# final-project

May 8, 2023

# 1 CÂU 1: Phân cụm dữ liệu

#### 1.1 Installation

```
[]: from google.colab import drive
     drive.mount('/content/drive')
    Drive already mounted at /content/drive; to attempt to forcibly remount, call
    drive.mount("/content/drive", force_remount=True).
[]: !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.
      \hookrightarrow 1-bin-hadoop3.2.tqz
     !cp drive/MyDrive/MMDS-data/spark-3.1.1-bin-hadoop3.2.tgz .
     !tar xf spark-3.1.1-bin-hadoop3.2.tgz
     !pip install -q findspark
[]: 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()
[]: from pyspark import SparkContext
     sc = SparkContext(master="local", appName="final")
[]: from pyspark.sql import SQLContext
     from pyspark.sql.types import IntegerType, StructType, StringType, ArrayType,
      →StructField
     sqlc = SQLContext(sc)
[]: import matplotlib.pyplot as plt
     import numpy as np
     import pyspark.sql.functions as F
     from pyspark.sql.types import IntegerType
```

```
from pyspark.sql import SQLContext
from pyspark.sql.types import StructType, StringType, ArrayType,

StructField,DoubleType
```

```
[]: from pyspark.sql import SparkSession spark = SparkSession(sc)
```

```
[]: # Tạo RDD từ dữ liệu

rdd = spark.sparkContext.textFile("mnist_small_test.csv")

# Chuyển RDD thành DataFrame

df = rdd.map(lambda line: line.split(",")).toDF()

# Đặt tên cho các cột

df = df.withColumnRenamed("_1", "label")

df = df.selectExpr("cast(label as int) as label", *["cast(_{0}) as int) as_

→pixel_{0}".format(i) for i in range(2, 786)])
```

```
[]: # Hàm chuyển đổi vector ảnh thành ma trận

def reshape_image(row):
    pixels = row[1:]
    return (row[0], [pixels[i:i+28] for i in range(0, len(pixels), 28)])

# Áp dụng hàm chuyển đổi vào DataFrame

df = df.rdd.map(reshape_image).toDF(["label", "image"])
```

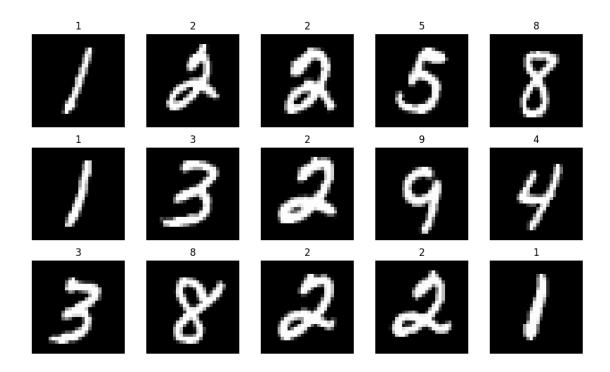
# 1.2 YC1\_1: Hiển thị

```
[]: # Lấy 15 bức ảnh đầu tiên
sample_df = df.limit(15).toPandas()

# Tạo lưới 3 x 5 và hiển thị ảnh
fig, axes = plt.subplots(nrows=3, ncols=5, figsize=(10, 6))

for i, ax in enumerate(axes.flatten()):
    ax.imshow(sample_df["image"][i], cmap="gray")
    ax.set_title(str(sample_df["label"][i]))
    ax.axis("off")

plt.tight_layout()
plt.show()
```



```
[]: from pyspark.ml.clustering import KMeans
from pyspark.ml.evaluation import ClusteringEvaluator
from pyspark.ml.feature import VectorAssembler
from pyspark.ml.clustering import KMeansModel
from pyspark.sql.functions import udf

[]: # Lựa chọn các cột pixel để tạo DataFrame chứa đữ liệu đầu vào cho KMeans
feature_cols = ["pixel_{{}}".format(i) for i in range(2,786)]

# Tạo VectorAssembler để tạo cột "features" từ các cột pixel
assembler = VectorAssembler(inputCols=feature_cols, outputCol="features")

# Tạo DataFrame chỉ chứa các cột pixel
input_data = df.select(*feature_cols)

# Biến đổi DataFrame để thêm cột "features"
input_data = assembler.transform(input_data)
```

## 1.3 YC1\_2: Phân cụm

```
[]: # Chọn giá trị k
k_values = [5, 10, 15]

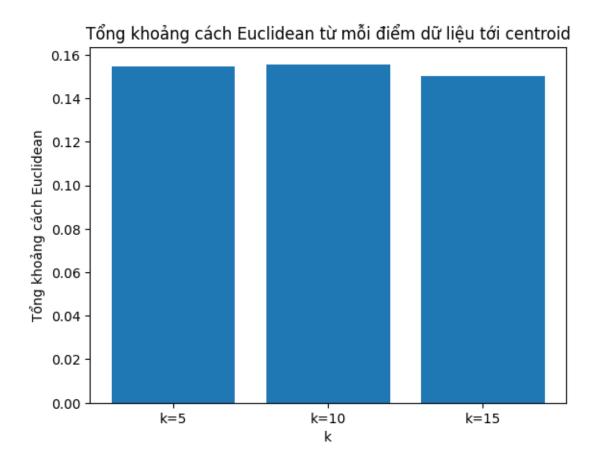
# Tạo mô hình và tính toán tổng khoảng cách Euclidean cho mỗi giá trị k
models = []
```

```
distances = []
for k in k_values:
    # Tao mô hình KMeans với qiá tri k
   kmeans = KMeans().setK(k).setSeed(42)
   # Huấn luyên mô hình
   model = kmeans.fit(input_data)
    # Phân cum dữ liêu
   predictions = model.transform(input_data)
    # Lưu mô hình xuống file
   model.write().overwrite().save("kmeans_model_k{}".format(k))
   # Load mô hình từ file
   loaded_model = KMeansModel.load("kmeans_model_k{}".format(k))
    # Đánh giá mô hình
   evaluator = ClusteringEvaluator()
   silhouette = evaluator.evaluate(predictions)
    # Tính tổng khoảng cách Euclidean từ mỗi điểm dữ liệu tới centroid tương ứng
   evaluator = ClusteringEvaluator()
   distance = evaluator.evaluate(loaded_model.transform(input_data))
   distances.append(distance)
   models.append(loaded_model)
```

