

# DA5020.A7.Nithya.Sarabudla

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2023-10-25

## loading the packages

```
library(rvest)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(ggplot2)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v forcats   1.0.0      v stringr   1.5.0
## v lubridate 1.9.2      v tibble   3.2.1
## v purrr     1.0.2      v tidyr    1.3.0
## v readr     2.1.4
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter()      masks stats::filter()
## x readr::guess_encoding() masks rvest::guess_encoding()
## x dplyr::lag()         masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

extract the tabular data on the “Percent of

population living on less than \$1.15, \$2.60 and \$3.85 a day” from the Wikipedia page.

## Question 1

### PART 1

Scrape the data from the webpage and extract the following fields: Country, < \$1.15, <

\$2.60, < \$3.85, Year and Continent. Prepare the data for analysis and ensure that the columns have

meaningful names.

```
# Read the HTML page from the given URL
page = read_html("https://en.wikipedia.org/w/index.php?title=List_of_sovereign_states_by_percentage_of_")

# Extract the third table from the HTML page
table <- html_table(page)[3]
table
```

```
## [[1]]
## # A tibble: 166 x 6
##   Country    '$1.15' '$2.60' '$3.85'   Year Continent
##   <chr>      <chr>   <chr>   <chr>   <int> <chr>
## 1 Albania    13.35    26.60%  38.40%   2023 Europe
## 2 Algeria     0.32%    2.23%   20.83%   2019 Africa
## 3 Angola     51.40%   72.79%  89.13%   2019 Africa
## 4 Argentina   1.60%    5.80%   18.20%   2020 South America
## 5 Armenia     0.40%    6.90%   44.70%   2020 Asia
## 6 Australia   0.50%    0.74%    0.74%   2019 Oceania
## 7 Austria     0.60%    0.70%    0.80%   2019 Europe
## 8 Azerbaijan 0.00%    0.00%    0.00%   2019 Asia
## 9 Bangladesh 6.62%   37.44%  76.01%   2019 Asia
## 10 Belarus    0.00%    0.00%    0.10%   2020 Europe
## # i 156 more rows
```

```
# Convert the table to a data frame
df <- data.frame(table)

# Rename the columns
colnames(df) <- c("Country", "less_than_1.15", "less_than_2.60", "less_than_3.85", "Year", "Continent")

# Clean and convert the "less_than_1.15" column by removing any trailing characters and converting to numeric
df$less_than_1.15 <- as.array(df$less_than_1.15)
df$less_than_1.15 <- substring(df$less_than_1.15,1,nchar(df$less_than_1.15)-1)
```

```
df$less_than_1.15 <- as.numeric(df$less_than_1.15)

# Clean and convert the "less_than_2.60" column by removing any trailing characters and converting to numeric
df$less_than_2.60<- as.array(df$less_than_2.60)
df$less_than_2.60 <- substring(df$less_than_2.60,1,nchar(df$less_than_2.60)-1)
df$less_than_2.60 <- as.numeric(df$less_than_2.60)

# Clean and convert the "less_than_3.85" column by removing any trailing characters and converting to numeric
df$less_than_3.85 <- as.array(df$less_than_3.85)
df$less_than_3.85 <- substring(df$less_than_3.85,1,nchar(df$less_than_3.85)-1)
df$less_than_3.85<- as.numeric(df$less_than_3.85)

# Display the first few rows of the data frame
head(df)
```

```
##      Country less_than_1.15 less_than_2.60 less_than_3.85 Year      Continent
## 1  Albania      13.30      26.60      38.40 2023      Europe
## 2  Algeria       0.32       2.23     20.83 2019      Africa
## 3   Angola     51.40     72.79     89.13 2019      Africa
## 4 Argentina      1.60       5.80     18.20 2020 South America
## 5  Armenia       0.40       6.90     44.70 2020      Asia
## 6 Australia      0.50       0.74      0.74 2019     Oceania
```

## PART 2

Calculate the mean and the standard deviation of the percent of the population living under

\$3.85 per day for each continent. Perform a comparative analysis (i.e. explanation) of the data from

each continent

