

# Homework 2: Married children on the Titanic

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2023-07-17

## Overview

For exploring, I decided to take the Titanic manifest down to just those under 18. In doing so, I noticed several females with “Mrs.” in the Name column, and realized at that time, that was probably plenty old enough to be married. Back then, it was likely that children were either referred to by their name, or perhaps Miss or Master. As well, the age of consent was very different in the 1910’s.

```
library(sqldf)

## Loading required package: gsubfn
## Loading required package: proto
## Loading required package: RSQLite

library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

library(stringr)

#7. BONUS - place the original .csv in a github file and read from a link.
loc<-"https://raw.githubusercontent.com/tonythor/pyspark-env/main/data/titanic.csv"
titanic_dataset <- loc
titanic <- read.csv(titanic_dataset)

#3. Create new column names for the new data frame.
titanic <- dplyr::rename(titanic, c("Class" = "Pclass"), c("Gender" = "Sex"))
titanic[c('LastName', 'FirstName')]<- str_split_fixed(titanic$Name, ',', 2)

#2. Create a new data frame with a subset of the columns and rows. Make
#   sure to rename it.
t_children = sqldf("select  Survived, Class, LastName, FirstName,
                          Gender, Age from titanic where age < 18
                          order by Class, Age desc")

#1 Use the summary function to gain an overview of the data set. Then
#   display the mean and median for at least two attributes.
```

```
summary(t_children)
```

```
##      Survived      Class      LastName      FirstName
## Min.   :0.0000 Min.   :1.000 Length:113 Length:113
## 1st Qu.:0.0000 1st Qu.:2.000 Class :character Class :character
## Median :1.0000 Median :3.000 Mode  :character Mode  :character
## Mean   :0.5398 Mean   :2.584
## 3rd Qu.:1.0000 3rd Qu.:3.000
## Max.   :1.0000 Max.   :3.000
##      Gender      Age
## Length:113 Min.   : 0.420
## Class :character 1st Qu.: 3.000
## Mode  :character Median : 9.000
##              Mean   : 9.041
##              3rd Qu.:16.000
##              Max.   :17.000
```

```
sprintf("P Class : mean:%s median:%s",
        round(mean(t_children$Class), digits=4),
        round(median(t_children$Class), digits=4))
```

```
## [1] "P Class : mean:2.5841 median:3"
```

```
sprintf("Survived: mean:%s median:%s",
        round(mean(t_children$Survived), digits=4),
        round(median(t_children$Survived), digits=4))
```

```
## [1] "Survived: mean:0.5398 median:0.5398"
```

```
# Add that column in there based on salutations
```

```
t_children = t_children %>%
  mutate(PossiblyMarried = case_when(
    grepl("Mrs.", FirstName) ~ "True",
    grepl("Mister.", FirstName) ~ "True",
    grepl("Mr.", FirstName) ~ "True",
    .default = ""
  ))
```

```
#5 For at least 3 values in a column please rename so that every value in
# that column is renamed.
```

```
replacements = c('Miss.'='', 'Master.'='', 'Mrs.'='', 'Mister.'='', 'Mr.'='')
t_children$FirstName<- str_replace_all(t_children$FirstName, replacements )
```

```
# The first name column is pretty long. Let's truncate it so it'll fit on one page.
```

```
t_children = t_children %>% mutate(FirstName = str_trunc(FirstName, width = 10))
```

```
married = nrow(sqldf("select PossiblyMarried from t_children where length(PossiblyMarried) > 1"))
sprintf('Number of children in this dataset: %s', nrow(t_children))
```

```
## [1] "Number of children in this dataset: 113"
```

```
sprintf('Number of children who might already be married: %s', married)
```

```
## [1] "Number of children who might already be married: 26"
```

*#6. Display enough rows to see examples of all of steps 1-5 above.*  
`head(t_children, 50)`

##	Survived	Class	LastName	FirstName	Gender	Age	PossiblyMarried
## 1	1	1	Penasco y Castellana	Victo...	female	17.00	True
## 2	1	1	Thayer	John ...	male	17.00	True
## 3	1	1	Dick	Alber...	female	17.00	True
## 4	1	1	Hippach	Jean ...	female	16.00	
## 5	1	1	Maioni	Roberta	female	16.00	
## 6	1	1	Lines	Mary ...	female	16.00	
## 7	1	1	Madill	Georg...	female	15.00	
## 8	1	1	Carter	Lucil...	female	14.00	
## 9	1	1	Carter	Willi...	male	11.00	
## 10	1	1	Dodge	Washi...	male	4.00	
## 11	0	1	Allison	Helen...	female	2.00	
## 12	1	1	Allison	Hudso...	male	0.92	
## 13	1	2	Ilett	Bertha	female	17.00	
## 14	1	2	Lehmann	Bertha	female	17.00	
## 15	0	2	Gaskell	Alfred	male	16.00	True
## 16	0	2	Mudd	Thoma...	male	16.00	True
## 17	1	2	Nasser	Nicho...	female	14.00	True
## 18	1	2	Mellinger	Madel...	female	13.00	
## 19	1	2	Collyer	Marjo...	female	8.00	
## 20	1	2	Davies	John ...	male	8.00	
## 21	1	2	Hart	Eva M...	female	7.00	
## 22	1	2	Harper	Annie...	female	6.00	
## 23	1	2	West	Const...	female	5.00	
## 24	1	2	Becker	Mario...	female	4.00	
## 25	1	2	Wells	Joan	female	4.00	
## 26	1	2	Laroche	Simon...	female	3.00	
## 27	1	2	Navratil	Michel M	male	3.00	
## 28	1	2	Richards	Willi...	male	3.00	
## 29	1	2	Navratil	Edmon...	male	2.00	
## 30	1	2	Quick	Phyll...	female	2.00	
## 31	1	2	Becker	Richa...	male	1.00	
## 32	1	2	Mallet	Andre	male	1.00	
## 33	1	2	Caldwell	Alden...	male	0.83	
## 34	1	2	Richards	Georg...	male	0.83	
## 35	1	2	Hamalainen	Viljo	male	0.67	
## 36	1	3	Andersson	Erna ...	female	17.00	
## 37	0	3	Attalah	Malake	female	17.00	
## 38	0	3	Calic	Jovo	male	17.00	True
## 39	0	3	Kallio	Nikol...	male	17.00	True
## 40	0	3	Calic	Petar	male	17.00	True
## 41	0	3	Elias	Josep...	male	17.00	True
## 42	0	3	Jensen	Svend...	male	17.00	True
## 43	0	3	Culumovic	Jeso	male	17.00	True
## 44	0	3	Goodwin	Lilli...	female	16.00	
## 45	0	3	Ford	Willi...	male	16.00	True
## 46	0	3	Osen	Olaf ...	male	16.00	True
## 47	1	3	Gilnagh	Kathe...	female	16.00	
## 48	1	3	Carr	Helen...	female	16.00	
## 49	1	3	Sunderland	Victo...	male	16.00	True
## 50	0	3	Panula	Ernes...	male	16.00	True