## 1.4 YC1\_3: Trực quan hoá kết quả

```
[]: # Chia cột x thành các nhóm
x_ticks = np.arange(len(k_values))
group_labels = ['k={}'.format(k) for k in k_values]

# Vẽ biểu đồ cột
plt.bar(x_ticks, distances, tick_label=group_labels)
plt.xlabel('k')
plt.ylabel('Tổng khoảng cách Euclidean')
plt.title('Tổng khoảng cách Euclidean từ mỗi điểm dữ liệu tới centroid')
plt.show()
```



# 2 Câu 2: Giảm số chiều với SVD

```
[]: from pyspark.mllib.linalg.distributed import RowMatrix
import pandas as pd
import numpy as np

[]: def reduce_dimensions(file_path, output_file, k=196):
```

```
[]: def reduce_dimensions(file_path, output_file, k=196):
    # Doc dû liệu từ file CSV
    data = pd.read_csv(file_path, header=None)
    labels = data.iloc[:, 0]
    features = data.iloc[:, 1:]

# Chuyển đổi dũ liệu thành RowMatrix
    rows = sc.parallelize(features.values)
    mat = RowMatrix(rows)

# Tính toán SVD và giảm số chiều
    svd = mat.computeSVD(k, computeU=True)
    U = svd.U
```

```
s = svd.s
V = svd.V

U = np.array(U.rows.collect())
s = np.array(s)
V = np.array(V.toArray()).transpose()[:, :k]

reduced_matrix = U @ np.diag(s) @ V
# Luu kết quả vào file CSV
reduced_data = pd.concat([labels, pd.DataFrame(reduced_matrix)], axis=1)
reduced_data.to_csv(output_file, index=False, header=False)
sc.stop()
```

```
[]: #Câu 2_1: Giảm số chiều tập train
reduce_dimensions('/content/drive/MyDrive/BigData/Final/datasets/

⇔mnist_small_train.csv', '/content/drive/MyDrive/BigData/Final/datasets/

⇔mnist_small_train_svd.csv')
```

```
[]: #Câu 2_2: Giảm số chiều tập test
reduce_dimensions('/content/drive/MyDrive/BigData/Final/datasets/
⇔mnist_small_test.csv', '/content/drive/MyDrive/BigData/Final/datasets/
⇔mnist_small_test_svd.csv')
```

# 3 Câu 3: Khuyến nghị sản phẩm với Collaborative Filtering

```
root
|-- index: integer (nullable = true)
```

```
|-- user: integer (nullable = true)
     |-- item: integer (nullable = true)
     |-- rating: double (nullable = true)
    +----+
    |index|user|item|rating|
        32|
              1 | 167 |
                        3.51
      130
              1 | 422 |
                        3.5
    390
                        5.0
              1 | 352 |
    | 1188|
              1 | 168 |
                        5.0|
    | 1544|
                        5.01
              1 | 163 |
       50|
              2 | 413 |
                        3.5|
      139|
                        5.01
              2 | 183 |
      238
                        4.01
              2 | 434 |
    | 482|
              2 | 251 |
                        5.01
    674
              2| 288|
                        5.0|
    | 757|
                        4.51
              2 | 204 |
    | 1097|
              2 | 82 |
                        4.5|
    11081
              2 | 310 |
                        2.01
    | 1180|
                        5.0
              2 | 176 |
    | 1213|
                        3.5
              2|
                   0|
    | 1274|
              2 | 199 |
                        4.5|
    | 1399|
              2 | 320 |
                        2.0
    | 1485|
              2 | 271 |
                        4.0|
    | 1568|
              2 | 216 |
                        1.0|
    | 1622|
              2 | 294 |
                        4.5
    +----+
    only showing top 20 rows
[]: # Loc và tách dữ liêu thành tâp huấn luyên và tâp kiểm tra
     train_data = ratings
     test_data = ratings.filter((ratings.user >= 71) & (ratings.item >= 401))
      # Lấy dữ liệu của người dùng từ 71 đến hết và items từ 401 đến hết
[]: from pyspark.ml.recommendation import ALS
     als = ALS(rank = 5,
               maxIter = 5,
               userCol="user",
               itemCol="item",
               ratingCol="rating",
               numUserBlocks = 60,
               regParam = 0.01)
    model = als.fit(train_data)
```

```
[]: predictions = model.transform(test_data)
    evaluator = RegressionEvaluator(metricName="mse", labelCol="rating",
                                     predictionCol="prediction")
    mse = evaluator.evaluate(predictions)
    print("MSE = ",mse)
    MSE = 0.09646192568250929
[]: predictions.show()
    +----+
    |index|user|item|rating|prediction|
    +----+
    | 2208 | 72 | 451 |
                        4.0| 4.0065737|
    | 1898|
            72 | 436 |
                        4.0 | 3.8972464 |
       843 l
            721 4061
                        2.01 2.70492771
    l 1585 l
            72 | 412 |
                        3.0 | 2.7101414 |
    | 1208|
                        3.0| 3.0890405|
            73 | 412 |
                        3.01
      324
            72 | 460 |
                               3.00493
    | 1788|
            72 | 417 |
                        2.0 | 1.837924 |
       299 l
            73 | 417 |
                        3.0| 3.0867054|
            72| 444|
      157
                        3.0 | 3.223478 |
        41|
            72 | 435 |
                        4.0 | 3.8670328 |
       694
            72 | 440 |
                        3.0 | 4.2921796 |
    1 12281
            72 | 432 |
                        5.0 | 4.4767704 |
    | 2043|
            72| 452|
                        4.0| 4.050189|
    | 2132|
           72 | 425 |
                        3.01
                               3.10914
       200
            72 | 447 |
                        4.0 | 4.0065737 |
    | 1897| 72| 462|
                        4.0| 3.8262565|
    | 1035|
            72 | 437 |
                        5.0 | 4.826022|
    | 1701|
            72| 456|
                        5.0 | 4.3451123 |
                        2.0 | 2.1292784 |
    1 7971
            72 | 446 |
    | 1587 | 72 | 453 |
                        4.0 | 4.0065737 |
    +----+
    only showing top 20 rows
[]: predictions.printSchema()
    print("Pearson Correlation: ",end="")
    predictions.corr("rating", "prediction")
    root
     |-- index: integer (nullable = true)
     |-- user: integer (nullable = true)
     |-- item: integer (nullable = true)
     |-- rating: double (nullable = true)
     |-- prediction: float (nullable = false)
```