```
# Calculate the mean and standard deviation of "less_than_3.85" percentages for each continent
less_than_3.85_df <- df %>%
  select(Continent, less_than_3.85) %>%
  group_by(Continent) %>%
  summarise(Mean = mean(less_than_3.85, na.rm = TRUE), std_dev = sd(less_than_3.85, na.rm = TRUE))

less_than_3.85_df
```

```
## # A tibble: 7 x 3
##   Continent      Mean std_dev
##   <chr>      <dbl>  <dbl>
## 1 Africa      74.3    25.4
## 2 Asia        33.8    30.2
## 3 Asia, Europe  6.74    4.90
## 4 Europe       5.03    9.22
## 5 North America 28.5    20.7
```

```
## 6 Oceania          49.2    27.4
## 7 South America    21.3    12.8
```

```
# Find and display the continent with the lowest mean poverty level
min_mean_continent <- less_than_3.85_df %>%
  filter(Mean == min(Mean)) %>%
  select(Continent) %>%
  pull()
cat("Continent with the lowest mean poverty level:", min_mean_continent, "\n")
```

```
## Continent with the lowest mean poverty level: Europe
```

```
# Find and display the continent with the lowest standard deviation
min_std_dev_continent <- less_than_3.85_df %>%
  filter(std_dev == min(std_dev)) %>%
  select(Continent) %>%
  pull()
cat("Continent with the lowest standard deviation:", min_std_dev_continent, "\n")
```

```
## Continent with the lowest standard deviation: Asia, Europe
```

```
# Find and display the continent with the highest mean poverty level
max_mean_continent <- less_than_3.85_df %>%
  filter(Mean == max(Mean)) %>%
  select(Continent) %>%
  pull()
cat("Continent with the highest mean poverty level:", max_mean_continent, "\n")
```

```
## Continent with the highest mean poverty level: Africa
```

```
# Find and display the continent with the highest standard deviation
max_std_dev_continent <- less_than_3.85_df %>%
  filter(std_dev == max(std_dev)) %>%
  select(Continent) %>%
  pull()
cat("Continent with the highest standard deviation:", max_std_dev_continent, "\n")
```

```
## Continent with the highest standard deviation: Asia
```

## PART 3

What are the 10 countries with the highest percentage of the population having an income

of less than \$3.85 per day? Using a suitable chart, display the country name, the percentage and

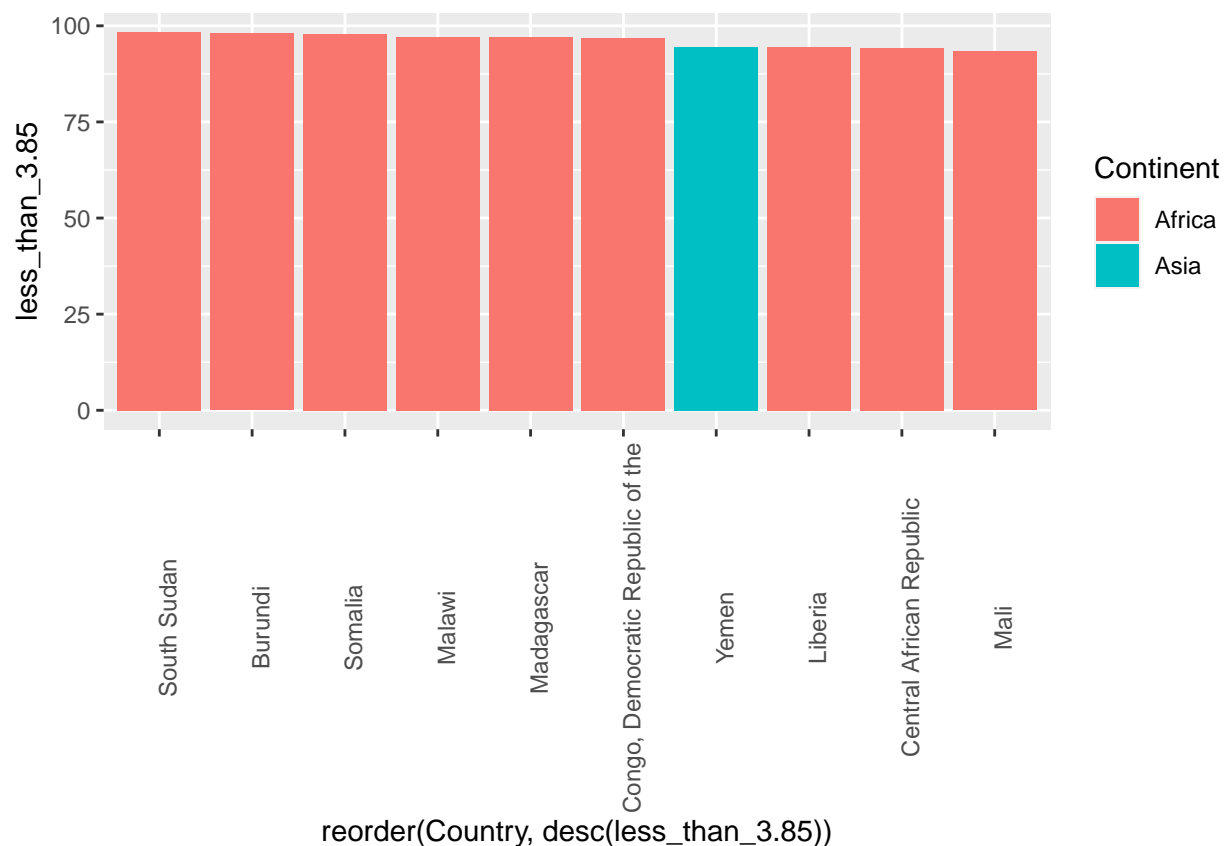
color- code by the Continent. Summarize your findings

