Sistemas de Información y Telemedicina. *

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November 25, 2019

1 Preámbulo

```
1
     names of variables
   labels = ['age', 'leptin', 'bmi', 'adiponectin', 'glucose',
            'resistin', 'insulin', 'MCP1', 'HOMA']
3
4
5
     loads data
6
   data = np.loadtxt (open (r'../../data.csv', 'rb'), delimiter = ',')
     rewrites data as all the rows of data w/out nan cells
8
9
   data = data [~np.isnan (data).any (axis=1)]
10
      separates parameters into matrix x
11
                                              for x in range (len (data))])
12
        = np.array ([list (data [x][:-1])
13
      and class (1, 2) into vector y
14
                                             for x in range (len (data))])
        = np.array ([int (data [x][ -1])
15
16
     removes outliers
17
18
   data_no = data [(np.abs (stats.zscore (data)) < 3).all (axis = 1)]
        \uparrow = No Outliers
19
20
   x_no = np.array ([list (data_no [x][:-1]) for x in range (len (data_no))])
21
   y_no = np.array ([int (data_no [x][ -1]) for x in range (len (data_no))])
22
```

Listing 1: Importaciones iniciales y preparacion de datos.

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2 Histogramas

En este apartado dibujamos los histogramas comparativos.

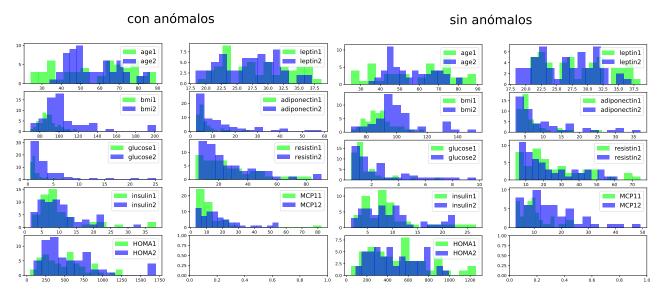


Fig. 1: Histogramas para datos con y sin anomalias.

```
1
   import matplotlib as mpl
2
   import matplotlib.pyplot as plt
3
   \# load preprocessed data, x and y are raw, x_no and y_no contain no outliers
4
   from preprocessing import x, y, x_no, y_no, labels
6
7
   # colours for the histograms
   fc = [(), (0, 1, 0, 0.6), (0, 0, 1, 0.6)]
              (R, G, B, \alpha )\leftarrow transparency
9
10
   fig, ax = plt.subplots (nrows = 5, ncols = 2, figsize = (13, 10))
11
   ax = ax.flatten ()
12
13
   # draws each of the histograms, two for each variable
14
   for i in range (0, 9):
15
       for j in [1, 2]:
16
            ax[i].hist (x [y == j, i], bins = 15, fc = fc [j], label = labels [i] + str <math>\searrow
17
                (j))
            ax[i].legend (loc = 1, prop={'size': 15})
18
19
20
   fig.suptitle ('con anómalos', fontsize = 30)
   fig.savefig ('../images/hist.pdf', bbox_inches = 'tight', pad_inches = 0)
21
```

Listing 2: Código generador de los histogramas con datos anómalos.

3 Kernel Density

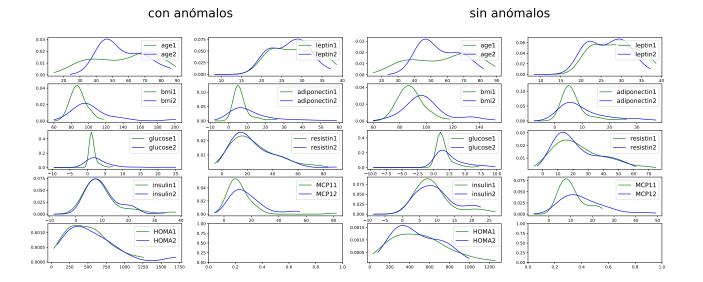


Fig. 2: Kernel Density para datos con y sin anomalias.

