### Node Embeddings

This notebook demonstrates different methods for node embeddings and how to further reduce their dimensionality to be able to visualize them in a 2D plot.

Node embeddings are essentially an array of floating point numbers (length = embedding dimension) that can be used as "features" in machine learning. These numbers approximate the relationship and similarity information of each node and can also be seen as a way to encode the topology of the graph.

#### Considerations

Due to dimensionality reduction some information gets lost, especially when visualizing node embeddings in two dimensions. Nevertheless, it helps to get an intuition on what node embeddings are and how much of the similarity and neighborhood information is retained. The latter can be observed by how well nodes of the same color and therefore same community are placed together and how much bigger nodes with a high centrality score influence them.

If the visualization doesn't show a somehow clear separation between the communities (colors) here are some ideas for tuning:

- Clean the data, e.g. filter out very few nodes with extremely high degree that aren't actually that important
- Try directed vs. undirected projections
- Tune the embedding algorithm, e.g. use a higher dimensionality
- Tune t-SNE that is used to reduce the node embeddings dimension to two dimensions for visualization.

It could also be the case that the node embeddings are good enough and well suited the way they are despite their visualization for the down stream task like node classification or link prediction. In that case it makes sense to see how the whole pipeline performs before tuning the node embeddings in detail.

### Note about data dependencies

PageRank centrality and Leiden community are also fetched from the Graph and need to be calculated first. This makes it easier to see if the embeddings approximate the structural information of the graph in the plot. If these properties are missing you will only see black dots all of the same size.

#### References

- jqassistant
- Neo4j Python Driver
- Tutorial: Applied Graph Embeddings
- Visualizing the embeddings in 2D
- scikit-learn TSNE
- AttributeError: 'list' object has no attribute 'shape'
- Fast Random Projection (neo4j)
- HashGNN (neo4j)
- node2vec (neo4j) computes a vector representation of a node based on second order random walks in the graph.
- Complete guide to understanding Node2Vec algorithm

The openTSNE version is: 1.0.2 The pandas version is: 2.2.3

## Dimensionality reduction with t-distributed stochastic neighbor embedding (t-SNE)

The following function takes the original node embeddings with a higher dimensionality, e.g. 64 floating point numbers, and reduces them into a two dimensional array for visualization.

It converts similarities between data points to joint probabilities and tries to minimize the Kullback-Leibler divergence between the joint probabilities of the low-dimensional embedding and the high-dimensional data.

(see https://opentsne.readthedocs.io)

### 1. Java Packages

## 1.1 Generate Node Embeddings using Fast Random Projection (Fast RP) for Java Packages

Fast Random Projection is used to reduce the dimensionality of the node feature space while preserving most of the distance information. Nodes with similar neighborhood result in node embedding with similar vectors.

← Hint: To skip existing node embeddings and always calculate them based on the parameters below edit Node\_Embeddings\_0a\_Query\_Calculated so that it won't return any results.

Received notification from DBMS server: {severity: WARNING} {code: Neo.ClientNotifica tion.Statement.UnknownRelationshipTypeWarning} {category: UNRECOGNIZED} {title: The p rovided relationship type is not in the database.} {description: One of the relations hip types in your query is not available in the database, make sure you didn't misspe ll it or that the label is available when you run this statement in your application (the missing relationship type is: HAS ROOT)} {position: line: 9, column: 44, offset: 696} for query: '// Query already calculated and written node embeddings on nodes wit h label in parameter \$dependencies projection node including a communityId and centra lity. Variables: dependencies projection node, dependencies projection write propert y. Requires "Add file name and extension.cypher".\n \n MATCH (codeUnit)\n WHERE \$de pendencies projection node IN LABELS(codeUnit)\n AND codeUnit[\$dependencies projec tion write property] IS NOT NULL\n // AND codeUnit.notExistingToForceRecalculation IS NOT NULL // uncomment this line to force recalculation\n OPTIONAL MATCH (artifact: Java:Artifact)-[:CONTAINS]->(codeUnit)\n WITH \*, artifact.name AS artifactName\n 0 PTIONAL MATCH (projectRoot:Directory)<-[:HAS ROOT]-(proj:TS:Project)-[:CONTAINS]->(co WITH \*, last(split(projectRoot.absoluteFileName, \'/\')) AS projectName \n RETURN DISTINCT \n coalesce(codeUnit.fgn, codeUnit.globalFgn, codeUnit.fi leName, codeUnit.signature, codeUnit.name) AS codeUnitName\n ,codeUnit.name ,coalesce(artifactName, projectName) AS shortCodeUnitName\n AS projectName\n , coalesce(codeUnit.communityLeidenId, 0) AS communi ,coalesce(codeUnit.centralityPageRank, 0.01) AS centrality\n ,codeUnit[\$dependencies projection write property] AS embedding\n ORDER BY communit yId'