```

# Extract the top 10 countries with the highest percentage of income less than $3.85
top_10_countries <- df %>%
  arrange(desc(less_than_3.85)) %>%
  head(10)

# Create a bar plot to visualize the top 10 countries' income percentages
df %>%
  arrange(desc(less_than_3.85)) %>%
  select(Country, less_than_3.85, Continent) %>%
  head(10) %>%
  ggplot(aes(x = reorder(Country, desc(less_than_3.85)), y = less_than_3.85, fill = Continent)) +
  geom_bar(stat = 'identity') +
  theme(axis.text.x = element_text(angle = 90))

```



```

# Summarize your findings
summary(top_10_countries)

```

```

##      Country      less_than_1.15  less_than_2.60  less_than_3.85
## Length:10      Min.      :42.26   Min.      :74.64   Min.      :93.29
## Class :character 1st Qu.:58.18   1st Qu.:82.98   1st Qu.:94.48
## Mode  :character Median :70.88   Median :88.64   Median :96.94
##                Mean  :66.58   Mean  :86.52   Mean  :96.19
##                3rd Qu.:75.79   3rd Qu.:90.39   3rd Qu.:97.68
##                Max.   :80.71   Max.   :93.27   Max.   :98.44

```

```
##      Year      Continent
## Min.   :2019   Length:10
## 1st Qu.:2019   Class :character
## Median :2019   Mode  :character
## Mean   :2019
## 3rd Qu.:2019
## Max.   :2019
```

From the above plot, it's evident that the majority of the countries with the highest percentage of the population living on less than \$3.85 per day are from Africa. Specifically, out of the top 10 countries in this category, 9 are from Africa, while only 1 country is from Asia. South Sudan ranks as the top country with the highest percentage of its population earning less than \$3.85 per day. The 10th country on the list with the highest population living on less than \$3.85 per day is Mali, also from Africa.

## PART 4

Explore the countries with the lowest percentage of the population having an income of

less than \$3.85 per day. What are the 5 countries with the lowest percentage, and how does the

results compare to the other income groups (i.e. \$1.15 and \$2.60)?

```
# Arrange the data frame in ascending order of the "less_than_3.85" column
# and select the top 5 rows with the lowest income percentages
df %>% arrange(less_than_3.85) %>% head(5)
```