#### Pearson Correlation:

#### []: 0.9584556304650566

# 4 Câu 4: Dự đoán giá chứng khoán.

```
[]: data = sqlc.read.option("header", True) \
                     .option("inferSchema", True) \
                     .csv("/content/drive/MyDrive/BigData/data/stockHVN2022.csv")
[]: data.show()
           Ngay| HVN|
    |18/11/2022| 9.3|
    |17/11/2022| 9.42|
    |16/11/2022| 9.21|
    |15/11/2022| 8.61|
    |14/11/2022| 9.25|
    |11/11/2022| 9.9|
    |10/11/2022| 9.76|
    |09/11/2022| 10.4|
    |08/11/2022|10.25|
    |07/11/2022| 10.1|
    |04/11/2022| 10.7|
    |03/11/2022|10.85|
    |02/11/2022| 10.9|
    |01/11/2022| 10.9|
    |31/10/2022| 10.8|
    [28/10/2022]10.75]
    |27/10/2022|10.85|
    [26/10/2022]10.55]
    |25/10/2022|10.55|
    |24/10/2022|10.55|
    +----+
    only showing top 20 rows
[]: from pyspark.sql.functions import split
     rdd_stock = data.select(split(data.Ngay, '/').getItem(1).alias("month"), data.
      →HVN) \
                     .rdd \
                     .sortBy(lambda x: x[0]) \
                     .collect()
```

```
[]: def pivot_list(arr):
        rs = []
        for i in range(len(arr) - 5):
            tempL = arr[i:i+5]
            tempR = arr[i+5]
            rs.append((tempL, tempR))
        return rs
[]: temp_1 = [rdd_stock[i][1] for i in range(len(rdd_stock)) if rdd_stock[i][0] <=__
     temp_2 = [rdd_stock[i][1] for i in range(len(rdd_stock)) if rdd_stock[i][0] > [
     ن80'∍
    training = pivot list(temp 1)
    testing = pivot_list(temp_2)
[]: from pyspark.ml.linalg import Vectors, VectorUDT
    from pyspark.sql.functions import udf
    from pyspark.ml.feature import VectorAssembler
     # đinh nghĩa một UDF để chuyển đổi từ array float sang 1 dense vector
    to_dense_vector_udf = udf(lambda arr: Vectors.dense(arr), VectorUDT())
    df_training = sqlc.createDataFrame(training, ['features', 'label'])
    df_testing = sqlc.createDataFrame(testing, ['features', 'label'])
     # sử dung UDF cho côt features để chuyển đổ sang dense vector
    df_training = df_training.withColumn("features",__

    dense_vector_udf(df_training["features"]))

    df testing = df testing.withColumn("features", ...
      →to_dense_vector_udf(df_testing["features"]))
    assembler = VectorAssembler(inputCols=["features"], outputCol="features_vector")
    df_training = assembler.transform(df_training).select("features_vector",_

¬"label")
    df testing = assembler.transform(df testing).select("features vector", "label")
[]: df_training.show()
    df_training.printSchema()
    +----+
         features_vector|label|
    +----+
    |[24.3,24.0,24.1,2...| 22.7|
    |[24.0,24.1,23.35,...|22.75|
    |[24.1,23.35,22.8,...| 22.7|
    |[23.35,22.8,22.7,...|22.95|
    |[22.8,22.7,22.75,...| 22.9|
```

```
|[22.7,22.75,22.7,...| 22.7|
    | [22.75,22.7,22.95...|22.65|
    |[22.7,22.95,22.9,...|22.95|
    |[22.95,22.9,22.7,...|23.05|
    |[22.9,22.7,22.65,...| 23.2|
    |[22.7,22.65,22.95...|22.65|
    |[22.65,22.95,23.0...|22.85|
    |[22.95,23.05,23.2...| 23.1|
    |[23.05,23.2,22.65...| 23.3|
    |[23.2,22.65,22.85...|25.35|
    |[22.65,22.85,23.1...| 25.9|
    |[22.85,23.1,23.3,...|25.65|
    |[23.1,23.3,25.35,...|26.25|
    |[23.3,25.35,25.9,...| 26.1|
    |[25.35,25.9,25.65...| 26.8|
    +----+
    only showing top 20 rows
    root
     |-- features_vector: vector (nullable = true)
     |-- label: double (nullable = true)
[]: df testing.show()
     df_testing.printSchema()
          features_vector|label|
         ----+
    |[13.35,13.6,13.65...| 14.6|
    |[13.6,13.65,14.05...| 14.8|
    |[13.65,14.05,14.2...|14.65|
    |[14.05,14.25,14.6...|14.95|
    |[14.25,14.6,14.8,...|14.55|
    |[14.6,14.8,14.65,...|14.85|
    |[14.8,14.65,14.95...|15.15|
    |[14.65,14.95,14.5...|14.75|
    |[14.95,14.55,14.8...| 14.8|
    |[14.55,14.85,15.1...|14.65|
    |[14.85,15.15,14.7...|14.85|
    |[15.15,14.75,14.8...|15.15|
    |[14.75,14.8,14.65...|16.25|
    |[14.8,14.65,14.85...|16.85|
    |[14.65,14.85,15.1...| 17.0|
    |[14.85,15.15,16.2...| 10.8|
    |[15.15,16.25,16.8...|10.75|
```

