

Classical and Quantum Optics

Assignment-2 Answers

Varghese Reji

Problem 1

The python code to solve this question is given here.

To create the upper confidence limit, we use the formula,

$$P(x < x_1|\mu) = 1 - \alpha \quad (1)$$

and for central interval, we use

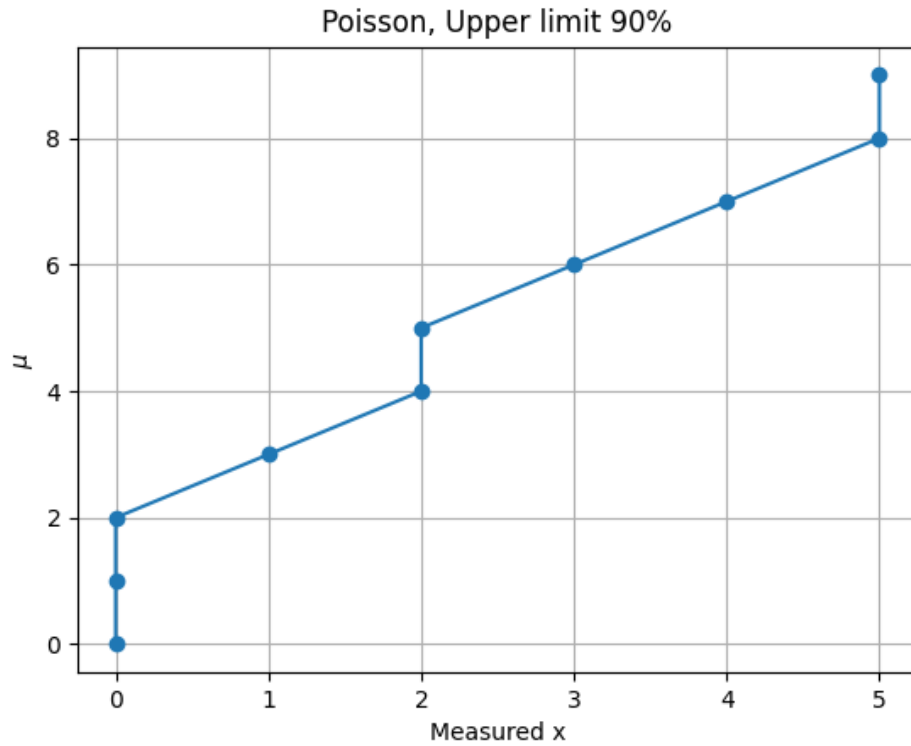
$$P(x < x_1|\mu) = P(x > x_2) = \frac{(1 - \alpha)}{2} \quad (2)$$

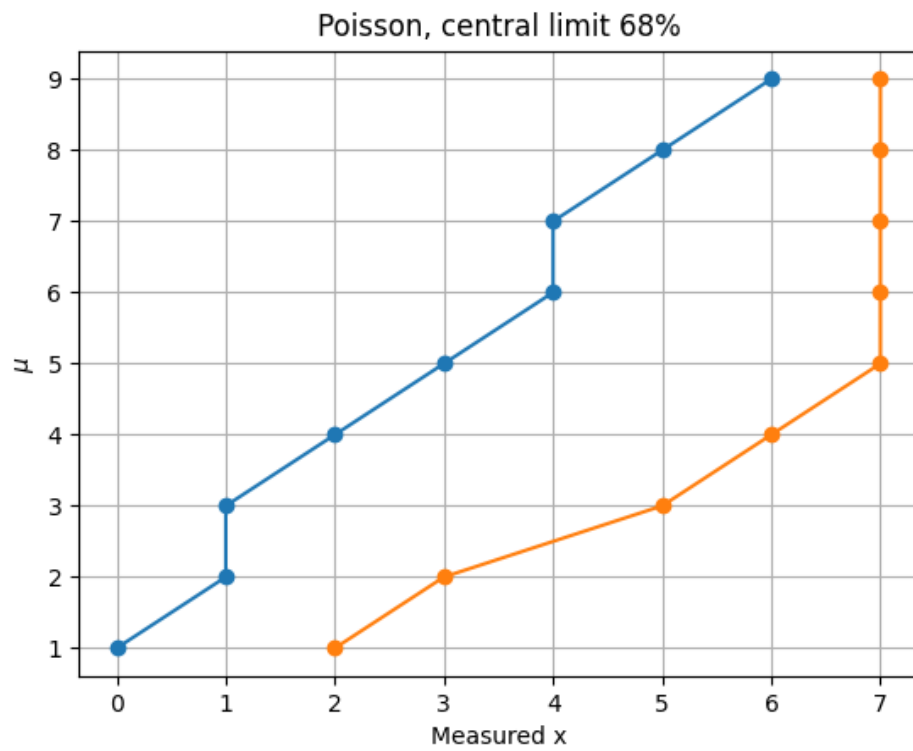
We will take the central interval 68% and upper limit 90%.

