = 12, color = "black"), axis.title.x = element\_text(face = "bold", size = 12, margin =

margin(t = 10, r = 0, b = 0, l = 0))

axis.title.y = element\_text(face = "bold", size = 12), plot.title = element\_text(size = 16, face = "bold"), plot.subtitle = element\_markdown()

) +

geom\_rect(data = df.top.dumbbell, aes(xmin = max(yr2020) \* 1.09, xmax

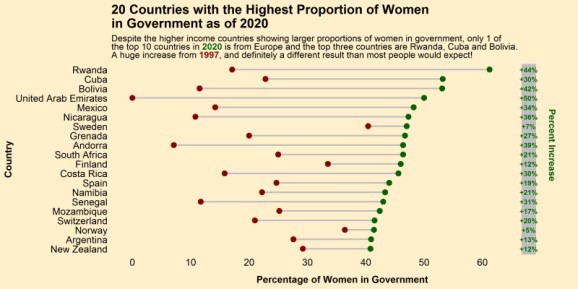
= max(yr2020) \* 1.13, ymin = -Inf, ymax = Inf), fill = "grey") + annotate(geom="text", x = 72, y= df.top.dumbbell$country\_name[9],

angle = 270,

label=c("Percent Increase"), color="darkgreen", size = 4, fontface = 'bold') +

geom\_text(data = df.top.dumbbell, aes(label = paste0("+", round(change), "%"), y = country\_name, x = max(yr2020) \* 1.11), fontface = "bold", size = 3, color = "darkgreen") top.dumbbell.plot

ggsave("plots/TopdumbbellPlot.png", top.dumbbell.plot, height = 5, width = 10)



This chart is telling because **despite all the progress you can see that only three countries actually break the 50% barrier in terms of representation**: Rwanda, Cuba and Bolivia.

Obviously 3/246 countries evaluated being over 50% female representation is not even close to equality, so the world still has a way to go. One interesting thing to note from this chart is the lack of European and North American countries in the top 10, showing that **maybe income is not always the best indicator of representation**!

## Graph #5 - Female Representation in Government by Country Income Level

Finally, the last chart for my Women’s Day post. This dumbbell chart looks at the **largest increases in representation by country.**

# The second dumbbell chart

change.dumbbell.plot <- ggplot(df.change.dumbbell, aes(y = country\_name, x = yr1997, xend = yr2020, group=country\_name)) +

geom\_dumbbell(size= 1, size\_x = 3, size\_xend = 3, colour\_x = "darkred", colour\_xend = "darkgreen", colour = "grey") +

labs(

y = "Country",

x = "Percentage of Women in Government",

title = "The 20 Countries with the Largest Increases of Women \nin Government from 1997 to 2020",

subtitle = "Many of the countries that have grown their proportion of women in government had less than 15% of

women in government in **1997**, but have since seen those levels grow 25-50 percentage

points by **2020**."

)+

scale\_x\_continuous(breaks = seq(from = 0, to = 60, by = 10)) + scale\_y\_discrete(limits = rev(levels(df.change.dumbbell$

country\_name))) + theme(

panel.background = element\_rect(fill = "#FFEFCB", color = NA), plot.background = element\_rect(fill = "#FFEFCB", color = NA), panel.grid = element\_blank(),

axis.ticks = element\_blank(),

axis.text = element\_text(size = 12, color = "black"),

axis.title.x = element\_text(face = "bold", size = 12, margin = margin(t = 10, r = 0, b = 0, l = 0))

axis.title.y = element\_text(face = "bold", size = 12), plot.title = element\_text(size = 16, face = "bold"), plot.subtitle = element\_markdown()

