

S1 - HOPE PCA_v2

November 15, 2020

0.1 Import data CSV.

```
[1]: # pip install pymysql
from sqlalchemy import create_engine
import pymysql
import pandas as pd
import numpy as np
```

```
[2]: dfPCA = pd.read_csv('hope_dataset_cleaned.csv')
```

```
[3]: dfPCA.head(10)
```

```
[3]:    pedido.data.attributes.age  pedido.data.attributes.diagnostic_main \
0                               75.0                                FISTULA PERITONEAL
1                               75.0                                FISTULA PERITONEAL
2                               75.0                                FISTULA PERITONEAL
3                               75.0                                FISTULA PERITONEAL
4                               75.0                                FISTULA PERITONEAL
5                               75.0                                FISTULA PERITONEAL
6                               75.0                                FISTULA PERITONEAL
7                               75.0                                FISTULA PERITONEAL
8                               75.0                                FISTULA PERITONEAL
9                               75.0                                FISTULA PERITONEAL

    pedido.data.attributes.gender  articulo  respuesta.articlesRevisedYear \
0                               male  27395425                             2018
1                               male  28560554                             2018
2                               male  28641726                             2017
3                               male  26245344                             2016
4                               male  28942543                             2018
5                               male  24782153                             2014
6                               male  28002229                             2018
7                               male  27505109                             2017
8                               male  24850546                             2015
9                               male  29371050                             2019

    respuesta.articlesRevisedMonth \
0                               1
```

1	4
2	12
3	12
4	6
5	6
6	9
7	4
8	1
9	4

	respuesta.pubmed_keys	utilidad
0	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	1.0
1	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
2	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
3	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
4	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
5	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
6	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
7	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
8	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN
9	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	NaN

Get only data with 'utilidad' value defined

```
[4]: dfPCA = dfPCA[pd.notnull(dfPCA['utilidad'])]

print(dfPCA.shape[0])
```

51

```
[5]: dfPCA.head(10)
```

```
[5]:      pedido.data.attributes.age  pedido.data.attributes.diagnostic_main \
0      75.0      FISTULA PERITONEAL
32     75.0      FISTULA PERITONEAL
230    36.0      INSUFICIENCIA RESPIRATORIA
290    51.0      POLITRAUMATISMO
299    51.0      POLITRAUMATISMO
300    18.0      ABDOMEN AGUDO
303    18.0      ABDOMEN AGUDO
304    18.0      ABDOMEN AGUDO
305    18.0      ABDOMEN AGUDO
311    76.0      TORACOTOMIA

      pedido.data.attributes.gender  articulo  respuesta.articlesRevisedYear \
0      male  27395425      2018
32     male  28694230      2017
```

230	male	28805236	2011
290	male	27537587	2011
299	male	28148670	2019
300	male	25055513	2019
303	male	29279563	2017
304	male	29279563	2017
305	male	28065368	2017
311	male	30762794	2019

	respuesta.articlesRevisedMonth \
0	1
32	12
230	3
290	3
299	3
300	3
303	2
304	2
305	11
311	3

	respuesta.pubmed_keys	utilidad
0	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	1.0
32	Abdomen,Blood Culture,Catharsis,Diuresis,Drug ...	1.0
230	Abdomen,Analgesics,Antitubercular Agents,Cipro...	0.0
290	Abdomen,Analgesics,Bone,Catharsis,Electroconvu...	0.0
299	Abdomen,Analgesics,Bone,Catharsis,Electroconvu...	1.0
300	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	1.0
303	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	0.0
304	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	0.0
305	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	1.0
311	Abdomen,Amiodarone,Analgesia,Angiodysplasia,Hy...	1.0

