

Tệp quá lớn nên không hiển thị được. [Tải xuống](#)

▼ Đồ án cuối kỳ

Môn: Xử lý dữ liệu lớn

Học kỳ 1 - Năm học 2022-2023

Giảng viên: Th.S. Nguyễn Thành An

```
from google.colab import drive
drive.mount('/content/drive')
```

```
Mounted at /content/drive
```

▼ Cài đặt PySpark

```
!apt-get install openjdk-8-jdk-headless -qq > /dev/null
!wget -q http://archive.apache.org/dist/spark/spark-3.1.1/spark-3.1.1-bin-hadoop3.2.tgz
!cp drive/MyDrive/tmp/spark-3.1.1-bin-hadoop3.2.tgz .
!tar xf spark-3.1.1-bin-hadoop3.2.tgz
!pip install -q findspark
```

```
cp: cannot stat 'drive/MyDrive/tmp/spark-3.1.1-bin-hadoop3.2.tgz': No such file or directory
```

```
import os
os.environ["JAVA_HOME"] = "/usr/lib/jvm/java-8-openjdk-amd64"
os.environ["SPARK_HOME"] = "/content/spark-3.1.1-bin-hadoop3.2"
```

```
import findspark
findspark.init()
```

▼ Yêu cầu

▼ Spark Context

```
from pyspark import SparkContext
from pyspark.sql import SQLContext
```

```
sc = SparkContext("local", 'CK')
sqlc = SQLContext(sc)
```

▼ Câu 1: Giảm số chiều với SVD

```
import matplotlib.pyplot as plt
import numpy as np
from pyspark.mllib.linalg import Vectors
from pyspark.mllib.linalg.distributed import RowMatrix
from pyspark.sql import Row
```

```
def preprocess(x):
    x = x.split(',')
    matrix = np.array(x[2:]).astype(np.int)
    return (int(x[0]),int(x[1]),matrix)
```

```
rdd_train = sc.textFile("/content/drive/MyDrive/New Version/oxford_pet3_train.csv") #rdd to taming
rdd_test = sc.textFile("/content/drive/MyDrive/New Version/oxford_pet3_test.csv")
rdd_display_images = rdd_train.map(preprocess)\
    .map(lambda x: (x[0],x[1],x[2].reshape(128,128,3)))
```

```
# df = pd.read_csv("/content/drive/MyDrive/datasets/oxford_pet3_train.csv")
def plot_images_grid(rdd,rows):
    r = rdd.take(rows)
    fig = plt.figure(figsize = (12,8))
    for i in range(rows):
        title,image = r[i][1],r[i][2]
        plt.subplot(3,5,i+1)
        plt.imshow(image/255)
        plt.title(title)
    fig.show()
plot_images_grid(rdd_display_images,15)
```



```
#1.1
rows = rdd_train.map(preprocess)\
                .map(lambda v: Vectors.dense(v[2]))
#pick r = 100
mat = RowMatrix(rows)
svd = mat.computeSVD(80, computeU=True)
U = svd.U
s = svd.s
V = svd.V

l = np.array(U.rows.collect()) #out of method :^
sig = np.array(s)
r = np.array(V.toArray()).transpose()
print(l.shape)
print(sig.shape)
print(r.shape)

(500, 80)
(80,)
(80, 49152)

#reduce quality of image
rec_mat = l @ np.diag(sig) @ r
rec_mat.shape

(500, 49152)

pet3_r80_train = rdd_train.map(preprocess)\
                .map(lambda x: (x[0],x[1], str([float(n) for n in rec_mat[x[0]]])))\
                .toDF()

!rm -rf "/content/drive/MyDrive/datasets/oxford_pet3_80_train.csv"
pet3_r80_train.repartition(1).write.csv('/content/drive/MyDrive/datasets/oxford_pet3_80_train.csv')
```

```
df = sqlc.read.csv("/content/drive/MyDrive/datasets/oxford_pet3_80_train.csv")
df.show()
```

```
+---+---+-----+
|_c0|_c1|          _c2|
+---+---+-----+
| 0| 33|[11.2485358633465...|
| 1| 12|[20.5779916161160...|
| 2| 9|[12.2953497514603...|
| 3| 32|[104.467091904726...|
| 4| 33|[234.462074508373...|
| 5| 2|[186.288582218599...|
| 6| 22|[182.895838829834...|
| 7| 22|[135.102005692965...|
| 8| 36|[55.9788320216787...|
| 9| 5|[56.3647122849480...|
|10| 6|[175.741976022630...|
|11| 28|[235.354870458183...|
|12| 32|[167.915536528234...|
|13| 10|[30.3593886657564...|
|14| 14|[77.8572907709398...|
|15| 3|[32.3623421192620...|
|16| 22|[136.257473434922...|
|17| 12|[27.8430395051883...|
|18| 32|[76.7253793715973...|
|19| 5|[209.856118307185...|
+---+---+-----+
only showing top 20 rows
```

```
# #1.2
rows = rdd_test.map(preprocess)\
               .map(lambda v: v[2])
mat = RowMatrix(rows)

svd = mat.computeSVD(80, computeU=True)
U_test = svd.U
s_test = svd.s
V_test = svd.V

l_test = np.array(U_test.rows.collect())
sig_test = np.array(s_test)
r_test = np.array(V_test.toArray()).transpose()

rec_mat_test = l_test @ np.diag(sig_test) @ r_test
print(rec_mat_test.shape)
#reformat to save as csv
#((title,row),index) -> index, title, row
pet3_r100_test = rdd_train.map(preprocess)\
                        .map(lambda x: (x[0],x[1], str([float(n) for n in rec_mat_test[x[0]]])))\
                        .toDF()

!rm -rf "/content/drive/MyDrive/datasets/oxford_pet3_80_test.csv"
pet3_r100_test.repartition(1).write.csv('/content/drive/MyDrive/datasets/oxford_pet3_80_test.csv')

(500, 49152)
```