|[16.25,16.85,17.0...|10.85| |[16.85,17.0,10.8,...|10.55|

```
|[17.0,10.8,10.75,...|10.55|
    +----+
    only showing top 20 rows
    root
     |-- features_vector: vector (nullable = true)
     |-- label: double (nullable = true)
[]: from pyspark.ml.regression import LinearRegression
    # Xây dựng mô hình Linear Regression
    lr = LinearRegression(featuresCol='features_vector', labelCol='label', maxIter_
      →= 10, regParam = 0.3, elasticNetParam = 0.8)
    lrModel = lr.fit(df training)
[]: # Lưu mô hình xuống tâp tin
    lrModel.write().overwrite().save('/content/drive/MyDrive/BigData/linreg_model')
[]: # Doc mô hình từ tập tin
    from pyspark.ml.regression import LinearRegressionModel
    loaded model = LinearRegressionModel.load("/content/drive/MyDrive/BigData/
      →linreg_model")
[]: # Dư đoán kết quả đối với tâp train
    pred results train = loaded model.evaluate(df training)
    pred_results_train.predictions.show()
    +----+
         features vector|label|
                                      prediction
       -----+
    |[24.3,24.0,24.1,2...| 22.7| 22.78503637901166|
    |[24.0,24.1,23.35,...|22.75|22.525248799412786|
    |[24.1,23.35,22.8,...| 22.7|22.479104805032346|
    |[23.35,22.8,22.7,...|22.95| 22.45424213888017|
    |[22.8,22.7,22.75,...| 22.9| 22.58730056292722|
    | [22.7,22.75,22.7,...| 22.7| 22.61631111282257|
    |[22.75,22.7,22.95...|22.65| 22.51382650308939|
    |[22.7,22.95,22.9,...|22.95| 22.43177614449911|
    |[22.95,22.9,22.7,...|23.05|22.570447649708235|
    |[22.9,22.7,22.65,...| 23.2|22.696329728174693|
    | [22.7,22.65,22.95...|22.65|22.832618231345464|
    |[22.65,22.95,23.0...|22.85|22.568715590755996|
    |[22.95,23.05,23.2...| 23.1|22.559221487384093|
    [23.05,23.2,22.65...] 23.3 22.69982666883175
    | [23.2,22.65,22.85...|25.35|22.892286992821546|
    |[22.65,22.85,23.1...| 25.9|24.119474302226482|
```

[]: # Dự đoán kết quả đối với tập test

pred\_results\_test = loaded\_model.evaluate(df\_testing)

pred\_results\_test.predictions.show()

```
-----+
     features vector|label|
                                  prediction|
   ------
|[13.35,13.6,13.65...| 14.6|14.728308463118806|
|[13.6,13.65,14.05...| 14.8|15.011011362898554|
|[13.65,14.05,14.2...|14.65|15.228151888774045|
|[14.05,14.25,14.6...|14.95|15.224570550850132|
| [14.25,14.6,14.8,...|14.55|15.374664352382649|
|[14.6,14.8,14.65,...|14.85|15.209749380869303|
|[14.8,14.65,14.95...|15.15|15.307168302239857|
| [14.65,14.95,14.5... | 14.75 | 15.514169264944371 |
|[14.95,14.55,14.8...| 14.8|15.389869603916251|
|[14.55,14.85,15.1...|14.65| 15.34640281257304|
|[14.85,15.15,14.7...|14.85| 15.23810898767854|
| [15.15,14.75,14.8...| 15.15| 15.318310067297144|
|[14.75,14.8,14.65...|16.25|15.523194889496644|
|[14.8,14.65,14.85...|16.85|16.235183880198836|
|[14.65,14.85,15.1...| 17.0|16.871868685111334|
|[14.85,15.15,16.2...| 10.8| 17.20376319412876|
| [15.15,16.25,16.8...| 10.75| 13.800971568969587|
| [16.25,16.85,17.0...| 10.85| 12.256160346342142|
|[16.85,17.0,10.8,...|10.55|11.740585808243909|
|[17.0,10.8,10.75,...|10.55|11.591690342596952|
+----+
only showing top 20 rows
```

```
[]: # Tinh độ đo MSE cho tập train và test

train_mse = pred_results_train.meanSquaredError

test_mse = pred_results_test.meanSquaredError

print("Mean squared error of train data", train_mse)

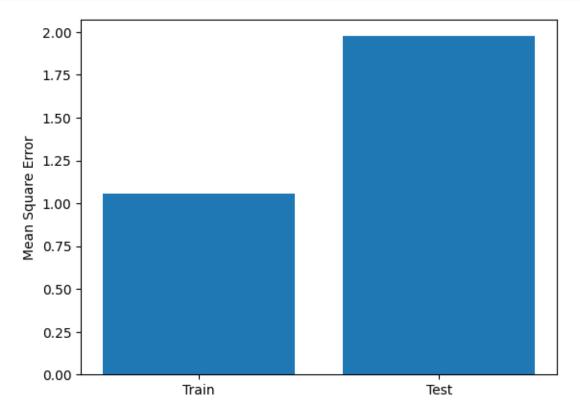
print("Mean squared error of test data", test_mse)
```

