PortafolioImplementacion-M5-TC3007C

December 2, 2022

```
[]: import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    from scipy.stats import anderson, probplot, multivariate_normal, chi2
    from pingouin import multivariate_normality
    %matplotlib inline
[]: url = "https://raw.githubusercontent.com/crisb-7/Mercurio/main/mercurio.csv"
    columnNames = ["ID", "Lago", "Alcalinidad", "pH", "Calcio", "Clorofila", u
      "N_Peces", "MinMercurio", "MaxMercurio", "Estimacion", "Edad"]
    df = pd.read_csv(url, names = columnNames, header = 0)
[]: df.head()
[]:
        ID
                         Alcalinidad
                                            Calcio
                                                   Clorofila Mercurio N_Peces
                                       pН
                                 5.9 6.1
                                                                   1.23
              Alligator
                                               3.0
                                                          0.7
                                                                               5
    1
                   Annie
                                  3.5 5.1
                                               1.9
                                                          3.2
                                                                   1.33
                                                                               7
    2
                 Apopka
                               116.0 9.1
                                              44.1
                                                        128.3
                                                                   0.04
                                                                               6
    3
        4
           Blue Cypress
                                39.4 6.9
                                              16.4
                                                          3.5
                                                                   0.44
                                                                              12
                                  2.5 4.6
                                                          1.8
                  Brick
                                               2.9
                                                                   1.20
                                                                              12
       MinMercurio MaxMercurio Estimacion Edad
              0.85
                           1.43
                                        1.53
    0
              0.92
                           1.90
                                        1.33
    1
                                                0
              0.04
                           0.06
                                       0.04
                                                0
    3
              0.13
                           0.84
                                        0.44
                                                0
              0.69
                           1.50
                                        1.33
                                                1
[]: df = df.drop(columns = "ID")
    columnNames.remove("ID")
[]: df.describe()
```

```
[]:
                                                  Clorofila
                                                                           N_Peces \
            Alcalinidad
                                        Calcio
                                                              Mercurio
                                 pН
     count
              53.000000 53.000000 53.000000
                                                  53.000000 53.000000
                                                                        53.000000
     mean
              37.530189
                           6.590566
                                     22.201887
                                                  23.116981
                                                              0.527170
                                                                         13.056604
     std
              38.203527
                           1.288449
                                     24.932574
                                                  30.816321
                                                              0.341036
                                                                          8.560677
    min
               1.200000
                           3.600000
                                      1.100000
                                                   0.700000
                                                              0.040000
                                                                          4.000000
     25%
               6.600000
                           5.800000
                                      3.300000
                                                   4.600000
                                                              0.270000
                                                                         10.000000
     50%
              19.600000
                           6.800000
                                     12.600000
                                                  12.800000
                                                              0.480000
                                                                         12.000000
     75%
              66.500000
                           7.400000
                                     35.600000
                                                  24.700000
                                                              0.770000
                                                                         12.000000
             128.000000
                           9.100000
                                     90.700000
                                                 152.400000
                                                              1.330000
                                                                        44.000000
    max
            MinMercurio
                         MaxMercurio
                                       Estimacion
                                                         Edad
     count
              53.000000
                            53.000000
                                        53.000000
                                                    53.000000
                             0.874528
                                                     0.811321
     mean
               0.279811
                                         0.513208
     std
               0.226406
                             0.522047
                                         0.338729
                                                     0.394998
     min
               0.040000
                             0.060000
                                         0.040000
                                                     0.000000
     25%
                                         0.250000
               0.090000
                             0.480000
                                                     1.000000
     50%
               0.250000
                             0.840000
                                         0.450000
                                                     1.000000
     75%
               0.330000
                             1.330000
                                         0.700000
                                                     1.000000
               0.920000
                             2.040000
                                         1.530000
                                                     1.000000
     max
```

1 Análisis de Normalidad

```
[ ]: numericVars =list(df.describe().columns)
numericVars
```

Prueba de normalidad univariada de cada variable

```
[]: # Anderson() SciPy

# If the returned statistic is larger than these critical values then for the

corresponding significance level,

# the null hypothesis that the data come from the chosen distribution can be

rejected.

# The returned statistic is referred to as 'A2' in the references.
```

