

PSY6422_module_project

2023-05-11

Analysis and Visualisation for the PSY6422 module project

Background

Vandenbroucke (2016) claimed that there is generally an inverse correlation between income and the total fertility rate within and between nations. Higher income allows people to have an access to the higher the degree of education. It was found that fewer children are born in any developed country. According Rowthorn (2011), the married women who share the same religion with her spouse tend to have more babies, and less likely to not produce any. Also, Storm (2017) found that lower income is associated with more religiosity, a negative relationship. In this analysis I wanted to confirm if their claim is to be true. Thus following three hypotheses were formulated.

Hypotheses:

- 1: Countries with greater PPP per capita would show lower fertility rate.
- 2: Countries with greater religiosity would show greater fertility rate.
- 3: Countries with greater PPP per capita would be less religious.

Methods

Data collection

For 'PPP per capita', International Monetary Fund (IMF) data from Statistics Times (2021) is used for the analysis. For 'fertility rate', Dataset for World Population Review (2023) is used for analysis.

Analysis

```
#libraries  
library(conflicted)
```

```
## Warning: package 'conflicted' was built under R version 4.2.3
```

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 4.2.3
```

```
## Warning: package 'ggplot2' was built under R version 4.2.3
```

```
## Warning: package 'tibble' was built under R version 4.2.3

## Warning: package 'dplyr' was built under R version 4.2.3

## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.2      v readr      2.1.4
## v forcats    1.0.0      v stringr    1.5.0
## v ggplot2     3.4.2      v tibble     3.2.1
## v lubridate  1.9.2      v tidyr      1.3.0
## v purrr       1.0.1

conflict_prefer("filter", "dplyr")

## [conflicted] Will prefer dplyr::filter over any other package.

conflict_prefer("lag", "dplyr")

## [conflicted] Will prefer dplyr::lag over any other package.

library(readxl) # Load the readxl package; data is excel file

## Warning: package 'readxl' was built under R version 4.2.3

library(ggplot2)
library(ggrepel)

## Warning: package 'ggrepel' was built under R version 4.2.3

library(gganimate)

## Warning: package 'gganimate' was built under R version 4.2.3

#library(ggrepel) for naming the dots in regional scatter?

# Set the working directory to the folder 'PSY6422_module_project'
setwd("C:/Users/kinu1/Documents/PSY6422_module_project")

# Import the data using the read_excel function
data <- read_excel("data/Fertility_data.xlsx")

# Remove rows with missing values
df<- data[complete.cases(data), ]

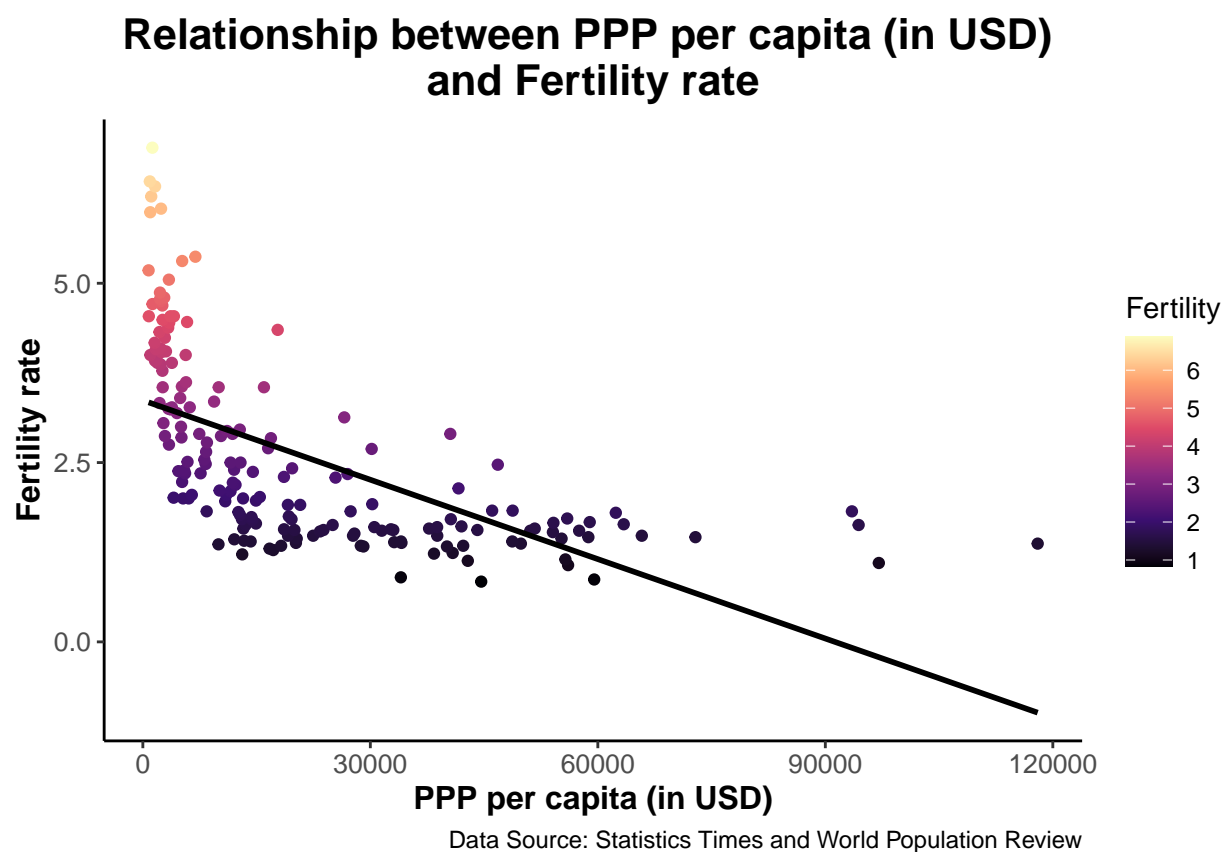
# Rename the variables
colnames(df) <- c("Country", "PPP", "Religiosity", "Fertility", "Region")
```

Scatterplots

The General relationship between PPP per capita and Fertility rate

```
ggplot(df, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita")
  scale_color_viridis_c(option = "magma") +
  geom_smooth(method = "lm", se = FALSE, color = "black") +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



The general relationship between fertility rate and PPP show a negative correlation. Therefore, hypothesis 1 is supported here.

Regional difference in the relationship

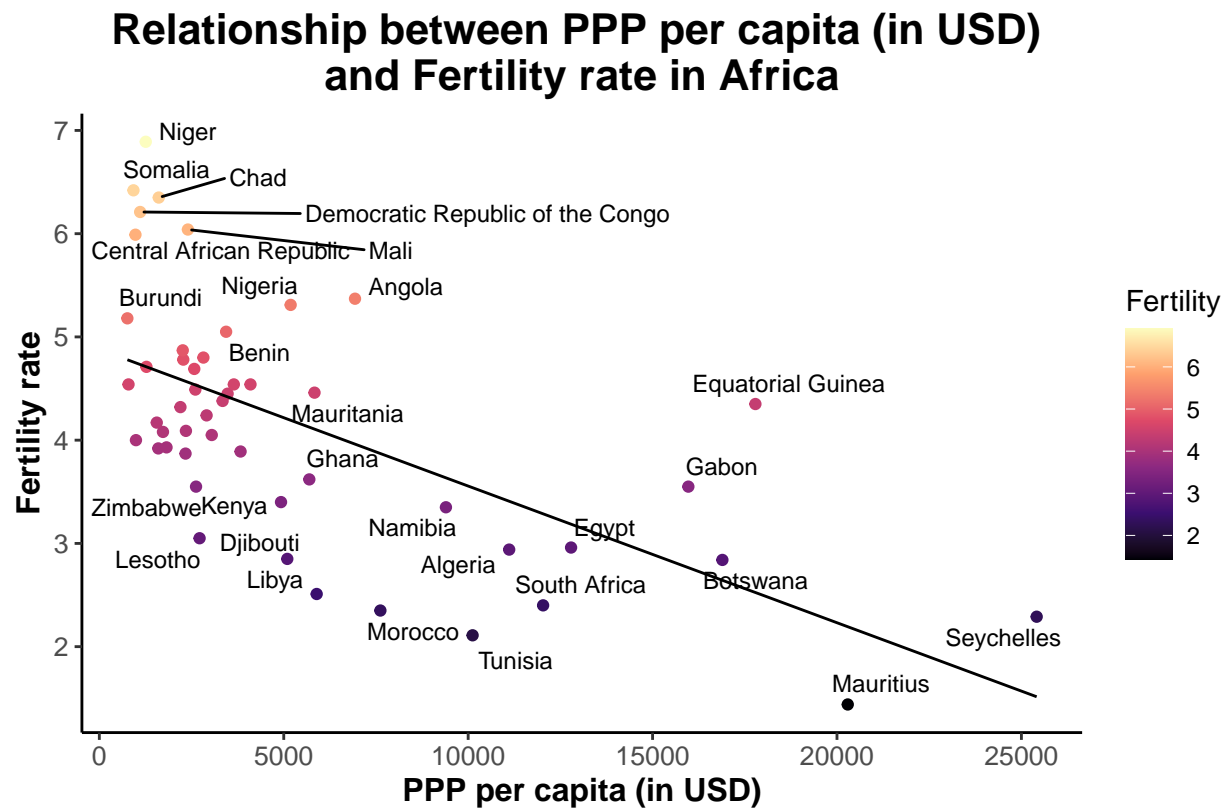
Africa

```
# Subset the dataframe to include only Africa data
df_africa <- subset(df, Region == "Africa")
```

```
ggplot(df_africa, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita and Fertility rate in Africa") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## Warning: ggrepel: 22 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



Data Source: Statistics Times and World Population Review

African fertility rate and PPP show a negative correlation.

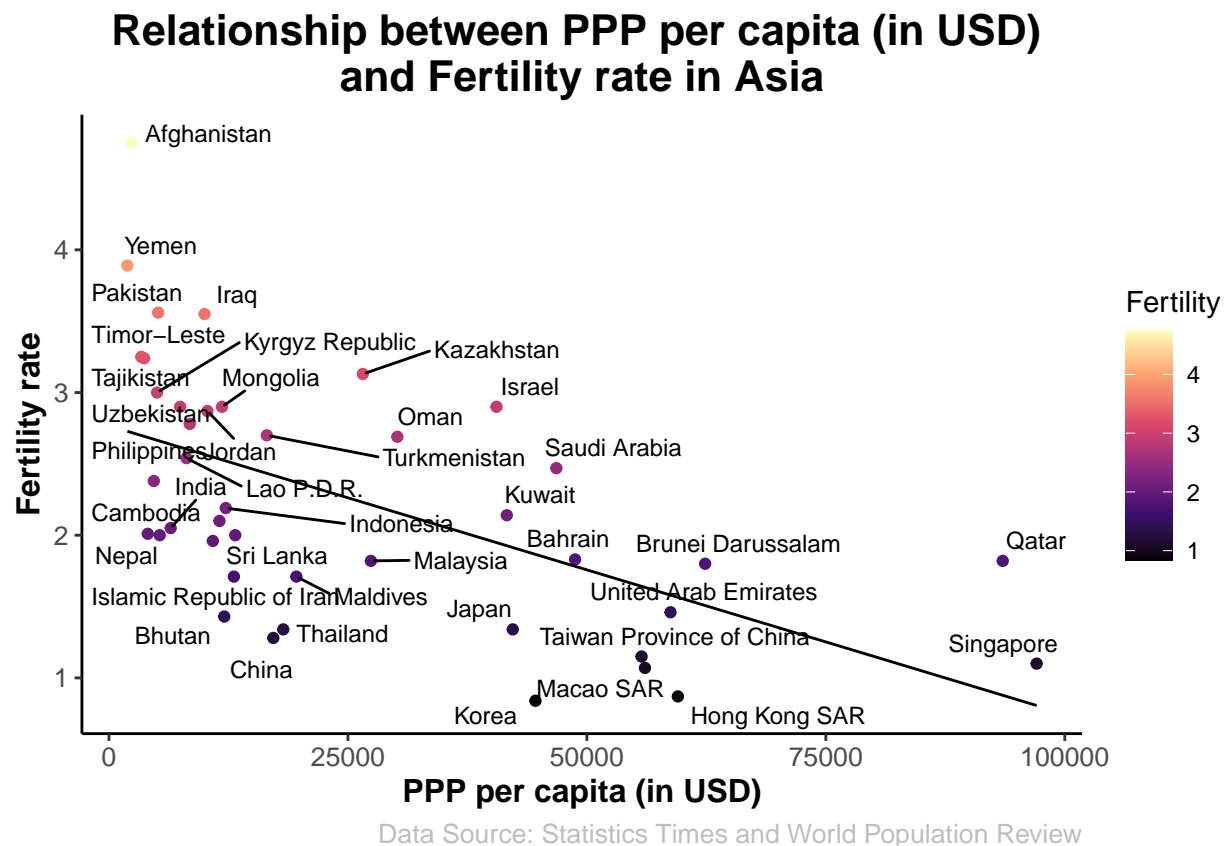
Asia

```
# Subset the dataframe to include only Asia data
df_asia <- subset(df, Region == "Asia")

ggplot(df_asia, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita and Fertility rate in Asia") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))

## 'geom_smooth()' using formula = 'y ~ x'

## Warning: ggrepel: 3 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



Asian fertility rate and PPP show a negative correlation.

