exercicio7

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1 Redes Neurais Artificiais

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1.1 K-means RBF

```
[]: import matplotlib.pyplot as plt import numpy as np import pandas as pd
```

1.1.1 Funções de treinamento e inferência da RBF

```
[]: import numpy as np
     from sklearn.cluster import KMeans
     def trainRBF(xin, yin, p):
         # Função Radial Gaussiana
         def pdfnvar(x, m, K, n):
             if(n == 1):
                r = np.sqrt(int(K))
                px = (1/(np.sqrt(2*np.pi*r*r)))*np.exp(-0.5 * ((x-m)/(r))**2)
             else:
                 px = ((1/np.sqrt((2*np.pi)**n * (np.linalg.det(K))))) * np.exp(-0.
      \rightarrow5*((x-m) @ (np.linalg.inv(K) @ np.transpose(x-m))))
             return px
         nSamples = xin.shape[0] # Numero de amostras.
         nDimension = xin.shape[1]
                                     # Dimensao de entrada.
         xin = np.matrix(xin) # garante que xin seja matriz
         yin = np.matrix(yin) # garante que yin seja matriz
         #clusteriza os dados de entrada por meio do algorítmo K-médias
         xclust = KMeans(n_clusters=p).fit(xin)
```

```
# Armazena vetores de centros das funções
  m = np.matrix(xclust.cluster_centers_)
   covlist = []
   # Estima matrizes de covarância para todos os centros
  for i in range(p):
       ici = np.where(xclust.labels_ == i)
       xci = xin[ici,]
       if nDimension == 1:
           covi = np.var(xci)
       else:
           covi = np.cov(xci[0], rowvar=False)
       covlist.append(covi)
  H = np.zeros((nSamples, p))
   #calcula matriz H
  for j in range(nSamples):
       for i in range(p):
           mi = m[i,]
           covi = np.transpose(covlist[i]) + 0.001*np.identity(nDimension)
           aux = pdfnvar(xin[j,], mi, covi, nSamples)
           aux = np.asarray(aux)
           H[j][i] = aux[0]
  Haug = pd.DataFrame(H)
  Haug.insert(H.shape[1], H.shape[1], 1)
  Haug.to_numpy()
  W = (np.linalg.inv(Haug.T @ Haug) @ Haug.T) @ yin.T # W<-( solve(_
\rightarrow t (Hauq) %*% Hauq) %*% t (Hauq) ) %*% yin
  return [m, covlist, W, H]
```

```
[]: # Calcula a saida da rede RBF
def YRBF(xin, modRBF):

    # Função Radial Gaussiana
    def pdfnvar(x, m, K, n):
        if(n == 1):
            r = np.sqrt(int(K))
            px = (1/(np.sqrt(2*np.pi*r*r)))*np.exp(-0.5 * ((x-m)/(r))**2)
        else:
            px = ((1/np.sqrt((2*np.pi)**n * (np.linalg.det(K))))) * np.exp(-0.

→5*((x-m) @ (np.linalg.inv(K) @ np.transpose(x-m))))
        return px
```

```
nSamples = xin.shape[0] # Numero de amostras.
nDimension = xin.shape[1]
                            # Dimensao de entrada.
m = modRBF[0]
covlist = modRBF[1]
p = len(covlist) # Numero de funções radiais
W = modRBF[2]
H = np.zeros((nSamples, p))
#calcula matriz H
for j in range(nSamples):
   for i in range(p):
       mi = m[i,]
        covi = np.transpose(covlist[i]) + 0.001*np.identity(nDimension)
        aux = pdfnvar(xin[j,], mi, covi, nSamples)
        aux = np.asarray(aux)
        H[j][i] = aux[0]
Haug = pd.DataFrame(H)
Haug.insert(H.shape[1], H.shape[1], 1)
Haug.to_numpy()
Yhat = Haug @ W
return Yhat
```