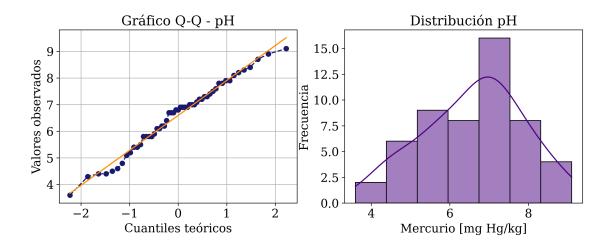
```
[]: aN = [] # Anderson Normal - Boolean
aS = [] # Anderson Statistic
```

```
aC = [] # Anderson Critical Values
    aL = [] # Anderson Significance Levels
    for var in numericVars:
      ds = df[var]
      nt = anderson(ds, dist="norm")
      aS.append(nt.statistic)
      aC.append(nt.critical_values)
      aL.append(nt.significance level)
       if nt.statistic < nt.critical_values[2]:</pre>
         aN.append(True)
      else:
         aN.append(False)
    univar_normal = pd.DataFrame({"Variable":numericVars, "Normal":aN, "Statistic":
      →aS, "Crit":aC, "Significance":aL})
    univar_normal.head(10)
[]:
          Variable Normal
                                                                     Crit \
                            Statistic
      Alcalinidad
                     False
                             3.672490
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    0
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    1
                рΗ
                      True
                             0.349555
    2
                     False
                             4.050986
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
            Calcio
    3
         Clorofila
                    False
                             5.428596 [0.54, 0.615, 0.738, 0.861, 1.024]
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    4
          Mercurio
                    False
                             0.925285
           N Peces False 8.694305
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    5
    6 MinMercurio
                    False
                            1.977048
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    7 MaxMercurio
                     True
                             0.658470
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    8
        Estimacion
                     False
                             1.046948
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
    9
                     False 14.334971
                                       [0.54, 0.615, 0.738, 0.861, 1.024]
              Edad
                      Significance
    0 [15.0, 10.0, 5.0, 2.5, 1.0]
    1 [15.0, 10.0, 5.0, 2.5, 1.0]
    2 [15.0, 10.0, 5.0, 2.5, 1.0]
    3 [15.0, 10.0, 5.0, 2.5, 1.0]
    4 [15.0, 10.0, 5.0, 2.5, 1.0]
    5 [15.0, 10.0, 5.0, 2.5, 1.0]
    6 [15.0, 10.0, 5.0, 2.5, 1.0]
    7 [15.0, 10.0, 5.0, 2.5, 1.0]
    8 [15.0, 10.0, 5.0, 2.5, 1.0]
    9 [15.0, 10.0, 5.0, 2.5, 1.0]
[]: print("Hay", univar_normal.Normal.sum(), "variable(s) Anderson normale(s)")
```

Hay 2 variable(s) Anderson normale(s)

Variable normal de acuerdo con la prueba de Anderson-Darling

```
[]: univar_normal.loc[aN]
[]:
          Variable Normal Statistic
                                                                      Crit \
                                        [0.54, 0.615, 0.738, 0.861, 1.024]
     1
                Нq
                      True
                              0.349555
     7 MaxMercurio
                      True
                                        [0.54, 0.615, 0.738, 0.861, 1.024]
                              0.658470
                      Significance
     1 [15.0, 10.0, 5.0, 2.5, 1.0]
     7 [15.0, 10.0, 5.0, 2.5, 1.0]
[]: plt.rcParams['figure.dpi'] = 300
     plt.rcParams['font.family'] = 'serif'
     plt.rcParams['font.size'] = 15
     # plt.rcParams["text.usetex"] = True
     plt.rcParams["axes.titlesize"] = 17
     # plt.rcParams.update({"figure.dpi": 180, "font.family": "serif", "text.usetex":
     → True, "font.size": 14, "axes.titlesize": 16,})
     fig, axes = plt.subplots(1,2, figsize=(12,4))
     probplot(df.pH, plot = axes[0])
     axes[0].get_lines()[0].set_marker('o')
     axes[0].get_lines()[0].set_linestyle('--')
     axes[0].get_lines()[0].set_color('midnightblue')
     axes[0].get_lines()[1].set_color('darkorange')
     axes[0].set_title("Gráfico Q-Q - pH")
     axes[0].set_xlabel("Cuantiles teóricos")
     axes[0].set_ylabel("Valores observados")
     axes[0].grid()
     sns.histplot(data = df, x = "pH", kde = True, color="indigo", ax = axes[1])
     axes[1].set_title("Distribución pH")
     axes[1].set_xlabel("Mercurio [mg Hg/kg]")
     axes[1].set_ylabel("Frecuencia")
     plt.show()
```



