Data Wrangling

Missing Data

- 1. Drop data
 - a. Drop the whole row
 - b. Drop the whole column
- 2. Replace data
 - a. Replace it by mean
 - b. Replace it by frequency
 - c. Replace it based on other functions

Evaluate missing data by counting per column df.isnull().sum()

Correct data format

Check : df.dtypes Convert with .astype

Data Standardization

Columns use common format so they can be compared (e.g. in same units)

Data Normalization

Transforming values of several variables into a similar range Examples:

- Variable average is 0
- Variance is 1
- Variable values range from 0 to 1

To scale values to be from 0 to 1: replace original with original / max value df['height'] = df['height']/df['height'].max()

Binning

Transforming continuous numerical variables into discrete categorical 'bins' for grouped analysis

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Ex: Change horsepower which ranges from 48 to 288 with 59 unique values, into low, medium and high categories.

Utilize np.linspace(start_value, end_value, numbers_generated)

3 bins -> 4 dividers
bins = np.linspace(min(df["horsepower"]), max(df["horsepower"]), 4)

Set bin group names:
group_names = ['Low', 'Medium', 'High']

Apply:
df['horsepower-binned'] = pd.cut(df['horsepower'], bins, labels=group_names, include_lowest=True )

Visualize bins w/ histogram:
import matplotlib as plt
from matplotlib import pyplot
```

Indicator Variable (or Dummy Variable)

plt.pyplot.hist(df["horsepower"], bins = 3)

```
Numerical value used to label categories
Typically used to make categorical variables fit for regression analysis
Ex: fuel type is gas or diesel, convert these to 0 and 1

dummy_variable_1 = pd.get_dummies(df["fuel-type"])
Reformat column names:
dummy_variable_1.rename(columns={'gas':'fuel-type-gas', 'diesel':'fuel-type-diesel'}, inplace=True)

# merge data frame "df" and "dummy_variable_1"
df = pd.concat([df, dummy_variable_1], axis=1)

# drop original column "fuel-type" from "df"
df.drop("fuel-type", axis = 1, inplace=True)
```