import matplotlib as mpl 1 import matplotlib.pyplot as plt import numpy as np 3 from scipy.stats import gaussian_kde 4 5 6 # load preprocessed data, x and y are raw, x_no and y_no contain no outliers from preprocessing import x, y, x_no, y_no, labels 7 8 9 # colours fc = ['', 'green', 'blue'] 10 11 fig, ax = plt.subplots (nrows = 5, ncols = 2, figsize = (13, 10)) 12 ax = ax.flatten ()13 14 15 # same loop in principle as before for i in range (0, 9): 16 for j in [1, 2]: 17 $kde = gaussian_kde (x_ := x [y == j, i])$ 18 $xs = np.linspace(np.min (x_) - 10, np.max (x_), num=len (x_))$ 19 ax[i].plot (xs, kde(xs), c = fc[j], label = labels [i] + str (j)) 20 ax[i].legend (loc = 1, prop={'size': 15}) 21 22

Listing 3: Código generador de los kernel density plots con datos anómalos.

fig.savefig ('../images/kden.pdf', bbox_inches = 'tight', pad_inches = 0)

fig.suptitle ('con anómalos', fontsize = 30)

23

4 Boxplot

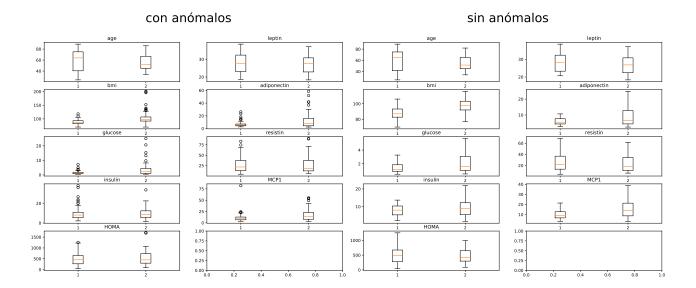


Fig. 3: Boxplots para datos con y sin anomalias.

```
import matplotlib as mpl
2
   import matplotlib.pyplot as plt
3
   \# load preprocessed data, x and y are raw, x_no and y_no contain no outliers
4
   from preprocessing import x, y, x_no, y_no, labels
5
6
7
   fig, ax = plt.subplots (nrows = 5, ncols = 2, figsize = (13, 10))
8
   ax = ax.flatten ()
9
   for i in range (0, 9):
10
       ax[i].boxplot ([x [y == 1, i], x [y == 2, i]])
11
12
       ax[i].title.set_text (labels [i])
13
   fig.suptitle ('con anómalos', fontsize = 30)
14
   fig.savefig ('../images/boxp.pdf', bbox_inches = 'tight', pad_inches = 0)
```

Listing 4: Código generador de los boxplots con datos anómalos.

5 QQplot

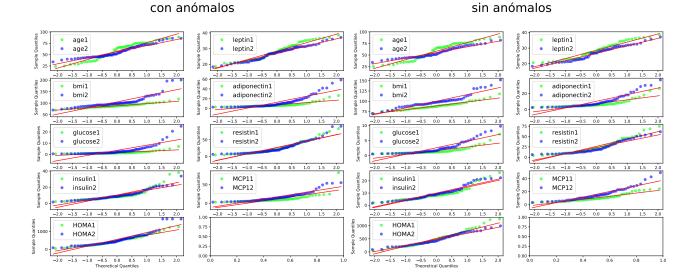


Fig. 4: QQplots para datos con y sin anomalias.

```
1
   import matplotlib as mpl
2
   import matplotlib.pyplot as plt
3
   \# load preprocessed data, x and y are raw, x_no and y_no contain no outliers
4
   from preprocessing import x, y, x_no, y_no, labels
   import statsmodels.api as sm
6
8
   fc = [(), (0, 1, 0, 0.6), (0, 0, 1, 0.6)]
9
   fig, ax = plt.subplots (nrows = 5, ncols = 2, figsize = (13, 10))
10
   ax = ax.flatten ()
11
12
   for i in range (0, 9):
13
       for j in [1, 2]:
14
           sm.qqplot (x [y == j, i], ax = ax[i], c = fc[j],
15
                    line = 's', label = labels [i] + str (j))
16
           ax[i].legend (loc = 2, prop={'size': 15})
17
18
   fig.suptitle ('con anómalos', fontsize = 30)
19
   fig.savefig ('../images/qqp.pdf', bbox_inches = 'tight', pad_inches = 0)
20
```

Listing 5: Código generador de los Q Q
plots con datos anómalos.