

## Question 3 and 4

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### Data

For referencing, please refer to the referencing Felix has completed, as I have taken his code for importing and cleaning the data, and have added my own code

```
# !!! DON'T FORGET TO CHANGE THE WORKING DIRECTORY TO YOUR OWN DIRECTORY !!!
setwd("/Users/charliedicker/Downloads/Big_Issue-main/Data")
knitr::opts_chunk$set(echo = TRUE)
library(ggplot2)
library(tidyverse)
library(readr)
library(readxl)
library(patchwork)

# Importing Kenya Inequality
Ken_W_Ineq <- read_delim("Kenya_Wealth_Inequality.csv",
  delim = ";", escape_double = FALSE, col_names = FALSE,
  trim_ws = TRUE, skip = 1
) # This is why I would rather use xls file

Ken_I_Ineq <- read_delim("Kenya_Income_Inequality.csv",
  delim = ";", escape_double = FALSE, col_names = FALSE,
  trim_ws = TRUE, skip = 1
)

# Importing Kenya Income Growth
Ken_I_Growth <- read_excel("Bang_Income.xlsx")

# Importing Bangladesh Inequality
Bang_W_Ineq <- read_excel("Bang_Wealth_Inequality.xlsx", col_names = FALSE)
Bang_I_Ineq <- read_excel("Bang_Income_Inequality.xlsx", col_names = FALSE)

# Importing Bangladesh Income Growth
Bang_I_Growth <- read_excel("Kenya_Income.xlsx")

# Importing Kenya Trade
Trade_df <- read_excel("Bang_and_Ken_Trade.xlsx", col_names = FALSE)
```

```
# Importing Bangladesh Trade
```

```
# Make a function for cleaning data sets
```

```
clean_data_inequality <- function(x) {  
  colnames(x) <- c("Country", "Indicator", "Percentile", "Year", "Value")  
  # Columns of Inequality data sets are all in this order, check when use for others  
  x <- x %>%  
    pivot_wider(  
      names_from = Percentile,  
      values_from = Value  
    ) %>%  
    filter(!if_all(c(pall, p0p50, p50p90, p90p100, p99p100), is.na)) %>%  
    # filter out those rows where all the values are NA  
    select(Country, Year, pall, p0p50, p50p90, p90p100, p99p100) %>%  
    # To ensure the columns are in correct order and delete indicator column  
    group_by(Country, Year) %>%  
    summarise(  
      across(  
        c(pall, p0p50, p50p90, p90p100, p99p100),  
        ~ first(na.omit(.))  
      ),  
      .groups = "drop"  
    )  
    # Note that previously we have 5 lines for a single year, and each  
    # line only shows a single indicator. By doing this, we combine the data together.  
    colnames(x) <- c(  
      "Country", "Year", "Gini_Coeff", "Share_Bottom50",  
      "Share_Middle40", "Share_Top10", "Share_Top1"  
    )  
    return(x)  
}
```

```
#I added this function
```

```
clean_data_growth <- function(x) {  
  colnames(x) <- c("Year", "National Income")  
  x <- x %>% mutate("Growth" = ((`National Income` - lag(`National Income`))/ lag(`National Income`) *  
  filter(!is.na(Growth))  
  
  return(x)  
}
```

```
# I added this function to clean trade data
```

```
clean_trade <- function(x) {  
  colnames(x) <- c("Percentile", "Year", "Bang_Trade", "Ken_Trade")  
  x$Percentile <- NULL  
  x <- x %>% filter(!is.na(Bang_Trade) & !is.na(Ken_Trade))  
  x$Bang_Trade <- as.numeric(x$Bang_Trade)  
  x$Ken_Trade <- as.numeric(x$Ken_Trade)  
  
  return(x)  
}
```

```

Bang_I_Ineq_wider <- clean_data_inequality(Bang_I_Ineq)
Bang_W_Ineq_wider <- clean_data_inequality(Bang_W_Ineq)

#I added this
Bang_I_Growth_cleaned <- clean_data_growth(Bang_I_Growth)

Ken_I_Ineq_wider <- clean_data_inequality(Ken_I_Ineq)
Ken_W_Ineq_wider <- clean_data_inequality(Ken_W_Ineq)

# I added this
Ken_I_Growth_cleaned <- clean_data_growth(Ken_I_Growth)

# I added this
Trade_cleaned <- clean_trade(Trade_df)

## Warning in clean_trade(Trade_df): NAs introduced by coercion
## Warning in clean_trade(Trade_df): NAs introduced by coercion

Trade_cleaned <- Trade_cleaned[-1, ]

longer_format <- function(x) {
  x %>% pivot_longer(
    cols = c(
      Gini_Coeff, Share_Bottom50, Share_Middle40,
      Share_Top10, Share_Top1
    ),
    names_to = "Indicator",
    values_to = "Value"
  )
}

Bang_I_Ineq_longer <- longer_format(Bang_I_Ineq_wider)
Bang_W_Ineq_longer <- longer_format(Bang_W_Ineq_wider)
Ken_I_Ineq_longer <- longer_format(Ken_I_Ineq_wider)
Ken_W_Ineq_longer <- longer_format(Ken_W_Ineq_wider)

```

## Question 3

First, we need to join inequality and growth data. I have written a quick function, using `inner_join` in order to remove any years with only 1 value

```

# Function to join Inequality and Growth data
to_join <- function(x, y) {
  x <- x %>%
    inner_join(y, by = "Year")
  return(x)
}

```

I then wrote 3 functions, to calculate correlation between the Gini Coefficient and National Income Growth. I used Pearson Correlation Coefficient with and without outliers, and the Spearman Correlation Coefficient with outliers.

