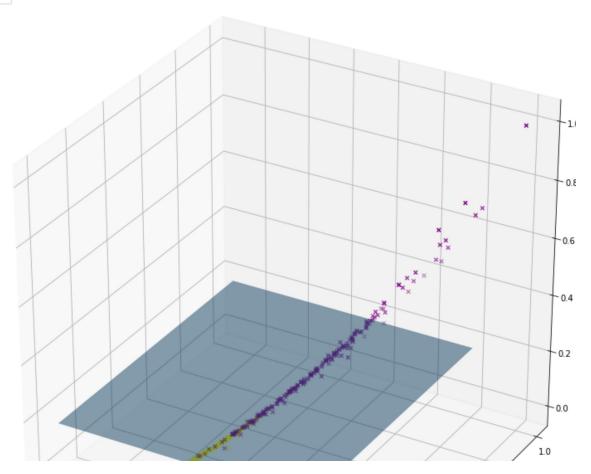
Ćwiczenie 2 Celem ćwiczenia jest samodzielne wykorzystanie wiedzy na temat wczytywania i przetwarzania języku Python. In [1]: import matplotlib.pyplot as plt 2 from mpl toolkits.mplot3d import Axes3D import pandas as pd import numpy as np from sklearn.datasets import load breast cancer from sklearn import preprocessing 7 data = load breast cancer() from sklearn.feature selection import SelectKBest from sklearn.feature selection import SelectPercentile, f regression, mutual inf In [2]: dir (data) # dir zwraca listę atrybutów danego obiektu ['DESCR', 'data', 'feature_names', 'filename', 'target', 'target_names'] Mając do dyspozycji bazę danych dotyczącą raka piersi: zwizualizuj znormalizowane dane w postaci wykresu 3D wybierz trzy cechy, które pozwalają na rozróżnienie od siebie przynajmniej dwóch klas użyj funkcji plot_surface do narysowania płaszczyny pozwalającej na rozróżnienie dwóch kl muszą być idealnie oddzielone)

```
In [3]:
              X data, y data = data.data, data.target #podział danych na macierz cech i wektor
              print (X data.shape, y data.shape)
          3
              print (X data[0], y data[0])
              print (data.target names)
                                                 #nazwy klas
          4
          5
              print (data.feature names)
                                                 #nazwy cech
           6
              #print (data.DESCR)
          (569, 30) (569,)
          [1.799e+01 1.038e+01 1.228e+02 1.001e+03 1.184e-01 2.776e-01 3.001e-01
           1.471e-01 2.419e-01 7.871e-02 1.095e+00 9.053e-01 8.589e+00 1.534e+02
           6.399e-03 4.904e-02 5.373e-02 1.587e-02 3.003e-02 6.193e-03 2.538e+01
           1.733e+01 1.846e+02 2.019e+03 1.622e-01 6.656e-01 7.119e-01 2.654e-01
           4.601e-01 1.189e-01] 0
          ['malignant' 'benign']
          ['mean radius' 'mean texture' 'mean perimeter' 'mean area'
           'mean smoothness' 'mean compactness' 'mean concavity'
           'mean concave points' 'mean symmetry' 'mean fractal dimension'
           'radius error' 'texture error' 'perimeter error' 'area error'
           'smoothness error' 'compactness error' 'concavity error'
           'concave points error' 'symmetry error' 'fractal dimension error'
           'worst radius' 'worst texture' 'worst perimeter' 'worst area'
           'worst smoothness' 'worst compactness' 'worst concavity'
           'worst concave points' 'worst symmetry' 'worst fractal dimension']
In [4]:
              skala = preprocessing.MinMaxScaler()
          2
              pandy = pd.DataFrame(data = skala.fit transform(data.data),
           3
                                                 index = data.target, columns = data.feature names)
In [5]:
              pandy
                                                                                           mean
               mean
                        mean
                                   mean
                                           mean
                                                         mean
                                                                       mean
                                                                                  mean
                                                                                                      mean
                                                                                         concave
              radius
                      texture perimeter
                                            area smoothness compactness
                                                                             concavity
                                                                                                  symmetry
                                                                                           points
                                                                                                             dime
           0.521037 0.022658 0.545989
                                         0.363733  0.593753
                                                                              0.703140
                                                                                                             0.605
                                                               0.792037
                                                                                         0.731113
                                                                                                  0.686364
           0.643144 0.272574 0.615783
                                         0.501591 0.289880
                                                                0.181768
                                                                              0.203608
                                                                                         0.348757
                                                                                                  0.379798
                                                                                                             0.141
            0.601496 0.390260 0.595743
                                         0.449417 0.514309
                                                                0.431017
                                                                              0.462512
                                                                                         0.635686
                                                                                                  0.509596
                                                                                                             0.211:
            0.210090 0.360839 0.233501
                                         0.102906 0.811321
                                                               0.811361
                                                                              0.565604
                                                                                         0.522863
                                                                                                  0.776263
                                                                                                             1.000
                                         0.489290 0.430351
          0
            0.629893 0.156578 0.630986
                                                               0.347893
                                                                              0.463918
                                                                                         0.518390
                                                                                                  0.378283
                                                                                                             0.186
                     ...