```
##      Country less_than_1.15 less_than_2.60 less_than_3.85 Year
## 1      Azerbaijan          0          0.0          0.00 2019
## 2 United Arab Emirates          0          0.0          0.00 2019
## 3          Iceland          0          0.0          0.04 2019
## 4          Belarus          0          0.0          0.10 2020
## 5          Finland          0          0.1          0.10 2019
##      Continent
## 1      Asia
## 2      Asia
## 3     Europe
## 4     Europe
## 5     Europe
```

Among the 5 countries with the lowest population percentage living on less than \$3.85 per day, 2 are in Asia (Azerbaijan the United Arab Emirates) with a 0% population in this category, while the remaining 3 are from Europe (Iceland, Belarus, and Finland) with percentages ranging from 0.04% to 0.1%. For the less than \$2.60 per day threshold, only Finland has 0.1%, with the rest at 0%. In all these countries, when considering less than \$1.15 per day, the population percentage is 0%.

## PART 5

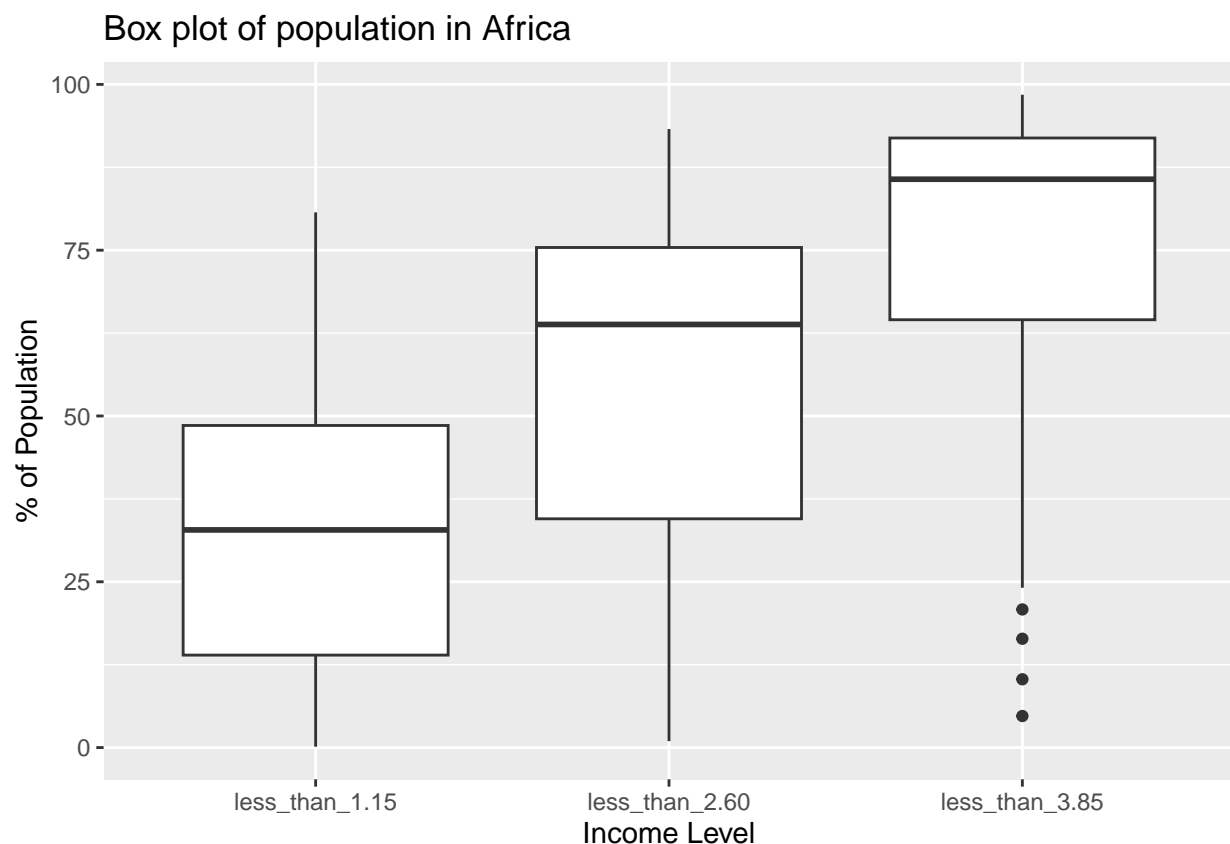
Extract the data for any two continents of your choice. For each continent, visualize the

percent of the population living on less than \$1.90, \$3.20 and \$5.50 using box plots. Compare and

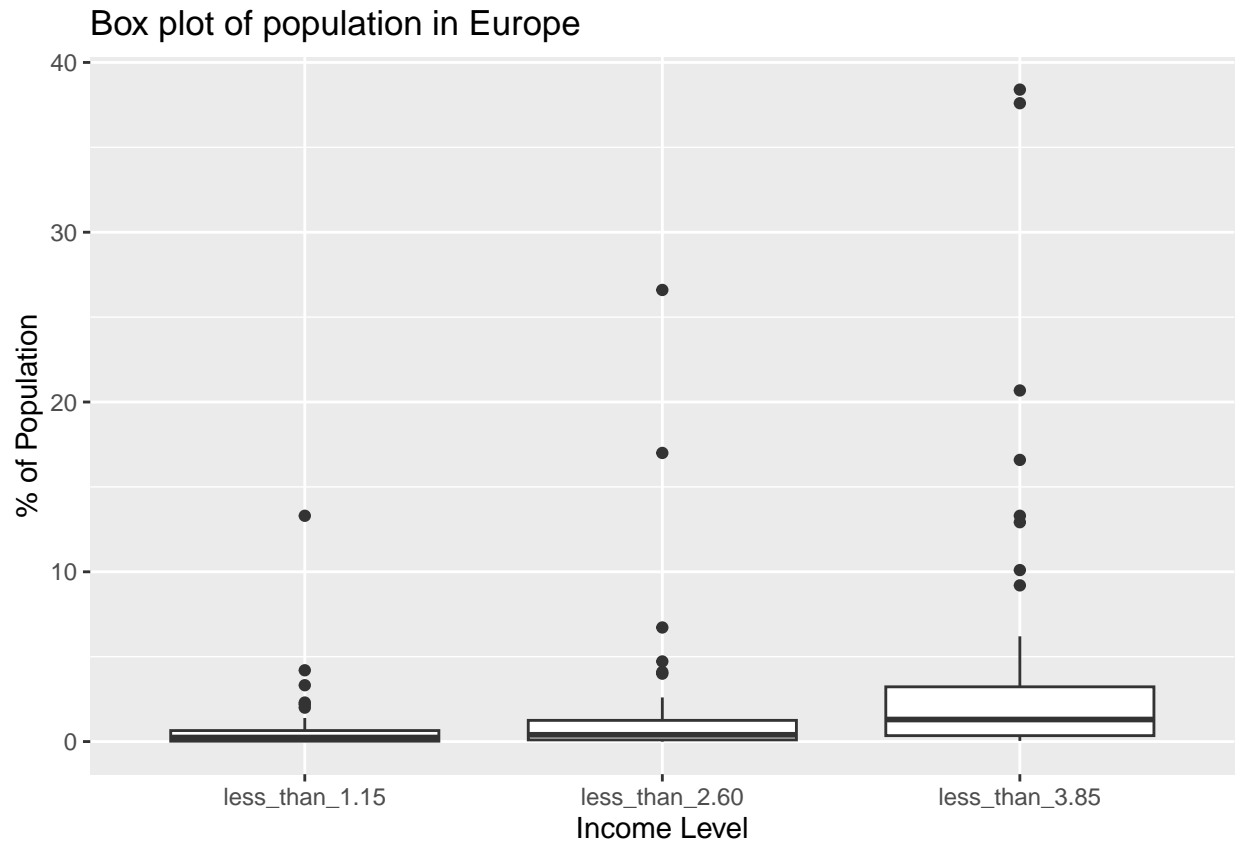
contrast the results, while ensuring that you discuss the distribution, skew and any outliers that are

evident.

