

EDA Proposal Statistical

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```
#libraries
```

```
library(lubridate)
```

Warning: package 'lubridate' was built under R version 4.4.2

```
library(dplyr)
```

Warning: package 'dplyr' was built under R version 4.4.2

```
library(ggplot2)
```

Warning: package 'ggplot2' was built under R version 4.4.2

```
library(GGally)
```

Warning: package 'GGally' was built under R version 4.4.2

```
tuesdata <- tidytuesdayR::tt_load('2023-04-25')  
tuesdata <- tidytuesdayR::tt_load(2023, week = 17)
```

```
winners <- tuesdata$winners  
london_marathon <- tuesdata$london_marathon
```

```

winners$Time.Seconds <- period_to_seconds(hms(winners$Time))

# str(winners)

# Factoring the variables
winners$Category <- factor(winners$Category)
winners$Athlete <- factor(winners$Athlete)
winners$Nationality <- factor(winners$Nationality)

# str(london_marathon)

# Handling the NA in Raised
london_marathon$Raised[is.na(london_marathon$Raised)] = 0
london_marathon <- london_marathon[rowSums(is.na(london_marathon)) <= 2,]

```

```
summary(winners)
```

	Category	Year	Athlete
Men	:43	Min. :1981	David Weir : 8
Wheelchair Men	:39	1st Qu.:1992	Tanni Grey-Thompson : 6
Wheelchair Women	:39	Median :2002	David Holding : 4
Women	:42	Mean :2002	Eliud Kipchoge : 4
		3rd Qu.:2012	Francesca Porcellato: 4
		Max. :2022	Ingrid Kristiansen : 4
			(Other) :133
	Nationality	Time	Time.Seconds
United Kingdom	:44	Length:163	Min. : 5187
Kenya	:30	Class1:hms	1st Qu.: 6550
United States	:11	Class2:difftime	Median : 7675
Switzerland	:10	Mode :numeric	Mean : 7608
Ethiopia	: 9		3rd Qu.: 8418
Norway	: 7		Max. :16143
(Other)	:52		

```
summary(london_marathon)
```

Date	Year	Applicants	Accepted
Min. :1981-03-29	Min. :1981	Min. : 20000	Min. : 77
1st Qu.:1991-01-20	1st Qu.:1991	1st Qu.: 78750	1st Qu.:33057
Median :2000-10-18	Median :2000	Median : 94500	Median :43057

Mean	:2000-10-23	Mean	:2000	Mean	:133354	Mean	:39269
3rd Qu.	:2010-07-23	3rd Qu.	:2010	3rd Qu.	:163232	3rd Qu.	:49903
Max.	:2020-10-04	Max.	:2020	Max.	:457861	Max.	:56398

Starters	Finishers	Raised	Official charity
Min. : 77	Min. : 61	Min. : 0.00	Length:40
1st Qu.:24488	1st Qu.:23252	1st Qu.: 0.00	Class :character
Median :31369	Median :30584	Median : 0.00	Mode :character
Mean :28886	Mean :28145	Mean :17.67	
3rd Qu.:35671	3rd Qu.:35326	3rd Qu.:48.05	
Max. :42906	Max. :42549	Max. :66.40	

- The Years of data span from 1981 to 2022 in winners whereas, there is data from 1981 to 2020 on london marathons.
- The highest time to finish a marathon is 16143 seconds which is a outlier.
- In one of the marathons only 77 applicants were accepted and started the marathon.

