

Homework Example

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Problem 1

a) Save the parameter values as objects.

```
Population_Mean<- c(29.3)
Standard_Deviation <-c(9.9)
Significant_Level<- c(0.05)
```

Part b

```
mytest<-function(x){
  samp<-rnorm(x,29.3,9.9)
  ts<-(mean(samp)-29.3)/(9.9/sqrt(x))
  pval<- 2*pnorm(-abs(ts))
  pval<= Significant_Level
}
mytest(26)

## [1] FALSE
```

Part c. What proportion of tests reject the null hypothesis?

```
Replicate<-replicate(10000, mytest(26))
sum(Replicate)/10000
```

```
## [1] 0.047
```

##d. Theoretically, what should the proportion resulting from part (c) be?

It should be equal to the significant level

##e. Write a function that will output the #proportion described in part (c) for a given sample size #(again, without using any loops). Execute your function #for the sample sizes 7, 26, and 50.

```
prop<-function(x){
  Replicate1<-replicate(10000, mytest(x))
  sum(Replicate1)/10000
}
prop(7)
```

```
## [1] 0.0497
```

```
prop(26)
```

```
## [1] 0.0498
```

```
prop(50)
```

```
## [1] 0.0517
```

(f) Without using any loops,

#execute the function written in part (e) #for every sample size from 3 through 55.

```
sapply(3:55,prop)
```

```
## [1] 0.0485 0.0485 0.0474 0.0452 0.0468 0.0525 0.0464 0.0479 0.0512 0.0517
## [11] 0.0484 0.0495 0.0486 0.0519 0.0544 0.0506 0.0486 0.0467 0.0532 0.0511
## [21] 0.0516 0.0503 0.0498 0.0540 0.0505 0.0506 0.0475 0.0504 0.0459 0.0505
## [31] 0.0515 0.0522 0.0502 0.0482 0.0511 0.0482 0.0524 0.0496 0.0538 0.0499
## [41] 0.0502 0.0481 0.0443 0.0497 0.0497 0.0506 0.0501 0.0506 0.0464 0.0524
## [51] 0.0481 0.0489 0.0501
```

#(g) What do you notice about the general trend of the resulting proportions? Does sample size appear to have any effect on the results?

```
# The general trend that it is
#close to the significant level which is 0.05.
#Even the sample size is different,
#the prop is almost 0.05
```

Problem 2 #a. Read in the file and save the data

#in a data frame called nym2019.

```
nym2019 <-  
read.table("/Users/mailuu/Desktop/STAT3080/nym2019.txt", header=TRUE,  
stringsAsFactors =FALSE,  
fileEncoding = "latin1")  
head(nym2019)
```

##	Sex	Age	Place	DivPlace	DIV	Time	BostonQualifier	HomeStateOrCountry
## 1	M	26	7660	751	M25-29	216.40	N	FL
## 2	M	29	768	157	M25-29	172.02	Y	NY
## 3	M	52	6028	494	M50-54	209.52	N	FRA
## 4	M	42	9247	1252	M40-44	222.30	N	GER
## 5	M	40	5819	861	M40-44	208.77	N	FRA
## 6	M	50	859	31	M50-54	173.45	Y	IL

2b. Determine the number of finishers' times that are contained in this data set.

```
length(nym2019$Time)
```

```
## [1] 250
```

2c. Determine the number of finishers in the data whose home country is the U.S.

```
sum(nchar(as.character(nym2019$HomeStateOrCountry))==2)
```

```
## [1] 121
```

2d. Determine the number of finishers representing each country.

```
home<-as.character((nym2019$HomeStateOrCountry))  
home[nchar(home)==2]<-"USA"  
table(home)
```

```
## home
## ARG ARU AUS BEL BRA CAN CHI CHN CRC CZE ESP ETH FIN FRA GBR GER INA IRL ISR ITA
## 1 1 6 1 4 4 3 7 2 1 10 1 1 21 6 10 1 2 1 8
## JPN KEN MEX NED NOR NZL PAN PAR POL POR RSA RUS SIN SLO SUI SWE TPE UKR URU USA
## 5 1 4 8 2 1 1 1 1 1 1 2 1 1 3 1 1 1 1 121
## VEN
## 1
```

2e. Determine the number of countries represented in the data.

```
nrow(table(home))
```

```
## [1] 41
```

f. Determine the age of the youngest and

#oldest finishers given in the data

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0 --
```

```
## v ggplot2 3.2.1      v purrr 0.3.3
```

```
## v tibble 2.1.3       v dplyr 0.8.3
```

```
## v tidyr 1.0.2        v stringr 1.4.0
```

```
## v readr 1.3.1        v forcats 0.4.0
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag() masks stats::lag()
```

```
Youngest_Oldest<-group_by(nym2019) %>%
  summarize(YoungestAge=min(Age),OldestAge=max(Age))
Youngest_Oldest
```

```
## # A tibble: 1 x 2
```

```
##   YoungestAge OldestAge
```

```
##       <int>      <int>
```

```
## 1         23         66
```

(g) Determine the age of the fastest and slowest finishers given in the data.

```
nym2019[nym2019$Time == min(nym2019$Time),"Age"]
```

```
## [1] 25
```

```
nym2019[nym2019$Time == max(nym2019$Time),"Age"]
```

```
## [1] 37
```

h. #9. number who qualifies. and their information

```
Top10_Div<-nym2019[nym2019$DivPlace<= 10,]
nrow(Top10_Div)
```

```
## [1] 9
```

```
Top10_Div
```

##	Sex	Age	Place	DivPlace	DIV	Time	BostonQualifier	HomeStateOrCountry
## 36	F	35	919	10	F35-39	174.15	Y	NY
## 37	M	36	95	5	M35-39	153.43	Y	NJ
## 66	F	29	748	6	F25-29	171.68	Y	NC
## 71	F	30	50	2	F30-34	147.12	Y	AUS
## 86	F	24	265	1	F20-24	162.35	Y	ETH
## 152	F	43	92	4	F40-44	153.07	Y	FL
## 214	F	42	43	2	F40-44	146.38	Y	AUS
## 225	M	66	5854	6	M65-69	208.88	Y	OH
## 236	M	25	2	2	M25-29	128.60	Y	KEN

#i. Determine the divisions of the finishers who finished in the Top 10 of their division.

```
Divisions_Top_Ten<-Top10_Div[,5]
sort(unique(Divisions_Top_Ten))
```

```
## [1] "F20-24" "F25-29" "F30-34" "F35-39" "F40-44" "M25-29" "M35-39" "M65-69"
```

(j) Display all information for finishers who finished in the Top 5 of their division.

```
Top_5_Div<-nym2019[nym2019$DivPlace<= 5,]
Top_5_Div
```

##	Sex	Age	Place	DivPlace	DIV	Time	BostonQualifier	HomeStateOrCountry
## 37	M	36	95	5	M35-39	153.43	Y	NJ
## 71	F	30	50	2	F30-34	147.12	Y	AUS
## 86	F	24	265	1	F20-24	162.35	Y	ETH
## 152	F	43	92	4	F40-44	153.07	Y	FL
## 214	F	42	43	2	F40-44	146.38	Y	AUS
## 236	M	25	2	2	M25-29	128.60	Y	KEN

##(k) Determine the average age of finishers who did and who did not qualify for the Boston Marathon.

```
tapply(nym2019$Age, nym2019$BostonQualifier, mean)
```

```
##           N           Y
## 39.86364 36.72034
```

References

1. <https://www.mathsisfun.com/numbers/addition.html>
2. <https://www.mathsisfun.com/numbers/subtraction.html>