

# CT5100 Data Visualisation: Assignment 2

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## Introduction

Matt Ridley's hypothesis states that "fertility rates everywhere will converge to 2.1 in a few decades, and the world population will stabilise at 9 billion people" This is known as the replacement rate. When this rate is hit the population will begin to shrink.

Our project uses compiled datasets of fertility rates, education and population growth using data collected from the World Bank and United Nations data sources. Our collective objective is to illustrate through appropriate visualisations the evidence that might support Matt Ridley's hypothesis. Our four main visualisations are as follows:

1. Percentage increase in world population from 1950 till present.
2. Changing fertility rates from 1950 to the present for selected countries of the world.
3. Regional view of the above visualisation
4. The observed relationship between fertility & per capita income & education

To execute these visualizations, we used a programming language known as R. To create the best visualizations, we utilized many different R packages such as; scales; ggplot2; reshape; dplyr and many more. After exploring and creating the visualizations we then uploaded them to a HTML page rendered from a RMarkdown script.

## 1. Percentage increase in world population.

The dataset below was obtained from the UN data source:

1. World Population <http://data.un.org/Data.aspx?q=world+population&d=PopDiv&f=variableID%3a12%3bcrID%3a900>.

```
head(world_pop)
```

##	Country.or.Area	Year.s.	Variant	Value
## 1	World	2018	No change	7637740
## 2	World	2017	No change	7553284
## 3	World	2016	No change	7468369
## 4	World	2015	No change	7383009
## 5	World	2014	No change	7298453
## 6	World	2013	No change	7213426

## Plots:

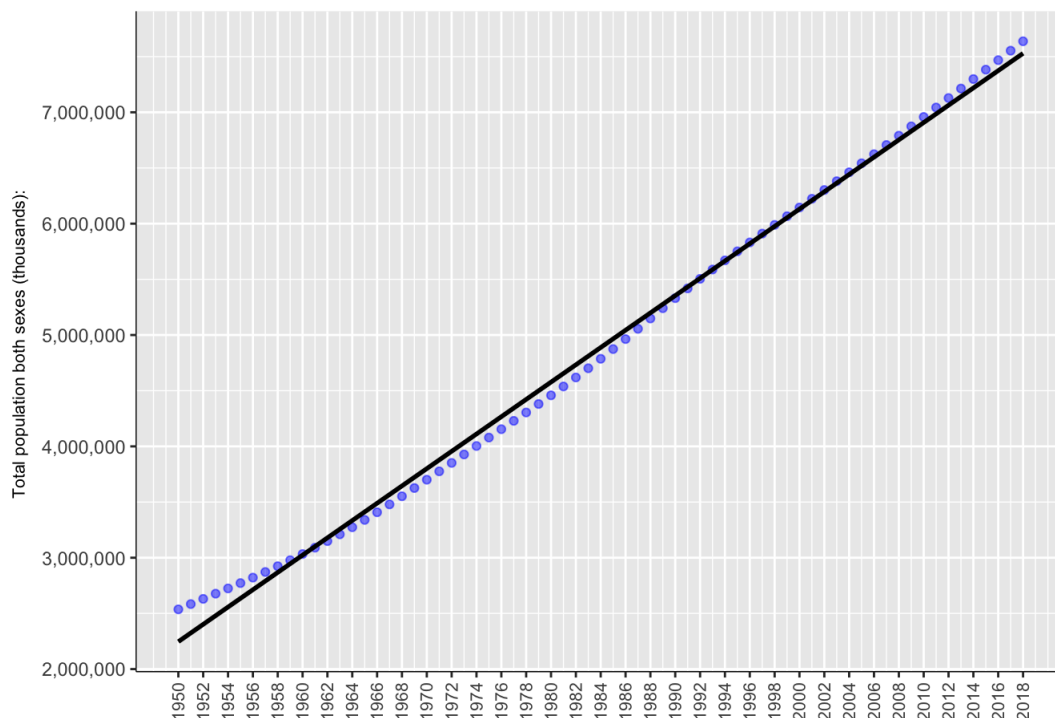
Plot 1.1 below illustrates the world population growth since 1950. The graph shows the gradual increase in the world's population.