I cleaned the data and got rid of outliers using the method of anything which is less than the 1st quartile - (1.5 x the interquartile range), or greater than the 3rd quartile + (1.5 x the interquartile range). This is a simple and common way to remove outliers.

GDP growth often contains extreme but economically relevant fluctuations, while Gini coefficients vary within a narrower range (between 0 and 1). Therefore, I removed outliers only from the growth variable to provide a robustness check, but retain inequality values in full. This approach prevents the loss of important economic information while allowing me to assess whether extreme growth shocks drive the correlation.

```
# Function calculates Pearson Correlation Coefficient (with outliers)
calc_correlation <- function(x) {
  return(cor(x$Gini_Coeff, x$Growth))
}

# Function calculates Pearson Correlation Coefficient (without outliers) (I used IQR)
calc_correlation_noOutliers <- function(x) {
  Q1 <- quantile(x$Growth, 0.25, na.rm = TRUE)
  Q3 <- quantile(x$Growth, 0.75, na.rm = TRUE)
  IQR <- Q3 - Q1

  y <- x %>%
    filter(Growth > Q1 - 1.5*IQR & Growth < Q3 + 1.5*IQR)

  return(cor(y$Gini_Coeff, y$Growth))
}

# Function calculates Spearman Correlation Coefficient (with outliers)
calc_correlation_spearman <- function(x){
  return(cor(x$Gini_Coeff, x$Growth, method = "spearman"))
}
```

I then created 2 functions. One to create a graph with the outliers, and one to create a graph without outliers. This allows me to view if outliers are driving correlation, or vice versa, if there is a higher correlations without volatile fluctuations in national income growth.

```
# Function to create the graphs (with outliers)
graph <- function(x, y) {
  return(ggplot(x, aes(x = Growth, y = Gini_Coeff)) +
    geom_point(color = "#1f77b4", size = 1, alpha = 0.7) + # points: muted blue
    geom_smooth(method = "lm", color = "#333333", size = 0.5) + # line: dark gray
    theme_minimal(base_size = 6) + # clean minimal theme
    theme(
      panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
      panel.grid.major = element_line(color = "#dcdcdc"), # soft grid lines
      panel.grid.minor = element_line(color = "#eaeaea"),
      axis.title = element_text(color = "#333333", face = "bold"),
      axis.text = element_text(color = "#333333"),
      plot.title.position = "plot",
      plot.title = element_text(
        lineheight = 1.1,
        margin = margin(b = 10, t = 5), # spacing above/below
      )
    )
  )
}
```

```

    hjust = 0.5 )) +
  labs(
    title = y,
    x = "Growth (%)",
    y = "Gini Coefficient"
  )
)
}

# Function to create the graphs (without outliers)
graph_noOutliers <- function(x, y) {
  Q1 <- quantile(x$Growth, 0.25, na.rm = TRUE)
  Q3 <- quantile(x$Growth, 0.75, na.rm = TRUE)
  IQR <- Q3 - Q1

  z <- x %>%
    filter(Growth > Q1 - 1.5*IQR & Growth < Q3 + 1.5*IQR)

  return(ggplot(z, aes(x = Growth, y = Gini_Coeff)) +
    geom_point(color = "#1f77b4", size = 1, alpha = 0.7) +      # points: muted blue
    geom_smooth(method = "lm", color = "#333333", size = 0.5) + # line: dark gray
    theme_minimal(base_size = 6) +                             # clean minimal theme
    theme(
      panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
      panel.grid.major = element_line(color = "#dcdcdc"),           # soft grid lines
      panel.grid.minor = element_line(color = "#eaeaea"),
      axis.title = element_text(color = "#333333", face = "bold"),
      axis.text = element_text(color = "#333333"),
      plot.title.position = "plot",
      plot.title = element_text(
        lineheight = 1.1,
        margin = margin(b = 10, t = 5),      # spacing above/below
        hjust = 0.5 )) +
    labs(
      title = stringr::str_wrap(y, width = 40),
      x = "Growth (%)",
      y = "Gini Coefficient"
    ))
}

```

This next section is just passing data into the functions, and getting returned values or graphs.

```

# Joining both inequality data and income growth by the year
Bang_I_Ineq_and_Growth <- to_join(Bang_I_Growth_cleaned, Bang_I_Ineq_wider)
Bang_W_Ineq_and_Growth <- to_join(Bang_I_Growth_cleaned, Bang_W_Ineq_wider)
Ken_I_Ineq_and_Growth <- to_join(Ken_I_Growth_cleaned, Ken_I_Ineq_wider)
Ken_W_Ineq_and_Growth <- to_join(Ken_I_Growth_cleaned, Ken_W_Ineq_wider)

# Calculating the pearson correlation coefficient with outliers
Bang_I_Ineq_and_Growth_Correlation <- calc_correlation(Bang_I_Ineq_and_Growth)
Bang_W_Ineq_and_Growth_Correlation <- calc_correlation(Bang_W_Ineq_and_Growth)
Ken_I_Ineq_and_Growth_Correlation <- calc_correlation(Ken_I_Ineq_and_Growth)