) +

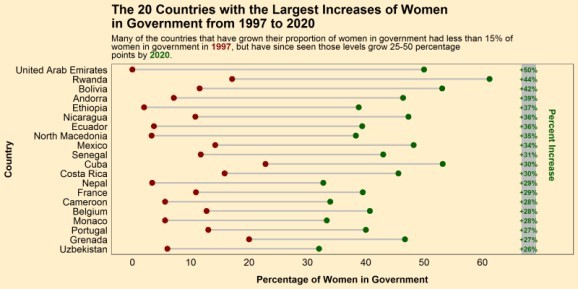
geom\_rect(data = df.change.dumbbell, aes(xmin = max(yr2020) \* 1.09, xmax = max(yr2020) \* 1.13, ymin = -Inf, ymax = Inf), fill = "grey") +

annotate(geom="text", x = 72, y= df.change.dumbbell$country\_name[9], angle = 270,

label=c("Percent Increase"), color="darkgreen", size = 4, fontface = 'bold') +

geom\_text(data = df.change.dumbbell, aes(label = paste0("+", round(change), "%"), y = country\_name, x = max(yr2020) \* 1.11), fontface = "bold", size = 3, color = "darkgreen") change.dumbbell.plot

ggsave("plots/ChangedumbbellPlot.png", change.dumbbell.plot, height = 5, width = 10)



The leader in this chart is the United Arab Emirates, which I find very surprising and peculiar. A bit more research shows that the UAE is not a democracy, so these are not democratically elected women, but appointed by the monarch. So this isn’t really equality in my mind.

The Full R code

# Women In Parliament

# Set wd

setwd("~/Coding/Blog/Programming/WomenParliament")

# Load the necessary packages

if(!require("tidyverse")) install.packages("tidyverse") # Our rock in data analysis (includes ggplot2)

if(!require("janitor")) install.packages("janitor") # cleans data like the best janitor out there!

if(!require("ggsci")) install.packages("ggsci") # My favourite palettes

if(!require("ggalt")) install.packages("ggalt") # Necessary for the dumbbell chart

if(!require("ggtext")) install.packages("ggtext") # ggtext allows for great text to include for charts

# Load the csv data file

df <- read\_csv("WomenInGovernment\_Data.csv")

# Change the column names to account for each year of the data 1997-2020

names <- colnames(df)[4:27]

names <- substring(names, 7,12)

colnames(df)[4:27] <- names

colnames(df) <- janitor::make\_clean\_names(colnames(df))

# Change all the ..s in the data to NAs and then convert the columns to numeric

df[,4:27] <- na\_if(df[,4:27], "..")

df[,4:27] <- sapply(df[,4:27],as.numeric) %>%

round(digits = 1)

# Create a dataset for each chart, just to make it .

# For this we filter the data for the region, income or country, then make the graph longer using the pivot\_longer() function

# Regional chart dataset

df.region <- df %>%

filter(category == "Region" | category == "World") %>%

pivot\_longer(cols = starts\_with("yr"), names\_to = "year", names\_prefix = "yr",

values\_to = "value", values\_drop\_na = TRUE)

# Income chart dataset

df.income <- df %>%

filter(category == "Income" | category == "World") %>%

pivot\_longer(cols = starts\_with("yr"), names\_to = "year", names\_prefix = "yr",

values\_to = "value", values\_drop\_na = TRUE)

# Selected Countries chart (Chose countries that are either near and dear to me or where my readers tend to be)

selected\_countries <- c("World", "Canada", "China", "Denmark", "France", "Germany", "India",

"Italy", "Norway", "Sweden", "United Kingdom", "United States")

df.country <- df %>%

filter(country\_name %in% selected\_countries| category == "World") %>%

pivot\_longer(cols = starts\_with("yr"), names\_to = "year", names\_prefix = "yr",

values\_to = "value", values\_drop\_na = TRUE)

# Top Female representation by country dumbbell chart

df.top.dumbbell <- df %>%

filter(category == "Country")

df.top.dumbbell$change <- df.top.dumbbell$yr2020 - df.top.dumbbell$yr1997

df.top.dumbbell <- head(arrange(df.top.dumbbell, desc(yr2020)), n = 20) %>%

droplevels()

df.top.dumbbell$country\_name <- factor(df.top.dumbbell$country\_name, levels = as.character(df.top.dumbbell$country\_name)) # to retain the order in plot.

# Top Female Increase in representation dumbbell chart

df.change.dumbbell <- df %>%

filter(category == "Country")

df.change.dumbbell$change <- df.change.dumbbell$yr2020 - df.change.dumbbell$yr1997

df.change.dumbbell <- head(arrange(df.change.dumbbell, desc(change)), n = 20) %>%

droplevels()

df.change.dumbbell$country\_name <- factor(df.change.dumbbell$country\_name, levels = as.character(df.change.dumbbell$country\_name)) # to retain the order in plot.