Instead of designing this graph with only one line we decided to opt for the actual population growth followed by a fitted linear regression line to emphasize the upwards trend in world population. The solid black line represents the upwards population trend, while the actual population growth is displayed by the spotted blue line. With simplicity in mind we decided to separate our X-axis into two year break ticks instead of individual years. We felt that this left our graph less cluttered, while still allowing you to visualize as much information as possible.

```
p1.1 <- ggplot(world_pop , aes(x=Year.s., y=Value))
p1.1 + geom_point(na.rm = TRUE, alpha=0.5, color = "blue") +
  ggtitle("1.1 World Population Growth 1950-Present:") +
  geom_smooth(method=lm, se=FALSE, color = "black") +
  scale_x_log10(labels = trans_format("log10", math_format(10^x)), breaks = scales::trans_breaks("log10", function(x) 10^x)) +
  scale_size_area(max_size = 10, labels = comma) +
  scale_x_continuous(lim=c(1950,2018), breaks=seq(1950,2018,2)) +
  scale_y_continuous(name="Total population both sexes (thousands): ", labels = comma) +
  theme(axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 12, face = "bold"), axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5), title = element_text(size=8), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))
```

```
## Scale for 'x' is already present. Adding another scale for 'x', which
## will replace the existing scale.
```

### 1.1 World Population Growth 1950-Present:



Plot 1.2 below illustrates the world population percentage growth versus the previous year since 1950. The graph shows the gradual downward trend in the percentage increase compared to the previous year.

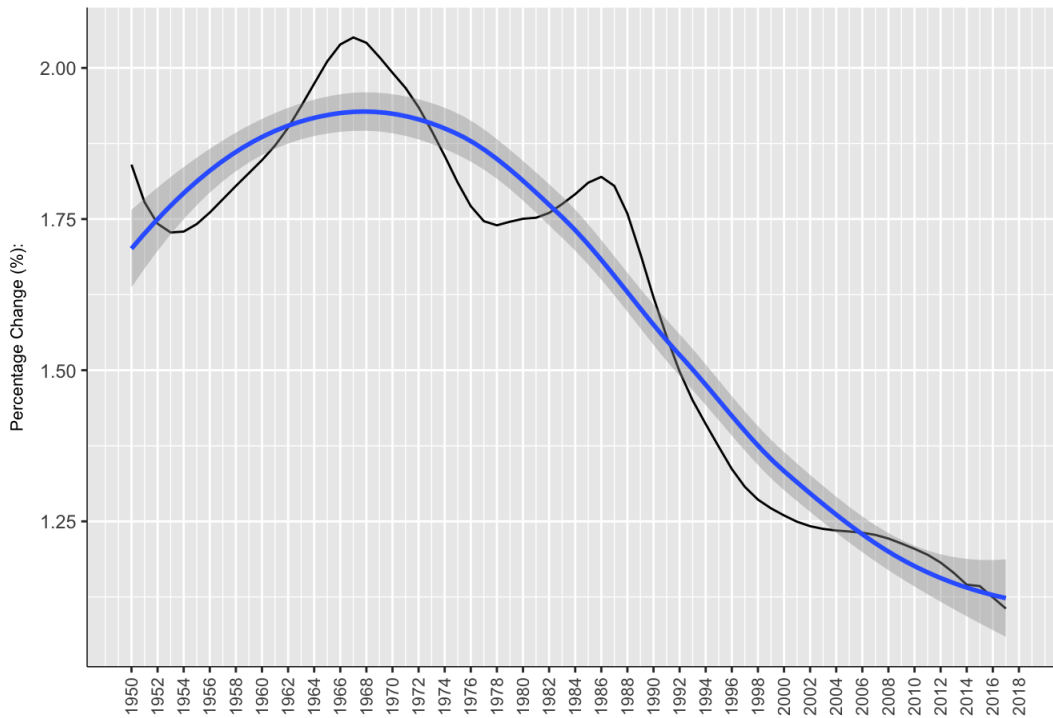
Instead of designing this graph with only one line we decided to opt for the actual percentage values followed by a regression line to emphasize the downwards trend in percentage increase compared to the previous year. The solid black line represents the actual percentage values, while the blue line is a regression line. Continuing with the above design in mind we decided to separate our X-axis into two year break ticks instead of individual years. We felt that this left our graph less cluttered, while still allowing you to visualize as much information as possible.

```
# Percentage increase for each year
Year <- world_pop $Year.s.
Population <- world_pop $Value
df <- data_frame(Year, Population)

df %>% mutate(Previous_Year = lag(Population, 1), Change = Previous_Year - Population, Percent_Change = Change/Previous_Year*100) %>%
  ggplot(aes(x = Year, y = Percent_Change)) +
  geom_line() +
  geom_smooth() +
  xlab("Year") +
  ylab("Percentage Change (%)") +
  ggtitle("1.2 Actual vs Average percentage change in World Population vs previous year 1950-Present:") +
  scale_x_continuous(lim=c(1950,2018), breaks=seq(1950,2018,2)) +
  theme(axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 12, face = "bold"),
        axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5),
        title = element_text(size=8), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))
```

```
## `geom_smooth()` using method = 'loess'
```