```

```

Ken_W_Ineq_and_Growth_Correlation <- calc_correlation(Ken_W_Ineq_and_Growth)

# Calculating the pearson correlation coefficient without outliers
Bang_I_Ineq_and_Growth_Correlation_NoOutliers <- calc_correlation_noOutliers(Bang_I_Ineq_and_Growth)
Bang_W_Ineq_and_Growth_Correlation_NoOutliers <- calc_correlation_noOutliers(Bang_W_Ineq_and_Growth)
Ken_I_Ineq_and_Growth_Correlation_NoOutliers <- calc_correlation_noOutliers(Ken_I_Ineq_and_Growth)
Ken_W_Ineq_and_Growth_Correlation_NoOutliers <- calc_correlation_noOutliers(Ken_W_Ineq_and_Growth)

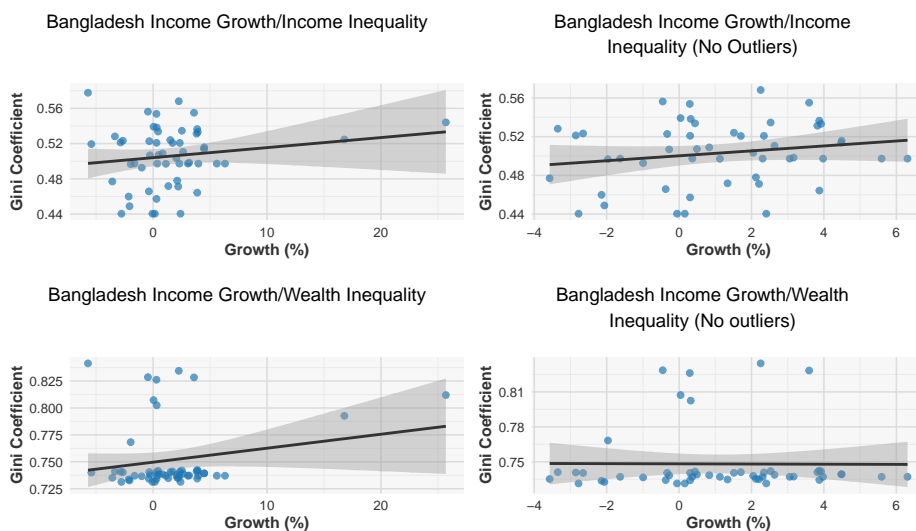
# Calculating the spearman correlation coefficient with outliers
Bang_I_Ineq_and_Growth_Correlation_Spearman <- calc_correlation_spearman(Bang_I_Ineq_and_Growth)
Bang_W_Ineq_and_Growth_Correlation_Spearman <- calc_correlation_spearman(Bang_W_Ineq_and_Growth)
Ken_I_Ineq_and_Growth_Correlation_Spearman <- calc_correlation_spearman(Ken_I_Ineq_and_Growth)
Ken_W_Ineq_and_Growth_Correlation_Spearman <- calc_correlation_spearman(Ken_W_Ineq_and_Growth)

# Plotting the graphs of Bangladesh
(graph(Bang_I_Ineq_and_Growth, "Bangladesh Income Growth/Income Inequality") |
graph_noOutliers(Bang_I_Ineq_and_Growth, "Bangladesh Income Growth/Income Inequality (No Outliers)")) /
(graph(Bang_W_Ineq_and_Growth, "Bangladesh Income Growth/Wealth Inequality") |
graph_noOutliers(Bang_W_Ineq_and_Growth, "Bangladesh Income Growth/Wealth Inequality (No outliers)"))

## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

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

```

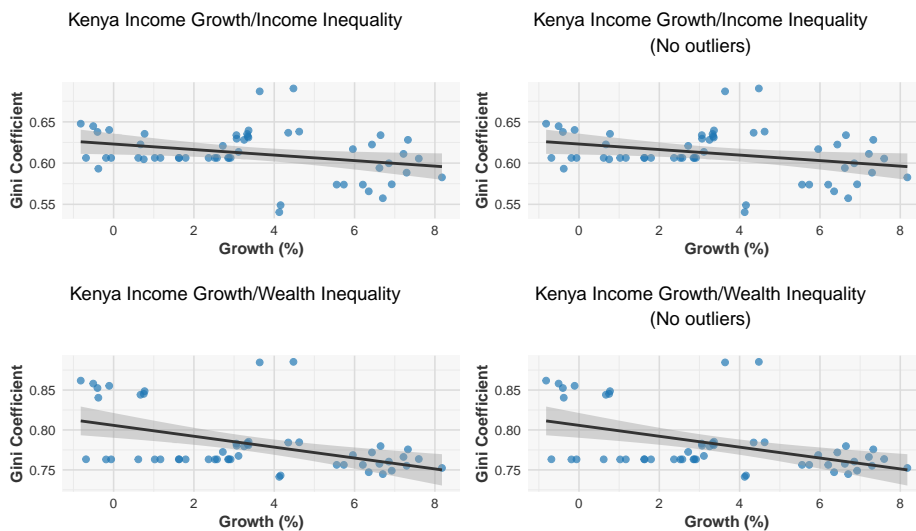


```

# Plotting the graphs of Kenya
(graph(Ken_I_Ineq_and_Growth, "Kenya Income Growth/Income Inequality") |
graph_noOutliers(Ken_I_Ineq_and_Growth, "Kenya Income Growth/Income Inequality (No outliers)")) /
(graph(Ken_W_Ineq_and_Growth, "Kenya Income Growth/Wealth Inequality") |
graph_noOutliers(Ken_W_Ineq_and_Growth, "Kenya Income Growth/Wealth Inequality (No outliers)"))

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

