

Placement

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Note: the github repo for this work is: <https://github.com/sheny2/Placement>

Q1

Task 1: Read the data Scrape the wikipedia page on natural disasters

```
url <- "https://en.wikipedia.org/wiki/List_of_natural_disasters_by_death_toll"

webpage <- read_html(url)

tables <- webpage %>% html_nodes("table.wikitable")

# 20th and 21st century data are the 2nd and 3rd table of this page
table_20th <- tables[[2]] %>% html_table(fill = TRUE)
table_21st <- tables[[3]] %>% html_table(fill = TRUE)

disasters <- rbind(table_20th, table_21st) %>% as_tibble()

head(disasters)
```

```
## # A tibble: 6 x 6
##   Year 'Death toll' Event 'Countries affected' Type Date
##   <int> <chr>      <chr>      <chr>      <chr> <chr>
## 1  1900 6,000-12,000 1900 Galveston hurricane United States Trop~ Sept~
## 2  1901 9,500      1901 eastern United State~ United States Heat~ June~
## 3  1902 29,000      1902 eruption of Mount Pe~ Martinique Volc~ Apri~
## 4  1903 3,500      1903 Manzikert earthquake Turkey Eart~ Apri~
## 5  1904 400        1904 Sichuan earthquake China Eart~ Augu~
## 6  1905 20,000+    1905 Kangra earthquake India Eart~ Apri~
```

Task 2: Clean the data Convert the death toll to numbers using the midpoints when a range is given and the bound when an upper or lower bound is given.

```
disasters_clean = disasters
# remove unnecessary footnote, words and symbols
disasters_clean$`Death toll` = disasters$`Death toll` %>%
  str_remove_all(",") %>%
  str_remove_all("\\[.?!\\]") %>%
  str_remove_all("\\(.?!\\)") %>%
  str_remove_all("[a-zA-Z]") %>%
  str_trim()
```

```
# which entries need special attention to clean
index_to_clean = is.na(as.numeric(gsub(",", "", disasters_clean$`Death toll`)))
disasters$`Death toll`[index_to_clean]
```

```
## [1] "6,000-12,000"      "20,000+"           "12,000-15,000"
## [4] "75,000-82,000"     "6,000-8,000"       "50,000-220,000"
## [7] "942-1,900"         "29,978-32,610"     "2,000-10,000"
## [10] "258,707-273,407"   "50,000-100,000+"    "105,385-142,800"
## [13] "4,112+"            "3,257-3,800"       "2,000-8,000"
## [16] "422,499-4,000,000" "3,103+"            "6,865-9,300"
## [19] "10,700-12,000"     "5,000+"            "715+"
## [22] "32,700-32,968"     "2,824-5,000"       "10,000-110,000"
## [25] "1,023+"            "300,000-500,000"   "2,175-2,204"
## [28] "8,210+"            "26,000-240,000"    "242,419-655,000"
## [31] "10,000-50,000"     "15,000-25,000"     "2,633-5,000"
## [34] "25,000-50,000"     "35,000-45,000"     "17,126-18,373"
## [37] "700-800"           "13,805-20,023"     "86,000-87,351"
## [40] "5,749-5,778"       "100,000-316,000"   "3,951+"
## [43] "59,259-62,013"     "320-600(estimate)"
```

```
# use regular expression here
convert_death_toll <- function(toll) {
  if (str_detect(toll, "\\d+\\s*-\\s*\\d+")) {
    nums <- str_extract_all(toll, "\\d+")[1] %>% as.numeric()
    return(mean(nums))
  } else if (str_detect(toll, "\\d+\\s*-\\s*\\d+")) {
    nums <- str_extract_all(toll, "\\d+")[1] %>% as.numeric()
    return(mean(nums))
  } else if (str_detect(toll, ">\\s*\\d+")) {
    return(as.numeric(str_extract(toll, "\\d+")))
  } else if (str_detect(toll, "\\d+")) {
    return(as.numeric(str_extract(toll, "\\d+")))
  }
}

death_toll_clean = round(sapply(disasters_clean$`Death toll`, convert_death_toll))
death_toll_clean[index_to_clean] # check the results of those irregular ones
```

```
##      6000-12000      20000+      12000-15000      75000-82000      6000-8000
##           9000           20000           13500           78500           7000
##      50000-220000      942-1900      29978-32610      2000-10000      258707-273407
##           135000           1421           31294           6000           266057
##      50000-100000+      105385-142800           4112+           3257-3800           2000-8000
##           75000           124092           4112           3528           5000
##      422499-4000000           3103+           6865-9300      10700-12000           5000+
##           2211250           3103           8082           11350           5000
##           715+      32700-32968           2824-5000      10000-110000           1023+
##           715           32834           3912           60000           1023
##      300000-500000           2175-2204           8210+      26000-240000      242419-655000
##           400000           2190           8210           133000           448710
##           10000-50000      15000-25000           2633-5000      25000-50000           35000-45000
##           30000           20000           3816           37500           40000
##           17126-18373           700-800           13805-20023           86000-87351           5749-5778
```

```
##           17750           750           16914           86676           5764
## 100000-316000      3951+    59259-62013      320-600
##           208000           3951           60636           460
```

