features

```
import pandas as pd
import numpy as np
from sklearn.datasets import load_iris
# Load iris dataset into a DataFrame
df_iris = pd.DataFrame(load_iris().data, columns=load_iris().feature_names)
# Convert DataFrame to NumPy array
iris_array = df_iris.values
# Display the first few rows of the DataFrame
print(df_iris.head())
\# Display the NumPy array
print(iris_array)
      [5.8 2.6 4. 1.2]
      [5. 2.3 3.3 1. ]
      [5.6 2.7 4.2 1.3]
      [5.7 3. 4.2 1.2]
      [5.7 2.9 4.2 1.3]
      [6.2 2.9 4.3 1.3]
      [5.1 2.5 3. 1.1]
      [5.7 2.8 4.1 1.3]
      [6.3 3.3 6. 2.5]
      [5.8 2.7 5.1 1.9]
      [7.1 3. 5.9 2.1]
      [6.3 2.9 5.6 1.8]
      [6.5 3. 5.8 2.2]
      [7.6 3. 6.6 2.1]
      [4.9 2.5 4.5 1.7]
      [7.3 2.9 6.3 1.8]
      [6.7 2.5 5.8 1.8]
      [7.2 3.6 6.1 2.5]
      [6.5 3.2 5.1 2. ]
      [6.4 2.7 5.3 1.9]
      [6.8 3. 5.5 2.1]
      [5.7 2.5 5. 2. ]
      [5.8 2.8 5.1 2.4]
      [6.4 3.2 5.3 2.3]
      [6.5 3. 5.5 1.8]
      [7.7 3.8 6.7 2.2]
      [7.7 2.6 6.9 2.3]
      [6. 2.2 5. 1.5]
      [6.9 3.2 5.7 2.3]
      [5.6 2.8 4.9 2. ]
      [7.7 2.8 6.7 2. ]
      [6.3 2.7 4.9 1.8]
      [6.7 3.3 5.7 2.1]
      [7.2 3.2 6. 1.8]
      [6.2 2.8 4.8 1.8]
      [6.1 3. 4.9 1.8]
      [6.4 2.8 5.6 2.1]
      [7.2 3. 5.8 1.6]
      7.4 2.8 6.1 1.9]
      [7.9 3.8 6.4 2. ]
      [6.4 2.8 5.6 2.2]
      [6.3 2.8 5.1 1.5]
      [6.1 2.6 5.6 1.4]
      [7.7 3. 6.1 2.3]
      [6.3 3.4 5.6 2.4]
      [6.4 3.1 5.5 1.8]
      [6. 3. 4.8 1.8]
      [6.9 3.1 5.4 2.1]
      [6.7 3.1 5.6 2.4]
      [6.9 3.1 5.1 2.3]
      [5.8 2.7 5.1 1.9]
      [6.8 3.2 5.9 2.3]
      [6.7 3.3 5.7 2.5]
      [6.7 3. 5.2 2.3]
      [6.3 2.5 5. 1.9]
      [6.5 3. 5.2 2. ]
      [6.2 3.4 5.4 2.3]
      [5.9 3. 5.1 1.8]]
features = np.array(iris_array)
```

https://colab.research.google.com/drive/1fOfPDsBOQdw-v-4XgiHCDaWvJkGSMwm0#scrollTo=UnKEr-pjR1W9&printMode=true

```
[6.2, 2.9, 4.3, 1.3],
[5.1, 2.5, 3. , 1.1],
[5.7, 2.8, 4.1, 1.3],
[6.3, 3.3, 6., 2.5],
[5.8, 2.7, 5.1, 1.9],
[7.1, 3., 5.9, 2.1],
[6.3, 2.9, 5.6, 1.8],
[6.5, 3., 5.8, 2.2],
[7.6, 3., 6.6, 2.1],
[4.9, 2.5, 4.5, 1.7],
[7.3, 2.9, 6.3, 1.8],
[6.7, 2.5, 5.8, 1.8],
[7.2, 3.6, 6.1, 2.5],
[6.5, 3.2, 5.1, 2. ],
[6.4, 2.7, 5.3, 1.9],
[6.8, 3., 5.5, 2.1],
[5.7, 2.5, 5., 2.],
[5.8, 2.8, 5.1, 2.4],
[6.4, 3.2, 5.3, 2.3],
[6.5, 3., 5.5, 1.8],
[7.7, 3.8, 6.7, 2.2],
[7.7, 2.6, 6.9, 2.3],
[6., 2.2, 5., 1.5],
[6.9, 3.2, 5.7, 2.3],
[5.6, 2.8, 4.9, 2.],
[7.7, 2.8, 6.7, 2.],
[6.3, 2.7, 4.9, 1.8],
[6.7, 3.3, 5.7, 2.1],
[7.2, 3.2, 6., 1.8],
[6.2, 2.8, 4.8, 1.8],
[6.1, 3., 4.9, 1.8],
[6.4, 2.8, 5.6, 2.1],
[7.2, 3., 5.8, 1.6],
[7.4, 2.8, 6.1, 1.9],
[7.9, 3.8, 6.4, 2.],
[6.4, 2.8, 5.6, 2.2],
[6.3, 2.8, 5.1, 1.5],
[6.1, 2.6, 5.6, 1.4],
[7.7, 3., 6.1, 2.3],
[6.3, 3.4, 5.6, 2.4],
