

CBAM Data Analysis

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Loading Data

Statistical Capability Data

```
SPI_index = read_delim("~/Downloads/SPI_index.csv")

## Rows: 3488 Columns: 79
## -- Column specification -----
## Delimiter: ","
## chr (4): country, iso3c, income, region
## dbl (75): date, SPI.INDEX.PIL1, SPI.INDEX.PIL2, SPI.INDEX.PIL3, SPI.INDEX.PI...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
SPI_index_2019 = SPI_index[grepl('2019', SPI_index$date),]
stats_capab_data = SPI_index_2019[c(1,2,9,76,77)]
colnames(stats_capab_data) = c("Country_Name",
  ↳ "Country_Code", "Statistical_Capability", "Income", "Region")
stats_capab_data$Income <- factor(stats_capab_data$Income , levels=c("Low
  ↳ income", "Lower middle income", "Upper middle income", "High income"))
```

```
dim(stats_capab_data)
```

```
## [1] 218 5
```

LDC Column

```
ldcs = c("Angola", "Benin", "Burkina Faso", "Burundi", "Central African
  ↳ Republic", "Chad", "Comoros", "Congo, Dem. Rep.", "Djibouti", "Eritrea",
  ↳ "Ethiopia", "Gambia, The", "Guinea", "Guinea-Bissau", "Lesotho", "Liberia",
  ↳ "Madagascar", "Malawi", "Mali", "Mauritania", "Mozambique", "Niger",
  ↳ "Rwanda", "Sao Tome and Principe", "Senegal", "Sierra Leone", "Somalia",
  ↳ "South Sudan", "Sudan", "Togo", "Uganda", "Tanzania", "Zambia",
  ↳ "Afghanistan", "Bangladesh", "Bhutan", "Cambodia", "Lao PDR", "Myanmar",
  ↳ "Nepal", "Timor-Leste", "Yemen, Rep.", "Haiti", "Kiribati", "Solomon
  ↳ Islands", "Tuvalu")

stats_capab_data$LDC = with(stats_capab_data,
  ↳ ifelse(stats_capab_data$Country_Name %in% ldcs, 1, 0))
table(stats_capab_data$LDC)
```

```
##  
##    0    1  
## 172  46
```

Export Data

```
comtrade_aluminum = read.csv('~Downloads/comtrade (4).csv')  
aluminum_76 = comtrade_aluminum[c(13,14,32)]  
colnames(aluminum_76) = c("Country_Name", "Country_Code", "Aluminum_Exports")  
dim(aluminum_76)
```

```
## [1] 161    3
```

Energy Data

```
iea_data = read_excel('~Downloads/energy statistics data.xlsx')  
dim(iea_data)
```

```
## [1] 146    9
```

Data Cleaning

Standardizing country names.

```
aluminum_76 <- mutate(aluminum_76, Country_Name=recode(Country_Name, "Bolivia"  
  ↪  (Plurinational State of)" = "Bolivia",  
    "Bosnia Herzegovina" = "Bosnia and Herzegovina",  
    "Congo" = "Congo, Rep.",  
    "Czechia" = "Czech Republic",  
    "Dominican Rep." = "Dominican Republic",  
    "China, Hong Kong SAR" = "Hong Kong SAR, China",  
    "Côte d'Ivoire" = "Cote d'Ivoire",  
    "Rep. of Korea" = "Korea, Rep.",  
    "Kyrgyzstan" = "Kyrgyz Republic",  
    "Lao People's Dem. Rep." = "Lao PDR",  
    "China, Macao SAR" = "Macao SAR, China",  
    "Slovakia" = "Slovak Republic",  
    "Viet Nam" = "Vietnam",  
    "United Rep. of Tanzania" = "Tanzania",  
    "USA" = "United States",  
    "Yemen" = "Yemen, Rep.",
```

```

"Bahamas" = "Bahamas, The",
"Central African Rep." = "Central African Republic",
"Dem. Rep of Congo" = "Congo, Dem. Rep.",
"Gambia" = "Gambia, The",
"Rep. of Moldova" = "Moldova",
"Saint Lucia" = "St. Lucia",
"Egypt" = "Egypt, Arab Rep.",
"Venezuela" = "Venezuela, RB",
"Syria" = "Syrian Arab Republic",
"Solomon Isds" = "Solomon Islands",
"Br. Virgin Isds" = "British Virgin Islands",
"Cayman Isds" = "Cayman Islands",
"Dem. Rep. of the Congo" = "Congo, Dem. Rep.",
"Faeroe Isds" = "Faroe Islands",
"State of Palestine" = "West Bank and Gaza",
"Iran" = "Iran, Islamic Rep.",
"Curaçao" = "Curacao",
"Saint Maarten" = "Sint Maarten (Dutch part)",
"Marshall Isds" = "Marshall Islands",
"Turks and Caicos Isds" = "Turks and Caicos Islands",
"Saint Vincent and the Grenadines" = "St. Vincent and the Grenadines"))

```

```

df_stats_exports_combined <- inner_join(aluminum_76, stats_capab_data,
  ↪ by="Country_Code")

```

```

df_stats_exports_combined = df_stats_exports_combined[-c(4)]

```

```

iea_data <- mutate(iea_data, Country_Name_New =
  ↪ stringr::str_replace_all(Country_Name,"[:space:]", " "))

