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Load the RDF File Using Apache Jena

Screenshot -

```
vedant@VEDANT-PC: ~/spark/conf
scala> val rdfFilePath = "file:///home/vedant/BDALab/LabAssign7/tb_symptoms.ttl"
rdfFilePath: String = file:///home/vedant/BDALab/LabAssign7/tb_symptoms.ttl

scala> val model = ModelFactory.createDefaultModel()
25/03/28 13:34:58 ERROR JenaXMLInput: Problem setting StAX property - name: "http://javax.xml.XMLConst
ants/property/accessExternalDTD" - value: "" - error: Unrecognized property 'http://javax.xml.XMLConst
ants/property/accessExternalDTD'
25/03/28 13:34:58 ERROR JenaXMLInput: Problem setting StAX property - name: "http://javax.xml.XMLConst
ants/property/accessExternalDTD" - value: "" - error: Unrecognized property 'http://javax.xml.XMLConst
ants/property/accessExternalDTD'
25/03/28 13:34:58 ERROR JenaXMLInput: Problem setting StAX property - name: "http://javax.xml.XMLConst
ants/property/accessExternalDTD" - value: "" - error: Unrecognized property 'http://javax.xml.XMLConst
ants/property/accessExternalDTD'
model: org.apache.jena.rdf.model.Model = <ModelCom   {} | >

scala> model.read(rdfFilePath, "TTL")
res2: org.apache.jena.rdf.model.Model = <ModelCom   {tb:patient/1597190594 schema1:name "Myrvyn"; tb:p
atient/1597190594 schema1:gender "Female"; tb:patient/1597190594 schema1:date "2021-01-21"^^xsd:date;
tb:patient/1597190594 tb:time "15:12:00"^^xsd:time; tb:patient/1597190594 tb:hasSymptom tb:symptom/spu
tum_mixed_with_blood; tb:patient/1597190594 tb:hasSymptom tb:symptom/shortness_of_breath; tb:patient/1
597190594 tb:hasSymptom tb:symptom/loss_of_appetite; tb:patient/1597190594 tb:hasSymptom tb:symptom/fe
ver_for_two_weeks; tb:patient/1597190594 tb:hasSymptom tb:symptom/body_feels_tired; tb:patient/2518367
500 schema1:name "Godart"; tb:patient/2518367500 schema1:gender "Male"; tb:patient/2518367500 schema1:
date "2020-03-03"^^xsd:date; tb:patient/2518367500 tb:tim...

scala> println(s"Loaded RDF model with ${model.size} triples")
Loaded RDF model with 8943 triples

scala> 
```

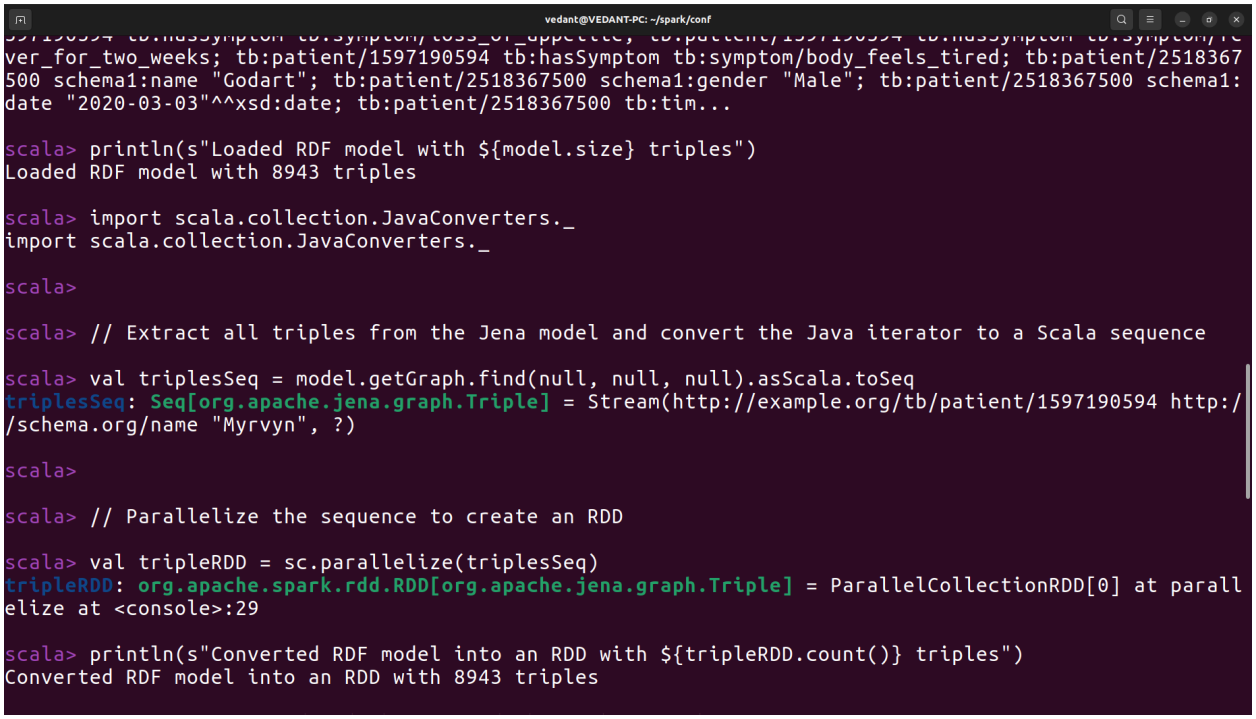
Convert the RDF Model into an RDD of Triples