            0.690000 0.428813 0.678668
                                         0.566490 0.526948
                                                                0.296055
                                                                              0.571462
                                                                                         0.690358
                                                                                                  0.336364
                                                                                                             0.132
            0.622320 0.626987 0.604036
                                         0.474019 0.407782
                                                               0.257714
                                                                              0.337395
                                                                                         0.486630
                                                                                                  0.349495
                                                                                                             0.113
            0.455251 0.621238 0.445788
                                         0.303118 0.288165
                                                               0.254340
                                                                              0.216753
                                                                                         0.263519
                                                                                                  0.267677
                                                                                                             0.137
            0.644564 0.663510 0.665538
                                         0.475716 0.588336
                                                               0.790197
                                                                              0.823336
                                                                                         0.755467
                                                                                                  0.675253
                                                                                                             0.425
            0.036869 0.501522 0.028540
                                         0.015907 0.000000
                                                                0.074351
                                                                              0.000000
                                                                                         0.000000
                                                                                                  0.266162
                                                                                                             0.187
         569 rows × 30 columns
In [6]:
              Bestest = SelectKBest(mutual info regression, k=3)
          2
              Bestest.fit(pandy, data.target)
              columns = Bestest.get support(indices=True)
```

```
In [7]:
             Bestpandy = pandy.iloc[:,columns]
          2
             Bestpandy
            worst radius worst perimeter worst area
         0 0.620776
                         0.668310
                                        0.450698
            0.606901
                         0.539818
                                        0.435214
            0.556386
                         0.508442
                                        0.374508
                                        0.094008
            0.248310
                         0.241347
            0.519744
                         0.506948
                                        0.341575
            0.623266
                         0.576174
                                        0.452664
         0
            0.560655
                                        0.379915
                         0.520892
                                        0.230731
            0.393099
                         0.379949
            0.633582
                         0.668310
                                        0.402035
            0.054287
                         0.043578
                                        0.020497
        569 rows × 3 columns
In [8]:
          1
             X1, X2, Y1, Y2, Z1, Z2 = Bestpandy['worst radius'][0], Bestpandy['worst radius']
          2
                                           Bestpandy['worst perimeter'][0], Bestpandy['worst perim
          3
                                           Bestpandy['worst area'][0], Bestpandy['worst area'][1]
In [9]:
             dist = [[(Z1.median()+Z2.median())/2]*2]*2
In [ ]:
          1
```

```
In [10]:
            import matplotlib.pyplot as plt
         2
            from mpl toolkits.mplot3d import Axes3D
         3
            fig = plt.figure(figsize=(15, 15))
         4
         5
         6
            ax = fig.add subplot(111, projection='3d')
         7
            ax.scatter(X1, Y1, Z1, c='purple', marker='x', label=data.target names[0]);
         8
            ax.scatter(X2, Y2, Z2, c='yellow', marker='*', label=data.target names[1]);
         9
            ax.plot surface(np.array([[-.3, -.3], [.8, .8]]),
        10
                                np.array([[0, 1], [0, 1]]),
                                 np.array(dist), alpha=.5);
        11
        12
        13
            ax.set xlabel(Bestpandy.columns[0])
        14
            ax.set ylabel(Bestpandy.columns[1])
        15
        16
            ax.set zlabel(Bestpandy.columns[2])
        17
            plt.legend(loc='upper left')
        18
        19
        20
           plt.show()
```

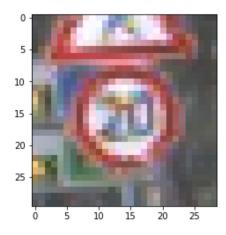
malignantbenign



In []:	1								
In []:	1								
In [11]:	1 %matplotlib inline								
	Wczytwanie i przetwarzanie danych w postaci c z wykorzystaniem formatu csv								
	Ćwiczenie 3								
	Celem ćwiczenia jest: • wyświetlanie macierzy jako obrazu; • zapoznanie z formatem csv; • przećwiczenie składni Pythona podczas implementacji algorytmu wczytywania bazy danych								
	Pobierz i rozpakuj bazę danych GTSRB http://benchmark.ini.rub.de/?section=gtsrb&subsection=dataset#Downloads) Wczytwanie bazy danych z pliku:								
In [12]:	<pre>from imageio import imread file_path = r"C:\Users\sticz\Desktop\Magisterka sezon drugi\Podstawy Uczenia Mas img = imread(file_path) img.shape (30, 29, 3)</pre>								
In [13]:	1 img.size 2610								
In [14]:	<pre>import sklearn as sk import matplotlib.pyplot as plt</pre>								
	Jaki format ma jeden rekord? Jest to obraz w formacie .ppm. Jakiego rozmiaru jest to obrazek?								