```

```

df_stats_exports_combined <- mutate(df_stats_exports_combined, Country_Name_New =
  ↪ stringr::str_replace_all(Country_Name.x,"[:space:]", " "))

```

Creating Final Data Set

```

df_all <- inner_join(df_stats_exports_combined, iea_data, by="Country_Name_New")

```

```

df_filtered=df_all[-c(9, 12, 13, 14, 15)]
colnames(df_filtered) = c("Country_Name", "Country_Code", "Trade_Value",
  ↪ "Statistical_Capability", "Income", "Region", "LDC", "Country_Name_New",
  ↪ "Population", "GDP", "Electricity_Consumption", "CO2_Emissions")

```

```
dim(df_filtered)
```

```
## [1] 108 12
```

```
head(df_filtered)
```

```
## Country_Name Country_Code Trade_Value Statistical_Capability
## 1 Albania ALB 37139568 75.38292
## 2 Algeria DZA 1834154 55.14917
## 3 Angola AGO 243849 54.94583
## 4 Azerbaijan AZE 34976688 68.14125
## 5 Argentina ARG 22199243 64.59583
## 6 Australia AUS 63475543 88.24167
## Income Region LDC Country_Name_New
## 1 Upper middle income Europe & Central Asia 0 Albania
## 2 Lower middle income Middle East & North Africa 0 Algeria
## 3 Lower middle income Sub-Saharan Africa 1 Angola
## 4 Upper middle income Europe & Central Asia 0 Azerbaijan
## 5 Upper middle income Latin America & Caribbean 0 Argentina
## 6 High income East Asia & Pacific 0 Australia
## Population GDP Electricity_Consumption CO2_Emissions
## 1 2.9 13.0 6.6 4.0
## 2 43.1 177.4 71.5 142.4
## 3 31.8 109.8 13.7 18.8
## 4 10.0 53.5 22.5 34.1
## 5 44.9 570.5 129.4 162.2
## 6 25.4 1338.6 251.1 380.7
```

```
table(df_filtered$LDC,df_filtered$Income)
```

```
##
## Low income Lower middle income Upper middle income High income
## 0 2 24 38 26
## 1 8 10 0 0
```

Calculated Columns

- Aluminum exports originally were in kg of aluminum, and needed to be divided by 1000 to get metric tonne
- Energy intensity is in kwh/tonne (15474 is the default value for Europe from the International Aluminium Institute for 2019)
- The conversion factor is in CO_2 /kwh (using the average across all countries to simulate default values)

- GDP is in billion 2015 USD
- Formulas:
 - Tonne Aluminum * (kwh/tonne) * (tCO_2 /kwh) = tonne CO_2 total
 - Percent Cost = Default Cost / ($GDP/10^7$)

```
df_filtered$Aluminum_Exports = df_filtered$Trade_Value/1.794

df_filtered$Conversion_Factor = (df_filtered$CO2_Emissions *
  ↪ 10^6)/(df_filtered$Electricity_Consumption * 10^9)

df_filtered$Default_CO2_Output = (df_filtered$Aluminum_Exports /1000) * 15474 *
  ↪ mean(df_filtered$Conversion_Factor)

df_filtered$Default_Cost = df_filtered$Default_CO2_Output * 86.284

df_filtered$Default_Percent_Cost =
  ↪ (df_filtered$Default_Cost/(df_filtered$GDP*10^7))
```

Visualizations

Statistical Capability Data

```
table(stats_capab_data$Region,stats_capab_data$Income)
```

```
##
##               Low income Lower middle income Upper middle income
## East Asia & Pacific                1                12                10
## Europe & Central Asia                1                 4                15
## Latin America & Caribbean            1                 4                20
## Middle East & North Africa           2                 6                 5
## North America                       0                 0                 0
## South Asia                          1                 6                 1
## Sub-Saharan Africa                  23                18                 5
##
##               High income
## East Asia & Pacific                15
## Europe & Central Asia               38
## Latin America & Caribbean           17
## Middle East & North Africa           8
## North America                       3
## South Asia                          0
## Sub-Saharan Africa                  2
```

```
table(stats_capab_data$LDC,stats_capab_data$Income)
```

