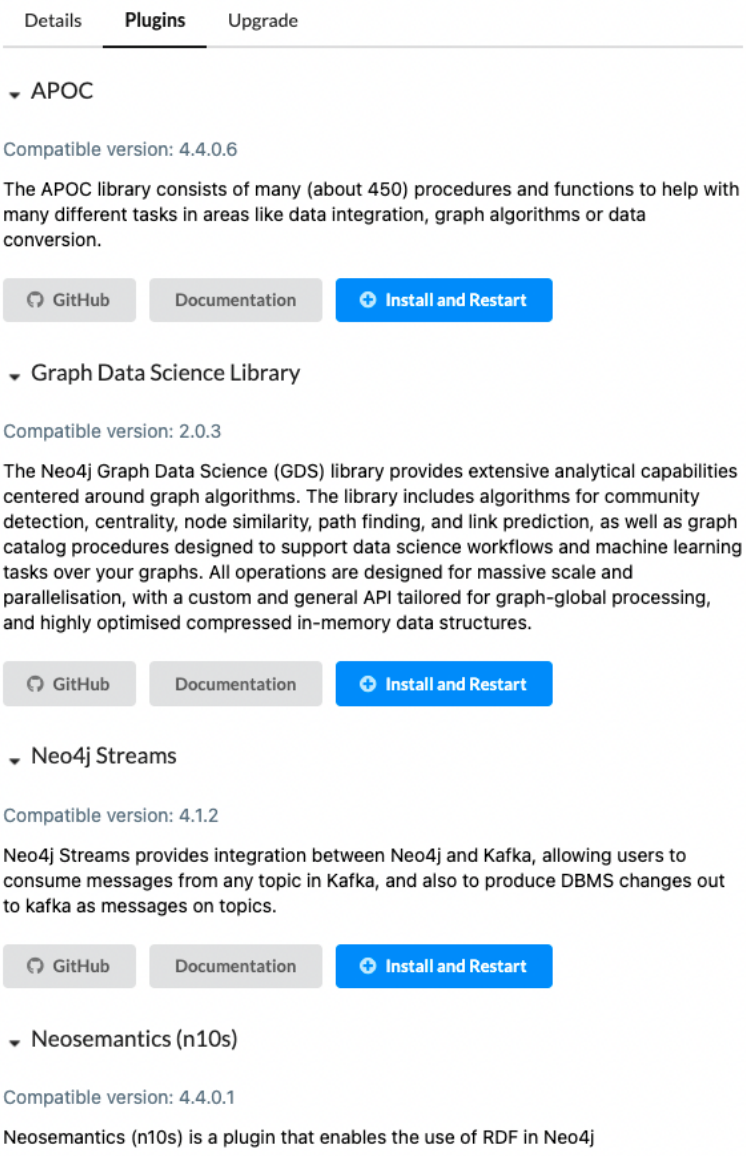
GDS 簡單介紹



若使用 desktop 可放在

- 基本結構

```
CALL gds[.<tier>].<algorithm>.<execution-mode>[.<estimate>](
  graphName: String,
  configuration: Map
)
```

- Execution mode
  - stream
    - 標準的查找模式，也是 cypher 的預設執行模式
    - 不擅長處理大數據，因此通常搭配使用 “top N-style” 的模式執行
  - stats
    - 將演算法的結果統計輸出（次數 or 百分比）
    - 不會對任何的東西做更改
    - 是 mutate 和 write 的基礎
  - mutate （變異）
    - 將統計結果寫回 projected graph （也就是被計算的 graph）
    - 可將複數的演算法寫在一起，無需再額外寫入 projected
      - 某些演算法有互相依賴的問題
      - 可直接將結果 轉換成 cypher （原本須額外的轉換 gds.util.nodeproperty）
  - write
    - 與 mutate 相似，可將統計結果直接寫進去 DB
      - projected graph 是被存在 memory 的 graph
- 使用 estimate 來估計運算資源！

> If the estimation shows that there is a very high probability of the execution going over its memory limitations, the execution is prohibited

Memory Estimation - Neo4j Graph Data Science

The graph algorithms library operates completely on the heap, which means we'll need to configure our Neo4j Server with a much larger heap size than we would for transactional

https://neo4j.com/docs/graph-data-science/current/common-usage/memory-estimation/#estimate-heap-control

The #1 Database for Connected Data

- 關於 graph catalog
  - catalog 是 將圖給提取出來放置記憶體中進行運算或研究（開發使用）
  - 延伸問題：如果有這個，還會需要radis 嗎？（開發的時候）

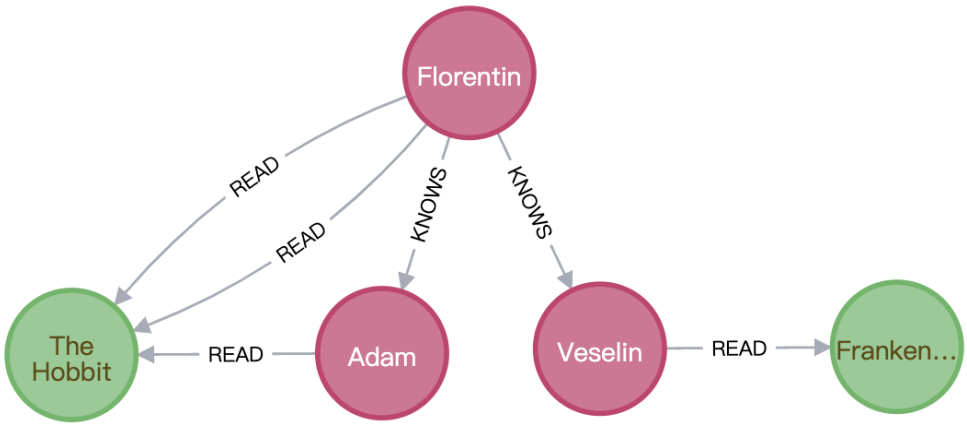
Graph Catalog - Neo4j Graph Data Science