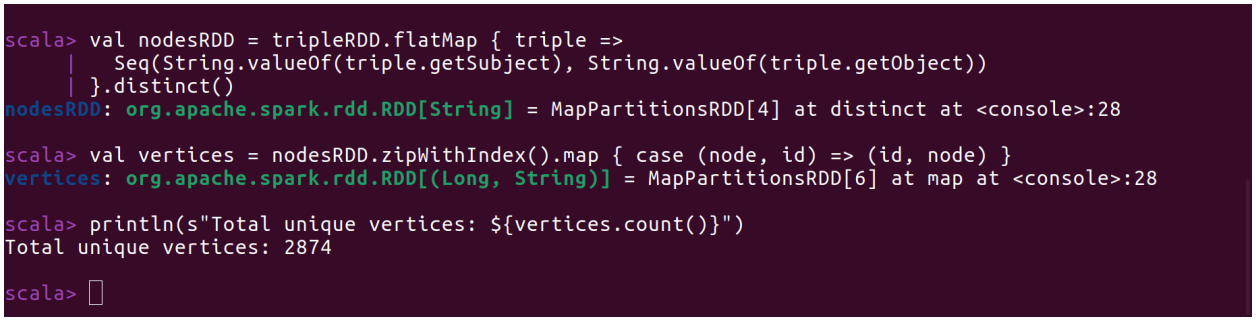
Screenshot -

```
vedant@VEDANT-PC: ~/spark/conf
500 schema1:name "Godart"; tb:patient/2518367500 schema1:gender "Male"; tb:patient/2518367500 schema1:
date "2020-03-03"^^xsd:date; tb:patient/2518367500 tb:tim...

scala> println(s"Loaded RDF model with ${model.size} triples")
Loaded RDF model with 8943 triples

scala> import scala.collection.JavaConverters._
import scala.collection.JavaConverters._

scala>

scala> // Extract all triples from the Jena model and convert the Java iterator to a Scala sequence

scala> val triplesSeq = model.getGraph.find(null, null, null).asScala.toSeq
triplesSeq: Seq[org.apache.jena.graph.Triple] = Stream(http://example.org/tb/patient/1597190594 http:/
/schema.org/name "Myrvyn", ?)

scala>

scala> // Parallelize the sequence to create an RDD

scala> val tripleRDD = sc.parallelize(triplesSeq)
tripleRDD: org.apache.spark.rdd.RDD[org.apache.jena.graph.Triple] = ParallelCollectionRDD[0] at parall
elize at <console>:29

scala> println(s"Converted RDF model into an RDD with ${tripleRDD.count()} triples")
Converted RDF model into an RDD with 8943 triples

scala> 
```