	Wyświetlamy wszystkie kanały osobno za pomoca matplotlib.								

```
f, (ax1, ax2, ax3) = plt.subplots(1, 3)
ax1.imshow(img[:,:,0], cmap='Reds')
ax2.imshow(img[:,:,1], cmap='Greens')
ax3.imshow(img[:,:,2], cmap='Blues')
plt.show()
```

Wyświetlamy wszystkie kanały na raz



Korzystając z pliku csv wczytaj obrazy z bazy danych. Następnie wyświetl 25 losowych obrazów na jednej kanwie, np. używając subplots.

```
import pandas
pandy = pandas.read_csv(r"C:\Users\sticz\Desktop\Magisterka sezon drugi\Podstawy
```

```
[n [18]:
               pandy
                     Filename Width Height Roi.X1 Roi.Y1 Roi.X2 Roi.Y2 ClassId
               00000_00000.ppm 29
                                       30
                                                       6
                                                                                0
          0
                                               5
                                                                24
                                                                        25
          1
               00000_00001.ppm 30
                                       30
                                               5
                                                        5
                                                                        25
                                                                                0
                                                                25
          2
               00000_00002.ppm 30
                                       30
                                               5
                                                        5
                                                                25
                                                                        25
                                                                                0
          3
               00000_00003.ppm 31
                                       31
                                               5
                                                        5
                                                                26
                                                                        26
                                                                                0
          4
               00000_00004.ppm 30
                                       32
                                                                25
                                               5
                                                        6
                                                                        26
                                                                                0
                                       ...
                                                       ...
                                                                ...
                                                                        ...
          ...
                                                                                ...
          205 00006_00025.ppm 85
                                       87
                                               8
                                                       7
                                                                77
                                                                        80
                                                                                0
          206 00006_00026.ppm 92
                                       95
                                               8
                                                        8
                                                                83
                                                                        87
                                                                                0
               00006_00027.ppm 97
                                       100
                                                                        92
          207
                                               9
                                                        8
                                                                89
                                                                                0
              00006_00028.ppm 105
                                       109
                                                                        100
          208
                                               9
                                                        10
                                                                95
                                                                                0
          209
              00006_00029.ppm 112
                                       118
                                               10
                                                        11
                                                                103
                                                                        108
                                                                                0
         210 rows × 8 columns
In [19]:
           1
               pandy int = pandy[['Filename','ClassId']]
           2
               display(pandy int)
                     Filename ClassId
          0
              00000_00000.ppm 0
          1
              00000_00001.ppm 0
          2
              00000_00002.ppm 0
          3
              00000_00003.ppm 0
          4
              00000_00004.ppm 0
          ...
