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## Homework Exercises

### Exercise 2 - 1

You are meeting an old friend. As you are catching up, he tells you that he now has two children. What is the probability that these are two brothers?

- a) If you have **no further information**?
- b) If he tells you that **the first born is a boy**?
- c) If he tells you that **at least one of them is a boy**?
- d) If he tells you that **one of them is a boy born on a Monday**?
- e) As in (d), **but a week has  $n$  days**?
- f) Examine the cases  $n = 1$  and  $n = \infty$  and explain why some of the probabilities from (a)-(c) are reproduced! What happens in between these extremes?

### Exercise 2 - 2

Bob flips a (probably) manipulated coin as long as he gets tails. The moment the coin lands with head up, he stops the tossing.  $n$  denotes the number of tails he got. The coins' probability to land with tail up may be  $\mu$ . Bob's strategy is denoted by  $B$ .

- a) Calculate  $\mathcal{P}(n | \mu, B)$ .
- b) Calculate the expected number of tails, i.e.  $E[n]_{\mathcal{P}(n|\mu,B)}$ .
- c) Bob performs one tossing experiment from a) and gets  $n$  tails in a row, which he tells Alice. So far Alice does not know how the experiment was conducted and likes to infer the unfairness of the coin.
  - Until now Alice believes that Bob performed a coin toss experiment of predetermined length  $n + 1$ . This strategy is called  $A$ . Calculate the most probable  $\mu$  using  $\mathcal{P}(\mu | n, A)$ .
  - Now Bob tells Alice that he ended the tossing when he got the first head. She therefore infers the most probable  $\mu$  using  $\mathcal{P}(n | \mu, B)$  and Bayes Theorem. Calculate  $\mathcal{P}(\mu | n, B)$ .
  - Compare the results and discuss whether the finding is surprising.
- d) As Alice knows that the maximum of a non-symmetric probability distribution is not equal to its expectation value she uses a computer algebra system of her choice to plot the probability distribution of  $\mu$ .  
Calculate the posterior mean  $E[\mu]_{\mathcal{P}(\mu|n,B)}$  and compare with your results from b).

### Exercise 2 - 3

- a) Consider some experiment in which we record events that originate from two distinct sources ( $x$  and  $y$ ), each following independently a Poisson distribution with rates  $\lambda_x$  and  $\lambda_y$ , respectively. The detector observes their sum  $z = x + y$ . Show that this random variable also follows a Poisson distribution and express it in terms of  $\lambda_x$  and  $\lambda_y$ .
- b) Next we want to consider a *thinned Poisson process*. Here we have a random number of occurrences,  $N$ , distributed according to a Poisson distribution with rate  $\mu$ . Each of the  $N$  occurrences,  $X_n$ , can take on values of 1, with probability  $\mu$ , or 0, with probability  $(1 - \mu)$ , following

a Bernoulli distribution. We want to derive the probability distribution for

$$X = \sum_{n=1}^N X_n. \quad (1)$$

Show that the probability distribution is given by

$$\mathcal{P}(X) = \frac{e^{-\lambda\mu}(\lambda\mu)^X}{X!}. \quad (2)$$

This scenario, for example, corresponds to experiments that produce some kind of physical event, but the detector only has a certain probability to record an event.

#### Exercise 2 - 4

You have a stick of length 1 and you break it into two pieces. The breaking point appears at a random location.

- What is the average length of the smaller of the two pieces?
- Calculate the average length ratio of the smaller to the larger piece.
- Calculate the average length ratio of the larger to the smaller piece.
- You want to break another stick, but this time you use too much force and it breaks into three pieces. What is the mean length of the smallest piece?

#### Exercise 2 - 5

As a child, Nora always looked forward to getting the toy that was in the cereal box. She opened a new box every week, and looked for the toy. There were  $N$  different kinds of toys and they are distributed equally among the boxes.

- Consider Nora already found  $m$  toys ( $0 \leq m \leq N$ ), what is the probability to find a new toy in the next box?
- Given the  $m$  toys, what is the expectation value of boxes to find the next new toy?
- What is the expectation value for the number weeks until she had at least one of each kind of toy?

#### Exercise 2 - 6

The goal of this exercise is the implementation of the linear regression of a curve. The data consists of multiple pairs of  $(x, y)$  and we assume  $y$  to be a noisy measurement of some underlying function  $y = f(x) + n$ . We assume independent, additive, Gaussian noise  $n \sim \mathcal{N}(n|0, N)$  with known variance  $N$  (e.g.  $N = 0.1$ , but could depend on the chosen function  $f(x)$ ). We want to learn this function in terms of the polynomial basis  $f(x) = \sum_{i=0}^M a_i x^i$  up to some order  $M$  in terms of the coefficients  $a$ . A priori we assume the coefficients to follow a Gaussian distribution  $a \sim \mathcal{P}(a) = \mathcal{N}(a|0, A)$  with unit variance ( $A = \mathbb{1}$ ).

- We start by generating data according to the described process. Choose some (reasonably smooth) function  $f(x)$  as ground-truth (e.g.  $\sin(x)$ ,  $e^x$ , some polynomial, combinations of those). Generate several sampling locations  $x$  for this function (either randomly or equidistant, maybe 10 – 30 locations). Generate the corresponding  $y$  by applying the function to  $x$  and add a noise contribution  $n$ , drawn from the corresponding Gaussian.

- b) Plot the obtained data points together with the true function and indicate the error bars in  $y$ -direction in terms of the standard deviation.
- c) Implement the prior covariance  $A$ , the noise covariance  $N$ , and the design matrix  $R_{ij} = x_i^j$  as matrices. (exact definition, see script)
- d) Implement the posterior covariance  $D = (R^T N^{-1} R + A^{-1})^{-1}$  and calculate  $j = R^T N^{-1} y$
- e) Use the posterior covariance to calculate the posterior mean of the coefficients  $m = D R^T N^{-1} y$ .
- f) Plot the function corresponding to the mean coefficients  $m$  together with the data and ground-truth.
- g) Draw several posterior samples from the Gaussian  $a \sim \mathcal{N}(a|m, D)$  and add the corresponding functions to the previous plot (thin, transparent lines) to illustrate the uncertainty.
- h) Now we want to determine which polynomial order  $M$  we should use. Calculate evidences for polynomial orders  $M \in [0, \dots, 10]$  and compare their values. What is the best polynomial order for your setup? Plot the evidence for the different  $M$ .

Hint: The evidence  $P(y|M)$  is:

$$\mathcal{P}(y|M) = \frac{|2\pi D|^{\frac{1}{2}}}{|2\pi N|^{\frac{1}{2}} |2\pi A|^{\frac{1}{2}}} e^{-\frac{1}{2} y^T N^{-1} y + \frac{1}{2} j^T D j}, \text{ with} \quad (3)$$

$$D = (R^T N^{-1} R + A^{-1})^{-1} \quad \text{and} \quad j = R^T N^{-1} y \quad (4)$$

- i) How does this optimal order depend on the noise level? Repeat the same analysis for several noise covariances (and corresponding data) and plot your results.