The results have been provided by the query filename: ../cypher/Node\_Embeddings/Node\_Embeddings\_0a\_Query\_Calculated.cypher

embedding	centrality	communityId	projectName	${\bf short Code Unit Name}$	codeUnitName	
[0.07319182902574539, -0.11561159789562225, 0	0.012156	0	axon-disruptor- 4.11.0	commandhandling	org.axonframework.disruptor.commandhandling	0
[0.0698334202170372, -0.16919010877609253, 0.2	0.033013	0	axon-modelling- 4.11.0	command	org.axonframework.modelling.command	1
[0.04140711575746536, -0.1820932924747467, 0.2	0.019850	0	axon-modelling- 4.11.0	inspection	org. ax on framework. modelling. command. in spection	2
[-0.06166009232401848, -0.15027368068695068, 0	0.011941	0	axon-modelling- 4.11.0	legacyjpa	org.axonframework.modelling.command.legacyjpa	3
[0.07438604533672333, -0.14707809686660767, 0	0.017131	0	axon- eventsourcing- 4.11.0	eventsourcing	org.axonframework.eventsourcing	4

# 1.2 Dimensionality reduction with t-distributed stochastic neighbor embedding (t-SNE)

This step takes the original node embeddings with a higher dimensionality, e.g. 64 floating point numbers, and reduces them into a two dimensional array for visualization. For more details look up the function declaration for

<sup>&</sup>quot;prepare\_node\_embeddings\_for\_2d\_visualization".

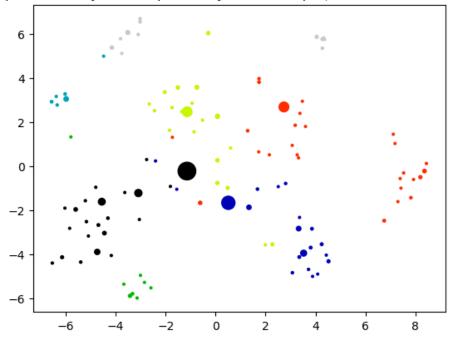
```
TSNE(early exaggeration=12, random state=47, verbose=1)
______
===> Finding 90 nearest neighbors using exact search using euclidean distance...
    --> Time elapsed: 0.09 seconds
===> Calculating affinity matrix...
    --> Time elapsed: 0.00 seconds
===> Calculating PCA-based initialization...
    --> Time elapsed: 0.00 seconds
===> Running optimization with exaggeration=12.00, lr=9.67 for 250 iterations...
             50, KL divergence -1.0225, 50 iterations in 0.0429 sec
Iteration 100, KL divergence 1.2483, 50 iterations in 0.0108 sec
Iteration 150, KL divergence 1.2483, 50 iterations in 0.0100 sec
Iteration 200, KL divergence 1.2483, 50 iterations in 0.0100 sec
Iteration 250, KL divergence 1.2483, 50 iterations in 0.0099 sec
    --> Time elapsed: 0.08 seconds
===> Running optimization with exaggeration=1.00, lr=116.00 for 500 iterations...
            50, KL divergence 0.2085, 50 iterations in 0.0391 sec
Iteration 100, KL divergence 0.1648, 50 iterations in 0.0466 sec
Iteration 150, KL divergence 0.1603, 50 iterations in 0.0466 sec
Iteration 200, KL divergence 0.1602, 50 iterations in 0.0452 sec
Iteration 250, KL divergence 0.1603, 50 iterations in 0.0534 sec
Iteration 300, KL divergence 0.1601, 50 iterations in 0.0451 sec
Iteration 350, KL divergence 0.1602, 50 iterations in 0.0443 sec
Iteration 400, KL divergence 0.1600, 50 iterations in 0.0441 sec
Iteration 450, KL divergence 0.1601, 50 iterations in 0.0440 sec
Iteration 500, KL divergence 0.1601, 50 iterations in 0.0442 sec
   --> Time elapsed: 0.45 seconds
(116, 2)
                                        artifact communityId centrality
                                 axon-disruptor-4.11.0
   org.axonframework.disruptor.commandhandling
                                                     0 0.012156 -5.092271 -3.171525
       org. ax on framework. modelling. command\\
                                 axon-modelling-4.11.0
                                                     0 0.033013 -4.451940 -3.038192
2 org.axonframework.modelling.command.inspection
                                 axon-modelling-4.11.0
                                                     0 0.019850 -4.692068 -2.670332
3 org.axonframework.modelling.command.legacyjpa
                                 axon-modelling-4.11.0
                                                     0 0.011941 -5.843010 -2.820116
```

