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Simple example of using Spark to compute Kmeans

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from pyspark.mllib.clustering import KMeans
from numpy import array, random
from math import sqrt
from pyspark import SparkConf, SparkContext
from sklearn.preprocessing import scale
K = 5
# Standard Spark:
conf = SparkConf().setMaster("local").setAppName("SparkKMeans")
sc = SparkContext(conf = conf)
#Create fake income/age clusters for N people in k clusters
def createClusteredData(N, k):
  random.seed(10)
  pointsPerCluster = float(N)/k
  X = []
  for i in range (k):
    incomeCentroid = random.uniform(20000.0, 200000.0)
    ageCentroid = random.uniform(20.0, 70.0)
    for j in range(int(pointsPerCluster)):
      X.append([random.normal(incomeCentroid, 10000.0), random.normal(ageCentroid, 2.0)])
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X = array(X)
  return X
random.seed(0)
# Create an RDD, load the data; I am normalizing it with scale()
data = sc.parallelize(scale(createClusteredData(100, K)))
# Build the model (cluster the data)
clusters = KMeans.train(data, K, maxIterations=10,
    initializationMode="random")
# Print out the cluster assignments
resultRDD = data.map(lambda point: clusters.predict(point)).cache()
#the cache call is avoid re-computing results
print("Counts by value:")
counts = resultRDD.countByValue()
print(counts)
print("Cluster assignments:")
results = resultRDD.collect()
print(results)
# Evaluate clustering by computing Within Set Sum of Squared Errors
def error(point):
  center = clusters.centers[clusters.predict(point)]
  return sqrt(sum([x**2 for x in (point - center)]))
WSSSE = data.map(lambda point: error(point)).reduce(lambda x, y: x + y)
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print("Within Set Sum of Squared Error = " + str(WSSSE))