1 PCA

1.1 Transform (factorice) from Categories to continuous atributes

Transform 'pedido.data.attributes.diagnostic_main' attribute

```
[6]: categoriesORGDiagnosticMain = dfPCA['pedido.data.attributes.diagnostic_main'].
    ↪ value_counts()

print("total: " + str(categoriesORGDiagnosticMain.size))

categoriesORGDiagnosticMain
```

total: 12

```
[6]: DOLOR ABDOMINAL          13
      INFECCION DE PARTES BLANDAS    9
      INFECCION URINARIA            5
      TORACOTOMIA                  4
      ABDOMEN AGUDO                 4
      HEMORRAGIA DIGESTIVA          3
      ACV.ISQUEMICO                 3
      CETOACIDOSIS DIABETICA        3
      POLITRAUMATISMO              2
      FISTULA PERITONEAL            2
      DISNEA                        2
      INSUFICIENCIA RESPIRATORIA    1
      Name: pedido.data.attributes.diagnostic_main, dtype: int64
```

```
[7]: dataDiagnosticMain, categoriesDiagnosticMain = pd.factorize(dfPCA['pedido.data.
      ↪attributes.diagnostic_main'])
```

```
categoriesDiagnosticMain
```

```
[7]: Index(['FISTULA PERITONEAL', 'INSUFICIENCIA RESPIRATORIA', 'POLITRAUMATISMO',
      'ABDOMEN AGUDO', 'TORACOTOMIA', 'INFECCION DE PARTES BLANDAS',
      'DOLOR ABDOMINAL', 'INFECCION URINARIA', 'HEMORRAGIA DIGESTIVA',
      'ACV.ISQUEMICO', 'DISNEA', 'CETOACIDOSIS DIABETICA'],
      dtype='object')
```

0 => first element found => 'FISTULA PERITONEAL'

1 => second element found => 'INSUFICIENCIA RESPIRATORIA'

...

```
[8]: dfPCA['pedido.data.attributes.diagnostic_main'] = dataDiagnosticMain
```

```
dfPCA.head(10)
```

```
[8]:      pedido.data.attributes.age  pedido.data.attributes.diagnostic_main  \
0                                75.0                                0
32                               75.0                                0
230                              36.0                                1
290                              51.0                                2
299                              51.0                                2
300                              18.0                                3
303                              18.0                                3
304                              18.0                                3
305                              18.0                                3
311                              76.0                                4

      pedido.data.attributes.gender  articulo  respuesta.articlesRevisedYear  \
```

0	male	27395425	2018
32	male	28694230	2017
230	male	28805236	2011
290	male	27537587	2011
299	male	28148670	2019
300	male	25055513	2019
303	male	29279563	2017
304	male	29279563	2017
305	male	28065368	2017
311	male	30762794	2019

	respuesta.articlesRevisedMonth \
0	1
32	12
230	3
290	3
299	3
300	3
303	2
304	2
305	11
311	3

	respuesta.pubmed_keys	utilidad
0	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	1.0
32	Abdomen,Blood Culture,Catharsis,Diuresis,Drug ...	1.0
230	Abdomen,Analgesics,Antitubercular Agents,Cipro...	0.0
290	Abdomen,Analgesics,Bone,Catharsis,Electroconvu...	0.0
299	Abdomen,Analgesics,Bone,Catharsis,Electroconvu...	1.0
300	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	1.0
303	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	0.0
304	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	0.0
305	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	1.0
311	Abdomen,Amiodarone,Analgesia,Angiodysplasia,Hy...	1.0

Transform 'gender' atribute

```
[9]: categoriesORGGender = dfPCA['pedido.data.attributes.gender'].value_counts()

print("total: " + str(categoriesORGGender.size))

categoriesORGGender
```

total: 1

```
[9]: male    51
      Name: pedido.data.attributes.gender, dtype: int64
```

```
[10]: dataGender, categoriesGender = pd.factorize(dfPCA['pedido.data.attributes.
      ↪gender'])
```

```
categoriesGender
```

```
[10]: Index(['male'], dtype='object')
```

```
[11]: dfPCA['pedido.data.attributes.gender'] = dataGender
```

```
dfPCA.head(10)
```

```
[11]:
```

	pedido.data.attributes.age	pedido.data.attributes.diagnostic_main	\
0	75.0	0	
32	75.0	0	
230	36.0	1	
290	51.0	2	
299	51.0	2	
300	18.0	3	
303	18.0	3	
304	18.0	3	
305	18.0	3	
311	76.0	4	

	pedido.data.attributes.gender	articulo	respuesta.articlesRevisedYear	\
0	0	27395425	2018	
32	0	28694230	2017	
230	0	28805236	2011	
290	0	27537587	2011	
299	0	28148670	2019	
300	0	25055513	2019	
303	0	29279563	2017	
304	0	29279563	2017	
305	0	28065368	2017	
311	0	30762794	2019	

	respuesta.articlesRevisedMonth	\
0	1	
32	12	
230	3	
290	3	
299	3	
300	3	
303	2	
304	2	
305	11	
311	3	

	respuesta.pubmed_keys	utilidad
0	Abdomen,Adenocarcinoma,Antiemetics,Blood Cultu...	1.0
32	Abdomen,Blood Culture,Catharsis,Diuresis,Drug ...	1.0
230	Abdomen,Analgesics,Antitubercular Agents,Cipro...	0.0
290	Abdomen,Analgesics,Bone,Catharsis,Electroconvu...	0.0
299	Abdomen,Analgesics,Bone,Catharsis,Electroconvu...	1.0
300	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	1.0
303	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	0.0
304	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	0.0
305	Abdomen,Anti-Bacterial Agents,Diuresis,Operati...	1.0
311	Abdomen,Amiodarone,Analgesia,Angiodysplasia,Hy...	1.0

Transform 'respuesta.pubmed_keys' attribute

```
[12]: categoriesORGPubMedKeys = dfPCA['respuesta.pubmed_keys'].value_counts()

print("total: " + str(categoriesORGPubMedKeys.size))

categoriesORGPubMedKeys
```

total: 25

```
[12]: Abdomen,Abdominal Pain,Amebiasis,Amebic,Catharsis,Ciprofloxacin,Diarrhea,Diuresi
s,Dysentery,Extremities,Hemorrhage,Hydrocortisone,Intestines,Lower
Extremity,Metronidazole,Pain,Pain Management,Signs and
Symptoms,Tachycardia,Therapeutics,Venous Thrombosis
5
Abdomen,Anti-Bacterial Agents,Diuresis,Operative,Peritonitis,Pregnancy,Signs and
Symptoms,Surgical Procedures,Surgical Wound,Therapeutics,Wounds and Injuries
4
Abdomen,Amiodarone,Analgesia,Angiodysplasia,Hypoventilation,Lung,Operative,Surgi
cal Procedures,Thoracotomy,Tramadol
4
Abdomen,Abdominal
Pain,Amylases,Breast,Catharsis,Cholangiopancreatography,Diuresis,Endoscopic
Retrograde,Extremities,Hyperbilirubinemia,Hypertension,Hypoventilation,Lower
Extremity,Pain,Signs and Symptoms,Skin,Transaminases,Venous Thrombosis
3
Abdomen,Brain
Diseases,Catharsis,Dehydration,Diuresis,Extremities,Fibrosis,Hepatic
Encephalopathy,Lower Extremity,Quarantine,Signs and
Symptoms,Skin,Therapeutics,Venous Thrombosis
3
Abdomen,Anti-Bacterial Agents,Antihypertensive Agents,Catharsis,Diuresis,Hernia,
Hypoventilation,Ileostomy,Intestines,Loperamide,Respiratory Mechanics,Surgical
Wound,Wounds and Injuries
3
```

Abdomen, Amiodarone, Anemia, Atrial Fibrillation, Catharsis, Ceftriaxone, Cephalexin, Colic, Communicable Diseases, Diuresis, Edema, Enzymes, Hematocrit, Hematuria, Hypokalemia, Hypoventilation, Infection, Pulmonary Atelectasis, Radiography, Renal Colic, Skin, Tachycardia, Therapeutics, Urinary Tract Infections, Urologic Diseases, Work

3

Abdomen, Catharsis, Diuresis, Kidney Calculi, Lithiasis, Methods, Nephrolithiasis, Pain, Pain Management, Renal Colic, Urologic Diseases