# Okay time to visualize!!!

# Let's set the theme and create a folder to store your output

theme\_set(theme\_bw())

dir.create("plots")

# First create a plot of all the regional representation of women in government

region.plot <- ggplot(df.region, aes(x=year,y=value,group=country\_name,color=country\_name)) +

geom\_line(size = 1.5) +

# My top color palette

scale\_color\_simpsons() +

# Set the x-axis scale

scale\_x\_discrete(breaks=c(1997,2002,2007,2012,2017)) +

# Update the labels of the axes

labs(x = "Year",

y = "Percentage of Women in Government",

title = "Female Representation in Government by World Region",

subtitle = "Unsurprisingly Europe leads the way in female representation, followed by Latin America \n& the Caribbean and then North America. Since 1997, the world's average has increased \nfrom 11.7% to 25.2%",

color = "Region") +

#Adjust the axes

theme(plot.title = element\_text(face="bold", size =14),

axis.title.x = element\_text(face="bold", size = 12),

axis.title.y = element\_text(face="bold", size = 12),

legend.title = element\_text(face="bold", size = 12))

region.plot

# Save the plot!

ggsave("plots/RegionalPlot.png", region.plot, height = 5, width = 8)

# Next comes all the countries by income level

income.plot <- ggplot(df.income, aes(x=year,y=value,group=country\_name,colour=country\_name)) +

geom\_line(size = 1.5) +

scale\_color\_simpsons() +

scale\_x\_discrete(breaks=c(1997,2002,2007,2012,2017)) +

labs(x = "Year",

y = "Percentage of Women in Government",

title = "Female Representation in Government by Country Income Level",

subtitle = "Looking at it by income basically replicates the regional charts, with OECD and high income \ncountries leading the way with around 30% female representation. Interestingly heavily indebted \npoor countries have made the biggest rise, gaining about 17 percentage points over the past 25 years",

color = "Country Income Level Group") +

theme(plot.title = element\_text(face="bold", size =14),

axis.title.x = element\_text(face="bold", size = 12),

axis.title.y = element\_text(face="bold", size = 12),

legend.title = element\_text(face="bold", size = 12))

income.plot

ggsave("plots/IncomePlot.png", income.plot, height = 5, width = 8)

# Third comes my selected countries

country.plot <- ggplot(df.country, aes(x=year,y=value,group=country\_name,colour=country\_name)) +

geom\_line(size = 1.5) +

scale\_color\_simpsons() +

scale\_x\_discrete(breaks=c(1997,2002,2007,2012,2017)) +

labs(x = "Year",

y = "Percentage of Women in Government",

title = "Female Representation in Government from Selected Countries",

subtitle = "Sweden, Germany, Denmark and Norway have all had a higher representation of women in government \nfor almost all of the past 25 years. On the other hand, there has been large growth in representation \nfrom countries like Italy, the UK and France. India and the US lag against these other countries",

color = "Selected Countries") +

theme(plot.title = element\_text(face="bold", size =14),

axis.title.x = element\_text(face="bold", size = 12),

axis.title.y = element\_text(face="bold", size = 12),

legend.title = element\_text(face="bold", size = 12))

country.plot

ggsave("plots/CountryPlot.png", country.plot, height = 5, width = 8)

# Okay now for the pretty dumbbell charts

# The first dumbbell chart shows

top.dumbbell.plot <- ggplot(df.top.dumbbell, aes(y = country\_name, x = yr1997, xend = yr2020, group=country\_name)) +

geom\_dumbbell(size= 1, size\_x = 3, size\_xend = 3, colour\_x = "darkred", colour\_xend = "darkgreen", colour = "grey") +

labs(

y = "Country",

x = "Percentage of Women in Government",

title = "20 Countries with the Highest Proportion of Women \nin Government as of 2020",

subtitle = "Despite the higher income countries showing larger proportions of women in government, only 1 of <br>the top 10 countries in <span style = 'color:darkgreen'><b>2020</b></span> is from Europe and the top three countries are Rwanda, Cuba and Bolivia. <br>Definitely different than most people would expect!"

