

# Summative Assessment EMATM0061 - Part B

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12/01/2022

## Section B

### B.1 Faulty Security System Probability

#### Given

##### Assumption - passage of time is per minute  $p^0$  - conditional probability that the sensor makes a sound if there is no person within one meter of the gate

$p^1$  - conditional probability that the sensor makes at least once, if there is at least one person present

$q$  - probability that at least one person walks within one meter of the gate

$\phi$  - conditional probability that at one person has passed during the minute, and the alarm made the sound.

therefore,  $q^0$  - probability that nobody walks within one meter of the gate

Part (a):

Function used ->

$$\phi = (p^1/q)/(p^0/q^0 + p^1/q)$$

```
c_prob_person_given_alarm = function(p0,p1,q){
  (p1/q) / ((p0/(1-q)) + (p1/q))
}
```

Part (b):

Value of  $\phi$

Given:  $p^0 = 0.05$ ,

$p^1 = 0.95$ ,

$q = 0.1$

```
phi <- c_prob_person_given_alarm(0.05,0.95,0.1)
print(phi)
```

```
## [1] 0.994186
```

Part (c):

Graph between  $\phi$  and  $q$  for varying value of  $q$

Given :  $p^0 = 0.05$ ,

$p^1 = 0.95$

```
library(ggplot2)
```

```
x1<-seq(from=0, to=1, by=.01)
```

```
print(x1)
```

```
## [1] 0.00 0.01 0.02 0.03 0.04 0.05 0.06 0.07 0.08 0.09 0.10 0.11 0.12 0.13 0.14
## [16] 0.15 0.16 0.17 0.18 0.19 0.20 0.21 0.22 0.23 0.24 0.25 0.26 0.27 0.28 0.29
## [31] 0.30 0.31 0.32 0.33 0.34 0.35 0.36 0.37 0.38 0.39 0.40 0.41 0.42 0.43 0.44
## [46] 0.45 0.46 0.47 0.48 0.49 0.50 0.51 0.52 0.53 0.54 0.55 0.56 0.57 0.58 0.59
## [61] 0.60 0.61 0.62 0.63 0.64 0.65 0.66 0.67 0.68 0.69 0.70 0.71 0.72 0.73 0.74
## [76] 0.75 0.76 0.77 0.78 0.79 0.80 0.81 0.82 0.83 0.84 0.85 0.86 0.87 0.88 0.89
## [91] 0.90 0.91 0.92 0.93 0.94 0.95 0.96 0.97 0.98 0.99 1.00
```

```
plotphi<- as.data.frame(x1)
```

```
dim(plotphi)
```

```
## [1] 101 1
```

```
x2<-seq(from=0, to=0.1, by=.01)
```

```
c_prob = function(p0,p1,input_vector){
  (p1/input_vector) / ((p0/(1-input_vector)) + (p1/input_vector))
}
```

```
x2 <-c_prob(0.05,0.95,x1)
```

```
x2[is.na(x2)]<-0
```

```
phivalue<-as.data.frame(x2)
```

```
plotphi$phivalue<-x2
```

```
plotphi <-setNames(plotphi,c("qvalue", "phivalue"))
```