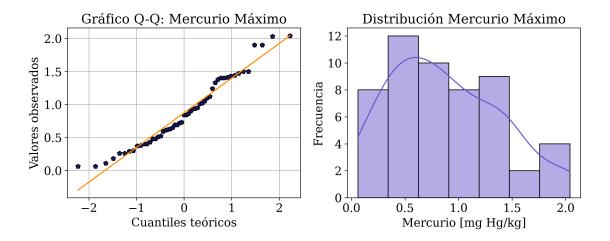
```
[]: df.MaxMercurio.skew()
```

[]: 0.4920966176197896

[]: df.MaxMercurio.kurtosis()

[]: -0.5149449394161021

```
[]: fig, axes = plt.subplots(1,2, figsize=(12,4))
     probplot(df.MaxMercurio, plot = axes[0])
     axes[0].get_lines()[0].set_marker('p')
     axes[0].get lines()[0].set markeredgecolor("k")
     # axes[0].get_lines()[0].set_linestyle('-')
     axes[0].get_lines()[0].set_color('midnightblue')
     axes[0].get_lines()[1].set_color('darkorange')
     axes[0].set_title("Gráfico Q-Q: Mercurio Máximo")
     axes[0].set_xlabel("Cuantiles teóricos")
     axes[0].set_ylabel("Valores observados")
     axes[0].grid()
     sns.histplot(data = df, x = "MaxMercurio", kde = True, color="slateblue", ax =__
      ⇒axes[1])
     axes[1].set_title("Distribución Mercurio Máximo")
     axes[1].set xlabel("Mercurio [mg Hg/kg]")
     axes[1].set_ylabel("Frecuencia")
     plt.show()
```



```
[]: multivariate_normality(df[["pH", "MaxMercurio"]], alpha=0.05)

[]: HZResults(hz=0.7695729297184755, pval=0.10247634232414388, normal=True)
```

Prueba de normalidad multivariada de todas las combinaciones

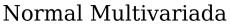
```
[]: numericVars
```

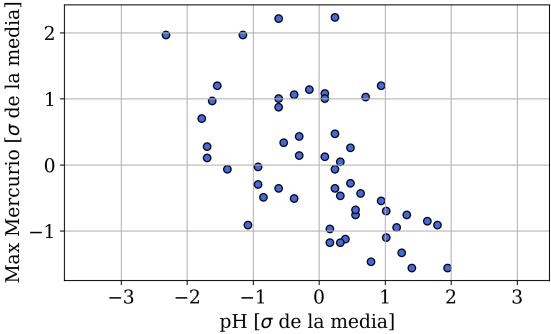
```
[]: from itertools import combinations

nn = len(numericVars)
c = 0

for i in range(2, nn):
    comb = combinations(numericVars, i)
    for j in comb:
        cols = list(j)
        # print(cols)
        data = df[cols]
        lol = multivariate_normality(data, alpha=0.05)
        if lol.normal:
```