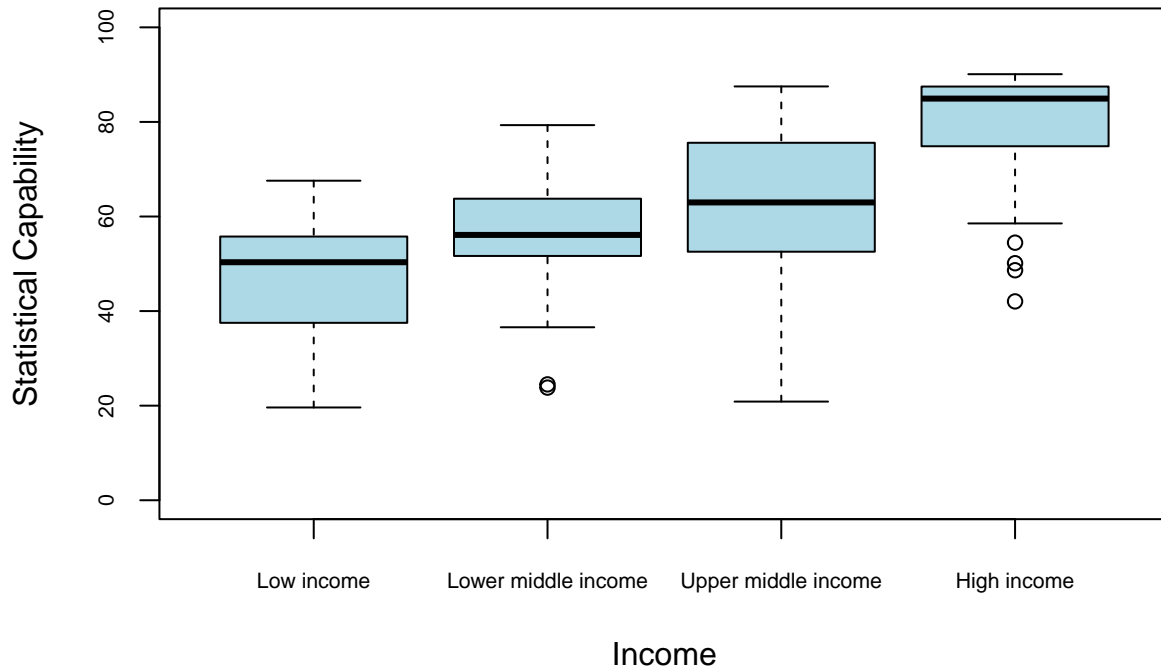
```
##
##      Low income Lower middle income Upper middle income High income
##    0          3          31          55          83
##    1          26          19          1          0
```

```
by(stats_capab_data$Statistical_Capability, stats_capab_data$Income, summary)
```

```
## stats_capab_data$Income: Low income
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##  19.62   37.77   50.33   46.96   55.23   67.57     3
## -----
## stats_capab_data$Income: Lower middle income
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##  23.82   51.66   56.11   56.68   63.76   79.32     1
## -----
## stats_capab_data$Income: Upper middle income
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##  20.86   52.73   62.98   61.76   75.49   87.51     8
## -----
## stats_capab_data$Income: High income
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##  42.05   74.86   84.93   78.53   87.48   90.09    32
```

```
boxplot(stats_capab_data$Statistical_Capability ~ stats_capab_data$Income, main =
  ↳ "Boxplots of Statistical Capability by Income Level", ylab="Statistical
  ↳ Capability", xlab="Income", cex.axis = 0.7, col = "light blue",
  ↳ ylim=c(0,100))
```

Boxplots of Statistical Capability by Income Level

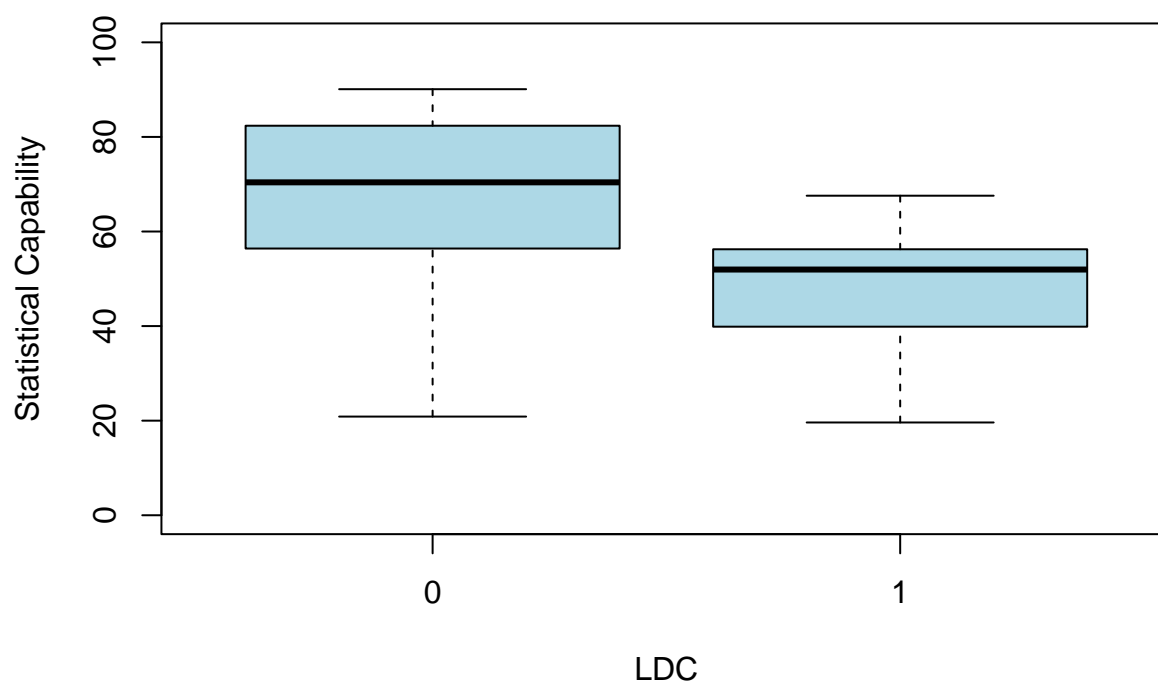


```
by(stats_capab_data$Statistical_Capability, stats_capab_data$LDC, summary)
```

```
## stats_capab_data$LDC: 0
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##   20.86  56.54   70.40   67.51  82.35   90.09    40
## -----
## stats_capab_data$LDC: 1
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##   19.62  40.32   51.95   48.98  56.23   67.57     4
```

```
boxplot(stats_capab_data$Statistical_Capability ~ stats_capab_data$LDC,
  ↪ ylab="Statistical Capability", xlab="LDC", col = "light blue", ylim=c(0,100),
  ↪ main = "Boxplots of Statistical Capability by LDC Status")
```


