

Data Analysis Practice

Michael V Cumbo

February 1, 2024

1 About this Data

Disaster Declarations Summaries is a summarized dataset describing all federally declared disasters. This dataset lists all official FEMA Disaster Declarations, beginning with the first disaster declaration in 1953 and features all three disaster declaration types: major disaster, emergency, and fire management assistance. The dataset includes declared recovery programs and geographic areas (county not available before 1964; Fire Management records are considered partial due to historical nature of the dataset).

```
library(tidyverse)
library(lubridate)
library(RSQLite)
library(DBI)
library(ggplot2)
library(dplyr)
library(forcats)
library(GGally)
library(stringr)
library(magrittr)

setwd("~/workbook")
con <- dbConnect(RSQLite::SQLite(), "Disaster_Data.db")
dbListTables(con)

## [1] "US_Declarations_2023"      "sqlite_master"
## [3] "us_disaster_declarations"

declarations_2023 <- as_tibble(dbGetQuery(
  con,
  "SELECT
  disasterNumber,
  state,
  declarationType,
  incidentType,
  declarationDate
  FROM US_Declarations_2023
  ORDER BY declarationDate;"
))
dbDisconnect(con)
```

```

# build an arrangement paired with mutate in
# DPLYR library than pass it to ggplot and build the graph
# this reorders the factors in the
# graph so its easier to break down and look at
v1 <- declarations_2023 %>%
  select("state", "incidentType", "Year") %>%
  group_by(incidentType) %>%
  summarise(n = n()) %>%
  arrange(n) %>%
  mutate(incidentType = factor(incidentType, levels = unique(incidentType))) %>%
  ggplot(aes(x = incidentType, y = n)) +
  geom_col() +
  theme_gray() +
  theme(axis.text.x = element_text(angle = 45, vjust = 0.5, hjust = 0.5))
print(v1)

```

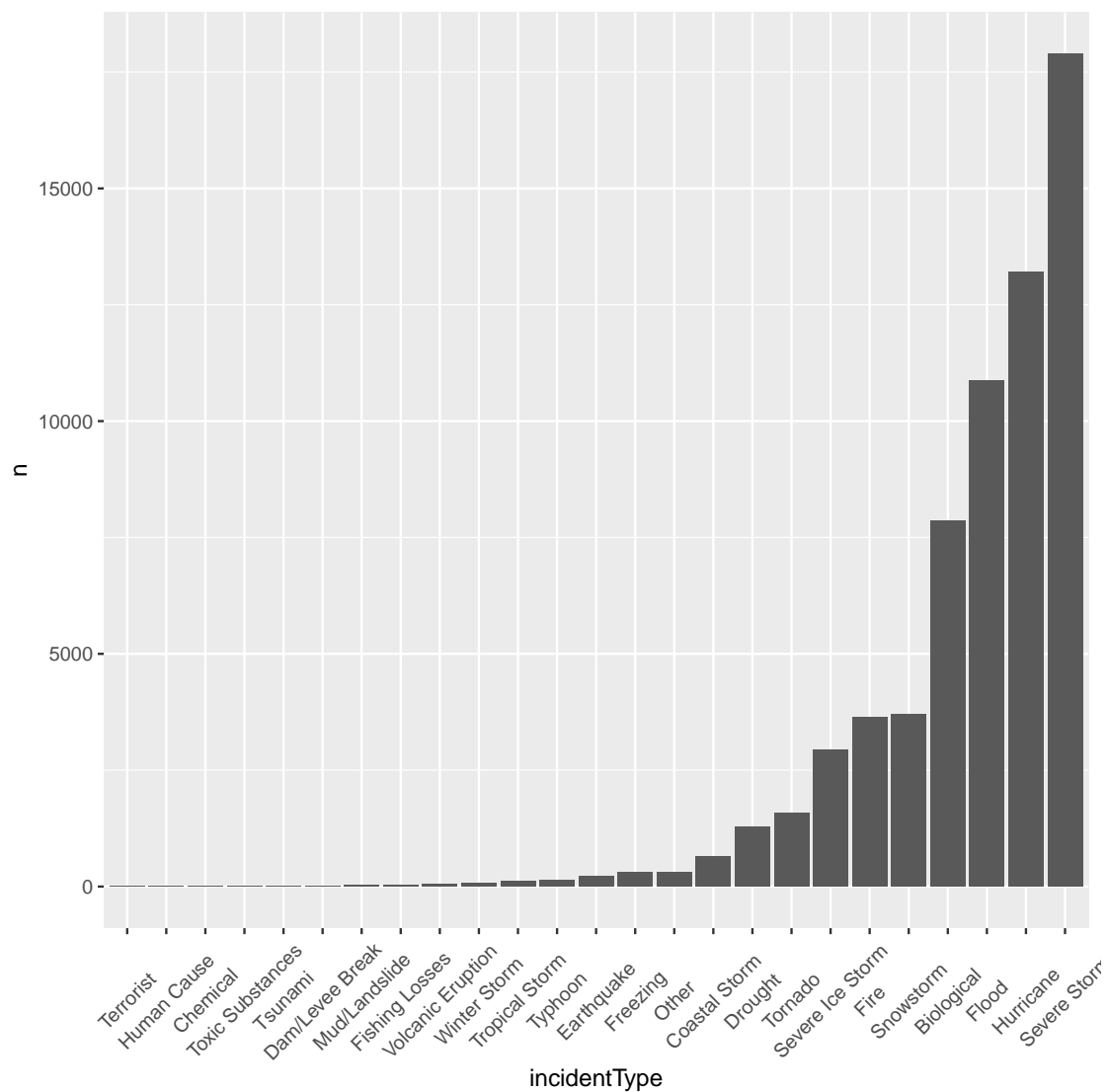
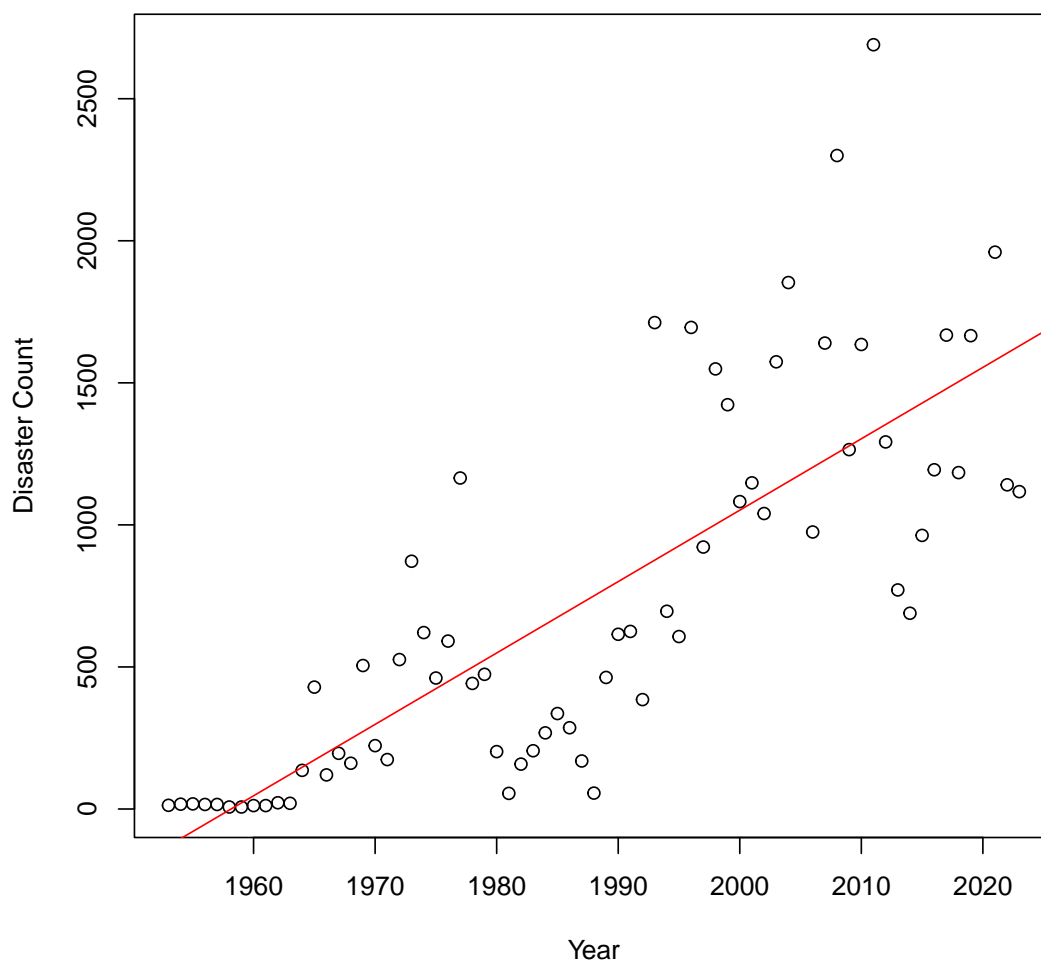


