

script

```
# load dataset that has been cropped for analysis
data <- read.csv("Cropped_Data.csv")

# the first column of the dataset is the names of the rows
# so we renamed the row names based on first column and
# deleted the column at the end
row_number <- nrow(data)
for (i in 1:row_number) {
  rownames(data)[i] = data[i,1]
}

# Delete the column
data <- data[,c(2:ncol(data))]

# Get rid of the first letter X for each of the years
colnames(data) <- gsub("X", "", colnames(data))

# Rename part of the row names
#(i.e. change .i Inapplicable into Inapplicable for better
#data representation)
rownames(data)[1] <- "Inapplicable"
rownames(data)[2] <- "No answer"
rownames(data)[3] <- "Do not Know/Cannot Choose"
rownames(data)[4] <- "Skipped on Web"

# Save the cleaned data
write.csv(data, "Cleaned_Data.csv")
```

Average Working Hour in 2008, 2021, 2022, and total

```
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# Save the cleaned data
write.csv(data, "Cleaned_Data.csv")

#Graphs
library(ggplot2)
data <- read.csv("Cleaned_Data.csv")
# Get rid of the first letter X for each of the years
colnames(data)[1] <- "work_hours"
colnames(data) <- gsub("X", "", colnames(data))

#Histogram of average working hours for 2008, 2021, 2022, and Total
hist_data <- select(data, work_hours, "2008", "2021", "2022", "Total")
hist_data <- hist_data[5:94,]
```

```

sum <- c(0,0,0,0)
total_people <- c(0,0,0,0)
for(i in 1:4){
  total_people[i] <- sum(hist_data[, i+1])
}
hist_data[, 1] <- sapply(hist_data[, 1], as.numeric)

```

Warning in lapply(X = X, FUN = FUN, ...): NAs introduced by coercion

```

hist_data[90,1] <- 90

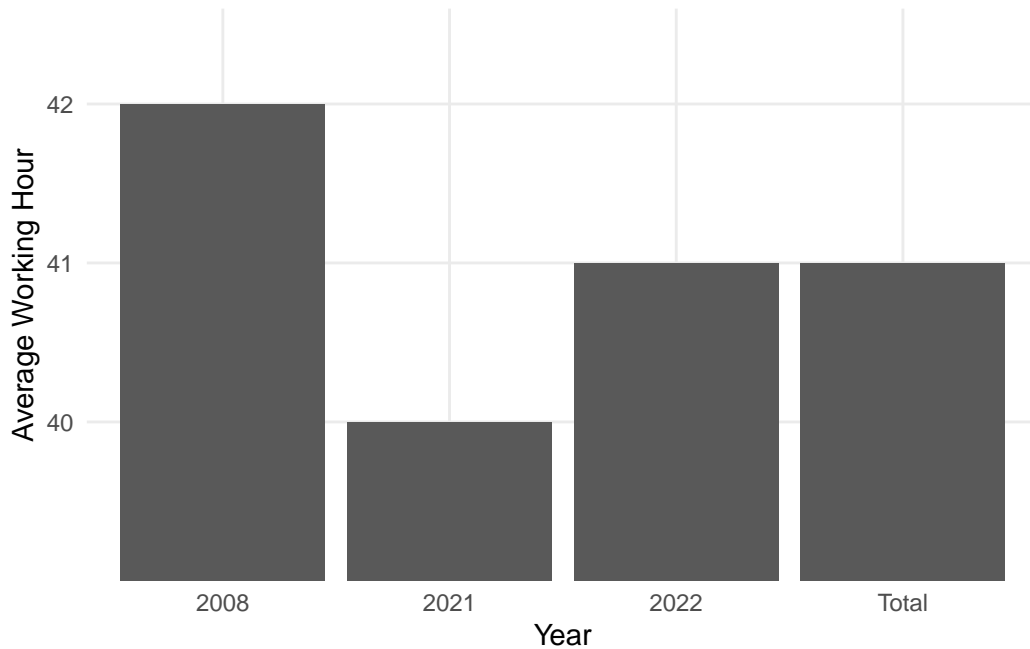
sum[1] <- sum(hist_data$"2008" * hist_data$work_hours)
sum[2] <- sum(hist_data$"2021" * hist_data$work_hours)
sum[3] <- sum(hist_data$"2022" * hist_data$work_hours)
sum[4] <- sum(hist_data$Total * hist_data$work_hours)

averages <- round(sum/total_people)
years <- c("2008","2021","2022" ,"Total")

average_hours <- data.frame(cbind(averages,years))

ggplot(average_hours,aes(x=years,y=averages)) +
  geom_bar(stat="identity") +
  theme_minimal() + # Make the theme neater
  labs(x = "Year", y = "Average Working Hour") +
  scale_color_brewer(palette = "Set1") +
  theme(legend.position = "bottom")