Boxplots of Statistical Capability by LDC Status



```
legendTitle = "Statistical Capability"
```

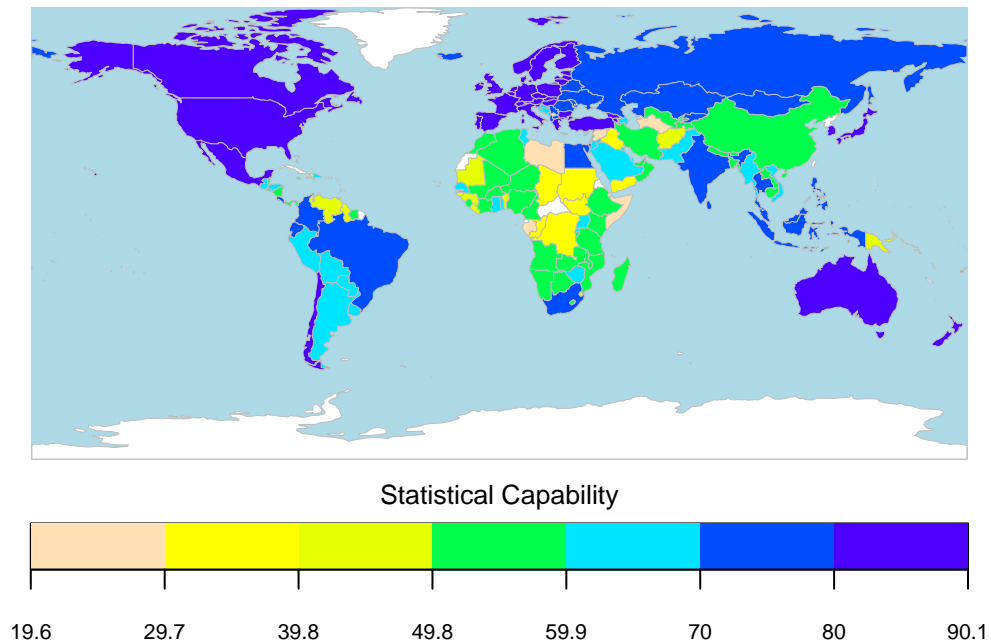
```
sdat <- joinCountryData2Map(stats_capab_data, joinCode="IS03",
  ↳ nameJoinColumn="Country_Code", verbose = TRUE)
```

```
## 215 codes from your data successfully matched countries in the map
## 3 codes from your data failed to match with a country code in the map
##      failedCodes
## [1,] "CHI"
## [2,] "GIB"
## [3,] "XKX"
## 28 codes from the map weren't represented in your data
```

```
mapParams = mapCountryData(sdat, nameColumnToPlot="Statistical_Capability",
  ↳ catMethod="fixedWidth", addLegend = FALSE, missingCountryCol = "white",
  ↳ mapTitle="Map of Statistical Capability by Country", oceanCol = "lightblue",
  ↳ colourPalette = "topo")
```

```
do.call(addMapLegend, c(mapParams, labelFontSize = 0.7, horizontal = TRUE,
  ↳ legendShrink = 0.928, legendLabels = "all", legendArgs=c(mtext(paste(""),
  ↳ side=3, adj=0.5, padj=0.4, cex=0.8),
  mtext(paste(legendTitle, sep=""),
  ↳ side=1, adj=0.5, padj=-0.8,
  ↳ cex=0.8)), digits=3))
```

Map of Statistical Capability by Country



```
sdat <- joinCountryData2Map(stats_capab_data, joinCode="IS03",  
  ↪ nameJoinColumn="Country_Code", verbose = TRUE)
```