```



We then calculate the p value, with a hypothesis test of:  $H_0: p = 0$   $H_1: p \neq 0$

With a significance level of 0.05

```

cor_p_value <- function(r, n) {
  t_value <- r * sqrt((n - 2) / (1 - r^2))
  degf <- n - 2
  p_value <- 2 * (1 - pt(abs(t_value), degf))
  p_value_char <- as.character(p_value)
  if (p_value < 0.05) {
    return(paste(p_value_char, "There is sufficient evidence showing correlation"))
  }
  else
  {
    return(paste(p_value_char, "There is insufficient evidence to show correlation"))
  }
}

```

```

# Function to create a dataframe with no outliers
no_outliers <- function(x) {
  Q1 <- quantile(x$Growth, 0.25, na.rm = TRUE)
  Q3 <- quantile(x$Growth, 0.75, na.rm = TRUE)

```

```

IQR <- Q3 - Q1

y <- x %>%
  filter(Growth > Q1 - 1.5*IQR & Growth < Q3 + 1.5*IQR)

  return(y)
}

Bang_I_Ineq_and_Growth_Correlation_Pearson_Sig = cor_p_value(Bang_I_Ineq_and_Growth_Correlation, nrow(B
Bang_I_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig = cor_p_value(Bang_I_Ineq_and_Growth_Correlat
Bang_I_Ineq_and_Growth_Correlation_Spearman_Sig = cor_p_value(Bang_I_Ineq_and_Growth_Correlation_Spearman

Bang_W_Ineq_and_Growth_Correlation_Pearson_Sig = cor_p_value(Bang_W_Ineq_and_Growth_Correlation, nrow(B
Bang_W_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig = cor_p_value(Bang_W_Ineq_and_Growth_Correlat
Bang_W_Ineq_and_Growth_Correlation_Spearman_Sig = cor_p_value(Bang_W_Ineq_and_Growth_Correlation_Spearman

Ken_I_Ineq_and_Growth_Correlation_Pearson_Sig = cor_p_value(Ken_I_Ineq_and_Growth_Correlation, nrow(Ken
Ken_I_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig = cor_p_value(Ken_I_Ineq_and_Growth_Correlation
Ken_I_Ineq_and_Growth_Correlation_Spearman_Sig = cor_p_value(Ken_I_Ineq_and_Growth_Correlation_Spearman

Ken_W_Ineq_and_Growth_Correlation_Pearson_Sig = cor_p_value(Ken_W_Ineq_and_Growth_Correlation, nrow(Ken
Ken_W_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig = cor_p_value(Ken_W_Ineq_and_Growth_Correlation
Ken_W_Ineq_and_Growth_Correlation_Spearman_Sig = cor_p_value(Ken_W_Ineq_and_Growth_Correlation_Spearman

```

This next chunk is taking all of the correlation values I have calculated and created a data frame out of it

```

correlation_df <- data.frame(
  Country = c("Bangladesh Income Pearson", "Bangladesh Wealth Pearson", "Kenya Income Pearson", "Kenya Wealth Pearson"),
  Correlation = c(
    Bang_I_Ineq_and_Growth_Correlation,
    Bang_W_Ineq_and_Growth_Correlation,
    Ken_I_Ineq_and_Growth_Correlation,
    Ken_W_Ineq_and_Growth_Correlation,
    Bang_I_Ineq_and_Growth_Correlation_NoOutliers,
    Bang_W_Ineq_and_Growth_Correlation_NoOutliers,
    Ken_I_Ineq_and_Growth_Correlation_NoOutliers,
    Ken_W_Ineq_and_Growth_Correlation_NoOutliers,
    Bang_I_Ineq_and_Growth_Correlation_Spearman,
    Bang_W_Ineq_and_Growth_Correlation_Spearman,
    Ken_I_Ineq_and_Growth_Correlation_Spearman,
    Ken_W_Ineq_and_Growth_Correlation_Spearman
  ),
  P_Value = c(
    Bang_I_Ineq_and_Growth_Correlation_Pearson_Sig,
    Bang_I_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig,
    Bang_I_Ineq_and_Growth_Correlation_Spearman_Sig,
    Bang_W_Ineq_and_Growth_Correlation_Pearson_Sig,
    Bang_W_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig,
    Bang_W_Ineq_and_Growth_Correlation_Spearman_Sig,
  )
)

```



```

Ken_I_Ineq_and_Growth_Correlation_Pearson_Sig,
Ken_I_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig,
Ken_I_Ineq_and_Growth_Correlation_Spearman_Sig,

Ken_W_Ineq_and_Growth_Correlation_Pearson_Sig,
Ken_W_Ineq_and_Growth_Correlation_Pearson_noOutliers_Sig,
Ken_W_Ineq_and_Growth_Correlation_Spearman_Sig
))

correlation_df

