510 DATA SCIENCE

Lecture 04

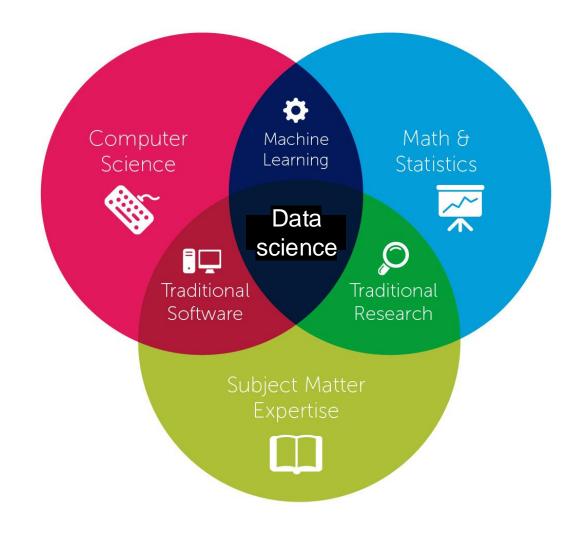
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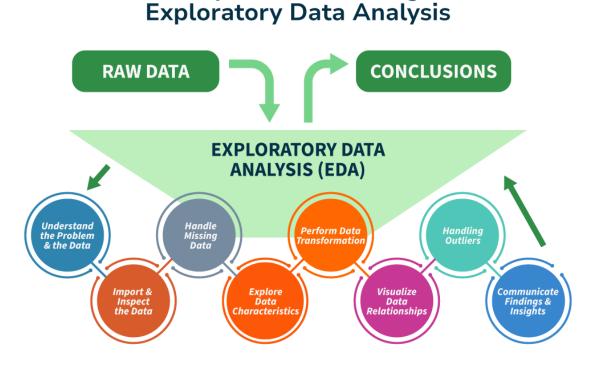


- EDA is understanding the main characteristics of a dataset, often with statistics and visual methods. The purpose is to discover patterns, spot anomalies and outliers, test hypotheses, and check assumptions.
- Data types and structures: df.dtypes, df.info().
- Desciptive statistics:

df.describe(), df.mean(), df.median(),
df.mode(), df.std().

Checking for Missing Values:

df.isnull(), df.isna().sum()



Steps for Performing

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Univariate analysis (one variable/column at a time)

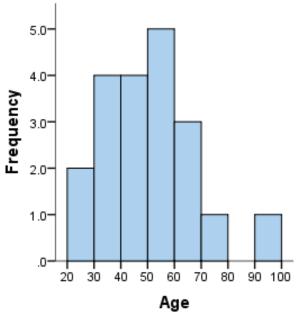
Data features (columns) are usually not independent. For example, salary is correlated with age of the employee and position in the company. Age and position are also related. Univariate ("single variable") analysis disregards these correlations and tries to analyze each column independently. For example, plotting historgrams is part of the univariate analysis.

Continuous Variables:

- Histograms: plt.hist(), df.plot(kind='hist')
- Boxplots: plt.boxplot(), df.plot(kind='box')

Categorical Variables:

- Bar plots: plt.bar(), df.value_counts().plot(kind='bar')
- Pie charts: plt.pie()

