

Weekly Exercise 4

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Probability distributions

(8 Points)

Exercise 1 (Probability distribution, 2 points). We throw two “fair” dice. Let us define a random variable X as the sum of the numbers showing on the dice. Define and draw the cumulative distribution function F_X .

Exercise 2 (Density function, 1 point). Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be a function defined as follows

$$f(x) = \begin{cases} x, & \text{if } 0 < x < 1, \\ 2 - x, & \text{if } 1 < x < 2, \\ 0, & \text{otherwise.} \end{cases}$$

Is it possible that f is a density function?

Exercise 3 (Random variable and expectation, 2 points). Let X be a discrete random variable with the possible values of 1, 2 and 3, where the corresponding probabilities are given as

$$P(X = 1) = \frac{1}{3}, \quad P(X = 2) = \frac{1}{2}, \quad P(X = 3) = \frac{1}{6}.$$

- Define and draw the cumulative distribution function F_X .
- What is the expected value of X ?

Exercise 4 (Random variable and expectation, 3 points). In order to express his gratitude, Siegfried invites Eduard to a pub for a couple of beers. There, they start playing a friendly game of darts. The dart board is a perfect disk of radius 10cm. If a dart falls within 1cm of the center, 100 points are scored. If the dart hits the board between 1 and 3cm from the center, 50 points are scored, if it is at a distance of 3 to 5cm 25 points are scored and if it is further away than 5cm 10 points are scored. As Siegfried and Eduard are both quite experienced dart players, they hit the dart board every time.

- Define a random variable X corresponding to the score of throws.
- What is the expected value of the scores?

Expectation-maximization algorithm**(4 Points)**

Exercise 5 (Expectation-maximization algorithm, 4 points). An alternative route in the derivation of the Expectation-maximization algorithm is to maximize the expected the *log-posterior* $\ln p(\boldsymbol{\theta} \mid \mathbf{X})$ instead of the expected *log-likelihood*. Show that for this case that the M-step yields

$$\boldsymbol{\theta}^{(t+1)} \in \operatorname{argmax}_{\boldsymbol{\theta}} \left(Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{(t)}) + \ln p(\boldsymbol{\theta}) \right) .$$

Therefore, one can assume the prior distribution of the parameters $\boldsymbol{\theta}$ (for example, to avoid singularities).

Hint: one may consider the maximization problem

$$\boldsymbol{\theta}^{(t+1)} \in \operatorname{argmax}_{\boldsymbol{\theta}} \mathbb{E}[\ln p(\boldsymbol{\theta} \mid \mathbf{X}, \mathbf{Z}) \mid \mathbf{X}, \boldsymbol{\theta}^{(t)}] .$$