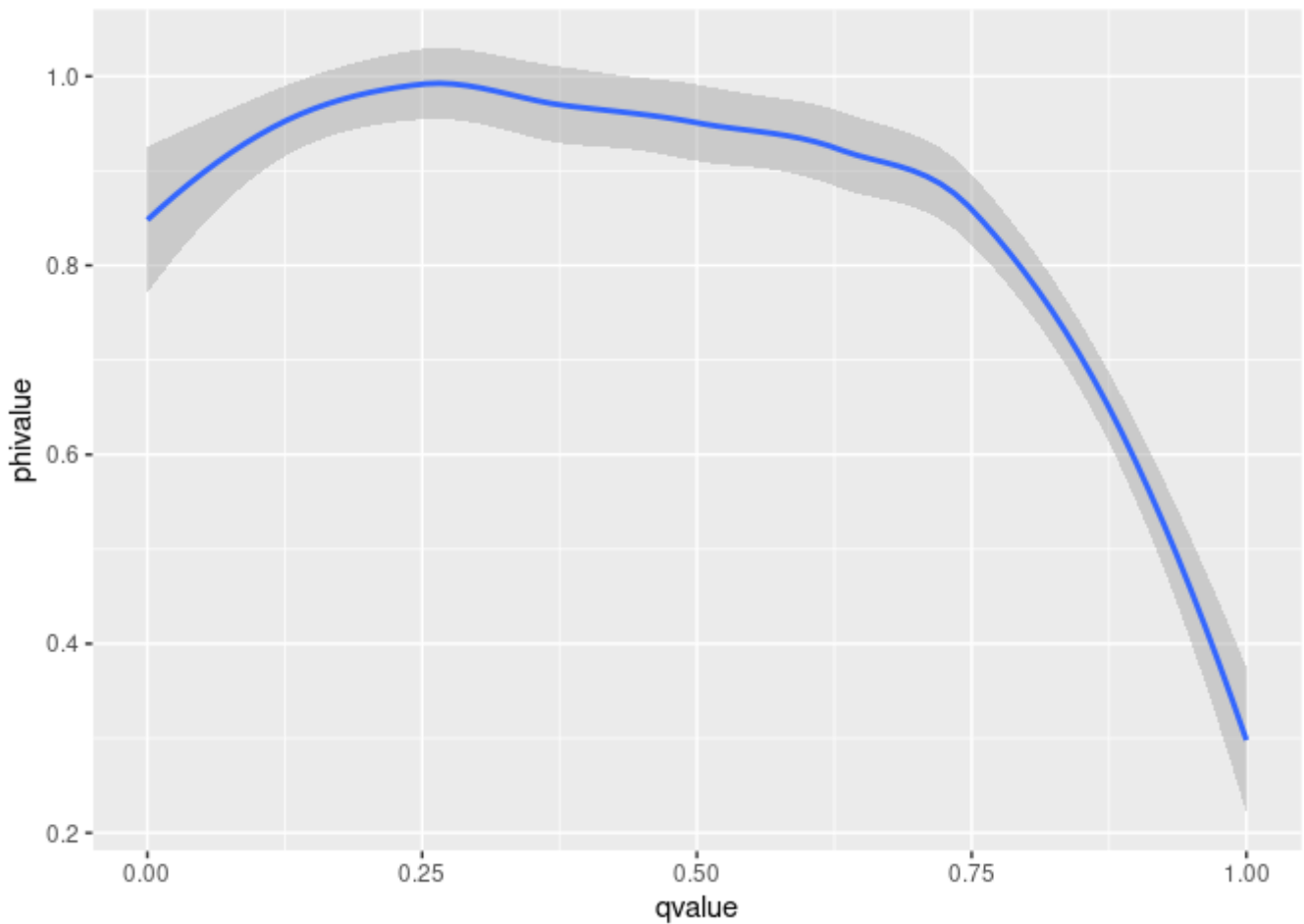
```
head(plotphi)
```

```
## qvalue phivalue
## 1 0.00 0.0000000
## 2 0.01 0.9994687
## 3 0.02 0.9989270
## 4 0.03 0.9983749
## 5 0.04 0.9978118
## 6 0.05 0.9972376
```

```
flow<- ggplot(plotphi, aes(x=qvalue,y=phivalue) ) +geom_smooth()
```

```
flow
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



## B.2 Discrete function

Given -

$$\alpha, \beta, \gamma \in$$

$$0, 1$$

$$X(\text{DNV}) - \{0, 1, 2, 5\}$$

$$P(X=1) = \alpha$$

$$P(X=2) = \beta$$

$$P(X=5) = \gamma$$

$$P(X \notin \{0, 1, 2, 5\}) = 0$$

1. Probability mass function =

$$P(x) = \{1 - \alpha - \beta - \gamma, \text{ if } x=0\}$$

$$\{\alpha, \text{ if } x=1\}$$

$$\{\beta, \text{ if } x=2\}$$

$$\{\gamma, \text{ if } x=5\}$$

$$\{0, \text{ otherwise}\}$$

$$2. E(X) = \alpha + 2\beta + 5\gamma$$

$$3. E(X^2) - E(X)^2 = (\alpha + 4\beta + 25\gamma) - (\alpha + 2\beta + 5\gamma)^2$$

$X_1, X_2, \dots, X_n$   
 $P(X_i=1) = \alpha$ ,  
 $P(X_i=2) = \beta$ ,  
 $P(X_i=5) = \gamma$  and  
 $P(X \notin \{0, 1, 2, 5\}) = 0$

let  $\bar{X} =$

$\frac{1}{n} \sum_{i=1}^n X_i$  be the sample mean

4.  $E(X) = \frac{1}{n} \sum_{i=1}^n X_i = (\alpha + 2\beta + 5\gamma)$

5.  $E(X^2) - E(X)^2 = \frac{1}{n} \sum_{i=1}^n X_i^2 - (\alpha + 2\beta + 5\gamma)^2$

6.