[6.4, 3.1, 5.5, 1.8],
[6., 3., 4.8, 1.8],
[6.9, 3.1, 5.4, 2.1],
[6.7, 3.1, 5.6, 2.4],
[6.9, 3.1, 5.1, 2.3],
[5.8, 2.7, 5.1, 1.9],
[6.8, 3.2, 5.9, 2.3],
[6.7, 3.3, 5.7, 2.5],
[6.7, 3., 5.2, 2.3],
[6.3, 2.5, 5. , 1.9],
[6.5, 3., 5.2, 2.],
[6.2, 3.4, 5.4, 2.3],
[5.9, 3., 5.1, 1.8]])
```

```
import numpy as np
from sklearn import preprocessing
minmax_scaler = preprocessing.MinMaxScaler(feature_range =(0,1))
scaled_feature = minmax_scaler.fit_transform(features)
scaled feature
```

```
[0.72222222, 0.5 , 0.79661017, 0.91666667],
            [0.36111111, 0.33333333, 0.66101695, 0.79166667],
            [0.94444444, 0.33333333, 0.96610169, 0.79166667],
            [0.5555556, 0.29166667, 0.66101695, 0.70833333],
            [0.66666667, 0.54166667, 0.79661017, 0.83333333],
            [0.80555556, 0.5
                                 , 0.84745763, 0.70833333],
            [0.52777778, 0.33333333, 0.6440678, 0.70833333],
                      , 0.41666667, 0.66101695, 0.70833333],
            [0.58333333, 0.33333333, 0.77966102, 0.83333333],
            [0.80555556, 0.41666667, 0.81355932, 0.625
            [0.86111111, 0.33333333, 0.86440678, 0.75
                      , 0.75
                                , 0.91525424, 0.79166667],
            [1.
            [0.58333333, 0.33333333, 0.77966102, 0.875
            [0.55555556, 0.33333333, 0.69491525, 0.58333333],
            [0.5 , 0.25 , 0.77966102, 0.54166667], [0.94444444, 0.41666667, 0.86440678, 0.91666667],
            [0.55555556, 0.58333333, 0.77966102, 0.95833333],
            [0.58333333, 0.45833333, 0.76271186, 0.70833333],
            [0.47222222, 0.41666667, 0.6440678, 0.70833333],
            [0.72222222, 0.45833333, 0.74576271, 0.83333333],
            [0.66666667, 0.45833333, 0.77966102, 0.95833333],
            [0.72222222, 0.45833333, 0.69491525, 0.91666667],
            [0.41666667, 0.29166667, 0.69491525, 0.75
            [0.69444444, 0.5
                                   , 0.83050847, 0.91666667],
            [0.66666667, 0.54166667, 0.79661017, 1.
            [0.66666667, 0.41666667, 0.71186441, 0.91666667],
            [0.55555556, 0.20833333, 0.6779661 , 0.75
            [0.61111111, 0.41666667, 0.71186441, 0.79166667],
            [0.52777778, 0.58333333, 0.74576271, 0.91666667],
            [0.44444444, 0.41666667, 0.69491525, 0.70833333]])
scaler = preprocessing. StandardScaler()
# transform the feature
standardized = scaler.fit_transform(features)
standardized
print("Mean {}".format(round (standardized.mean())))
print("Standard Deviation: {}".format(standardized.std()))
     Mean 0
     Standard Deviation: 1.0
# create scaler
robust_scaler = preprocessing. RobustScaler()
#transform feature
robust_scaler.fit_transform (features)
→ array([[-0.53846154, 1.