```
# Create a box plot to visualize the income distribution in Africa
df %>%
  filter(Continent == "Africa") %>%
  pivot_longer(cols = starts_with("less"), names_to = "Income Level", values_to = "Percentage") %>%
  ggplot(aes(x = `Income Level`, y = Percentage)) +
  geom_boxplot() +
  ggtitle("Box plot of population in Africa") +
  xlab("Income Level") +
  ylab("% of Population")
```



```
# Create a box plot to visualize the income distribution in Europe
df %>%
  filter(Continent == "Europe") %>%
  pivot_longer(cols = starts_with("less"), names_to = "Income Level", values_to = "Percentage") %>%
  ggplot(aes(x = `Income Level`, y = Percentage)) +
  geom_boxplot() +
  ggtitle("Box plot of population in Europe") +
  xlab("Income Level") +
  ylab("% of Population")
```



### Box Plot for Asia:

Distribution: The box plot for “Asia” shows that the distribution of the population living on less than \$1.15 is right skewed, while the distribution for less than \$2.60 is slightly left-skewed, and the distribution for less than \$3.85 is strongly left-skewed.

Skew: Right-skewed for less than \$1.15, with most people having lower income. Slightly left-skewed for less than \$2.60, where the majority have higher income. Strongly left-skewed for less than \$3.85, indicating that the majority have even higher income levels.

Outliers: There are 4 outliers in less\_than\_3.85 means there are a few places or individuals in Asia with much lower income compared to most people.



### **Box Plot for Europe :**

In Europe, the income distribution for less than \$1.15, less than \$2.60, and less than \$3.85 is strongly right-skewed. This indicates that the majority of the population in Europe has higher income levels, with a long tail of individuals who have lower income. There are a few outliers in the three income levels, which represent exceptional cases of individuals or regions with significantly lower income than the majority of the population.