```
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('purrr')
sample_X_0125 <- function(n, alpha, beta, gamma) {
  sample_X <- data.frame(U = runif(n)) %>%
    mutate(X = case_when(
      (0 <= U) & (U < alpha) ~ 1,
      (alpha <= U) & (U < alpha + beta) ~ 2,
      (alpha + beta <= U) & (U < alpha + beta + gamma) ~ 5,
      (alpha + beta + gamma <= U) & (U <= 1) ~ 0)) %>% pull(X)
  return(sample_X)
}
```

7.

```
n <- 100000
alpha <- 1/10
beta <- 2/10
gamma <- 3/10

sample_X <- sample_X_0125(n, alpha, beta, gamma)
mean(sample_X)
```

```
## [1] 1.99806
```

8.

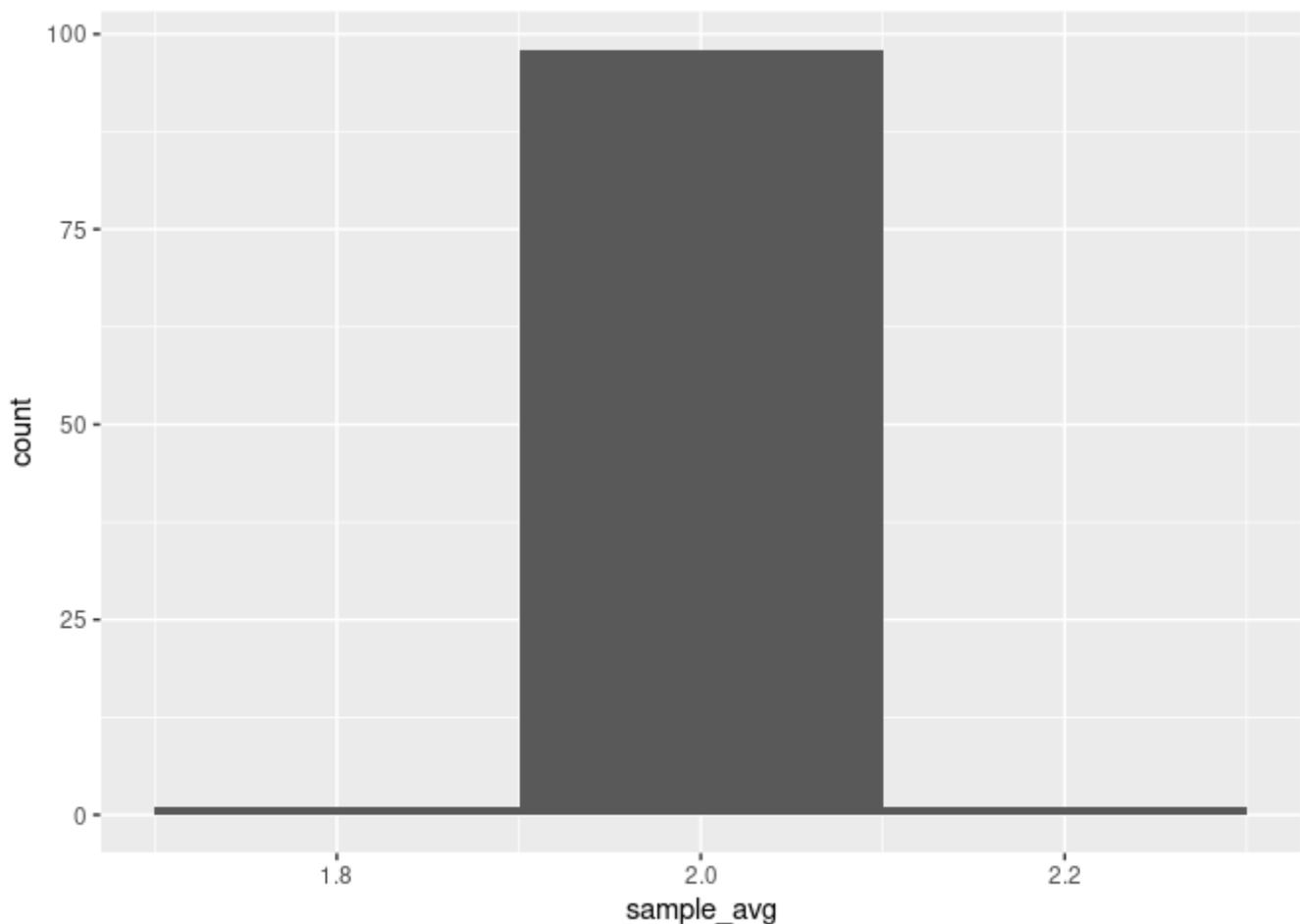
```
alpha<-1/10
beta<-2/10
gamma<-3/10
simulation_by_n<-data.frame(n=seq(0,10000,100))%>%
  mutate(sample_X=map(.x=n,~sample_X_0125(.x,alpha,beta,gamma)))%>%
  mutate(sample_avg=map_dbl(.x=sample_X,~mean(.x)))%>%
  select(-sample_X)%>%
  mutate(expectation=alpha+2*beta+5*gamma)
simulation_by_n%>%head(5)
```