## 1.3 Visualization of the node embeddings reduced to two dimensions

0 0.017131 -4.312332 -2.358277

org.axonframework.eventsourcing axon-eventsourcing-4.11.0

Java Package positioned by their dependency relationships (FastRP node embeddings + t-SNE)



#### 1.4 Node Embeddings for Java Packages using HashGNN

HashGNN resembles Graph Neural Networks (GNN) but does not include a model or require training. It combines ideas of GNNs and fast randomized algorithms. For more details see HashGNN. Here, the latter 3 steps are combined into one for HashGNN.

Received notification from DBMS server: {severity: WARNING} {code: Neo.ClientNotifica tion.Statement.UnknownRelationshipTypeWarning} {category: UNRECOGNIZED} {title: The p rovided relationship type is not in the database.} {description: One of the relations hip types in your query is not available in the database, make sure you didn't misspe ll it or that the label is available when you run this statement in your application (the missing relationship type is: HAS ROOT)} {position: line: 9, column: 44, offset: 696} for query: '// Query already calculated and written node embeddings on nodes wit h label in parameter \$dependencies projection node including a communityId and centra lity. Variables: dependencies projection node, dependencies projection write propert y. Requires "Add file name and extension.cypher".\n \n MATCH (codeUnit)\n WHERE \$de pendencies projection node IN LABELS(codeUnit)\n AND codeUnit[\$dependencies projec tion write property] IS NOT NULL\n // AND codeUnit.notExistingToForceRecalculation IS NOT NULL // uncomment this line to force recalculation\n OPTIONAL MATCH (artifact: WITH \*, artifact.name AS artifactName\n 0 Java:Artifact)-[:CONTAINS]->(codeUnit)\n PTIONAL MATCH (projectRoot:Directory)<-[:HAS ROOT]-(proj:TS:Project)-[:CONTAINS]->(co WITH \*, last(split(projectRoot.absoluteFileName, \'/\')) AS projectName \n RETURN DISTINCT \n coalesce(codeUnit.fgn, codeUnit.globalFgn, codeUnit.fi leName, codeUnit.signature, codeUnit.name) AS codeUnitName\n ,codeUnit.name AS shortCodeUnitName\n ,coalesce(artifactName, projectName) AS projectName\n ,coalesce(codeUnit.communityLeidenId, 0) AS communi AS centrality\n , coalesce(codeUnit.centralityPageRank, 0.01) ,codeUnit[\$dependencies projection write property] AS embedding\n ORDER BY communit yId'

The results have been provided by the query filename: ../cypher/Node\_Embeddings/Node\_ Embeddings 0a Query Calculated.cypher