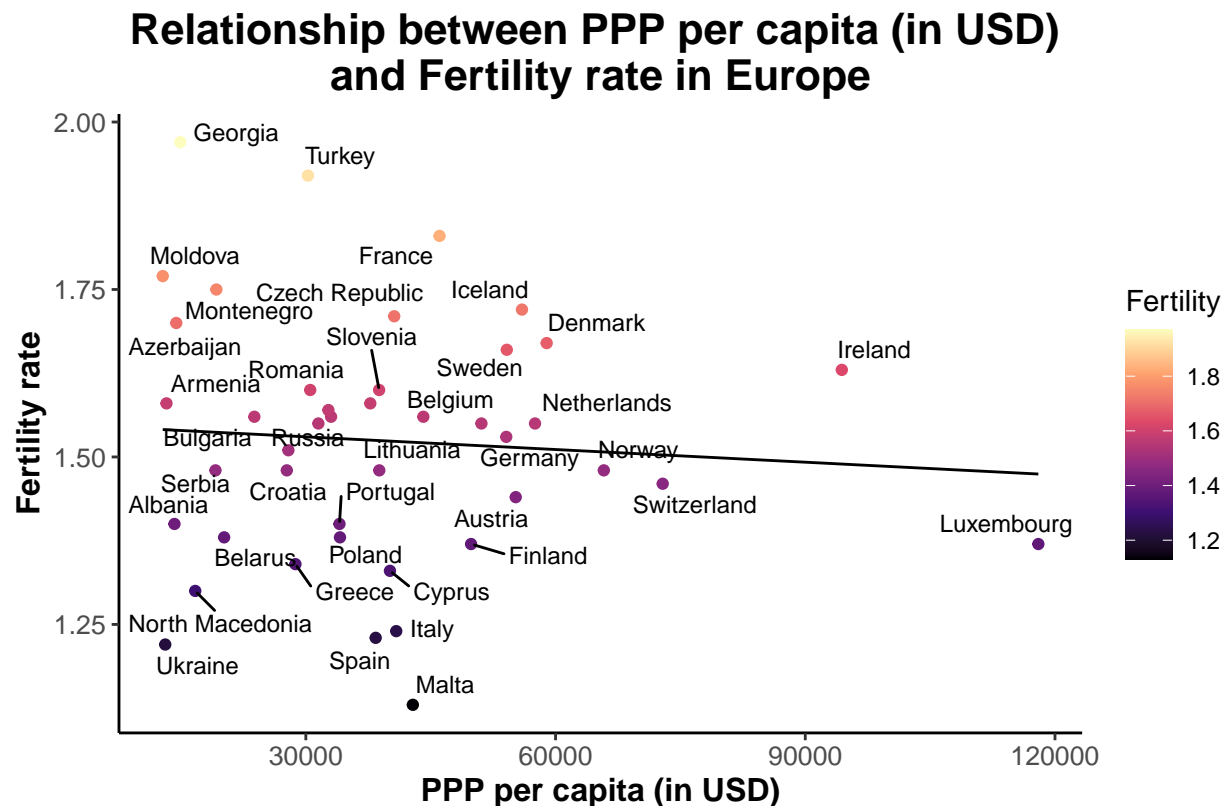
Europe

```
# Subset the dataframe to include only Europe data
df_europe <- subset(df, Region == "Europe")

ggplot(df_europe, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.le
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## Warning: ggrepel: 5 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



Data Source: Statistics Times and World Population Review

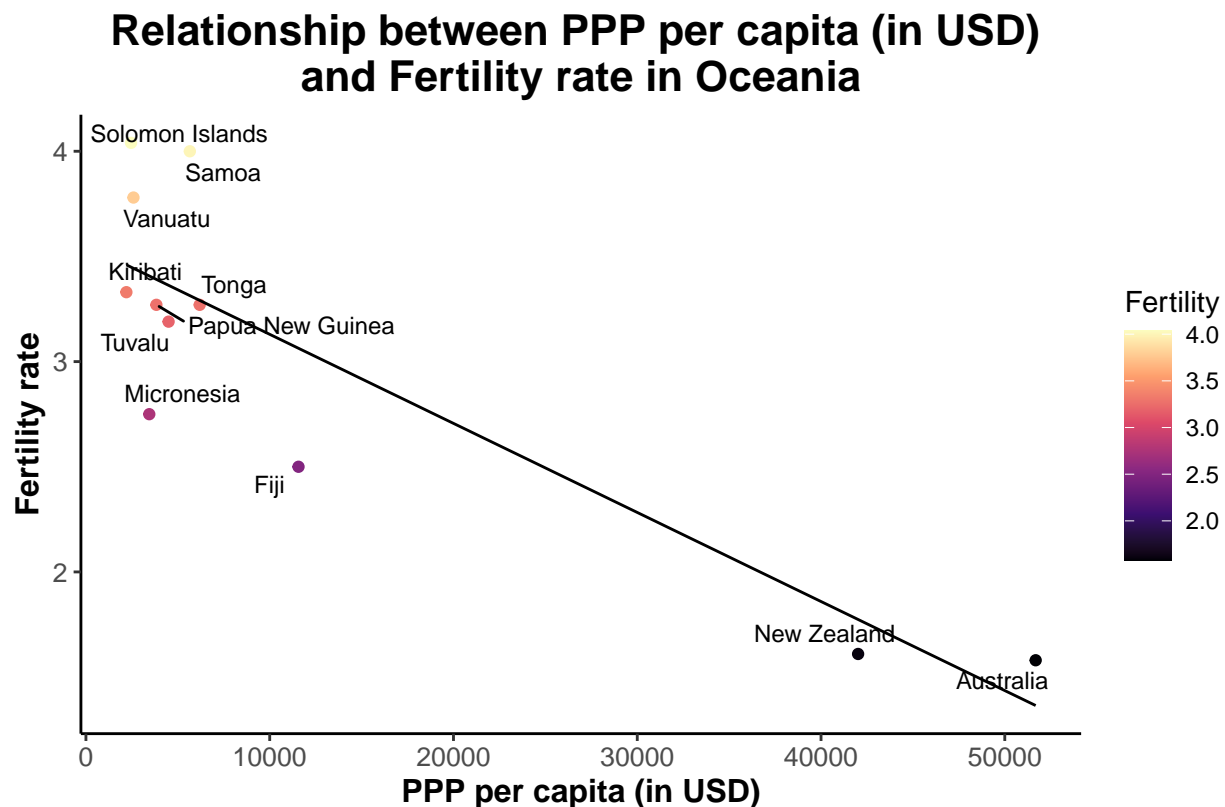
European fertility rate and PPP show a slightly negative correlation.

Oceania

```
# Subset the dataframe to include only Oceania data
df_oceania <- subset(df, Region == "Oceania")

ggplot(df_oceania, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita and Fertility rate in Oceania") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Data Source: Statistics Times and World Population Review

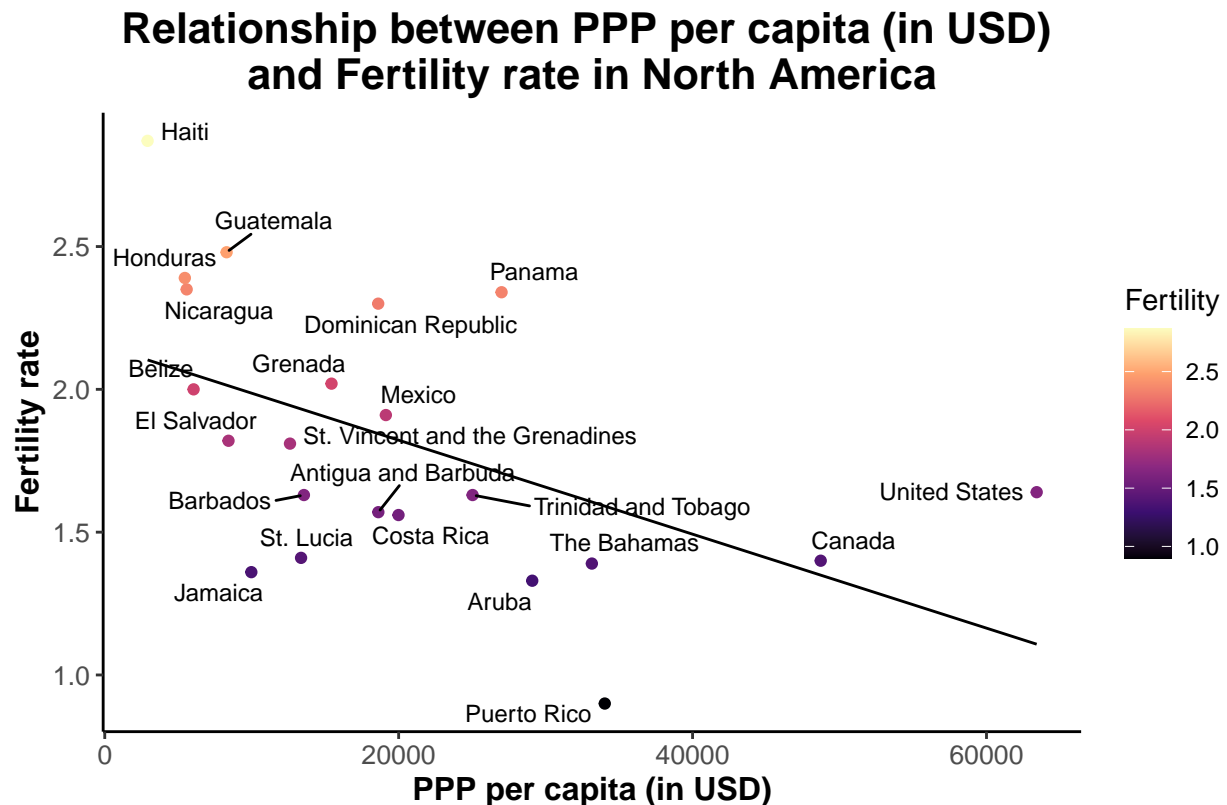
The relationship of the variables is found negative. Although there are two countries, New Zealand and Australia that seem to pulling the best-fit line, the correlation would still be slightly negative looking at the pattern of other countries.

North America

```
# Subset the dataframe to include only North America data
df_northa <- subset(df, Region == "North America")

ggplot(df_northa, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita and Fertility rate in North America") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Data Source: Statistics Times and World Population Review

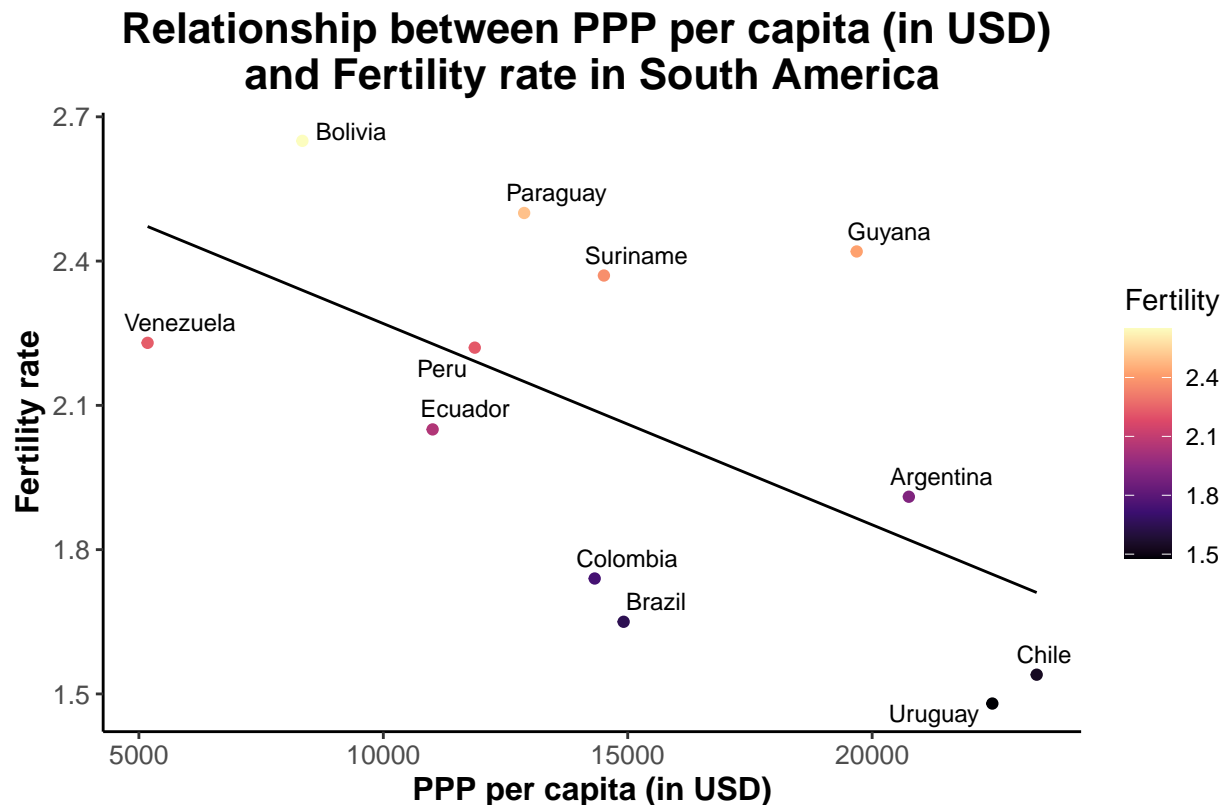
The relationship between fertility rate and PPP of North American countries is found negative.

South America