1.1.2 Teste das funções RBF com base de dados Iris

```
[]: from sklearn import datasets

# import
iris = datasets.load_iris()
X = iris.data[:, :2] # we only take the first two features.
y = iris.target

xc1 = iris.data[:49,]
xc2 = iris.data[50:99,]

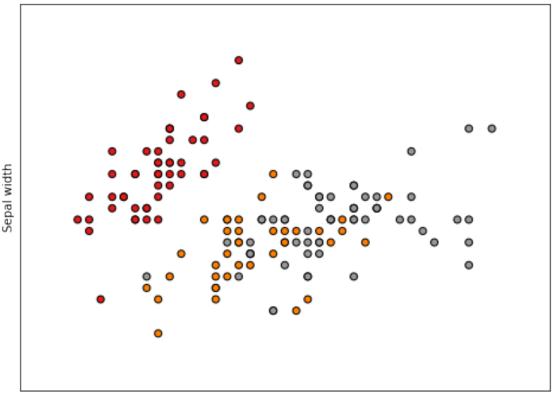
y1 = iris.target[:49,]
y2 = iris.target[50:99,]

# Plotting

x_min, x_max = X[:, 0].min() - 0.5, X[:, 0].max() + 0.5
y_min, y_max = X[:, 1].min() - 0.5, X[:, 1].max() + 0.5
```

```
plt.figure(2, figsize=(8, 6))
plt.clf()
# Plot the training points
{\tt plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Set1, edgecolor="k")}
plt.xlabel("Sepal length")
plt.ylabel("Sepal width")
plt.xlim(x_min, x_max)
plt.ylim(y_min, y_max)
plt.xticks(())
plt.yticks(())
plt.show()
# Selecionando 30 valores para treinamento
xcTrain1 = xc1[:29,]
xcTrain2 = xc2[:29,]
yTrain1 = y1[:29,]
yTrain2= y2[:29,]
xcTest1 = xc1[30:49,]
xcTest2 = xc2[30:49,]
yTest1 = y1[30:49,]
yTest2 = y2[30:49,]
yTest1 = pd.Series(yTest1)
yTest2 = pd.Series(yTest2)
# Treinamento
xcTrain = np.concatenate((xcTrain1, xcTrain2), axis=0)
yTrain = np.concatenate((yTrain1, yTrain2), axis=0)
retlist = trainRBF(xcTrain, yTrain, 6)
m = retlist[0]
covlist = retlist[1]
W = retlist[2]
H = retlist[3]
print("m: ", m)
print("covlist: ", covlist)
print("W: ", W)
print("H: ", H)
print("Vetor de pesos do Perceptron")
print("w: \n", W)
Yhat = YRBF(xcTrain, retlist)
```

```
print("Inferências")
print("Yhat: \n", Yhat)
```



