## Chapter 1

## Graph database

The DBSM chosen for the graph database is Neo4j, it was use to manege the social part of the site and also to keep track of the reviews made for the feed and suggestion functionality.

The entities used in the database are:

- User
- TopCritic
- Movie

The nodes themselves do not store a lot of data, we decided to keep the bare minimum by having the following attributes:

- for the User and TopCritic:
  - -id
  - name
- for Movie:
  - id
  - title

The relationship present are:

- User -[:FOLLOWS]-> TopCritic
- User -[:REVIEWED]-> Movie
- TopCritic -[:REVIEWED]-> Movie

In particular the REVIEWED relationship contains the following information:

- content
- date
- $\bullet$  freshness

Below is present a snapshot of the graph database taken from Neo4j.

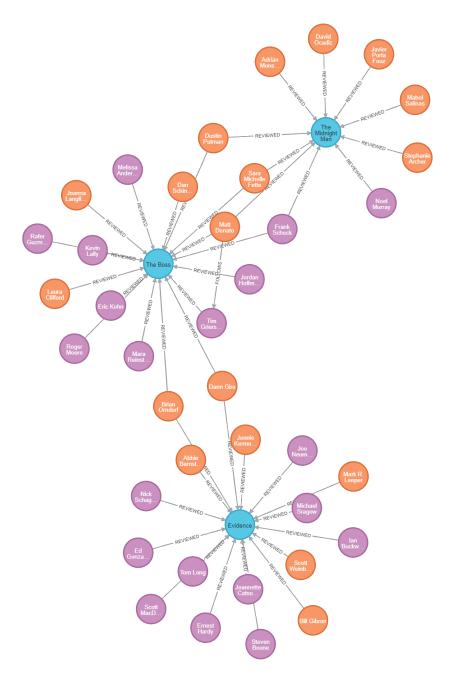


Figure 1.1: graph DB

As previously stated in the Feasibility study section 2, the graph database design was heavily influenced by the document one, it was in fact generated by python script (Appendix A .6) that starts from the movie collection and, after generating the nodes for the Movie entity, does the same for the user by dividing them between normal user and top critic. It then generates the REVIEWED relationship based upon the data stored in the document database and finally it generates the FOLLOWS relationship randomly

## 1 Index

The id attributes is common to all nodes and it stores a string which is also used as id by the document database in the  $\_id$  field, this choice was made so that we could have a way to identify the same object across the different databases with only one string from the application. Due to the OOP language used to build the latter the data was manipulated with classes containing the id of the object so, to improved the performance of the graph database we decided to use the id attribute as a index.

Below the information in a json format from the Neo4j DBSM about the new index (some attributes were omitted for space)

```
1
 2
      {
 3
        "id": 3,
        "name": "movie_id_index",
 4
        "type": "RANGE",
 5
        "entityType": "NODE",
 6
 7
        "labelsOrTypes": [
          "Movie"
 8
 9
10
        "properties": [
          "id"
11
12
13
     },
14
        "id": 5,
15
        "name": "topcritic_id_index",
16
17
        "type": "RANGE",
        "entityType": "NODE",
18
19
        "labelsOrTypes": [
20
          "TopCritic"
21
22
        "properties": [
          "id"
23
24
25
     },
26
27
        "id": 4,
        "name": "user_id_index",
28
        "type": "RANGE",
29
        "entityType": "NODE",
30
31
        "labelsOrTypes": [
32
          "User"
33
34
        "properties": [
          "id"
35
36
37
38
```

We then proceed to confront the profile of the queries with and without the above indexes, here we can see the result of this analysis:

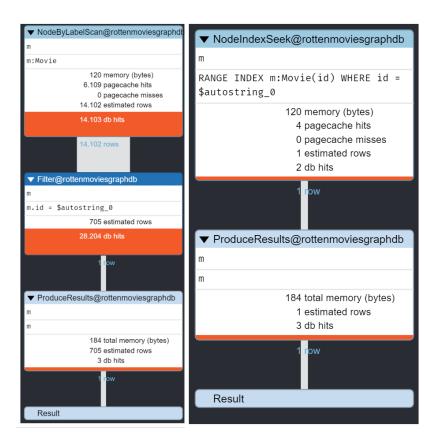


Figure 1.2: search by id without and with movie\_id \_index

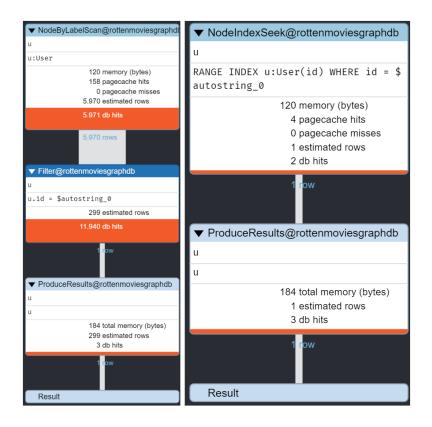


Figure 1.3: search by id without and with user\_id \_index

## 2 Queries

In the following section we will show the queries that drove the design of the graph database.

Graph-Centric Query	Domain-Specific Query
Which are the User nodes with the most out-	Which are the non top critic user with most
going edges of type REVIEWED	reviews made
Which are the TopCritic nodes with the most	Which are the most followed top critics
incoming edges of type FOLLOWS	
Which are the Movies nodes that have	Which are the latest reviewed movies by a
been most recently connected with TopCritic	followed top critics given a starting user
nodes by a REVIEWED relationship mean-	
while the latter are connected to a User node	
with a FOLLOWS relationship	
Which are the User and TopCritic nodes	Which are the non-followed top critics with
that have outgoing REVIEWED edges to the	the most affinity towards a given user
same Movie node without having a FOL-	
LOWS edge between them, which of these	
TopCritic nodes have more similar attribute	
on the REVIEWED relationship with the	
one on the same type going towards the same	
Movie node given a User node	
Which Movie nodes have different ratio of	Which movies have been targeted by review
freshness in the incoming REVIEWED edges	bombing in the last x month
spitted in base of their date attribute	