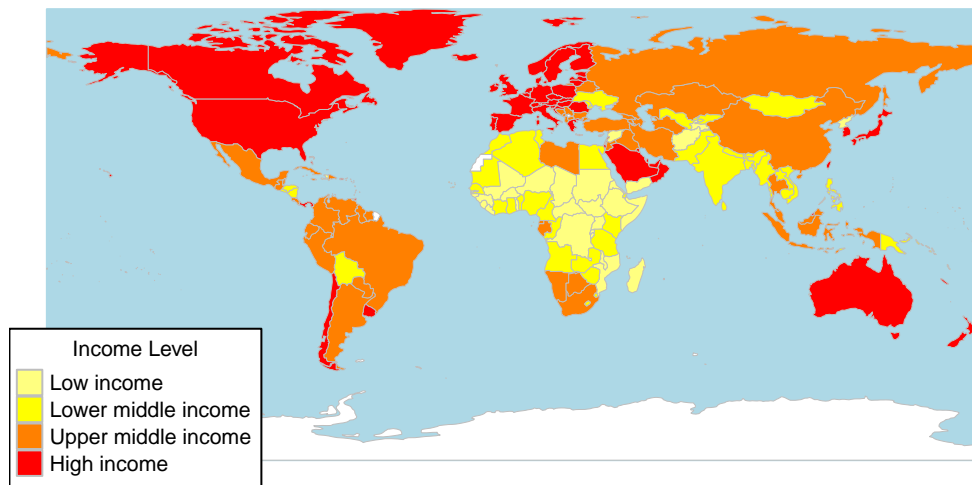
```
## 215 codes from your data successfully matched countries in the map  
## 3 codes from your data failed to match with a country code in the map  
##      failedCodes  
## [1,] "CHI"  
## [2,] "GIB"  
## [3,] "XKX"  
## 28 codes from the map weren't represented in your data
```

```
mapParams = mapCountryData(sdat, nameColumnToPlot="Income",  
  ↪ catMethod="Categorical", addLegend = FALSE, missingCountryCol = "white",  
  ↪ mapTitle="Map of Income Level by Country", oceanCol = "lightblue",  
  ↪ colourPalette = "heat")
```

```
## using catMethod='categorical' for non numeric data in mapCountryData
```

```
do.call( addMapLegendBoxes, c( mapParams, title = "Income Level", cex = 0.7))
```

Map of Income Level by Country



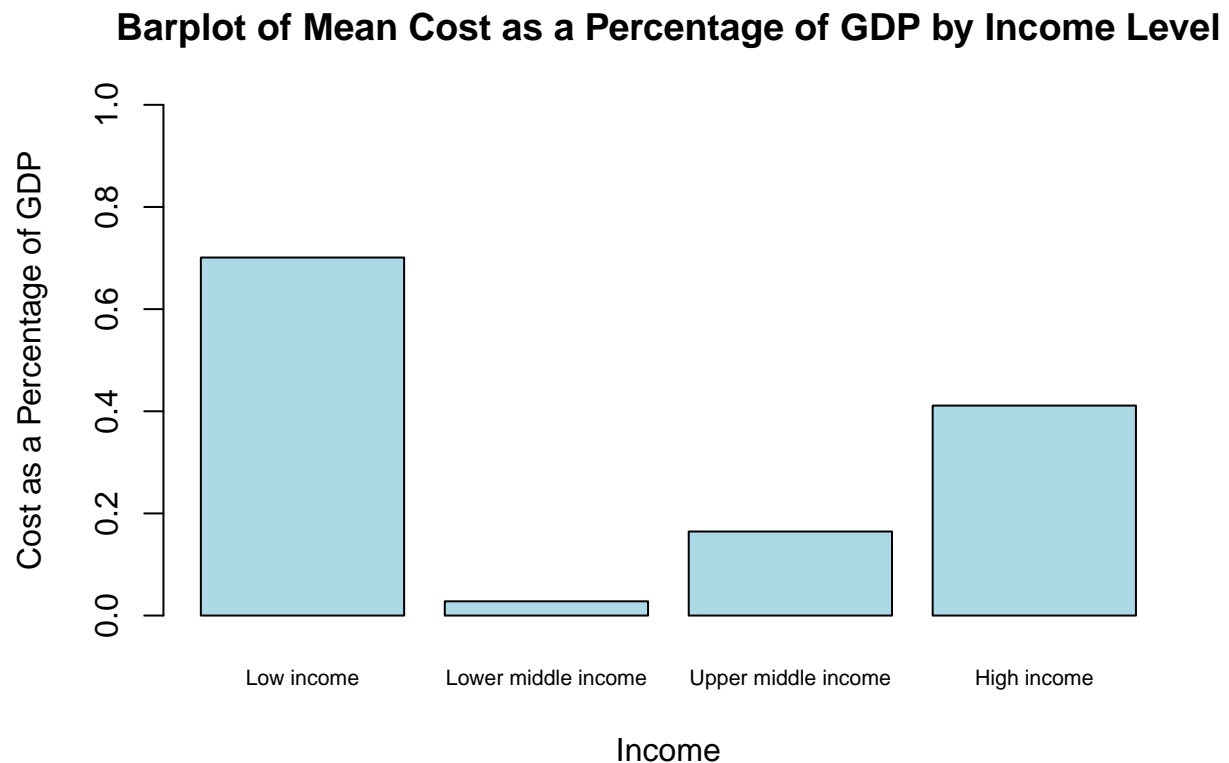
Cost Data

```
by(df_filtered$Default_Percent_Cost, df_filtered$Income, summary)
```

```
## df_filtered$Income: Low income
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.000032 0.000140 0.000437 0.701000 0.014752 6.911363
## -----
## df_filtered$Income: Lower middle income
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 2.250e-06 1.762e-04 1.422e-03 2.787e-02 1.805e-02 2.841e-01
## -----
## df_filtered$Income: Upper middle income
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.000056 0.001193 0.014673 0.164603 0.105081 1.765953
## -----
## df_filtered$Income: High income
```

