Homework 1

Foundations of Data Science TA team

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Introduction

This homework contains two parts. In the first part, you will have a chance to get back into programming with Python, as you will look at data on the quality and cost of care in hospitals in the United States of America. In the second part, you will use different visualisation techniques to explore data from a cohort of pediatric patients with suspected appendicitis admitted with abdominal pain to Children's Hospital St. Hedwig in Regensburg, Germany, between 2016 and 2021.

After this homework, you should be able to

- · describe data using pandas' built-in methods
- · identify missing values using pandas' built-in methods
- formulate hypotheses on reasons for missing values
- build on these hypotheses to replace missing values where possible
- explore a dataset with different visualisation techniques using matplotlib and seaborn
- identify the appropriate graph type based on your data and question at hand
- derive practical and actionable insights from data.

1 Recapitulation of Python for Data Science

1.1 Getting an overview

When encountering a new data set, the first step is to get an overview. For this purpose, answer the following questions:

- 1. How many rows and columns are in the data?
- 2. How many distinct hospitals have information reported in the data set? *Hint:* What information do you have to consider to uniquely identify a hospital?
- 3. How many states of the U.S. are represented in the data? Does that match your expectation? Explain.
- 4. Which hospital is the most expensive for *hip and knee* procedures? How much more expensive is this hospital compared to the least expensive?

5. Which hospital is the least expensive considering the sum of average costs of all treatment categories reported in the data? Does this low price have an effect on the quality reported? Explain your reasoning.

1.2 Missing data

- 1. Are there any values missing in the data? If yes, which column(s) are affected?
- 2. If there are column(s) with missing data, propose a way of handling the missing entries (for each column). *Note:* There is no need to actually implement the handling of missing values. Your script file still has to contain evidence of how you arrived at your answer(s), e.g. how you identified or summarised missing values.

1.3 Basic visualisation

- 1. Plot the distribution of the cost for treating pneumonia. Save the figure under the path output/131_pneumonia-cost.png. Don't forget to add units, axis labels and a plot title.
- 2. Plot the distribution of the cost for treating heart failure stratified by the quality of hospitals in this domain. *Note:* There are multiple ways to achieve this; you only have to make sure that all relevant information is represented in the plot. Save the figure under the path output/132_heart-failure-cost-vs-quality.png.

2 Data Visualization

2.1 Data exploration

In order to get an overview of a new dataset, it is a good practice to perform an initial exploratory data analysis. For this purpose, answer the following questions:

- 1. Missing values exploration:
 - (a) Create a function that generates a new dataframe to hold the total number and the percentage of missing values per column.
 - (b) Calculate the number of missing values per variable and the missing variables per patient. Hint: Consider the transpose method of pandas.
 - (c) Visualise the missing values per variable using a barplot and comment on your findings. Save the figure as 211c_mv_variables_barblot.png in the output folder.
 - (d) Visualise the top ten patients that have the most missing variables using a barplot. What do you observe? Save the figure as 211d_mv_patients_barblot.png in the output folder.
- 2. Visualise combinations of continuous and categorical variables: Visualise the age and the height of the patients by sex, as subplots of the same figure, using the appropriate graph type. Justify your choice and comment on your findings. Are your findings in alignment with your expectations? Save the figure as 212_age_height_graph.png in the output folder. Hint: Think of the data types to choose the appropriate graph and consider the study population to comment on the findings. *Note: More than one graph types may be appropriate.*

3. Visualise categorical variables: Visualise two potential target outcomes, i.e., Severity and Diagnosis, as subplots in the same figure, using the appropriate graph type. Justify your choice and comment on your findings. Save the figure as 213_diagnosis_severity_graph. png in the output folder.

2.2 Data Understanding

- 1. Visualise the Body_Temperature by Diagnosis for all patients, using a histogram, a boxplot and a violin plot. Present the three graphs as subplots in the same figure. Comment on your findings. Are there any outliers? Can they be explained? Save the figure as 221_body_temperature.png in the output folder.
- 2. Visualize six selected numerical variables (as defined in skeleton.ipynb) using boxplots. Present the results as subplots in the same figure. Comment on your findings. Which variable has the most outliers? Is there a valid explanation for them? Save the figure as 222_numerical_boxplots in the output folder.

2.3 Explore potential risk factors

- 1. Visualise CRP against WBC_Count by Diagnosis and Severity using scatterplot graphs. Present the two graphs as subplots in the same figure. Save the figure as 231_scatterplot. png in the output folder. Hint: The data points on each graph should have different colors based on Diagnosis and Severity respectively. Note: In most cases, appendicitis is presented with increased inflammation markers, such as CRP and WBC count.
- 2. Create a new variable combining the values of Contralateral_Rebound_Tenderness and Lower_Right_Abd_Pain. You may name the new variable Contralateral_Rebound _Tenderness-Lower_Right_Abd_Pain. Then, visualize the pairwise distribution of counts for Diagnosis against this new variable using a heatmap. Comment on your findings. Save the figure as 232_heatmap.png in the output folder. Hint: Consider the pandas crosstab method for creating the new variable. Note: Pain in the lower right abdomen is a common symptom of appendicitis and contralateral rebound tenderness is indicative of inflammation of the abdomen (more specifically in the peritoneum) that can be caused by acute appendicitis among other causes.