Create Vertices RDD from RDF Triples

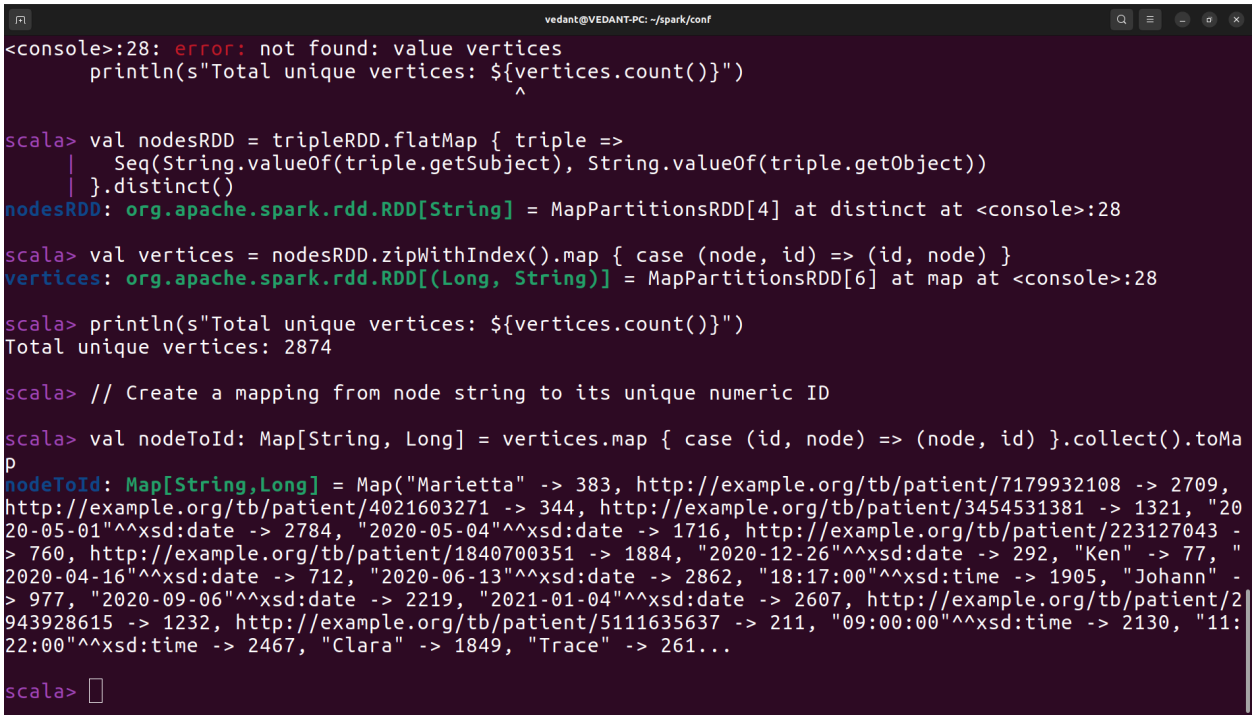
Screenshot -





Create a Mapping from Node URI to Vertex ID

Screenshot -



Create GraphX Graph -

Screenshot -

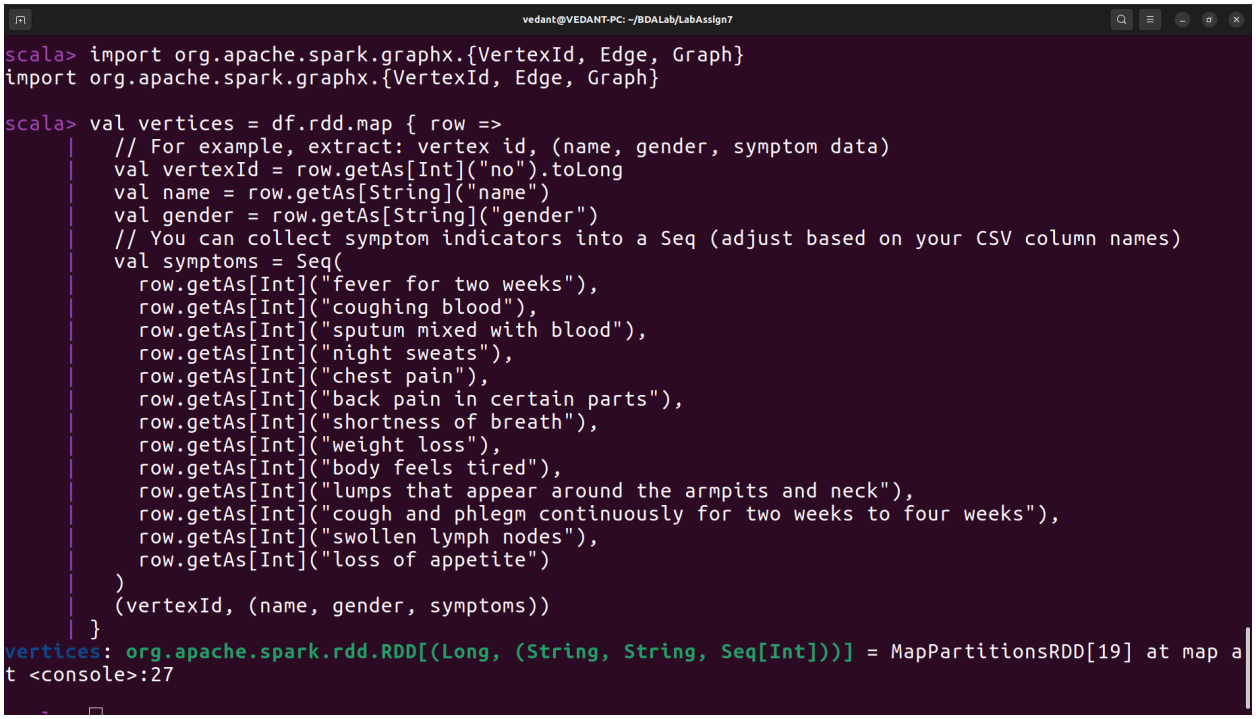
Create Vertices and Edges for the GraphX Graph :

Convert the DataFrame to an RDD and Create Vertices

Code -

```
import org.apache.spark.graphx.{VertexId, Edge, Graph}

// Create an RDD of vertices from the cleaned DataFrame.
val vertices = cleanedDF.rdd.map { row =>
  // Extract: vertex id, (name, gender, symptom data)
  val vertexId = row.getAs[Int]("no").toLong
  val name = row.getAs[String]("name")
  val gender = row.getAs[String]("gender")

  // Collect symptom indicators (adjusted column names)
  val symptoms = Seq(
          row.getAs[Int]("fever_for_two_weeks"),
          row.getAs[Int]("coughing_blood"),
          row.getAs[Int]("sputum_mixed_with_blood"),
          row.getAs[Int]("night_sweats"),
          row.getAs[Int]("chest_pain"),
          row.getAs[Int]("back_pain_in_certain_parts"),
          row.getAs[Int]("shortness_of_breath"),
          row.getAs[Int]("weight_loss"),
          row.getAs[Int]("body_feels_tired"),
          row.getAs[Int]("lumps_that_appear_around_the_armpits_and_neck"),
          row.getAs[Int]("cough_and_phlegm_continuously_for_two_weeks_to_four_weeks"),
          row.getAs[Int]("swollen_lymph_nodes"),
          row.getAs[Int]("loss_of_appetite")
  )

  (vertexId, (name, gender, symptoms))
}
```