```

```

##                Country Correlation
## 1      Bangladesh Income Pearson  0.161933870
## 2      Bangladesh Wealth Pearson  0.196492884
## 3              Kenya Income Pearson -0.300089929
## 4              Kenya Wealth Pearson -0.463112553
## 5 Bangladesh Income Pearson (no outliers)  0.186709229
## 6 Bangladesh Wealth Pearson (no outliers) -0.006729303
## 7              Kenya Income Pearson (no outliers) -0.300089929
## 8              Kenya Wealth Pearson (no outliers) -0.463112553
## 9              Bangladesh Income Spearman  0.139569609
## 10             Bangladesh Wealth Spearman  0.111406254
## 11              Kenya Income Spearman -0.285467035
## 12              Kenya Wealth Spearman -0.378578910
##
##                P_Value
## 1  0.242056224902433 There is insufficient evidence to show correlation
## 2   0.19418695207814 There is insufficient evidence to show correlation
## 3   0.314145323540167 There is insufficient evidence to show correlation
## 4   0.154428807283367 There is insufficient evidence to show correlation
## 5    0.96300711595396 There is insufficient evidence to show correlation
## 6   0.422546753733279 There is insufficient evidence to show correlation
## 7    0.0274759172689045 There is sufficient evidence showing correlation
## 8    0.0274759172689045 There is sufficient evidence showing correlation
## 9    0.0364010904097185 There is sufficient evidence showing correlation
## 10  0.000421328413568789 There is sufficient evidence showing correlation
## 11  0.000421328413568789 There is sufficient evidence showing correlation
## 12   0.0047609695818327 There is sufficient evidence showing correlation

```

Analysis on the data:

Sufficient evidence for Kenya Income and Wealth Pearson (no outliers) and Spearman, Bangladesh Income and Wealth Spearman

## Question 4

Use R to analyse the correlation between international trade and income inequality, and international trade and income growth for the two countries. [15 marks]

Chunk with functions calculate dfs with no outliers

```

calc_outliers_ <- function(x) {
  Q1 <- quantile(x, 0.25, na.rm = TRUE)
  Q3 <- quantile(x, 0.75, na.rm = TRUE)
  IQR <- Q3 - Q1

  x[x < Q1 - 1.5*IQR | x > Q3 + 1.5*IQR] <- NA
  return(x)
}

```

Chunk to pass in all previous data

```

# Merging the dfs into 2 large dfs
All_Data_Bang <- merge(Trade_cleaned, Bang_I_Ineq_and_Growth, by = "Year")
All_Data_Ken <- merge(Trade_cleaned, Ken_I_Ineq_and_Growth, by = "Year")

All_Data_Bang$Bang_Trade <- as.numeric(All_Data_Bang$Bang_Trade)
All_Data_Ken$Ken_Trade <- as.numeric(All_Data_Ken$Ken_Trade)

# Creating correlation data

Bang_Trade_Ineq_Corr <- cor(All_Data_Bang$Bang_Trade, All_Data_Bang$Gini_Coeff)
Bang_Trade_Growth_Corr <- cor(All_Data_Bang$Bang_Trade, All_Data_Bang$Growth)
Ken_Trade_Ineq_Corr <- cor(All_Data_Ken$Ken_Trade, All_Data_Ken$Gini_Coeff)
Ken_Trade_Growth_Corr <- cor(All_Data_Ken$Ken_Trade, All_Data_Ken$Growth)

# No outliers
All_Data_Bang$Bang_Trade_clean <- calc_outliers_(All_Data_Bang$Bang_Trade)
All_Data_Bang$Gini_Coeff_clean <- calc_outliers_(All_Data_Bang$Gini_Coeff)
Bang_clean <- na.omit(All_Data_Bang[, c("Bang_Trade_clean", "Gini_Coeff_clean")])
Bang_Trade_Ineq_Corr_noOutliers <- cor(Bang_clean$Bang_Trade_clean, Bang_clean$Gini_Coeff_clean)

All_Data_Bang$Bang_Trade_clean <- calc_outliers_(All_Data_Bang$Bang_Trade)
All_Data_Bang$Growth_clean <- calc_outliers_(All_Data_Bang$Growth)
Bang_clean <- na.omit(All_Data_Bang[, c("Bang_Trade_clean", "Growth_clean")])
Bang_Trade_Growth_Corr_noOutliers <- cor(Bang_clean$Bang_Trade_clean, Bang_clean$Growth_clean)

All_Data_Ken$Ken_Trade_clean <- calc_outliers_(All_Data_Ken$Ken_Trade)
All_Data_Ken$Gini_Coeff_clean <- calc_outliers_(All_Data_Ken$Gini_Coeff)
Ken_clean <- na.omit(All_Data_Ken[, c("Ken_Trade_clean", "Gini_Coeff_clean")])
Ken_Trade_Ineq_Corr_noOutliers <- cor(Ken_clean$Ken_Trade_clean, Ken_clean$Gini_Coeff_clean)

All_Data_Ken$Ken_Trade_clean <- calc_outliers_(All_Data_Ken$Ken_Trade)
All_Data_Ken$Growth_clean <- calc_outliers_(All_Data_Ken$Growth)
Ken_clean <- na.omit(All_Data_Ken[, c("Ken_Trade_clean", "Growth_clean")])
Ken_Trade_Growth_Corr_noOutliers <- cor(Ken_clean$Ken_Trade_clean, Ken_clean$Growth_clean)

# Spearman correlations

