

# MAP 531: Homework

*Paul-Antoine GIRARD & Adrien TOULOUSE*

## Problem 1: Estimating parameters of a Poisson distribution to model the number of goals scored in football

We recall that the Poisson distribution with parameter  $\theta > 0$  has a pdf given by  $(p(\theta, k), k \in \mathbb{N})$  w.r.t the counting measure on  $\mathbb{N}$ :

$$p(\theta, k) = e^{-\theta} \frac{\theta^k}{k!}$$

### Question 1:

The poisson distribution is a discrete distribution since it has a countable number of possible values ( $\mathbb{N}$ ).

In statistics, we use this distribution to compute the probability of a given number of (rare) events in a time period or the probability of a discrete waiting time until the next event (eg. number of minutes).

For example a poisson distribution can model:

- The number of patients arriving in an emergency room between 9 and 10am.
- The number of minutes we wait a bus at the bus stop.
- In quality control, the number of manufacturing defect.

### Question 2:

We assume that  $\mathbb{X}$  follows a Poisson distribution with parameter  $\theta > 0$ .

$$\begin{aligned}\mathbb{E}[\mathbb{X}] &= \sum_{i=0}^{\infty} (i * p(\theta, i)) = \sum_{i=0}^{\infty} (i * e^{-\theta} \frac{\theta^i}{i!}) = \theta * e^{-\theta} \sum_{i=1}^{\infty} (\frac{\theta^{i-1}}{(i-1)!}) = \theta * e^{-\theta} \sum_{i=0}^{\infty} (\frac{\theta^i}{i!}) = \theta * e^{-\theta} * e^{\theta} = \theta \\ \mathbb{E}[\mathbb{X}^2] &= \sum_{i=0}^{\infty} (i^2 * p(\theta, i)) = \sum_{i=0}^{\infty} (i^2 * e^{-\theta} \frac{\theta^i}{i!}) = \theta * e^{-\theta} \sum_{i=1}^{\infty} (i \frac{\theta^{i-1}}{(i-1)!}) = \theta * e^{-\theta} \sum_{i=0}^{\infty} ((i+1) \frac{\theta^i}{i!}) = \theta * e^{-\theta} [\sum_{i=0}^{\infty} (i \frac{\theta^i}{i!}) + \sum_{i=0}^{\infty} (\frac{\theta^i}{i!})] = \theta * e^{-\theta} [\theta * e^{\theta} + e^{\theta}] = \theta * e^{-\theta} [\theta + 1] * e^{\theta} = \theta(\theta + 1) \\ \mathbb{V}(\mathbb{X}) &= \mathbb{E}[\mathbb{X}^2] - \mathbb{E}[\mathbb{X}]^2 = \theta(\theta + 1) - \theta^2 = \theta\end{aligned}$$

### Question 3:

We are provided with  $n$  independent observations of a Poisson random variable of parameter  $\theta \in \Theta = \mathbb{R}_+^*$ .

Our observations are  $X_k \sim Pois(\theta), \forall k \in 1, \dots, n$ .

The corresponding statistical model is

$$\mathbb{M} = \{p(\cdot | \theta), \theta \in \Theta\}$$

We are trying to estimate the parameter  $\theta$ .

**Question 4:**

The likelihood function is the function on  $\theta$  that makes our  $n$  observations most likely.

$$l(\theta) = \prod_{k=1}^n p(\theta, x_k) = \prod_{k=1}^n e^{-\theta} \frac{\theta^{x_k}}{x_k!}, \text{ with } x_k \in \mathbb{N}, \forall k \in 1, \dots, n$$

$$L(\theta) = \log(l(\theta)) = \sum_{k=1}^n (-\theta + x_k \log(\theta) - \log(x_k!)) = -n\theta + \log(\theta) \sum_{k=1}^n x_k - \sum_{k=1}^n \log(x_k!)$$

