Exercises I3 Exercise 3.1

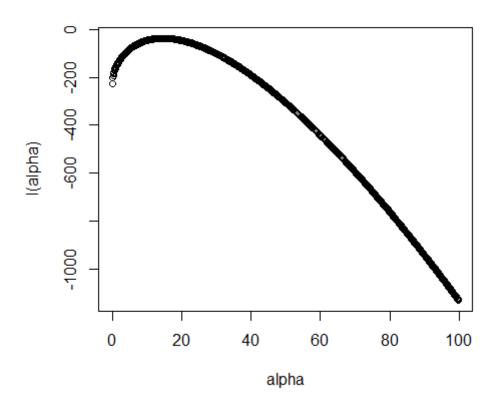
i) With the R script

```
o = c(5.1, 7.4, 10.9, 21.3, 12.3, 15.4, 25.4, 18.2,
17.4, 22.5)
loglikGamma = c()
alphas = seq(from = 0.01, to = 100, by = 0.1)

for (alpha_i in alphas) {
   loglikGamma =
append(loglikGamma, sum(log(dgamma(o,alpha_i))))
}

plot(alphas,loglikGamma,xlab = "alpha",ylab =
"l(alpha)")
```

we plot the log likelihood of the observed sample:



ii) With the R script

```
l_alpha = function(alphas) {
    loglikGamma = c()
    for (alpha_i in alphas) {
        loglikGamma = append(loglikGamma, sum(log(dgamma(o, alpha_i))))
    }
    return(loglikGamma)
}

derivative_l_alpha = function(alphas) {
    return(numDeriv::grad(l_alpha, alphas))
}
```

```
observed_info = function(alphas) {
  return(-numDeriv::grad(derivative_l_alpha, alphas))
}
newton_raphson = function(starting_value, n_iterations)
{
  alfa = starting_value
  for (i in 0:n_iterations) {
    print(alfa)
    alfa = alfa + (derivative_l_alpha(alfa) /
observed_info(alfa))
  }
  return(alfa)
interval = c(0.01, 100)
uniroot_result = uniroot(derivative_l_alpha, interval =
interval)$root
newton_raphson_result = newton_raphson(1, 100)
```

and we get 14.54997 using uniroot and 14.54997 using my implementation of Newton-Raphson with 100 iterations.

iii) We can use the <code>observed_info</code> function we defined previously to get $J(\hat{\alpha}) = 0.711444$

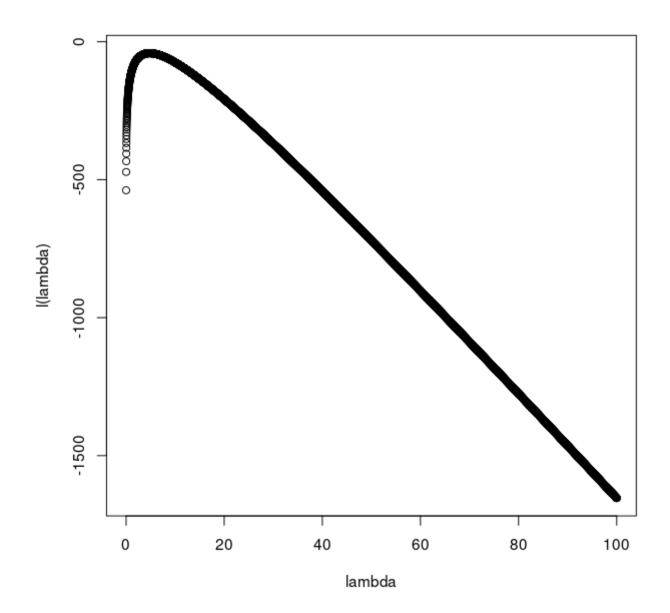
iv)...