Sepal length

```
m: [[6.65 3.04 4.66
                                  1.47
 [5.56666667 3.95 1.48333333 0.31666667]
           2.4625
                     3.7875
                              1.1375
 [4.6444444 3.14444444 1.33333333 0.17777778]
 [6.02727273 2.81818182 4.50909091 1.43636364]
 [5.06428571 3.45
                     1.55
                              0.26428571]]
covlist: [array([[ 0.04722222, -0.00555556, 0.01666667, -0.00722222],
      [-0.00555556, 0.02933333, -0.00266667, 0.00577778],
      [0.01666667, -0.00266667, 0.04044444, 0.01311111],
      [-0.00722222, 0.00577778, 0.01311111, 0.01344444]]), array([[
0.03466667, 0.022 , -0.00866667, -0.00533333],
      [ 0.022 , 0.059 , -0.009 , 0.011
      [-0.00866667, -0.009
                           , 0.04166667, 0.00633333],
      [-0.00533333, 0.011 , 0.00633333, 0.00966667]]), array([[
0.14857143, 0.02642857, 0.085 , -0.00357143],
      [0.02642857, 0.08839286, 0.01375, 0.02589286],
      [ 0.085 , 0.01375 , 0.08125 , 0.00910714],
```

```
[-0.00357143, 0.02589286, 0.00910714, 0.02839286]]), array([[
0.04277778, 0.00152778, 0.01708333, -0.00138889],
       [0.00152778, 0.05027778, -0.01916667, 0.00736111],
       [ 0.01708333, -0.01916667, 0.03 , 0.00083333],
       [-0.00138889, 0.00736111, 0.00083333, 0.00444444]]), array([[
0.05818182, -0.03054545, 0.00272727, -0.00909091],
       [-0.03054545, 0.07163636, -0.00618182, 0.01027273].
       [0.00272727, -0.00618182, 0.07090909, 0.01563636],
       [-0.00909091, 0.01027273, 0.01563636, 0.02654545]]), array([[
0.02401099, 0.00423077, -0.00653846, 0.00093407],
       [0.00423077, 0.035, -0.01038462, 0.00269231],
       [-0.00653846, -0.01038462, 0.02115385, 0.00192308],
       [0.00093407, 0.00269231, 0.00192308, 0.01016484]])]
W:
0 1.249814e+20
1 -1.640994e+20
2 3.676443e+20
3 -5.848053e+19
4 2.832649e+20
5 -6.344783e+19
6 5.508288e-01
H: [[1.02558146e-081 8.01192416e-023 1.19826354e-058 5.82847209e-022
 1.28294588e-057 9.31233004e-021]
 [7.34524595e-082 3.98859382e-025 1.26441699e-052 2.46241760e-021
 7.67512488e-060 6.13076977e-0231
 [6.35033071e-087 5.68911269e-027 1.66226488e-054 1.68022660e-020
  3.08187159e-063 9.51882577e-024]
 [7.72647347e-085 6.09495736e-028 5.15752414e-049 8.16059247e-021
  1.01523395e-061 1.59735951e-023]
 [1.06397983e-083 3.54789278e-024 9.93565152e-059 3.28383396e-022
  1.43600726e-058 6.30307023e-021]
 [1.47718587e-072 5.56047336e-021 2.10112739e-056 2.68006632e-027
  1.18045915e-050 7.45780914e-024]
 [1.37026083e-084 1.26684984e-027 3.84218975e-052 3.09826996e-021
  3.28914245e-061 2.22979942e-023]
 [3.93333875e-080 1.55759242e-023 1.43083957e-054 1.47521908e-021
  8.16910442e-057 1.19965502e-020]
 [6.24329073e-090 4.80119976e-031 4.35112277e-048 4.45969367e-021
  3.77859466e-067 6.81042656e-028]
 [2.48894049e-083 1.57711245e-025 1.36196309e-052 3.83204292e-021
  1.62793028e-059 5.63218361e-022]
 [4.27269356e-079 2.75184269e-021 1.65225007e-061 1.12637522e-024
  4.79355204e-055 5.15623611e-022]
 [1.20956492e-081 4.29995395e-026 2.98212480e-051 1.76181741e-022
  2.46472080e-057 3.77197215e-021]
 [3.22480721e-086 1.13417470e-026 5.06449392e-053 9.18334602e-021
  1.97620363e-062 1.70504637e-023]
 [1.99632690e-100 4.21965061e-036 1.53500756e-055 1.94089006e-021
```

