# Statistical Anova Analysis

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4/7/2022

## Cleaning our data

Down below is our functions. Everything below is already pre-run and nothing needs to be touched to examine the entire structure of this given code

```
remove_outliers <- function(x, na.rm = TRUE, ...) {
  qnt <- quantile(x, probs=c(.25, .75), na.rm = na.rm, ...)
  H <- 1.5 * IQR(x, na.rm = na.rm)
  y <- x
  y[x < (qnt[1] - H)] <- NA
  y[x > (qnt[2] + H)] <- NA
  y
}</pre>
```

This part is a portion of our data that is cleaned via R Programming

```
heartData <- read.csv("heart updated.csv")</pre>
strokeData <- read.csv("stroke_updated.csv")</pre>
strokeData[strokeData == "American Indian and Alaskan Native"] <- "AI/AN"</pre>
strokeData[strokeData == "Asian and Pacific Islander"] <- "A/PI"</pre>
heartData[heartData == "American Indian and Alaskan Native"] <- "AI/AN"
heartData[heartData == "Asian and Pacific Islander"] <- "A/PI"
#SUBSETTING DATA FOR OUTLIER REMOVAL
whiteheartData = subset(x = heartData, subset = Race.Ethnicity == "White")
whitestrokeData = subset(x = strokeData, subset = Race.Ethnicity == "White")
BlackheartData = subset(x = heartData, subset = Race.Ethnicity == "Black")
BlackstrokeData = subset(x = strokeData, subset = Race.Ethnicity == "Black")
APIheartData = subset(x = heartData, subset = Race.Ethnicity == "A/PI")
APIstrokeData = subset(x = strokeData, subset = Race.Ethnicity == "A/PI")
HispanicheartData = subset(x = heartData, subset = Race.Ethnicity == "Hispanic")
HispanicstrokeData = subset(x = strokeData, subset = Race.Ethnicity == "Hispanic")
AIANheartData = subset(x = heartData, subset = Race.Ethnicity == "AI/AN")
AIANstrokeData = subset(x = strokeData, subset = Race.Ethnicity == "AI/AN")
#REMOVING OUTLIERS
whiteheartData$Deaths.per.100.000 <- remove outliers(whiteheartData$Deaths.per.100.000)
whitestrokeData$Deaths.per.100.000 <- remove outliers(whitestrokeData$Deaths.per.100.000)
whitestrokeData <- na.omit(whitestrokeData)</pre>
whiteheartData <- na.omit(whiteheartData)</pre>
BlackheartData$Deaths.per.100.000 <- remove outliers(BlackheartData$Deaths.per.100.000)
BlackstrokeData$Deaths.per.100.000 <- remove_outliers(BlackstrokeData$Deaths.per.100.000)
BlackheartData <- na.omit(BlackheartData)</pre>
BlackstrokeData <- na.omit(BlackstrokeData)</pre>
APIheartData$Deaths.per.100.000 <- remove outliers(APIheartData$Deaths.per.100.000)
APIstrokeData$Deaths.per.100.000 <- remove outliers(APIstrokeData$Deaths.per.100.000)
APIheartData <- na.omit(APIheartData)</pre>
APIstrokeData <- na.omit(APIstrokeData)</pre>
HispanicheartData$Deaths.per.100.000 <- remove outliers(HispanicheartData$Deaths.per.100.000)
HispanicstrokeData$Deaths.per.100.000 <- remove outliers(HispanicstrokeData$Deaths.per.100.000)
HispanicstrokeData <- na.omit(HispanicstrokeData)</pre>
HispanicheartData <- na.omit(HispanicheartData)</pre>
AIANheartData$Deaths.per.100.000 <- remove outliers(AIANheartData$Deaths.per.100.000)
AIANstrokeData$Deaths.per.100.000 <- remove outliers(AIANstrokeData$Deaths.per.100.000)
AIANheartData <- na.omit(AIANheartData)
AIANstrokeData <- na.omit(AIANstrokeData)
#MERGING DATA
mergedheartData <- rbind(whiteheartData,BlackheartData, APIheartData, HispanicheartData, AIANhea
```

```
rtData)
mergedstrokeData <- rbind(whitestrokeData,BlackstrokeData, APIstrokeData, HispanicstrokeData, AI
ANstrokeData)
```