```
# Subset the dataframe to include only South America data
df_southa <- subset(df, Region == "South America")

ggplot(df_southa, aes(x = PPP, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Fertility rate", title = "Relationship between PPP per capita and Fertility rate in South America") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



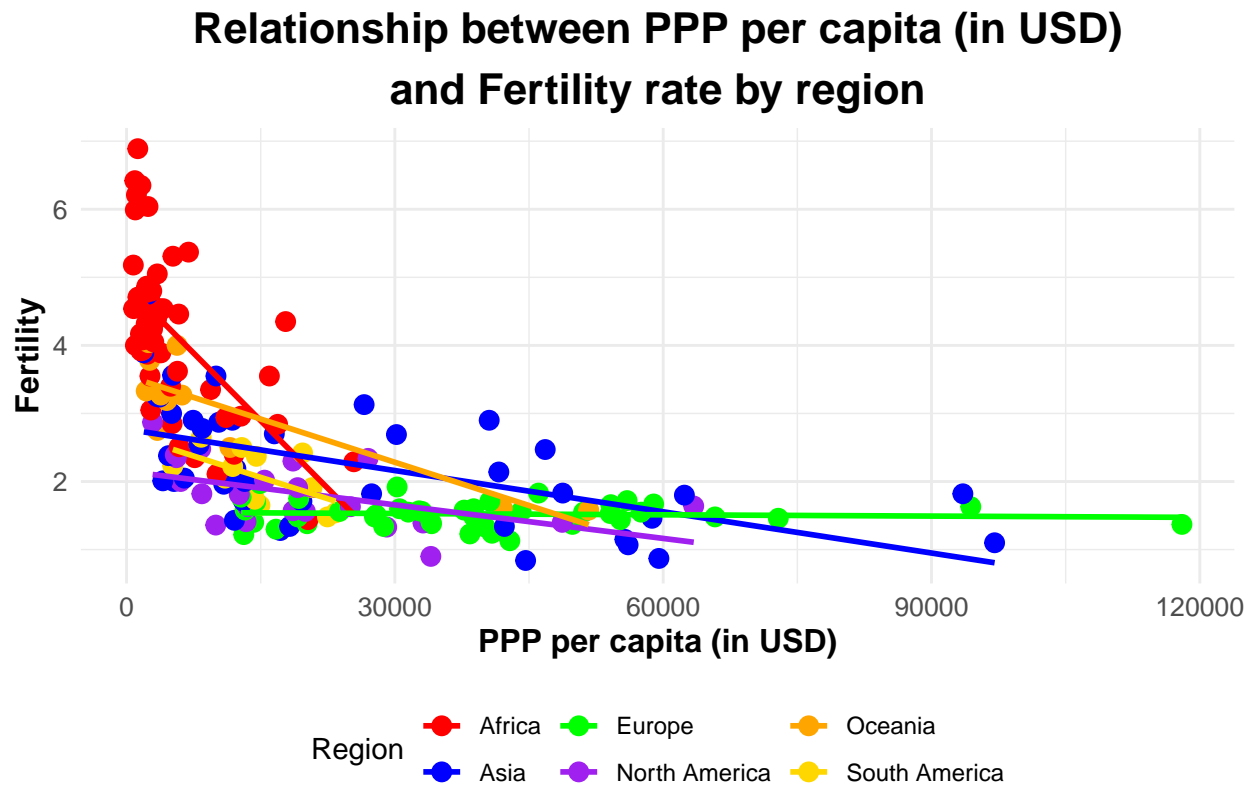
Data Source: Statistics Times and World Population Review

The relationship between two variables found negative in South American countries.

The general relationship regarding regional differences

```
ggplot(df, aes(x = PPP, y = Fertility, color = Region)) +  
  geom_point(size = 3) +  
  labs(x = "PPP per capita (in USD)", y = "Fertility",  
       title = "Relationship between PPP per capita (in USD)\nand Fertility rate by region", caption =  
  scale_color_manual(values = c("red", "blue", "green", "purple", "orange", "#FFD700")) +  
  geom_smooth(method = "lm", se = FALSE, aes(group = Region, color = Region)) +  
  theme_minimal() +  
  theme(  
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5, lineheight = 1.2),  
    axis.title = element_text(size = 12, face = "bold"),  
    axis.text = element_text(size = 10),  
    legend.position = "bottom")
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



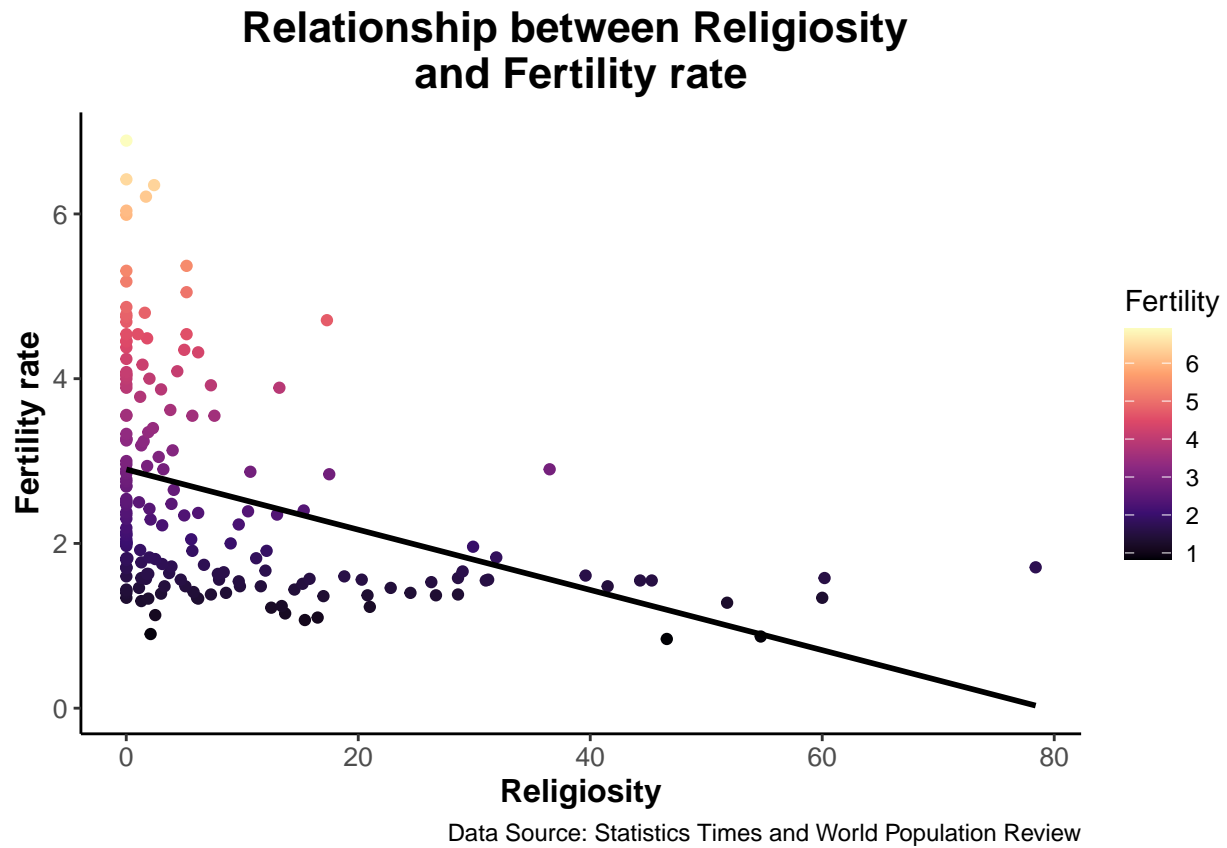
Data Source: Statistics Times and World Population Review

According to the scatter plot above, the six regions all demonstrated negative association between PPP per capita and religiosity. However, the strength of them were not similar. Region of country seems to have an effect on the correlations.

The General relationship between Religiosity and Fertility rate

```
ggplot(df, aes(x = Religiosity, y = Fertility, color = Fertility)) +  
  geom_point() +  
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity and Fertility rate") +  
  scale_color_viridis_c(option = "magma") +  
  geom_smooth(method = "lm", se = FALSE, color = "black") +  
  theme_classic() +  
  theme(  
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),  
    axis.title = element_text(size = 12, face = "bold"),  
    axis.text = element_text(size = 10))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Unlike the initial assumption, more religious countries did not show greater fertility rate. It turned out rather opposite. The hypothesis 2 that religious countries would have more babies, is rejected.

Regional difference in the relationship

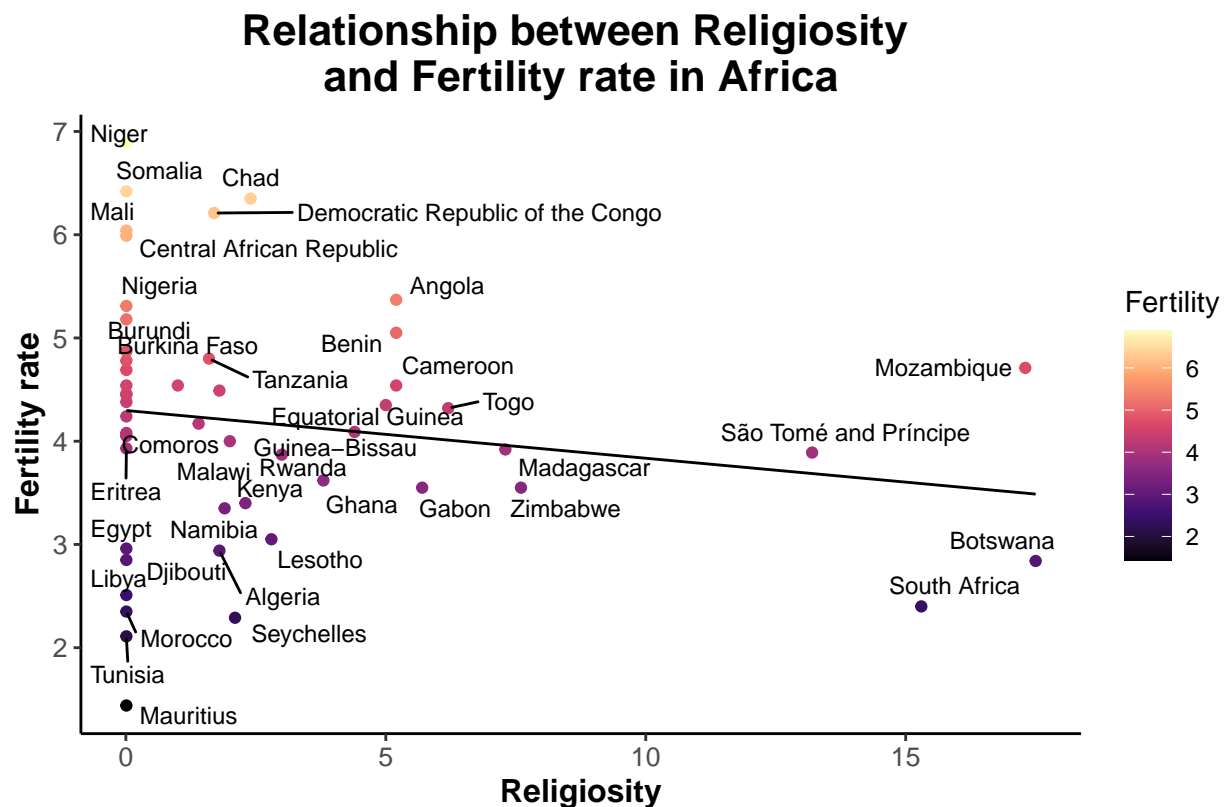
Africa

```
# Subset the dataframe to include only Africa data
df_africa <- subset(df, Region == "Africa")

ggplot(df_africa, aes(x = Religiosity, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity \nand Fertility") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## Warning: ggrepel: 11 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



Data Source: Statistics Times and World Population Review

African fertility rate and religiosity show a negative correlation.

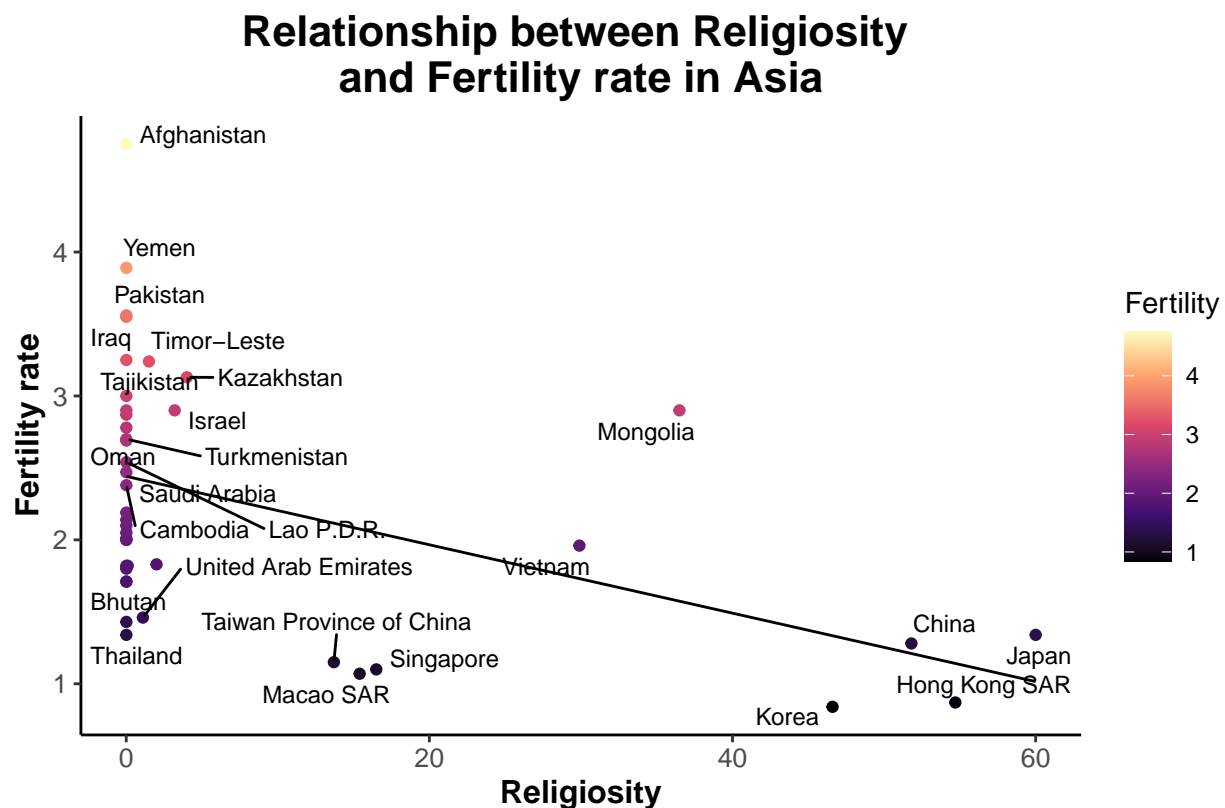
Asia

```
# Subset the dataframe to include only Asia data
df_asia <- subset(df, Region == "Asia")

ggplot(df_asia, aes(x = Religiosity, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity \nand Fertility") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## Warning: ggrepel: 17 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



Data Source: Statistics Times and World Population Review

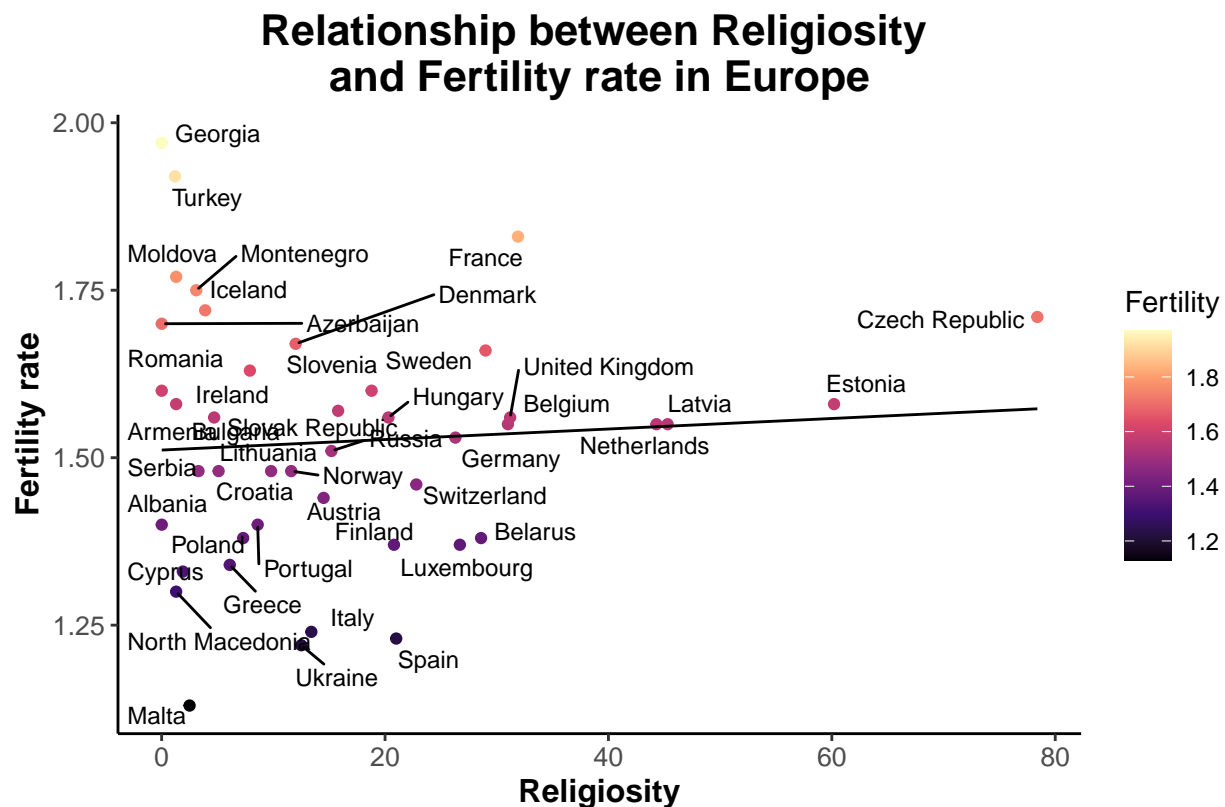
Asian fertility rate and religiosity show a negative correlation.

Europe

```
# Subset the dataframe to include only Europe data
df_europe <- subset(df, Region == "Europe")

ggplot(df_europe, aes(x = Religiosity, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity \nand Fertility") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Data Source: Statistics Times and World Population Review

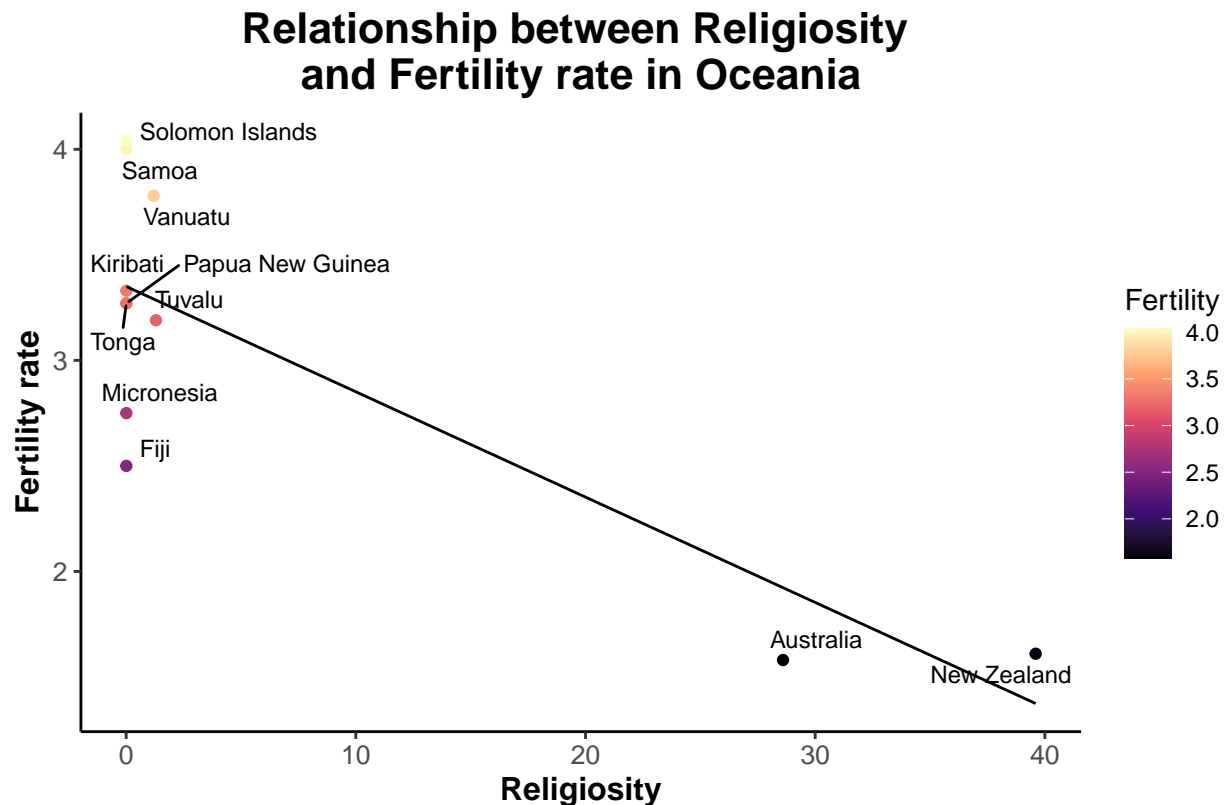
Unlike Africa and Asia, European fertility rate and religiosity had a slightly positive correlation.

Oceania

```
# Subset the dataframe to include only Oceania data
df_oceania <- subset(df, Region == "Oceania")

ggplot(df_oceania, aes(x = Religiosity, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity \nand Fertility") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Data Source: Statistics Times and World Population Review

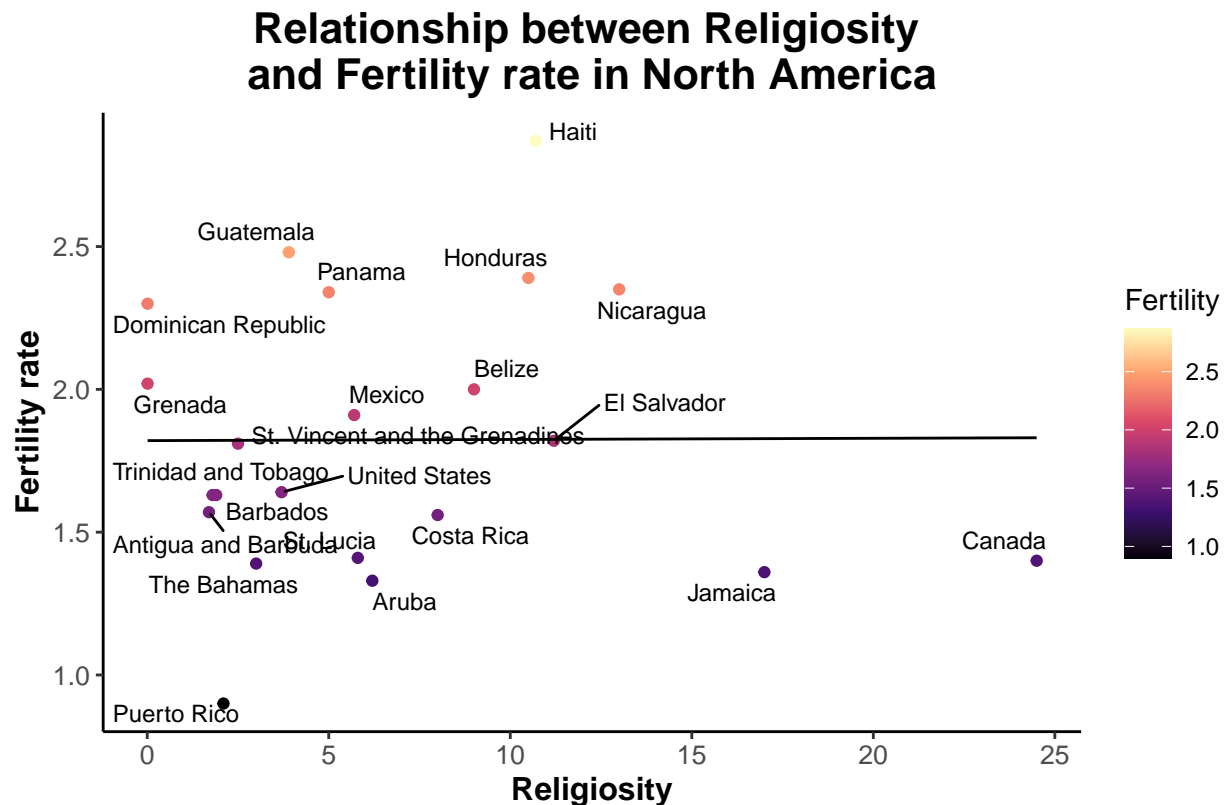
Again, fertility rate and religiosity of Oceanic countries is negatively correlated.