          205 00006_00025.ppm 0
          206 00006_00026.ppm 0
          207
              00006_00027.ppm 0
          208
              00006_00028.ppm 0
              00006_00029.ppm 0
         210 rows × 2 columns
```

00003_00021.ppm 0

```
In [20]:
              pandy_25 = pandy_int.sample(25)
           2
              pandy_25
                    Filename ClassId
          185 00006_00005.ppm 0
          108 00003_00018.ppm 0
          151 00005_00001.ppm 0
          195 00006_00015.ppm 0
          48
              00001_00018.ppm 0
          193 00006_00013.ppm 0
              00002_00019.ppm 0
          79
          125 00004_00005.ppm 0
              00002_00005.ppm 0
          65
          182 00006_00002.ppm 0
          118 00003_00028.ppm 0
              00004_00011.ppm 0
          131
          84
              00002_00024.ppm 0
          38
              00001_00008.ppm 0
          0
              00000_00000.ppm 0
          165 00005_00015.ppm 0
              00000_00011.ppm 0
          11
          39
              00001_00009.ppm 0
          26
              00000_00026.ppm 0
              00000_00008.ppm 0
          8
          128 00004_00008.ppm 0
              00002_00013.ppm 0
          73
          5
              00000_00005.ppm 0
          179 00005_00029.ppm 0
```

```
In [21]:
              filename = pandy_25[['Filename']]
              display(filename)
                    Filename
          185 00006_00005.ppm
         108
              00003_00018.ppm
         151
              00005_00001.ppm
              00006_00015.ppm
         195
         48
              00001_00018.ppm
         193 00006_00013.ppm
         79
              00002_00019.ppm
         125 00004_00005.ppm
         65
              00002_00005.ppm
         182
              00006_00002.ppm
              00003_00028.ppm
         118
              00004_00011.ppm
         131
         84
              00002_00024.ppm
         38
              00001_00008.ppm
         0
              00000_00000.ppm
         165 00005_00015.ppm
         11
              00000_00011.ppm
         39
              00001_00009.ppm
         26
              00000_00026.ppm
         8
              00000_00008.ppm
         128 00004_00008.ppm
         73
              00002_00013.ppm
         5
              00000_00005.ppm
             00005_00029.ppm
         179
              00003_00021.ppm
[n [22]:
```

filenames = filename.values.tolist()

```
In [23]:
              filenames
           [['00006_00005.ppm'],
            ['00003_00018.ppm'],
            ['00005_00001.ppm'],
            ['00006_00015.ppm'],
            ['00001_00018.ppm'],
            ['00006_00013.ppm'],
            ['00002_00019.ppm'],
            ['00004_00005.ppm'],
            ['00002_00005.ppm'],
            ['00006_00002.ppm'],
            ['00003_00028.ppm'],
            ['00004_00011.ppm'],
            ['00002_00024.ppm'],
            ['00001_00008.ppm'],
            ['00000_00000.ppm'],
            ['00005_00015.ppm'],
            ['00000_00011.ppm'],
            ['00001_00009.ppm'],
            ['00000_00026.ppm'],
            ['00000_00008.ppm'],
            ['00004_00008.ppm'],
            ['00002_00013.ppm'],
            ['00000_00005.ppm'],
            ['00005_00029.ppm'],
            ['00003_00021.ppm']]
In [24]:
               import keras
               from keras.preprocessing.image import ImageDataGenerator, load img
          Using TensorFlow backend.
In [25]:
               filenames full = r"C:\Users\sticz\Desktop\Magisterka sezon drugi\Podstawy Uczeni
```

n [26]:	1	filenames_full
		Filename
18	85	C:\Users\sticz\Desktop\Magisterka sezon drugi\
10	80	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1	51	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1:	95	C:\Users\sticz\Desktop\Magisterka sezon drugi\
4	8	C:\Users\sticz\Desktop\Magisterka sezon drugi\
19	93	C:\Users\sticz\Desktop\Magisterka sezon drugi\
79	9	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1:	25	C:\Users\sticz\Desktop\Magisterka sezon drugi\
6	5	C:\Users\sticz\Desktop\Magisterka sezon drugi\
18	82	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1	18	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1:	31	C:\Users\sticz\Desktop\Magisterka sezon drugi\
84	4	C:\Users\sticz\Desktop\Magisterka sezon drugi\
3	8	C:\Users\sticz\Desktop\Magisterka sezon drugi\
0)	C:\Users\sticz\Desktop\Magisterka sezon drugi\
10	65	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1	1	C:\Users\sticz\Desktop\Magisterka sezon drugi\
39	9	C:\Users\sticz\Desktop\Magisterka sezon drugi\
20	:6	C:\Users\sticz\Desktop\Magisterka sezon drugi\
8		C:\Users\sticz\Desktop\Magisterka sezon drugi\
1:	28	C:\Users\sticz\Desktop\Magisterka sezon drugi\
73	3	C:\Users\sticz\Desktop\Magisterka sezon drugi\
5	;	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1	79	C:\Users\sticz\Desktop\Magisterka sezon drugi\
1	11	C:\Users\sticz\Desktop\Magisterka sezon drugi\

```
n [27]:
                                                                              filenamesfull list = filenames full.values.tolist()
                                                                             filenamesfull list
                                                          \label{local-posterial} $$ ['C:\Users\] \agisterka sezon drugi\Podstawy Uczenia Maszynowego\PUM-master\Laboratorium or all properties of the properties of
                                                          ining\\Images\\00000\\\00006 00005.ppm'],
                                                              ['C:\V] \label{localize} I'C:\V] \label{loca