## 1.2 Actual vs Average percentage change in World Population vs previous year 1950-2018



Plots 1.1 & 1.2 show the increase in the world's population and also the decrease in the percentage increase compared to the previous year. This shows that as the world is getting closer to 9 billion people, it seems to be stabilising. This confirms part of Matt Ridley's hypothesis.

## 2. Changing fertility rates from 1950 to the present for selected countries of the world.

The datasets below were obtained from the World Bank & UN data source:

1. GDP <http://data.un.org/Data.aspx?q=world+population&d=PopDiv&f=variableID%3a12%3bcrID%3a900>.
2. Fertility Rates <http://data.un.org/Data.aspx?q=fertility%20rates&d=PopDiv&f=variableID%3a54>.

These datasets were then edited online prior to downloading.

```
head(countries)
```

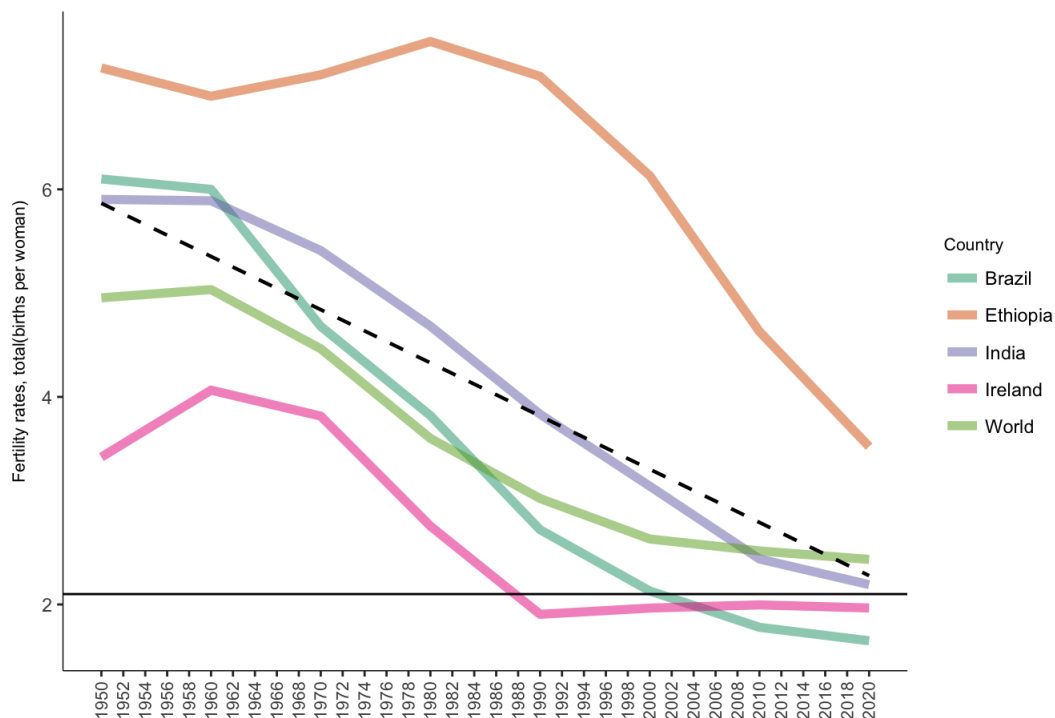
##	Country	Year	End_Year	Variant	Value
## 1	Brazil	2020	2025	Medium	1.65
## 2	Brazil	2010	2015	Medium	1.78
## 3	Brazil	2000	2005	Medium	2.13
## 4	Brazil	1990	1995	Medium	2.72
## 5	Brazil	1980	1985	Medium	3.82
## 6	Brazil	1970	1975	Medium	4.68