                                     , -0.84285714, -0.73333333],
                                     , -0.84285714, -0.73333333],
            [-0.69230769, 0.
                                    , -0.87142857, -0.73333333],
            [-0.84615385, 0.4
                                    , -0.81428571, -0.73333333],
            [-0.92307692, 0.2
                                     , -0.84285714, -0.73333333],
            [-0.61538462, 1.2
            [-0.30769231, 1.8
                                    , -0.75714286, -0.6
                                    , -0.84285714, -0.66666667],
, -0.81428571, -0.73333333],
            [-0.92307692, 0.8
            [-0.61538462, 0.8
            [-1.07692308, -0.2
                                    , -0.84285714, -0.733333333],
            [-0.69230769, 0.2
                                     , -0.81428571, -0.8
            [-0.30769231, 1.4
                                    , -0.81428571, -0.73333333],
                                    , -0.78571429, -0.733333333],
            [-0.76923077, 0.8
[-0.76923077, 0.
                                     , -0.84285714, -0.8
                                    , -0.92857143, -0.8
            [-1.15384615, 0.
                                    , -0.9
                        , 2.
                                                   , -0.73333333],
            [-0.07692308, 2.8
                                    , -0.81428571, -0.6
, -0.87142857, -0.6
            [-0.30769231, 1.8
                                    , -0.84285714, -0.66666667],
            [-0.53846154, 1.
[-0.07692308, 1.6
                                     , -0.75714286, -0.66666667],
                                    , -0.81428571, -0.66666667],
            [-0.53846154, 1.6
                                    , -0.75714286, -0.73333333],
, -0.81428571, -0.6 ],
            [-0.30769231, 0.8
            [-0.53846154, 1.4
            [-0.92307692, 1.2
                                    , -0.95714286, -0.73333333],
            [-0.53846154, 0.6
                                     , -0.75714286, -0.53333333],
            [-0.76923077, 0.8
                                     , -0.7
                                                  , -0.73333333],
                                    , -0.78571429, -0.73333333],
            [-0.61538462, 0.
            [-0.61538462, 0.8
                                     , -0.78571429, -0.6
            [-0.46153846, 1.
                                     , -0.81428571, -0.73333333],
            [-0.46153846, 0.8
                                     , -0.84285714, -0.73333333],
                                     , -0.78571429, -0.73333333],
            [-0.84615385, 0.4
            [-0.76923077, 0.2
                                     , -0.78571429, -0.73333333],
```

[0.01111111, 0.4100000/, 0./02/1180, 0./0833333],

 $\hbox{\tt [0.47222222, 0.08333333, 0.6779661, 0.58333333],}\\$

[0.94444444, 0.75 [0.94444444, 0.25

, 0.96610169, 0.875

, 1. , 0.91666667],

```
, -0.81428571, -0.6
[-0.30769231, 0.8
                             , -0.81428571, -0.8 ],
, -0.84285714, -0.73333333],
[-0.46153846, 2.2
[-0.23076923, 2.4
                             , -0.81428571, -0.73333333],
, -0.9 , -0.73333333],
, -0.87142857, -0.73333333],
[-0.69230769, 0.2
[-0.61538462, 0.4
[-0.23076923, 1.
[-0.69230769, 1.2
[-1.07692308, 0.
                         , -0.84285714, -0.8
                             , -0.87142857, -0.73333333],
, -0.81428571, -0.73333333],
[-0.53846154, 0.8
                         , -0.87142857, -0.66666667],
, -0.87142857, -0.66666667],
[-0.61538462, 1.
          , -1.4
[-1.
[-1.07692308, 0.4 , -0.87142857, -0.73333333],
                             , -0.78571429, -0.46666667],
, -0.7 , -0.6
[-0.61538462, 1.
                        , -0.7 , -0.6 , , -0.84285714, -0.66666667], , -0.78571429, -0.7333333], , -0.84285714, -0.73333333], -0.81428571, -0.73333333],
[-0.53846154, 1.6
[-0.76923077, 0.
[-0.53846154, 1.6
[-0.92307692, 0.4
[-0.38461538, 1.4 , -0.81428571, -0.73333333],
[-0.61538462, 0.6 , -0.84285714, -0.73333333],
                             , -0.84285714, -0.73333333],
, 0.1 , 0.06666667].