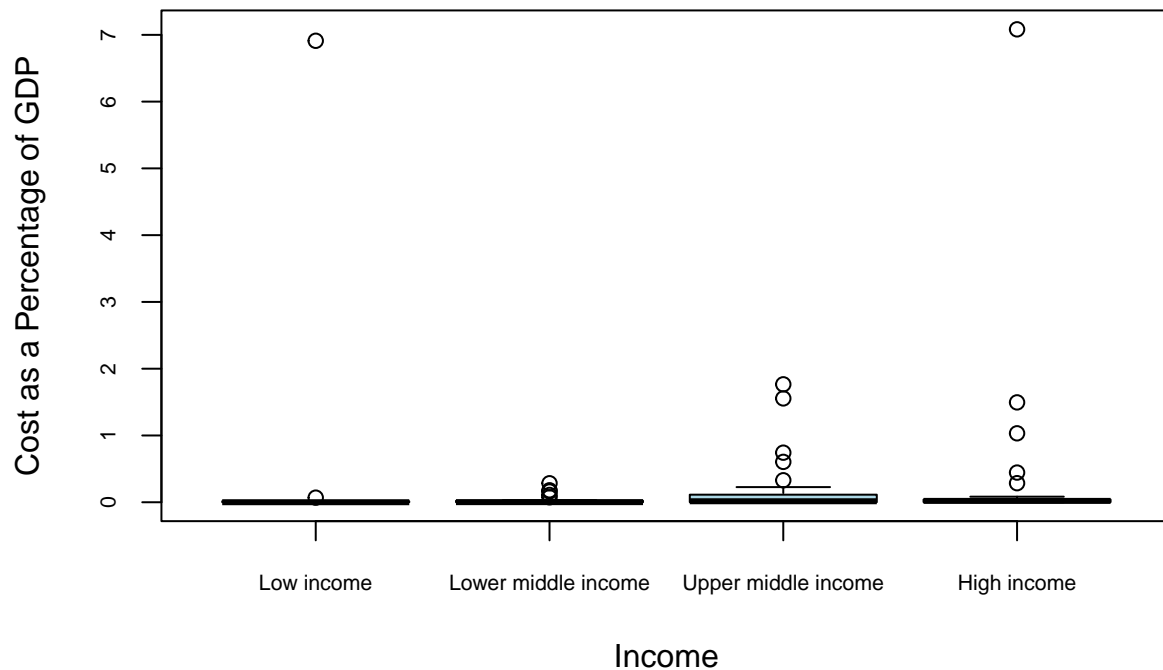
```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.000075 0.003079 0.020292 0.411137 0.038213 7.082377
```

```
barplot(by(df_filtered$Default_Percent_Cost, df_filtered$Income, mean), cex.names
  ↳ = 0.7, ylim = c(0,1), main = "Barplot of Mean Cost as a Percentage of GDP by
  ↳ Income Level", ylab = "Cost as a Percentage of GDP", xlab = "Income", col =
  ↳ "lightblue")
```



```
boxplot(df_filtered$Default_Percent_Cost ~ df_filtered$Income, cex.axis = 0.7,
  ↳ ylab = "Cost as a Percentage of GDP", xlab = "Income", main = "Boxplots of
  ↳ Cost as a Percentage of GDP by Income Level", col = "lightblue")
```

Boxplots of Cost as a Percentage of GDP by Income Level

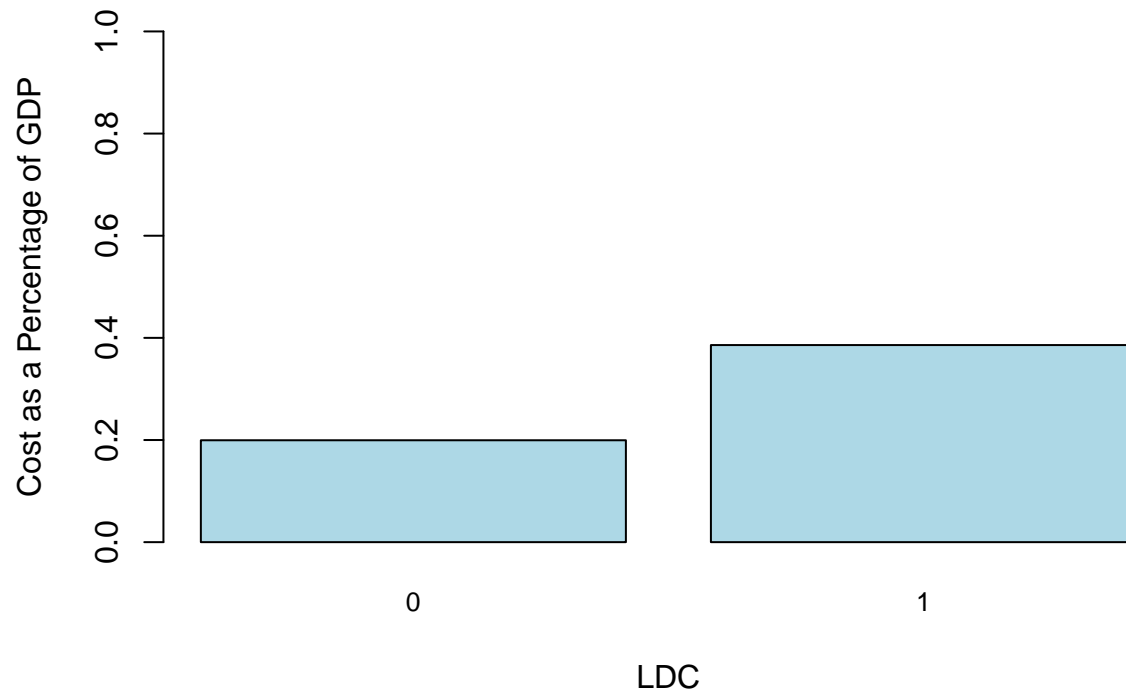


```
by(df_filtered$Default_Percent_Cost, df_filtered$LDC, summary)
```

```
## df_filtered$LDC: 0
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.000006 0.001334 0.009919 0.199508 0.072774 7.082377
## -----
## df_filtered$LDC: 1
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.000002 0.000028 0.000125 0.385901 0.000491 6.911363
```