## Plots:

Graph 2 below is a comparison visualization of four different countries' rates versus the world average fertility rate since 1950. Each country represents either a 'rich' (Ireland), 'poor' (Ethiopia), 'newly developed' (Brazil) or 'developing' (India) country. To best differentiate the countries, we used the 'Dark2' palette. We felt that this clearly distinguished each country from one another. The plot includes a black line, which is Ridley's hypothesis of a 2.1 fertility rate. This was executed by using the 'geom\_hline' function in the ggplot2 package. As there are several times our data overlaps we decided to add some transparency to the graph. We did this by setting the alpha function to 0.5, which resulted in a much clearer plot. The black dashed line is a downward sloping regression line, which allows the viewer to visualize the trajectory of fertility rates.

```
countries$Year <- as.integer(countries$Year)
p2 <- ggplot(countries, aes(x=Year, y=Value, color=Country))
p2 + geom_line(alpha = 0.5, size = 2 ) +
  stat_smooth(method=lm, se= FALSE, colour = "black", size = 0.8, linetype="dashed")+
  scale_x_continuous(lim=c(1950,2020), breaks=seq(1950,2020,2)) +
  ylab("Fertility rates, total (births per woman)") +
  ggtitle("2. Changing fertility rates from 1950-2020:") +
  geom_hline(yintercept=2.1) +
  theme( panel.background = element_blank() ) +
  scale_color_brewer(palette='Dark2') +
  theme(panel.grid.major = element_blank(), panel.background = element_blank(), axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 12, face = "bold"),
        axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5),
        title = element_text(size=8), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))
```

## 2. Changing fertility rates from 1950-2020:



Originally, Brazil had a higher fertility rate than India, but from 1960 onwards Brazil's fertility rate was less than India's. This is important because Brazil is seen as a newly developed country, while India is seen as a developing country.

This graph highlights the downward trend of worldwide fertility rates. It shows this rate getting closer to the 2.1 replacement rate.

## 3. Regional view of the above visualisation

The datasets below were obtained from the World Bank & UN data source:

1. GDP <http://data.un.org/Data.aspx?q=world+population&d=PopDiv&f=variableID%3a12%3bcrID%3a900>.
2. Fertility Rates <http://data.un.org/Data.aspx?q=fertility%20rates&d=PopDiv&f=variableID%3A54>.

These datasets were then edited online prior to downloading.

```
head(region)
```

```
##      Regions Year End_Year Variant Value
## 1 Central Asia 2020      2025   Medium 2.436
## 2 Central Asia 2010      2015   Medium 2.707
## 3 Central Asia 2000      2005   Medium 2.514
## 4 Central Asia 1990      1995   Medium 3.572
## 5 Central Asia 1980      1985   Medium 4.134
## 6 Central Asia 1970      1975   Medium 5.024
```