### **ANOVA TEST**

```
# Hypotheses from Mean Model
# H_0: mu_A/PI = mu_AI/AN = mu_Black = mu_Hispanic = mu_White
# H_1: at least one of the means is different

# ANOVA method
dataAnova <- aov(mergedheartData$Deaths.per.100.000 ~ mergedheartData$Race.Ethnicity)
dataAnova2 <- aov(mergedstrokeData$Deaths.per.100.000 ~ mergedstrokeData$Race.Ethnicity)
summary(dataAnova)</pre>
```

```
summary(dataAnova2)
```

```
## Df Sum Sq Mean Sq F value Pr(>F)

## mergedstrokeData$Race.Ethnicity 4 2137609 534402 1695 <2e-16 ***

## Residuals 6584 2075619 315

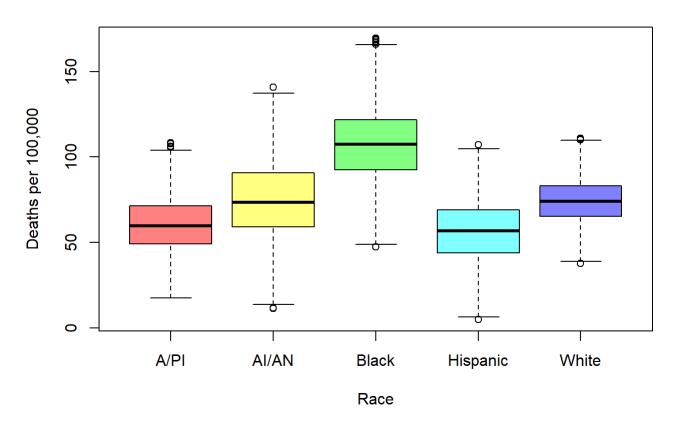
## ---

## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

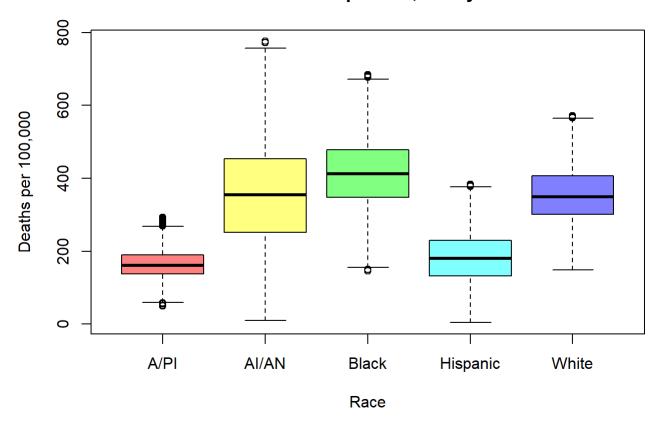
```
# p-value is represented by Pr(>F). It is very small for both tests, reject H_0. # Conclude at least one mean is different statistically. There is a significant # difference in the mortality trend between races.
```

### Additional Exploratory Data Analysis

#### Stroke fatalities per 100,000 by Race



#### Heart fatalities per 100,000 by Race



#Boxplots to distinguish outliers and present visualization on significant #differences between multiple race/ethnicities

case.vector = tapply(mergedheartData\$Deaths.per.100.000, mergedheartData\$Race.Ethnicity, sum)
case.vector2 = tapply(mergedstrokeData\$Deaths.per.100.000, mergedstrokeData\$Race.Ethnicity, sum)

case.vector #Total number of deaths for this year from heart disease

```
## A/PI AI/AN Black Hispanic White
## 489847.8 276762.2 864993.3 308981.1 1122818.0
```

case.vector2 #Total number of deaths for this year from stroke

```
## A/PI AI/AN Black Hispanic White
## 38853.7 16646.0 169423.0 58519.5 232077.7
```

aggregate(mergedheartData\$Deaths.per.100.000, list(mergedheartData\$Race.Ethnicity), FUN=mean) #A verage number of deaths

```
## Group.1 x
## 1 A/PI 165.7691
## 2 AI/AN 358.0365
## 3 Black 412.8846
## 4 Hispanic 186.4702
## 5 White 357.6993
```

aggregate(mergedstrokeData\$Deaths.per.100.000, list(mergedstrokeData\$Race.Ethnicity), FUN=mean)
#Average number of deaths

```
## Group.1 x

## 1 A/PI 60.89922

## 2 AI/AN 74.31250

## 3 Black 107.98152

## 4 Hispanic 56.10690

## 5 White 74.50327
```