Graph algorithms run on a graph data model which is a projection of the Neo4j property graph data model. A graph projection can be seen as a materialized view over the stored

https://neo4j.com/docs/graph-data-science/current/management-ops/graph-catalog-ops/

The #1 Database for Connected Data

常用的指令可以查看上面的網頁



- 加入 graph catalog 需備三個基本元素
  - graph name
  - node
  - relation
- 加入的方法可以分為三種
  - native projection （預設）

```
CALL gds.graph.project(
  'graphWithProperties',                                  ①
  {                                                       ②
    Person: {properties: 'age'},                          ③
    Book: {properties: {price: {defaultValue: 5.0}}}      ④
  },
  ['KNOWS', 'READ'],                                      ⑤
  {nodeProperties: 'ratings'}                             ⑥
)
YIELD
  graphName, nodeProjection, nodeCount AS nodes, relationshipCount AS rels
RETURN graphName, nodeProjection.Book AS bookProjection, nodes, rels
```

Table 9. Results

graphName	bookProjection	nodes	rels
"graphWithProperties"	`{label=Book, properties={price={defaultValue=5.0, property=price}, ratings={defaultValue=null, property=ratings}}}`	`5`	`6`

  - cypher

```
CALL gds.graph.project.cypher(
  'graphWithProperties',
  'MATCH (n)
   WHERE n:Book OR n:Person
   RETURN
    id(n) AS id,
    labels(n) AS labels,
    coalesce(n.age, 18) AS age,
    coalesce(n.price, 5.0) AS price,
    n.ratings AS ratings',
  'MATCH (n)-[r:KNOWS|READ]->(m) RETURN id(n) AS source, id(m) AS target, type(r) AS type'
)
YIELD
  graphName, nodeCount AS nodes, relationshipCount AS rels
RETURN graphName, nodes, rels
```

使用 cypher 找到需要的 node 與 relations，並將點與關係轉換為 id 後回傳

node: 須轉換 id、對應的 label，其餘的會作為 properties

relation: id、realation type，其餘的會作為 properties

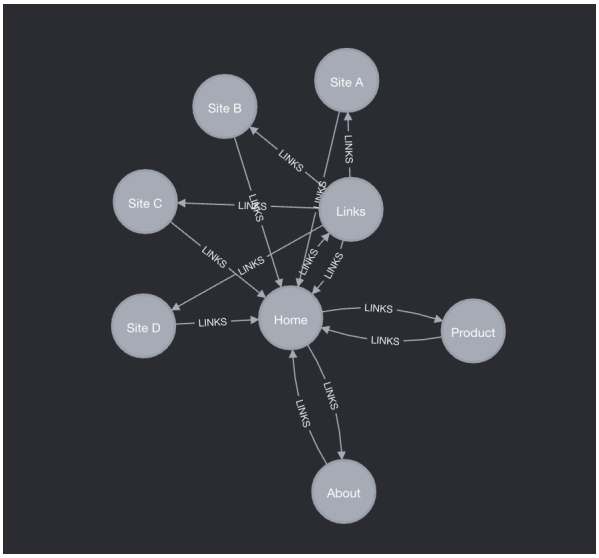
- aggregation
  - 傳入指定的點
  - 還要研究一下
- project graph 是 user define 的，當你切換尸

algorithms 分類

PageRank

PageRank is introduced in the original Google paper as a function that solves the following equation:

$$PR(A) = (1-d) + d(\frac{PR(T_1)}{C(T_1)} + ... + \frac{PR(T_n)}{C(T_n)})$$




```
CALL gds.pageRank.stream(
  graphName: String,
  configuration: Map
)
YIELD
  nodeId: Integer,
  score: Float
```

```
CALL gds.graph.project(
  'myGraph',
  'Page',
  'LINKS',
  {
    relationshipProperties: 'weight'
  }
)
```

Estimate

```
CALL gds.pageRank.write.estimate('myGraph', {
  writeProperty: 'pageRank',
  maxIterations: 20,
  dampingFactor: 0.85
})
YIELD nodeCount, relationshipCount, bytesMin, bytesMax, requiredMemory
```

Table 13. Results

nodeCount	relationshipCount	bytesMin	bytesMax	requiredMemory
8	14	696	696	"696 Bytes"

Stream output

```
CALL gds.pageRank.stream('myGraph')
YIELD nodeId, score
RETURN gds.util.asNode(nodeId).name AS name, score
ORDER BY score DESC, name ASC
```

Table 14. Results

name	score
"Home"	3.215681999884452
"About"	1.0542700552146722
"Links"	1.0542700552146722
"Product"	1.0542700552146722
"Site A"	0.3278578964488539
"Site B"	0.3278578964488539
"Site C"	0.3278578964488539
"Site D"	0.3278578964488539