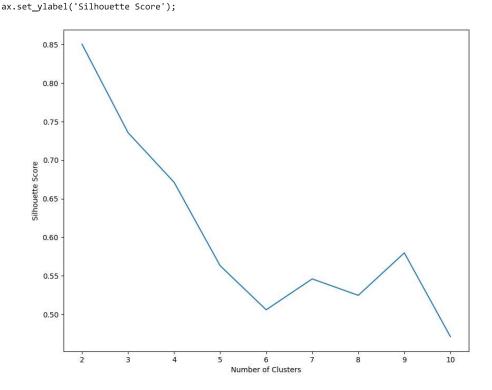
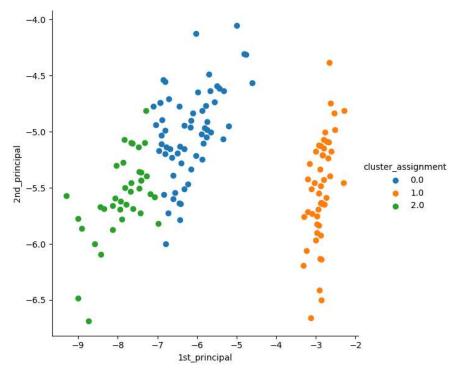
```
!pip install pyspark
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Requirement already satisfied: pyspark in /usr/local/lib/python3.10/dist-packages (3.4.0)
     Requirement already satisfied: py4j==0.10.9.7 in /usr/local/lib/python3.10/dist-packages (from pyspark) (0.10.9.7)
from pyspark.sql import SparkSession
from pyspark import SparkContext
spark = SparkSession.builder.getOrCreate()
spark
□ SparkSession - in-memory
     SparkContext
     Spark UI
     Version
          v3.4.0
     Master
          local[*]
     AppName
          pyspark-shell
# Read data
df = spark.read.csv("iris.csv",header= True)
# Remove columns
df = df.drop("Id").drop("Species")
from pyspark.sql.functions import col
# Convert string col to float col
float_cols = ["SepalLengthCm", "SepalWidthCm", "PetalLengthCm", "PetalWidthCm"]
df = df.select([col(col_name).cast("float").alias(col_name) for col_name in float_cols])
df.show()
     |SepalLengthCm|SepalWidthCm|PetalLengthCm|PetalWidthCm|
                5.1
                            3.5
                                          1.4
               4.9
                            3.0
                                          1.4
                                                       0.21
               4.7
                            3.2
                                          1.3
                                                       0.2
                4.6
                            3.1
                                          1.5
                                                       0.2
                5.0
                            3.6
                                          1.4
                                                       0.2
                                          1.7
                5.4
                            3.9
                                                       0.4
                4.6
                            3.4
                                          1.4
                                                       0.3
                5.0
                             3.4
                                          1.5
                                                       0.2
                4.4
                             2.9
                                           1.4
                                                       0.2
                4.9
                             3.1
                                          1.5
                                                       0.1
                5.4
                             3.7
                                          1.5
                                                       0.2
                4.8
                             3.4
                                           1.6
                                                       0.2
                4.8
                            3.0
                                          1.4
                                                       0.1
                4.3
                             3.0
                                           1.1
                                                       0.1
                                           1.2
                5.8
                             4.0
                                                       0.2
                5.7
                             4.4
                                          1.5
                                                       0.4
                5.4
                             3.9
                                           1.3
                                                       0.4
                5.1
                             3.5
                                           1.4
                                                       0.3
                                           1.7
                                                       0.3
                5.7I
                             3.8
                5.1
                             3.8
                                           1.5
                                                       0.3
     only showing top 20 rows
# Create feature vector
from pyspark.ml.feature import VectorAssembler
assemble=VectorAssembler(inputCols=[
'SepalLengthCm',
'SepalWidthCm',
'PetalLengthCm',
'PetalWidthCm'],outputCol = 'iris_features')
```

```
assembled_data=assemble.transform(df)
# Using the silhouette method as an evaluation metric for clustering algorithms.
from pyspark.ml.clustering import KMeans
from pyspark.ml.evaluation import ClusteringEvaluator
silhouette_scores=[]
evaluator = ClusteringEvaluator(featuresCol='iris_features', \
metricName='silhouette', distanceMeasure='squaredEuclidean')
for K in range(2,11):
    KMeans_=KMeans(featuresCol='iris_features', k=K)
   KMeans_fit=KMeans_.fit(assembled_data)
   KMeans_transform=KMeans_fit.transform(assembled_data)
    evaluation_score=evaluator.evaluate(KMeans_transform)
    silhouette_scores.append(evaluation_score)
# Visualize sihouette score with the number of clusters
import matplotlib.pyplot as plt
fig, ax = plt.subplots(1,1, figsize =(10,8))
ax.plot(range(2,11),silhouette_scores)
ax.set_xlabel('Number of Clusters')
```



```
#Build the K-Means Clustering model
KMeans_=KMeans(featuresCol='iris_features', k=3)
KMeans_Model=KMeans_.fit(assembled_data)
KMeans_Assignments=KMeans_Model.transform(assembled_data)
#Using PCA to visualize clustering
from pyspark.ml.feature import PCA as PCAml
```

```
pca = PCAml(k=2, inputCol="iris_features", outputCol="pca")
pca_model = pca.fit(assembled_data)
pca_transformed = pca_model.transform(assembled_data)
# Extract the principal components
import numpy as np
x_pca = np.array(pca_transformed.rdd.map(lambda row: row.pca).collect())
# Retrieve the cluster assignments from k-means assignments
cluster\_assignment = np.array(\texttt{KMeans\_Assignments.rdd.map(lambda row: row.prediction).collect()).reshape(-1,1)
# Show clusters
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
pca_data = np.hstack((x_pca,cluster_assignment))
pca_df = pd.DataFrame(data=pca_data, columns=("1st_principal", "2nd_principal", "cluster_assignment"))
sns.FacetGrid(pca_df,hue="cluster_assignment", height=6).map(plt.scatter, '1st_principal', '2nd_principal' ).add_legend()
plt.show()
```



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