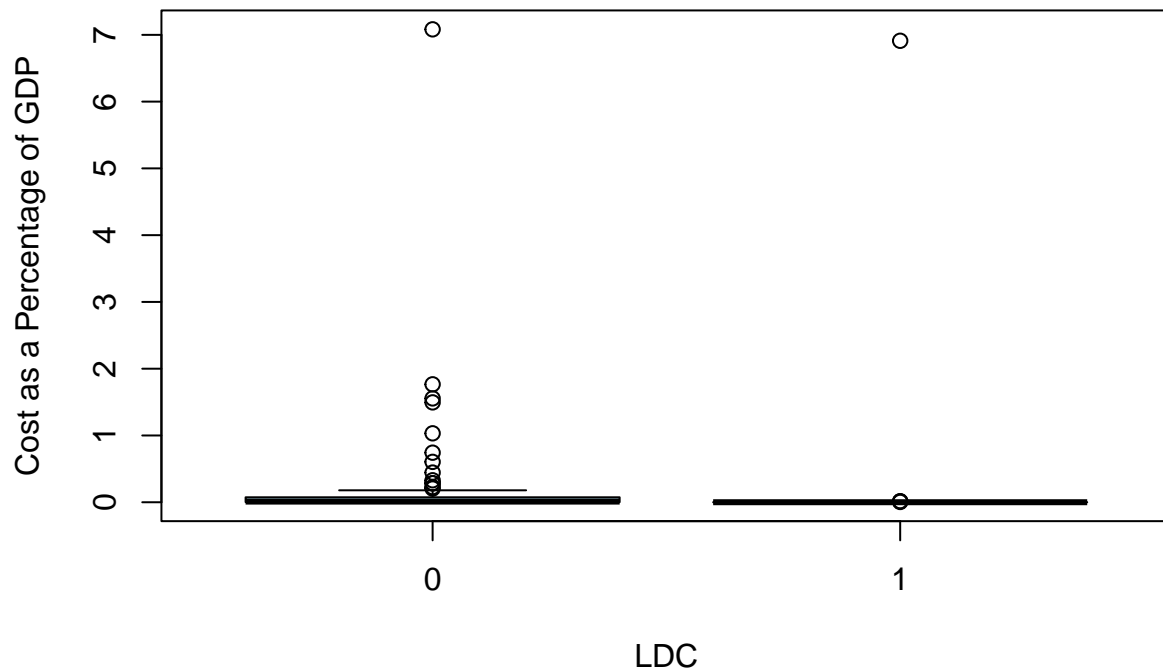
```
barplot(by(df_filtered$Default_Percent_Cost, df_filtered$LDC, mean), cex.names =
  ↪ 0.83, ylim = c(0,1), main = "Barplot of Mean Cost as a Percentage of GDP by
  ↪ LDC Status", xlab = "LDC", ylab = "Cost as a Percentage of GDP", col =
  ↪ "lightblue")
```

Barplot of Mean Cost as a Percentage of GDP by LDC Status



```
boxplot(df_filtered$Default_Percent_Cost ~ df_filtered$LDC, ylab = "Cost as a  
↳ Percentage of GDP", xlab = "LDC", main = "Boxplots of Cost as a Percentage of  
↳ GDP by LDC Status", col = "lightblue")
```