```



Data Cleaning: Modify hours into 1-20,20-40,40-60,60-80, 80+ categories

```
data <- read.csv("Cleaned_Data.csv")
colnames(data)[1] <- "work_hours"
colnames(data) <- gsub("X", "", colnames(data))

cate_data <- data
year <- colnames(data)
Hours <- c("No Response", "0-20", "21-40", "41-60", "61-80", "80+", "Total")

col_number <- ncol(cate_data) - 1

filtered_df1 <- cate_data %>%
  filter(work_hours < 20)

tweenties <- rep(0, 35)
for(i in 1:col_number){
  tweenties[i] <- sum(filtered_df1[, i+1])
}

filtered_df2 <- cate_data %>%
```

```

filter(work_hours < 40) %>%
filter(work_hours > 20)

forties <- rep(0, 35)
for(i in 1:col_number){
  forties[i] <- sum(filtered_df2[, i+1])
}

filtered_df3 <- cate_data %>%
  filter(work_hours < 60) %>%
  filter(work_hours > 40)

sixties <- rep(0, 35)
for(i in 1:col_number){
  sixties[i] <- sum(filtered_df3 [, i+1])
}

filtered_df4 <- cate_data %>%
  filter(work_hours < 80) %>%
  filter(work_hours > 60)

eighties <- rep(0, 35)
for(i in 1:col_number){
  eighties[i] <- sum(filtered_df4[, i+1])
}

filtered_df5 <- cate_data %>%
  filter(work_hours > 80)
filtered_df5 <- filtered_df5[5:14,]

more <- rep(0, 35)
for(i in 1:col_number){
  more[i] <- sum(filtered_df5[, i+1])
}

filtered_df6 <- cate_data %>%
  filter(work_hours > 80)
filtered_df6 <- filtered_df6[1:4,]

No_Response <- rep(0, 35)
for(i in 1:col_number){
  No_Response[i] <- sum(filtered_df6[, i+1])
}

```

```

}

rm(filtered_df1,filtered_df2,filtered_df3,filtered_df4,filtered_df5,filtered_df6)

total <- data[95,2:36]

cate_data <- rbind(No_Response,twenties,forties,sixties,eighties,more,total)
cate_data <- data.frame(cbind(Hours,cate_data))
colnames(cate_data) <- year

write.csv(cate_data,"cleaned_categorized_data.csv")

data<- read.csv("cleaned_categorized_data.csv")
# Delete the column
data <- data[c(2:ncol(data))]
colnames(data)[1] <- "Work Hours/Years"
# Get rid of the first letter X for each of the years
colnames(data) <- gsub("X", "", colnames(data))
#kable(data,row.names = FALSE)|>
# kable_styling() |>
#row_spec(6, hline_after = TRUE)

```

Comparson between the years around 2008

Percentage of Non-response rate

```

# Calculate percentages
data <- read.csv("cleaned_categorized_data.csv")
data <- data[c(2:ncol(data))]
data <- data[1:6,]

```

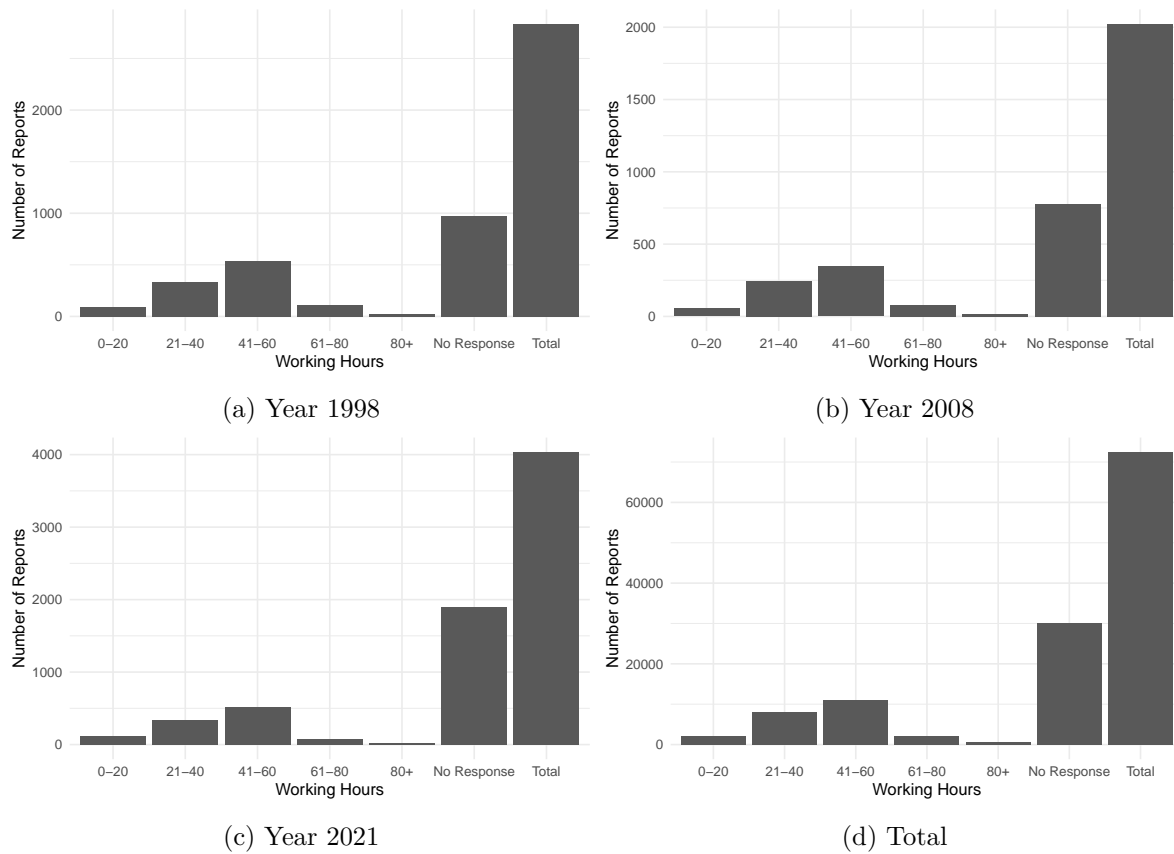


Figure 1: Beliefs in Free will vs. Job Satisfaction between Two Populations