```

```

Bang_Trade_Ineq_Corr_spearman <- cor(All_Data_Bang$Bang_Trade, All_Data_Bang$Gini_Coeff, method = "spearman")
Bang_Trade_Growth_Corr_spearman <- cor(All_Data_Bang$Bang_Trade, All_Data_Bang$Growth, method = "spearman")
Ken_Trade_Ineq_Corr_spearman <- cor(All_Data_Ken$Ken_Trade, All_Data_Ken$Gini_Coeff, method = "spearman")
Ken_Trade_Growth_Corr_spearman <- cor(All_Data_Ken$Ken_Trade, All_Data_Ken$Growth, method = "spearman")

```

Chunk with functions to display the graph

```

# Graphing Trade vs Gini for Bangladesh
(ggplot(All_Data_Bang, aes(x = Bang_Trade, y = Gini_Coeff)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) + # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) + # line: dark gray
  theme_minimal(base_size = 6) + # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"), # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5), # spacing above/below
      hjust = 0.5 )) +
  labs(
    title = "Bangladesh: Trade/Gini",
    x = "Current Account ($bn)",
    y = "Gini Coefficient"
  ) |
ggplot(All_Data_Bang, aes(x = Bang_Trade_clean, y = Gini_Coeff_clean)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) + # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) + # line: dark gray
  theme_minimal(base_size = 6) + # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"), # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5), # spacing above/below
      hjust = 0.5 )) +
  labs(
    title = "Bangladesh: Trade/Gini (no outliers)",
    x = "Current Account ($bn)",
    y = "Gini Coefficient"
  )) /
(ggplot(All_Data_Bang, aes(x = Bang_Trade, y = Growth)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) + # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) + # line: dark gray
  theme_minimal(base_size = 6) + # clean minimal theme
  theme(

```

```

panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
panel.grid.major = element_line(color = "#dcdcdc"),           # soft grid lines
panel.grid.minor = element_line(color = "#eaeaea"),
axis.title = element_text(color = "#333333", face = "bold"),
axis.text = element_text(color = "#333333"),
plot.title.position = "plot",
plot.title = element_text(
  lineheight = 1.1,
  margin = margin(b = 10, t = 5),    # spacing above/below
  hjust = 0.5 )) +
labs(
  title = "Bangladesh: Trade/Growth",
  x = "Growth (%)",
  y = "Gini Coefficient"
) |
ggplot(All_Data_Bang, aes(x = Bang_Trade_clean, y = Growth_clean)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) +      # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) +    # line: dark gray
  theme_minimal(base_size = 6) +                               # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"),           # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5),    # spacing above/below
      hjust = 0.5 )) +
    labs(
      title = "Bangladesh: Trade/Growth (no outliers)",
      x = "Growth (%)",
      y = "Gini Coefficient"
    )
  ))

```

```

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

```

```

## Warning: Removed 6 rows containing non-finite outside the scale range
## ('stat_smooth()').

```

```

## Warning: Removed 6 rows containing missing values or values outside the scale range
## ('geom_point()').

```

```

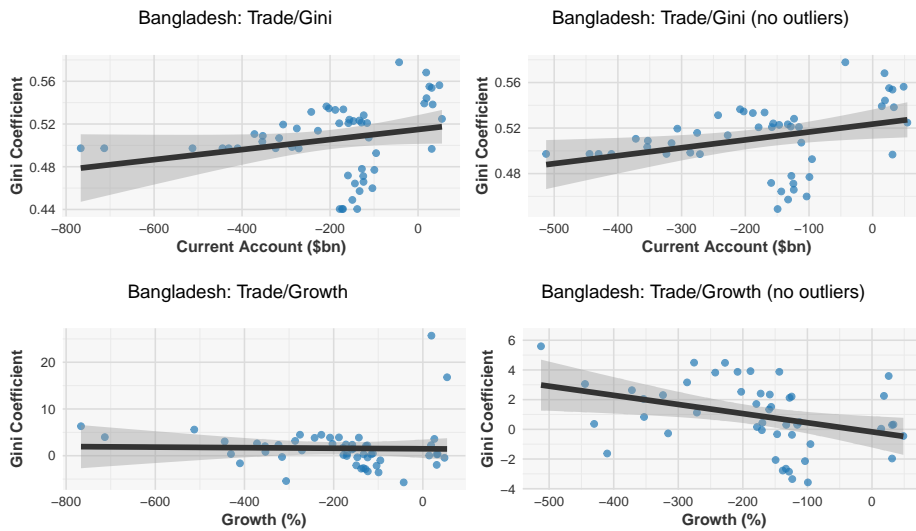
## 'geom_smooth()' using formula = 'y ~ x'
## 'geom_smooth()' using formula = 'y ~ x'

```

```

## Warning: Removed 6 rows containing non-finite outside the scale range ('stat_smooth()').
## Removed 6 rows containing missing values or values outside the scale range
## ('geom_point()').