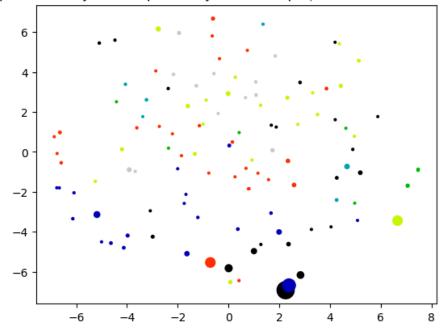
	codeUnitName	shortCodeUnitName	projectName	communityId	centrality	embedding
0	org.axonframework.disruptor.commandhandling	commandhandling	axon-disruptor- 4.11.0	0	0.012156	[-0.6495190411806107, -0.6495190411806107, -1
1	org.axonframework.modelling.command	command	axon-modelling- 4.11.0	0	0.033013	[0.4330126941204071, -1.5155444294214249, 0.21
2	org. ax on framework. modelling. command. in spection	inspection	axon-modelling- 4.11.0	0	0.019850	[-0.8660253882408142, -1.5155444294214249, 0.2
3	org.axonframework.modelling.command.legacyjpa	legacyjpa	axon-modelling- 4.11.0	0	0.011941	[-0.6495190411806107, -1.5155444294214249, 0.4
4	org.axonframework.eventsourcing	eventsourcing	axon- eventsourcing- 4.11.0	0	0.017131	[-1.0825317353010178, -1.948557123541832, -1.0
-	 SNE(early_exaggeration=12,	random state=	 =47, verbos	se=1)		
-	===> Finding 90 nearest neig	hhore using a	evact sear	 Sh usina	euclide	an distance
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=	===> Calculating affinity ma					
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]	Iteration 250, KL divergenc	e 0.6261, 50	iterations	in 0.05	08 sec	
]	Iteration 300, KL divergenc	e 0.6262, 50	iterations	in 0.05	01 sec	
]	Iteration 350, KL divergenc					
]	Iteration 400, KL divergenc					
]	Iteration 450, KL divergenc	e 0.6261, 50	iterations	in 0.05	05 sec	
]	Iteration 500, KL divergenc	e 0.6262, 50	iterations	in 0.05	08 sec	
	> Time elapsed: 0.49 se	conds				
(	(116, 2)					
_	codeUnit	artif	act communitylo	d centrality	х	у
0	org.axonframework.disruptor.commandhandling	axon-disruptor-4.1	11.0	0.012156	1.881062 1.	245779
1	org.axonframework.modelling.command	axon-modelling-4.1			5.195499 -1.	
2	org.axonframework.modelling.command.inspection	axon-modelling-4.1			4.268606 -1.	
3	org.axonframework.modelling.command.legacyjpa	axon-modelling-4.1	11.0	0.011941	5.885445 1.	768883

org.axonframework.eventsourcing axon-eventsourcing-4.11.0

4

0 0.017131 2.820696 3.477451

Java Package positioned by their dependency relationships (HashGNN node embeddings + t-SNE)



#### 2.5 Node Embeddings for Java Packages using node2vec

Received notification from DBMS server: {severity: WARNING} {code: Neo.ClientNotifica tion.Statement.UnknownRelationshipTypeWarning} {category: UNRECOGNIZED} {title: The p rovided relationship type is not in the database.} {description: One of the relations hip types in your query is not available in the database, make sure you didn't misspe ll it or that the label is available when you run this statement in your application (the missing relationship type is: HAS ROOT)} {position: line: 9, column: 44, offset: 696} for query: '// Query already calculated and written node embeddings on nodes wit h label in parameter \$dependencies projection node including a communityId and centra lity. Variables: dependencies projection node, dependencies projection write propert y. Requires "Add\_file\_name and\_extension.cypher".\n \n MATCH (codeUnit)\n WHERE \$de pendencies projection node IN LABELS(codeUnit)\n AND codeUnit[\$dependencies projec tion write property] IS NOT NULL\n // AND codeUnit.notExistingToForceRecalculation IS NOT NULL // uncomment this line to force recalculation\n OPTIONAL MATCH (artifact: Java:Artifact)-[:CONTAINS]->(codeUnit)\n WITH \*, artifact.name AS artifactName\n 0 PTIONAL MATCH (projectRoot:Directory)<-[:HAS\_ROOT]-(proj:TS:Project)-[:CONTAINS]->(co WITH \*, last(split(projectRoot.absoluteFileName, \'/\')) AS projectName deUnit)\n coalesce(codeUnit.fgn, codeUnit.globalFgn, codeUnit.fi \n RETURN DISTINCT \n leName, codeUnit.signature, codeUnit.name) AS codeUnitName\n ,codeUnit.name AS shortCodeUnitName\n ,coalesce(artifactName, projectName) , coalesce(codeUnit.communityLeidenId, 0) AS projectName\n AS communi , coalesce(codeUnit.centralityPageRank, 0.01) AS centrality\n tyId\n ,codeUnit[\$dependencies projection write property] AS embedding\n ORDER BY communit

