Statistical Machine Learning Approaches to Liver Disease Prediction

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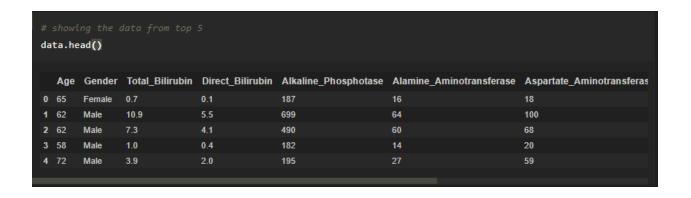**TeamMember:** Sakthisubashini

Exploratory Data Analysis

The exploratory data analysis (EDA) notebook is designed to assist you with discovering patterns in data, checking data sanity, and summarizing there levant data for predictive models.

The EDA notebook example was optimized with web-based data in mindand consists of two parts. Part one starts with using Query Service to view trends and data snapshots. Next, with a goal in mind for exploratory data analysis, the data is aggregated at the profile and visitor level.

Part two starts by performing descriptive analysis on aggregated data using Python libraries. This note book show cases visualizations such as histograms, scatterplots, boxplots and a correlation matrix to derive actionable insights used to determine which features are most likely to be helpful in predicting a goal.

[4]:		Year	Month	Count_days	First_date	Last_date	Count_hits
	0	2020	1	1	31	31	117060
	1	2020	2	29	1	29	3503948

head(): To check the first five n rows of the dataset, we have a function called **head( ).**

```
# showing the data from top 5
data.head()
```

	Age	Gender	Total_Bilirubin	Direct_Bilirubin	Alkaline_Phosphotase	Alamine_Aminotransferase	Aspartate_Aminotransferas
0	65	Female	0.7	0.1	187	16	18
1	62	Male	10.9	5.5	699	64	100
2	62	Male	7.3	4.1	490	60	68
3	58	Male	1.0	0.4	182	14	20
4	72	Male	3.9	2.0	195	27	59

Tail(): To check the last five rows of the dataset, we have a function called **tail().**

```
data.tail()
```

	Age	Gender	Total_Bilirubin	Direct_Bilirubin	Alkaline_Phosphotase	Alamine_Aminotransferase	Aspartate_Aminotransfer
578	60	Male	0.5	0.1	500	20	34
579	40	Male	0.6	0.1	98	35	31
580	52	Male	0.8	0.2	245	48	49
581	31	Male	1.3	0.5	184	29	32
582	38	Male	1.0	0.3	216	21	24

Will see how our dataset is, by using the info() method.

```
data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 583 entries, 0 to 582
Data columns (total 11 columns):
 #   Column                      Non-Null Count  Dtype
---  ------                      --------------  -----
 0   Age                         583 non-null    int64
 1   Gender                      583 non-null    object
 2   Total_Bilirubin             583 non-null    float64
 3   Direct_Bilirubin            583 non-null    float64
 4   Alkaline_Phosphotase        583 non-null    int64
 5   Alamine_Aminotransferase    583 non-null    int64
 6   Aspartate_Aminotransferase  583 non-null    int64
 7   Total_Protiens              583 non-null    float64
 8   Albumin                     583 non-null    float64
 9   Albumin_and_Globulin_Ratio  579 non-null    float64
 10  Dataset                     583 non-null    int64
dtypes: float64(5), int64(5), object(1)
memory usage: 50.2+ KB
```

describe(): functions are used to compute values like count, mean, standard deviation and IQR (InterQuantile Ranges) and give a summary of numeric type data.

```
data.describe()
```

```
data.describe()
```

	Age	Total_Bilirubin	Direct_Bilirubin	Alkaline_Phosphotase	Alamine_Aminotransferase	Aspartate_Aminotransfera
count	583.000000	583.000000	583.000000	583.000000	583.000000	583.000000
mean	44.746141	3.298799	1.486106	290.576329	80.713551	109.910806
std	16.189833	6.209522	2.808498	242.937989	182.620356	288.918529
min	4.000000	0.400000	0.100000	63.000000	10.000000	10.000000
25%	33.000000	0.800000	0.200000	175.500000	23.000000	25.000000
50%	45.000000	1.000000	0.300000	208.000000	35.000000	42.000000
75%	58.000000	2.600000	1.300000	298.000000	60.500000	87.000000
max	90.000000	75.000000	19.700000	2110.000000	2000.000000	4929.000000