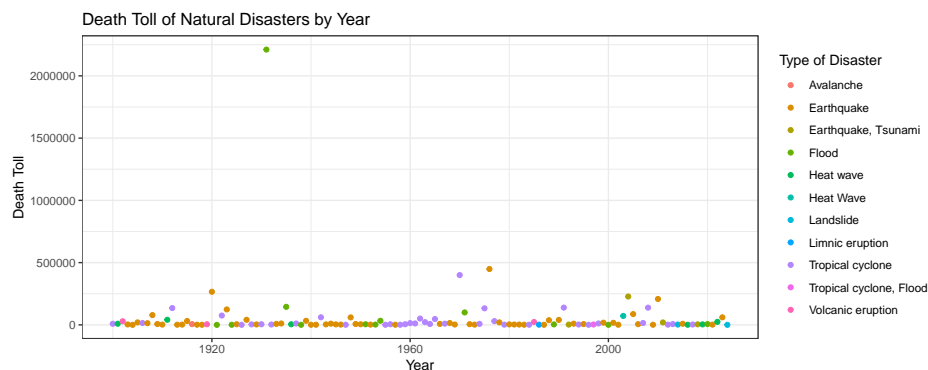
```
disasters_clean$`Death toll` = death_toll_clean
disasters_clean
```

```
## # A tibble: 125 x 6
##   Year 'Death toll' Event 'Countries affected' Type Date
##   <int>      <dbl> <chr>      <chr>      <chr> <chr>
## 1 1900          9000 1900 Galveston hurricane United States Trop~ Sept~
## 2 1901          9500 1901 eastern United Stat~ United States Heat~ June~
## 3 1902         29000 1902 eruption of Mount P~ Martinique Volc~ Apri~
## 4 1903          3500 1903 Manzikert earthquake Turkey Eart~ Apri~
## 5 1904           400 1904 Sichuan earthquake China Eart~ Augu~
## 6 1905         20000 1905 Kangra earthquake India Eart~ Apri~
## 7 1906         15000 1906 Hong Kong typhoon Hong Kong,China Trop~ Sept~
## 8 1907         13500 1907 Qaratog earthquake Uzbekistan Eart~ Octo~
## 9 1908         78500 1908 Messina earthquake Italy Eart~ Dece~
## 10 1909          7000 1909 Borujerd earthquake Iran Eart~ Janu~
## # i 115 more rows
```