▼ Câu 2: Khuyến nghị sản phẩm với Collaborative Filtering

```
from pyspark.ml.evaluation import RegressionEvaluator
from pyspark.ml.recommendation import ALS
from pyspark.sql import Row
ratingsRDD = sqlc.read.csv('/content/drive/MyDrive/ratings2k.csv',header = "True")\
               .rdd\
               .map(lambda x:Row(index = int(x[0]),user = int(x[1]),item = int(x[2]), rating = float(x[3])))
ratings = sqlc.createDataFrame(ratingsRDD).orderBy("user")

ratings.printSchema()
ratings.show()
```

```
root
|-- index: long (nullable = true)
|-- user: long (nullable = true)
|-- item: long (nullable = true)
|-- rating: double (nullable = true)
```

```
+-----+-----+-----+-----+
|index|user|item|rating|
+-----+-----+-----+
| 390| 1| 352| 5.0|
| 32| 1| 167| 3.5|
| 1188| 1| 168| 5.0|
| 130| 1| 422| 3.5|
| 1544| 1| 163| 5.0|
| 674| 2| 288| 5.0|
| 1568| 2| 216| 1.0|
| 482| 2| 251| 5.0|
| 757| 2| 204| 4.5|
| 50| 2| 413| 3.5|
| 1108| 2| 310| 2.0|
| 139| 2| 183| 5.0|
| 1274| 2| 199| 4.5|
| 1485| 2| 271| 4.0|
| 1622| 2| 294| 4.5|
| 1097| 2| 82| 4.5|
| 1180| 2| 176| 5.0|
| 1213| 2| 0| 3.5|
| 1399| 2| 320| 2.0|
| 238| 2| 434| 4.0|
+-----+-----+-----+
only showing top 20 rows
```

```
test = ratings.filter(ratings.user > 70) #4 last user
train = ratings
```

```
from pyspark.ml.recommendation import ALS
als = ALS(rank = 5,
          maxIter = 5,
          userCol="user",
          itemCol="item",
          ratingCol="rating",
          numUserBlocks = 70,
          regParam = 0.01)
model = als.fit(train)
```

```
predictions = model.transform(test)
evaluator = RegressionEvaluator(metricName="mse", labelCol="rating",
                               predictionCol="prediction")
mse = evaluator.evaluate(predictions)
print("MSE = ",mse)
```

```
MSE = 0.19381224376720063
```

```
predictions.show()
```

```
+-----+-----+-----+-----+
|index|user|item|rating|prediction|
+-----+-----+-----+-----+
| 276| 75| 148| 3.0| 2.5445342|
| 868| 72| 85| 3.0| 3.0942104|
| 140| 72| 251| 5.0| 4.1459146|
| 146| 75| 251| 4.0| 4.0674276|
| 463| 71| 251| 5.0| 5.4261575|
| 1931| 74| 251| 4.0| 3.488337|
| 2208| 72| 451| 4.0| 4.1256137|
| 2075| 72| 255| 3.0| 3.1679611|
| 565| 72| 322| 4.0| 3.928509|
| 311| 75| 322| 4.0| 3.9876685|
| 2202| 71| 322| 5.0| 5.201237|
| 1617| 74| 322| 3.0| 3.366821|
| 1505| 72| 321| 2.0| 2.9125924|
| 919| 75| 321| 5.0| 3.2928898|
| 962| 72| 108| 2.0| 2.0628068|
| 1754| 72| 34| 4.0| 3.9747963|
| 1433| 75| 34| 3.5| 3.8938198|
| 1270| 72| 193| 4.0| 4.14925|
| 1958| 75| 193| 3.0| 3.0069091|
| 665| 74| 193| 1.0| 1.1009812|
```

```
+-----+-----+-----+-----+
only showing top 20 rows

predictions.printSchema()
print("Pearson Correlation: ",end="")
predictions.corr("rating","prediction")

root
|-- index: long (nullable = true)
|-- user: long (nullable = true)
|-- item: long (nullable = true)
|-- rating: double (nullable = true)
|-- prediction: float (nullable = false)

Pearson Correlation: 0.9135376220745998
```