## Plots:

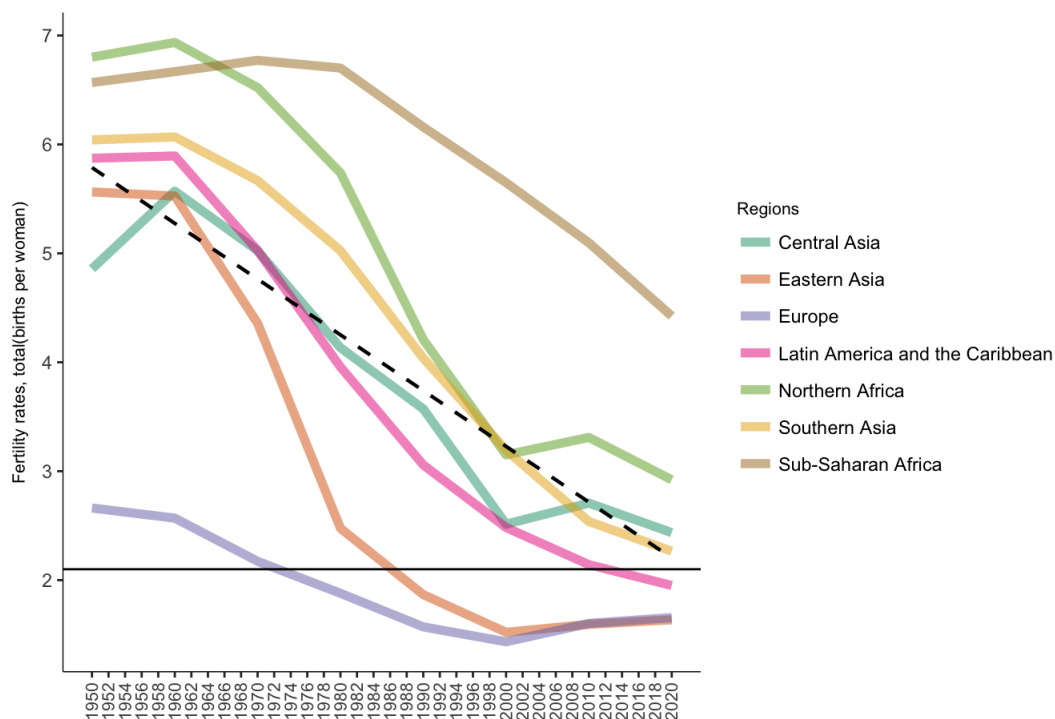
Graph 3 below is compiled from the same data from the regions designated in this week's dataset. The seven different regions are clearly

distinguished by the 'Dark2' palette. We incorporated a regression to illustrate which direction fertility rates are headed, which is negatively sloped. Since the dataset goes back all the way to 1950 we felt that it was better to separate the X-axis by every two years. As there were many lines in this visualization transparency within the plot was crucial, so we set the alpha to 0.5

```
region$Year <- as.integer(region$Year)

p3 <- ggplot(region, aes(x=Year, y=Value, color=Regions))
p3 + geom_line(alpha = 0.5, size = 2) +
  stat_smooth(method=lm, se= FALSE, colour = "black", size = 0.8, linetype="dashed")+
  scale_x_continuous(lim=c(1950,2020), breaks=seq(1950,2020,2)) +
  ylab("Fertility rates, total(births per woman)") +
  ggtitle("3. Regional view of the above visualisation:") +
  geom_hline(yintercept=2.1) +
  theme(panel.background = element_blank()) +
  scale_color_brewer(palette='Dark2') +
  theme(panel.grid.major = element_blank(), panel.background = element_blank(), axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 12, face = "bold"),
        axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5),
        title = element_text(size=8), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))
```

### 3. Regional view of the above visualisation:



This graph highlights the downward trend of regional fertility rates. It shows this rate getting closer to the 2.1 replacement rate.

## 4. Observed relationship between fertility & per capita income & education

The datasets below were obtained from the World Bank data source:

1. GDP <https://data.worldbank.org/indicator/NY.GDP.PCAP.CD?end=2016&start=1970&view=chart>.
2. Literacy Rates <https://data.worldbank.org/indicator/SE.ADT.LITR.ZS>.
3. Fertility Rates <https://data.worldbank.org/indicator/SP.DYN.TFRT.IN?end=2016&start=1970>.

