## Models\_AdrianBrenner\_IB\_ResearchDesign

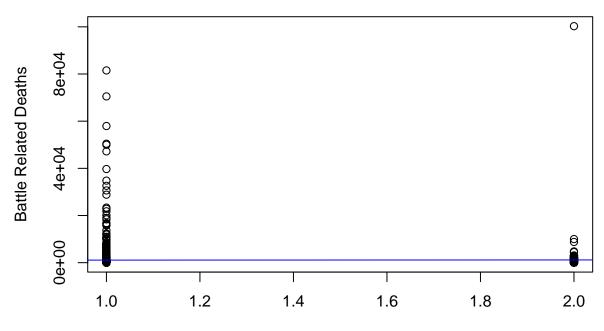
## Adrian Brenner

## 2023-12-13

```
# This is the code for importing and cleaning the chosen Datasets.
if (!requireNamespace("readxl", quietly = TRUE)) {
  install.packages("readxl")
library(readxl)
## Warning: Paket 'readxl' wurde unter R Version 4.3.2 erstellt
xlsx_file <- "C:/Users/AD/Desktop/data/BattleDeaths_v23_1.xlsx"</pre>
data <- read xlsx(xlsx file)
# In the first cleaning step i removed all conflict, not taking place in region = 4 (Africa)
new_data <- data[data$region == 4, ]</pre>
# The second step involves removing all cases where there is only one belligerent on the second side, t
cleaned_data <- new_data[complete.cases(new_data$gwno_b_2nd), ]</pre>
cleaned_data_2 <- new_data[complete.cases(new_data$gwno_a_2nd), ]</pre>
# The third step is to combine both data sets into one and remove the duplicates.
library(dplyr)
## Attache Paket: 'dplyr'
## Die folgenden Objekte sind maskiert von 'package:stats':
##
##
       filter, lag
## Die folgenden Objekte sind maskiert von 'package:base':
       intersect, setdiff, setequal, union
# Assuming 'cleaned_data' and 'cleaned_data_2' are your datasets
# Combine the datasets and remove duplicates
combined_data <- bind_rows(cleaned_data, cleaned_data_2) %>%
 distinct()
\#The\ final\ Dataset\ cleaned\ for\ the\ linear\ regression\ model\ of\ 4.2.1\ now\ contains\ 238\ distinct\ cases.\ I
```

```
#TO create the opposite Dataset (the one with no intervention) I just subtract the combined_data from t
rest_data <- anti_join(data, combined_data)</pre>
## Joining with `by = join_by(conflict_id, dyad_id, location_inc, side_a,
## side_a_id, side_a_2nd, side_b, side_b_id, side_b_2nd, incompatibility,
## territory_name, year, bd_best, bd_low, bd_high, type_of_conflict,
## battle_location, gwno_a, gwno_a_2nd, gwno_b, gwno_b_2nd, gwno_loc, gwno_battle,
## region, version)`
# In this next step I will create the regression model to include the battle related Deaths count of th
# Assuming 'rest data' and 'combined data' are your datasets
# Assuming 'Battle_Related_Deaths' is the dependent variable
# Create a binary predictor variable
predictor_variable <- rep(c(0, 1), c(nrow(rest_data), nrow(combined_data)))</pre>
# Combine the datasets
all_data <- bind_rows(rest_data, combined_data)</pre>
# Create a linear regression model
model <- lm(bd_best ~ as.factor(predictor_variable), data = all_data)</pre>
# Plot the linear regression model
plot(all_data$bd_best ~ as.factor(predictor_variable),
     xlab = "Foreign State Involvement (Binary)",
     ylab = "Battle Related Deaths",
    main = "Linear Regression Model")
# Add the regression line to the plot
abline(model, col = "blue")
```

## **Linear Regression Model**



Foreign State Involvement (Binary)

```
# I now extrace its coefficients, slope and intercept.
# Get coefficients of the linear regression model
coefficients <- coef(model)</pre>
# Extract slope and intercept
slope <- coefficients[2] # Coefficient for the predictor variable</pre>
intercept <- coefficients[1] # Intercept term</pre>
# Print the results
cat("Slope:", slope, "\n")
## Slope: 73.76704
cat("Intercept:", intercept, "\n")
## Intercept: 1013.859
summary(model)
##
## Call:
## lm(formula = bd_best ~ as.factor(predictor_variable), data = all_data)
## Residuals:
##
      Min
              1Q Median
                             ЗQ
                                   Max
    -1062
##
            -975
                    -898
                           -545
                                 99194
##
```

```
## Coefficients:
##
                                  Estimate Std. Error t value Pr(>|t|)
                                               121.60
                                                                 <2e-16 ***
## (Intercept)
                                   1013.86
                                                        8.337
## as.factor(predictor_variable)1
                                     73.77
                                               338.85
                                                        0.218
                                                                  0.828
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 4879 on 1846 degrees of freedom
## Multiple R-squared: 2.567e-05, Adjusted R-squared: -0.000516
## F-statistic: 0.04739 on 1 and 1846 DF, p-value: 0.8277
# T-test for the coefficient of the predictor variable
t_test <- summary(model)$coefficients["as.factor(predictor_variable)1", "t value"]
# Print the t-test result
cat("T-test for the coefficient of the predictor variable:", t_test, "\n")
## T-test for the coefficient of the predictor variable: 0.2176953
# Extracting the residual standard error
residual_standard_error <- summary(model)$sigma</pre>
# Degrees of freedom
df <- length(model$coefficients) - 1 # Subtract 1 for the intercept term
# Chi-square test statistic
chi_square <- (t_test^2) / residual_standard_error^2</pre>
# P-value for chi-square test
p_value_chi_square <- 1 - pchisq(chi_square, df)</pre>
# Print the chi-square test results
cat("Chi-square test statistic:", chi_square, "\n")
## Chi-square test statistic: 1.990538e-09
cat("Degrees of freedom:", df, "\n")
## Degrees of freedom: 1
cat("P-value for chi-square test:", p_value_chi_square, "\n")
## P-value for chi-square test: 0.9999644
Adrian Brenner (12422308) for GSI LMU - IB Research Design (15103) - Dr. Oliver Pamp
```