

# urop-new

November 28, 2023

## 0.1 Preprocessing

```
[43]: import pandas as pd

# Assuming your CSV file is named 'your_file.csv'
file_path = 'oasis_longitudinal_new.csv'

# Read the CSV file into a DataFrame
df = pd.read_csv(file_path)

# Drop rows with missing values
df = df.dropna()
cleaned_file_path = 'urop.csv'
df_cleaned.to_csv(cleaned_file_path, index=False)
```

```
[44]: df.columns
```

```
[44]: Index(['Visit', 'OR Delay', 'M/F', 'Age', 'EDUC', 'SES', '00SE', 'CDR', 'eTIV',
        'nWBV', 'AS1', 'Group'],
        dtype='object')
```

```
[45]: df.info
```

```
[45]: <bound method DataFrame.info of
CDR  eTIV  nWBV  AS1  \
0      1      0  0  87  14  2.0  27.0  0.0  1987  0.696  0.883
1      2     457  0  88  14  2.0  30.0  0.0  2004  0.681  0.876
5      1      0  1  88  18  3.0  28.0  0.0  1215  0.710  1.444
6      2     538  1  90  18  3.0  27.0  0.0  1200  0.718  1.462
7      1      0  0  80  12  4.0  28.0  0.0  1689  0.712  1.039
..    ...    ...  ...  ...  ...  ...  ...  ...  ...  ...  ...
368    2     842  0  82  16  1.0  28.0  0.5  1693  0.694  1.037
369    3    2297  0  86  16  1.0  26.0  0.5  1688  0.675  1.040
370    1      0  1  61  13  2.0  30.0  0.0  1319  0.801  1.331
371    2     763  1  63  13  2.0  30.0  0.0  1327  0.796  1.323
372    3    1608  1  65  13  2.0  30.0  0.0  1333  0.801  1.317
```

Group

```

0      0
1      0
5      0
6      0
7      0
..     ...
368    1
369    1
370    0
371    0
372    0

```

[354 rows x 12 columns]>

```
[46]: df.head(50)
```

```

[46]:   Visit  OR Delay  M/F  Age  EDUC  SES  OOSE  CDR  eTIV  nWBV  AS1  Group
0      1         0    0   87   14  2.0  27.0  0.0  1987  0.696  0.883    0
1      2       457    0   88   14  2.0  30.0  0.0  2004  0.681  0.876    0
5      1         0    1   88   18  3.0  28.0  0.0  1215  0.710  1.444    0
6      2       538    1   90   18  3.0  27.0  0.0  1200  0.718  1.462    0
7      1         0    0   80   12  4.0  28.0  0.0  1689  0.712  1.039    0
8      2      1010    0   83   12  4.0  29.0  0.5  1701  0.711  1.032    0
9      3      1603    0   85   12  4.0  30.0  0.0  1699  0.705  1.033    0
13     1         0    1   93   14  2.0  30.0  0.0  1272  0.698  1.380    0
14     2       742    1   95   14  2.0  29.0  0.0  1257  0.703  1.396    0
15     1         0    0   68   12  2.0  27.0  0.5  1457  0.806  1.205    1
16     2       576    0   69   12  2.0  24.0  0.5  1480  0.791  1.186    1
17     1         0    1   66   12  3.0  30.0  0.5  1447  0.769  1.213    1
18     2       854    1   68   12  3.0  29.0  0.5  1482  0.752  1.184    1
19     1         0    1   78   16  2.0  29.0  0.0  1333  0.748  1.316    0
20     2       730    1   80   16  2.0  29.0  0.0  1323  0.738  1.326    0
21     3      1598    1   83   16  2.0  29.0  0.0  1323  0.718  1.327    0
22     1         0    1   81   12  4.0  30.0  0.0  1230  0.715  1.427    0
23     2       643    1   82   12  4.0  30.0  0.0  1212  0.720  1.448    0
24     3      1456    1   85   12  4.0  29.0  0.0  1225  0.710  1.433    0
25     1         0    0   76   16  3.0  21.0  0.5  1602  0.697  1.096    1
26     2       504    0   77   16  3.0  16.0  1.0  1590  0.696  1.104    1
27     1         0    0   88    8  4.0  25.0  0.5  1651  0.660  1.063    1
28     2       707    0   90    8  4.0  23.0  0.5  1668  0.646  1.052    1
29     1         0    0   80   12  3.0  29.0  0.0  1783  0.752  0.985    0
30     3       617    0   81   12  3.0  27.0  0.5  1814  0.759  0.968    0
31     4      1861    0   85   12  3.0  30.0  0.0  1820  0.755  0.964    0
32     5      2400    0   86   12  3.0  27.0  0.0  1813  0.761  0.968    0
33     1         0    1   87   14  1.0  30.0  0.0  1406  0.715  1.248    2
34     3       489    1   88   14  1.0  29.0  0.0  1398  0.713  1.255    2
35     4      1933    1   92   14  1.0  27.0  0.5  1423  0.696  1.234    2