Screenshot -

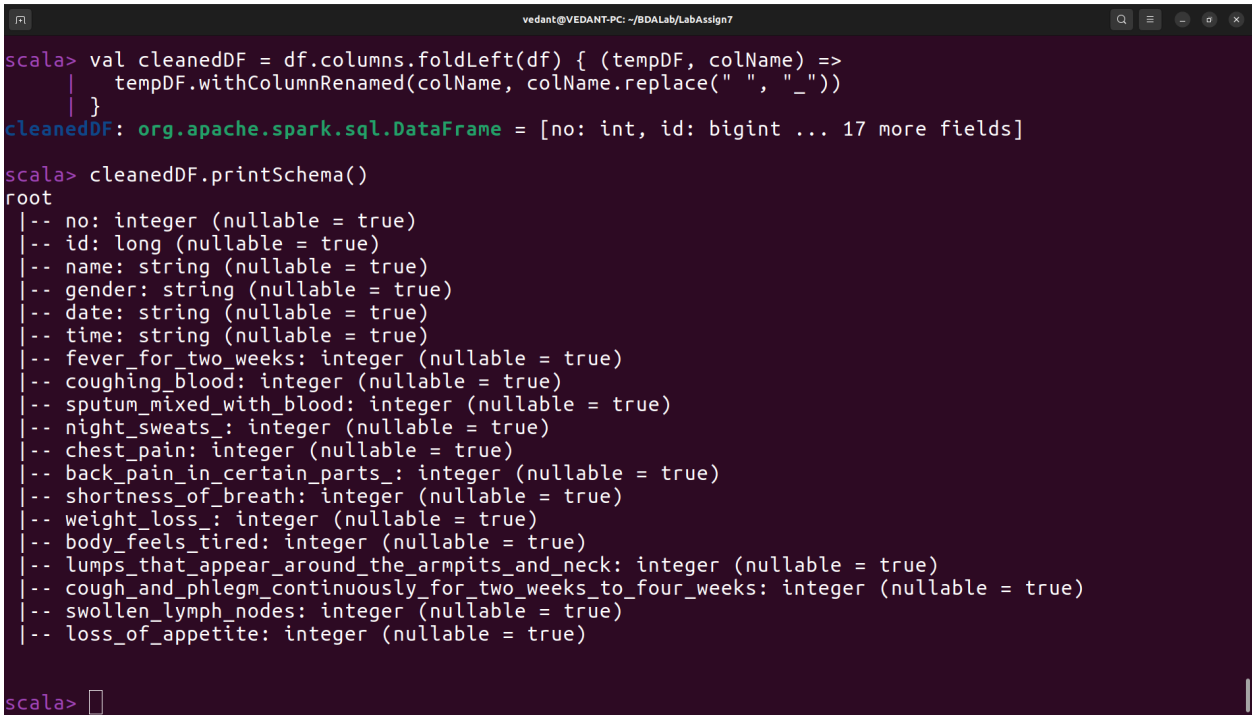


Clean Column Names, Replace space with Underscore

Code -

```
val cleanedDF = df.columns.foldLeft(df)((tempDF, colName) =>
  tempDF.withColumnRenamed(colName, colName.trim.replaceAll("\\s+", "_").replaceAll("^_|_$", "").toLowerCase))

// Show schema after renaming
cleanedDF.printSchema()
```

Screenshot -

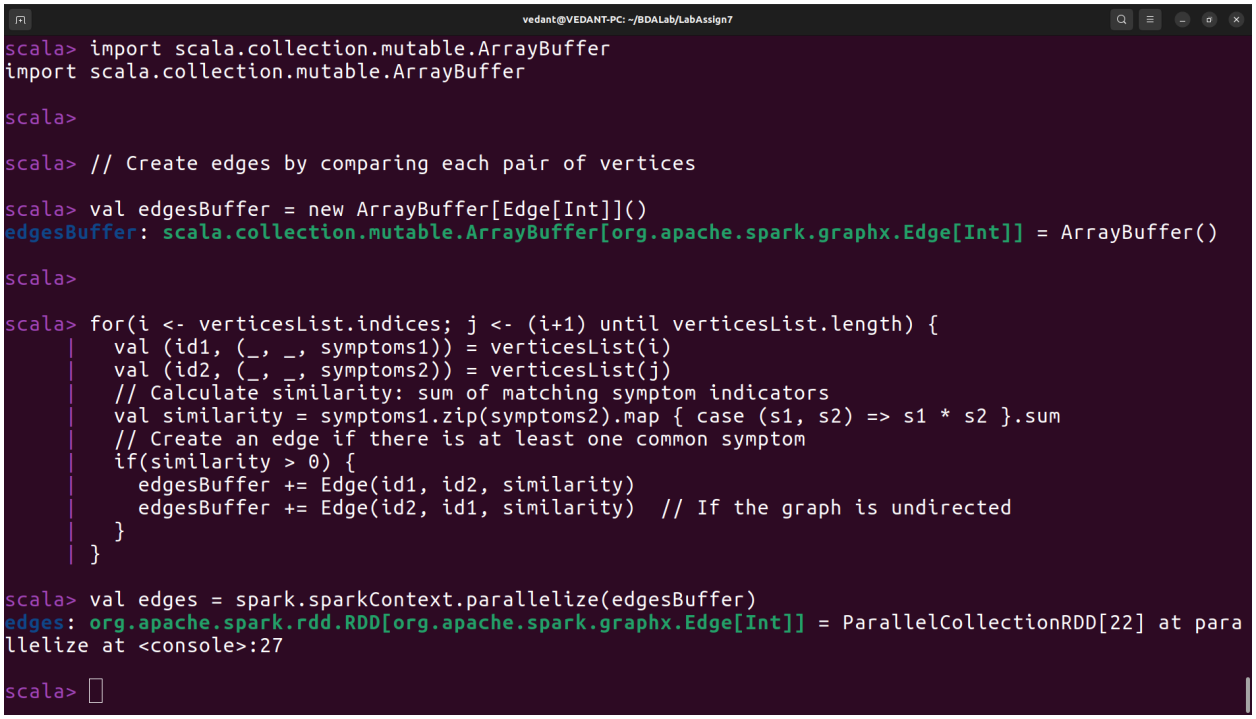


Create an RDD of Edges

Code -

```
// Create a list of vertices (this approach works for a small dataset)
val verticesList = vertices.collect()
import scala.collection.mutable.ArrayBuffer

// Create edges by comparing each pair of vertices
val edgesBuffer = new ArrayBuffer[Edge[Int]]()

for(i <- verticesList.indices; j <- (i+1) until verticesList.length) {
  val (id1, (_, _, symptoms1)) = verticesList(i)
  val (id2, (_, _, symptoms2)) = verticesList(j)
  // Calculate similarity: sum of matching symptom indicators
  val similarity = symptoms1.zip(symptoms2).map { case (s1, s2) => s1 * s2 }.sum
  // Create an edge if there is at least one common symptom
  if(similarity > 0) {
        edgesBuffer += Edge(id1, id2, similarity)
        edgesBuffer += Edge(id2, id1, similarity)  // If the graph is undirected
  }
}
val edges = spark.sparkContext.parallelize(edgesBuffer)
```