```



```
# Graphing Trade vs Gini for Kenya
(ggplot(All_Data_Ken, aes(x = Ken_Trade, y = Gini_Coeff)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) + # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) + # line: dark gray
  theme_minimal(base_size = 6) + # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"), # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5), # spacing above/below
      hjust = 0.5 )) +
  labs(
    title = "Kenya: Trade/Gini",
    x = "Current Account ($bn)",
    y = "Gini Coefficient"
  ) |
ggplot(All_Data_Ken, aes(x = Ken_Trade_clean, y = Gini_Coeff_clean)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) + # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) + # line: dark gray
  theme_minimal(base_size = 6) + # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"), # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5), # spacing above/below
```

```

    hjust = 0.5 ))+
  labs(
    title = "Kenya: Trade/Gini (no outliers)",
    x = "Current Account ($bn)",
    y = "Gini Coefficient"
  ) /
(ggplot(All_Data_Ken, aes(x = Ken_Trade, y = Growth)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) +      # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) +    # line: dark gray
  theme_minimal(base_size = 6) +                               # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"),          # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5),      # spacing above/below
      hjust = 0.5 )) +
  labs(
    title = "Kenya: Trade/Growth",
    x = "Growth (%)",
    y = "Gini Coefficient"
  ) |
ggplot(All_Data_Ken, aes(x = Ken_Trade_clean, y = Growth_clean)) +
  geom_point(color = "#1f77b4", size = 1, alpha = 0.7) +      # points: muted blue
  geom_smooth(method = "lm", color = "#333333", size = 1) +    # line: dark gray
  theme_minimal(base_size = 6) +                               # clean minimal theme
  theme(
    panel.background = element_rect(fill = "#f7f7f7", color = NA), # light gray background
    panel.grid.major = element_line(color = "#dcdcdc"),          # soft grid lines
    panel.grid.minor = element_line(color = "#eaeaea"),
    axis.title = element_text(color = "#333333", face = "bold"),
    axis.text = element_text(color = "#333333"),
    plot.title.position = "plot",
    plot.title = element_text(
      lineheight = 1.1,
      margin = margin(b = 10, t = 5),      # spacing above/below
      hjust = 0.5 )) +
  labs(
    title = "Kenya: Trade/Growth (no outliers)",
    x = "Growth (%)",
    y = "Gini Coefficient"
  ))

```

```

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

```

```

## Warning: Removed 5 rows containing non-finite outside the scale range
## ('stat_smooth()').

```

```

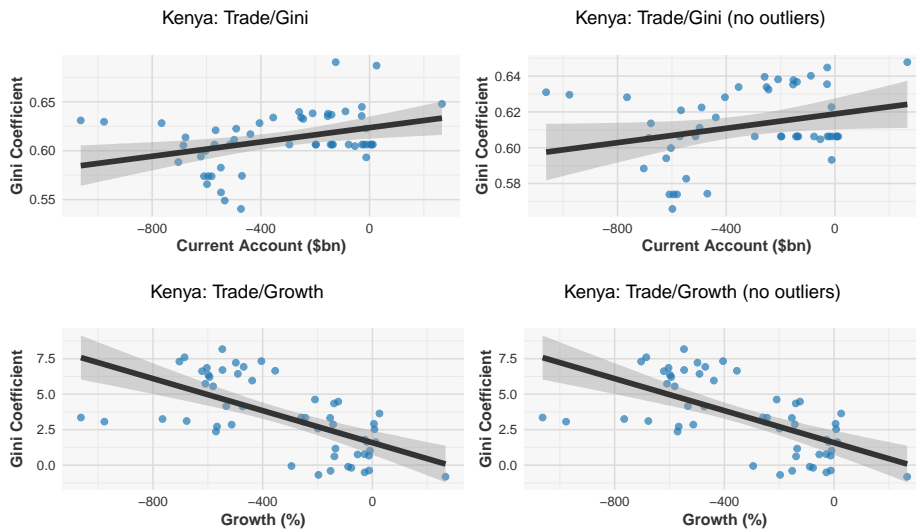
## Warning: Removed 5 rows containing missing values or values outside the scale range

```

```
## ('geom_point()').
```

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

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



We then calculate the p value, with a hypothesis test of:  $H_0: p = 0$   $H_1: p \neq 0$