```
##      n sample_avg expectation
## 1    0         NaN          2
## 2 100    2.140000          2
## 3 200    1.885000          2
## 4 300    1.943333          2
## 5 400    2.007500          2
```

1.

```
plotit<- ggplot(simulation_by_n, aes(x=sample_avg)) + geom_histogram(binwidth=0.2)
plotit
```

```
## Warning: Removed 1 rows containing non-finite values (stat_bin).
```



8.

```
expectation<- alpha + 2*beta + 5*gamma
print(expectation)
```

```
## [1] 2
```

```
var(sample_X)
```

```
## [1] 4.40424
```

## B.3 Exponential Distribution.

1. Formula for population mean and variance

Population mean for exponential random variable X with parameter  $\lambda = \frac{1}{\lambda}$

Variance for exponential variable random variable X with parameter  $\lambda = \frac{2}{\lambda^2}$

2. Formula for cumulative distribution function and quantile function for exponential random variables with parameter  $\lambda$

Cummulative function is given by =

$$F\lambda(x) = \begin{cases} 0, & \text{if } x \leq 0 \\ 1 - e^{-\lambda x}, & \text{if } x > 0 \end{cases}$$

Quantile function is given by =

$$F^{-1}\lambda(x) = \inf\{x \in \mathbb{R} : F\lambda(x) \leq p\} = \begin{cases} -\infty, & \text{if } p=0 \\ \frac{1}{\lambda} \ln(1-p), & \text{if } p \in (0,1) \end{cases}$$

- 3.

Maximum Likelihood Estimation =  $\bar{X}$

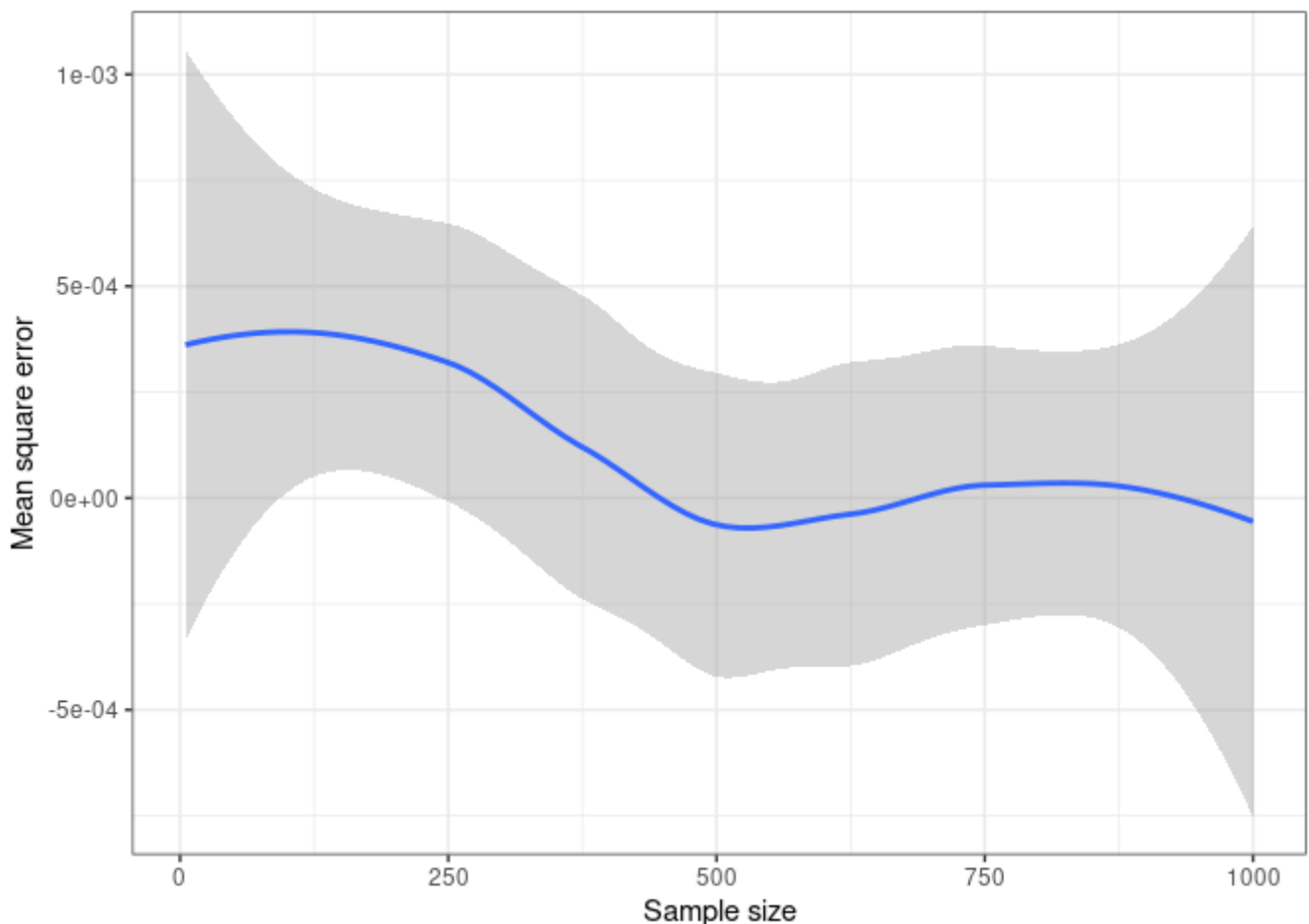
- 4.