Figure 1: histogram of the total number of disasters declared by the United States

```
##      1      2      3      4      5      6      7      8
## 1654.88 1680.01 1705.14 1730.27 1755.40 1780.53 1805.66 1830.79
##      9     10     11     12
## 1855.92 1881.05 1906.18 1931.32
##
## Call:
## lm(formula = DisasterCount ~ Year, data = annual_disasters)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -714.57 -312.30  -49.99  177.93 1361.83
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -49209.962   4821.925  -10.21 2.80e-15 ***
## Year          25.131      2.426    10.36 1.51e-15 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 409.4 on 67 degrees of freedom
## Multiple R-squared:  0.6156, Adjusted R-squared:  0.6098
## F-statistic: 107.3 on 1 and 67 DF,  p-value: 1.514e-15
```

Disaster Count Over Years



Model Summary

The summary of the linear model offers key insights:

Coefficients:

- The intercept is -49209.962 , implying the model's prediction for *DisasterCount* when *Year* is 0, which is not applicable in this context.
- The slope coefficient for *Year* is 25.131. This indicates an annual increase of approximately 25.131 in *DisasterCount*, as per the model.
- These predictions indicate an upward trend in *DisasterCount* over the years.

Statistical Significance:

Both the intercept and the slope demonstrate statistical significance with a p-value < 0.001 .

Model Fit:

- The R-squared value is 0.6156, signifying that approximately 61.56% of the variability in *DisasterCount* is explained by the year. However, a significant portion of variability remains unexplained.
- The Residual Standard Error (RSE) is 409.4, indicating the average deviation of data points from the fitted line.

Residuals:

The residuals range from -714.57 to 1361.83 , suggesting variability around the regression line.

Interpretation and Considerations

- The model indicates a significant upward trend in disaster counts over the years.
- The presence of significant residuals and a R-squared value of 0.6156 implies that, while there is a discernible trend, other unaccounted factors might also be influencing *DisasterCount*.
- Predictions for future years should be approached with caution due to the simplicity of the model and its exclusion of other potential predictive factors.
- When interpreting these results and making future decisions, the limitations of a simple linear regression model and the impact of external factors not included in the model should be considered.