### **Homework 3 Question Math Formula Part**

#### **Poisson Distribution**

Event: must happen in the limited space and time.

Example:

Eg1 One year time, the specific size of meteors which are dropped on the earth.

Eg2 Numbers of patients who visited the emergency room between specific times such as 10:00 pm - 11:00pm.

Eg3 Given an ellismate to the starbuck how many sandwiches should be prepared for the morning section in a specific time period.

**Build Model Basic Requirement:** 

- 1. K should be a variable which represents the events happening time, k range is 0 to n.
- 2. Each event should be an independently event
- 3. The constant value which can represent a probability in the same time period should be the same.
- 4. Just one event happened at the same time.
- 5. There is a ratio into probability and time period.
- 6. The binomial distribution should exist in the Poisson Distribution.

Calculate the probability for the lambda:

We can use lambda to represent the specific period the event happened probability.

### Here is the Poisson Distribution Form:

$$p(X|\lambda) = \frac{\exp^{-\lambda} \lambda^{x_i}}{x_i!}$$

We need to derive the formula for this rule and Elismate the max likelihood from this formula.

Here are the steps for derivation of this formula.

## Step1:

We writing out the likelihood function(L), let  $\theta = \lambda$ 

$$L(\theta) = \prod_{i=1}^{N} \frac{e^{-\lambda} \lambda^{x_i}}{x_i!}$$

So the formula will change to

# Step2:

We write the log toe of the function by taking the log.

$$LL(\theta) = \log \left( \prod_{i=1}^{N} \frac{e^{-\lambda} \lambda^{x_i}}{x_i!} \right)$$

Step3:

Let us use some simplifications for the formula.

$$LL(\theta) = \log \left( \prod_{i=1}^{N} \frac{e^{-\lambda} \lambda^{x_i}}{x_i!} \right)$$

$$= \sum_{i=1}^{N} \log \left( \frac{e^{-\lambda} \lambda^{x_i}}{x_i!} \right)$$

Because we have some log properties:

$$\log \prod_{i}^{N} x_i = \sum_{i}^{N} \log x_i$$

Then we can use quotient rule, power rule to simplify the formula again.

$$= \sum_{i=1}^{N} \log \left( \frac{e^{-\lambda} \lambda^{x_i}}{x_i!} \right)$$

$$= \sum_{i=1}^{N} [\log (e^{-\lambda}) + \log(\lambda^{x_i}) - \log (x_i!)]$$

$$= \sum_{i=1}^{N} \left[ -\lambda \log \left( e \right) + x_{i} \log(\lambda) - \log \left( x_{i} ! \right) \right]$$

$$= \sum_{i=1}^{N} \left[ -\lambda + x_i \log(\lambda) - \log(x_i!) \right]$$

### Step4:

Let us use summation rules to simplify depth.

## (Using online resources:

https://www.mathdoubts.com/constant-multiple-rule-of-derivatives-proof/ https://tutorial.math.lamar.edu/classes/calci/summationnotation.aspx https://dwstockburger.com/Introbook/sbk10.htm

$$LL(\theta) = \sum_{i=1}^{N} [-\lambda + x_i \log(\lambda) - \log(x_i!)]$$

Rules we using(refer to the links):

$$= \sum_{i=1}^{N} -\lambda + \sum_{i=1}^{N} x_i \log (\lambda) - \sum_{i=1}^{N} \log (x_i!) \qquad \sum_{i=1}^{n} (a_i + b_i) = \sum_{i=1}^{n} a_i + \sum_{i=1}^{n} b_i \qquad \sum_{i=1}^{n} (a_i - b_i) = \sum_{i=1}^{n} a_i - \sum_{i=1}^{n} b_i$$

$$= -\lambda N + \log(\lambda) \sum_{i=1}^{N} x_i - \sum_{i=1}^{N} \log(x_i!) \qquad \sum_{i=1}^{n} ca_i = c * \sum_{i=1}^{n} a_i$$

$$\sum_{i=1}^n c = n * c$$

# Step5:

Calculate the derivative of the log likelihood function with the lambda.

$$LL(\theta) = -\lambda N + \log{(\lambda)} \sum_{i=1}^{N} x_i - \sum_{i=1}^{N} \log{(x_i!)}$$

$$\frac{\partial LL(\theta)}{\partial \lambda} = -1 * N + \frac{1}{\lambda} \sum_{i=1}^{N} x_i$$

We treat derivative of  $\lambda$  with respect  $\lambda$  is 1.  $\log(\lambda)$  with respect to  $\lambda$  is  $1/\lambda$ . We can drop the last summation because there is no  $\lambda$  in there.