Screenshot -

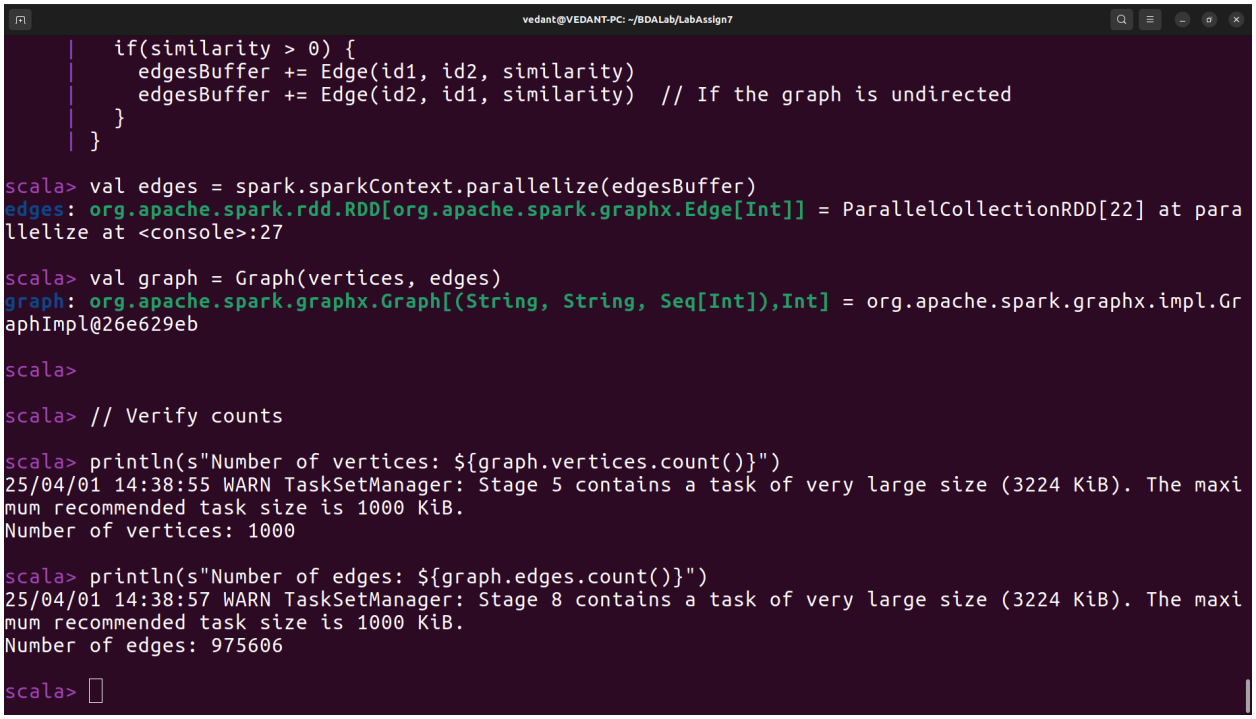


Build the GraphX Graph

Code -

```
val graph = Graph(vertices, edges)

// Verify counts
println(s"Number of vertices: ${graph.vertices.count()}")
println(s"Number of edges: ${graph.edges.count()}")
```

Screenshot -



Part 3: Graph Operations :