                                                          ining\\Images\\00000\\\00003_00018.ppm'],
                                                               ['C:\V] \label{localize} I'C:\V] \label{loca
                                                         ining\\Images\\00000\\\00005_00001.ppm'],
                                                               ining\\Images\\00000\\\00006_00015.ppm'],
                                                              ining\\Images\\00000\\\00001_00018.ppm'],
                                                              ining\\Images\\00000\\\00006_00013.ppm'],
                                                               ining\\Images\\00000\\\00002 00019.ppm'],
                                                              ining\\Images\\00000\\\\00004 00005.ppm'],
                                                             ining\\Images\\00000\\\00002_00005.ppm'],
                                                              ining\\Images\\00000\\\00006_00002.ppm'],
                                                               ining\\Images\\00000\\\00003_00028.ppm'],
                                                               ining\\Images\\00000\\\00004_00011.ppm'],
                                                              ['C:\V] \label{local_posterior} I construct the construction of 
                                                          ining\\Images\\00000\\\00002_00024.ppm'],
                                                              ['C:\V] \label{local_posterior} I construct the construction of 
                                                          ining\\Images\\00000\\\00001 00008.ppm'],
                                                               ining\\Images\\00000\\\00000_00000.ppm'],
                                                              ['C:\V] \label{local_posterior} I construct the construction of 
                                                          ining\\Images\\00000\\\00005 00015.ppm'],
                                                              ining\\Images\\00000\\\00000 00011.ppm'],
                                                               ining\\Images\\00000\\\\00001 00009.ppm'],
                                                               ['C:\\Users\\sticz\\Desktop\\Magisterka sezon drugi\\Podstawy Uczenia Maszynowego\\PUM-master\\Laboratorium and the control of the control of
                                                         ining\\Images\\00000\\\\00000 00026.ppm'],
                                                              ining\\Images\\00000\\\00000 00008.ppm'],
                                                              ining\\Images\\00000\\\00004 00008.ppm'],
                                                               ining\\Images\\00000\\\00002 00013.ppm'],
                                                              ['C:\V] \label{local_posterior} I construct the construction of 
                                                          ining\\Images\\00000\\\00000 00005.ppm'],
                                                             ['C:\\Users\\sticz\\Desktop\\Magisterka sezon drugi\\Podstawy Uczenia Maszynowego\\PUM-master\\Laboratorium
                                                          ining\\Images\\00000\\\00005_00029.ppm'],
                                                              ['C:\V] \label{local_posterior} I construct the construction of 
                                                          ining\\Images\\00000\\\00003 00021.ppm']]
In [28]:
                                                                              from pathlib import Path
```

```
In [29]:
             img = imread(Path(filenamesfull_list[0][0]))
             plt.figure()
          3
             plt.imshow(img)
          4
             plt.show()
           0
           5
          10
          15
          20
          25
          30
          35
                          20
            ó
                   10
                                 30
In [30]:
             IMAGE_WIDTH=128
          2
             IMAGE_HEIGHT=128
             IMAGE_SIZE=(IMAGE_WIDTH, IMAGE_HEIGHT)
```

In [31]:	1	pandy_25	
	ı	Filename	Classid
1	85	00006_00005.ppm	
	08	00003_00018.ppm	
	51	00005_00001.ppm	
1!	95	00006_00015.ppm	
4	8	00001_00018.ppm	
19	93	00006_00013.ppm	0
79	9	00002_00019.ppm	0
1:	25	00004_00005.ppm	0
6	5	00002_00005.ppm	0
18	82	00006_00002.ppm	0
1	18	00003_00028.ppm	0
1:	31	00004_00011.ppm	0
84	4	00002_00024.ppm	0
38	8	00001_00008.ppm	0
0		00000_00000.ppm	0
10	65	00005_00015.ppm	0
1	1	00000_00011.ppm	0
3	9	00001_00009.ppm	0
20	6	00000_00026.ppm	0
8		00000_00008.ppm	0
1:	28	00004_00008.ppm	0
7:	3	00002_00013.ppm	0
5		00000_00005.ppm	0
1	79	00005_00029.ppm	0
1	11	00003_00021.ppm	0

```
In [32]:
            sample_test = pandy_25.sample(25)
         2
            sample_test.head(25)
         3
            plt.figure(figsize=(50, 48))
         4
            a = 0
         5
            lista = []
            for index, row in sample test.iterrows():
         7
                index = a
         8
                filename = row['Filename']
         9
                category = row['ClassId']
                img = load_img(filenamesfull_list[a][0], target_size=IMAGE_SIZE)
        10
                plt.subplot(10, 5, index+1)
        11
        12
                plt.imshow(img)
                plt.xlabel(filename + '(' + "{}".format(category) + ')' ) #filename +
        13
        14
                a = a+1
        15
                lista.append(category)
        16
           plt.tight_layout()
        17 plt.show()
        Type Markdown and LaTeX: \alpha^2
In [ ]:
```