

Data Science Intern Case Study Summary

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1. Exploratory Data Analysis (EDA):

At this stage, the dimensions of the data set, column names and data types, missing data were examined. Basic statistical summary was created. Categorical variables were identified and the unique values and their numbers in each categorical column were examined. Data that may have a relationship between them were examined with graphics. Pandas, Matplotlib and Seaborn libraries were used. More detailed visualization was applied after the preprocessing step.

Coursel sütumlen isin özet istatistikleni

| | | Sayısa | l sütunlar için özet istatisti | cler: | | | |
|------------------------------------|----------------|------------|---|---------------|--------------------------|----------------------------------|-----|
| | | | Kullanici_id | Dogum_Tarihi | \ | | |
| | | count | 2357.000000 | 2357 | | | |
| | | mean | n 1.000000 1939-10-12 00:00:00 % 47.000000 1959-02-05 00:00:00 | | | Eksik veri analizi: | |
| | | min | | | | Kullanici id | 0 |
| | | 25% 50% | | | | Cinsiyet | 778 |
| Veri setinin boyutları: (2357, 19) | | | 97.000000 1973-09-09 00:00:00 | | | 9 | |
| Sütunlar ve veri tipleri: | | 75% | 146.000000 1992-03-24 00:00:00 196.000000 2011-04-25 00:00:00 | | Dogum_Tarihi | | |
| Kullanici_id | int64 | max | 57.017200 NaN | | Uyruk | 0 | |
| Cinsiyet | object | Stu | 57.01/200 | IVdIV | | Il | 227 |
| Dogum_Tarihi | datetime64[ns] | | Ilac Baslangic Tarihi | 1 | lac Bitis Tarihi \ | Ilac_Adi | 0 |
| Uyruk | object | count | 2357 | | 2357 | Ilac Baslangic Tarihi | 0 |
| Il | object | mean | 2022-01-07 10:47:36.173101312 | 2022-03-10 16 | :25:27.365294848 | Ilac Bitis Tarihi | 9 |
| Ilac_Adi | object | min | 2022-01-01 00:00:00 | | 2-03-02 00:00:00 | Yan Etki | 9 |
| Ilac_Baslangic_Tarihi | datetime64[ns] | 25% | 2022-01-04 00:00:00 2022-03-06 00:00:00 | | | (179) | |
| Ilac_Bitis_Tarihi | datetime64[ns] | 50% | 2022-01-07 00:00:00 2022-03-11 00:00:00 | | Yan_Etki_Bildirim_Tarihi | 0 | |
| Yan_Etki | object | 75% | 2022-01-11 00:00:00 2022-03-15 00:00:00 | | Alerjilerim | 484 | |
| Yan_Etki_Bildirim_Tarihi | datetime64[ns] | max | 2022-01-14 00:00:00 2022-03-19 00:00:00 | | Kronik Hastaliklarim | 392 | |
| Alerjilerim | object | std | NaN | | NaN | Baba Kronik Hastaliklari | 156 |
| Kronik Hastaliklarim | object | | Yan Etki Bildirim Tarihi | Kilo | Boy | Anne Kronik Hastaliklari | 217 |
| Baba Kronik Hastaliklari | object | count | 2357 | | 2243.000000 | Kiz Kardes Kronik Hastaliklari | 97 |
| Anne Kronik Hastaliklari | object | mean | 2022-02-10 17:09:30.742044928 | 80.863857 | 174.638431 | Erkek Kardes Kronik Hastaliklari | 121 |
| Kiz Kardes Kronik Hastaliklari | object | min | 2022-02-01 04:34:33 | 50.000000 | 145.000000 | | |
| Erkek Kardes Kronik Hastaliklari | object | 25% | 2022-02-04 05:29:20 | 65.000000 | 160.000000 | Kan Grubu | 347 |
| Kan Grubu | object | 50% | 2022-02-09 20:53:54 | 83.000000 | 176.000000 | Kilo | 293 |
| Kilo | float64 | 75% | 2022-02-17 07:08:01 | 96.000000 | 187.000000 | Boy | 114 |
| Boy | float64 | max | 2022-02-19 21:47:39 | 110.000000 | 203.000000 | dtype: int64 | |
| dtype: object | | std | NaN | 18.635269 | 16.516552 | deype. Incor | |