```
london_marathon[london_marathon$Starters == 77,]
```

```
# A tibble: 1 x 8
```

	Date	Year	Applicants	Accepted	Starters	Finishers	Raised
	<date>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	2020-10-04	2020	457861	77	77	61	0

```
# i 1 more variable: `Official charity` <chr>
```

```
options(scipen = 999)
# Accepted participants vs finishers by amount raised
london_marathon %>%
  filter(Raised > 0 ) %>%
  ggplot(aes(x=Applicants, y = Accepted, size = Raised)) +
  geom_point(alpha = 0.7) +
  scale_size_continuous(name = "Raised (in Millions)") +
  labs(title="Applications by Accepted counts with raised amount") +
  theme_bw()
```

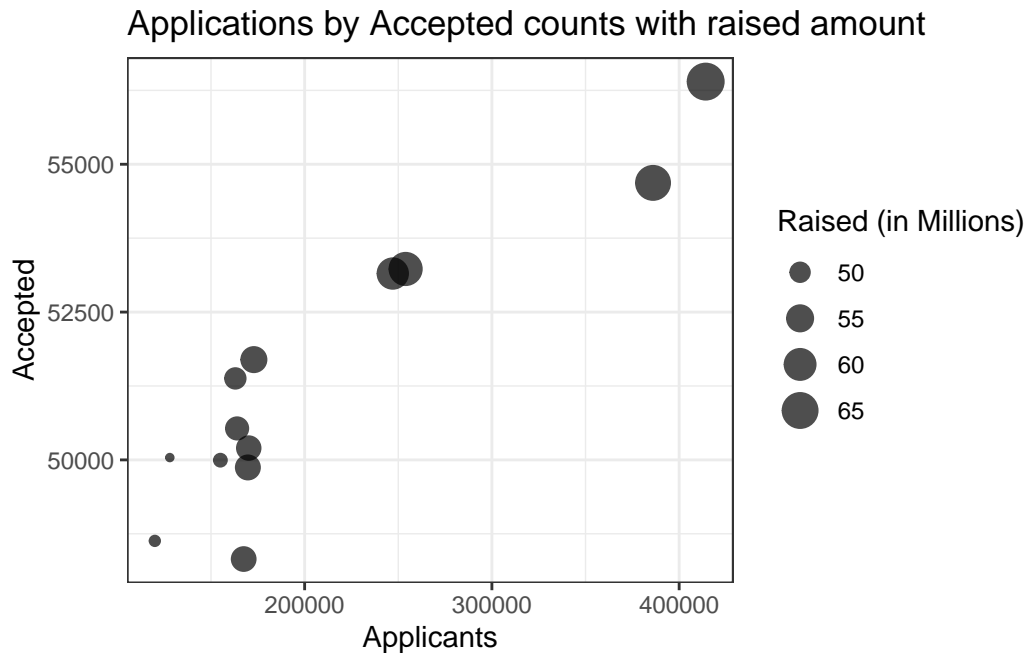


Figure 1: Scatter plot for total Applicants vs. Accepted applications and funds raised by each race

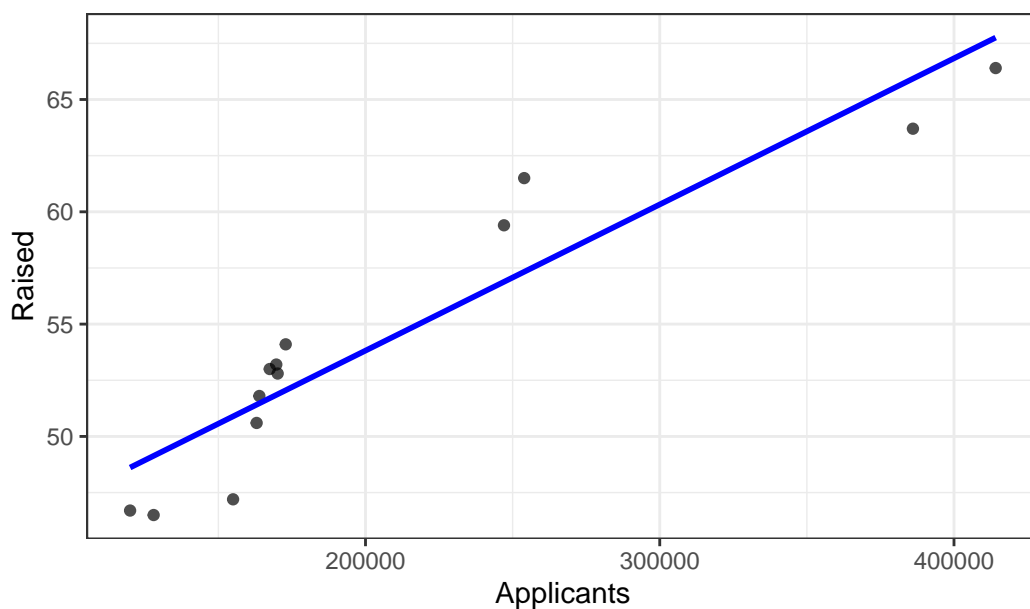
Question: Does the amount gets raised when the applicants are accepted more? Proposed Solution: Fitting a linear model can help.