(a) Poisson Discrete random variable.

$$P(x|\mu) = \frac{\mu^x}{x!} e^{-\mu} \quad (3)$$

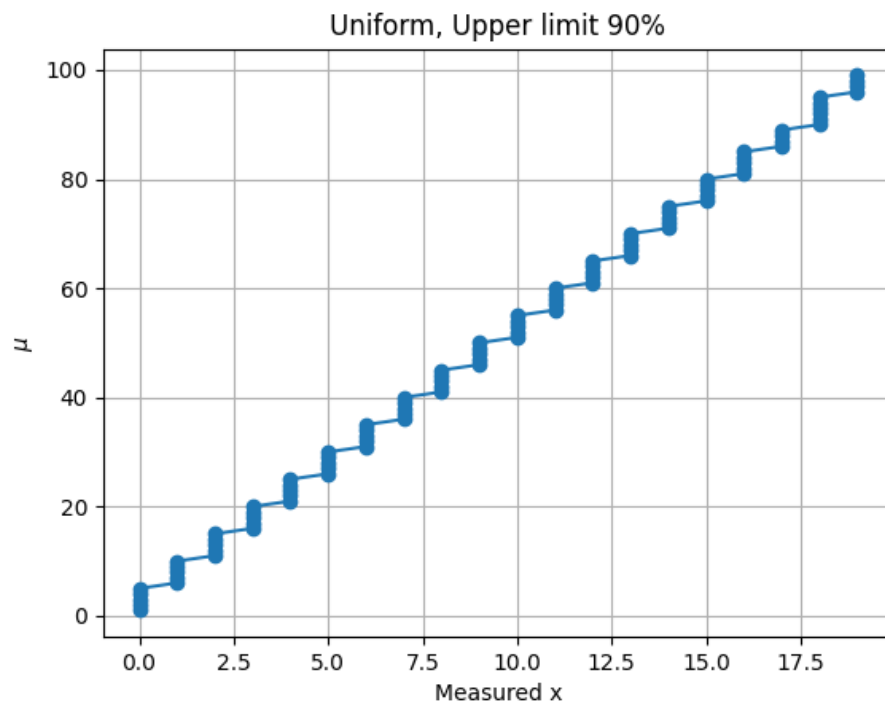
The plots are shown below.

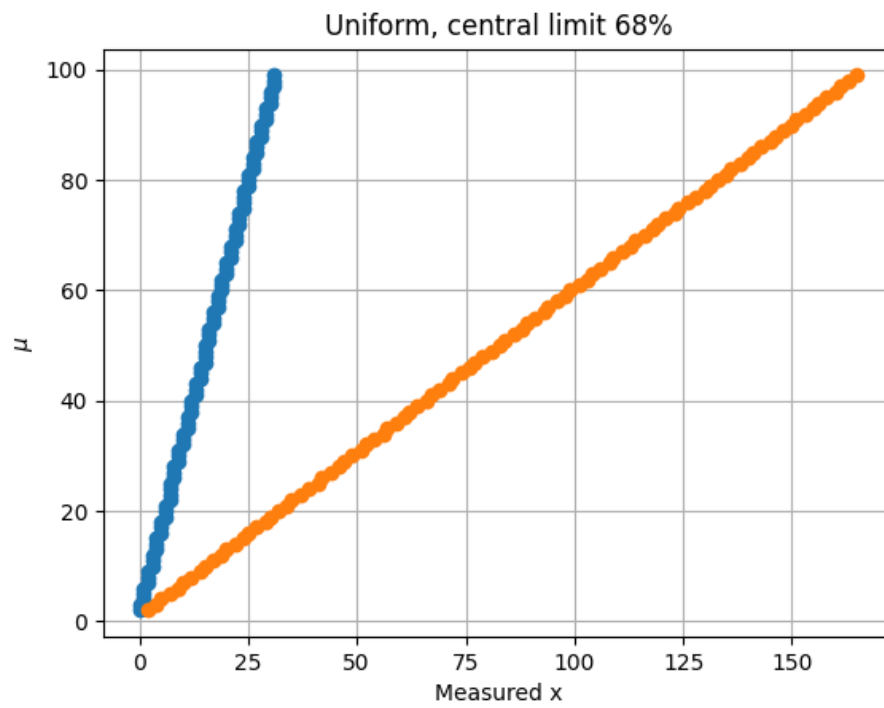




(b) Uniform distribution. Here, $k = 2\mu$.