▼ Câu 3: Dự đoán giá chứng khoán

```
from pyspark.sql.types import StructType, FloatType, StringType
schema = StructType() \
    .add("Ngày", StringType(), True) \
    .add("HVN", FloatType(), True)
df = sqlc.read.option("header",True).schema(schema).csv("/content/stockHVN2022.csv")

# Lấy dữ liệu tháng 1-8 làm tập train và từ tháng 9 trở lên làm tập test
def splitData(rdd):
    dataForTrain = []
    dataForTest = []
    for row in range(len(rdd)):
        month = int(rdd[row][0].split("/")[1])
        if month > 8:
            dataForTest.append([rdd[row][0], rdd[row][1]])
        else:
            dataForTrain.append([rdd[row][0], rdd[row][1]])
    return (dataForTrain,dataForTest)

# Chuyển đổi dữ liệu thành dạng cứ 5 ngày sẽ dự đoán 1 ngày
def convertDataForModel(data):
    result = []
    for i in range(len(data)-5):
        result.append([data[i+1][1],data[i+2][1],data[i+3][1],data[i+4][1],data[i+5][1],data[i][1]])
    return result

rdd = df.rdd.map(lambda x: x).collect()
dataForTrain = splitData(rdd)[0]
dataForTest = splitData(rdd)[1]

# Tạo dataframe cho tập train
df_train_data = sqlc.createDataFrame(convertDataForModel(dataForTrain),["Day1", "Day2", "Day3", "Day4", "Day5", "Nextday"])

# Tạo dataframe cho tập test
df_test_data = sqlc.createDataFrame(convertDataForModel(dataForTest),["Day1", "Day2", "Day3", "Day4", "Day5", "Nextday"])

df_train_data.show()
```