Total 2 different categories, Column: Cinsiyet

Total 1 different category, Column: Uyruk

Total 13 different categories, Column: Il

Total 151 different categories, Column: İlaç_Adi

Total 22 different categories, Column: Yan_Etki

Total 28 different categories, Column: Alerjilerim

Total 80 different categories, Column: Kronik Hastaliklarim

Total 92 different categories, Column: Baba Kronik Hastaliklari

Total 84 different categories, Column: Anne Kronik Hastaliklari

Total 85 different categories, Column: Kiz Kardes Kronik Hastaliklari

Total 90 different categories, Column: Erkek Kardes Kronik Hastaliklari

Total 8 different categories. Column: Kan Grubu

When the graphs were examined:

It was determined that people with the B rh(-) blood group had a different taste in their mouth as a side effect, people with the AB rh(-) blood group had high blood pressure, people with the B rh(+) blood group had a different taste in their mouth, people with the 0 rh(-) blood group had a different taste in their mouth, people with the AB rh(+) blood group felt tired, people with the 0rh(+) blood group had high blood pressure, people with the A rh(+) blood group had high blood pressure, and people with the A rh(-) blood group had a feeling of tiredness as a side effect. The most common side effect was bruising in Çanakkale, fatigue in Trabzon, high blood pressure in Adana, a different taste in the mouth in İzmir, a different taste in the mouth in Mersin, high blood pressure in Antalya, a different taste in the mouth in Eskişehir, high blood pressure in Samsun, high blood pressure in Ankara, fatigue in Bursa, fatigue in İstanbul, fatigue, a different taste in the mouth and weakness in Malatya, and finally blurred vision in Kayseri. And like these examples, the relationship between drug name and side effect, the relationship between chronic diseases and side effects, the relationship between allergies and side effects, and the relationship between blood type and

chronic diseases have also been made interpretable. These relationships should be tested with a detailed regression analysis and models can be created for specific groups.

2. Data Pre-Processing:

The numpy library was used at this stage. The missing values in the "Il" column were filled with the most common values. The column was deleted because the values in the "Uyruk" column were the same for everyone. The "Kullanıcı id" column was deleted. The column was deleted because there was too much missing data in the "Cinsiyet" column and it could be misleading. The missing values in the "Boy" and "Kilo" columns were filled by taking their averages. The "İlaç Kullanım Süresi" column was created using the drug start and end dates. The duration of side effects was calculated using the drug start date and side effect notification dates. The "Yan Etkilerin Süresi" column was created. The body mass index was calculated using the weight and height values and a new column named "Vki" was created. The missing values in the "Kan Grubu" column were filled with the most frequently used values. A new column named "Yaş" was created using the values in the date of birth column. For categorical values, the unique values examined in the EDA phase were listed in an ordered manner by creating a mapping dictionary. This process was applied for the columns "İlaç Adı", "Yan Etki", "Alerji", "Il", "Kan Grubu", "Kronik Hastaliklari", "Baba Kronik Hastaliklar", "Anne Kronik Hastaliklari", "Kiz Kardes Kronik Hastaliklari", "Erkek Kardes Kronik Hastaliklari". For chronic disease columns, all missing nan values were deleted at these stages. New columns were created and columns belonging to old values were deleted. New values consisting entirely of numerical data were converted to integers. Date columns were updated in date format. Finally a new dataset was created.

3. Data Visualization:

In order to better analyze the new dataset, the relationships between the values were examined with various graphs. The relationship between drug name and chronic diseases was visualized with the heat map, the relationship between blood group and side effects according to drug name was visualized with the facetgrid, the relationship between age and duration of drug use was visualized with scatterplot, the most common drug and side effect combinations were visualized with barplot, the relationship between weight and height and side effects was visualized with scatterplot, the most common allergy and side effect combinations were visualized with barplot, blood group and side effect combinations were visualized with barplot, the relationship between drug name and chronic diseases was visualized with heat map.

4. Future Plan:

In the next step; Comparisons can be made with the visualized relationships in the EDA stage and preprocessing stage, and they can be grouped and evaluated. Time series analysis can be performed. Future predictions can be made by interpreting with various machine learning and deep learning methods. In this way, the analyzed data can be made guiding with meaningful relationships.