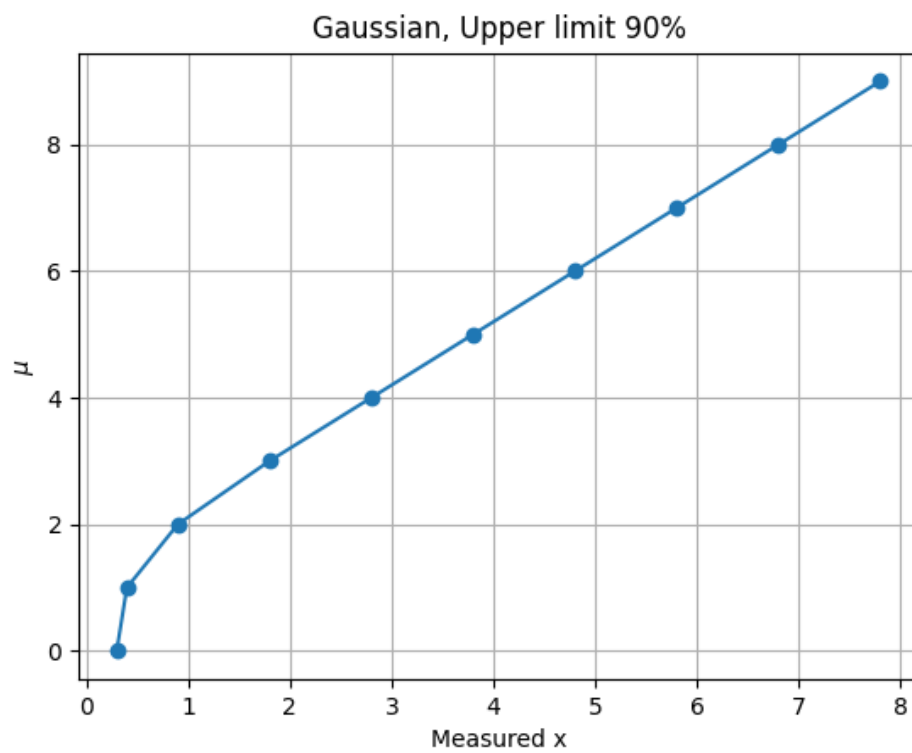
Here, I took $k = 100$. Plots are shown here.

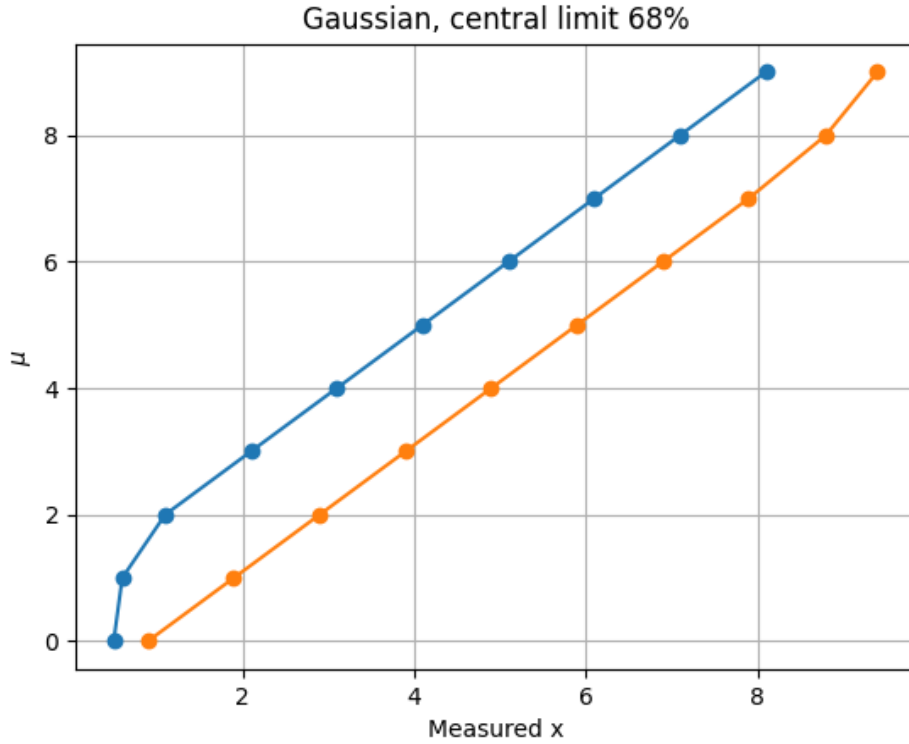




(c) Gaussian function with $\sigma = 1$.

$$P(x|\mu) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(x - \mu)^2}{2}\right) \quad (4)$$





Problem 2

Let us consider an experiment done by Physicist X. He makes the following statements.

- (1) If the result x is less than 3σ , I will state an upper limit from the standard tables. If the result is less than 3σ , I will state a central confidence interval from the standard tables.
- (2) If my measured value of a physically positive quantity is negative, I will pretend that I measured zero when quoting a confidence interval.[1]

The first one is called 'flip-flopping'. Second one will introduce some conservatism. Using these statements, we can make the plot shown in fig 1. For each value of the measured x , we can estimate the segment $[\mu_1, \mu_2]$ by drawing a vertical line. Then we can examine the collection of vertical confidence intervals to see what horizontal acceptance intervals it implies. But in some cases, it does not satisfy the equation

$$P(x \in [x_1, x_2] | \mu) = \alpha \quad (5)$$

Suppose $\mu=2.0$, the acceptance interval has $x_1=2-1.28$ and $x_2=2+1.64$. But this interval only contains 85% of the probability. That means, 5 is not satisfied. The interval is undercover for a significant range of μ : they are not confidence intervals or conservative confidence intervals.

But without flip-flopping, using the second statement only, the result will be unsatisfying when we get x as negative values. In that case, when we draw a vertical line as directed and find that the confidence interval is an empty set. So these are the issues we are facing at the moment, and these can be solved by using ordering principle. That's why ordering principle is relevant.

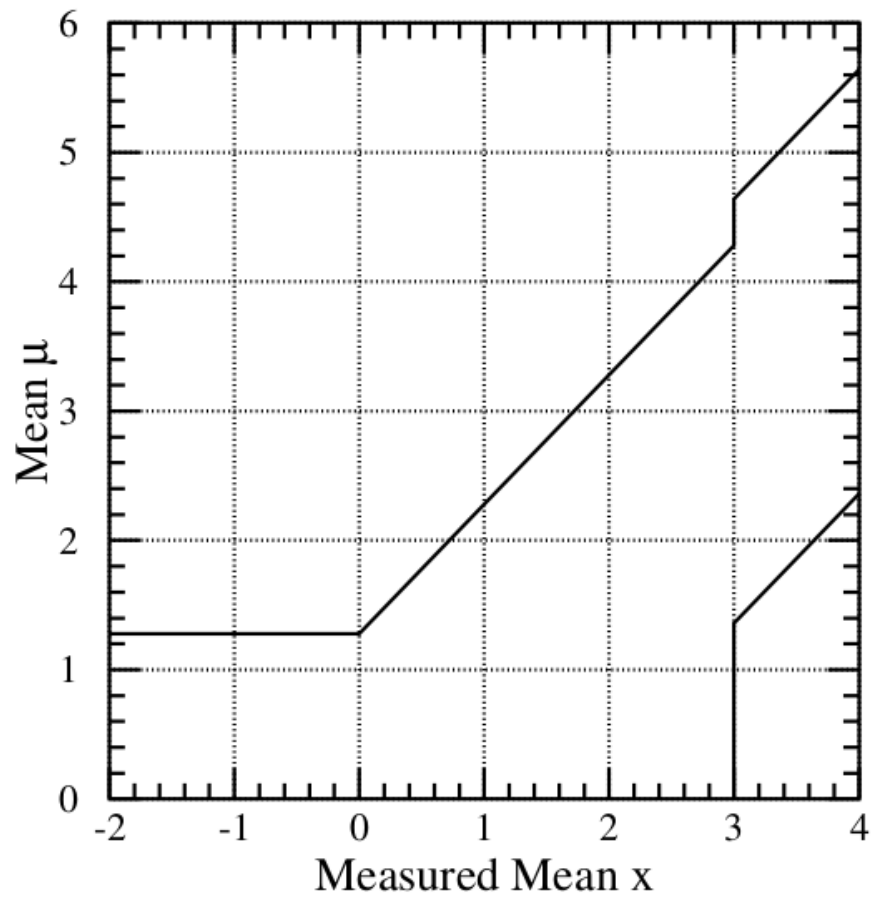


Figure 1: Plot of confidence belts made based on statements. [1]

References

References

- [1] Gary J. Feldman and Robert D. Cousins. Unified approach to the classical statistical analysis of small signals. *Physical Review D*, 57(7):3873–3889, Apr 1998.