```

36	1	0	0	80	20	1.0	29.0	0.0	1587	0.693	1.106	2
37	2	756	0	82	20	1.0	28.0	0.5	1606	0.677	1.093	2
38	3	1563	0	84	20	1.0	26.0	0.5	1597	0.666	1.099	2
39	1	0	0	72	20	1.0	26.0	0.5	1911	0.719	0.919	1
40	2	1164	0	76	20	1.0	25.0	0.5	1926	0.736	0.911	1
41	1	0	1	61	16	3.0	30.0	0.0	1313	0.805	1.337	0
42	2	828	1	64	16	3.0	29.0	0.0	1316	0.796	1.333	0
43	1	0	1	86	12	4.0	21.0	0.5	1247	0.662	1.407	1
44	2	578	1	87	12	4.0	21.0	0.5	1250	0.652	1.405	1
45	1	0	0	82	12	3.0	27.0	0.5	1420	0.713	1.236	1
46	2	673	0	84	12	3.0	27.0	0.5	1445	0.695	1.214	1
47	1	0	1	69	12	3.0	29.0	0.0	1365	0.783	1.286	0
48	2	609	1	71	12	3.0	30.0	0.0	1360	0.782	1.291	0
49	3	1234	1	73	12	3.0	30.0	0.0	1358	0.775	1.293	0
50	4	1779	1	74	12	3.0	30.0	0.0	1353	0.772	1.297	0
51	1	0	0	64	18	2.0	22.0	0.5	1547	0.737	1.134	1
52	2	610	0	66	18	2.0	21.0	1.0	1562	0.717	1.124	1
53	1	0	1	77	12	4.0	29.0	0.0	1377	0.734	1.275	0
54	2	1099	1	80	12	4.0	30.0	0.0	1390	0.735	1.263	0
55	1	0	1	60	18	1.0	30.0	0.0	1402	0.822	1.252	0

## 0.2 ifsm\_final\_mod.py CODE

```
[47]: import csv
import statistics
import time

# Load data from CSV
data = []
with open("urop.csv", "r") as csvfile:
    reader = csv.reader(csvfile)
    for row in reader:
        data.append([float(val) for val in row])

totalObjCount = len(data)
print('Total no. of objects=', totalObjCount)
totalLen = len(data[0])
print('Total no. of attributes=', totalLen)

# Define functions
def attrBroadcast(i):
    attr = [col[i] for col in data]
    return attr

def sNorm(a, b):
    return a + b
```

```

def truncate(n, decimals=0):
    multiplier = 10 ** decimals
    return int(n * multiplier) / multiplier

def findSim(x, y, std):
    valA = x - y + std
    valA = valA / std

    valB = y - x + std
    valB = valB / std

    s = min(valA, valB)
    sRes = max(s, 0)
    sRes = truncate(sRes, 2)
    return sRes

def similarityMatrix_ID(descId, snorm, R):
    listA = []
    decValue = decisionVariable
    attributeValue = attrBroadcastVariable
    for attrId in range(0, len(attributeValue)):
        if decValue[descId] != decValue[attrId] and descId < attrId:
            listA.append((descId, attrId, snorm, R))
    return listA

def similarityMatrix(x1, x2, Reduct, std):
    attrVal = attrBC_Var
    dSimVal = 1 - findSim(attrVal[x1], attrVal[x2], std)
    norm = sNorm(dSimVal, Reduct)
    norm = truncate(norm, 2)
    return (x1, x2, norm, Reduct)

def similarityBackward(x1, x2, Reduct, std):
    attrVal = attrBC_Var2
    dSimVal = 1 - findSim(attrVal[x1], attrVal[x2], std)
    norm = Reduct - dSimVal
    norm = truncate(norm, 2)
    return (x1, x2, norm, Reduct)

# Main
st = time.time()

decisionVariable = [row[0] for row in data]
reductOutput = []
d = {}

posReg, currPosReg = 0.0, 0.0

