Question 5

Steps for expressing the MLE are as follows:

- a) Expressing the joint probability distribution as a product assuming independence $P(X \mid 1, X \mid 2, ..., X \mid n) = P(X \mid 1) P(X \mid 2) * * * * P(X \mid n)$
- b) The joint probability distribution becomes the likelihood function $L(\theta)$ $L(\theta) = P(\theta \ 1) P(\theta \ 2) * * * P(\theta \ n)$

$$L(\theta) = \prod_{i=1}^{n} P(\theta_i)$$

c) Taking the natural log of the likelihood function becomes the log likelihood function $l(\theta) = log(L(\theta))$

$$l(\theta) = log(\prod_{i=0}^{n} P(\theta_i))$$

d) The value of θ that renders a maximum value of the log likelihood function becomes the maximum likelihood estimator

$$l'(\hat{\theta}) = 0 = \frac{d}{d\hat{\theta}}log(\prod_{i=0}^{n} P(\hat{\theta}_i))$$

e) To see if θ is the max estimator (Notice that we are deriving the function twice.)

$$l''(\hat{\theta}) = \frac{d^2}{d\hat{\theta}^2} log(\prod_{i=0}^n P(\hat{\theta}_i)) < 0$$

1. <u>https://newonlinecourses.science.psu.edu/stat414/node/191/</u>. Refer this link for further clarification. As we are finding the maximum likelihood estimator $\lambda^{\hat{}}$ using probability density function for the Poisson distribution:

$$f(x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

Hence start with

$$L(\lambda) = \prod_{i} \frac{\lambda^{x_i} e^{-\lambda}}{x!} \ i \in (1, n)$$

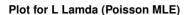
$$l(\lambda) = log(\prod_i \frac{\lambda^{x_i} e^{-\lambda}}{x!}) \ i \in (1, n)$$

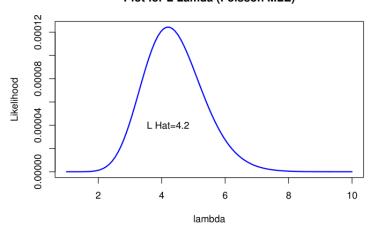
Then obtain

and finally get
$$\hat{\lambda}$$
 by solving $\frac{dl(\lambda)}{d\lambda} = 0$

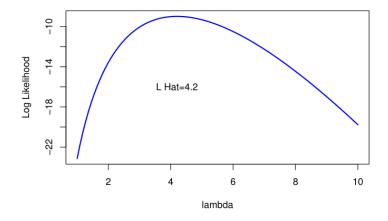
- 2. By performing a double derivate on the function l(y)
- 3. Show that this second derivative is < 0. (Note: Use to fact that log is a monotonically increasing function to justify your answer). Alternatively, to prove $\lambda^{\hat{}}$ is indeed maximum we need to show second derivative of $l(\lambda)$ is negative when evaluated at $\lambda = mean \times (bar)$.
- 4. Show that the maximum of $L(\theta)$ is equal to the maximum of $l(\theta)$
- 5. Skeleton of the function

```
myml = function(x)
{
# Create lambda vector as a set of values ranging from 0 to 2*max(x)
lambda=seq(.01,2*max(x),0.5)
# Calculate Poisson MLE
lambdahat = round(sum(x)/length(x),4)
#the likelihood function
lik <- exp(-length(x)*lambda)*(lambda^sum(x))/prod(factorial(x))
#the loglikelihood function
loglik <- -length(x)*lambda+sum(x)*log(lambda)+log(prod(factorial(x)))
return(list(mle=lambdahat.....))
Expected Plots :</pre>
```





Plot for Ln L Lamda (Poisson MLE)



Question 6

Hint: From results above we found $\hat{\lambda} = \overline{x}$, the sample mean. We need to prove $E(x) = \lambda$. Start with proving for a random variable $Y \sim Poisson(\lambda)$, $E(Y) = \lambda$

Question 8

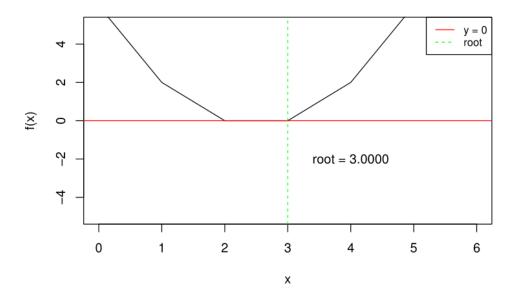
mynr() Uses the Newton-Raphson method to calculate a root. As inputs, it requires a function to be passed, a derivative to be passed, an initial guess, a tolerance, and a maximum number of iterations.

or

https://rpubs.com/aaronsc32/newton-raphson-method

Expected Answer:

Newton Raphson Method



φ - (x) - (