- 1.29202747e-074 1.08819019e-032]
- [2.06284471e-090 2.69637183e-021 1.45427270e-078 8.27014859e-031
- 4.49562587e-061 1.34408511e-026]
- [3.54517370e-085 2.07161287e-021 9.07116573e-071 6.24489278e-033
- 1.59744824e-057 9.75522367e-030]
- [2.80640312e-083 2.48374668e-021 2.26655111e-066 4.39117845e-027
- 9.05018498e-057 1.11003054e-022]
- [2.19438932e-079 2.66887143e-022 2.25773037e-057 2.14849612e-022
- 2.45662228e-056 9.17655187e-021]
- [3.59889253e-072 2.45764990e-021 6.09580187e-060 4.50855449e-028
- 8.89856336e-051 3.48313590e-026]
- [4.74377825e-080 1.03740973e-022 3.70697686e-058 6.42236110e-024
- 1.35244600e-055 2.96806312e-021]
- [2.45753462e-072 7.41236508e-022 4.49529914e-054 7.42890576e-024
- 3.05749062e-051 2.79817435e-022]
- [1.03027912e-076 5.24981362e-022 8.85550406e-056 5.66444749e-024
- 4.51585760e-054 3.91582149e-021]
- [2.48938033e-097 9.29604048e-032 2.29940262e-064 1.42597643e-021
- 4.25204604e-069 3.30942017e-027]
- [4.94191789e-068 9.03961103e-025 3.94018676e-047 1.24765796e-026
- 1.39044307e-049 7.02614695e-022]
- [3.14472370e-079 2.68672226e-027 8.93882753e-047 2.90862268e-027
- 1.71310394e-053 3.11903209e-022]
- [1.04586830e-076 2.78301593e-024 2.93884012e-049 2.61329084e-021
- 9.44513980e-056 8.53272115e-022]
- [4.32749365e-073 5.67670696e-023 5.95631868e-050 2.78840220e-023
- 7.75747095e-053 6.31061207e-021]
- [1.04471276e-078 4.34346591e-022 2.24713484e-057 1.45039190e-022
- 2.01004115e-055 9.43701019e-021]
- [4.07651358e-080 3.23455318e-022 8.45306818e-059 2.94160184e-022
- 9.07372135e-057 6.23684369e-021]
- [1.70848387e-021 3.28649456e-187 1.31168631e-025 2.85513507e-156
- 7.67554381e-028 1.69267957e-209]
- [4.86515440e-021 4.03908345e-156 6.30403844e-024 3.68716870e-153
- 4.87660995e-023 1.31871410e-169]
- [4.24264172e-021 1.59058206e-201 1.09464519e-025 4.95978948e-173
- 1.33498229e-026 4.24097202e-222]
- [8.16247629e-033 1.02602732e-121 6.64052614e-022 1.33538343e-118
- 1.45526424e-025 9.51908204e-104]
- [2.58612129e-021 1.95159885e-184 3.60730635e-024 1.00273355e-158
- 2.33163368e-022 1.65862455e-177]
- [8.52180525e-030 2.32176293e-116 6.52115346e-024 4.10929893e-135
- 4.91637762e-022 8.99344323e-138]
- [1.02444632e-021 1.15336637e-160 5.55962828e-025 1.26944562e-171
- 3.13413092e-023 6.61273447e-185]
- [1.17787121e-045 9.88706882e-066 4.46776632e-022 7.06372067e-075
- 6.70202277e-035 2.86798855e-059]
- [1.82612418e-021 1.25724251e-161 1.38762288e-023 1.05296909e-139

```
4.00150378e-023 3.15089473e-177]
 [1.15102293e-032 1.71641696e-101 4.12059876e-022 2.59490647e-122
 4.78380432e-026 1.93706534e-096]
 [1.77432496e-048 1.35741479e-083 5.46749333e-022 3.04986912e-083
 5.14931090e-036 8.75173193e-070]
 [1.54736725e-023 3.29839236e-133 1.07387828e-022 1.19890921e-140
  1.04605718e-021 2.26003830e-131]
 [1.16425166e-031 1.05787828e-120 3.71980035e-022 6.05551427e-095
 2.45857558e-024 2.22119370e-109]
 [2.05802325e-024 5.93581898e-145 4.13807723e-024 1.65371421e-151
  1.70924592e-021 2.34644003e-167]
 [4.16805492e-029 6.38926109e-097 1.95395390e-022 2.80271657e-103
  1.88936449e-024 3.18512248e-086]
 [3.38399103e-021 1.38511502e-166 1.50991591e-024 3.65686457e-142
  1.96105081e-024 3.40828452e-171]
 [2.27451284e-027 2.35163772e-123 1.36947141e-024 3.86656693e-153
  6.57235200e-022 1.66170898e-142]
 [5.39280177e-032 8.40251836e-093 2.75798458e-022 4.75152717e-097
 2.13282250e-023 6.18289329e-109]
 [6.38162549e-028 9.13246701e-198 3.76889232e-025 3.30198741e-160
  1.14876226e-022 1.90454951e-158]
 [6.12863476e-032 2.91578135e-097 2.26111769e-021 2.10437390e-096
 1.20932196e-024 5.92794086e-095]
 [3.47669323e-024 1.69623083e-168 5.14983766e-027 2.33187045e-198
  1.47418247e-022 6.70802174e-185]
 [4.98905326e-024 1.91991507e-128 1.58824351e-022 1.36897627e-116
 4.76506856e-022 1.15687960e-118]
 [1.38184684e-024 2.23913715e-192 1.92114115e-025 1.48171719e-170
 4.82506703e-022 3.03029409e-190]
 [3.84339653e-029 9.02541418e-130 2.33253267e-024 2.12567532e-136
 4.28591628e-022 5.25039615e-162]
 [1.09600064e-021 1.67769160e-144 5.88209111e-023 3.15829266e-127
  3.43805479e-022 7.20637126e-148]
 [4.79547193e-021 4.73861891e-165 5.11670207e-024 1.92860263e-141
 2.42518239e-023 5.28135803e-166]
 [3.41919008e-021 5.66106063e-196 8.86766559e-025 2.42039759e-158
 3.11688928e-024 1.86298672e-203]
 [1.19331450e-021 2.32529160e-221 8.72946101e-027 8.28609555e-198
 3.96651472e-025 1.07811395e-225]
 [5.41933247e-023 4.26595307e-147 2.59607750e-023 4.71244963e-152
  2.78525494e-021 7.58455050e-153]]
Vetor de pesos do Perceptron
w:
               0
```