```
lmod <- lm(Raised ~ Applicants + Accepted + Starters + Finishers, london_marathon)

# lmod <- lm(Raised ~ Applicants, london_marathon[london_marathon$Raised > 0,])

london_marathon %>%
  filter(Raised > 0) %>%
  # filter(Raised > 0 & Applicants < 450000) %>%
  filter(Applicants < 450000) %>%
  ggplot(aes(x=Applicants, y = Raised)) +
  geom_point(alpha = 0.7) +
  scale_size_continuous(name = "Raised (in Millions)") +
  labs(title="Applications by Accepted counts with raised amount") +
  theme_bw() +
  geom_smooth(method = "lm", formula = y ~ x, se = FALSE, color = "blue")
```

Applications by Accepted counts with raised amount



```
summary(lmod)
```

Call:

```
lm(formula = Raised ~ Applicants + Accepted + Starters + Finishers,
    data = london_marathon)
```

Residuals:

Min	1Q	Median	3Q	Max
-30.6615	-12.6324	-0.8866	10.7487	26.8530

Coefficients:

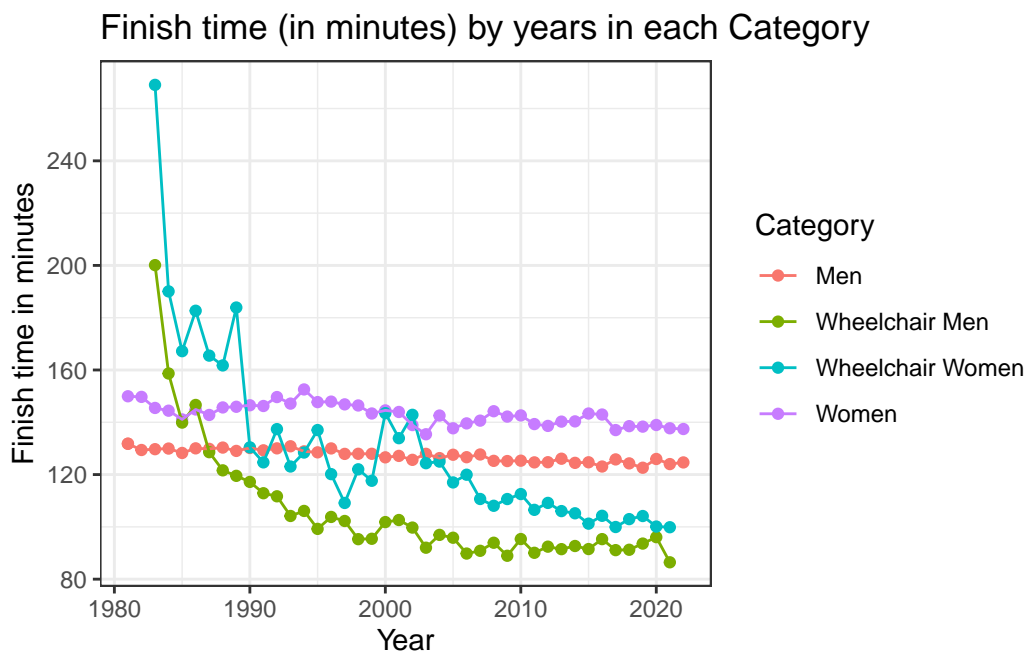
	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-40.94549533	12.37057088	-3.310	0.002171	**
Applicants	0.00011584	0.00003147	3.681	0.000777	***
Accepted	-0.00192098	0.00196776	-0.976	0.335653	
Starters	0.00046931	0.00631594	0.074	0.941190	
Finishers	0.00373246	0.00673803	0.554	0.583143	

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 15.8 on 35 degrees of freedom

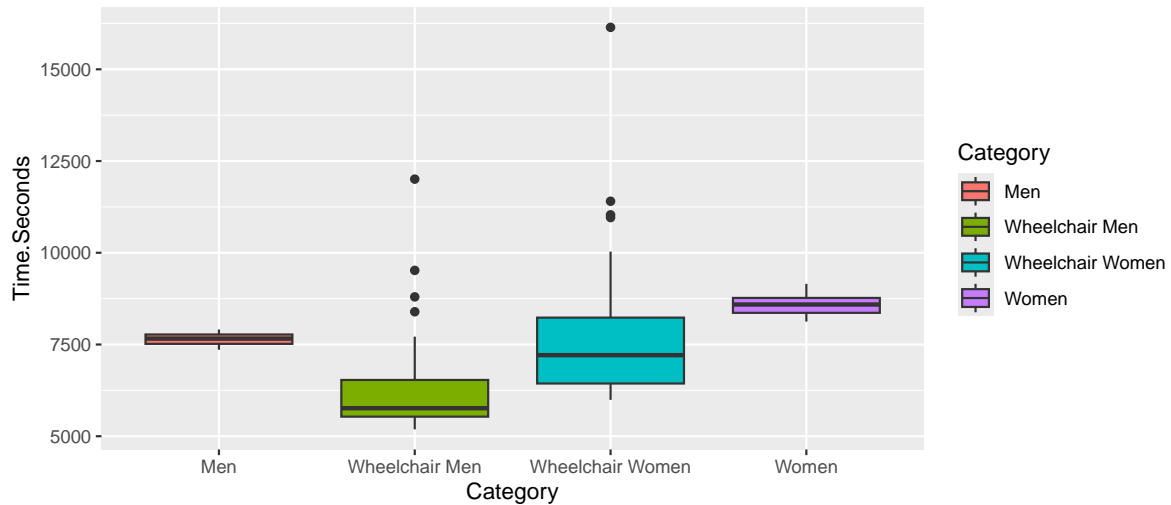
Multiple R-squared: 0.6696, Adjusted R-squared: 0.6319
F-statistic: 17.73 on 4 and 35 DF, p-value: 0.00000004866