Exercise 3.2

i) With:

```
o = c(7, 4, 2, 4, 3, 2, 5, 10, 7, 7, 3, 5, 5, 5, 4, 3,
7, 3, 6, 4)
samples=o
l_lambda = function(lambda) {
  loglikPoi = c()
  for (lambda_i in lambda) {
    loglikPoi = append(loglikPoi,
sum(log(dpois(samples, lambda_i))))
  }
  return(loglikPoi)
}
lambdas = seq(from = 0.01, to = 100, by = 0.01)
plot(lambdas,l_lambda(lambdas),xlab = "lambda",ylab =
"l(lambda)")
```

we plot:



and using again uniroot we find that $\hat{lpha}=4.80001$

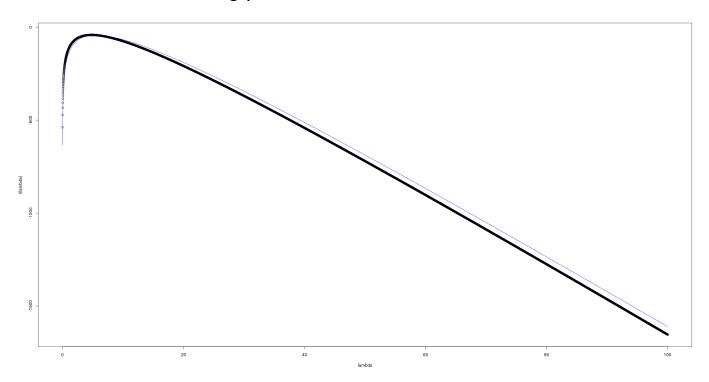
iii) with the script

```
o = c(7, 4, 2, 4, 3, 2, 5, 10, 7, 7, 3, 5, 5, 5, 4, 3,
7, 3, 6, 4)
samples=o

l_lambda = function(lambda) {
    loglikPoi = c()
```

```
for (lambda_i in lambda) {
    loglikPoi = append(loglikPoi,
sum(log(dpois(samples, lambda_i))))
  }
  return(loglikPoi)
}
lambdas = seq(from = 0.01, to = 100, by = 0.01)
plot(lambdas,l_lambda(lambdas),xlab = "lambda",ylab =
"l(lambda)")
derivative_l_lambda = function(lambda) {
  return(numDeriv::grad(l_lambda, lambda))
}
interval = c(0.01, 100)
uniroot_result = uniroot(derivative_l_lambda, interval
= interval)$root
resampled_o = sample(o,20,replace = TRUE)
samples=resampled_o
lines(lambdas,l_lambda(lambdas),col="blue")
```

we obtain the following plot:



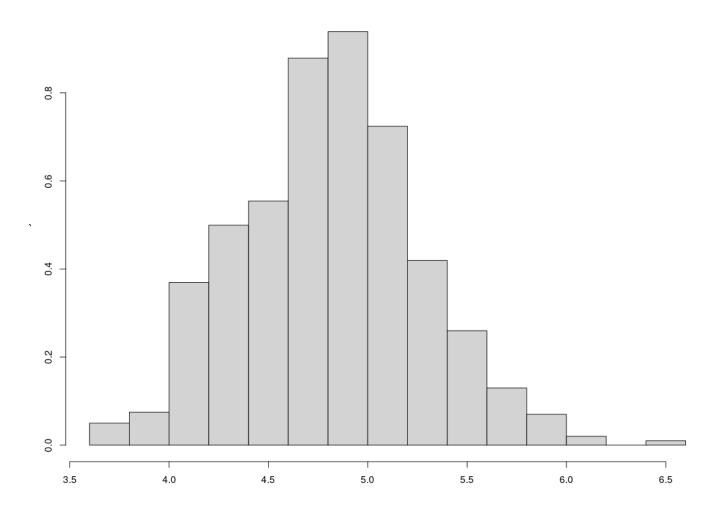
where the blue line is the one obtained drawing 20 numbers with replacement from the sample.

iii) With the script:

```
maximum_likelihood_vector = c()
for(i in (0:1000)){
    resampled_o = sample(o,20,replace = TRUE)
    ML = uniroot(function(lambda)
    derivative_l_lambda(lambda,resampled_o), interval = interval)$root

maximum_likelihood_vector=append(maximum_likelihood_vector,ML)
}
hist(maximum_likelihood_vector,freq = FALSE)
```

Histogram of maximum_likelihood_vector



Exercise 3.3

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