Mean squared error of train data 1.0539797517581313 Mean squared error of test data 1.9755975428083403

```
[]: import matplotlib.pyplot as plt

# Vẽ biểu đồ cột thể hiện giá trị Mean Square Error trên tập huấn luyện và tậpunkiểm tra
x = ["Train", "Test"]
y = [train_mse, test_mse]

plt.bar(x, y)
plt.ylabel("Mean Square Error")
plt.show()
```



# 5 Câu 5: Phân loại đa lớp với pyspark

# 5.1 Sử dụng tập train, test ban đầu

```
[]: from pyspark.ml.linalg import Vectors as ml_vectors
import numpy as np
def f(x):
    k = x.split(',')
    return int(k[0]), ml_vectors.dense(np.array(k[1:]).astype(int))
```

```
[]: train data = sc.textFile("/content/drive/MyDrive/BigData/Final/datasets/
     →mnist_small_train.csv").map(f).toDF(['label', 'features'])
    test data = sc.textFile("/content/drive/MyDrive/BigData/Final/datasets/
      →mnist small test.csv").map(f).toDF(['label', 'features'])
[]: train_set, val_data = train_data.randomSplit([0.8, 0.2], seed=42)
    5.1.1 MLP
[]: from pyspark.ml.classification import MultilayerPerceptronClassifier
    from pyspark.ml.evaluation import MulticlassClassificationEvaluator
    layers = [784, 256, 128, 10]
    mlp = MultilayerPerceptronClassifier(maxIter=100, layers=layers, blockSize=128, __
     ⇒seed=42)
[ ]: model_mlp = mlp.fit(train_set)
[]: # tính accuracy trên val_data
    result = model_mlp.transform(val_data)
    predictionAndLabels = result.select("prediction", "label")
    evaluator val mlp = MulticlassClassificationEvaluator(metricName="accuracy")
    accuracy_val_mlp = evaluator_val_mlp.evaluate(predictionAndLabels)
    print("MLP Validation set accuracy = " + str(accuracy_val_mlp))
    MLP Validation set accuracy = 0.9214659685863874
[]: # Tinh accuracy trên test_data
    result = model_mlp.transform(test_data)
    predictionAndLabels = result.select("prediction", "label")
    evaluator_test_mlp = MulticlassClassificationEvaluator(metricName="accuracy")
    accuracy_test_mlp = evaluator_test_mlp.evaluate(predictionAndLabels)
    print("MLP Test set accuracy = " + str(accuracy_test_mlp))
    5.1.2 Random Forest
[]: from pyspark.ml.classification import RandomForestClassifier
    rf = RandomForestClassifier(labelCol='label', featuresCol='features',
      []: model_rf = rf.fit(train_set)
[]: # tính accuracy trên val_data
```

result = model\_rf.transform(val\_data)

```
predictionAndLabels = result.select("prediction", "label")
evaluator_val_rf = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_rf = evaluator_val_rf.evaluate(predictionAndLabels)
print("Random Forest Validation set accuracy = " + str(accuracy_val_rf))
```

Random Forest Validation set accuracy = 0.9334330590875094

```
[]: # Tinh accuracy trên test_data
    result = model_rf.transform(test_data)
    predictionAndLabels = result.select("prediction", "label")
    evaluator_test_rf = MulticlassClassificationEvaluator(metricName="accuracy")
    accuracy_test_rf = evaluator_test_rf.evaluate(predictionAndLabels)
    print("Random Forest Test set accuracy = " + str(accuracy_test_rf))
```

Random Forest Test set accuracy = 0.95

#### 5.1.3 Linear Support Vector Machine

#### LSVC phân biệt số 0

```
from pyspark.sql.functions import when, col
# label khác 0 thì thay = 1, 0 thì giữ nguyên 0 (otherwise(0))
test_data_0 = test_data.withColumn("label", when(col("label") != 0, 1).
otherwise(0))
train_set_0 = train_set.withColumn("label", when(col("label") != 0, 1).
otherwise(0))
val_data_0 = val_data.withColumn("label", when(col("label") != 0, 1).
otherwise(0))
```

```
[]: from pyspark.ml.classification import LinearSVC

# Load training data
lsvc = LinearSVC(maxIter=100, regParam=0.1)

# Fit the model
model_lsvc_0 = lsvc.fit(train_set_0)
```

Linear Support Vector Machine Validation set accuracy prediction 0 = 0.9895287958115183

```
[]: # Tinh accuracy trên test_data
result = model_lsvc_0.transform(test_data_0)
predictionAndLabels = result.select("prediction", "label")
evaluator_test_lsvc_0 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_test_lsvc_0 = evaluator_test_lsvc_0.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Test set accuracy prediction 0 = " +___

str(accuracy_test_lsvc_0))
```

Linear Support Vector Machine Test set accuracy prediction 0 = 0.992

### LSVC phân biệt số 1

```
[]: # Load training data
lsvc = LinearSVC(maxIter=100, regParam=0.1)

# Fit the model
model_lsvc_1 = lsvc.fit(train_set_1)
```

Linear Support Vector Machine Validation set accuracy prediction 1 = 0.9902767389678384

Linear Support Vector Machine Test set accuracy prediction 1 = 0.993

```
[]: test_data_2 = test_data.withColumn("label", when(col("label") != 2, 0).
      ⇔otherwise(1))
     train_set_2 = train_set.withColumn("label", when(col("label") != 2, 0).
      →otherwise(1))
     val_data_2 = val_data.withColumn("label", when(col("label") != 2, 0).

otherwise(1))
[]: # Load training data
     lsvc = LinearSVC(maxIter=100, regParam=0.1)
     # Fit the model
     model_lsvc_2 = lsvc.fit(train_set_2)
[]: # tính accuracy trên val_data
     result = model_lsvc_2.transform(val_data_2)
     predictionAndLabels = result.select("prediction", "label")
     evaluator val lsvc 2 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val_lsvc_2 = evaluator_val_lsvc_2.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Validation set accuracy prediction 2 = " + "
      ⇔str(accuracy_val_lsvc_2))
    Linear Support Vector Machine Validation set accuracy prediction 2 =
    0.9730740463724757
[]: # Tính accuracy trên test_data
     result = model lsvc 2.transform(test data 2)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_test_lsvc_2 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_test_lsvc_2 = evaluator_test_lsvc_2.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Test set accuracy prediction 2 = " +_{\sqcup}

¬str(accuracy_test_lsvc_2))
    Linear Support Vector Machine Test set accuracy prediction 2 =
    0.9726666666666667
```

```
[]: model_lsvc_3 = lsvc.fit(train_set_3)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_3.transform(val_data_3)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_3 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_3 = evaluator_val_lsvc_3.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation set accuracy prediction 3 = " +__

str(accuracy_val_lsvc_3))
```