With a significance level of 0.05

```
# Function to calculate the p value
p_value_calc <- function(r, n)
{
  t <- r * sqrt(n - 2) / sqrt(1 - r^2)
  df <- n - 2
  p <- 2 * (1 - pt(abs(t), df))
  p_char <- as.character(p)

  if(p < 0.05) {
    return(paste(p_char, "There is sufficient evidence showing correlation"))
  }
  else {
    return(paste(p_char, "There is insufficient evidence to showing correlation"))
  }
}

Bang_Trade_Ineq_p <- p_value_calc(Bang_Trade_Ineq_Corr, nrow(All_Data_Bang))
Bang_Trade_Growth_p <- p_value_calc(Bang_Trade_Growth_Corr, nrow(All_Data_Bang))
Ken_Trade_Ineq_p <- p_value_calc(Ken_Trade_Ineq_Corr, nrow(All_Data_Ken))
Ken_Trade_Growth_p <- p_value_calc(Ken_Trade_Growth_Corr, nrow(All_Data_Ken))

All_Data_Bang_clean <- All_Data_Bang[, c("Bang_Trade_clean", "Gini_Coeff_clean")]
All_Data_Bang_clean <- na.omit(All_Data_Bang_clean)
```

```

All_Data_Ken_clean <- All_Data_Ken[, c("Ken_Trade_clean", "Gini_Coeff_clean")]
All_Data_Ken_clean <- na.omit(All_Data_Ken_clean)

Bang_Trade_Ineq_p_noOutliers <- p_value_calc(Bang_Trade_Ineq_Corr_noOutliers, nrow(All_Data_Bang_clean))
Bang_Trade_Growth_p_noOutliers <- p_value_calc(Bang_Trade_Growth_Corr_noOutliers, nrow(All_Data_Bang_clean))
Ken_Trade_Ineq_p_noOutliers <- p_value_calc(Ken_Trade_Ineq_Corr_noOutliers, nrow(All_Data_Ken_clean))
Ken_Trade_Growth_p_noOutliers <- p_value_calc(Ken_Trade_Growth_Corr_noOutliers, nrow(All_Data_Ken_clean))

Bang_Trade_Ineq_p_spearman <- p_value_calc(Bang_Trade_Ineq_Corr_spearman, nrow(All_Data_Bang))
Bang_Trade_Growth_p_spearman <- p_value_calc(Bang_Trade_Growth_Corr_spearman, nrow(All_Data_Bang))
Ken_Trade_Ineq_p_spearman <- p_value_calc(Ken_Trade_Ineq_Corr_spearman, nrow(All_Data_Ken))
Ken_Trade_Growth_p_spearman <- p_value_calc(Ken_Trade_Growth_Corr_spearman, nrow(All_Data_Ken))

correlation_df_4 <- data.frame(
  Country_Trade_Comp = c("Bangladesh Inequality Pearson", "Bangladesh Growth Pearson", "Kenya Inequality",
    "Kenya Growth", "Kenya Inequality Spearman", "Kenya Growth Spearman"),
  Correlation_ = c(
    Bang_Trade_Ineq_Corr,
    Bang_Trade_Growth_Corr,
    Ken_Trade_Ineq_Corr,
    Ken_Trade_Growth_Corr,
    Bang_Trade_Ineq_Corr_noOutliers,
    Bang_Trade_Growth_Corr_noOutliers,
    Ken_Trade_Ineq_Corr_noOutliers,
    Ken_Trade_Growth_Corr_noOutliers,
    Bang_Trade_Ineq_Corr_spearman,
    Bang_Trade_Growth_Corr_spearman,
    Ken_Trade_Ineq_Corr_spearman,
    Ken_Trade_Growth_Corr_spearman
  ),
  P_Value_ = c(
    Bang_Trade_Ineq_p,
    Bang_Trade_Growth_p,
    Ken_Trade_Ineq_p,
    Ken_Trade_Growth_p,
    Bang_Trade_Ineq_p_noOutliers,
    Bang_Trade_Growth_p_noOutliers,
    Ken_Trade_Ineq_p_noOutliers,
    Ken_Trade_Growth_p_noOutliers,
    Bang_Trade_Ineq_p_spearman,
    Bang_Trade_Growth_p_spearman,
    Ken_Trade_Ineq_p_spearman,
    Ken_Trade_Growth_p_spearman
  ))
correlation_df_4

```

```

##               Country_Trade_Comp Correlation_
## 1 Bangladesh Inequality Pearson    0.24152786

```



```

## 2          Bangladesh Growth Pearson -0.02158368
## 3          Kenya Inequality Pearson  0.36294178
## 4          Kenya Growth Pearson    -0.62513074
## 5 Bangladesh Inequality Pearson (no outliers) 0.33389853
## 6          Bangladesh Growth Pearson (no outliers) -0.35802412
## 7          Kenya Inequality Pearson (no outliers) 0.28321365
## 8          Kenya Growth Pearson (no outliers) -0.62513074
## 9          Bangladesh Inequality Spearman 0.28494031
## 10         Bangladesh Growth Spearman -0.34903755
## 11         Kenya Inequality Spearman 0.35124622
## 12         Kenya Growth Spearman -0.63765962
##
##                                     P_Value_
## 1 0.0784956098191993 There is insufficient evidence to showing correlation
## 2 0.876889318374959 There is insufficient evidence to showing correlation
## 3 0.00699019520730992 There is sufficient evidence showing correlation
## 4 4.34593126596994e-07 There is sufficient evidence showing correlation
## 5 0.0203765573776791 There is sufficient evidence showing correlation
## 6 0.0124715626918328 There is sufficient evidence showing correlation
## 7 0.0486214606888931 There is sufficient evidence showing correlation
## 8 1.57313313930985e-06 There is sufficient evidence showing correlation
## 9 0.036762748462349 There is sufficient evidence showing correlation
## 10 0.00968746866326264 There is sufficient evidence showing correlation
## 11 0.00920666913912882 There is sufficient evidence showing correlation
## 12 2.15304707396058e-07 There is sufficient evidence showing correlation

```

Analysis:

#AI Usage I have used AI in producing my graphs, specifically, for making the graphs have a nice theme. The prompt used: using ggplot2, write me code to change the colour of the points, of the line, and change the background to be nice