	Day1	Day2	Day3	Day4	Day5	Nextday
	17.100000381469727	17.049999237060547	17.649999618530273	17.899999618530273	17.899999618530273	17.25
	17.049999237060547	17.649999618530273	17.899999618530273	17.899999618530273	17.450000762939453	17.100000381469727
	17.649999618530273	17.899999618530273	17.899999618530273	17.450000762939453	17.399999618530273	17.049999237060547
	17.899999618530273	17.899999618530273	17.450000762939453	17.399999618530273	17.399999618530273	17.649999618530273
	17.899999618530273	17.450000762939453	17.399999618530273	17.399999618530273	17.350000381469727	17.899999618530273
	17.450000762939453	17.399999618530273	17.399999618530273	17.350000381469727	17.350000381469727	17.899999618530273
	17.399999618530273	17.399999618530273	17.350000381469727	17.350000381469727	17.700000762939453	17.450000762939453
	17.399999618530273	17.350000381469727	17.350000381469727	17.700000762939453	17.649999618530273	17.399999618530273
	17.350000381469727	17.350000381469727	17.700000762939453	17.649999618530273	17.850000381469727	17.399999618530273

```
|17.350000381469727|17.700000762939453|17.649999618530273|17.850000381469727|17.799999237060547|17.350000381469727| |
|17.700000762939453|17.649999618530273|17.850000381469727|17.799999237060547|18.100000381469727|17.350000381469727|
|17.649999618530273|17.850000381469727|17.799999237060547|18.100000381469727|17.549999237060547|17.700000762939453|
|17.850000381469727|17.799999237060547|18.100000381469727|17.549999237060547|17.600000381469727|17.649999618530273|
|17.799999237060547|18.100000381469727|17.549999237060547|17.600000381469727|17.799999237060547|17.850000381469727|
|18.100000381469727|17.549999237060547|17.600000381469727|17.799999237060547|17.100000381469727|17.799999237060547|
|17.549999237060547|17.600000381469727|17.799999237060547|17.100000381469727|17.049999237060547|18.100000381469727|
|17.600000381469727|17.799999237060547|17.100000381469727|17.049999237060547|16.950000762939453|16.950000762939453|
|17.799999237060547|17.100000381469727|17.049999237060547|16.950000762939453|16.950000762939453|16.5|17.799999237060547|
|17.100000381469727|17.049999237060547|16.950000762939453|16.950000762939453|16.5|16.100000381469727|17.100000381469727|
+-----+-----+-----+-----+-----+-----+
only showing top 20 rows
```

```
from pyspark.ml.feature import VectorAssembler

# Chuyển các feature cho model học dưới dạng vector
featureassembler = VectorAssembler(inputCols=["Day1", "Day2", "Day3", "Day4", "Day5"], outputCol="Independent Features")

df_train_data = featureassembler.transform(df_train_data)
df_test_data = featureassembler.transform(df_test_data)

df_train_data.show()
```

	Day1	Day2	Day3	Day4	Day5	Nextday	Ind
17.100000381469727	17.049999237060547	17.649999618530273	17.899999618530273	17.899999618530273	17.899999618530273	17.25	17
17.049999237060547	17.649999618530273	17.899999618530273	17.899999618530273	17.899999618530273	17.450000762939453	17.100000381469727	17
17.649999618530273	17.899999618530273	17.899999618530273	17.450000762939453	17.399999618530273	17.399999618530273	17.049999237060547	17
17.899999618530273	17.899999618530273	17.450000762939453	17.399999618530273	17.399999618530273	17.399999618530273	17.649999618530273	17
17.899999618530273	17.450000762939453	17.399999618530273	17.399999618530273	17.399999618530273	17.350000381469727	17.899999618530273	17
17.450000762939453	17.399999618530273	17.399999618530273	17.399999618530273	17.350000381469727	17.350000381469727	17.899999618530273	17
17.399999618530273	17.399999618530273	17.350000381469727	17.350000381469727	17.700000762939453	17.450000762939453	17.450000762939453	17
17.399999618530273	17.350000381469727	17.350000381469727	17.700000762939453	17.649999618530273	17.399999618530273	17.399999618530273	17
17.350000381469727	17.350000381469727	17.700000762939453	17.649999618530273	17.850000381469727	17.399999618530273	17.399999618530273	17
17.700000762939453	17.649999618530273	17.850000381469727	17.799999237060547	18.100000381469727	17.350000381469727	17.350000381469727	17
17.649999618530273	17.850000381469727	17.799999237060547	18.100000381469727	17.549999237060547	17.700000762939453	17.700000762939453	17
17.850000381469727	17.799999237060547	18.100000381469727	17.549999237060547	17.600000381469727	17.649999618530273	17.649999618530273	17
17.799999237060547	18.100000381469727	17.549999237060547	17.600000381469727	17.799999237060547	17.850000381469727	17.850000381469727	17
18.100000381469727	17.549999237060547	17.600000381469727	17.799999237060547	17.100000381469727	17.799999237060547	17.799999237060547	18
17.549999237060547	17.600000381469727	17.799999237060547	17.100000381469727	17.049999237060547	18.100000381469727	18.100000381469727	17
17.600000381469727	17.799999237060547	17.100000381469727	17.049999237060547	16.950000762939453	16.950000762939453	16.950000762939453	17
17.799999237060547	17.100000381469727	17.049999237060547	16.950000762939453	16.950000762939453	16.5	17.799999237060547	17
17.100000381469727	17.049999237060547	16.950000762939453	16.950000762939453	16.5	16.100000381469727	17.100000381469727	17

only showing top 20 rows

```
# Chỉ giữ 2 cột Features và kết quả cho các dataframe cho mô hình học và dự đoán
train_data = df_train_data.select("Independent Features", "Nextday")
test_data = df_test_data.select("Independent Features", "Nextday")

train_data.show()
```