Linear Support Vector Machine Validation set accuracy prediction 3 = 0.9805534779356769

```
[]: # Tinh accuracy trên test_data
result = model_lsvc_3.transform(test_data_3)
predictionAndLabels = result.select("prediction", "label")
evaluator_test_lsvc_3 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_test_lsvc_3 = evaluator_test_lsvc_3.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Test set accuracy prediction 3 = " +__

str(accuracy_test_lsvc_3))
```

Linear Support Vector Machine Test set accuracy prediction 3 = 0.9783333333333334

### LSVC phân biệt số 4

```
[]: # Load training data
lsvc = LinearSVC(maxIter=100, regParam=0.1)

# Fit the model
model_lsvc_4 = lsvc.fit(train_set_4)
```

Linear Support Vector Machine Validation set accuracy prediction 4 = 0.975317875841436

```
[]: # Tinh accuracy trên test_data
result = model_lsvc_4.transform(test_data_4)
predictionAndLabels = result.select("prediction", "label")
evaluator_test_lsvc_4 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_test_lsvc_4 = evaluator_test_lsvc_4.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Test set accuracy prediction 4 = " +___

str(accuracy_test_lsvc_4))
```

Linear Support Vector Machine Test set accuracy prediction 4 = 0.978666666666667

### LSVC phân biệt số 5

```
[]: model_lsvc_5 = lsvc.fit(train_set_5)
```

Linear Support Vector Machine Validation set accuracy prediction 5 = 0.9738219895287958

Linear Support Vector Machine Test set accuracy prediction 5 = 0.969666666666667

```
[]: test_data_6 = test_data.withColumn("label", when(col("label") != 6, 0).

otherwise(1))
```

```
train_set_6 = train_set.withColumn("label", when(col("label") != 6, 0).
      ⇔otherwise(1))
     val_data_6 = val_data.withColumn("label", when(col("label") != 6, 0).

otherwise(1))
[]: model_lsvc_6 = lsvc.fit(train_set_6)
[]: # tính accuracy trên val data
     result = model_lsvc_6.transform(val_data_6)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_val_lsvc_6 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val_lsvc_6 = evaluator_val_lsvc_6.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Validation set accuracy prediction 6 = " +_{\sqcup}
      ⇔str(accuracy_val_lsvc_6))
    Linear Support Vector Machine Validation set accuracy prediction 6 =
    0.9730740463724757
[]: # Tinh accuracy trên test data
     result = model lsvc 6.transform(test data 6)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_test_lsvc_6 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_test_lsvc_6 = evaluator_test_lsvc_6.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Test set accuracy prediction 6 = " +_{\sqcup}
      ⇔str(accuracy_test_lsvc_6))
    Linear Support Vector Machine Test set accuracy prediction 6 =
    0.98866666666666
    LSVC phân biệt số 7
[]: test_data_7 = test_data.withColumn("label", when(col("label") != 7, 0).
      ⇔otherwise(1))
     train_set_7 = train_set.withColumn("label", when(col("label") != 7, 0).
      ⇔otherwise(1))
     val_data_7 = val_data.withColumn("label", when(col("label") != 7, 0).

otherwise(1))
[]: model_lsvc_7 = lsvc.fit(train_set_7)
[]: # tính accuracy trên val_data
     result = model_lsvc_7.transform(val_data_7)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_val_lsvc_7 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val_lsvc_7 = evaluator_val_lsvc_7.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Validation set accuracy prediction 7 = " + ⊔
```

⇔str(accuracy\_val\_lsvc\_7))

Linear Support Vector Machine Validation set accuracy prediction 7 = 0.9745699326851159

Linear Support Vector Machine Test set accuracy prediction 7 = 0.9923333333333333

### LSVC phân biệt số 8

```
[]: model_lsvc_8 = lsvc.fit(train_set_8)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_8.transform(val_data_8)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_8 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_8 = evaluator_val_lsvc_8.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation set accuracy prediction 8 = " +___

str(accuracy_val_lsvc_8))
```

Linear Support Vector Machine Validation set accuracy prediction 8 = 0.9648466716529543

Linear Support Vector Machine Test set accuracy prediction 8 = 0.968

```
[]: model_lsvc_9 = lsvc.fit(train_set_9)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_9.transform(val_data_9)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_9 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_9 = evaluator_val_lsvc_9.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation set accuracy prediction 9 = " +__

str(accuracy_val_lsvc_9))
```

Linear Support Vector Machine Validation set accuracy prediction 9 = 0.9573672400897532

Linear Support Vector Machine Test set accuracy prediction 9 = 0.9626666666666667

Linear Support Vector Machine Validation set accuracy = 0.9752430815258041 Linear Support Vector Machine Test set accuracy = 0.979600000000001