[ 0.92307692, 0.4
                             , 0.04285714, 0.133333333],
[ 0.46153846, 0.4
[ 0.84615385, 0.2
                             , 0.15714286, 0.13333333],
, -0.1 , 0. ],
                                                , 0.
[-0.23076923, -1.4
                             , 0.07142857, 0.13333333],
[ 0.53846154, -0.4
                               , 0.04285714, 0. ],
[-0.07692308, -0.4
[ 0.38461538, 0.6
                               , 0.1 , 0.2
[-0.69230769, -1.2
                               , -0.3
                                               , -0.2
```

```
import numpy as np
from sklearn.preprocessing import Normalizer
normalizer11 =Normalizer(norm='11')
normalizer12 =Normalizer(norm='12')
normalizerMax =Normalizer(norm='max')
print("11 normalization\n",normalizer11.transform(features))
print("\n12 normalization\n",normalizer12.transform(features))
print("\nmax normalization\n",normalizerMax.transform(features))
```

```
4/24/24, 2:15 PM
```

```
[1.
                  0.44927536 0.7826087
                                        0.30434783]
      [1.
                 0.46268657 0.8358209 0.35820896]
      [1.
      [1.
                 0.44927536 0.73913043 0.33333333]
                 0.46551724 0.87931034 0.32758621]
                 0.47058824 0.86764706 0.33823529]
      [1.
      ٢1.
                 0.49253731 0.85074627 0.37313433]
                 0.44776119 0.7761194 0.34328358]
      [1.
                 0.3968254 0.79365079 0.3015873 ]
      [1.
      Γ1.
                 0.46153846 0.8
                                        0.307692317
      [1.
                 0.5483871 0.87096774 0.37096774]
      Г1.
                 0.50847458 0.86440678 0.30508475]]
import pandas as pd
from sklearn.datasets import make_blobs
from sklearn.cluster import KMeans
features,_= make_blobs (n_samples = 150,
                       n features=3,
                        centers = 3.
                        random_state= 1)
df = pd.DataFrame(features, columns= ["feature_1", "feature_2", "feature_3"])
# make k-means clusterer
clusterer = KMeans (3, random_state=0)
# fit clusterer
clusterer.fit(features)
# predict values
df ['group'] = clusterer.predict(features)
df.head()
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10
       warnings.warn(
        feature_1 feature_2 feature_3 group
                                                  ☶
        -0.941970
                     4.155785 -10.049242
         -6.475523
                    -2.792061
                               -3.001938
                                             2
         -6.785592
                    -1.208691
                               -1.827924
                                             2
         -0.411260
                     3.648816
                               -9.409418
         -4.013822
                    -8.093843
                               -7.546288
 Next steps:
             Generate code with df
                                     View recommended plots
import numpy as np
features[~np.isnan(features).any(axis=1)]
     array([[ -0.94197008,
                            4.15578496, -10.04924243],
              -6.4755234 , -2.79206127, -3.00193779],
            [ -6.7855919 , -1.20869056, -1.82792398],
             -0.41126012, 3.64881573, -9.40941834],
              -4.01382189, -8.09384267, -7.54628777],
            [ -3.79439472.
                           5.22333705, -8.81354119],
             -7.25507738,
                          -5.05861909, -1.34389165],
             -6.06498262,
                           -2.8606084 , -0.72361683],
            [ -1.23199783, 4.48049947, -10.45088111],
              -6.86332082,
                          -1.66231803, -2.37843746],
            [ -0.84702472,
                           4.6173919 , -9.57593335],
            [ -4.87044829, -3.20920741, -1.23242237],
             -6.12205907,
                           -4.14707068, -2.166599 ],
            [ -6.07198971, -2.23698036, -2.14545449],
            [ -6.64354615, -2.02316807, -1.99049784],
              -1.5905901 ,
                            4.87973126, -10.64806753],
            [ -4.40197918, -6.59432259, -9.55180633],
            [ -5.88991027, -4.25764374, -0.81167373],
              -1.47892914,
                          -7.39720703, -7.37608159],
            [ -5.28177321, -6.89951903, -8.12568374],
             -4.39609607,
                           -4.52586773, -1.18619469],
              -3.07096007,
                           -6.32891233, -9.50459007],
            [ -1.85909282,
                           4.63932217, -9.84231924],
             -5.32987837,
                           -3.01353107, -2.4451363 ],
            [ -1.54634187,
                            3.16374415, -8.39822943],
            [ -2.85245239, -5.95397939, -8.40109229],
              -3.0524769 ,
                                         -7.93554555],
                            -5.66239553,
            [ -5.51136714,
                           -2.13552151,
                                         -2.49889127],
            [ -4.79217079,
                           -7.74341308, -8.21911732],
              -2.43866686.
