

## Exercises - Introduction to Probability and Statistics

- (1) For the following probability density

$$P(x) = xe^{-x} \quad 0 \leq x < \infty$$

- (a) Find the mean and standard deviation of  $x$ . What is the probability content in the interval (mean-standard deviation, mean+standard deviation).
- (b) Find the median and 68 % central interval
- (c) Find the mode and 68 % smallest interval

- (2) Consider the data in the table: Starting with a flat prior for each energy, find an

Energy	Trials	Successes
0.5	100	0
1.0	100	4
1.5	100	20
2.0	100	58
2.5	100	92
3.0	1000	987
3.5	1000	995
4.0	1000	998

estimate for the efficiency (success parameter  $p$ ) as well as an uncertainty. For the estimate of the parameter, take the mode of the posterior probability for  $p$  and use the smallest interval to find the 68 % probability range. Make a plot of the result.

- (3) Analyze the data in the table from a frequentist perspective by finding the 90 % confidence level interval for  $p$  as a function of energy. Use the Central Interval to find the 90 % CL interval for  $p$ .
- (4) 9 events are observed in an experiment modeled with a Poisson probability distribution.
- (a) What is the 95 % probability lower limit on the Poisson expectation value  $\nu$  ?  
Take a flat prior for your calculations.
  - (b) What is the 68 % confidence level interval for  $\nu$  using the Neyman construction and the smallest interval definition?

- (5) Repeat the previous exercise, assuming you had a known background of 3.2 events.
- Find the Feldman-Cousins 68 % Confidence Level interval
  - Find the Neyman 68 % Confidence Level interval
  - Find the 68 % Credible interval for  $\nu$

- (6) (Only required for the TUM students) In this problem, we look at the relationship between an unbinned likelihood and a binned Poisson probability. We start with a one dimensional density  $f(x|\lambda)$  depending on a parameter  $\lambda$  and defined and normalized in a range  $[a, b]$ .  $n$  events are measured with  $x$  values  $x_i$   $i = 1, \dots, n$ . The unbinned likelihood is defined as the product of the densities

$$\mathcal{L}(\lambda) = \prod_{i=1}^n f(x_i|\lambda).$$

Now we consider that the interval  $[a, b]$  is divided into  $K$  subintervals (bins). Take for the expectation in bin  $j$

$$\nu_j = \int_{\Delta_j} f(x|\lambda) dx$$

where the integral is over the  $x$  range in interval  $j$ , which is denoted as  $\Delta_j$ . Define the probability of the data as the product of the Poisson probabilities in each bin.

We consider the limit  $K \rightarrow \infty$  and, if no two measurements have exactly the same value of  $x$ , then each bin will have either  $n_j = 0$  or  $n_j = 1$  event. Show that this leads to

$$\lim_{K \rightarrow \infty} \prod_{j=1}^K \frac{e^{-\nu_j} \nu_j^{n_j}}{n_j!} \propto \prod_{i=1}^n f(x_i|\lambda).$$

I.e., the unbinned likelihood is proportional to the limit of the product of Poisson probabilities for an infinitely fine binning.

- (7) (Only required for the TUM students) In the *Extended Likelihood* approach, the overall normalization is taken into account so as to have a penalty if the observed number of events is not in line with expectations. Referring to the notation in the previous problem, we define

$$\Lambda = \int_a^b f(x|\lambda) dx$$

The normalized unbinned likelihood is now

$$\mathcal{L}(\lambda)_{\text{normalized}} = \prod_{i=1}^n \frac{f(x_i|\lambda)}{\Lambda}$$

and the extended likelihood is defined as

$$\mathcal{L}(\lambda)_{\text{extended}} = \frac{\Lambda^n e^{-\Lambda}}{n!} \prod_{i=1}^n \frac{f(x_i|\lambda)}{\Lambda}.$$

Compare the value of the unbinned likelihood, the binned Poisson probability, and the binned Poisson probability for the data set in the table for the model  $f(x) = 10x \quad 0 \leq x \leq 1$ . Discuss the results.

Set	$x_1$	$x_2$	$x_3$	$x_4$
1	0.5	0.5	0.5	0.5
2	0.2	0.5	0.75	0.9

(8) In this exercise, we want to fit a Sigmoid function to the following data:

Energy ( $E_i$ )	Trials ( $N_i$ )	Successes ( $r_i$ )
0.5	100	0
1.0	100	4
1.5	100	22
2.0	100	55
2.5	100	80
3.0	100	97
3.5	100	99
4.0	100	99

The Sigmoid function to fit is defined as:

$$s(E|A, E_0) = \frac{1}{1 + e^{-A(E-E_0)}}$$

- Find the posterior probability distribution for the parameters  $(A, E_0)$ . You should clearly state what priors you assumed and your reasoning for the choice.
- Define a suitable test statistic and find the frequentist 68 % Confidence Level region for  $(A, E_0)$ .
- Repeat the analysis of the data with the function

$$\epsilon(E) = \sin(A(E - E_0))$$

Find the posterior probability distribution for the parameters  $(A, E_0)$  and the 68 % CL region for  $(A, E_0)$

- discuss the results