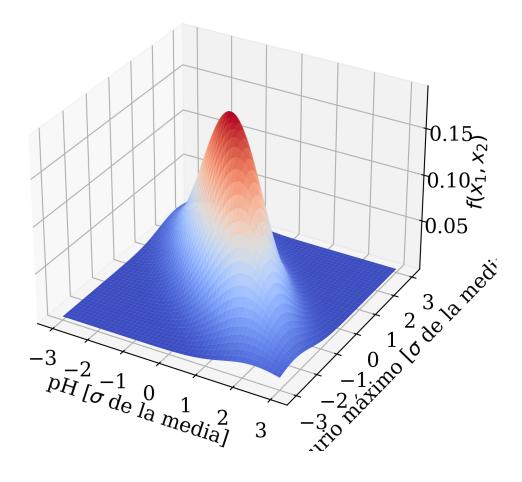
```
c += 1
           print(list(data.columns))
     print("Hay", c, "normal(es) multivariada(s)")
    ['pH', 'MaxMercurio']
    Hay 1 normal(es) multivariada(s)
[]: X1 = (df.pH - df.pH.mean())/df.pH.std()
     X2 = (df.MaxMercurio - df.MaxMercurio.mean())/df.MaxMercurio.std()
     plt.scatter(X1, X2, color = "royalblue", edgecolor="k")
     fig = plt.gcf()
     fig.set_size_inches(7, 4)
     plt.title("Normal Multivariada")
     plt.ylabel(r"Max Mercurio [$\sigma$ de la media]")
     plt.xlabel(r"pH [$\sigma$ de la media]")
     plt.grid()
     plt.axis("equal")
     plt.show()
```





```
[]: S = np.cov(X1, X2)
     S
[]: array([[1.
                        , -0.55181523],
            [-0.55181523, 1.
                                      ]])
[]: rho = -S[0,1]/(np.sqrt(S[0,0])*np.sqrt(S[1,1]))
     rho
[]: 0.5518152316870063
[]: lol = np.ones(len(df))*(-0.05)
[]: x = np.linspace(-3,3, 200)
     y = np.linspace(-3,3, 200)
     rv = np.random.multivariate_normal([X1.mean(), X2.mean()], S, size = 250)
     X, Y = np.meshgrid(x,y)
     f = (1/(2*np.pi*np.sqrt(1-rho**2)))*np.exp(-(1/(2*(1-rho**2))) * (X**2 +_{\square}))
      \hookrightarrowY**2) - 2*rho*X*Y )
[]: # from mpl_toolkits.mplot3d import axes3d
     fig = plt.figure(figsize=(12,6))
     ax = fig.add_subplot(projection="3d")
     ax.plot_surface(X, Y, f, cmap="coolwarm")
     # ax.scatter(X1, X2, zs=-0.2, color="k")
     ax.set_title("Normal Multivariada: pH y Mercurio máximo")
     ax.set_xlabel(r"pH [$\sigma$ de la media]")
     ax.set_ylabel(r"Mercurio máximo [$\sigma$ de la media]")
     ax.set_zlabel(r"$f(x_1, x_2)$")
     plt.show()
```

Normal Multivariada: pH y Mercurio máximo



```
[]: plt.contour(X, Y, f, 8, cmap='coolwarm', linestyles="solid")

fig = plt.gcf()
fig.set_size_inches(9, 5)

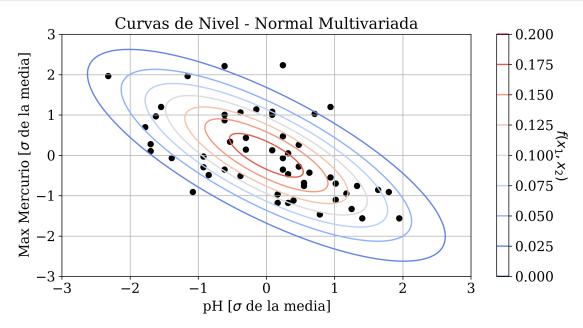
ax.set_figure
cbar = plt.colorbar()
cbar.ax.get_yaxis().labelpad = 15
cbar.ax.set_ylabel(r"\$f(x_1, x_2)\$", rotation=270)

plt.scatter(X1, X2, color = "k")

plt.title("Curvas de Nivel - Normal Multivariada")
ax.set_xlabel(r"pH [\$\sigma\$ de la media]")
ax.set_ylabel(r"Mercurio máximo [\$\sigma\$ de la media]")
plt.ylabel(r"Max Mercurio [\$\sigma\$ de la media]")
```

```
plt.xlabel(r"pH [$\sigma$ de la media]")
plt.grid()
plt.tight_layout()

plt.show()
```



```
[]: Md = pd.DataFrame(np.array([X1, X2]).T)
lam, v = np.linalg.eig(S)
```

```
def get_confidence_interval(alpha):
    gl = 2
    significance = 1 - alpha
    ci = chi2.ppf(significance, gl)
    print(ci)
    theta = np.linspace(0, 2*np.pi, len(X1));
    ab = np.sqrt(ci*lam[None,:])
    return (ab * v) @ np.array([np.sin(theta), np.cos(theta)])

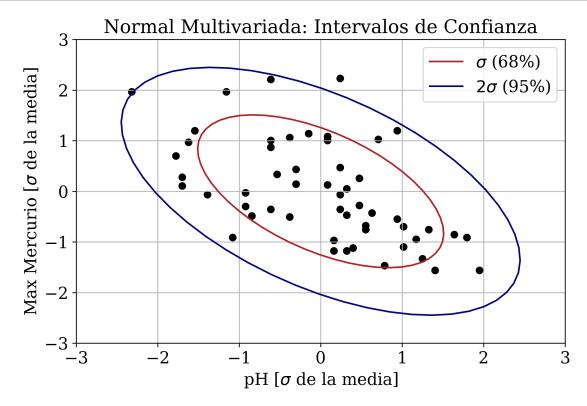
ellipsis68 = get_confidence_interval(0.32)
    ellipsis95 = get_confidence_interval(0.05)
```

