## Foundations of data science, summer 2020

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## 11. Exercise sheet Hand in solutions until Thursday, 2 July 2020, 12:00

**Exercise 11.1** (Mixture of densities).

(8 points)

Suppose you are given some random variables  $X^{(i)} \stackrel{\text{\tiny def}}{\longleftarrow} \mathbb{R}$  with density  $p_i$ . 8 For the computer scientist: some routine Xi produces samples of  $X^{(i)}$ . How do you construct a routine X that samples acc. to the overlayed density  $\sum_{i < k} w_i p_i$ ? Prove correctness:

**Theorem.** Consider  $X^{(i)} \stackrel{\text{\tiny de}}{\longleftarrow} p_i$  for i < k and  $\hat{\imath} \stackrel{\text{\tiny de}}{\longleftarrow} w$ , reading w as a distribution on  $\mathbb{N}_{< k}$ . Finally, let  $X \leftarrow X^{(\hat{\imath})}$ . Then  $X \sim p$ .

*Hint*:  $X \sim p$  means that p is the density of X, ie. prob  $(X \in [a, b]) = \int_a^b p(x) dx$  for all a < b.

Remark: This generalizes to random variables with other outputs instead of values in  $\mathbb{R}$ .

**Solution.** To show that the density of *X* is *p* means that we have to show

$$\operatorname{prob}\left(X \in [a,b]\right) = \int_{a}^{b} p(x) \, dx$$

by definition. We have

$$X \in [a,b] \Leftrightarrow \exists i < k \colon X^{(i)} \in [a,b] \land w = i.$$

This directly follows from the construction of *X*. So

$$\begin{split} \operatorname{prob}\left(X \in [a,b]\right) &= \sum_{i < k} \operatorname{prob}\left(X \in [a,b] \,\middle|\, \hat{\imath} = i\right) \cdot \operatorname{prob}\left(\hat{\imath} = i\right) \\ &= \sum_{i < k} \operatorname{prob}\left(X^{(i)} \in [a,b] \,\middle|\, \hat{\imath} = i\right) \cdot \operatorname{prob}\left(\hat{\imath} = i\right). \end{split}$$

Since  $X^{(i)}$  and  $\hat{\imath}$  are independent, we have

$$\operatorname{prob}(X^{(i)} \in [a, b] | \hat{i} = i) = \operatorname{prob}(X^{(i)} \in [a, b]).$$

Thus

$$\begin{split} \operatorname{prob}\left(X \in [a,b]\right) &= \sum_{i < k} \operatorname{prob}\left(X^{(i)} \in [a,b]\right) \cdot w_i \\ &= \sum_{i < k} w_i \cdot \int_a^b p_i(x) \ \mathrm{d}x \\ &= \int_a^b \sum_{i < k} w_i p_i(x) \ \mathrm{d}x \\ &= \int_a^b p(x) \ \mathrm{d}x \,, \end{split}$$

which was what we wanted.

## **Exercise 11.2** (Application of the SVD).

(0+13 points)

In this exercise you shall play with the example from

Alex Thomo (2009). Latent Semantic Analysis (Tutorial).

- (i) Reprogram it, denote by k the used dimension.
  - (ii) Examine the resulting ranking if...
    - (a) ... you modify  $k \in \{2, 3, 4, 5\}$ .
    - (b) ... you omit the scaling step.
    - (c) ...you change the selection of words by omitting words that only occur in a single document or by adding more words.
    - (d) ... you use the Euclidean metric instead of the angle metric.

That's a total of at least 24 cases. You need a careful analysis to isolate important insights.

(iii) Redo similar analysis with a larger dataset: You will find documents 11-document\*.txt in the exercises folder, which contain (parts of) the short overviews of some Wikipedia articles.

*Hint*: We expect you to present an analysis with insights, explanations and arguments. So, no large tables or thelike.

**Solution** (Hints). There is no optimal solution to this exercise, because your choices in the implementation and the sheer possibilities in experimenting can lead to a lot of different results, that might not necessarily be always true. The important aspect in a solution are well-planned experiments and a critical treatment of the result.

In general, a higher k leads to more accurate results, but for some queries, even k high did not always lead to the expected solution, as our sample size was quite small.