## 5.2 Sử dụng tập training\_svd, test\_svd từ câu b)

```
\lceil \ \rceil : \ def \ f(x) :
      k = x.split(',')
       return int(k[0]), ml_vectors.dense(np.array(k[1:]))
[]: train_data_svd = sc.textFile("/content/drive/MyDrive/BigData/Final/datasets/

¬mnist_small_train_svd.csv").map(f).toDF(['label', 'features'])

     test_data_svd = sc.textFile("/content/drive/MyDrive/BigData/Final/datasets/
      omnist_small_test_svd.csv").map(f).toDF(['label', 'features'])
[]: train_set_svd, val_data_svd = train_data_svd.randomSplit([0.8, 0.2], seed=42)
    5.2.1 MLP
[]: from pyspark.ml.classification import MultilayerPerceptronClassifier
     from pyspark.ml.evaluation import MulticlassClassificationEvaluator
     layers = [196, 256, 128, 10]
     mlp_svd = MultilayerPerceptronClassifier(maxIter=100, layers=layers,__
      ⇔blockSize=128, seed=42)
[]: model_mlp_svd = mlp_svd.fit(train_set_svd)
[]: # tính accuracy trên val_data
     result = model mlp svd.transform(val data svd)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_val_mlp_svd = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val_mlp_svd = evaluator_val_mlp_svd.evaluate(predictionAndLabels)
     print("MLP Validation SVD set accuracy = " + str(accuracy_val_mlp_svd))
    MLP Validation SVD set accuracy = 0.6013462976813763
[]: # Tinh accuracy trên test data
     result = model_mlp_svd.transform(test_data_svd)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_test_mlp_svd =_
      →MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_test_mlp_svd = evaluator_test_mlp_svd.evaluate(predictionAndLabels)
     print("MLP Test SVD set accuracy = " + str(accuracy_test_mlp_svd))
```

MLP Test SVD set accuracy = 0.5556666666666666

#### 5.2.2 Random Forest

# Load training data

# Fit the model

lsvc = LinearSVC(maxIter=100, regParam=0.1)

model\_lsvc\_0 = lsvc.fit(train\_set\_0)

```
[]: from pyspark.ml.classification import RandomForestClassifier
     rf_svd = RandomForestClassifier(labelCol='label', featuresCol='features', __
      →numTrees=50, seed=1234, maxDepth=10)
[]: model_rf_svd = rf_svd.fit(train_set_svd)
[]: # tính accuracy trên val_data
     result = model_rf_svd.transform(val_data_svd)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_val_rf_svd = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val rf_svd = evaluator_val_rf_svd.evaluate(predictionAndLabels)
     print("Random Forest Validation SVD set accuracy = " + str(accuracy_val_rf_svd))
    Random Forest Validation SVD set accuracy = 0.6192969334330591
[]: # Tính accuracy trên test data
     result = model_rf_svd.transform(test_data_svd)
     predictionAndLabels = result.select("prediction", "label")
     evaluator test rf svd = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy test rf svd = evaluator test rf svd.evaluate(predictionAndLabels)
     print("Random Forest Test SVD set accuracy = " + str(accuracy_test_rf_svd))
    Random Forest Test SVD set accuracy = 0.57133333333333334
    5.2.3 Linear Support Vector Machine
    LSVC phân biệt số 0
[]: from pyspark.sql.functions import when, col
     # label khác 0 thì thay = 1, 0 thì giữ nguyên 0 (otherwise(0))
     test_data_0 = test_data_svd.withColumn("label", when(col("label") != 0, 1).

otherwise(0))
     train_set_0 = train_set_svd.withColumn("label", when(col("label") != 0, 1).
      ⇔otherwise(0))
     val_data_0 = val_data_svd.withColumn("label", when(col("label") != 0, 1).

otherwise(0))
[]: from pyspark.ml.classification import LinearSVC
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_0.transform(val_data_0)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_0 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_0 = evaluator_val_lsvc_0.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation SVD set accuracy prediction 0 =

" + str(accuracy_val_lsvc_0))
```

Linear Support Vector Machine Validation SVD set accuracy prediction 0 = 0.9184741959611069

Linear Support Vector Machine Test SVD set accuracy prediction 0 = 0.4813333333333334

### LSVC phân biệt số 1

```
[]: # Load training data
lsvc = LinearSVC(maxIter=100, regParam=0.1)

# Fit the model
model_lsvc_1 = lsvc.fit(train_set_1)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_1.transform(val_data_1)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_1 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_1 = evaluator_val_lsvc_1.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation SVD set accuracy prediction 1 =

" + str(accuracy_val_lsvc_1))
```

Linear Support Vector Machine Validation SVD set accuracy prediction 1 = 0.8915482423335827

Linear Support Vector Machine Test SVD set accuracy prediction 1 = 0.5003333333333333

### LSVC phân biệt số 2

```
[]: # Load training data
lsvc = LinearSVC(maxIter=100, regParam=0.1)

# Fit the model
model_lsvc_2 = lsvc.fit(train_set_2)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_2.transform(val_data_2)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_2 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_2 = evaluator_val_lsvc_2.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation set accuracy prediction 2 = " +__

str(accuracy_val_lsvc_2))
```

Linear Support Vector Machine Validation set accuracy prediction 2 = 0.9184741959611069

Linear Support Vector Machine Test set accuracy prediction 2 = 0.5066666666666667

```
LSVC phân biệt số 3
```

```
[ ]: model_lsvc_3 = lsvc.fit(train_set_3)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_3.transform(val_data_3)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_3 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_3 = evaluator_val_lsvc_3.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation set accuracy prediction 3 = " +___

str(accuracy_val_lsvc_3))
```

Linear Support Vector Machine Validation set accuracy prediction 3 = 0.8945400149588631

Linear Support Vector Machine Test set accuracy prediction 3 = 0.5153333333333333

```
[]: # Load training data
lsvc = LinearSVC(maxIter=100, regParam=0.1)

# Fit the model
model_lsvc_4 = lsvc.fit(train_set_4)
```