0 1.249814e+20

1 -1.640994e+20

2 3.676443e+20

3 -5.848053e+19

- 4 2.832649e+20
- 5 -6.344783e+19
- 6 5.508288e-01

Inferências

Yhat:

0

- 0 -0.087251
- 0.402870 1
- 2 -0.432382
- 3 0.072579
- 4 0.131126
- 5 -0.362115
- 6 0.368225
- 7 -0.299154
- 8 0.290024
- 9
- 0.290968
- 10 0.066472
- 11 0.301195
- 12 0.012698
- 13 0.437325
- 14 0.108355
- 15 0.210878
- 16 0.136204
- 17 -0.087764
- 18 0.147528
- 19 0.345112
- 20 0.411004
- 21 0.215898
- 22 0.467437
- 23 0.506100
- 24 0.531039
- 25 0.343407
- 26 0.139488
- 27 -0.127687
- 28 0.084833
- 29 0.764406
- 30 1.175014
- 31 1.081124
- 32 0.795005
- 33 0.941419
- 34 0.692490
- 35 0.687948
- 36 0.715084
- 37 0.795497
- 38 0.702334
- 39 0.751838
- 40 0.888555
- 41 0.688282

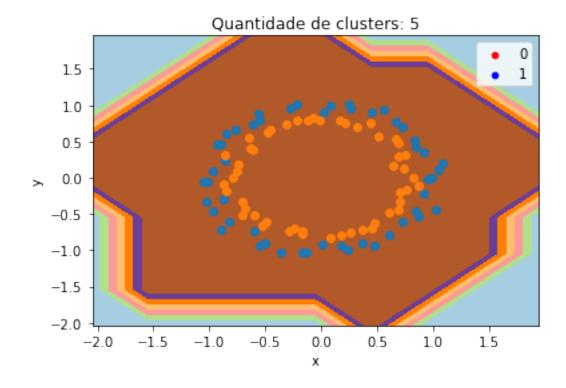
```
42 1.036777
43 0.623200
44 0.974875
45 0.737504
46 0.658266
47 0.583508
48 1.382458
49 0.593024
50 0.744821
51 0.687749
52 0.673091
53 0.806822
54 1.158924
55 0.979373
56 0.700087
57 1.356111
```