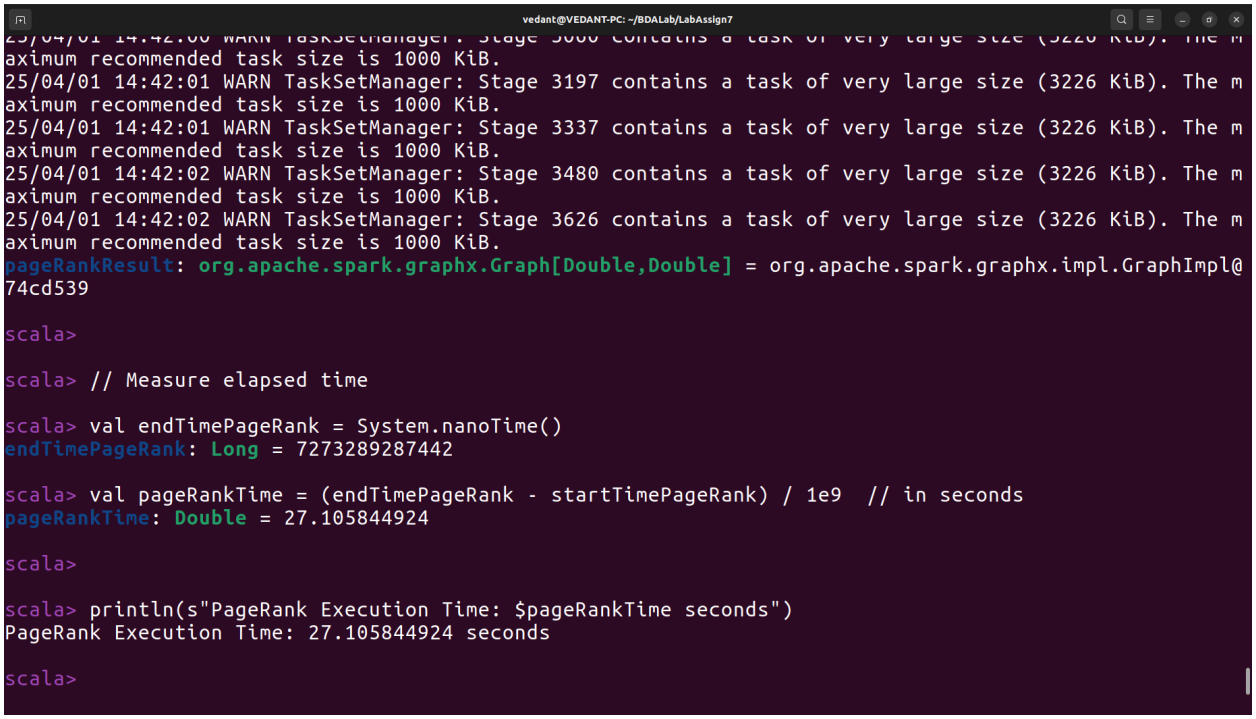
a) **PageRank**

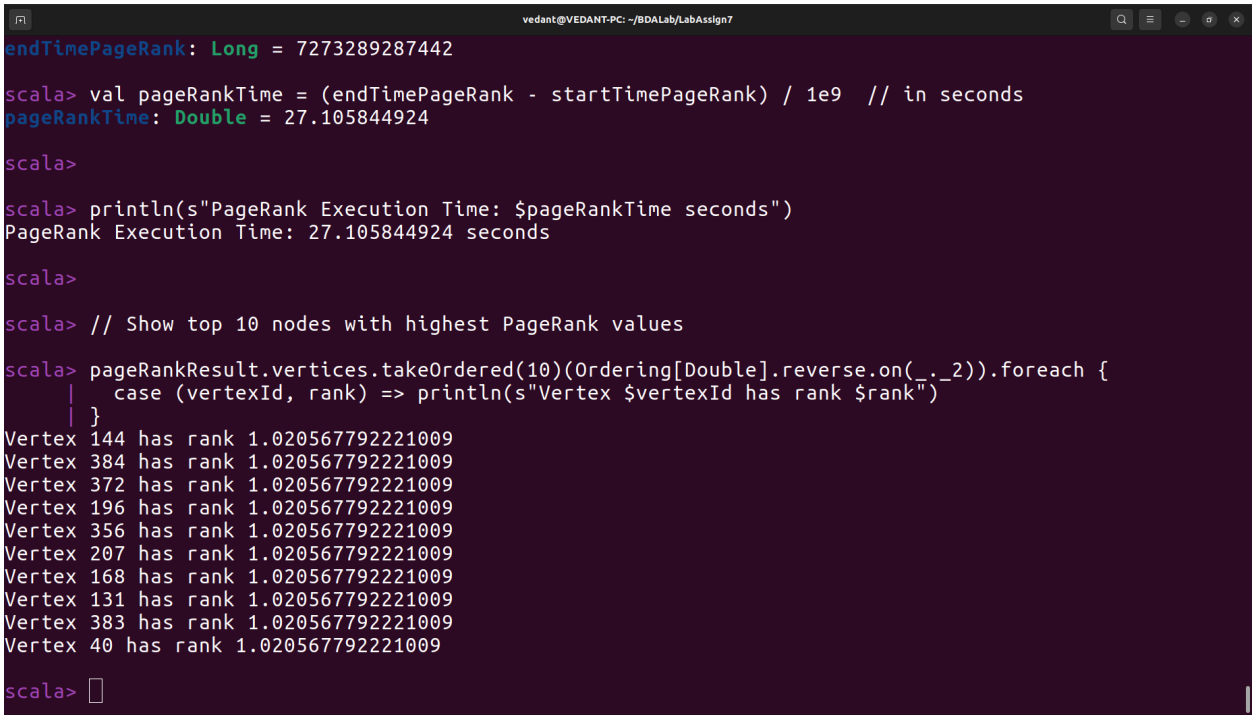
PageRank is used to determine the most influential nodes in a graph based on their connections. For this, you can apply `graph.pageRank` on the created graph.

Code -

```
import org.apache.spark.graphx.PageRank

// Measure execution time for PageRank
val startTimePageRank = System.nanoTime()

// Perform PageRank
val pageRankResult = graph.pageRank(0.0001)  // Convergence threshold

// Measure elapsed time
val endTimePageRank = System.nanoTime()
val pageRankTime = (endTimePageRank - startTimePageRank) / 1e9  // in seconds

println(s"PageRank Execution Time: $pageRankTime seconds")
```

// Show top 10 nodes with highest PageRank values
pageRankResult.vertices.takeOrdered(10)(Ordering[Double].reverse.on(_._2)).foreach {
  case (vertexId, rank) => println(s"Vertex $vertexId has rank $rank")
}
Screenshot -