```
# Year vs Time by Category
winners %>%
  ggplot(aes(x = Year, y = Time.Seconds / 60, color = Category)) +
  geom_point() +
  geom_line() +
  labs(
    title = "Finish time (in minutes) by years in each Category",
    x = "Year",
    y = "Finish time in minutes"
  ) +
  theme_bw()
```



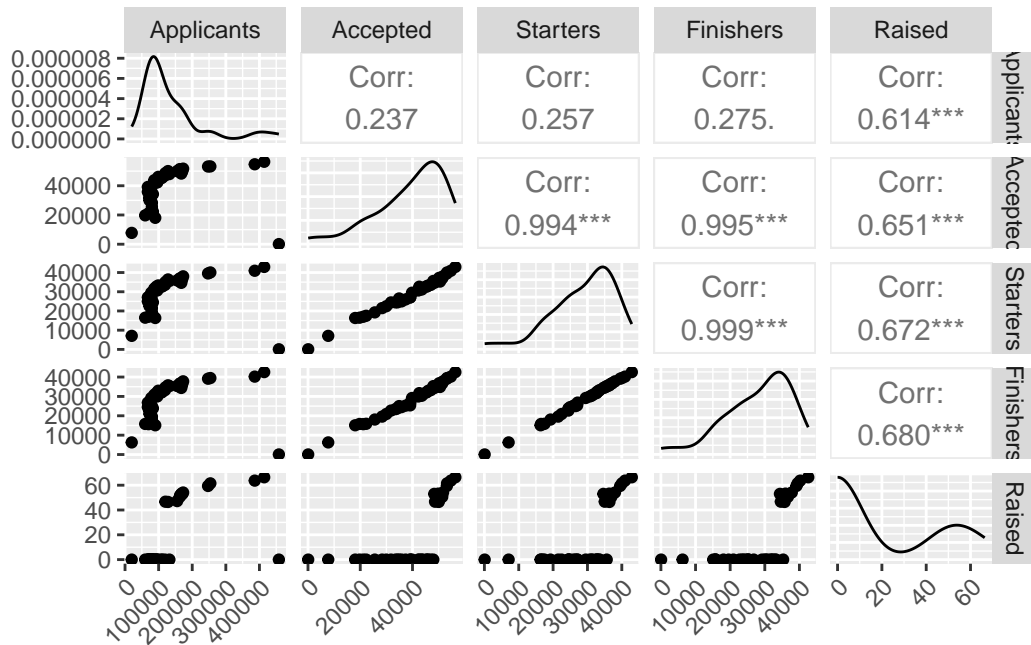
Question: Wheelchair individuals have some correlation with time to finish the race? Proposed Solution: ANOVA test to identify the relationship between Category and time to finish race

```
winners %>%
  ggplot(aes(x = Category, y = Time.Seconds, fill = Category)) +
  geom_boxplot()
