Exploratory Data Analysis (EDA) for Patient Diagnostic

Sarthak Khandelwal

1. Introduction

The Jupyter Notebook titled **eda.ipynb** conducts an **Exploratory Data Analysis (EDA)** on a patient diagnostic dataset. The analysis includes loading the data, cleaning and exploring demographic and diagnostic data, identifying a prospective study cohort, and investigating the relationship between hypertension and diabetes.

2. Loading and Cleaning

The script begins by loading three datasets: **demo.csv**, **dx.csv**, and **dx_map.csv**. The data is loaded using the Pandas library, and initial information is displayed using the info() method. It is noted that **birth_date** and **death_date** columns need to be converted to datetime format, which is promptly addressed.

3. Descriptive Statistics

Descriptive statistics are provided for both demographic and diagnostic data using the describe() method. The 5 number summary statistics help understand the central tendency and distribution of the data.

4. Data Merging

The three datasets are merged into a single dataframe (df_merged) using patient IDs and diagnostic codes. Duplicate columns are removed, and appropriate column renaming is performed.

5. Prospective Study Cohort

A prospective study cohort is identified, consisting of living patients aged 18 or older. The script calculates age for each patient and filters the dataset accordingly. A box plot is used to visualize the age distribution, revealing an outlier aged 129. This outlier is removed based on the assumption of erroneous data.



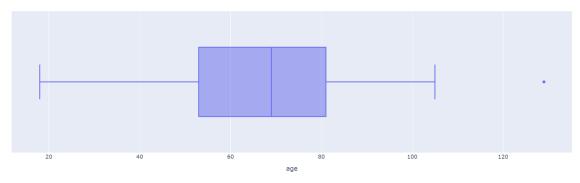
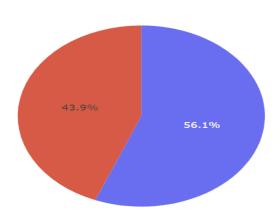


Fig.1: Age Distribution of the cohort of living patients

6. Data Exploration

The script explores missing data, removes unnecessary columns, checks for duplicates, and corrects case sensitivity in the gender column. Visualization includes a pie chart of gender distribution, bar charts for site and dx method distribution, and a box plot for age distribution.



Females have more health problems

Fig.2: Gender Distribution of the Cohort

7. Relationship Between Hypertension and Diabetes

The script identifies patients with hypertension, diabetes, or both by filtering the dataset. A bar chart visually represents the overlap between patients with hypertension and diabetes. Additionally, a count of patients with both diagnoses is calculated, resulting in 158 individuals.

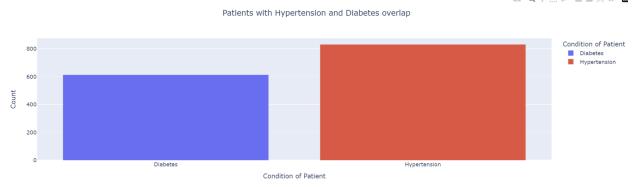


Fig.3: Distribution of Patients with Diabetes and Hypertension

8. Age Distribution for Diabetes and Hypertension

Two separate dataframes (Diabetes_df and Hypertension_df) are created for patients with diabetes and hypertension, respectively. Histograms are plotted to visualize the age distribution for both conditions.

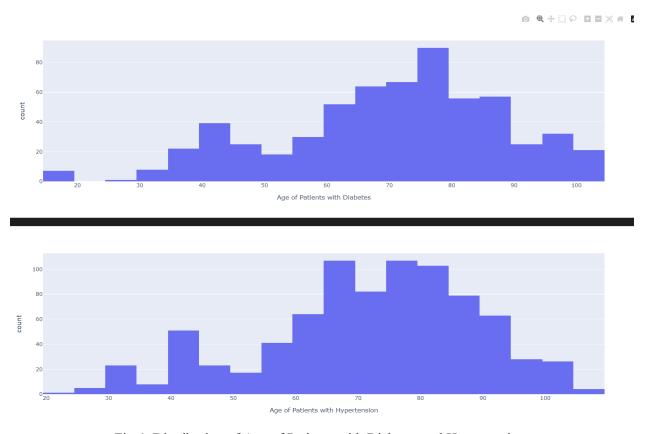


Fig.4: Distribution of Age of Patients with Diabetes and Hypertension

9. Conclusion

The EDA script successfully explores the patient diagnostic dataset, identifies a prospective study cohort, and investigates the relationship between hypertension and diabetes. Visualizations and statistical summaries enhance the understanding of the data, providing valuable insights for further analysis and decision-making.

10. Future Work

Future work includes tasks such as examining the number of people with hypertension and diabetes compared to those with other diseases. This could lead to a more comprehensive understanding of the dataset and potential correlations between various health conditions.