Independent Features	Nextday
17.1000003814697...	17.25
17.0499992370605...	17.100000381469727
17.6499996185302...	17.049999237060547
17.8999996185302...	17.649999618530273
17.8999996185302...	17.899999618530273
17.4500007629394...	17.899999618530273
17.3999996185302...	17.450000762939453
17.3999996185302...	17.399999618530273
17.3500003814697...	17.399999618530273
17.3500003814697...	17.350000381469727
17.7000007629394...	17.350000381469727
17.6499996185302...	17.700000762939453
17.8500003814697...	17.649999618530273
17.7999992370605...	17.850000381469727
18.1000003814697...	17.799999237060547
17.5499992370605...	18.100000381469727
17.6000003814697...	17.549999237060547
17.7999992370605...	17.600000381469727
17.1000003814697...	17.799999237060547

```
|[17.0499992370605...|17.100000381469727|
+-----+-----+
only showing top 20 rows

from pyspark.ml.regression import LinearRegression

# Xây dựng mô hình Linear Regression
regressor = LinearRegression(featuresCol='Independent Features', labelCol='Nextday')
regressor = regressor.fit(train_data)

# Dự đoán kết quả đối với tập train
pred_results_train = regressor.evaluate(train_data)
pred_results_train.predictions.show()
```

Independent Features	Nextday	prediction
[17.1000003814697...]	17.25	17.153309991632323
[17.0499992370605...]	17.100000381469727	17.09404553004526
[17.6499996185302...]	17.049999237060547	17.631954486885153
[17.8999996185302...]	17.649999618530273	17.843904760661808
[17.8999996185302...]	17.899999618530273	17.93048536357782
[17.4500007629394...]	17.899999618530273	17.445659516454693
[17.3999996185302...]	17.450000762939453	17.325259991838923
[17.3999996185302...]	17.399999618530273	17.399598897962125
[17.3500003814697...]	17.399999618530273	17.341683127360668
[17.3500003814697...]	17.350000381469727	17.312126745594203
[17.7000007629394...]	17.350000381469727	17.661468128349224
[17.6499996185302...]	17.700000762939453	17.708736498786223
[17.8500003814697...]	17.649999618530273	17.87009558639704
[17.7999992370605...]	17.850000381469727	17.672610617644214
[18.1000003814697...]	17.799999237060547	18.258077980994376
[17.5499992370605...]	18.100000381469727	17.572919334078545
[17.6000003814697...]	17.549999237060547	17.521354224809713
[17.7999992370605...]	17.600000381469727	17.845516246141653
[17.1000003814697...]	17.799999237060547	17.16581983930224
[17.0499992370605...]	17.100000381469727	17.127399725805372

only showing top 20 rows

```
# Dự đoán kết quả đối với tập test
pred_results_test = regressor.evaluate(test_data)
pred_results_test.predictions.show()

# Tính độ đo MSE cho tập train và test
print("Mean squared error of train data",pred_results_train.meanSquaredError)
print("Mean squared error of test data",pred_results_test.meanSquaredError)

Mean squared error of train data 0.19694910868467175
Mean squared error of test data 0.129352449053747
```

▼ Câu 4: Phân loại đa lớp với pyspark

```

from pyspark.ml.linalg import Vectors as ml_vectors
import numpy as np
def f(x):
    k = x.split(',')
    return int(k[1]), ml_vectors.dense(np.array(k[2:]).astype(int))

from pyspark.ml.classification import LogisticRegression, OneVsRest
from pyspark.ml.evaluation import MulticlassClassificationEvaluator
# Data gốc
pet3_train = sc.textFile('/content/drive/MyDrive/New Version/oxford_pet3_train.csv').map(f).toDF(['label', 'features'])
pet3_test = sc.textFile('/content/drive/MyDrive/New Version/oxford_pet3_test.csv').map(f).toDF(['label', 'features'])

# Data đã chỉnh số chiều ở câu 1
df1 = sqlc.read.csv("/content/drive/MyDrive/datasets/oxford_pet3_80_train.csv")
df2 = sqlc.read.csv("/content/drive/MyDrive/datasets/oxford_pet3_80_test.csv")
# df.show()
pet3_r80_train = df1.rdd.map(lambda x: (int(x[0]),int(x[1]),list(map(float,(x[2][1:-1].split(','))))) ).toDF(['ID','label', 'features'])
pet3_r80_test = df2.rdd.map(lambda x: (int(x[0]),int(x[1]),list(map(float,(x[2][1:-1].split(','))))) ).toDF(['ID','label', 'features'])
pet3_r80_train.show()
pet3_r80_test.show()