Task 3: Plot the data

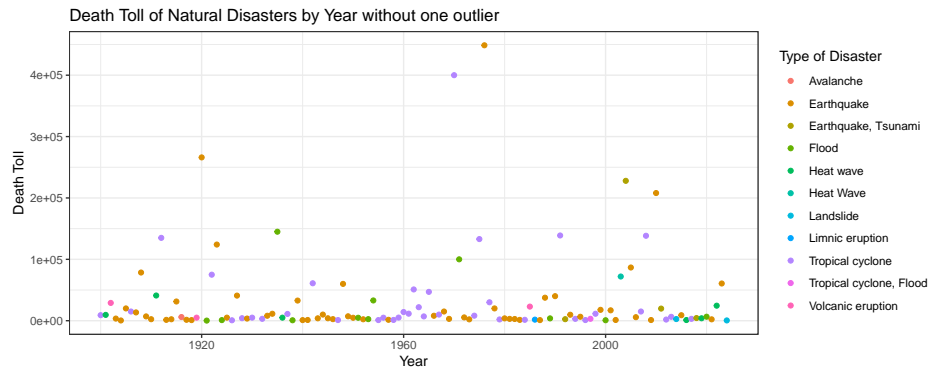
Disasters have happened frequently throughout the years. While at one time floods caused one of the highest death tolls, it should be noted that earthquakes are often the leading ones causing fatalities. Both topical cyclones and earthquakes seem to be the most frequent and deadly disasters during the studied period.

```
ggplot(disasters_clean, aes(x = Year, y = `Death toll`, color = Type)) +
  geom_point() +
  labs(title = "Death Toll of Natural Disasters by Year",
       x = "Year",
       y = "Death Toll",
       color = "Type of Disaster")
```



```
ggplot(disasters_clean %>% filter(Event != "1931 China floods"),
  aes(x = Year, y = `Death toll`, color = Type)) +
  geom_point() +
  labs(title = "Death Toll of Natural Disasters by Year without one outlier",
       x = "Year",
```

```
y = "Death Toll",
color = "Type of Disaster")
```



Q2

First we derive the form of gradient by computing the derivative w.r.t b:

$$\frac{\partial ||y - bx||^2}{\partial b} = 2 \left(\frac{\partial \sum_i^n (y_i - bx_i)}{\partial b} \right) = 2 \left(\frac{-\sum_i^n x_i (y_i - bx_i)}{n} \right)$$

```
# write the gradient descent function
gradient_descent <- function(x, y, learning_rate, num_iter) {
  b <- 0 # start at 0, or somewhere
  n <- length(x)

  for (i in 1:num_iter) {
    gradient <- -2 * sum(x * (y - b * x)) / n # compute gradient
    b <- b - learning_rate * gradient # update here
  }
  return(b)
}

# Test the function using randomly generated normal vectors
set.seed(8848)
n <- 100
x <- rnorm(n)
b <- 5
y <- b * x + rnorm(n)

S <- 1000 # long enough

b_grad <- gradient_descent(x, y, learning_rate = 0.05, num_iter = S) # choose e to be 0.05

# Compare
b_solution <- x %*% y / (norm(x, type="2"))^2
b_solution %>% as.vector()
```

```
## [1] 5.015827
```

```
b_grad # success!
```

```
## [1] 5.015827
```

```
# To test different learning rates
test_learning_rates <- function(x, y, b, learning_rates, num_iter) {
  results <- data.frame(learning_rate = c(), b_est = c())

  for (e in learning_rates) {
    b_est <- gradient_descent(x, y, e, num_iter)
    results <- rbind(results, data.frame(learning_rate = e, b_est = b_est))
  }

  return(results)
}
```

Findings: The algorithm can perform well as long as we choose a reasonable value for the learning rate.

We can see from below that the learning rate should not be too small. A too-small step size can make it inefficient for the algorithm to explore and reach convergence or get stuck at a local min instead of global mean, thus unable to find the correct solution.

Meanwhile, we also do not want the step size to be too large at each update, which could cause our estimate to oscillate too much and miss the optimal point as well.

```
learning_rates <- seq(0.001, 0.1, by = 0.001)
results <- test_learning_rates(x, y, b_true, learning_rates, num_iter = S)

# Plot the results
ggplot(results, aes(x = learning_rate, y = b_est)) +
  geom_line() +
  labs(title = "Estimate vs Learning Rate", x = "Learning Rate", y = "Estimate")
```