```



Question: Is Womens' time normally distributed? Proposed Solution: Histogram, Boxplot, qqplot, and shapiro-wilks

```
london_marathon[,c("Applicants", "Accepted", "Starters", "Finishers", "Raised")] %>%
  # filter(Raised > 0) %>%
  ggpairs() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) # chat gpt helped me
```



OpenAI. (2024). ChatGPT [Large language model]. <https://chatgpt.com>

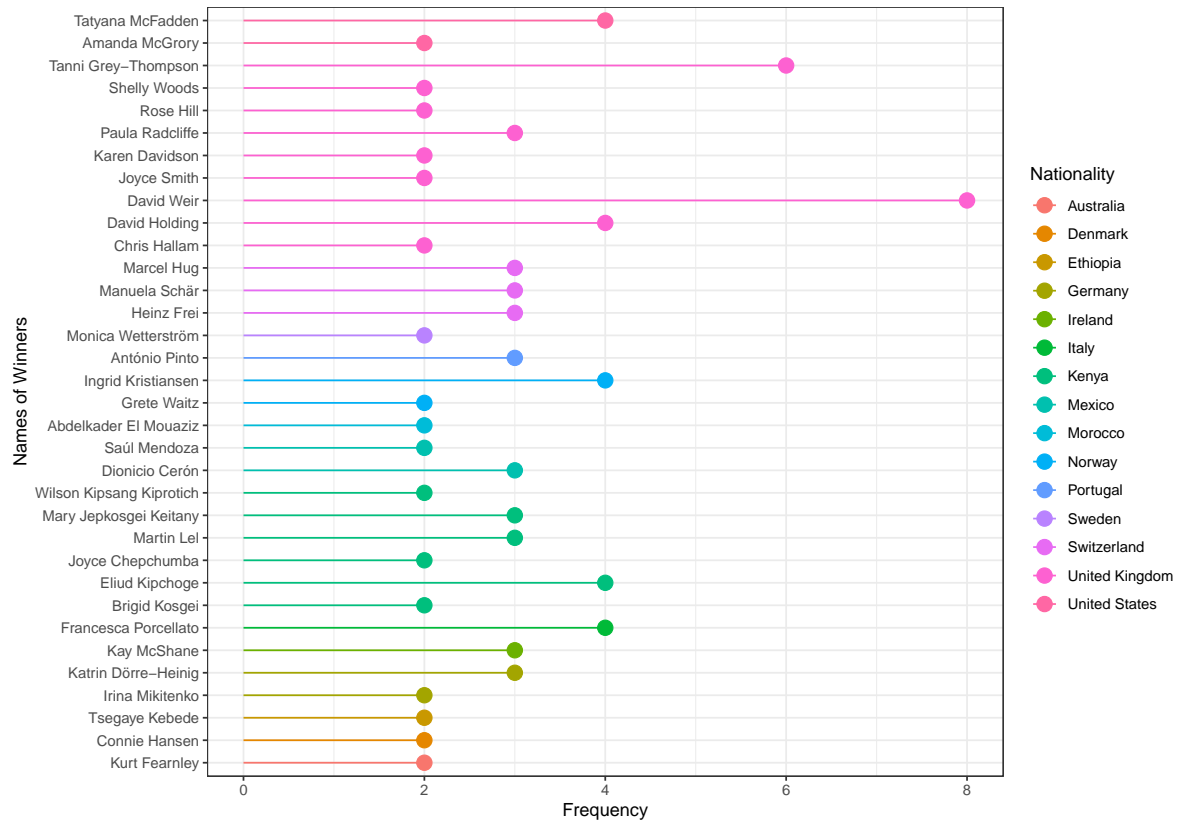
(OpenAI, 2024)

Question: Can we predict the Raised amount based on the Applicants, Accepted, Starters and Finishers. proposed solution: Multiple Linear Regression