The results have been provided by the query filename: ../cypher/Node\_Embeddings/Node\_ Embeddings\_0a\_Query\_Calculated.cypher

	codeUnitName	shortCodeUnitName	projectName	communityId	centrality	embedding		
0	org.axonframework.disruptor.commandhandling	commandhandling	axon-disruptor- 4.11.0	0	0.012156	[0.24934718012809753, 0.05292239785194397, 0.3		
1	org.axonframework.modelling.command	command	axon-modelling- 4.11.0	0	0.033013	[0.1892593801021576, 0.075283944606781, 0.2512		
2	org. axon framework. modelling. command. in spection	inspection	axon-modelling- 4.11.0	0	0.019850	[0.4492984116077423, -0.08181710541248322, 0.1		
3	org.axonframework.modelling.command.legacyjpa	legacyjpa	axon-modelling- 4.11.0	0	0.011941	[0.3352530002593994, -0.054873015731573105, 0		
4	org.axonframework.eventsourcing	eventsourcing	axon- eventsourcing- 4.11.0	0	0.017131	[0.18305234611034393, -0.053913407027721405, 0		
- ]	TSNE(early_exaggeration=12, random_state=47, verbose=1)							
= = = = = = = = = = = = = = = = = = =	===> Finding 90 nearest neighbors using exact search using euclidean distance> Time elapsed: 0.00 seconds ==>> Calculating affinity matrix> Time elapsed: 0.00 seconds ==>> Calculating PCA-based initialization> Time elapsed: 0.00 seconds ===> Running optimization with exaggeration=12.00, lr=9.67 for 250 iterations  Iteration 50, KL divergence -0.4867, 50 iterations in 0.0471 sec  Iteration 100, KL divergence 1.1778, 50 iterations in 0.0120 sec  Iteration 150, KL divergence 1.1778, 50 iterations in 0.0101 sec  Iteration 200, KL divergence 1.1778, 50 iterations in 0.0101 sec  Iteration 250, KL divergence 1.1778, 50 iterations in 0.0100 sec> Time elapsed: 0.09 seconds ===> Running optimization with exaggeration=1.00, lr=116.00 for 500 iterations  Iteration 50, KL divergence 0.3508, 50 iterations in 0.0400 sec  Iteration 150, KL divergence 0.3409, 50 iterations in 0.0516 sec  Iteration 150, KL divergence 0.3390, 50 iterations in 0.0514 sec  Iteration 200, KL divergence 0.3375, 50 iterations in 0.0511 sec  Iteration 250, KL divergence 0.3357, 50 iterations in 0.0510 sec  Iteration 300, KL divergence 0.3357, 50 iterations in 0.0508 sec  Iteration 350, KL divergence 0.3356, 50 iterations in 0.0511 sec							
	Iteration 400, KL divergenc							
	Iteration 450, KL divergenc							
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0	org.axonframework.disruptor.commandhandling	axon-disruptor-4.1	1.0	0 0.012156 2	2.167051 0	.360958		
1	org.axonframework.modelling.command	axon-modelling-4.1	.1.0	0 0.033013 2	2.885397 -0	.596777		
2	org. ax on framework. modelling. command. in spection	axon-modelling-4.1	1.0	0 0.019850 3	3.875641 -0	.630526		

**3** org.axonframework.modelling.command.legacyjpa axon-modelling-4.11.0

org.axonframework.eventsourcing axon-eventsourcing-4.11.0

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0 0.011941 4.538305 1.458984

0 0.017131 2.335855 -0.509846

Java Package positioned by their dependency relationships (node2vec node embeddings + t-SNE)