- 2.278868566376729
- 5.991464547107979

```
[]: # plt.scatter(Md[0], Md[1], color = "blue")
     plt.scatter(X1, X2, color = "k", edgecolor="k")
     fig = plt.gcf()
     fig.set_size_inches(8, 5)
     plt.plot(ellipsis68[0,:], ellipsis68[1,:], color = "firebrick", __
      →label=r"$\sigma$ (68%)")
     plt.plot(ellipsis95[0,:], ellipsis95[1,:], color = "navy", label=r"$2\sigma$_\_

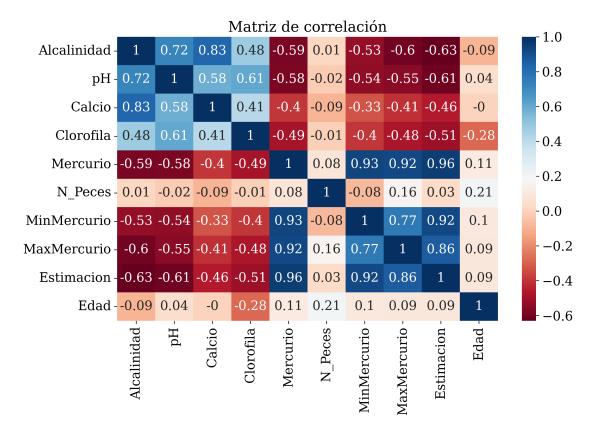
√(95%)")

     plt.title("Normal Multivariada: Intervalos de Confianza")
     plt.xlabel(r"pH [$\sigma$ de la media]")
     plt.ylabel(r"Max Mercurio [$\sigma$ de la media]")
     plt.xlim([-3, 3])
     plt.ylim([-3,3])
     plt.legend()
     plt.grid()
     plt.show()
```



2 Análisis de Componentes Principales

```
[]: cols = df.corr(numeric_only=True).columns
    dfs = df[cols]
[]: from sklearn.preprocessing import MinMaxScaler, StandardScaler
[]: | # scaler = MinMaxScaler()
    scaler = StandardScaler()
    dfs = scaler.fit_transform(dfs)
    dfs = pd.DataFrame(dfs, columns = cols)
[]: dfs.head()
[]:
       Alcalinidad
                                Calcio Clorofila Mercurio
                                                              N Peces \
                          рΗ
         -0.835862 -0.384385 -0.777523 -0.734400 2.080592 -0.950124
    0
    1
         -0.899285 -1.167939 -0.822064 -0.652498 2.376623 -0.714262
          2.073650 1.966277 0.886698 3.445888 -1.442171 -0.832193
          0.049412 0.242458 -0.234930 -0.642669 -0.258049 -0.124606
    3
         -0.925711 -1.559716 -0.781572 -0.698363 1.991783 -0.124606
       MinMercurio MaxMercurio Estimacion
                                                 Edad
    0
          2.542537
                       1.074209
                                   3.030510 0.482243
    1
          2.854675
                                   2.434418 -2.073644
                       1.983127
         -1.069346
                      -1.575190 -1.410376 -2.073644
    3
         -0.668026
                      -0.066773
                                  -0.218192 -2.073644
    4
          1.829079
                       1.209580
                                   2.434418 0.482243
[]: R = dfs.corr()
    lam, v = np.linalg.eig(R)
[]: # PCA on similar correlation variables and compare them???
    fig, ax = plt.subplots(figsize=(10, 6))
    sns.heatmap(R.round(decimals=2), cmap="RdBu", annot=True, ax = ax)
    plt.title("Matriz de correlación")
    plt.show()
```



Eigenvalores

[]: pd.DataFrame(lam).head()

[]: 0

0 5.361226

1 1.254261

2 1.216681

3 0.909433

4 0.591417

Eigenvectores

[]: pd.DataFrame(v).head()