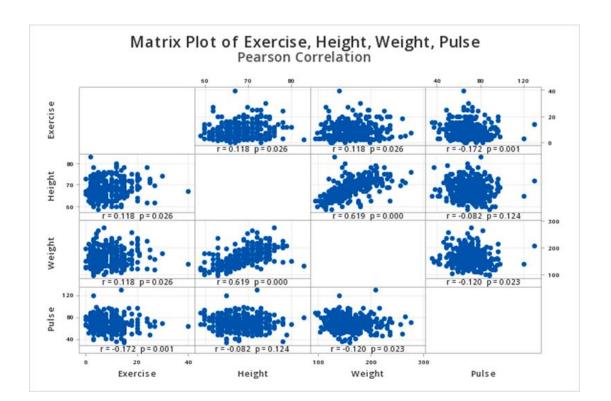


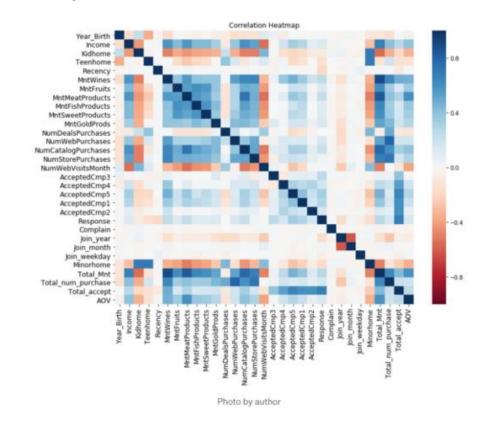
Multivariate Analysis (examining relationships between columns)

This is analysis where more than one column/variable/predictor/feature is used.

Correlation matrix: It shows correlation between predictors.

Scatter Plots: plt.scatter()





Advanced Visualization:

Heatmaps: Visual representation of data where values are depicted by color.

sns.heatmap()

Faceting: Creating multiple plots which share the same variables.

sns.FacetGrid()

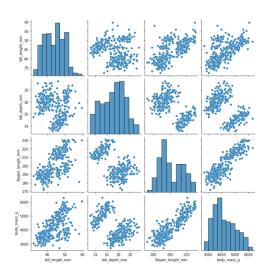
Time Series Visualization:

Line charts: plt.plot(), df.plot()

Interactive Visualization (Zoom, rotation etc):

• Libraries like Plotly, Bokeh, ...





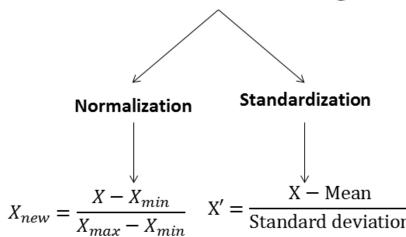
Data Transformation for Visualization:
 Normalization and Standardization: Most of

times this helps increase model performance.

Normalization: Scale data to interval [0,1].

Standardization: Make data's μ =0 and σ =1.

Feature scaling

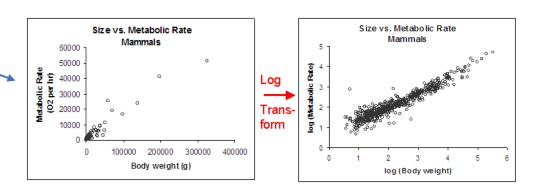


Log Transformations: Useful for skewed data.

np.log()

Binning Data: Converting continuous data into categorical.

pd.cut(), pd.qcut()

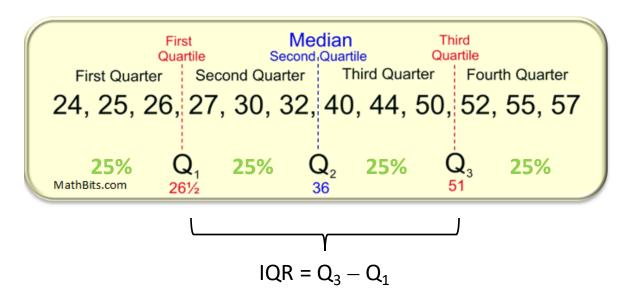


Outlier Detection:

Visual Methods: Using boxplots and scatter plots.

Statistical Methods: Z-score, IQR (interquartile range)

For the data=[24,25,26,27,30,32,40,44,50,52,55,57]



Data points below Q_1 - 1.5 IQR and above Q_3 + 1.5 IQR are considered outliers.

Whiskers usually show $Q_1 - 1.5 IQR$ and $Q_3 + 1.5 IQR$.

