

PYTHON CODE USED TO PRODUCE FIGURE 7

Contents

| | |
|---|---|
| #IMPORTING LIBRARIES | 1 |
| #IMPORTING DATA | 1 |
| #COLUMN INDEX..... | 2 |
| #HISTOGRAM PLOT OF DATA IN EACH COLUMN..... | 2 |
| #PAIRWISE CORRELATIONS OF COLUMN DATA..... | 2 |
| #HEATMAP OF PAIRWISE CORRELATIONS | 2 |
| #PRINCIPAL COMPONENT ANALYSIS (PCA) | 2 |
| #STANDARDIZE DATA USING StandardScaler (applied for columns not rows) | 2 |
| #VIEW DATA Z-SCORES PER COLUMN | 3 |
| #CONCATENATE ORIGINAL DATA TO Z-SCORE DATA | 3 |
| #STANDARDIZE (CONTINUATION) | 3 |
| #SCREE PLOT (Shows the % of each PC)..... | 3 |
| #HEATMAPS OF PCs | 5 |
| #PCA PLOT (2D)..... | 6 |

#IMPORTING LIBRARIES

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.decomposition import PCA
from sklearn.preprocessing import StandardScaler
%matplotlib inline
```

#IMPORTING DATA

```
data = r'C:\Users\HP\Desktop\Dr. Auwal\MAN1\Supporting information_Unsupervised ML
data.csv'
```

```
df = pd.read_csv(data)
```

```
df
```

#COLUMN INDEX

```
df.columns
```

#HISTOGRAM PLOT OF DATA IN EACH COLUMN

```
df[cols].hist(layout=(1, len(cols)), figsize=(3 * len(cols), 3.5));
```

```
plt.rcParams['figure.dpi'] = 200
```

#PAIRWISE CORRELATIONS OF COLUMN DATA

```
df.corr()
```

#HEATMAP OF PAIRWISE CORRELATIONS

```
dcorr=df[cols].corr()
```

```
#dcorr
```

```
mask = np.zeros_like(dcorr)
```

```
#mask.shape
```

```
mask[np.triu_indices_from(mask)] = True
```

```
fig, ax = plt.subplots(figsize=(7,5))
```

```
sns.heatmap(dcorr, cmap=sns.diverging_palette(10, 145, n=100),
```

```
          vmin=-1, vmax=1, center=0, linewidths=1, annot=True, mask=mask, ax=ax);
```

#PRINCIPAL COMPONENT ANALYSIS (PCA)

```
#Defining dataframe columns as X
```

```
X=df[cols]
```

#STANDARDIZE DATA USING StandardScaler (applied for columns not rows)

```
scalar = StandardScaler()
```

```
X_t=scalar.fit_transform(X)
```

```
scalar.mean_
```

```
scalar.var_
```

```
X_t.round(4) #round up to 4 digits
```

#VIEW DATA Z-SCORES PER COLUMN

```
dz=pd.DataFrame(X_t.round(4), columns=[f'z_{c}' for c in cols])
```

```
dz
```

#CONCATENATE ORIGINAL DATA TO Z-SCORE DATA

```
pd.concat([df, dz], axis='columns')
```

#STANDARDIZE (CONTINUATION)

```
(df['% AGEs formed]-scalar.mean_[0])/np.sqrt(scalar.var_[0])
```

```
X_t[:, 0]
```

```
X_t[:, 0]
```

```
X_t[:, 0].mean().round(4)
```

```
np.sqrt(X_t[:, 0].var())
```

```
X.head()
```

```
X_t[:5]
```

```
X_t.shape
```

```
X_t.shape[1]
```

```
#pca = PCA(n_components=X_t.shape[1])
```

```
pca = PCA(n_components=4)
```

```
pca.fit_transform(X_t)
```

#SCREE PLOT (Shows the % of each PC)

```
def scree_plot(X, n_components, with_cumulative=False, show_data_label=False, figsize=(10, 7)):
```