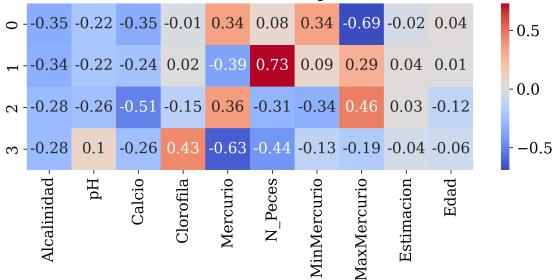
```
[]:
               0
                         1
                                   2
                                             3
                                                                  5
                                                                            6
                                                                               \
     0 -0.350659 -0.216916 -0.347291 -0.009131
                                                0.340505
                                                           0.075475
                                                                     0.338235
     1 -0.337004 -0.219409 -0.236097
                                      0.017242 -0.393960
                                                           0.731210
                                                                     0.086296
     2 -0.281683 -0.262507 -0.511378 -0.146950
                                                0.362059 -0.313423 -0.343122
     3 -0.283342 0.101951 -0.263961
                                      0.432676 -0.630934 -0.441122 -0.134352
        0.398308 -0.121042 -0.299664
                                      0.080630 -0.030469
                                                           0.074369
                                                                     0.013778
```

```
7
     0 -0.686230 -0.022398  0.042840
     1 0.287692 0.044453
                            0.013636
     2 0.455688 0.026347 -0.115083
     3 -0.190070 -0.039824 -0.063331
     4 0.016748 -0.848276 0.062433
    Varianza explicada
[]: var_expl = (lam/lam.sum())*100
     pd.DataFrame(var_expl).round(decimals=4).head()
[]:
              0
     0 53.6123
     1 12.5426
     2 12.1668
     3
        9.0943
         5.9142
[]: np.cumsum(var_expl)
[]: array([53.61226408,
                           66.15487501,
                                         78.32168878,
                                                       87.41601552,
             93.33018911, 96.36166322,
                                         98.42902665,
                                                       99.29723991,
             99.48360976, 100.
                                      ])
[]: df.head()
[]:
                                    pH Calcio
                      Alcalinidad
                                                Clorofila Mercurio
                                                                     N_{Peces}
                Lago
     0
           Alligator
                              5.9
                                  6.1
                                           3.0
                                                      0.7
                                                                1.23
                                                                            5
                                                                            7
     1
               Annie
                              3.5
                                   5.1
                                           1.9
                                                      3.2
                                                                1.33
                            116.0 9.1
                                          44.1
                                                    128.3
                                                                0.04
                                                                            6
     2
              Apopka
     3
       Blue Cypress
                             39.4
                                  6.9
                                          16.4
                                                      3.5
                                                                0.44
                                                                           12
                                           2.9
                                                      1.8
                                                                1.20
               Brick
                              2.5 4.6
                                                                           12
        MinMercurio MaxMercurio Estimacion Edad
               0.85
                            1.43
                                        1.53
     0
     1
               0.92
                            1.90
                                        1.33
                                                 0
     2
               0.04
                            0.06
                                        0.04
                                                 0
     3
               0.13
                            0.84
                                        0.44
                                                 0
                                        1.33
     4
               0.69
                            1.50
                                                 1
[]: coeffs = pd.DataFrame(v, columns = cols).loc[0:4, :]
     coeffs = coeffs.round(decimals=2)#.abs()
     # coeffs = (coeffs > 0.69/2).astype(int)
     if coeffs.min().min() > 0:
       colormap = "Purples"
     else:
```

```
colormap = "coolwarm"

# fig, ax = plt.subplots(figsize=(10,6))
fig, ax = plt.subplots(figsize=(10,3))
sns.heatmap(coeffs.loc[0:3, :], cmap=colormap, annot=True, ax = ax)
plt.title("Coeficientes de Descomposición")
# plt.title("Eigenvectores de S")
plt.show()
```

Coeficientes de Descomposición



[]: df.head()

[]:	Lago	Alcalinidad	pН	Calcio	Clorofila	Mercurio	N_Peces	١
0	Alligator	5.9	6.1	3.0	0.7	1.23	5	
1	Annie	3.5	5.1	1.9	3.2	1.33	7	
2	Apopka	116.0	9.1	44.1	128.3	0.04	6	
3	Blue Cypress	39.4	6.9	16.4	3.5	0.44	12	
4	Brick	2.5	4.6	2.9	1.8	1.20	12	