```

```

snorm, R = 0, 0

arrBroadCast = attrBroadcast(1)
attrBroadCastVariable = arrBroadCast

resultFirst = []
for descId in range(totalObjCount):
    resultFirst.extend(similarityMatrix_ID(descId, snorm, R))

for i in range(1, totalLen - 1):
    BC_List = attrBroadcast(i)
    attrBC_Var = BC_List
    std_dev = statistics.stdev(BC_List)
    if std_dev == 0:
        continue
    d[i] = std_dev

    resultRdd = []
    for x in resultFirst:
        resultRdd.append(similarityMatrix(x[0], x[1], x[3], std_dev))

    combinedRdd = []
    for x in resultRdd:
        combinedRdd.extend([(x[0], x[2]), (x[1], x[2])])

    storeKeyVal = {}
    for col in combinedRdd:
        key = col[0]
        val = min(1, col[1])
        if key not in storeKeyVal:
            storeKeyVal[key] = val
        else:
            storeKeyVal[key] = min(storeKeyVal[key], val)

    values1 = list(storeKeyVal.values())
    currPosReg1 = sum(values1)

    if posReg < currPosReg1:
        reductOutput.append(i)
        posReg = currPosReg1
        finalRes = []
        for x in resultRdd:
            finalRes.append((x[0], x[1], 0.0, x[2]))
        resultFirst = finalRes

    if posReg == totalObjCount:
        break

```

```

print("reductOutput after forward process :: ", reductOutput)

for i in list(reductOutput):
    BC_List = attrBroadcast(i)
    attrBC_Var2 = BC_List
    std_dev2 = d.get(i)
    if std_dev2 == 0:
        continue

    resultRdd = []
    for x in resultFirst:
        resultRdd.append(similarityBackward(x[0], x[1], x[3], std_dev2))

    combinedRdd = []
    for x in resultRdd:
        combinedRdd.extend([(x[0], x[2]), (x[1], x[2])])

    storeKeyVal = {}
    for col in combinedRdd:
        key = col[0]
        val = min(1, col[1])
        if key not in storeKeyVal:
            storeKeyVal[key] = val
        else:
            storeKeyVal[key] = min(storeKeyVal[key], val)

    values1 = list(storeKeyVal.values())
    currPosReg1 = sum(values1)

    if posReg == currPosReg1:
        reductOutput.remove(i)
        finalRes = []
        for x in resultRdd:
            finalRes.append((x[0], x[1], 0.0, x[2]))
        resultFirst = finalRes

print("reductOutput after backward elimination :: ", reductOutput)
et = time.time()

elapsed_time = et - st
print('Execution time:', elapsed_time, 'seconds')

# In[ ]:

```

```
# In[ ]:
```

Total no. of objects= 354

Total no. of attributes= 12

reductOutput after forward process :: [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]

reductOutput after backward elimination :: [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]

Execution time: 5.458693742752075 seconds

### 0.3 SVM and KNN classification

```
[48]: import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score

# Load your CSV file
file_path = 'urop.csv'
df = pd.read_csv(file_path)

# Assuming the last column is the target variable and the rest are features
selected_columns = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]
X = df.iloc[:, selected_columns]
y = df.iloc[:, -1]

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    random_state=42)

# Standardize the features (important for SVM)
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)

# K-Nearest Neighbors (KNN)
knn_model = KNeighborsClassifier(n_neighbors=5)
knn_model.fit(X_train, y_train)
y_pred_knn = knn_model.predict(X_test)
accuracy_knn = accuracy_score(y_test, y_pred_knn)
print(f'KNN Accuracy: {accuracy_knn:.2f}')

# Support Vector Machine (SVM)
svm_model = SVC(kernel='linear')
```

```
svm_model.fit(X_train_scaled, y_train)
y_pred_svm = svm_model.predict(X_test_scaled)
accuracy_svm = accuracy_score(y_test, y_pred_svm)
print(f'SVM Accuracy: {accuracy_svm:.2f}')
```

KNN Accuracy: 0.54

SVM Accuracy: 0.90

[ ]: