kNN Classification of members of congress using similarity algorithms in Neo4j.

Tenemos que instalar los siguientes plugins.

- Neo4j graph algorithms plugin
- Neo4j APOC plugin
- 1. Código para importar a Neo4j.

LOAD CSV FROM "http://archive.ics.uci.edu/ml/machine-learning-databases/voting-records/house-votes-84.data" as row

CREATE (p:Person)

SET p.class = row[0],

p.features = row[1..];

2. Verificamos el numero de miembros del congreso que tienen menos un voto.



3. En el resultado anterior, se pude ver que casi la mitad tienen votos faltantes, por lo que con el siguiente codigo se procede a revisar cual es la distribución de los votos faltantes.

MATCH (p:Person)

WHERE '?' in p.features

WITH p,apoc.coll.occurrences(p.features,'?') as missing

RETURN missing,count(*) as times ORDER BY missing ASC

 massing
 times

 A rest
 1
 124

 2
 43
 3

 3
 16
 6

 5
 5
 6

 7
 1
 9

 14
 1
 1

 15
 1
 1

 16
 1
 1

4. Datos de entrenamiento.

Started streaming 11 records after 2 ms and completed after 67 ms.

80% datos de entrenamiento.

20% datos de prueba.

Total 344 nodos.

MATCH (p:Person)

WITH p LIMIT 344

SET p:Training;

5. Marcar datos de entrenamiento.

neo4j\$ MATCH (p:Person) WITH p SKIP 344 SET p:Test



Added 91 labels, completed after 16 ms.

6. Vector de caracteristicas.

Vamos a mapear el vector de las siguientes características; 'y' para 1, 'n' para 0, y '?' para 0.5

Transformamos el vector de características.

MATCH (n:Person)

UNWIND n.features as feature

WITH n,collect(CASE feature WHEN 'y' THEN 1

WHEN 'n' THEN 0

ELSE 0.5 END) as feature_vector

SET n.feature vector = feature vector

7. Algoritmo clasificador KNN.

MATCH (test:Test)

WITH test,test.feature_vector as feature_vector

CALL apoc.cypher.run('MATCH (training:Training)

// calculate euclidian distance between each test node and all training nodes

 $WITH\ training, algo. similarity. euclidean Distance (\$ feature_vector,\ training. feature_vector)\ AS\ similarity. In the property of the p$

// return only top 3 nodes

ORDER BY similarity ASC LIMIT 3

RETURN collect(training.class) as classes',

 $\{feature_vector: feature_vector\}) \ YIELD \ value$

 $WITH\ test. class\ as\ class,\ apoc. coll. sortMaps (apoc. coll. frequencies (value. classes),\ '`count') [-1]. item\ as\ predicted_class$

WITH sum(CASE when class = predicted_class THEN 1 ELSE 0 END) as correct_predictions, count(*) as total_predictions

 $RETURN\ correct_predictions, total_predictions,\ correct_predictions/\ toFloat(total_predictions)\ as\ ratio$