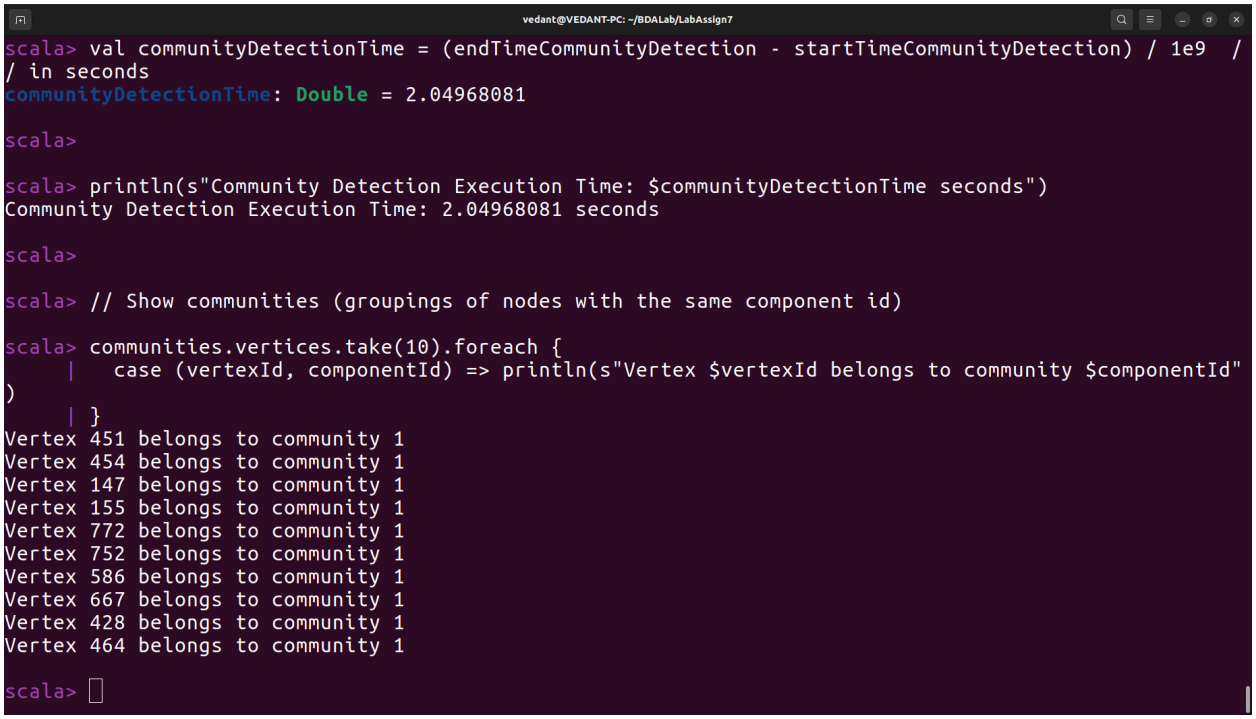
b) Community Detection

Community detection can be done using `GraphX`'s `connectedComponents` or more advanced methods like `Label Propagation`. For simplicity, let's use `connectedComponents` here:

Code -

```
// Measure execution time for Community Detection
val startTimeCommunityDetection = System.nanoTime()

// Perform community detection
val communities = graph.connectedComponents()

// Measure elapsed time
val endTimeCommunityDetection = System.nanoTime()
val communityDetectionTime = (endTimeCommunityDetection - startTimeCommunityDetection)
/ 1e9  // in seconds

println(s"Community Detection Execution Time: $communityDetectionTime seconds")

// Show communities (groupings of nodes with the same component id)
communities.vertices.take(10).foreach {
  case (vertexId, componentId) => println(s"Vertex $vertexId belongs to community
$componentId")
}
```

Screenshot -



c) Connected Components

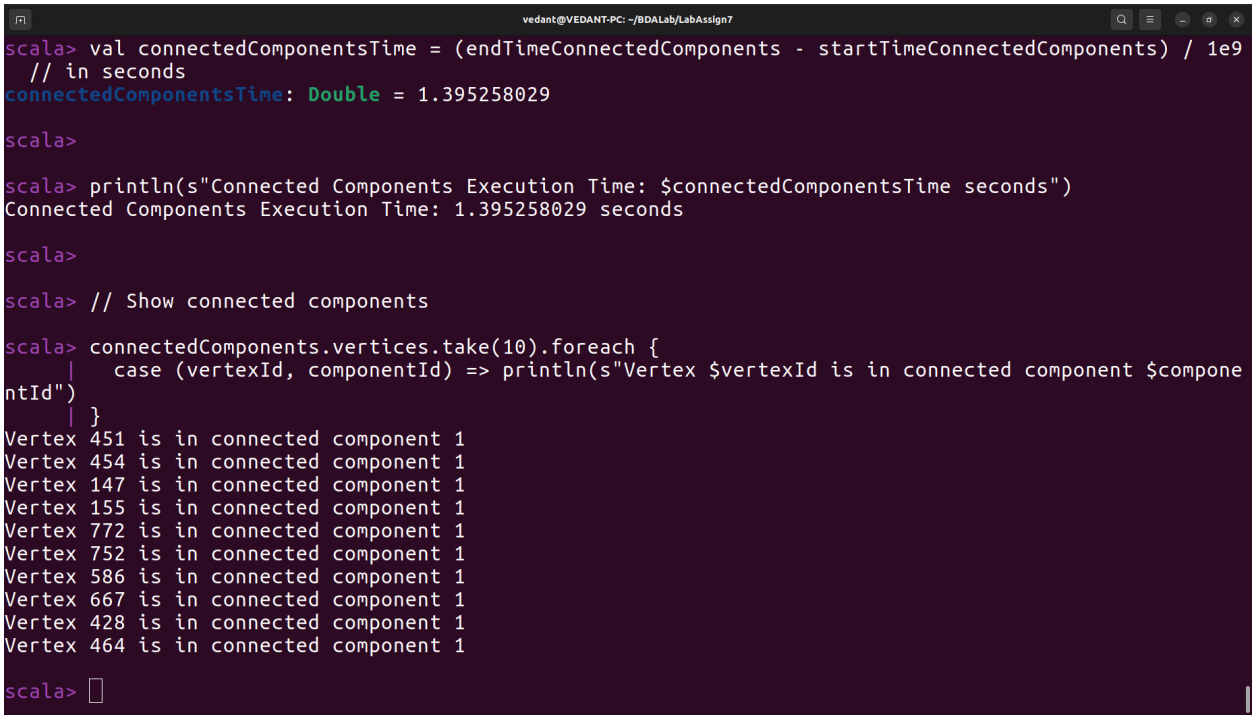
`connectedComponents` is used for identifying isolated groups (connected components) in the graph. This is already included in the community detection part, but you can also run it independently:

Code -

```
// Measure execution time for Connected Components
val startTimeConnectedComponents = System.nanoTime()

// Get connected components
val connectedComponents = graph.connectedComponents()

// Measure elapsed time
val endTimeConnectedComponents = System.nanoTime()
val connectedComponentsTime = (endTimeConnectedComponents -
startTimeConnectedComponents) / 1e9  // in seconds

println(s"Connected Components Execution Time: $connectedComponentsTime seconds")

// Show connected components
connectedComponents.vertices.take(10).foreach {
```

```
  case (vertexId, componentId) => println(s"Vertex $vertexId is in connected component
$componentId")
}
```

Screenshot -



d) Shortest Path

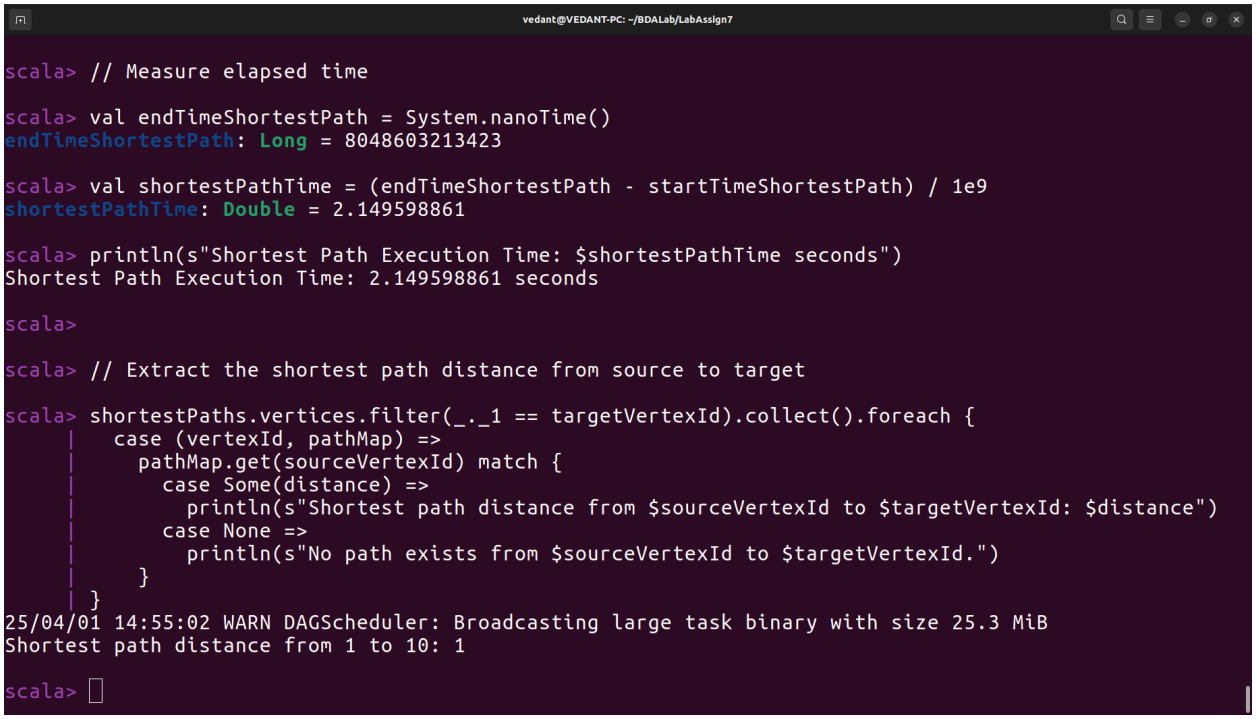
You can compute the shortest path using `GraphX`'s `shortestPaths` function. Here's an example for calculating the shortest path between two nodes:

Code -

```
import org.apache.spark.graphx.{Graph, VertexId}
import org.apache.spark.graphx.lib.ShortestPaths // Import the ShortestPaths library

// Measure execution time for Shortest Path
val startTimeShortestPath = System.nanoTime()

// Specify source and target vertices
val sourceVertexId: VertexId = 1L
val targetVertexId: VertexId = 10L

// Compute shortest paths from the source vertex using ShortestPaths.run()
```

```
val shortestPaths = ShortestPaths.run(graph, Seq(sourceVertexId))

// Measure elapsed time
val endTimeShortestPath = System.nanoTime()
val shortestPathTime = (endTimeShortestPath - startTimeShortestPath) / 1e9
println(s"Shortest Path Execution Time: $shortestPathTime seconds")

// Extract the shortest path distance from source to target
shortestPaths.vertices.filter(_._1 == targetVertexId).collect().foreach {
  case (vertexId, pathMap) =>
        pathMap.get(sourceVertexId) match {
        case Some(distance) =>
        println(s"Shortest path distance from $sourceVertexId to $targetVertexId: $distance")
        case None =>
        println(s"No path exists from $sourceVertexId to $targetVertexId.")
        }
}
```

Screenshot -



Execution Time :

Operation	Execution Time
PageRank	27.105 sec
Community Detection	2.049 sec
Connected Components	1.395 sec
Shortest Path	2.149 sec