```
winners_count <- data.frame(table(winners$Athlete))
names(winners_count) <- c("Athlete", "Frequency")

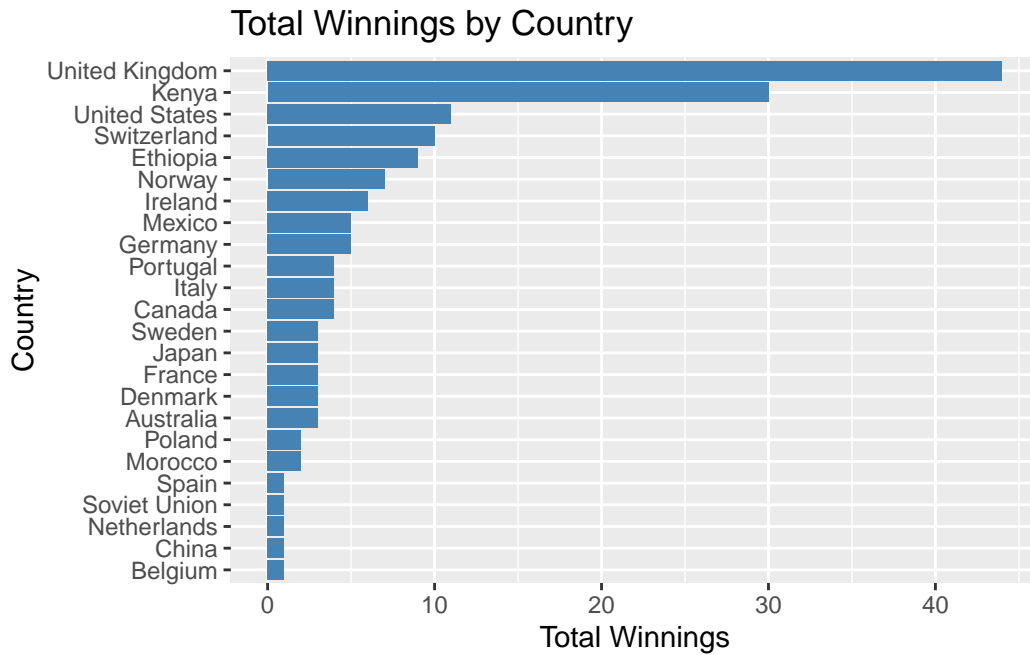
winners_nationality <- unique(left_join(winners_count,
                                         winners[,c("Athlete", "Nationality")],
                                         by="Athlete"))