3

Abdomen, Acromegaly, Arteries, Breast, Bundle-Branch Block, Catharsis, Chronic Obstructive, Cough, Craniotomy, Diabetes Mellitus, Diabetes Mellitus, Diuresis, Echocardiography, Edema, Electrocardiography, Extremities, Goiter, Heart Failure, Hypertension, Hypertension, Hypoventilation, Ischemia, Lower Extremity, Mastectomy, Myocardial Ischemia, Nose, Oxygen Inhalation Therapy, Piperacillin, Pulmonary, Pulmonary Artery, Pulmonary Disease, Pulmonary Embolism, Radiotherapy, Respiratory Insufficiency, Respiratory Sounds, Segmental, Tazobactam, Therapeutics, Thromboembolism, Thrombosis, Type 2, Venous Thrombosis, Volition 2

Abdomen, Diuresis, Extremities, Hernia, Hernia, Hiatal, Hypoventilation, Intestine, Intestines, Lower Extremity, Pain, Signs and Symptoms, Small, Venous Thrombosis, Wounds and Injuries

2

Abdomen, Adrenal Cortex Hormones, Amebiasis, Amebic, Catharsis, Ciprofloxacin, Colon, Cysts, Diarrhea, Disease, Diuresis, Dysentery, Edema, Heart Murmurs, Hematocrit, Inflammatory Bowel Diseases, Intestines, Metronidazole, Signs and Symptoms, Speech, Therapeutics

2

Abdomen, Analgesics, Bone, Catharsis, Electroconvulsive Therapy, Extremities, Fractures, Immunologic Memory, Lung, Medical History Taking, Signs and Symptoms

2

Abdomen, Diuresis, Extremities, Hyperplasia, Hypertension, Hypoventilation, Ileus, Intestines, Ischemia, Lower Extremity, Pain, Pain Management, Respiratory Sounds, Signs and Symptoms, Venous Thrombosis, Work, Wounds and Injuries

2

Abdomen, Aphasia, Aphasia, Atrial Appendage, Broca, Catharsis, Diabetes Mellitus, Diuresis, Hemiplegia, Hypertension, Neck, Obesity, Palpation, Rehabilitation, Respiratory Sounds, Stroke

2

Abdomen, Adenocarcinoma, Antiemetics, Blood Culture, Catharsis, Diuresis, Fistula, Gastrectomy, Incisional Hernia, Intestines, Muscles, Nausea, Pain, Pain Threshold, Palpation, Piperacillin, Pleural Effusion, Pneumonia, Quarantine, Respiratory Sounds, Signs and Symptoms, Surgical Wound, Tazobactam, Therapeutics, Thorax, Tomography, Wounds and Injuries, X-Ray Computed


```

1
Abdomen,Blood Culture,Catharsis,Ciprofloxacin,Communicable Diseases,Diuresis,Hyp
erplasia,Infection,Pain,Pelvis,Prostatitis,Therapeutics,Tramadol
1
Abdomen,Clindamycin,Diuresis,Edema,Inflammation,Molar,Therapeutics,Tomography,X-
Ray Computed
1
Abdomen,Alzheimer Disease,Communicable
Diseases,Disease,Infection,Skin,Sleep,Urinary Tract Infections
1
Diabetes Mellitus,Diabetic Ketoacidosis,Diagnosis,Ketosis
1
Abdomen,Analgesics,Antitubercular
Agents,Ciprofloxacin,Defecation,Diuresis,Intention,Intestines,Lupus
Erythematosis,Pain,Pain Management,Parenteral Nutrition,Rifampin,Streptomycin,Sy
ndrome,Systemic,Therapeutics,Tuberculosis,Wounds and Injuries
1
Chronic,Chronic,Diabetes Mellitus,Diabetic Ketoacidosis,Diagnosis,Ketosis,Kidney
Failure,Renal Insufficiency,Renal Insufficiency,Urologic Diseases
1
Abdomen,Arteries,Catharsis,Diabetes
Mellitus,Diuresis,Extremities,Hypertension,Hypoventilation,Lower
Extremity,Oxygen Inhalation Therapy,Signs and
Symptoms,Speech,Stroke,Therapeutics,Venous Thrombosis
1
Diabetes Mellitus,Diabetes Mellitus,Diabetic
Ketoacidosis,Diagnosis,Ketosis,Therapeutics,Type 1
1
Abdomen,Blood Culture,Catharsis,Diuresis,Drug
Therapy,Extremities,Fistula,Hiccup,Intestines,Lower
Extremity,Morphine,Nausea,Pain,Palpation,Piperacillin,Pneumonia,Respiratory
Sounds,Tazobactam,Therapeutics,Wounds and Injuries
1
Abdomen,Adrenal Cortex
Hormones,Catharsis,Ciprofloxacin,Cysts,Disease,Grief,Heart Murmurs,Inflammatory
Bowel Diseases,Infliximab,Intestines,Mesalamine,Metronidazole,Signs and
Symptoms,Syncope,Therapeutics
1
Name: respuesta.pubmed_keys, dtype: int64

```

```
[13]: dataPubMedKeys, categoriesPubMedKeys = pd.factorize(dfPCA['respuesta.
↳pubmed_keys'])
```

```
[14]: dfPCA['respuesta.pubmed_keys'] = dataPubMedKeys
```

```
[15]: dfPCA.head(10)
```

```
[15]: pedido.data.attributes.age  pedido.data.attributes.diagnostic_main  \
0                                75.0                                0
32                               75.0                                0
230                              36.0                                1
290                              51.0                                2
299                              51.0                                2
300                              18.0                                3
303                              18.0                                3
304                              18.0                                3
305                              18.0                                3
311                              76.0                                4

      pedido.data.attributes.gender  articulo  respuesta.articlesRevisedYear  \
0                                0  27395425                                2018
32                               0  28694230                                2017
230                              0  28805236                                2011
290                              0  27537587                                2011
299                              0  28148670                                2019
300                              0  25055513                                2019
303                              0  29279563                                2017
304                              0  29279563                                2017
305                              0  28065368                                2017
311                              0  30762794                                2019

      respuesta.articlesRevisedMonth  respuesta.pubmed_keys  utilidad
0                                1                            0        1.0
32                               12                            1        1.0
230                              3                            2        0.0
290                              3                            3        0.0
299                              3                            3        1.0
300                              3                            4        1.0
303                              2                            4        0.0
304                              2                            4        0.0
305                              11                            4        1.0
311                              3                            5        1.0
```

1.2 Standardize the Data

```
[16]: from sklearn.preprocessing import StandardScaler

features = ['pedido.data.attributes.age',
            'pedido.data.attributes.diagnostic_main',
            'pedido.data.attributes.gender',
            'respuesta.articlesRevisedYear',
            'respuesta.articlesRevisedMonth',
            'respuesta.pubmed_keys',
            'articulo']
```

```

# Separating out the features
x = dfPCA.loc[:, features].values# Separating out the target
dfPCA['utilidad']=pd.Categorical(dfPCA['utilidad'])
my_color=dfPCA['utilidad'].cat.codes

featuresTransformed = StandardScaler().fit_transform(x)

featuresTransformed

```

```

[16]: array([[ 0.91709628, -2.24479066,  0.          ,  0.48066047, -1.27483261,
        -1.81819247, -0.19317962],
       [ 0.91709628, -2.24479066,  0.          ,  0.11478459,  1.60896212,
        -1.65608091,  0.31397864],
       [-0.8747549 , -1.85670821,  0.          , -2.08047067, -0.75050629,
        -1.49396934,  0.35732434],
       [-0.18558137, -1.46862576,  0.          , -2.08047067, -0.75050629,
        -1.33185777, -0.13766811],
       [-0.18558137, -1.46862576,  0.          ,  0.84653634, -0.75050629,
        -1.33185777,  0.100948  ],
       [-1.70176313, -1.0805433 ,  0.          ,  0.84653634, -0.75050629,
        -1.16974621, -1.10687007],
       [-1.70176313, -1.0805433 ,  0.          ,  0.11478459, -1.01266945,
        -1.16974621,  0.54253988],
       [-1.70176313, -1.0805433 ,  0.          ,  0.11478459, -1.01266945,
        -1.16974621,  0.54253988],
       [-1.70176313, -1.0805433 ,  0.          ,  0.11478459,  1.34679897,
        -1.16974621,  0.06842018],
       [ 0.96304118, -0.69246085,  0.          ,  0.84653634, -0.75050629,
        -1.00763464,  1.12171293],
       [ 0.96304118, -0.69246085,  0.          ,  0.84653634, -0.75050629,
        -1.00763464,  1.12171293],
       [ 0.96304118, -0.69246085,  0.          ,  0.11478459,  1.60896212,
        -1.00763464,  0.66295943],
       [ 0.96304118, -0.69246085,  0.          ,  0.11478459, -0.22617998,
        -1.00763464, -1.74384989],
       [-1.19636921, -0.30437839,  0.          ,  0.84653634, -0.75050629,
        -0.84552307,  0.41640915],
       [-1.19636921, -0.30437839,  0.          ,  0.11478459,  1.34679897,
        -0.84552307,  0.32427757],
       [ 1.10087588,  0.08370406,  0.          ,  0.84653634, -0.22617998,
        -0.68341151, -0.59661724],
       [ 1.10087588,  0.08370406,  0.          ,  0.84653634, -0.22617998,
        -0.68341151, -0.59661724],
       [ 1.19276569,  0.47178651,  0.          ,  0.84653634, -0.22617998,
        -0.52129994,  0.25862449],
       [ 0.87115137,  0.08370406,  0.          ,  0.84653634, -0.22617998,