                            3,62882716,
                                         -9.375149711.
              0.02893715,
                             5.03230134, -11.60907632],
            [ -1.59946217,
                             4.86891066, -9.31287601],
```

```
[ -1.55639691,
                             3.78482302, -9.72199446],
            [ -6.39400275, -3.81457605, -3.13087409],
            [ -6.57996897, -1.26301168, -2.42684438],
              -1.31270057,
                             5.77352257, -9.32399643],
            [ -0.74236389, 4.48674046, -9.76947373],
            [ -1.07763472, 3.99628235, -7.7008464 ],
              -5.63202111, -2.55853869, -0.6185191],
            [ -4.53936962, -4.20390713, -2.19574027],
              -4.51388547, -6.924109 , -6.38562715],
              -5.81979901, -7.93820265, -7.85766481],
            [ \  \  \, -6.39390511, \quad \, -5.29714933, \quad \, -2.30126622],
            [ -4.54736797, -6.07666574, -7.97012404],
            [ -3.57365667, -7.18057694, -8.29984821],
            [ -6.47637244, -4.03144103, -2.23675048],
            [ -1.05540626, 6.79169568, -10.12159583],
[ -2.69615858, -6.60231014, -9.06826133],
            [ -1.75343694, 4.23671585, -10.53885713],
              -6.25378449, -3.35512243, -2.43646564],
            [ -5.50091176, -3.66043258, -1.81714196],
            [ -6.25460817, -5.11952152, -1.80730775],
              -1.29372156, 6.01399691, -10.23588987],
            [ -7.75883232, -3.04490815, -2.70598468],
              -4.72518456, -7.32984286, -7.04594588],
              -1.50104502,
                             5.2799081 , -10.10909587],
            [ -6.44056643, -2.38973725, -3.21985689],
import pandas as pd
df = pd.DataFrame (features, columns= ["feature_1", "feature_2", "feature_3"])
df.dropna()
           feature_1 feature_2 feature_3
                                              扁
            -0.941970
                       4.155785 -10.049242
            -6.475523
                       -2.792061
                                  -3.001938
       1
            -6.785592
                       -1.208691
                                  -1.827924
       3
            -0.411260
                       3.648816
                                  -9.409418
                       -8.093843
            -4.013822
                                  -7.546288
            -4.437912
                       -6.688998
                                  -7.405180
      145
      146
            -2.701437
                       -3.605530
                                  -1.793650
      147
            -5.138206
                       -3.407788
                                  -2.079798
      148
            -2.951123
                       3.558246 -10.164312
      149
            -6.666073 -2.469441
                                  -2 828343
     150 rows × 3 columns
import numpy as np
from sklearn.preprocessing import StandardScaler
from sklearn.impute import SimpleImputer
from sklearn.datasets import make_blobs
features, _ = make_blobs(n_samples=150, n_features=3, random_state=1)
scaler = StandardScaler()
standardized_features = scaler.fit_transform(features)
true_value = standardized_features[0, 0]
standardized_features[0, 0] = np.nan
mean_imputer = SimpleImputer(strategy="mean")
median_imputer = SimpleImputer(strategy="median")
mode_imputer = SimpleImputer(strategy="most_frequent")
features_mean_imputed = mean_imputer.fit_transform(standardized_features)
features_median_imputed = median_imputer.fit_transform(standardized_features)
features_mode_imputed = mode_imputer.fit_transform(standardized_features)
print("True Value: {}".format(true_value))
print("Mean Imputed Value: {}".format(features_mean_imputed[0, 0]))
print("Median Imputed Value: {}".format(features_median_imputed[0, 0]))
print("Mode Imputed Value: {}".format(features_mode_imputed[0, 0]))
```

True Value: 1.4286939379208594

Mean Imputed Value: -0.009588549918932624 Median Imputed Value: -0.016984995847737824 Mode Imputed Value: -2.181974840016922