

# MACHINE LEARNING

## PRACTICE LAB

### PREPROCESSING

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Implement the concept of data preprocessing on your assigned domain. Submit both pdf and .ipynb file

```
✓ 10s [4] import pandas as pd

# Import the dataset
from google.colab import files
uploaded=files.upload()

import pandas as pd
df=pd.read_csv('World-happiness-report-2024.csv')
```

↔ Choose Files World-happ...rt-2024.csv

- **World-happiness-report-2024.csv**(text/csv) - 8618 bytes, last modified: 12/3/2024 - 100% done

Saving World-happiness-report-2024.csv to World-happiness-report-2024.csv

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```
# Display the first few rows of the dataset
print("Dataset Preview:")
print(df.head())
```



Dataset Preview:

	Ladder score	upperwhisker	lowerwhisker	Log GDP per capita	\
0	7.741	7.815	7.667	1.844	
1	7.583	7.665	7.500	1.908	
2	7.525	7.618	7.433	1.881	
3	7.344	7.422	7.267	1.878	
4	7.341	7.405	7.277	1.803	

  

	Social support	Healthy life expectancy	Freedom to make life choices	\
0	1.572	0.695	0.859	
1	1.520	0.699	0.823	
2	1.617	0.718	0.819	
3	1.501	0.724	0.838	
4	1.513	0.740	0.641	

  

	Generosity	Perceptions of corruption	Dystopia + residual
0	0.142	0.546	2.082
1	0.204	0.548	1.881
2	0.258	0.182	2.050
3	0.221	0.524	1.658
4	0.153	0.193	2.298

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```
# Display basic information about the dataset
print("\nDataset Info:")
print(df.info())
```



```
Dataset Info:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 143 entries, 0 to 142
Data columns (total 10 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Ladder score                          143 non-null    float64
1   upperwhisker                         143 non-null    float64
2   lowerwhisker                         143 non-null    float64
3   Log GDP per capita                   140 non-null    float64
4   Social support                       140 non-null    float64
5   Healthy life expectancy               140 non-null    float64
6   Freedom to make life choices          140 non-null    float64
7   Generosity                           140 non-null    float64
8   Perceptions of corruption             140 non-null    float64
9   Dystopia + residual                   140 non-null    float64
dtypes: float64(10)
memory usage: 11.3 KB
None
```

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```
# Display summary statistics
print("\nSummary Statistics:")
print(df.describe())
```



Summary Statistics:

	Ladder score	upperwhisker	lowerwhisker	Log GDP per capita	\
count	143.000000	143.000000	143.000000	140.000000	
mean	5.527580	5.641175	5.413972	1.378807	
std	1.170717	1.155008	1.187133	0.425098	
min	1.721000	1.775000	1.667000	0.000000	
25%	4.726000	4.845500	4.606000	1.077750	
50%	5.785000	5.895000	5.674000	1.431500	
75%	6.416000	6.507500	6.319000	1.741500	
max	7.741000	7.815000	7.667000	2.141000	

	Social support	Healthy life expectancy	Freedom to make life choices	\
count	140.000000	140.000000	140.000000	
mean	1.134329	0.520886	0.620621	
std	0.333317	0.164923	0.162492	
min	0.000000	0.000000	0.000000	
25%	0.921750	0.398000	0.527500	
50%	1.237500	0.549500	0.641000	
75%	1.383250	0.648500	0.736000	
max	1.617000	0.857000	0.863000	

	Generosity	Perceptions of corruption	Dystopia + residual
count	140.000000	140.000000	140.000000
mean	0.146271	0.154121	1.575914
std	0.073441	0.126238	0.537459
min	0.000000	0.000000	-0.073000
25%	0.091000	0.068750	1.308250
50%	0.136500	0.120500	1.644500
75%	0.192500	0.193750	1.881750
max	0.401000	0.575000	2.998000

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# Perform basic analysis (Example: Check for missing values)  
print("\nMissing Values:")  
print(df.isnull().sum())

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Missing Values:  
Ladder score 0  
upperwhisker 0  
lowerwhisker 0  
Log GDP per capita 3  
Social support 3  
Healthy life expectancy 3  
Freedom to make life choices 3  
Generosity 3  
Perceptions of corruption 3  
Dystopia + residual 3  
dtype: int64