```

-0.35918837, 1.08237552],	
[0.87115137, 0.08370406, 0.	, -0.98284304, 1.08463581,
-0.35918837, 1.10153914],	
[0.87115137, 0.08370406, 0.	, 0.11478459, 1.34679897,
-0.35918837, 0.15789492],	
[0.45764726, 0.85986897, 0.	, 0.84653634, -0.22617998,
-0.19707681, 0.21066839],	
[0.45764726, 0.85986897, 0.	, -0.98284304, 1.08463581,
-0.19707681, 0.6678502],	
[0.45764726, 0.85986897, 0.	, -0.25109129, 1.34679897,
-0.19707681, 1.17931899],	
[0.41170236, 1.24795142, 0.	, 0.84653634, -0.48834314,
-0.03496524, -0.74218844],	
[0.41170236, 1.24795142, 0.	, -1.34871892, -0.75050629,
-0.03496524, -0.57050462],	
[1.10087588, 0.08370406, 0.	, 0.84653634, -1.01266945,
0.12714633, -0.77544256],	
[1.10087588, 0.08370406, 0.	, 0.11478459, -0.75050629,
0.12714633, -1.31346715],	
[1.10087588, 0.08370406, 0.	, -0.61696716, 0.82247265,
0.12714633, 0.11654809],	
[-1.19636921, -0.30437839, 0.	, -0.25109129, -1.01266945,
0.28925789, -2.6686866],	
[-1.19636921, -0.30437839, 0.	, 0.48066047, 1.34679897,
0.45136946, 0.89516893],	
[-1.19636921, -0.30437839, 0.	, 0.11478459, -0.75050629,
0.45136946, 0.51002923],	
[-1.19636921, -0.30437839, 0.	, 0.48066047, -0.48834314,
0.45136946, -1.57727317],	
[-1.19636921, -0.30437839, 0.	, 0.48066047, 1.60896212,
0.45136946, 0.41640915],	
[-1.19636921, -0.30437839, 0.	, -0.98284304, 0.82247265,
0.45136946, -1.10097381],	
[0.22792275, 0.47178651, 0.	, -0.25109129, 1.34679897,
0.61348103, -0.16001844],	
[0.22792275, 0.47178651, 0.	, 0.84653634, -1.01266945,
0.61348103, 0.79928405],	
[0.22792275, 0.47178651, 0.	, 0.84653634, -1.01266945,
0.61348103, 0.14857688],	
[1.10087588, 0.08370406, 0.	, -2.44634655, 0.56030949,
0.77559259, 0.98519776],	
[1.10087588, 0.08370406, 0.	, 0.84653634, -0.75050629,
0.77559259, -2.45688246],	
[0.31981255, 0.08370406, 0.	, -0.25109129, 1.34679897,
0.93770416, -1.61222666],	
[0.31981255, 0.08370406, 0.	, 0.84653634, -0.75050629,
0.93770416, 0.81440854],	

```
[ 0.31981255, 0.08370406, 0.          , 0.84653634, -1.01266945,
 0.93770416, 0.98519776],
[ 0.41170236, 1.24795142, 0.          , -0.98284304, 0.82247265,
 1.09981573, 1.04535678],
[ 0.45764726, 0.47178651, 0.          , 0.11478459, 0.82247265,
 1.26192729, 0.35655197],
[ 0.45764726, 1.63603387, 0.          , 0.84653634, -1.01266945,
 1.42403886, 0.24450666],
[ 0.45764726, 1.63603387, 0.          , -2.44634655, -0.75050629,
 1.42403886, -1.68007619],
[-0.64503039, -0.30437839, 0.          , 0.84653634, -1.01266945,
 1.58615043, -1.48790854],
[-1.51798352, 2.02411633, 0.          , -2.08047067, 1.60896212,
 1.74826199, 1.4155516 ],
[-1.51798352, 2.02411633, 0.          , -2.08047067, 1.60896212,
 1.91037356, 1.4155516 ],
[-1.51798352, 2.02411633, 0.          , 0.84653634, -1.01266945,
 2.07248513, 0.14101718]])
```

```
[17]: from sklearn.decomposition import PCA

pca = PCA(n_components=5)

pca.fit(featuresTransformed)

result=pd.DataFrame(pca.transform(featuresTransformed), columns=['PCA%i' % i_
↳for i in range(5)])

result.head(10)
```

```
[17]:      PCA0      PCA1      PCA2      PCA3      PCA4
0 -3.149930 -0.354331 -0.429642  0.398361  0.715131
1 -1.849930 -2.291115 -0.875543  0.750490 -1.002239
2 -1.339644 -1.848462  1.486153  0.957123  1.752901
3 -1.168393 -1.304228  0.969168  1.481622  1.631511
4 -2.193722 -0.367401  0.201702 -0.490824  0.143486
5 -1.949279  0.251310  1.872268 -0.348433 -0.477504
6 -1.459827 -0.731184  1.566880 -1.004888  0.762812
7 -1.459827 -0.731184  1.566880 -1.004888  0.762812
8 -0.750723 -1.795165  1.399297 -0.314486 -1.145681
9 -1.450976 -0.361905 -1.253764 -0.832103  0.674737
```

```
[18]: print('explained variance ratio (first three components): %s' %
str(pca.explained_variance_ratio_))
print('sum of explained variance (first three components): %s' %
str(sum(pca.explained_variance_ratio_)))
```

```

explained variance ratio (first three components): [0.32833981 0.23094825
0.17265137 0.14192254 0.09557452]
sum of explained variance (first three components): 0.9694364891147347

```

```

[19]: pd.DataFrame(pca.components_,columns=features,index =
↳ ['PC1','PC2','PC3','PC4','PC5'])

```

```

[19]:      pedido.data.attributes.age  pedido.data.attributes.diagnostic_main  \
PC1                -0.094054                0.603912
PC2                 0.135318                0.327845
PC3               -0.891259               -0.174252
PC4                 0.382620               -0.115174
PC5                 0.098479                0.051453

```

```

      pedido.data.attributes.gender  respuesta.articlesRevisedYear  \
PC1                -1.110223e-16                -0.365651
PC2                -5.273559e-16                 0.362710
PC3                -5.828671e-16                -0.236812
PC4                -1.929013e-15                -0.621075
PC5                 3.774758e-15                -0.540677

```

```

      respuesta.articlesRevisedMonth  respuesta.pubmed_keys  articulo
PC1                 0.338374                 0.585481  0.188307
PC2                -0.556061                 0.399721 -0.523132
PC3                -0.133609                 0.066053 -0.311447
PC4                 0.162456                -0.093185 -0.647709
PC5                -0.729252                -0.080884  0.396197

```

PC1 => diagnostic_main PC2 => pubmed_keys/Year PC3 => pubmed_keys PC4 => age PC5
=> articulo