)+

scale\_x\_continuous(breaks = seq(from = 0, to = 60, by = 10)) +

scale\_y\_discrete(limits = rev(levels(df.top.dumbbell$country\_name))) +

theme(

panel.background = element\_rect(fill = "#FFEFCB", color = NA),

plot.background = element\_rect(fill = "#FFEFCB", color = NA),

panel.grid = element\_blank(),

axis.ticks = element\_blank(),

axis.text = element\_text(size = 12, color = "black"),

axis.title.x = element\_text(face = "bold", size = 12, margin = margin(t = 10, r = 0, b = 0, l = 0)),

axis.title.y = element\_text(face = "bold", size = 12),

plot.title = element\_text(size = 16, face = "bold"),

plot.subtitle = element\_markdown()

) +

geom\_rect(data = df.top.dumbbell, aes(xmin = max(yr2020) \* 1.09, xmax = max(yr2020) \* 1.13, ymin = -Inf, ymax = Inf), fill = "grey") +

annotate(geom="text", x = 72, y= df.top.dumbbell$country\_name[9], angle = 270,

label=c("Percent Increase"), color="darkgreen", size = 4, fontface = 'bold') +

geom\_text(data = df.top.dumbbell, aes(label = paste0("+", round(change), "%"), y = country\_name, x = max(yr2020) \* 1.11), fontface = "bold", size = 3, color = "darkgreen")

top.dumbbell.plot

ggsave("plots/TopdumbbellPlot.png", top.dumbbell.plot, height = 5, width = 10)

# The second dumbbell chart

change.dumbbell.plot <- ggplot(df.change.dumbbell, aes(y = country\_name, x = yr1997, xend = yr2020, group=country\_name)) +

geom\_dumbbell(size= 1, size\_x = 3, size\_xend = 3, colour\_x = "darkred", colour\_xend = "darkgreen", colour = "grey") +

labs(

y = "Country",

x = "Percentage of Women in Government",

title = "The 20 Countries with the Largest Increases of Women \nin Government from 1997 to 2020",

subtitle = "Many of the countries that have grown their proportion of women in government had less than 15% of <br>women in government in <span style = 'color:darkred'><b>1997</b></span>, but have since seen those levels grow 25-50 percentage <br>points by <span style = 'color:darkgreen'><b>2020</b></span>."

)+

scale\_x\_continuous(breaks = seq(from = 0, to = 60, by = 10)) +

scale\_y\_discrete(limits = rev(levels(df.change.dumbbell$country\_name))) +

theme(

panel.background = element\_rect(fill = "#FFEFCB", color = NA),

plot.background = element\_rect(fill = "#FFEFCB", color = NA),

panel.grid = element\_blank(),

axis.ticks = element\_blank(),

axis.text = element\_text(size = 12, color = "black"),

axis.title.x = element\_text(face = "bold", size = 12, margin = margin(t = 10, r = 0, b = 0, l = 0)),

axis.title.y = element\_text(face = "bold", size = 12),

plot.title = element\_text(size = 16, face = "bold"),

plot.subtitle = element\_markdown()

) +

geom\_rect(data = df.change.dumbbell, aes(xmin = max(yr2020) \* 1.09, xmax = max(yr2020) \* 1.13, ymin = -Inf, ymax = Inf), fill = "grey") +

annotate(geom="text", x = 72, y= df.change.dumbbell$country\_name[9], angle = 270,

label=c("Percent Increase"), color="darkgreen", size = 4, fontface = 'bold') +

geom\_text(data = df.change.dumbbell, aes(label = paste0("+", round(change), "%"), y = country\_name, x = max(yr2020) \* 1.11), fontface = "bold", size = 3, color = "darkgreen")

change.dumbbell.plot

ggsave("plots/ChangedumbbellPlot.png", change.dumbbell.plot, height = 5, width = 10)

# Do it all in Plotly

library(plotly)

fig <- plot\_ly(df.income, x = ~year, y = ~value, name = ~country\_name, type = 'scatter', mode = 'lines')

fig