Normalize data using numpy

```
In [1]: import numpy as np

# Example data
data = np.array([1, 2, 3, 4, 5])

# Normalizing the data
normalized_data = (data - np.min(data)) / (np.max(data) - np.min(data))
print(normalized_data)

[0.  0.25 0.5  0.75 1. ]
```

Standardize data using numpy

```
In [2]: import numpy as np

# Example data
data = np.array([1, 2, 3, 4, 5])

# Standardizing the data
standardized_data = (data - np.mean(data)) / np.std(data)

print(standardized_data)

[-1.41421356 -0.70710678 0. 0.70710678 1.41421356]
```

Normalize using MinMaxScaler

```
In [3]: from sklearn.preprocessing import MinMaxScaler

# Example data
data = [[1], [2], [3], [4], [5]]

# Initialize the scaler
scaler = MinMaxScaler()

# Fit and transform the data
normalized_data = scaler.fit_transform(data)

print(normalized_data)

[[0. ]
    [0.25]
    [0.5]
    [0.75]
    [1. ]]
```

Standardize using StandardScaler

Normalize on dataframe

```
In [5]: import pandas as pd

# Example DataFrame
df = pd.DataFrame({'A': [1, 2, 3, 4, 5], 'B': [10, 20, 30, 40, 50]})

# Normalizing the DataFrame
df_normalized = (df - df.min()) / (df.max() - df.min())

print(df_normalized)

A B
0 0.00 0.00
1 0.25 0.25
2 0.50 0.50
3 0.75 0.75
4 1.00 1.00
```

Standardize on DataFrame

```
In [6]: import pandas as pd
# Example DataFrame
df = pd.DataFrame({'A': [1, 2, 3, 4, 5], 'B': [10, 20, 30, 40, 50]})
# Standardizing the DataFrame
df_standardized = (df - df.mean()) / df.std()
print(df_standardized)

A B
0 -1.264911 -1.264911
1 -0.632456 -0.632456
2 0.000000 0.000000
3 0.632456 0.632456
4 1.264911 1.264911
```

Feature Encoding

Label encoding on list

```
In [7]: from sklearn.preprocessing import LabelEncoder

# Example data
data = ['red', 'green', 'blue', 'green', 'blue', 'red']

# Initialize the encoder
encoder = LabelEncoder()

# Fit and transform the data
encoded_data = encoder.fit_transform(data)

print(encoded_data)

[2 1 0 1 0 2]
```

Label encoding on DataFrame

```
In [8]: import pandas as pd
        from sklearn.preprocessing import LabelEncoder
        # Example DataFrame
        df = pd.DataFrame({
             'color': ['red', 'green', 'blue', 'green', 'blue', 'red'], 'size': ['S', 'M', 'L', 'M', 'S', 'L']
        })
        # Initialize the LabelEncoder
        label_encoder = LabelEncoder()
        # Apply label encoding to the 'color' column
        df['color_encoded'] = label_encoder.fit_transform(df['color'])
        print(df)
            color size color_encoded
        0
             red S
           green
             blue
                                     0
         3 green M
             blue
                     S
             red
```

One Hot Encoding on DataFrame

```
In [9]: import pandas as pd
        df = pd.DataFrame({'color': ['red', 'green', 'blue', 'green', 'blue', 'red']})
        # One-hot encoding
        one_hot_encoded_data = pd.get_dummies(df, columns=['color'])
        print(one_hot_encoded_data)
           color_blue color_green color_red
        0
                    0
        1
                    0
                                1
                                           0
                                 0
        3
                    0
                                1
                                           0
                   1
```

One Hot Encoding using sklearn

```
In [10]: from sklearn.preprocessing import OneHotEncoder

# Example data
data = [['red'], ['green'], ['blue'], ['green'], ['blue'], ['red']]

# Initialize the encoder
encoder = OneHotEncoder(sparse_output=False)

# Fit and transform the data
one_hot_encoded_data = encoder.fit_transform(data)

print(one_hot_encoded_data)

[[0. 0. 1.]
[0. 1. 0.]
[1. 0. 0.]
[0. 1. 0.]
[1. 0. 0.]
[0. 0. 1.]]
```

Ordinal Encoding

```
In [11]: from sklearn.preprocessing import OrdinalEncoder

# Example data
data = [['low'], ['medium'], ['high'], ['medium'], ['high'], ['low']]

# Initialize the encoder
encoder = OrdinalEncoder(categories=[['low', 'medium', 'high']])

# Fit and transform the data
ordinal_encoded_data = encoder.fit_transform(data)

print(ordinal_encoded_data)

[[0.]
[1.]
[2.]
[1.]
[2.]
[0.]]
In [12]: #pip install category_encoders
```

Binary Encoding

```
In [13]: import category_encoders as ce
         # Example data
         data = pd.DataFrame({'city': ['New York', 'Los Angeles', 'Chicago', 'Houston', 'Phoenix']})
         # Initialize the encoder
         encoder = ce.BinaryEncoder(cols=['city'])
         # Fit and transform the data
         binary_encoded_data = encoder.fit_transform(data)
         print(binary_encoded_data)
            city_0 city_1 city_2
         a
                 a
                         0
         1
                 0
                         1
                                0
                 0
                                1
                         0
                                0
         3
                 1
                         0
```

Log Transformation

```
In [14]: import numpy as np
# Example data
data = np.array([1, 10, 100, 10000])
# Logarithmic transformation
log_transformed = np.log(data)
print(log_transformed)

[0. 2.30258509 4.60517019 6.90775528 9.21034037]
```

Log Transformation on single column

```
In [15]: import pandas as pd
         import numpy as np
         # Example DataFrame
         df = pd.DataFrame({
             'A': [1, 10, 100, 1000, 10000],
             'B': [5, 15, 25, 35, 45]
         })
         # Apply log transformation to column 'A'
         df['A_log'] = np.log(df['A'])
         print(df)
               Δ
                   В
                         A_log
         0
               1
                  5 0.000000
              10 15 2.302585
             100 25 4.605170
         2
            1000 35 6.907755
         4 10000 45 9.210340
```

log Transformation on Multiple Columns

```
In [16]: import pandas as pd
           import numpy as np
           # Example DataFrame
           df = pd.DataFrame({
                'A': [1, 10, 100, 1000, 10000],
                'B': [5, 15, 25, 35, 45],
                'C': [2, 20, 200, 2000, 20000]
           })
           # Apply log transformation to multiple columns
           columns_to_transform = ['A', 'B','C']
           df_log_transformed = df.copy()
           for col in columns_to_transform:
                df_log_transformed[col + '_log'] = np.log(df_log_transformed[col])
           print(df_log_transformed)
                                        A log
                                                    B log
                                                               C log
           0
                                2 0.000000 1.609438 0.693147
                   1 5
                              20 2.302585 2.708050 2.995732
           1
                  10 15

        100
        25
        200
        4.605170
        3.218876
        5.298317

        1000
        35
        2000
        6.907755
        3.555348
        7.600902

           2
           4 10000 45 20000 9.210340 3.806662 9.903488
```

Square root transformation

```
In [17]: import numpy as np
# Example data
data = np.array([1, 4, 9, 16, 25])
# Square root transformation
sqrt_transformed = np.sqrt(data)
print(sqrt_transformed)

[1. 2. 3. 4. 5.]
```

Exponential transformation

```
In [18]: import numpy as np
# Example data
data = np.array([0.1, 0.2, 0.3, 0.4, 0.5])
# Exponential transformation
exp_transformed = np.exp(data)
print(exp_transformed)
```

[1.10517092 1.22140276 1.34985881 1.4918247 1.64872127]