By derivating with respect to  $\theta$ , we have:

$$L'(\theta) = -n + \frac{\sum_{k=1}^n x_k}{\theta}$$

Then, we set this derivative equal to zero to obtain a critical point:

$$L'(\theta) = 0 \Leftrightarrow -n + \frac{\sum_{k=1}^n x_k}{\theta} = 0 \Leftrightarrow \hat{\theta} = \bar{x}$$

and this critical point is a local maximum, and we will assume that it is also a global maximum of the likelihood function:

$$L''(\theta) = -\frac{\sum_{k=1}^n x_k}{\theta^2} < 0$$

So, the maximum likelihood estimator is:

$$\hat{\theta}_{MLE} = \bar{x}$$

**Question 5:**

We have that:

$$\mathbb{E}[\bar{x}] = \frac{1}{n} \sum_{k=1}^n \mathbb{E}[x_k] = \mathbb{E}[x_1] = \theta$$

$$\mathbb{V}(\bar{x}) = \frac{1}{n^2} \sum_{k=1}^n \mathbb{V}(x_k) = \frac{1}{n} \mathbb{V}[x_1] = \frac{\theta}{n}$$

Applying the central limit theorem, we have that  $\sqrt{n}(\hat{\theta}_{MLE} - \theta)$  converges towards a Gaussian  $\mathcal{N}(0, \theta)$ .

**Question 6:**

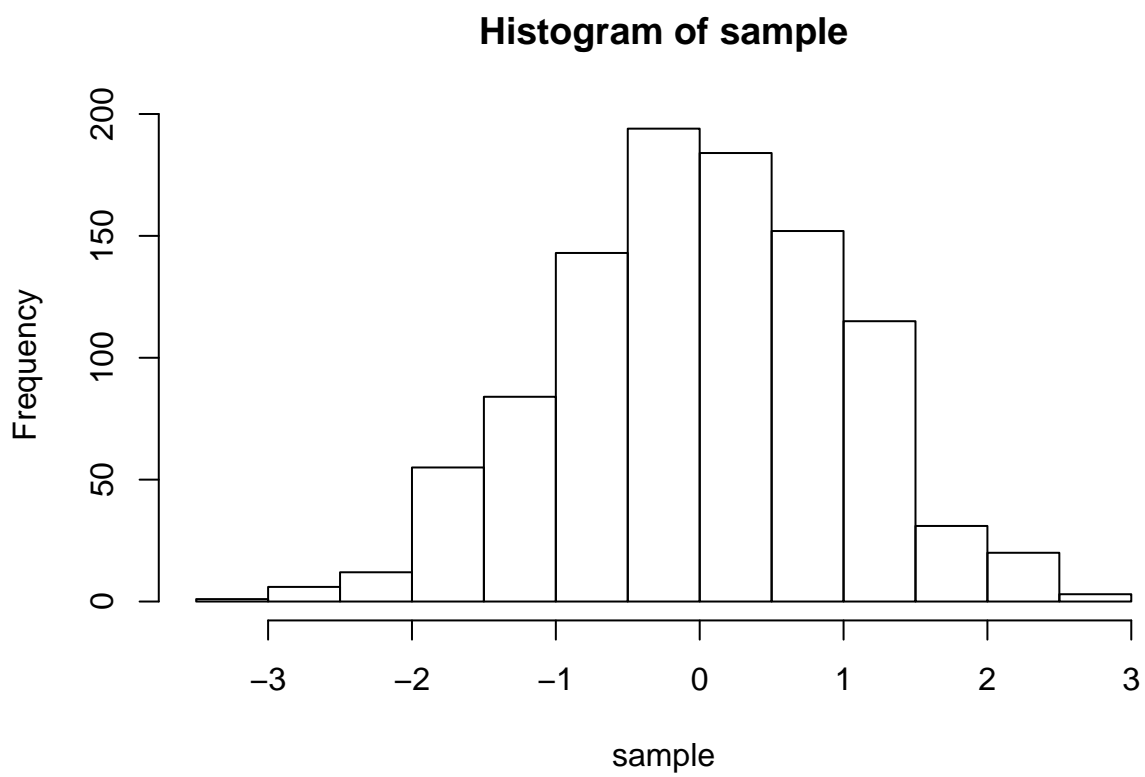
By continuous mapping,  $\sqrt{\hat{\theta}_{MLE}}$  converges in probability towards  $\sqrt{\theta}$ . Then, by Slutsky's theorem, we have that  $\sqrt{n} \frac{(\hat{\theta}_{MLE} - \theta)}{\sqrt{\hat{\theta}_{MLE}}}$  converges in law towards a gaussian  $\mathcal{N}(0, 1)$ .