Deliverables

You are asked to submit a **report** (two to five pages, **must be in PDF format**, acknowledgements of AI tools used do not count to page limit), and both the .py file and the .ipynb file completed by you to solve the homework. The script file and the jupyter notebook should contain all steps necessary to arrive at the answer to individual questions. Do not hardcode values, which are printed to the terminal. Your answers should be computed in the script and still be correct if the data changed. The report can be organised in sections similar to this question sheet and should include all plots you are asked to create or which you decide to create to help answer a question. Where necessary, add an interpretation to the plot. All questions, for which you should explictly include the answer in your report are highlighted. The Python script and the Jupyter notebook submitted have to run without having to make any modifications. Do not use absolute filepaths. We provide you with a .zip file which contains all the necessary files and the directory structure required. You can assume that the working directory during execution of your submitted script is the hw01/code/ directory in which skeleton.py is stored. Make sure that you save the plots you are asked to generate to the hw01/output/ directory. Once you have completed the homework you can submit the same directory, including the completed script, your report, and all plots as a .zip file again.

Data Description

The first data set you work with in this homework contains information on hospitals in the United States in which Medicare patients received treatment. The data set contains information on

- Hospital Name and Location: name, city, state, state abbreviation
- Ratings of Hospital Performance: ratings across multiple dimensions (Mortality, Safety, etc.) relative to the national average (below, same as, above national average), as well as an overall rating from 1 (worst) to 5 (best),
- **Information on Cost, Quality and Value**: multiple categories of treatments (heart attack, heart failure, hip and knee surgery, pneumonia).

The second dataset you work with in this homework contains information on paediatric patients with suspected appendicitis abdominal pain to Children's Hospital St. Hedwig in Regensburg, Germany, between 2016 and 2021 [1]. The dataset at hand consists of a subset from the original data and contains information on:

- Patient Demographics (5 variables): age, height, weight, BMI, sex.
- **Physical Examination** (6 variables): Alvarado Score (AS), Paedriatic Appendicitis Score (PAS), lower right abdominal pain, contralateral rebound tenderness, nausea, body temperature.
- **Ultrasound (US) information** (4 variables): US performed, US identifier, appendix appeared on US, appendix diameter.
- Lab tests (2 variables): white blood cell (WBC) count, C-reactive protein (CRP) levels.
- **Disease outcomes** (2 variables): diagnosis and severity.

A more detailed description of the data variables from the second dataset can be found in table 1.

Python environment

To be able to solve this homework, you need to have a working Python 3.13 installation and the following packages installed in the environment in which you execute the script:

- matplotlib (3.10)
- numpy (2.2)
- pandas (2.2)
- scipy (1.15)
- seaborn (0.13)
- jupyter (1.1)

The versions specified were used to create the homework. If you encounter problems related to a particular package, we strongly suggest you install the version specified here as a first step and check if the issue persists.

References

[1] Marcinkevičs, R., Reis Wolfertstetter, P., Klimiene, U., Ozkan, E., Chin-Cheong, K., Paschke, A., Zerres, J., Denzinger, M., Niederberger, D., Wellmann, S., Knorr, C., & Vogt, J. E. (2023). *Regensburg Pediatric Appendicitis Dataset (1.01)* [Data set]. Zenodo. https://doi.org/10.5281/zenodo.7669442

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Category	Variable	Description
	age	Patient's age in years.
	weight	Patient's weight in kg.
Demographics	height	Patient's height in cm.
	BMI	Patient's BMI in kg/m².
	Sex	Patient's gender (by birth).
	Alvarado_Score	Alvarado score (AS): Patient's score according to scoring system.
	Dandriatic Appendicitie Corre	Paediatric Appendicitis Score (PAS): Patient's score according
Dhyrical Exam	ו מפען זמרור_אף/פוועורו רוז_טרטן פ	to the scoring system.
r III yalcalı Exalii	Lower_Right_Abd_Pain	Right iliac fossa pain detected on palpation.
	00:10N	Feeling of sickness/ejection of contents from stomach through
	ואמטפים	the mouth.
	100000 Tourist Tourist	Pain of the contralateral side is felt on the release of
	Collei atacei at_neboulid_i elidei lless	pressure over the abdomen.
	Body_Temperature	Patient's temperature in °C (rectum or auditory canal).
	US_Performed	If an abdominal ultrasonography was performed or not.
Illtracont Info	US_Number	Ultrasound identifier.
Oluasouila muo	Appendix_on_US	Detectability of the vermiform appendix during sonographic examination.
	Appendix_Diameter	Maximal outer diameter of the appendix.
	**************************************	Number of white blodd cells in a unit volume of blood
Lab tests	אם ל	inflammation parameter $(\times 10^3)$.
	aac	C-reactive protein, elevated in case of inflammation,
	כאר	infection, or injury (mg/L).
		Patient's diagnosis, histologically confirmed for operated patients.
Disease outcomes	Diagnosis	Conservatively managed patients were labelled as having appendicitis
Discase outcomes		if they had an AS or PAS of ≥ 4 and an appendix diameter of ≥ 6 mm.
	· · · · · · · · · · · · · · · · · · ·	Severity of appendicitis: uncomplicated: subacute/ catharral, fibrosis;
	Sevel 1 Ly	phlegmonous or complicated: gangrenous, perforated, abscessed.

Table 1: Category and description of the patient variables[1]. Information adapted from the UCI Machine Learning Repository