

# 2024-06-23

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#sic\_lab #data\_exploration

## Merging dataframes

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Use students\_list.csv files

```
df = pd.read_csv('data_studentlist.csv')
df1 = df.loc[7: , ['NAME' , 'BLOODTYPE' , 'WEIGHT' , 'HEIGHT' ] ]
df2 = df.loc[:10, ]
```

- Merge two dataframes [inner, left, right, outer join]

## Concat data from different sources

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- Merge data from `data.csv` and `data.json` files
  - How can we concat concat function
- hat are the similarities and difference in the two sources

## Plot the following plots using dummy data in both matplotlib and seaborn

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Use any relevant dataset

- Bar plot
- Histogram (play with: density, bins, alpha)
- Boxplot [ Multiple variables, vertical, ]
- Lineplot [linestyle, markerstyles]
- Scatter plot

## Plot sine wave over the linear space of 0-10 having 100 steps in between and check different

- markers [o, v, s, ^]
- color [red, green ,.. ]
- linestyle [-, --, ...]
- figure size of plt.figure

## Using Axes:

Use [plt.figure and fig.add\_axes]

- Multiple plot in same axes [Sine and cos]
- Multiple plots in different axes
- Multiple plots using plt.subplots (draw 4 different plots in a 2x2 grid)

## Plot using Pandas using iris flower dataset

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Use: [iris dataset, based on type]

- histogram
- barplot
- scatterplot
- scatter\_matrix

## Plot using Seaborn using mpg dataset:

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Use [x=weight, y= mpg , hue=origin if required]

```
dat = sns.load_dataset('mpg')
```

- Histogram [kde, rug, bins, color, ]
- KDEplot
- jointplot [kind]
- Implot[hue, col, markers]
- barplot[ (origin, mpg) estimator]
- countplot [hue]
- boxplot [multiple with x/y=origin/mpg, notch, palette, hue=cylinders]
- violinplot [(origin, mpg), (cylinders, horsepower)]
- stripplot, swarmplot, violinplot + swarmplot overlap
- pairplot [hue=species]
- PairGrid
- FacetGrid
- heatmap of correlation

## Feature selection and Engineering

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Using the mpg dataset

```
sns.load_dataset("mpg")
```

- Based on correlation which feature are important / which can be skipped?
  - How do we get score of correlation between all the variables
- Using p-score
  - Use SelectKBest, f\_regression from sklearn
- Feature normalization
  - StandardScaler
  - MinMaxScaler
- Adding new variables combining existing features
  - eg BMI
  - polynomial feature using sklearn ()
- Encodings

```
data = {
    "Rec-no": range(10),
    "Temperature": ["Hot", "Cold", "Very Hot",
                    "Warm", "Hot", "Warm",
                    "warm", "hot", "hot", "cold"],
    "Color": ["Red", "Yellow", "Blue",
              "Blue", "Red", "Yellow",
              "Red", "Yellow", "Yellow", "Yellow"],
    "Target": [1, 1, 1, 0, 1, 0, 1, 0, 1, 1]
}
```

Is there any issue with the above data?

Explore the Different encodings we have learn about:

- One hot encoding
  - Using OHE from sklearn
  - pd.get\_dummies from pandas
- Label Encoding
- Ordinal value encoding
- Binary Encoding
- Frequency encoding
  - Mean

**what is the difference between**

- fittransform and fit + transform
- transform vs predict