```
    """
```

```
    PCA scree plot with cumulative
```

```
    """
```

```
    scalar = StandardScaler()
```

```

X_t=scalar.fit_transform(X)

max_components = min(X.shape)
x=np.arange(1, n_components+1)
pca = PCA(n_components=max_components)
pca.fit_transform(X_t)
y1=pca.explained_variance_ratio_[0:n_components]
y2=np.cumsum(pca.explained_variance_ratio_)[0:n_components]

plt.figure(figsize=figsize)

if n_components > 20:
    marker = None
else:
    marker = 'o'
if with_cumulative:
    plt.plot(x, y2, linestyle='--', marker=marker, label='cumulative')

plt.plot(x, y1, linestyle='-', marker=marker, label='individual')
plt.title('Explained variance ratio')
plt.xlabel('Number of components')
plt.ylabel('Proportion of variance explained')
plt.legend()
if with_cumulative:
    [plt.axhline(y=x1, color='.7', linestyle='--',) for x1 in [.8, .9, .95, 1]]
plt.grid(axis='x')

```

```

if show_data_label:
    for n, v, cv in zip(np.nditer(x, flags=['refs_ok']),
                       np.nditer(y1, flags=['refs_ok']),
                       np.nditer(y2, flags=['refs_ok'])):
        plt.text(n+.02, v+.02, f'{v*100:.2f}%', fontsize=10)
    if with_cumulative:
        plt.text(n+.02, cv+.02, f'{cv*100:.2f}%', fontsize=10)
screes_plot(X, 4, True, True)

pca.components_ #Eigenvectors
dpc=pd.DataFrame(pca.components_.T,
                 index=cols,
                 columns=[f'PC{n+1}' for n in range(pca.components_.shape[0])]).round(4)
dpc

```

#HEATMAPS OF PCs

```

sns.heatmap(dpc, cmap=sns.diverging_palette(10, 145, n=100), linewidths=1,
            center=0, annot=True, vmin=-1, vmax=1);

```

```

X_t[:5]
X_t.shape
pca.components_.T.shape
#multiply matrix
np.dot(X_t, pca.components_.T)[:5]
pca.transform(X_t[:5])
df.head()
X_t[:1]
pca.components_[:1]

```

```

np.sum(X_t[:1] * pca.components_[:1])
pca.n_components_
dd=pd.concat([pd.DataFrame(pca.transform(X_t),
                        columns=[f'PC{n}' for n in range(1, pca.n_components_ +1)]),
            df[['peptide']]], axis = 'columns')
dd.head()

```

#PCA PLOT (2D)

```

# pca = PCA(n_components=X.shape[1])
pca = PCA(n_components=2)

```

```

X_pca=pca.fit_transform(X_t)
X_pca.shape
X_pca[:5]

```

```

# pca = PCA(n_components=X.shape[1])
pca = PCA(n_components=2)

```

```

X_pca=pca.fit_transform(X_t)
X_pca.shape
X_pca[:5]

```

```

principalDf = pd.DataFrame(data = X_pca
                        , columns = ['principal component 1', 'principal component 2'])

```

```

finalDf = pd.concat([principalDf, df[['peptide']]], axis = 1)

```

```

fig = plt.figure(figsize = (8,8))

```

```
ax = fig.add_subplot(1,1,1)
ax.set_xlabel('PC 1 48.05%', fontsize = 15)
ax.set_ylabel('PC 21.24%', fontsize = 15)
ax.set_xlim(-3, 3)
ax.set_ylim(-2, 2)
#ax.set_title('2 component PCA', fontsize = 20)
peptides = ['AVIAIMF', 'GPADPF', 'NGDF', 'NGNW', 'GSH']
colors = ['y', 'g',]
for peptide, color in zip(peptides,colors):
    indicesToKeep = finalDf['peptide'] == peptide
    ax.scatter(finalDf.loc[indicesToKeep, 'principal component 1']
               , finalDf.loc[indicesToKeep, 'principal component 2']
               , c = color
               , s = 50,)

ax.legend(peptides)
ax.grid()
```