



## 1 Preprocess the dataset Diamonds2

This dataset is a subset of the dataset Diamonds retrieved from Kaggle:

<https://www.kaggle.com/datasets/shivam2503/diamonds>

Check there to find more info.

## 2. Execute the following steps loading the required libraries. Use the Cheat Sheets

Libraries: *pandas*, *numpy*, *os*, *seaborn*, *sklearn.preprocessing*, *matplotlib.pyplot*

1. Change to the directory with the dataset and load the dataset **diamonds2.csv**
2. Change the name of the very first column to *idx* and set this variable as the index of the dataset
  - Tip: first get the columns names into a list, change the first element of the list to "idx" and set the index with `set_index()`
3. Check how many (and which) features have non-null values
  - Tip: use the `.info()` function or sum all the `.isnull()` values per column
4. Create a subset of your original dataset with only the attributes: carat, cut, depth, price
  - Tip: Create a list with the selected features. e.g., `df[['width','length','species']]` --> Select multiple columns with specific names.

5. Use the function `pairplot()` in `seaborn` to plot all features in a pairwise manner.

- Tip:
  - Use the `hue` parameter to set the colors according to the value of the attribute "cut"
  - Set the parameter `vars` as the list of variables that you want to plot except the cut feature, i.e., `["carat", "depth", "price"]`

6. Check if there are negative values.

7. Set the variables that are negative to null values

- Tip:
  - Null values: use `float("NaN")`
  - Select attributes with negative values using `loc`: e.g. `df.loc[df['a'] < 0, ['a', 'c']]`
  - Select rows meeting logical condition, and only the specific columns

8. Set the null values to the mean of the features excluding the null values

- Tip:
  - a. test all entries in the feature that are not null, use `pandas isna()` function to test e.g. `test = pd.isna(df["a"]) == False`
  - b. select the cases where the above test holds using `.loc[]`, e.g. `df.loc[test, "a"]`
  - c. apply the mean function to the resulting column
  - d. set this value where the test above is True, i.e., `pd.isna(df["a"]) == True`

9. Set the null values in "cut" to the most frequent value

- Tip:
  - a. use the function `value_counts()` to find the most frequent value
  - b. set the attribute cut as the most frequent value for rows with null values, i.e., the rows that match the condition `pd.isna(df["cut"]) == True`

10. Redo the figure in step 5 and save to file `fig.pdf`

- Tip:
  - use the `savefig` function from `matplotlib.pyplot`

11. Check if the feature clarity needs processing

- Tip
  - convert all the values to upper case. Use the function `upper()` applied to strings, e.g. `x.upper()`
  - create a list of all the values in clarity converted to upper case
  - use list comprehension e.g. `[x.upper() for x in dmd.clarity]`
  - set the column clarity with this new value

- recheck that all values are in upper case (use the `value_counts()` applied to the column)

12. Encode the variable `cut` as a numeric value; create a new feature called `cut_num` as the numeric encoding of `cut`

- Tip:
  - use the function `LabelEncoder` from `sklearn.preprocessing`
  - check the sklearn cheat sheet for an example

13. encode the variable `cut_num` as a one hot encoding value; set a new feature called `cut_ohe`

- Tip:
  - use the function `OneHotEncoder` from `sklearn.preprocessing`
  - first use the `fit` function on the `OneHotEncoder` object
  - use `reshape(-1, 1)` in the selected column, e.g. `df.cut_num.reshape(-1, 1)`
  - apply the `transform` function on the `OneHotEncoder` object
  - use the `toarray()` to obtain the converted matrix

14. Calculate a new feature called `price_zscore` as the standard Z-score of the feature `price`

15. Calculate a new feature called `price_min_max` as the min max scaling of the feature `price`

- check its range visualizing as a histogram