winners_nationality[winners_nationality$Frequency > 1,] %>%
  arrange(Nationality) %>%
  mutate(name=factor(Athlete, levels=Athlete)) %>%
  ggplot( aes(x=name, y=Frequency, color=Nationality)) +
  geom_segment( aes(x=name, xend=name, y=Frequency, yend=0)) +
  geom_point( size=4) +
  coord_flip() +
  theme_bw() +
  xlab("Names of Winners")
```

```
grouped_nationality <- winners_nationality %>%
  group_by(Nationality) %>%
  summarise(Total_winnings = sum(Frequency))

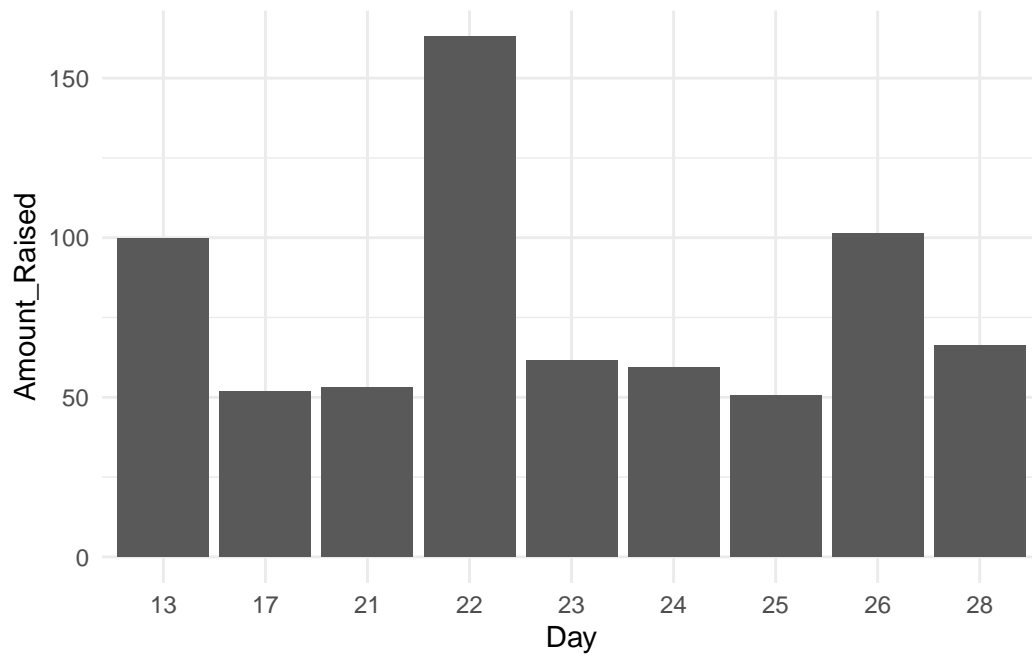
grouped_nationality %>%
  ggplot(aes(x=reorder(Nationality,Total_winnings), y=Total_winnings)) +
  geom_bar(stat="identity",fill="steelblue") +
  labs(title="Total Winnings by Country", x = "Country", y = "Total Winnings")+
  coord_flip()
```



Question: Does country have significant effect on total winnings? Proposed Solution: ANOVA

```
raised_mday <- data.frame(table(mday(london_marathon$Date), london_marathon$Raised))
raised_mday$Var2 <- as.numeric(as.character(raised_mday$Var2))
names(raised_mday) <- c("Day", "Amount_Raised", "Frequency")

raised_mday %>%
  filter(Amount_Raised > 0 & Frequency > 0) %>%
  ggplot(aes(x=Day, y=Amount_Raised)) +
  geom_col() +
  theme_minimal()
```



Question: Does the Day of month have significant effect on amount raised Proposed Solution: ANOVA