

Summer term 2023

## Visual Data Analysis

### Assignment Sheet 6

Solution has to be submitted via eCampus by May 12, 2023, 10:00 a.m.

If you have questions concerning the exercises, please use the forum on eCampus.

- Please work on this exercise in **small groups** of 3 students. Submit each solution only once, but clearly indicate who contributed to it by forming a team in eCampus. Remember that all team members have to be able to explain all answers.
- Please submit your answers in PDF format, and your scripts as \*.py/\*.ipynb files. If you are using [Jupyter notebook](#), please also export your scripts and results as PDF.

### Exercise 1 (Graph Visualization, 15 Points)

In this exercise you will learn how to visualize graphs in Python. We suggest the [Graphviz](#) package, which was used for our own example solution. In case you should already know and prefer an alternative Python package that allows you to perform the same tasks, you are free to use it.



Figure 1: Variable correlation graph.

- a) Read the [Breast Cancer Dataset](#) `breast-cancer-wisconsin.xlsx` (from Sheet 3) and fill in the missing values as before. Then compute the Pearson correlation between any pair of variables, and store them in a matrix. (2P)

- b) Create a graph from the correlation matrix and visualize it with a force-directed layout. Represent each variable as a node in the graph. Insert an edge between two variables whenever the Pearson correlation between them exceeds the threshold  $\rho > 0.6$ . (4P)
- c) Modify the visual attributes of edges to reflect the magnitude of the correlation. (3P)
- d) Produce an alternative visualization with a circular layout. Color the nodes so that there are four sets of nodes, one color for having at least one correlation more than 0.9 to other nodes, another for having at least a correlation  $0.8 < \rho_{\max} \leq 0.9$ , one for having a correlation  $0.6 < \rho_{\max} \leq 0.8$  and the last one for the remaining nodes. (3P)
- e) Answer the following questions:
  - At the selected threshold, which nodes are disconnected from the rest of the graph and what do they indicate? (1P)
  - If two nodes A and B are strongly correlated, and node C is strongly correlated with node B, can we conclude that node C will be also strongly correlated with node A? (1P)
  - Based on the visualization, which variables would you propose to predict the class? (1P)

## Exercise 2 (Large-Scale Graph Visualization, 10 Points)

The node-and-link diagrams that were discussed for graph visualization in the lecture do not scale well to large graphs. In this task, you will read about an alternative visualization approach that has been proposed for such cases, based on visualizing an adjacency matrix representation instead. The corresponding paper `elmqvist-zame-2008.pdf` is available from the lecture webpage.

Please answer the following questions in your own words. Remember that **we will not grant even partial credit for copy-pasted text**.

- a) The ZAME visualization tool uses a specific hierarchical data structure for storing graphs at multiple scales. At the lowest level, four integers are stored per vertex, and either six or four per edge. What do these integers describe? Which two are optional in case of the edges, and what is their purpose? (3P)
- b) The zoomable edge table stores edges in a particular order that makes it fast to search for an edge given its vertices. Write efficient pseudocode that returns the index within this table of an edge connecting vertices  $u$  and  $v$ , and returns `None` if the table does not contain such an edge. (4P)
- c) In the pseudocode listed in the paper's Figure 4, some modifications are highlighted in boldface, on lines starting with a bar. What is the purpose of these modifications? (2P)
- d) What is the difference between geometric zoom and detail zoom in the system? (1P)

## Exercise 3 (Interactive Visualization with Dash, 25 Points)

This task introduces you to Dash, a productive Python framework for building web applications. Dash allows you to build data visualization apps with custom user interfaces in pure Python. You should find the Dash tutorial, as well as the documentation chapter on Pattern Matching Callbacks, useful for solving this exercise (<https://dash.plotly.com/>).

- a) Read the file `Data_Cortex_Nuclear.xls` (from Sheet 3), and restrict it to the classes t-CS-s and c-CS-s. Run PCA, ISOMAP and t-SNE as dimensionality reduction techniques. Within the Dash framework, create a single scatter plot that will show the output from one of the techniques, as selected by the user. Add a **Radio** component from Dash to switch between the different techniques. A **callback** function is a function which is triggered by an event such as a mouse click, double click, or selection. In Dash, you can use `@app.callback` to define your event callback, with the dropdown **value** as the input. (10P)

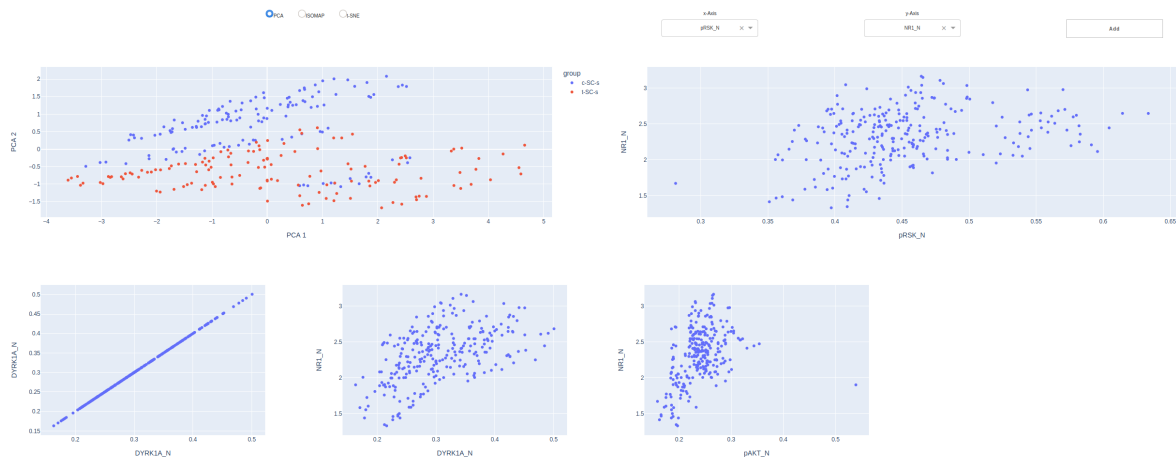


Figure 2: A screenshot of the interactive visualization you are asked to implement in this exercise.

- Add a second scatter plot with corresponding **dropdown** menus that should give you the opportunity to map any individual feature (i.e., protein) to any of the axes. (5P)
- Allow the user to show an arbitrary number of scatterplots by implementing an “Add” button. Clicking on it should replicate the scatterplot that is on the right side of your previously implemented app, keeping the current user-defined axis settings. The new plots should appear below the old ones, as in Figure 2. (10P)

*Hints:*

- If you haven’t looked at States yet - now is the time to do so. See the Basic Callbacks Tutorial.
- All the components in a Div container are stored in its “children” list. (This assignment usually happens automatically when you define your layout.) You can achieve a flexible layout by giving the Div container a component ID and using its “children” list as both a State and an Output for your callback.
- If the callback was triggered by the Add button, you want to append the new plot to the children list of the parent Div.

**Good Luck!**