

# GDS (Neo4j Graph Data Science)

## GDS Algorithms

### variants

- Named graph variant
  - The graph to operate over will be read from the graph catalog.
- Anonymous graph variant
  - The graph to operate over will be created and deleted as part of the algorithm execution

Each syntax variant additionally provides different execution modes. These are the supported execution modes:

- stream
  - Returns the results of the algorithm as a stream of records
- stats
  - Returns a single record of summary statistics, but does not write to the Neo4j database
- mutate
  - Writes the results of the algorithm to the in-memory graph and returns a single record of summary statistics. This mode is designed for the named graph variant, as its effects will be invisible on an anonymous graph.
- write
  - Writes the results of the algorithm to the Neo4j database and returns a single record of summary statistics.

### Centrality algorithms

- Production-quality
  - [Page Rank](#)
  - [Article Rank](#)
  - [Eigenvector Centrality](#)
  - [Betweenness Centrality](#)
  - [Degree Centrality](#)
- Alpha
  - [Closeness Centrality](#)
  - [Harmonic Centrality](#)
  - [HITS](#)
  - [Influence Maximization](#)

### Community detection algorithms

Community detection algorithms are used to evaluate how groups of nodes are clustered or partitioned, as well as their tendency to strengthen or break apart.

- Production-quality
  - [Louvain](#)
  - [Label Propagation](#)
  - [Weakly Connected Components](#)
  - [Triangle Count](#)
  - [Local Clustering Coefficient](#)

## Similarity algorithms

Similarity algorithms compute the similarity of pairs of nodes using different vector-based metrics.

- Production-quality
  - [Node Similarity](#)
- Beta
  - [K-Nearest Neighbors](#)
- Alpha
  - [Approximate Nearest Neighbors](#)
  - [Cosine Similarity](#)
  - [Euclidean Similarity](#)
  - [Jaccard Similarity](#)
  - [Overlap Similarity](#)
  - [Pearson Similarity](#)

## Path finding algorithms

- Production-quality
  - [Dijkstra Source-Target](#)
  - [Dijkstra Single-Source](#)
  - [A\\*](#)
  - [Yen's algorithm](#)

## Link Prediction algorithms

- Alpha
  - [Adamic Adar](#)
  - [Common Neighbors](#)
  - [Preferential Attachment](#)
  - [Resource Allocation](#)
  - [Same Community](#)
  - [Total Neighbors](#)

## Node embeddings

Node embedding algorithms compute low-dimensional vector representations of nodes in a graph. These vector, also called embeddings, can be used for machine learning.

- Production-quality
  - [FastRP](#)
- Beta
  - [GraphSAGE](#)
  - [Node2Vec](#)

## Machine Learning Models

The machine learning procedures in Neo4j [GDS](#) allow you to train supervised machine learning models. Models can then be accessed via the [Model Catalog](#) and used to make predictions about your graph.

To help with working with ML models, these are additional guides for pre-processing and hyperparameter tuning available in:

- [Pre-processing](#)
  - [Tuning parameters for training](#)
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- Alpha
  - [Node Classification](#)
  - [Link Prediction](#)

## Auxiliary procedures

Auxiliary procedures are extra tools that can be useful in your workflow. The Neo4j GDS library includes the following auxiliary procedures, grouped by quality tier:

- Beta
    - [Graph Generation](#)
  - Alpha
    - [Collapse Path](#)
    - [Scale Properties](#)
    - [One Hot Encoding](#)
    - [Split Relationships](#)
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- [Pregel API](#)
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### Plugin Backlinks:

From:

<https://jace.link/> - **Various Ways**

Permanent link:

<https://jace.link/open/gds>

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