North America

```
# Subset the dataframe to include only North America data
df_northa <- subset(df, Region == "North America")

ggplot(df_northa, aes(x = Religiosity, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity \nand Fertility") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Data Source: Statistics Times and World Population Review

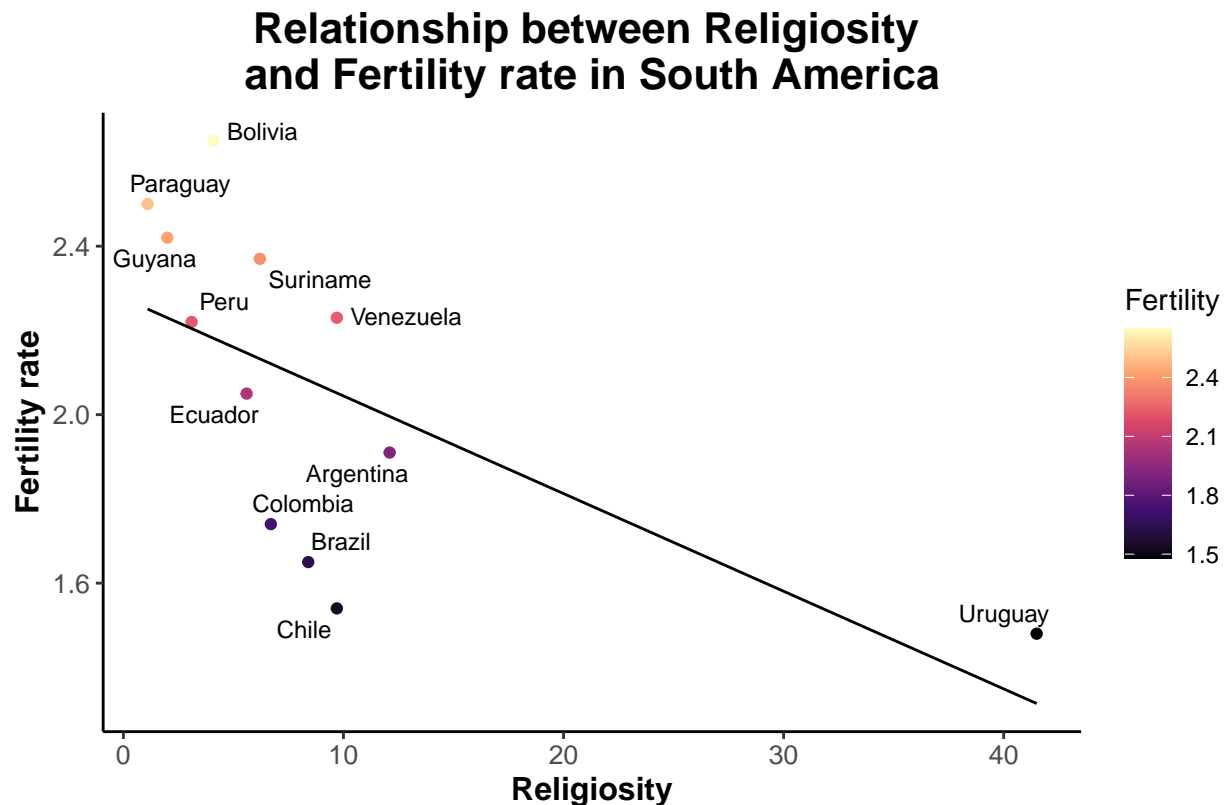
The relationship between fertility rate and religiosity of North American countries is found slightly positive like Europe.

South America

```
# Subset the dataframe to include only South America data
df_southa <- subset(df, Region == "South America")

ggplot(df_southa, aes(x = Religiosity, y = Fertility, color = Fertility)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility rate", title = "Relationship between Religiosity \nand Fertility") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



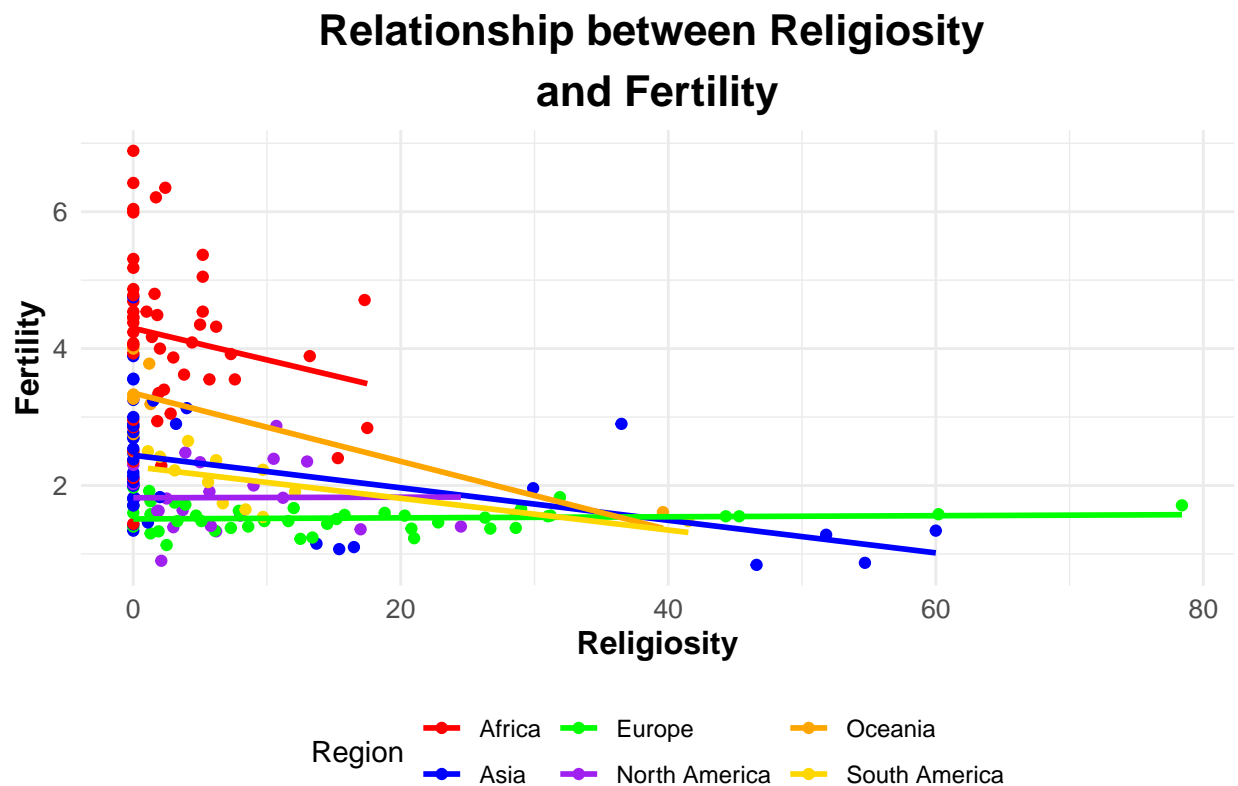
Data Source: Statistics Times and World Population Review

The relationship between the two variables found negative in South American countries.

The general relationship regarding regional differences

```
ggplot(df, aes(x = Religiosity, y = Fertility, color = Region)) +
  geom_point() +
  labs(x = "Religiosity", y = "Fertility", title = "Relationship between Religiosity \nand Fertility",
  scale_color_manual(values = c("red", "blue", "green", "purple", "orange", "#FFD700")) +
  geom_smooth(method = "lm", se = FALSE, aes(group = Region, color = Region)) +
  theme_minimal() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5, lineheight = 1.2),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    legend.position = "bottom")
```

'geom_smooth()' using formula = 'y ~ x'



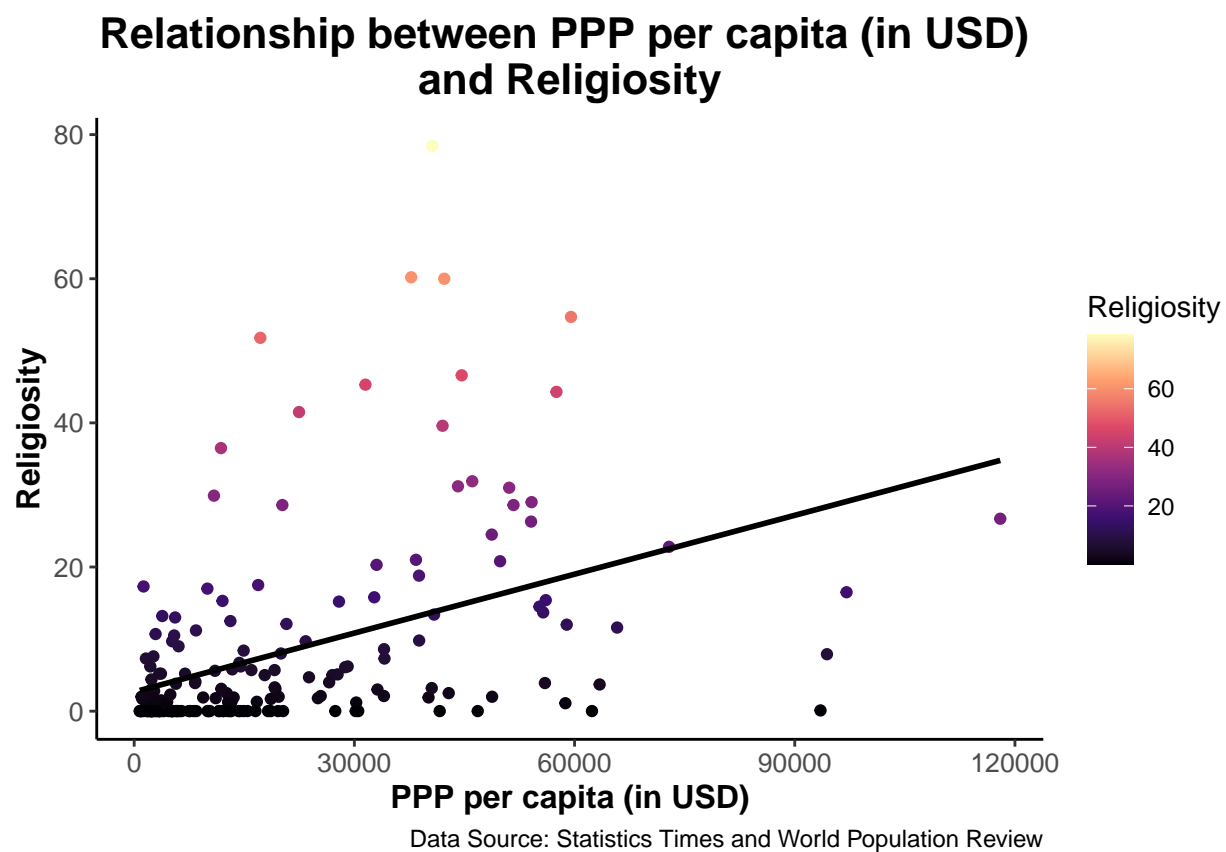
Data Source: Statistics Times and World Population Review

The scatterplots demonstrate more negative-like correlations between religiosity and fertility rate. Two of six regions, Europe and North America showed weak but positive relationship interestingly. Religiosity does not seem to be strong predictor of fertility rate compared to PPP per capita.

The General relationship between PPP per capita and Religiosity

```
ggplot(df, aes(x = PPP, y = Religiosity, color = Religiosity)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (")
  scale_color_viridis_c(option = "magma") +
  geom_smooth(method = "lm", se = FALSE, color = "black") +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



Again, hypothesis 3, expecting a negative relationship between PPP per capita and religiosity, is rejected.

Regional difference in the relationship

Africa

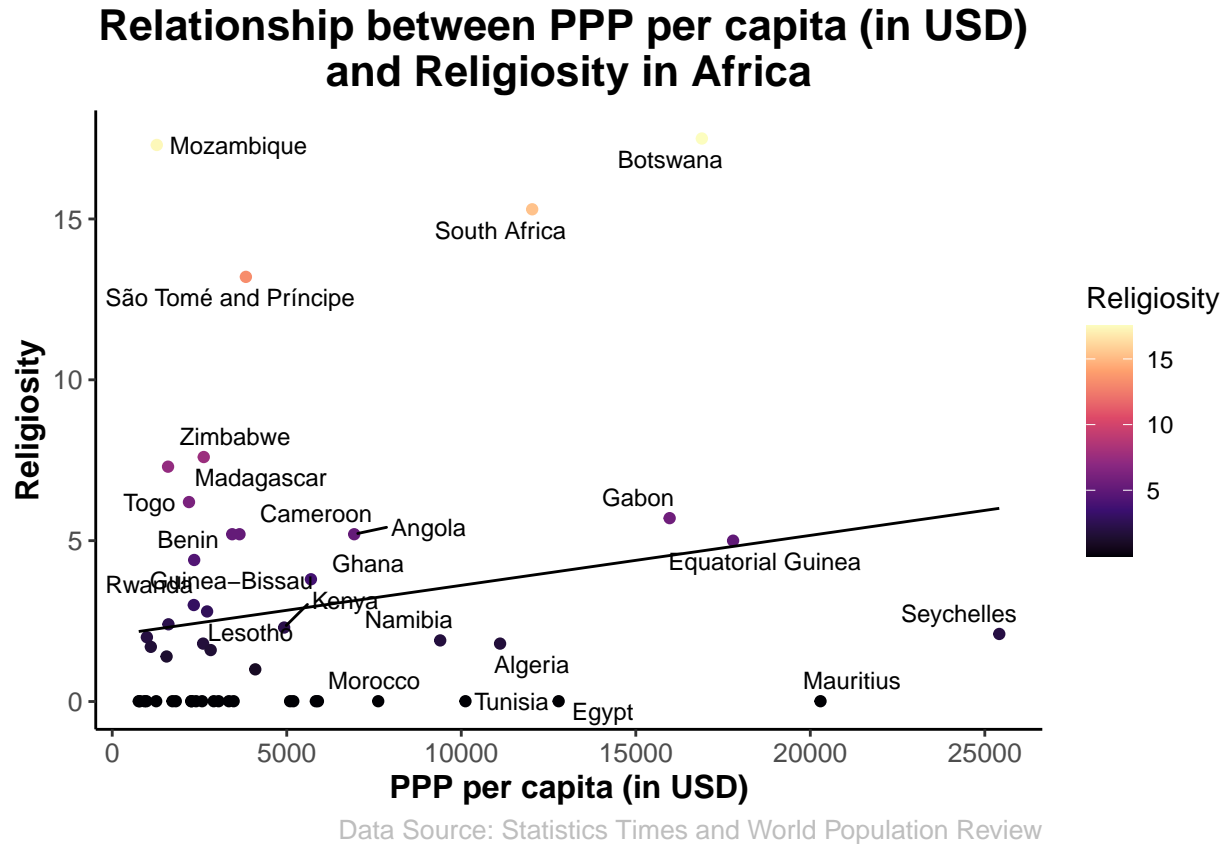
```
# Subset the dataframe to include only Africa data
df_africa <- subset(df, Region == "Africa")