Let's check this result in R by simulating 1000 times our random variable  $\sqrt{n} \frac{(\hat{\theta}_{MLE} - \theta)}{\sqrt{\hat{\theta}_{MLE}}}$  with a sample size of 100:

```
Nattempts = 1000
nsample = 100
theta = 3
sample = rep(0, 1000)
for (i in 1:Nattempts) # can be written without the for loop (nicer) !
{poisson_sample = rpois(nsample, theta)
  sample[i] = sqrt(nsample) * (mean(poisson_sample) - theta) / sqrt(mean(poisson_sample))}
```

```
}
```

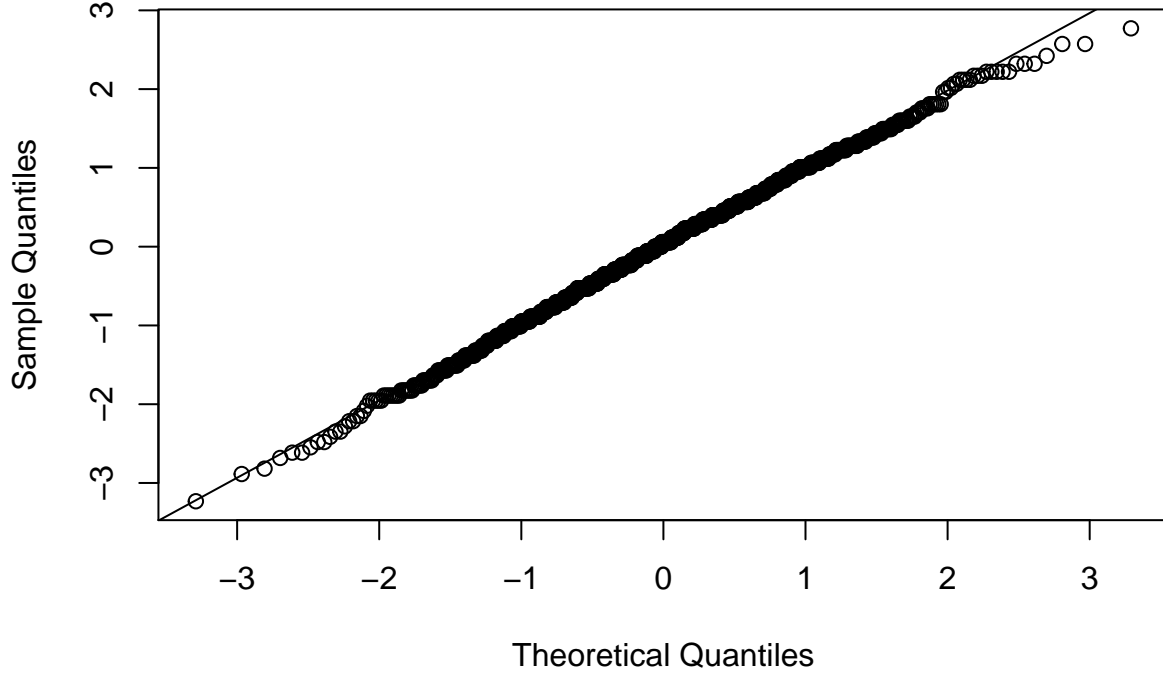
```
hist(sample)
```



```
qqnorm(sample)
```

```
qqline(sample)
```