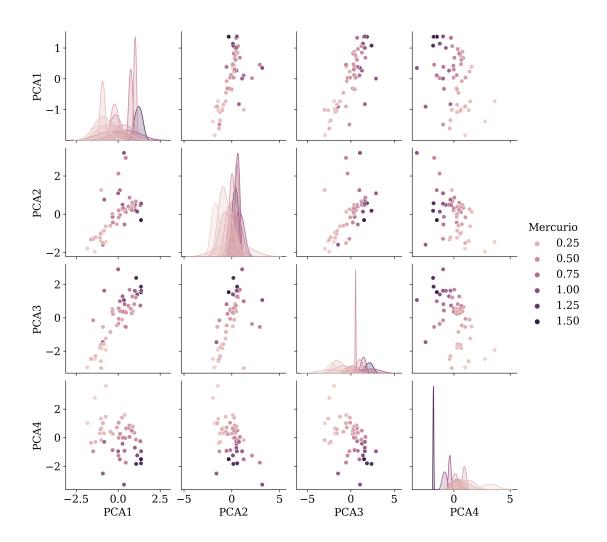
	MinMercurio	${ t MaxMercurio}$	Estimacion	Edad
0	0.85	1.43	1.53	1
1	0.92	1.90	1.33	0
2	0.04	0.06	0.04	0
3	0.13	0.84	0.44	0
4	0.69	1.50	1.33	1

PCA1 - Variacion del nivel de alcalinidad, ph y calcio en funcion del nivel de mercurio y nivel minimo de mercurio en el lago regulado por el nivel maximo de mercurio

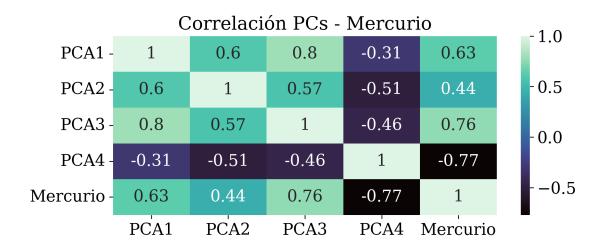
```
[]: M = np.array(dfs[cols])
[]: PCA = (v @ M.T)
    pca_columns = ["PCA1", "PCA2", "PCA3", "PCA4", "PCA5", "PCA6", "PCA7", "PCA8", __

¬"PCA9", "PCA10"]

    PCA = pd.DataFrame(PCA.T, columns = pca_columns)
    PCA.head()
[]:
           PCA1
                                       PCA4
                    PCA2
                              PCA3
                                                 PCA5
                                                          PCA6
                                                                    PCA7
    0 1.365559 -0.307733 1.534418 -1.505468 -2.734255 0.629991 2.632404
    1 1.076792 0.180404 2.387485 -1.840094 -2.270600 1.271905
                                                                1.296590
    2 -1.384938 -1.956949 -2.470802 2.777381 1.607941 0.805398 -1.852539
    3 -0.343811 -0.129749 0.514318 0.254899 0.052811 -0.870858 -1.323597
    4 1.364295 0.571494 1.864928 -1.797432 -2.061472 1.150706 1.991394
           PCA8
                    PCA9
                             PCA10
    1 -2.483843 2.595180 -0.322780
    2 1.073979 1.821683 -1.675864
    3 -0.100963 1.515232 0.355154
    4 -0.874413 0.356363 0.104145
[]: dummy = PCA.loc[:,PCA.columns[0:4]]
    dummy["Mercurio"] = df.Estimacion
    sns.pairplot(data=dummy, hue="Mercurio")
    plt.show()
```



```
[]: dummy = PCA.loc[:,PCA.columns[0:4]]
  dummy["Mercurio"] = df.Estimacion
  colormap = "mako"
  fig, ax = plt.subplots(figsize=(8,3))
  sns.heatmap(dummy.corr(), cmap=colormap, annot=True, ax = ax)
  plt.title("Correlación PCs - Mercurio")
  plt.show()
```



```
[]: plt.scatter(PCA.PCA1, PCA.PCA2, c=df.Mercurio, cmap="inferno", edgecolor="k")

# plt.colorbar()
ax.set_figure
cbar = plt.colorbar()
cbar.ax.get_yaxis().labelpad = 20
cbar.ax.set_ylabel("Mercurio [mg Hg/kg]", rotation=270)

plt.title("Componentes Principales")
plt.xlabel("PCA 1")
plt.ylabel("PCA 2")
plt.grid()
plt.axis("equal")
plt.xlim([-3, 3])
plt.ylim([-3, 3])
```