```

```

+---+-----+-----+
| ID|label|          features|
+---+-----+-----+
| 0| 33|[11.2485358633465...|
| 1| 12|[20.5779916161160...|
| 2| 9|[12.2953497514603...|
| 3| 32|[104.467091904726...|
| 4| 33|[234.462074508373...|
| 5| 2|[186.288582218599...|
| 6| 22|[182.895838829834...|
| 7| 22|[135.102005692965...|
| 8| 36|[55.9788320216787...|
| 9| 5|[56.3647122849480...|
|10| 6|[175.741976022630...|
|11| 28|[235.354870458183...|
|12| 32|[167.915536528234...|
|13| 10|[30.3593886657564...|
|14| 14|[77.8572907709398...|
|15| 3|[32.3623421192620...|
|16| 22|[136.257473434922...|
|17| 12|[27.8430395051883...|
|18| 32|[76.7253793715973...|
|19| 5|[209.856118307185...|
+---+-----+-----+
only showing top 20 rows

```

```

+---+-----+-----+
| ID|label|          features|
+---+-----+-----+
| 0| 33|[216.122411152008...|
| 1| 12|[40.0464279154003...|
| 2| 9|[121.656231051751...|
| 3| 32|[256.331047238208...|
| 4| 33|[180.024117922330...|
| 5| 2|[129.378204754546...|
| 6| 22|[124.671397161241...|
| 7| 22|[204.695844832180...|
| 8| 36|[139.411350452042...|
| 9| 5|[103.049397555513...|
|10| 6|[168.904242864282...|
|11| 28|[37.5230484286361...|
|12| 32|[168.059124695685...|
|13| 10|[61.3143609212698...|
|14| 14|[65.0413942257536...|
|15| 3|[44.3527603911445...|
|16| 22|[220.559316273806...|
|17| 12|[185.023626223027...|
|18| 32|[25.0300022562822...|
|19| 5|[20.2636891212269...|
+---+-----+-----+
only showing top 20 rows

```



```
from pyspark.ml.classification import LogisticRegression, OneVsRest
from pyspark.ml.evaluation import MulticlassClassificationEvaluator

# Cho model học tập dữ liệu gốc
lr = LogisticRegression (maxIter=100, tol=1E-6, fitIntercept=True, labelCol='label', featuresCol='features')
ovr = OneVsRest(classifier=lr)
model = ovr.fit(pet3_train)

train_result = model.transform(pet3_train)
test_result = model.transform(pet3_test)

evaluator = MulticlassClassificationEvaluator(metricName="accuracy")

acc_train = evaluator.evaluate(train_result)
acc_test = evaluator.evaluate(test_result)

# Cho model học tập dữ liệu sau khi giảm số chiều
lr = LogisticRegression (maxIter=100, tol=1E-6, fitIntercept=True, labelCol='label', featuresCol='features')
ovr = OneVsRest(classifier=lr)
model = ovr.fit(pet3_r80_train)

train_result_r80 = model.transform(pet3_r80_train)
test_result_r80 = model.transform(pet3_r80_test)

evaluator = MulticlassClassificationEvaluator(metricName="accuracy")

acc_train_r80 = evaluator.evaluate(train_result_r80)
acc_test_r80 = evaluator.evaluate(test_result_r80)

import matplotlib.pyplot as plt
fig = plt.figure()
ax = fig.add_axes([0,0,1,1])
dataSet = ['Train_data_ori', 'Test_data_ori', 'Train_data_r80', 'Test_data_r80']
accuracy = [acc_train, acc_test, acc_train_r80, acc_test_r80]
ax.bar(dataSet,accuracy)
plt.show()
```