ggplot(df_africa, aes(x = PPP, y = Religiosity, color = Religiosity)) +
```

```
geom_point() +
labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (in USD) and Religiosity in Africa") +
scale_color_viridis_c(option = "magma") +
geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
theme_classic() +
theme(
  plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
  axis.title = element_text(size = 12, face = "bold"),
  axis.text = element_text(size = 10),
  plot.caption = element_text(size = 10, color = "gray")
)
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## Warning: ggrepel: 26 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



African fertility rate and religiosity show a positive correlation.

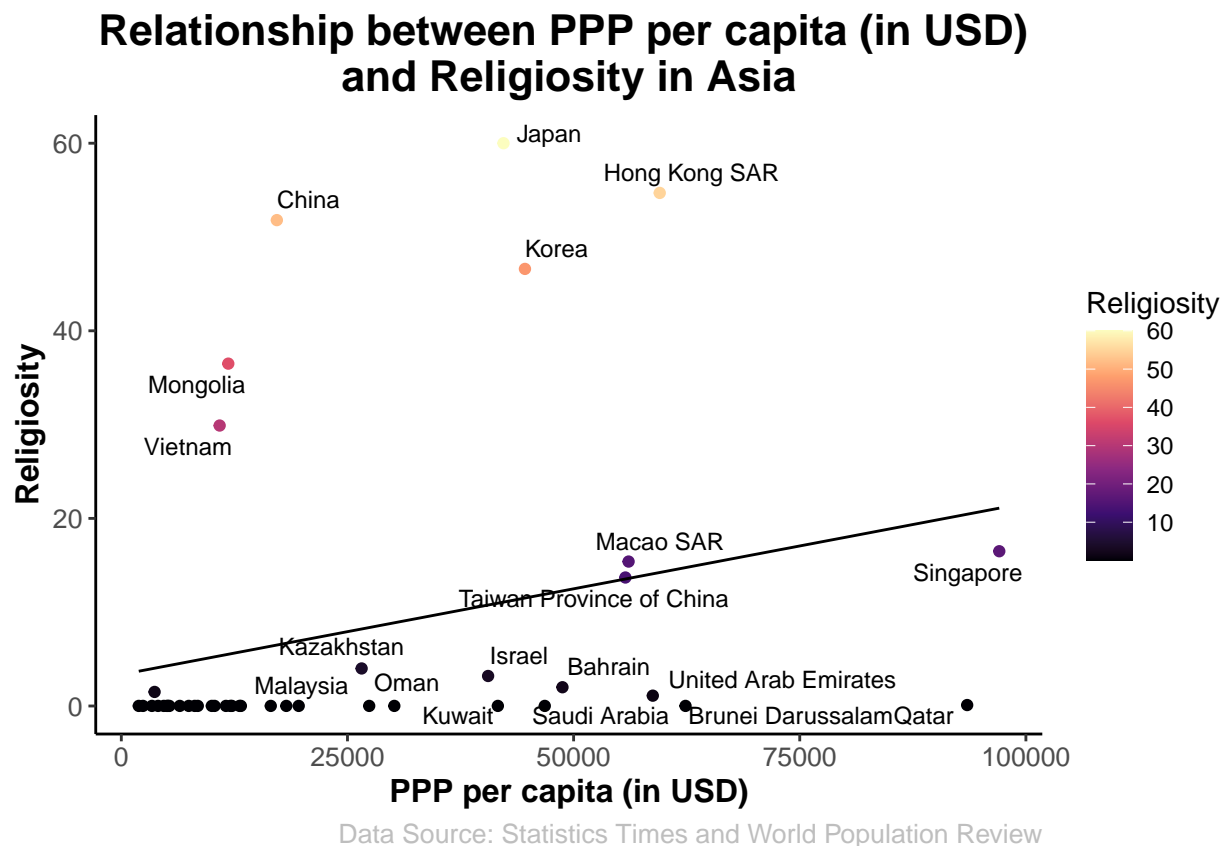
Asia

```
# Subset the dataframe to include only Asia data
df_asia <- subset(df, Region == "Asia")
```

```
ggplot(df_asia, aes(x = PPP, y = Religiosity, color = Religiosity)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (in USD) and Religiosity in Asia") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## Warning: ggrepel: 23 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



The relationship was observed to be positive for Asian countries. Although it seems there are number of outliers only looking at Asian countries, it would still show positive relationship even if they are removed.

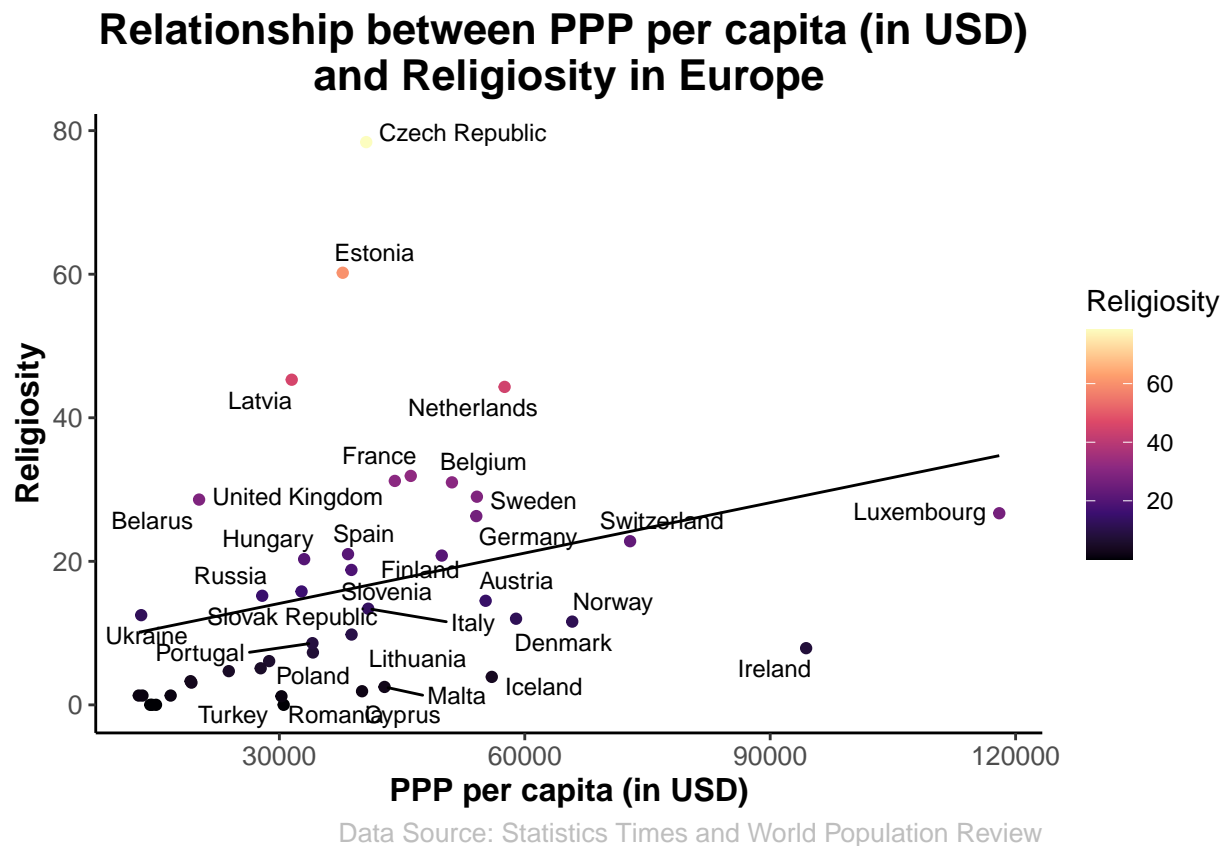
Europe

```
# Subset the dataframe to include only Europe data
df_europe <- subset(df, Region == "Europe")

ggplot(df_europe, aes(x = PPP, y = Religiosity, color = Religiosity)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (in USD) and Religiosity in Europe") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )

## 'geom_smooth()' using formula = 'y ~ x'

## Warning: ggrepel: 11 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



European PPP and religiosity show a positive correlation.

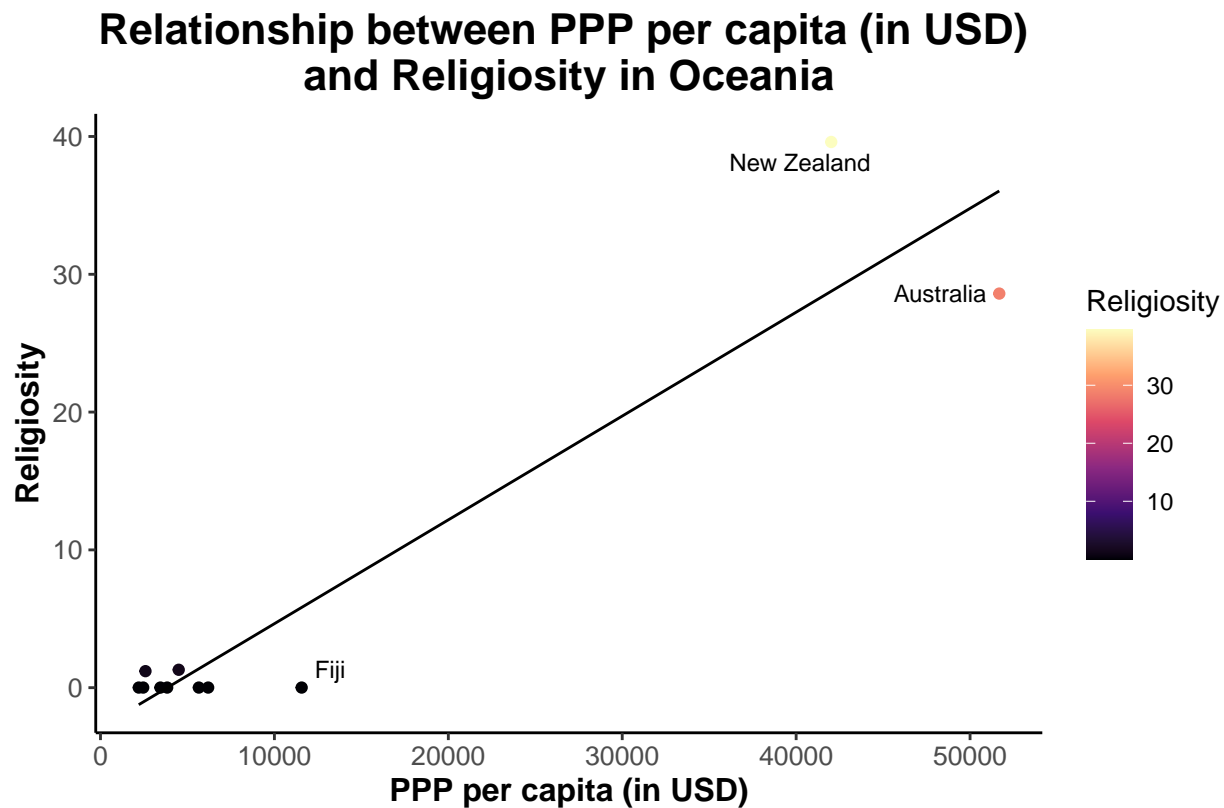
Oceania

```
# Subset the dataframe to include only Oceania data
df_oceania <- subset(df, Region == "Oceania")

ggplot(df_oceania, aes(x = PPP, y = Religiosity, color = Religiosity)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (in USD) and Religiosity in Oceania") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray"))

## 'geom_smooth()' using formula = 'y ~ x'

## Warning: ggrepel: 8 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



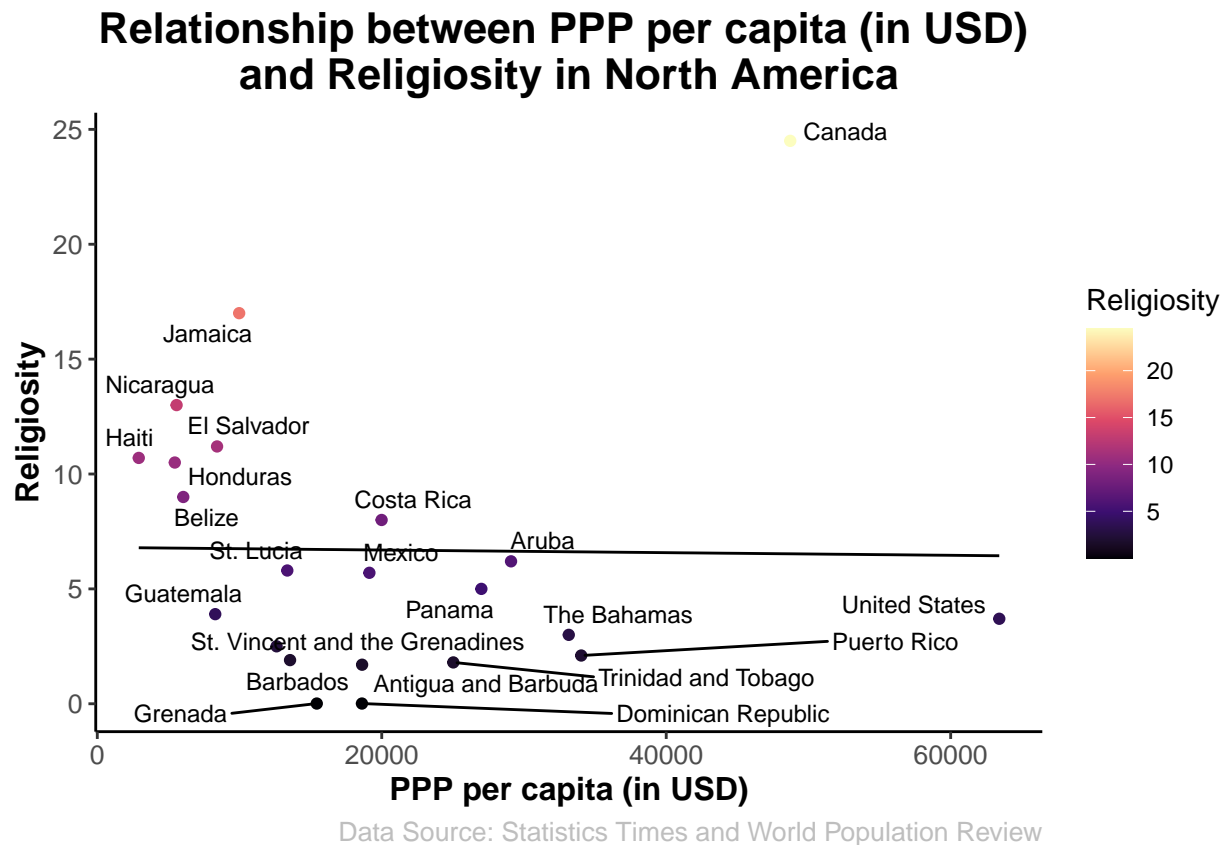
Again, PPP and religiosity of Oceanic countries is positively correlated. However, two countries, affecting the relationship as they have considerably greater PPP and religiosity.

North America

```
# Subset the dataframe to include only North America data
df_northa <- subset(df, Region == "North America")

ggplot(df_northa, aes(x = PPP, y = Religiosity, color = Religiosity)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (in USD) and Religiosity in North America") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



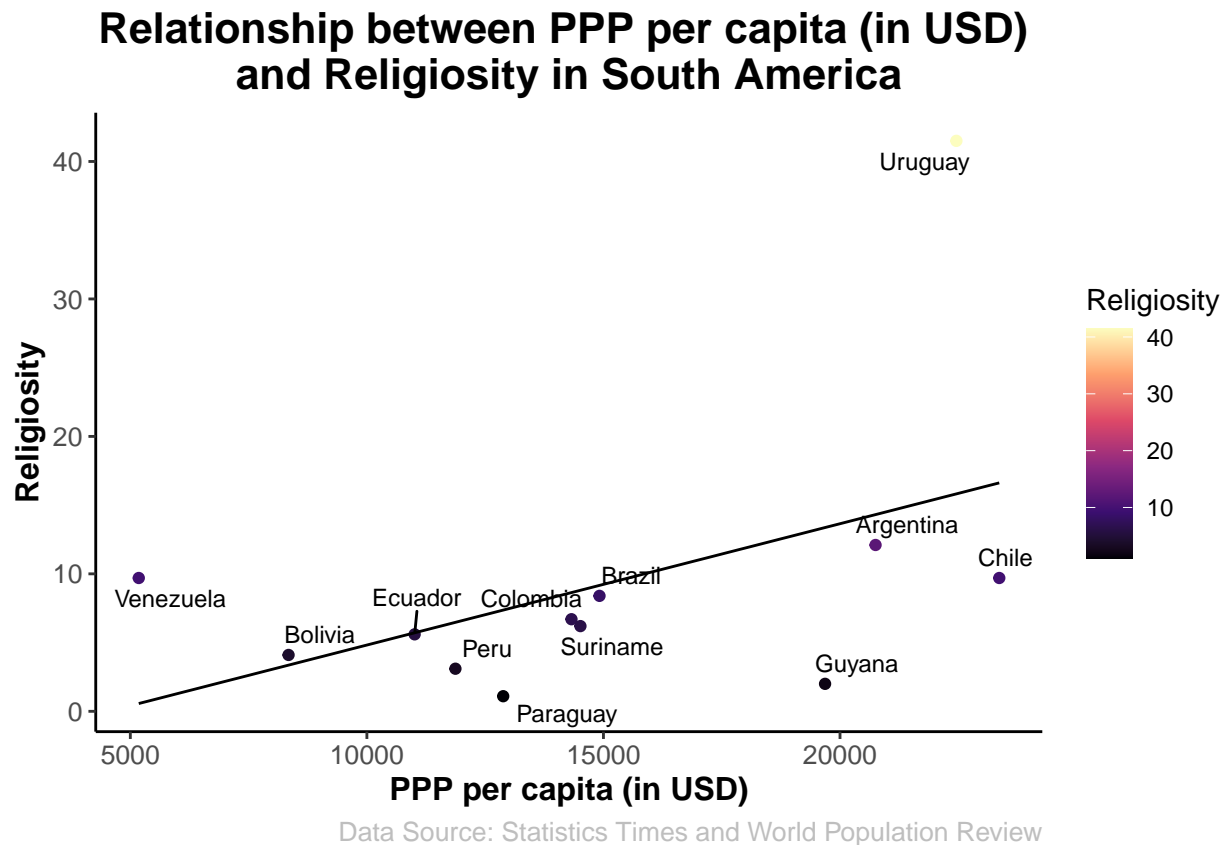
North America is the first sample that showed negative relationship. The strength might look small. However, two respective places seem to have an impact on the strength as outliers. If removed, the relationship would look more negative.

South America

```
# Subset the dataframe to include only South America data
df_southa <- subset(df, Region == "South America")

ggplot(df_southa, aes(x = PPP, y = Religiosity, color = Religiosity)) +
  geom_point() +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between PPP per capita (in USD) and Religiosity in South America") +
  scale_color_viridis_c(option = "magma") +
  geom_text_repel(aes(label = ifelse(!is.na(Country), Country, "")), size = 3, color = "black", show.legend = FALSE) +
  geom_smooth(method = "lm", se = FALSE, color = "black", linewidth = 0.5) +
  theme_classic() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    plot.caption = element_text(size = 10, color = "gray")
  )
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

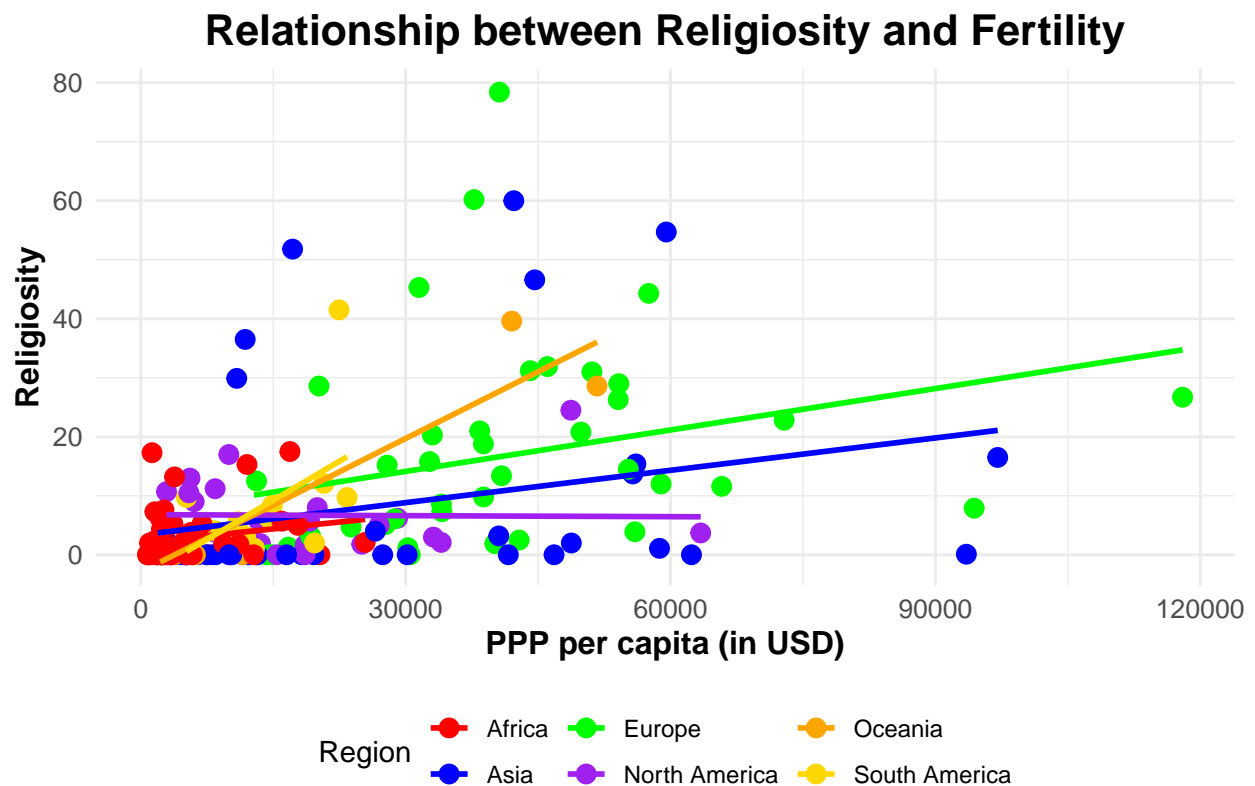


Unlike North America, PPP and religiosity of South American countries is positively correlated.

The general relationship regarding regional differences