Linear Support Vector Machine Validation set accuracy prediction 4 = 0.8967838444278234

```
[]: # Tinh accuracy trên test_data
result = model_lsvc_4.transform(test_data_4)
predictionAndLabels = result.select("prediction", "label")
evaluator_test_lsvc_4 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_test_lsvc_4 = evaluator_test_lsvc_4.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Test set accuracy prediction 4 = " +__

str(accuracy_test_lsvc_4))
```

Linear Support Vector Machine Test set accuracy prediction 4 = 0.501666666666667

### LSVC phân biệt số 5

```
[ ]: model_lsvc_5 = lsvc.fit(train_set_5)
```

```
[]: # tinh accuracy trên val_data
result = model_lsvc_5.transform(val_data_5)
predictionAndLabels = result.select("prediction", "label")
evaluator_val_lsvc_5 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_val_lsvc_5 = evaluator_val_lsvc_5.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Validation set accuracy prediction 5 = " +___

str(accuracy_val_lsvc_5))
```

Linear Support Vector Machine Validation set accuracy prediction 5 = 0.9229618548990277

```
[]: # Tinh accuracy trên test_data
result = model_lsvc_5.transform(test_data_5)
predictionAndLabels = result.select("prediction", "label")
evaluator_test_lsvc_5 = MulticlassClassificationEvaluator(metricName="accuracy")
```

```
accuracy_test_lsvc_5 = evaluator_test_lsvc_5.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Test set accuracy prediction 5 = " +__

str(accuracy_test_lsvc_5))
```

Linear Support Vector Machine Test set accuracy prediction 5 = 0.48733333333333334

#### LSVC phân biệt số 6

```
[]: model_lsvc_6 = lsvc.fit(train_set_6)
```

Linear Support Vector Machine Validation set accuracy prediction 6 = 0.9506357516828721

```
[]: # Tinh accuracy trên test_data
result = model_lsvc_6.transform(test_data_6)
predictionAndLabels = result.select("prediction", "label")
evaluator_test_lsvc_6 = MulticlassClassificationEvaluator(metricName="accuracy")
accuracy_test_lsvc_6 = evaluator_test_lsvc_6.evaluate(predictionAndLabels)
print("Linear Support Vector Machine Test set accuracy prediction 6 = " +___

str(accuracy_test_lsvc_6))
```

Linear Support Vector Machine Test set accuracy prediction 6 = 0.53

```
[ ]: model_lsvc_7 = lsvc.fit(train_set_7)
[]: # tính accuracy trên val data
     result = model_lsvc_7.transform(val_data_7)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_val_lsvc_7 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val_lsvc_7 = evaluator_val_lsvc_7.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Validation set accuracy prediction 7 = " +_{\sqcup}
      ⇔str(accuracy_val_lsvc_7))
    Linear Support Vector Machine Validation set accuracy prediction 7 =
    0.893792071802543
[]: # Tinh accuracy trên test data
     result = model_lsvc_7.transform(test_data_7)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_test_lsvc_7 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_test_lsvc_7 = evaluator_test_lsvc_7.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Test set accuracy prediction 7 = " + L
      str(accuracy_test_lsvc_7))
    Linear Support Vector Machine Test set accuracy prediction 7 =
    0.4696666666666667
    LSVC phân biệt số 8
[]: test_data_8 = test_data_svd.withColumn("label", when(col("label") != 8, 0).
     train_set_8 = train_set_svd.withColumn("label", when(col("label") != 8, 0).
      ⇔otherwise(1))
     val_data_8 = val_data_svd.withColumn("label", when(col("label") != 8, 0).
      ⇔otherwise(1))
[]: model_lsvc_8 = lsvc.fit(train_set_8)
[]: # tính accuracy trên val data
     result = model_lsvc_8.transform(val_data_8)
     predictionAndLabels = result.select("prediction", "label")
     evaluator_val_lsvc_8 = MulticlassClassificationEvaluator(metricName="accuracy")
     accuracy_val_lsvc_8 = evaluator_val_lsvc_8.evaluate(predictionAndLabels)
     print("Linear Support Vector Machine Validation set accuracy prediction 8 = " + \sqcup +
      str(accuracy_val_lsvc_8))
    Linear Support Vector Machine Validation set accuracy prediction 8 =
    0.9072550486163051
[]: # Tinh accuracy trên test_data
     result = model_lsvc_8.transform(test_data_8)
```

Linear Support Vector Machine Test set accuracy prediction 8 = 0.478

#### LSVC phân biệt số 9

```
[]: model_lsvc_9 = lsvc.fit(train_set_9)
```

Linear Support Vector Machine Validation set accuracy prediction 9 = 0.8975317875841436

Linear Support Vector Machine Test set accuracy prediction 9 = 0.526

```
[]: # trung binh công tất cả LSVC
accuracy_val_lsvc_arr = []
accuracy_test_lsvc_arr = []

for i in range(10):
   var_name_val = eval(f"accuracy_val_lsvc_{i}")
   var_name_test = eval(f"accuracy_test_lsvc_{i}")
   accuracy_val_lsvc_arr.append(var_name_val)
```

### 5.3 Thể hiện độ chính xác (visualize)

```
import matplotlib.pyplot as plt

val_acc = [accuracy_val_mlp, accuracy_val_rf, accuracy_val_lsvc]
test_acc = [accuracy_test_mlp, accuracy_test_rf, accuracy_test_lsvc]
val_acc_svd = [accuracy_val_mlp_svd, accuracy_val_rf_svd, accuracy_val_lsvc_svd]
test_acc_svd = [accuracy_test_mlp_svd, accuracy_test_rf_svd,_
accuracy_test_lsvc_svd]

# Dặt tên cho các model
models = ['MLP', 'Random Forest', 'LSVM']

plt.plot(models, val_acc, label='Validation Accuracy')
plt.plot(models, test_acc, label='Validation Accuracy SVD')
plt.plot(models, test_acc_svd, label='Validation Accuracy SVD')
plt.plot(models, test_acc_svd, label='Test Accuracy SVD')
plt.legend()
plt.show()
```