1.1.3 Exercicio

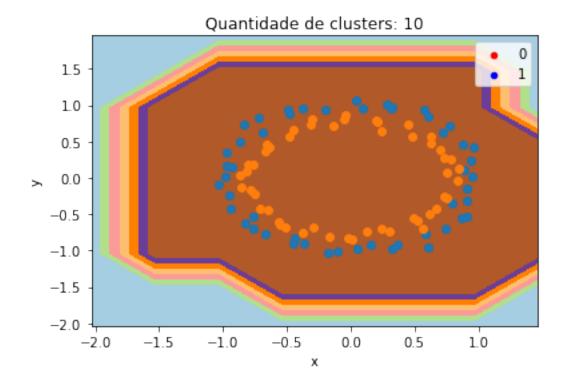
```
[]: from sklearn.datasets import make circles, make moons, make blobs
     from matplotlib import pyplot
     from pandas import DataFrame
     from numpy import where
     from numpy import meshgrid
     from numpy import arange
     from numpy import hstack
     def plotContour(format, nClusters):
         if(format == 1):
             X, y = make_circles(n_samples=100, noise=0.05)
         elif(format == 2):
             X, y = make_moons(n_samples=100, noise=0.05)
         # scatter plot, dots colored by class value
         # define bounds of the domain
         min1, max1 = X[:, 0].min()-1, X[:, 0].max()+1
         min2, max2 = X[:, 1].min()-1, X[:, 1].max()+1
         # define the x and y scale
         x1grid = arange(min1, max1, 0.5)
         x2grid = arange(min2, max2, 0.5)
         # create all of the lines and rows of the grid
         xx, yy = meshgrid(x1grid, x2grid)
```

```
# flatten each grid to a vector
   r1, r2 = xx.flatten(), yy.flatten()
   r1, r2 = r1.reshape((len(r1), 1)), r2.reshape((len(r2), 1))
    # horizontal stack vectors to create x1,x2 input for the model
   grid = hstack((r1,r2))
   df = DataFrame(dict(x=X[:,0], y=X[:,1], label=y))
   colors = {0:'red', 1:'blue'}
   fig, ax = pyplot.subplots()
   grouped = df.groupby('label')
   for key, group in grouped:
       group.plot(ax=ax, kind='scatter', x='x', y='y', label=key, __
retlist = trainRBF(X[:,:2], y, nClusters)
   m = retlist[0]
   covlist = retlist[1]
   W = retlist[2]
   H = retlist[3]
   #plotting contours
    # Make prediction from training process
   yhat = YRBF(grid, retlist)
   yhat = (yhat > 0.5).astype(int)
   yhat = pd.DataFrame(yhat).to_numpy()
   # reshape the predictions back into a grid
   zz = yhat.reshape(xx.shape)
    # plot the grid of x, y and z values as a surface
   pyplot.contourf(xx, yy, zz, cmap='Paired')
    # create scatter plot for samples from each class
   for class_value in range(2):
        # get row indexes for samples with this class
       row_ix = where(y == class_value)
        # create scatter of these samples
       pyplot.scatter(X[row_ix, 0], X[row_ix, 1], cmap='Paired')
    # show the plot
   if(nClusters == 5):
       plt.title('Quantidade de clusters: 5')
   elif(nClusters == 10):
       plt.title('Quantidade de clusters: 10')
   elif(nClusters == 30):
        plt.title('Quantidade de clusters: 30')
   pyplot.show()
plotContour(1,5)
```

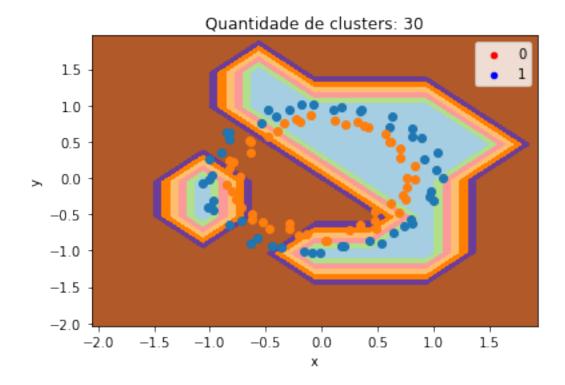
```
plotContour(1,10)
plotContour(1,30)
plotContour(2,5)
plotContour(2,10)
plotContour(2,30)
```



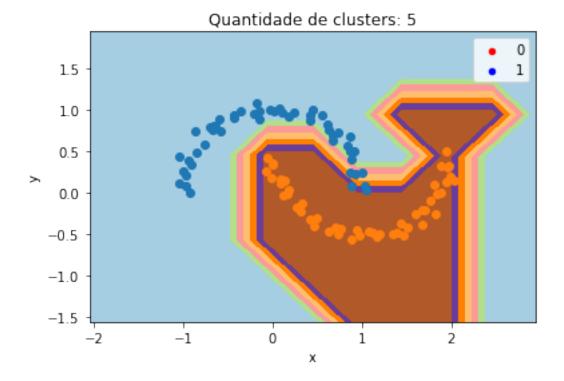
/opt/homebrew/lib/python3.9/site-packages/sklearn/utils/validation.py:593:
FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html warnings.warn(