```
ggplot(df, aes(x = PPP, y = Religiosity, color = Region)) +
  geom_point(size = 3) +
  labs(x = "PPP per capita (in USD)", y = "Religiosity", title = "Relationship between Religiosity and PPP",
    scale_color_manual(values = c("red", "blue", "green", "purple", "orange", "#FFD700")) +
  geom_smooth(method = "lm", se = FALSE, aes(group = Region, color = Region)) +
  theme_minimal() +
  theme(
    plot.title = element_text(size = 16, face = "bold", hjust = 0.5, lineheight = 1.2),
    axis.title = element_text(size = 12, face = "bold"),
    axis.text = element_text(size = 10),
    legend.position = "bottom")
```

'geom_smooth()' using formula = 'y ~ x'



Data Source: Statistics Times and World Population Review

Again, hypothesis 3, relatively financially comfortable countries would show low religiosity, is rejected except for North America. Although the spread of plots is expanding, the relationship between PPP and religiosity looks quite positive. The region, again, seemed to have effect on the differences in the strength of correlation. The interesting finding is that North America is the only group that showed different patterns compared to other groups. The reason for such phenomenon is not conclusive.

Additional analysis

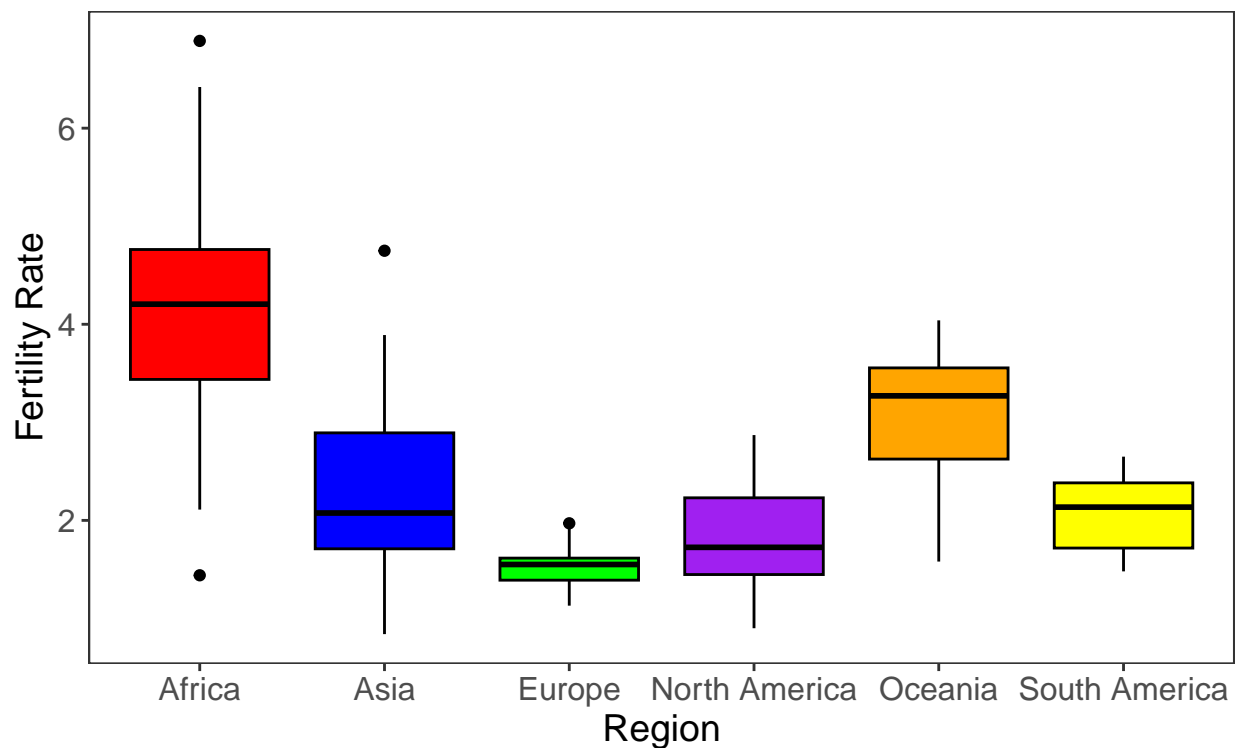
Box plot 1: Fertility difference by Regions

```
# Subset the data to only include the columns we need
df2 <- subset(df, select = c("Region", "Fertility"))

# Create the box plot
# Define the color palette for each region
region_colors <- c("Africa" = "red", "Asia" = "blue", "Europe" = "green", "North America" = "purple", "Oceania" = "orange", "South America" = "yellow")

# Create the box plot with customizations
ggplot(df2, aes(x = Region, y = Fertility, fill = Region)) +
  geom_boxplot(color = "black") +
  labs(x = "Region", y = "Fertility Rate",
       title = "Box Plot of Fertility Rate by Region", caption = "Data Source: Statistics Times and World Population Review") +
  theme_bw() +
  theme(plot.title = element_text(size = 18, face = "bold"),
        axis.text = element_text(size = 12),
        axis.title = element_text(size = 14),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        legend.position = "none") +
  scale_fill_manual(values = region_colors) +
  scale_color_manual(values = region_colors)
```

Box Plot of Fertility Rate by Region



Data Source: Statistics Times and World Population Review

When compared the fertility rates only by different regions, African countries showed considerably greater mean rate than the others. And Europe showed least mean rate and IQR.

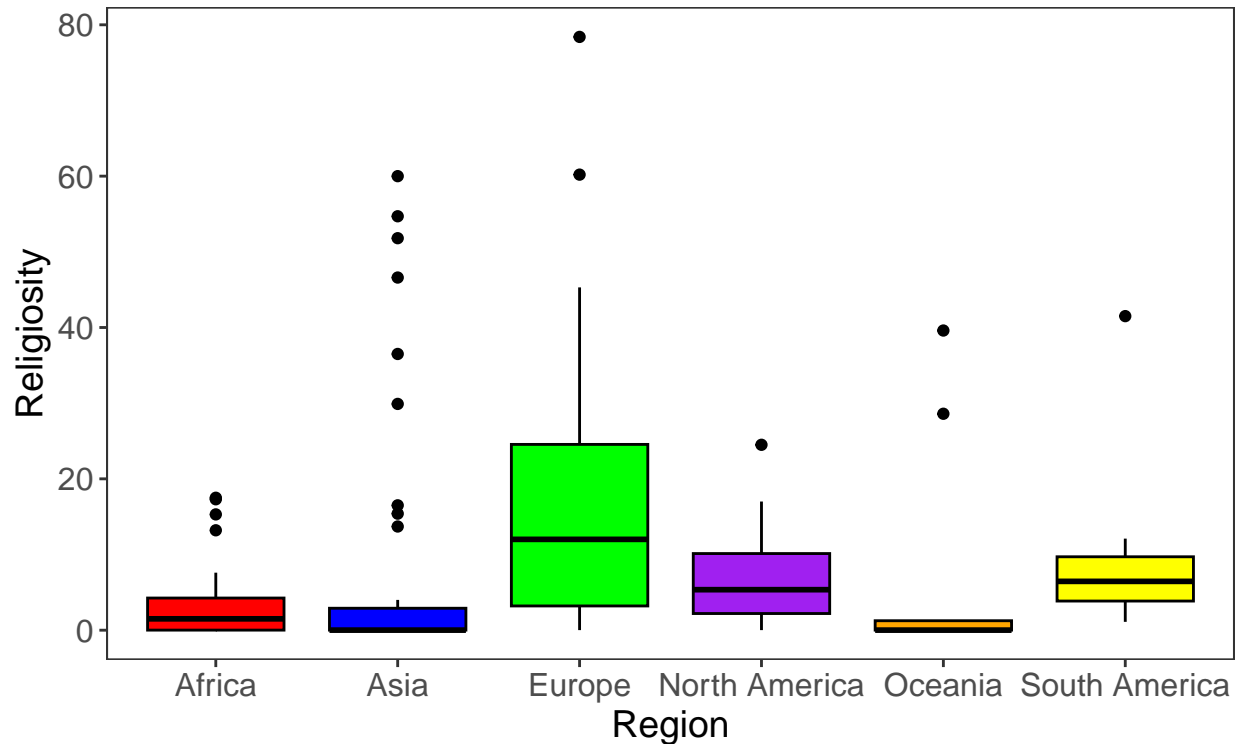
Box plot 2: Religiosity difference by Region

```
# Subset the data to only include the columns we need
df3 <- subset(df, select = c("Region", "Religiosity"))

# Create the box plot
# Define the color palette for each region
region_colors <- c("Africa" = "red", "Asia" = "blue", "Europe" = "green", "North America" = "purple", "South America" = "orange")

# Create the box plot with customizations
ggplot(df3, aes(x = Region, y = Religiosity, fill = Region)) +
  geom_boxplot(color = "black") +
  labs(x = "Region", y = "Religiosity",
       title = "Box Plot of Religiosity by Region", caption = "Data Source: Statistics Times and World Bank") +
  theme_bw() +
  theme(plot.title = element_text(size = 18, face = "bold"),
        axis.text = element_text(size = 12),
        axis.title = element_text(size = 14),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        legend.position = "none") +
  scale_fill_manual(values = region_colors) +
  scale_color_manual(values = region_colors)
```

Box Plot of Religiosity by Region



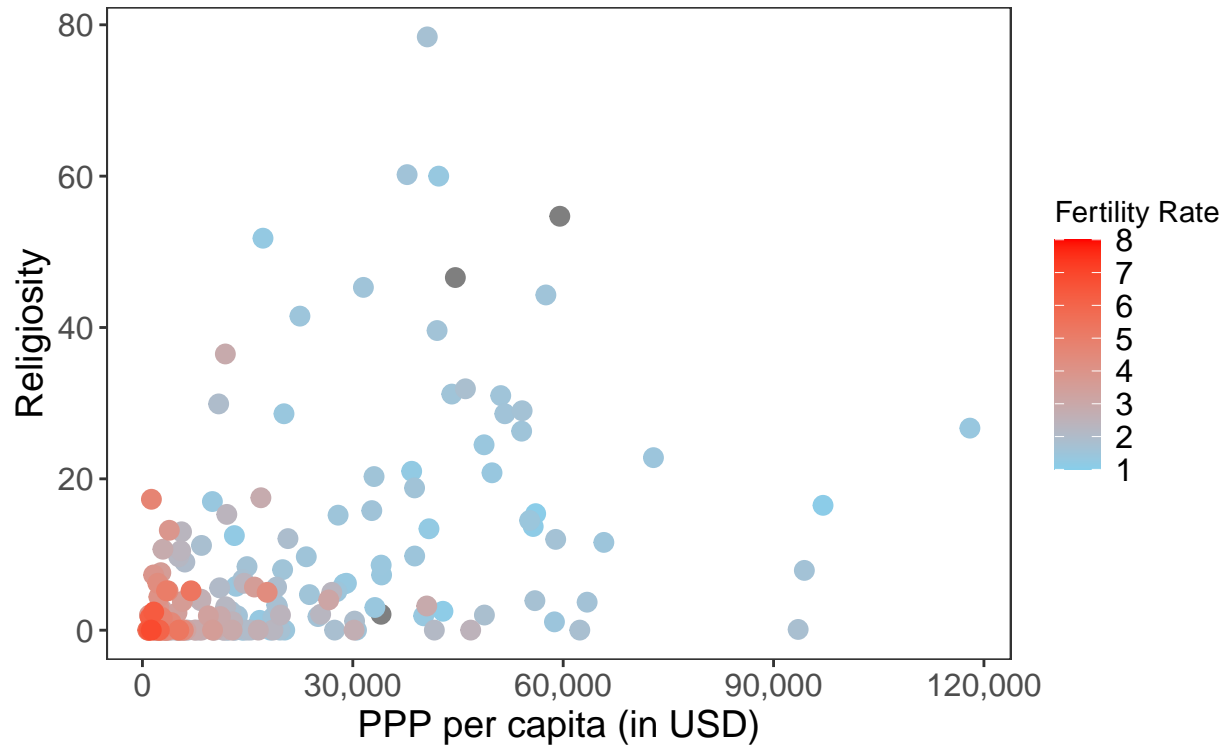
Data Source: Statistics Times and World Population Review

Another box plot was generated to compare the religiosity by region. Although Asia displayed number of outliers, the five regions except Europe showed similar level of religiosity. Europe showed greatest IQR. Yet, Europe showed the least fertility rate considering its greater average religiosity compared to the others.

Additional scatter plot 1 : Relationship between PPP per capita, Religiosity and Fertlity rate

```
ggplot(df, aes(x = PPP, y = Religiosity, color = Fertility)) +
  geom_point(size = 3) +
  scale_x_continuous(labels = scales::comma, name = "PPP per capita (in USD)") +
  scale_y_continuous(name = "Religiosity") +
  scale_color_gradient(low = "skyblue", high = "red", limits = c(1, 8),
    breaks = c(1, 2, 3, 4, 5, 6, 7, 8),
    guide = guide_colorbar(title = "Fertility Rate")) +
  labs(title = "Relationship between PPP, Religiosity, and Fertility",
    caption = "Data source: Statistics Times and World Population Review") +
  theme_bw() +
  theme(plot.title = element_text(size = 18, face = "bold"),
    axis.title = element_text(size = 14),
    axis.text = element_text(size = 12),
    legend.text = element_text(size = 12),
    legend.position = "right",
    panel.grid.major = element_blank(),
    panel.grid.minor = element_blank())
```

Relationship between PPP, Religiosity, and Fertility



Data source: Statistics Times and World Population Review

This plot clearly shows that there is a dense cluster at low religiosity and low PPP. Both GDP and PPP are strongly associated with education level and access to health medical services. The scatter plots suggest an environment that would provide less SES and no strong belief to organised religions is more likely to result greater fertility rate.

Counclusion

The first hypothesis, PPP per capita would be negatively correlated with fertility rate was not denied. However, the other two hypotheses, assumed positive relationship between religiosity and fertility rate and negative one between PPP per capita and religiosity, were rejected. Interestingly, countries with greater PPP per capita rather demonstrated deeper religiosity. Although possible effect of region as a covariate on these relationship was questioned, the effect seemed not big enough to change the direction of correlations except the strenght of association. However, there was interesting pattern found in North America as they showed more accepting those two rejected hypotheses. Further research is recommended to find out what is causing this different findings from the past research.

References

Least Religious Countries 2023. (2023). World Population Review. <https://worldpopulationreview.com/country-rankings/least-religious-countries>

List of Countries by GDP (PPP) per capita. (2021, Jun 21). Statistics Times.<https://statisticstimes.com/economy/countries-by-gdp-capita-ppp.php>

Rowthorn, R. (2011). Religion, fertility and genes: a dual inheritance model. *Proceedings of the Royal Society B: Biological Sciences*, 278(1717), 2519-2527.

Storm, I. (2017). Does economic insecurity predict religiosity? Evidence from the European social survey 2002–2014. *Sociology of Religion*, 78(2), 146-172.

Vandenbroucke, G. (2016). The link between fertility and income. Federal Reserve Bank of, St Louis, USA. Dec, 13.