These datasets were then combined and edited within Microsoft Excel.

```
head(rates)
```

##	Country	Year	Fertility.Rate	GDP	Literacy.Rate
## 1	World	1975	4.154705	1449.619	68.76034
## 2	World	1976	4.040733	1549.510	68.89088
## 3	World	1977	3.939673	1722.394	69.01872
## 4	World	1978	3.848233	1992.436	69.42679
## 5	World	1979	3.780500	2274.798	69.82894
## 6	World	1980	3.722039	2516.286	70.04151

## Plots:

Plot 4.1, 4.2 and 4.3 below illustrates the relationship between fertility & per capita income & education from 1975-2015. The graph shows the downward and upward trends of the relationships between the variables.

4.1 Graph 4.1 shows the average worldwide fertility rates. It shows the downward trend of the decrease in fertility rates worldwide. A regression line was fitted to show and emphasise this trend. As we can see the fertility rates have dramatically dropped since 1975. In this graph a 2.1 replacement line was fitted to show Matt Ripleys fertility hypothesis. As we can see from the graph and the fitted regression line we are no far off of the 2.1 replacement rate for fertility rates.

4.2 Graph 4.2 shows the average worldwide GDP per capita in USD from 1975 to 2015. It shows the upward trend of the increase in average worldwide GDP per capita. A regression line was also fitted to show and emphasise this trend. As we can see the average GDP per capita has increased by more than seven times since 1975.

4.3 Graph 4.3 shows the average worldwide adult literacy rate from 1965 to 2015. We decided to use literacy rates as a means to measure the level of education of a country. Datasets relation to literacy rates were also easier to find. A regression line was fitted to show the upward trend in literacy rates. Literacy rates have increased by almost 20% since 1975. Literacy rates also seem to be rising steadily compared to the other two graphs.

Continuing with the above design in mind we decided to separate our X-axis into two year break ticks instead of individual years. We felt that this left our graph less cluttered, while still allowing you to visualise as much information as possible.

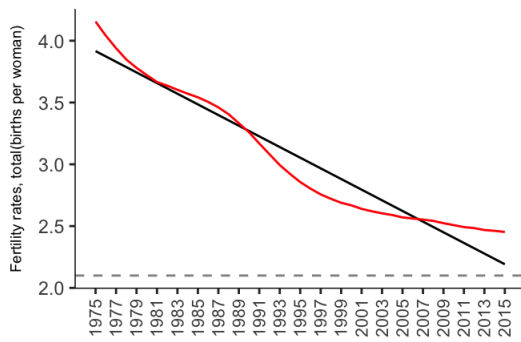
```
p4.1 <- ggplot(rates, aes(x = Year, y = Fertility.Rate)) +
  stat_smooth(method = lm, se = FALSE, color = "black", size = 0.5) +
  geom_line(color = "red") +
  geom_hline(yintercept=2.1, alpha = 0.5, linetype="dashed") +
  scale_x_continuous(lim=c(1975,2015), breaks=seq(1975,2015,2)) +
  ggtitle("4.1 Worldwide Fertility rates 1975-2015:") +
  ylab("Fertility rates, total (births per woman)") +
  theme(panel.grid.major = element_blank(), panel.background = element_blank(), axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 9, face = "bold"),
        axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5),
        title = element_text(size=7), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))

p4.2 <- ggplot(rates, aes(x=Year, y=GDP)) +
  stat_smooth(method = lm, se = FALSE, color = "black", size = 0.5) +
  geom_line(color = "green") +
  ggtitle("4.2 Avg Worldwide GDP per capita USD 1975-2015:") +
  ylab("GDP per capita (current USD)") +
  scale_x_continuous(lim=c(1975,2015), breaks=seq(1975,2015,2)) +
  theme(panel.grid.major = element_blank(), panel.background = element_blank(), axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 9, face = "bold"),
        axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5),
        title = element_text(size=7), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))

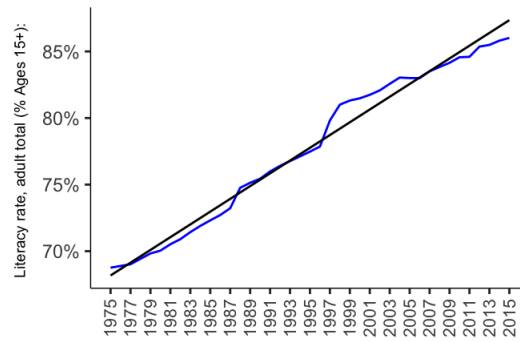
p4.3 <- ggplot(rates, aes(x=Year, y=Literacy.Rate)) +
  geom_line(color = "blue") +
  stat_smooth(method = lm, se = FALSE, color = "black", size = 0.5) +
  scale_x_continuous(lim=c(1975,2015), breaks=seq(1975,2015,2)) +
  scale_y_continuous(name="Literacy rate, adult total (% Ages 15+): ", labels = function(x){ paste0(x, "%")
}) +
  ggtitle("4.3 Avg Worldwide adult literacy rates 1975-2015:") +
  theme(panel.grid.major = element_blank(), panel.background = element_blank(), axis.line = element_line(colour = "black", size = 0.25), plot.title = element_text(size = 8, face = "bold"),
        axis.title.x=element_blank(), axis.text.x = element_text(size=8, angle = 90, hjust=1, vjust = .5),
        title = element_text(size=7), legend.key = element_rect(fill = NA, colour = NA, size = 0.25))

# plot the graphs together using the multiplot function.
multiplot(p4.1, p4.2, p4.3, cols=2)
```

