Assignment 1

Aim: Write a program using Multivariate Analysis methods on selected Big Data.

Objective: To apply multivariate analysis techniques on a large dataset using PySpark.

Theory:

To analyze the relationships between multiple variables efficiently in a distributed environment.

Multivariate Analysis is a set of statistical techniques used to analyze data that involves multiple variables. It helps in understanding relationships, patterns, and dependencies among multiple features simultaneously. This analysis is widely used in fields like finance, marketing, and machine learning to extract insights from complex datasets.

Principal Component Analysis (PCA):

Principal Component Analysis (PCA) is a dimensionality reduction technique used to transform highdimensional data into a lower-dimensional space while preserving as much variance as possible. It is widely used in machine learning, finance, image processing, and big data analytics for feature extraction, noise reduction, and visualization.

How PCA Works?

1. Standardization of Data

o Normalize the data to have a mean of zero and unit variance.

2. Compute the Covariance Matrix

o Understand relationships between different features.

3. Eigen Decomposition

Find eigenvalues and eigenvectors of the covariance matrix.

4. Select Principal Components

o Choose top k eigenvectors (principal components) based on eigenvalues.

5. Transform the Data

Project data onto the new feature space with reduced dimensions.

Dataset Used:

Flights Dataset from Kaggle

The dataset consists of flight records, including departure delays and arrival delays. It contains the following key attributes:

- **DayofMonth:** The day of the month the flight was scheduled.
- DayOfWeek: The day of the week the flight was scheduled.
- Carrier: The airline carrier code.
- OriginAirportID: The airport ID of the origin.

- **DestAirportID:** The airport ID of the destination.
- **DepDelay:** Departure delay in minutes.
- ArrDelay: Arrival delay in minutes.

Model Selection & Hyperparameters:

• PCA (n=2): Reduced data to 2 dimensions for visualization and analysis.

Implementation:

- 1. Initialize SparkSession Creates a Spark environment for distributed computing.
- 2. Load Dataset Reads flight data from a CSV file.
- 3. **Feature Selection** Selects relevant features for PCA.
- 4. **Standardization** Scales feature values to ensure uniform importance.
- 5. **Apply PCA** Reduces the dimensionality of data to 2 principal components.
- 6. **Display Results** Shows transformed data with principal components.

Interpretation of Results:

- PCA Output: The 2D scatter plot of PC1 vs PC2 displayed diverse flight delay behaviors.
- **Distinct Delay Patterns**: Flights with significant delays formed distinct groups, highlighting differences in departure and arrival delays.
- **Feature Significance**: The variance in PCA components indicated the importance of departure and arrival delay features.

This modified assignment uses the flights.csv dataset instead of financial transactions while maintaining the original structure and methodology.

multivariate-analysis-1

March 18, 2025

[1]: import kagglehub

```
# Download latest version
    path = kagglehub.dataset_download("tylerx/flights-and-airports-data")
    print("Path to dataset files:", path)
    C:\Users\Harshal\OneDrive\Desktop\py_spark project\myenv\Lib\site-
    packages\tqdm\auto.py:21: TqdmWarning: IProgress not found. Please update
    jupyter and ipywidgets. See
    https://ipywidgets.readthedocs.io/en/stable/user_install.html
      from .autonotebook import tqdm as notebook_tqdm
    Warning: Looks like you're using an outdated `kagglehub` version (installed:
    0.3.7), please consider upgrading to the latest version (0.3.10).
    Path to dataset files:
    C:\Users\Harshal\.cache\kagglehub\datasets\tylerx\flights-and-airports-
    data\versions\1
[2]: import numpy as np
    import pandas as pd
[3]: import pyspark
    from pyspark.sql import SparkSession
    from pyspark.sql.functions import col, sum, when
    from pyspark.ml.feature import StandardScaler, StringIndexer, VectorAssembler, u
      →VectorIndexer, OneHotEncoder
    from pyspark.ml.evaluation import MulticlassClassificationEvaluator
    from pyspark.mllib.evaluation import MulticlassMetrics
    from pyspark.ml import Pipeline
    from pyspark.ml.classification import LogisticRegression, __
      →DecisionTreeClassifier, RandomForestClassifier, GBTClassifier
[4]: spark = SparkSession.builder.appName("flight").master("local[1]").config("spark.
      ⇒sql.shuffle.partitions", "1").config("spark.driver.memory", "4g").
```

```
[5]: data = spark.read.csv('C:/Users/Harshal/.cache/kagglehub/datasets/tylerx/
      oflights-and-airports-data/versions/1/flights.csv', inferSchema=True, □
      →header=True)
     data.show(10)
    |DayofMonth|DayOfWeek|Carrier|OriginAirportID|DestAirportID|DepDelay|ArrDelay|
              191
                         5 l
                                DLl
                                              11433|
                                                             13303
                                                                          -3 l
                                                                                    11
             191
                         5 l
                                DLl
                                              14869
                                                             12478
                                                                          0|
                                                                                   -81
             19 l
                                DLl
                         5 l
                                              14057
                                                             14869
                                                                         -4 l
                                                                                  -15|
             19|
                         5|
                                DL|
                                              15016
                                                             11433|
                                                                          28|
                                                                                   24
             19 l
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                                                             12892 l
                                                                         -6 l
                                              11193
                                                                                  -11|
             19|
                         5|
                                DL
                                              10397
                                                             15016
                                                                         -1|
                                                                                  -19|
             19 l
                         5 I
                                DLI
                                                             10397
                                                                          0|
                                                                                   -1|
                                              15016
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                         51
                                DL
                                              10397
                                                             14869|
                                                                         15|
                                                                                   24|
             19 l
                         5 l
                                DL
                                              10397|
                                                             10423|
                                                                         33|
                                                                                   34|
                                                                        323|
              19|
                         51
                                DL
                                                             10397|
                                                                                  322|
                                              11278
    only showing top 10 rows
[6]: data.printSchema()
    root
     |-- DayofMonth: integer (nullable = true)
     |-- DayOfWeek: integer (nullable = true)
     |-- Carrier: string (nullable = true)
     |-- OriginAirportID: integer (nullable = true)
     |-- DestAirportID: integer (nullable = true)
     |-- DepDelay: integer (nullable = true)
     |-- ArrDelay: integer (nullable = true)
[7]:
     data.columns
[7]: ['DayofMonth',
      'DayOfWeek',
      'Carrier',
      'OriginAirportID',
      'DestAirportID',
      'DepDelay',
      'ArrDelay']
[8]: data = data.select("DayofMonth", "DayOfWeek", "Carrier",
      →"OriginAirportID", "DestAirportID", "DepDelay", "ArrDelay", ((col("ArrDelay")_
      →> 15).cast("Int").alias("Late")))
```