## Normal Q-Q Plot



### Question 7:

Let  $Z_n$  be our random variable, so that  $Z_n = \sqrt{n} \frac{(\hat{\theta}_{MLE} - \theta)}{\sqrt{\hat{\theta}_{MLE}}}$

$$\mathbb{P}(-z_{1-\alpha/2} \leq Z_n \leq z_{1-\alpha/2}) = 1 - \alpha \Leftrightarrow \mathbb{P}\left(-z_{1-\alpha/2} \sqrt{\frac{\hat{\theta}_{MLE}}{n}} \leq \hat{\theta}_{MLE} - \theta \leq z_{1-\alpha/2} \sqrt{\frac{\hat{\theta}_{MLE}}{n}}\right) = 1 - \alpha$$

For  $\alpha \in (0, 1)$ , an asymptotic confidence interval for  $\theta$  of level  $\alpha$  is therefore :

$$\left[ \hat{\theta}_{MLE} - z_{1-\alpha/2} \frac{\sqrt{\hat{\theta}_{MLE}}}{\sqrt{n}}; \hat{\theta}_{MLE} + z_{1-\alpha/2} \frac{\sqrt{\hat{\theta}_{MLE}}}{\sqrt{n}} \right]$$

### Question 8:

We apply the  $\delta$ -method with  $g(x) = 2 \times \sqrt{x}$  We have:  $g'(x) = \frac{1}{\sqrt{x}}$

So,

$$\sqrt{n}(\hat{\theta}_{MLE} - \theta) \xrightarrow{d} \mathcal{N}(0, g'(\theta)^2 \times \theta) \Leftrightarrow \sqrt{n}(\hat{\theta}_{MLE} - \theta) \xrightarrow{d} \mathcal{N}(0, 1)$$

### Question 9:

Let  $Z_n$  be our random variable, so that  $Z_n = \sqrt{n}(2\sqrt{\hat{\theta}_{MLE}} - 2\sqrt{\theta})$

We know that  $Z_n \xrightarrow{d} \mathcal{N}(0, 1)$

$$\mathbb{P}(-z_{1-\alpha/2} \leq Z_n \leq z_{1-\alpha/2}) = 1 - \alpha \Leftrightarrow \mathbb{P}\left(-\frac{z_{1-\alpha/2}}{2\sqrt{n}} \leq \sqrt{\hat{\theta}_{MLE}} - \sqrt{\theta} \leq \frac{z_{1-\alpha/2}}{2\sqrt{n}}\right) = 1 - \alpha$$

$$\Leftrightarrow \mathbb{P}(\sqrt{\hat{\theta}_{MLE}} - \frac{z_{1-\alpha/2}}{2\sqrt{n}} \leq \sqrt{\theta} \leq \sqrt{\hat{\theta}_{MLE}} + \frac{z_{1-\alpha/2}}{2\sqrt{n}}) = 1 - \alpha$$

For  $\alpha \in (0, 1)$ , an asymptotic confidence interval for  $\theta$  of level  $\alpha$  is therefore :

$$[\hat{\theta}_{MLE} - z_{1-\alpha/2} \frac{\sqrt{\hat{\theta}_{MLE}}}{\sqrt{n}}; \hat{\theta}_{MLE} + z_{1-\alpha/2} \frac{\sqrt{\hat{\theta}_{MLE}}}{\sqrt{n}}]$$

For  $\alpha \in (0, 1)$ , an asymptotic confidence interval for  $\theta$  of level  $\alpha$  is therefore :

$$[**]$$

### Question 10:

Based on the first moment of a poisson distribution, we easily have that:

$$\hat{\theta}_{MME} = \bar{x}$$

We then remark that  $\text{hat}\theta_{MME} = \text{hat}\theta_{MLE}$

Based on the second moment of a poisson distribution, we have:

$$n^{-1} \sum_{k=1}^n X_k^2 = \hat{\theta}_2(\hat{\theta}_2 + 1)$$

Let's define the function  $h(x) = x(x + 1)$

Its inverse on  $\mathbb{R}_+^*$  is  $h^{-1} = \frac{1}{2}(-1 + \sqrt{4x + 1})$  and then we have that:

$$\hat{\theta}_2 = \frac{1}{2}(-1 + \sqrt{4(n^{-1} \sum_{k=1}^n X_k^2) + 1})$$