```

library(dplyr)
set.seed(0)
num_trials_per_sample_size<-100
min_sample_size<-5
max_sample_size<-1000
sample_size_inc<-5
lambda_0<-0.01
poisson_simulation_df<-expand.grid(trial=seq(num_trials_per_sample_size),
sample_size=seq(min_sample_size,max_sample_size,
sample_size_inc))%>%
# create data frame of all pairs of sample_size and trial
mutate(simulation=pmap(.l=list(trial,sample_size),
.f=~rpois(.y,lambda=lambda_0)))%>%
# simulate sequences of Gaussian random variables
mutate(lambda_mle=map_dbl(.x=simulation,.f=mean))%>%
# compute the sample sd
group_by(sample_size)%>%
summarise(msq_error=mean((lambda_mle-lambda_0)*2))
poisson_simulation_df%>%
ggplot(aes(x=sample_size,y=msq_error))+
geom_smooth()+
theme_bw()+
xlab("Sample size")+ylab("Mean square error")

```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



```
library(readr)
bird_data<-read.csv("bird_data_EMATM0061.csv")

alpha<-0.00000004192201
sample_size<-length(bird_data$Time)
sample_mean<-mean(bird_data$Time)
sample_sd<-sd(bird_data$Time)
t<-qt(1-alpha/2,df=sample_size-1)
confidence_interval_l<-sample_mean-t*sample_sd/sqrt(sample_size)
confidence_interval_u<-sample_mean+t*sample_sd/sqrt(sample_size)
confidence_interval<-c(confidence_interval_l,confidence_interval_u)
print(confidence_interval)
```

```
## [1] 1476693 1554413
```

```
percentage<-(confidence_interval[1] / confidence_interval[2]) * 100
print(percentage)
```

```
## [1] 95
```

```
print("Confidence level of 95% achieved for alpha=0.00000004192201")
```

```
## [1] "Confidence level of 95% achieved for alpha=0.00000004192201"
```