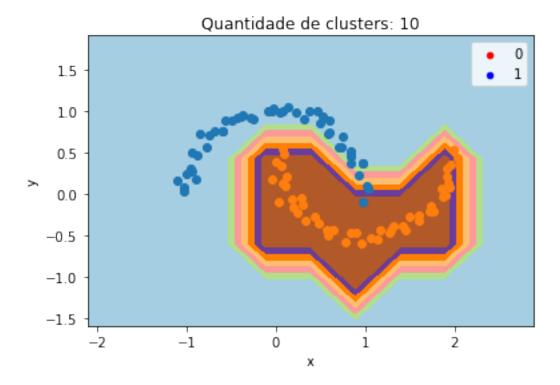


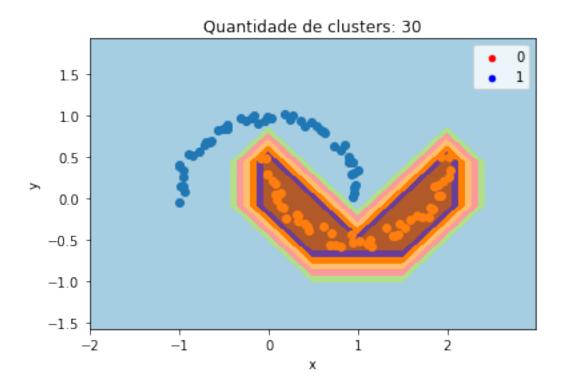
/opt/homebrew/lib/python3.9/site-packages/sklearn/utils/validation.py:593:
FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html warnings.warn(



/opt/homebrew/lib/python3.9/site-packages/sklearn/utils/validation.py:593:
FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html warnings.warn(







1.1.4 Aproximação da função Sinc

```
[]: xin = np.arange(0, 2*np.pi, 0.1*np.pi)

yin = pd.Series(np.sinc(xin))

retlist = trainRBF(xin, yin, 3)

yhat = YRBF(xin, retlist)

plt.plot(range(1, len(yhat) + 1), yhat)

plt.show()
```

```
IndexError Traceback (most recent call ⊔ → last)
```

/Users/jota/Documents/RNA/Artificial Neural Networks/RBF/part1/ \rightarrow exercicio7.ipynb Cell 12' in <cell line: 5>()

```
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000012?line=0'>1</
\rightarrowa> xin = np.arange(0, 2*np.pi, 0.1*np.pi)
         <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
-Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000012?line=2'>3</
→a> yin = pd.Series(np.sinc(xin))
   ----> <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000012?line=4'>5</
→a> retlist = trainRBF(xin, yin, 3)
         <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000012?line=6'>7</
\rightarrowa> yhat = YRBF(xin, retlist)
         <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000012?line=8'>9</
→a> plt.plot(range(1, len(yhat) + 1), yhat)
       /Users/jota/Documents/RNA/Artificial Neural Networks/RBF/part1/
→exercicio7.ipynb Cell 5' in trainRBF(xin, yin, p)
        <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000003?
→line=12'>13</a>
                      return px
        <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000003?
\rightarrowline=14'>15</a> nSamples = xin.shape[0]
                                               # Numero de amostras.
   ---> <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000003?
→line=15'>16</a> nDimension = xin.shape[1]
                                                 # Dimensao de entrada.
        <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000003?
→line=17'>18</a> xin = np.matrix(xin) # garante que xin seja matriz
        <a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>
→Artificial%20Neural%20Networks/RBF/part1/exercicio7.ipynb#ch0000003?
→line=18'>19</a> yin = np.matrix(yin) # garante que yin seja matriz
```

<a href='vscode-notebook-cell:/Users/jota/Documents/RNA/</pre>

IndexError: tuple index out of range