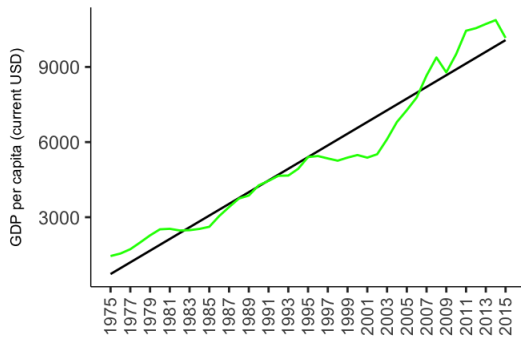
**4.1 Worldwide Fertility rates 1975-2015:**



**4.3 Avg Worldwide adult literacy rates 1975-2015:**



**4.2 Avg Worldwide GDP per capita USD 1975-2015:**



The above graphs show an interesting relationship between fertility & per capita income & education from 1975-2015. As literacy rates and GDP rates rise worldwide fertility rates decrease. You can definitely see a clear correlation between the 3 graphs. Graph 4.2 and 4.3 are almost mirrors of graph 4.1. You could almost overlay graph 4.2 onto 4.3. This goes with the trend of people being more educated having less kids and also people being richer having access to better health care and better education.

## Conclusion:

In conclusion if we take all the above graphs we can see evidence that supports Matt Ridley's hypothesis. In graphs 2, 3 & 4.1 we can see the world's fertility rates are nearing Matt's replacement rate of 2.1 births per women. Combine this with graphs 1.1 and 1.1 with the increase in population getting closer to the proposed 9 billion cut off point and followed by the decrease in the percentage increase compared to the previous year we have a lot of evidence in favour of the hypothesis. Finally looking at the last three graphs 4.1, 4.2 and 4.3 we see the downward trend in worldwide fertility rates correlating with the upward trend in GDP per capita and literacy rates worldwide. This suggests the world is getting richer and people are getting more education which again is leading to the replacement fertility rate of 2.1. We can safely say that the above visualisations are appropriate in supporting Matt Ridley's hypothesis.

## Group member contribution:

Kevin Derrane: I did part four: Observed relationship between fertility & per capita income & education. I did the three graphs for this followed by the write up about them. I also put together this R Markdown file.

Robbie Deegan: I did parts two: Changing fertility rates from 1950 to the present for selected countries of the world & also part three: Regional view of the above visualisation. I created the graphs and I also did my write up on these.

Jamie O'Halloran: I did part one: Percentage increase in world population. I created the graphs and I also did my write up on these.

Group: As a group we all feel we contributed evenly to this project. We all worked with each other to keep a theme for our graphs and helped each other where necessary. We did the introduction and conclusion together.