Boxplots of Cost as a Percentage of GDP by LDC Status



Scatterplots

Emissions vs. GDP

```
par(mfrow = c(1, 2))

cols = c("mediumpurple2", "lightblue", "darkseagreen", "lightpink")

plot(log(CO2_Emissions) ~ log(GDP), data=df_filtered, xlab = "GDP (billion 2015
↪ USD)", ylab = expression("CO"[2] * " Emissions (Mt)"), col =
↪ cols[df_filtered$Income], pch=16, cex.lab = 0.8, cex.axis = 0.8, cex.main =
↪ 1)

legend("bottomright", legend=levels(df_filtered$Income), fill=cols, title="Income
↪ Level", cex = 0.6)

abline(lm(log(df_filtered$CO2_Emissions)~log(df_filtered$GDP)), lty="dashed",
↪ col="darkgray")
```

```

cols = c("paleturquoise2","lightseagreen")

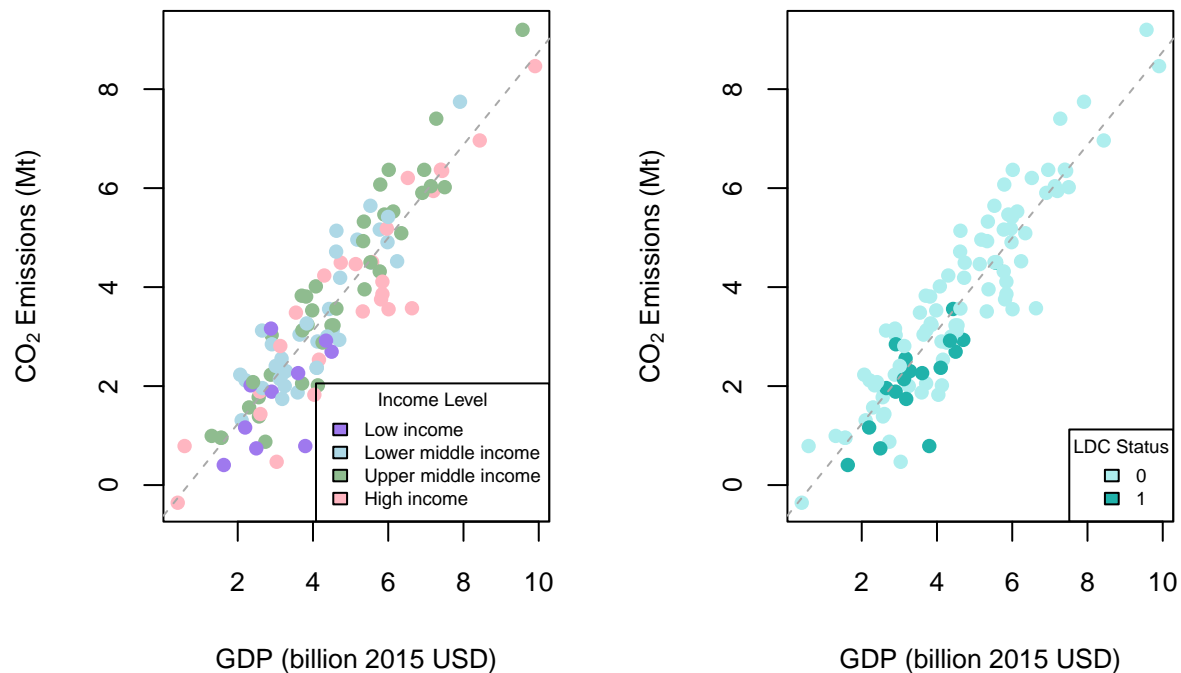
plot(log(CO2_Emissions) ~ log(GDP), data=df_filtered,xlab = "GDP (billion 2015
↳ USD)", ylab = expression("CO"[2] * " Emissions (Mt)"), col =
↳ cols[as.factor(df_filtered$LDC)], pch=16, cex.lab = 0.8, cex.axis = 0.8,
↳ cex.main = 1)

legend("bottomright", legend=levels(as.factor(df_filtered$LDC)), fill=cols,
↳ title="LDC Status", cex = 0.6)
abline(lm(log(df_filtered$CO2_Emissions) ~ log(df_filtered$GDP)), lty="dashed",
↳ col="darkgray")

mtext(expression(~bold("CO"[2] * " Emissions vs. GDP (log scale)")),
↳
    side = 3,
    line = - 2,
    outer = TRUE,
    font=2)

```

CO₂ Emissions vs. GDP (log scale)



Cost vs. Statistical Capability

```
summary(lm(log(CO2_Emissions) ~ log(GDP), data=df_filtered))
```

```
##
## Call:
## lm(formula = log(CO2_Emissions) ~ log(GDP), data = df_filtered)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.1362 -0.4683  0.0093  0.4828  1.4358
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.64252    0.18627  -3.449 0.000807 ***
## log(GDP)     0.93984    0.03892  24.149 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7393 on 106 degrees of freedom
## Multiple R-squared:  0.8462, Adjusted R-squared:  0.8447
## F-statistic: 583.2 on 1 and 106 DF,  p-value: < 2.2e-16
```

```
cor(log(df_filtered$CO2_Emissions), log(df_filtered$GDP))
```

```
## [1] 0.9198871
```

```
par(mfrow = c(1, 2))
```

```
cols = c("mediumpurple2", "lightblue", "darkseagreen", "lightpink")
```

```
plot(log(Default_Percent_Cost)~log(Statistical_Capability), data=df_filtered, col
↪ = cols[as.factor(df_filtered$Income)], pch=16, cex.lab = 0.8, cex.axis = 0.8,
↪ cex.main = 0.8, xlab = "Statistical Capability", ylab = "Cost as a Percentage
↪ of GDP")
```

```
legend("topleft", legend=levels(df_filtered$Income), fill=cols, title="Income
↪ Level", cex = 0.6)
```

```
cols = c("paleturquoise2", "lightseagreen")
```

```

plot(log(Default_Percent_Cost)~log(Statistical_Capability), data=df_filtered, col
↳ = cols[as.factor(df_filtered$LDC)], pch=16, cex.lab = 0.8, cex.axis = 0.8,
↳ cex.main = 1, xlab = "Statistical Capability", ylab = "Cost as a Percentage
↳ of GDP")

legend("topleft", legend=levels(as.factor(df_filtered$LDC)), fill=cols,
↳ title="LDC Status", cex = 0.6)

mtext("Cost as a Percentage of GDP vs. Statistical Capability (log scale)",
↳
side = 3,
line = - 2,
outer = TRUE,
font=2)

```

Cost as a Percentage of GDP vs. Statistical Capability (log scale)