[9]: data.show(10)

|DayofMonth|DayOfWeek|Carrier|OriginAirportID|DestAirportID|DepDelay|ArrDelay|La 5 l 19| DL 13303| -3| 1| 11433| 01 12478| 0| -8| 19| 5 l DLl 14869| 01 191 5| DL| 14057| 14869| -4| -15 l 01 191 5| DLl 24 15016 11433| 28| 11 19| 5| DL| -6| -11| 1 11193| 12892 01 19| 5| DL 10397| -1| -19| 15016| 01 19| 5| DL| 15016| 10397| 0| -1|

+-----

10397|

10397|

11278

14869|

10423|

10397|

15|

33|

323|

24|

341

322

only showing top 10 rows

19|

19|

19|

5|

5|

5 l

DL|

DL|

DL|

01

1|

1 | |

11

```
[10]: null_counts = {
        column: data.filter(col(column).isNull()).count() for column in data.columns
}

for column, count in null_counts.items():
        print(f"{column}: {count} null values")
```

DayofMonth: O null values DayOfWeek: O null values Carrier: O null values

OriginAirportID: O null values
DestAirportID: O null values
DepDelay: O null values
ArrDelay: O null values

Late: 0 null values

```
[11]: strIdx = StringIndexer(inputCol = "Carrier", outputCol = "CarrierIdx")
     str_encoder = OneHotEncoder(inputCol='CarrierIdx',outputCol='CarrierVec')
[12]: output = strIdx.fit(data).transform(data)
[13]: feature_cols = ['DayofMonth',
       'DayOfWeek',
       'CarrierIdx',
       'OriginAirportID',
       'DestAirportID',
       'ArrDelay',]
[14]: assembler = VectorAssembler(inputCols=feature_cols, outputCol="features")
[15]: output = assembler.transform(output)
[16]: final_data = output.select("features","Late")
     final_data.show(5)
     +----+
                 features | Late |
     +----+
     |[19.0,5.0,1.0,114...|
                           01
     [19.0,5.0,1.0,148...]
                           01
     |[19.0,5.0,1.0,140...|
                           0|
     |[19.0,5.0,1.0,150...|
                           1 |
     |[19.0,5.0,1.0,111...|
     +----+
     only showing top 5 rows
[17]: train_data,test_data = final_data.randomSplit([0.7, 0.3], seed = 1)
[18]: train_data, valid_data = train_data.randomSplit([0.9, 0.1], seed = 1)
[22]: | scaler = StandardScaler(inputCol="features", outputCol="scaled_features")
      scaler model = scaler.fit(train data)
      train_data = scaler_model.transform(train_data)
      test_data = scaler_model.transform(test_data)
      valid_data = scaler_model.transform(valid_data)
[23]: from pyspark.ml.feature import PCA
     pca = PCA(k=2, inputCol="scaled_features", outputCol="pca_features")
     pca_model = pca.fit(train_data)
```

```
train_pca = pca_model.transform(train_data)
test_pca = pca_model.transform(test_data)
valid_pca = pca_model.transform(valid_data)

# Show PCA output
train_pca.select("pca_features").show(truncate=False)
```

```
|pca_features
| [-4.713894746708608, -3.387776582510908]
|[-4.554491540208862,-3.2418025342565118] |
|[-3.450930879826006,-2.231212969418383]
|[-4.960337058640134,-3.6107821473702155]
|[-4.825457422371118,-3.4872656450011106]
|[-4.825457422371118,-3.4872656450011106] |
|[-4.8131956372557525,-3.4760368720584647]|
|[-4.8131956372557525,-3.4760368720584647]|
|[-4.800933852140387,-3.4648080991158188] |
| [-4.751886711678927, -3.4198930073452356] |
|[-4.617007075409911,-3.2963765049761307]
|[-4.757520619373999,-3.424850748306091]
|[-4.7452588342586335,-3.413621975363445]
|[-4.659426338451078,-3.335020564764924]
| [-4.610379197989618, -3.29010547299434]
|[-5.119452269907024,-3.750800770870742]
[-4.960049063407277,-3.604826722616345]
|[-4.678028005753881,-3.34656494493549]
|[-4.616719080177055,-3.2904210802222607]
|[-3.5376819900249292,-2.3022890612694233]|
only showing top 20 rows
```

[]: