

Large-Scale and Multi-Structured Databases

# Project Documentation

## GameCritic

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# Contents

# Chapter 1

## Introduction

- Allow users to retrieve various informations on videogames
- Allow users to review videogames and give them a score
- Allow users to follow other users, like and comment other users' reviews
- Give users recommendationson on who to follow and what games to play
- Allow videogame company managers to retrieve analytics on the sentiment of the reviewers for a particular game
- Allor the application manager to see which users are the most active on the site, so that he can promote them as the best reviewers

## Chapter 2

# Dataset and Web Scraping

- **Variety** : we exploited two real different sources, having different formats
- **Velocity/Variability** :
  - Comments are frequently added/liked/responded to
  - Most active users change periodically

### 2.1 Raw Data

Data was retrieved from:

- MobyGames
- MetaCritic
- generated data

### 2.2 Scraping Algorithms

We used *Python* for scraping and data generation

#### 2.2.1 Videogames Scraping

Data is scraped from *MobyGames* for videogames info

#### 2.2.2 Users and Review Scraping

Some data is scraped from *Metacritic* for reviews, while other reviews are generated. Comment on reviews are generated using algorithms

## 2.3 Resulting Dataset

The final volume wanders around 400 MB, which are the result of:

- 70k videogames
- 250k reviews
- 625k comments

# Chapter 3

## Design

### 3.1 Actors

There are three main actors:

- Reviewer (User)
- Company Manager (Super User)
- Administrator The *Reviewer* is the end-user of the application, who is able to search for *videogames*, review them, comment on other users' reviews, etc The *Company Manager* is an entity that represents the owner

### 3.2 Requirements

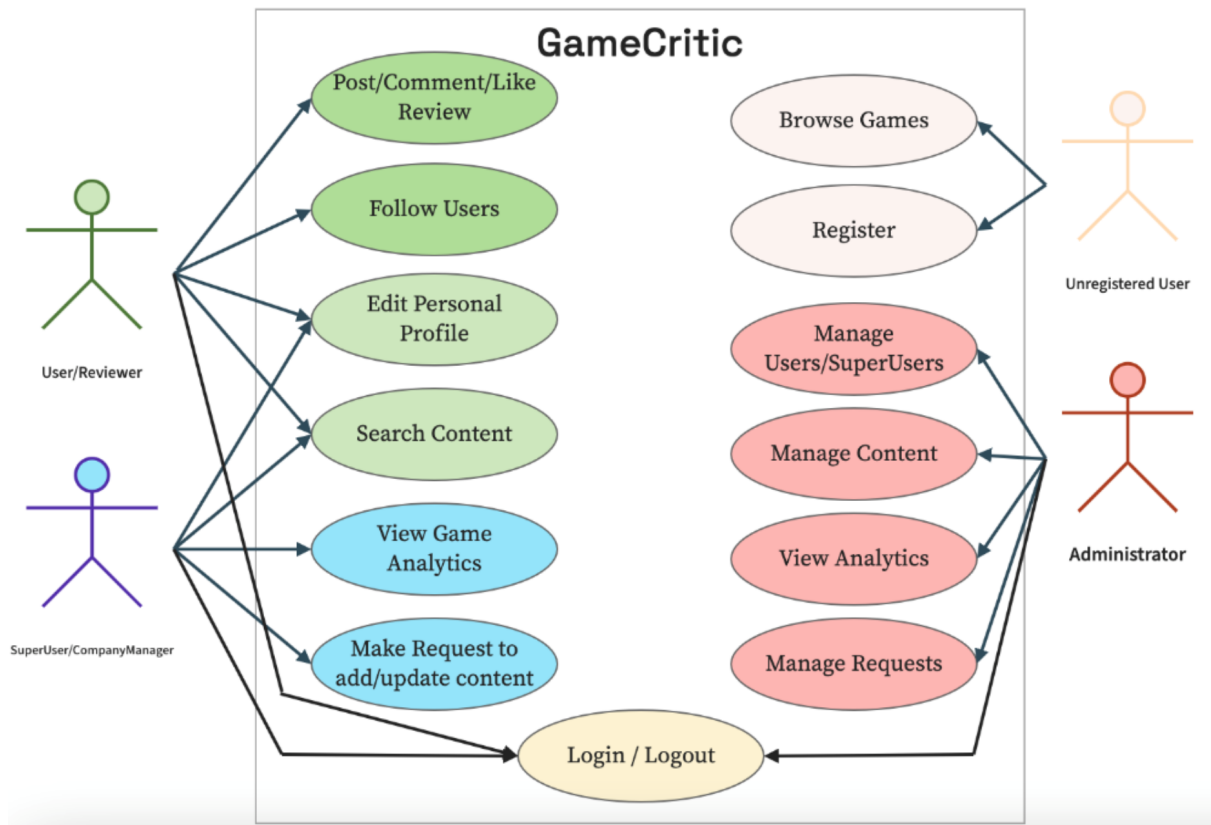
#### 3.2.1 Functional Requirements

**Handled by MongoDB**

- Average review score per game producer
- Most active Users (looking at the number of posted reviews)
- Most discussed reviews
- Most popular games in the latest week/month/year

**Handled by Neo4j**

- New friend recommendations
- Videogame recommendations



**Figure 3.1:** Actors and main supported functionalities

- identify the most influential users

### 3.2.2 Non-Functional Requirements

## 3.3 Use Case Diagram

## 3.4 UML Class Diagram

## 3.5 Data Models and Implementation

### 3.5.1 Document DB Collections

- Videogames
- Users
- Reviews
- Comments



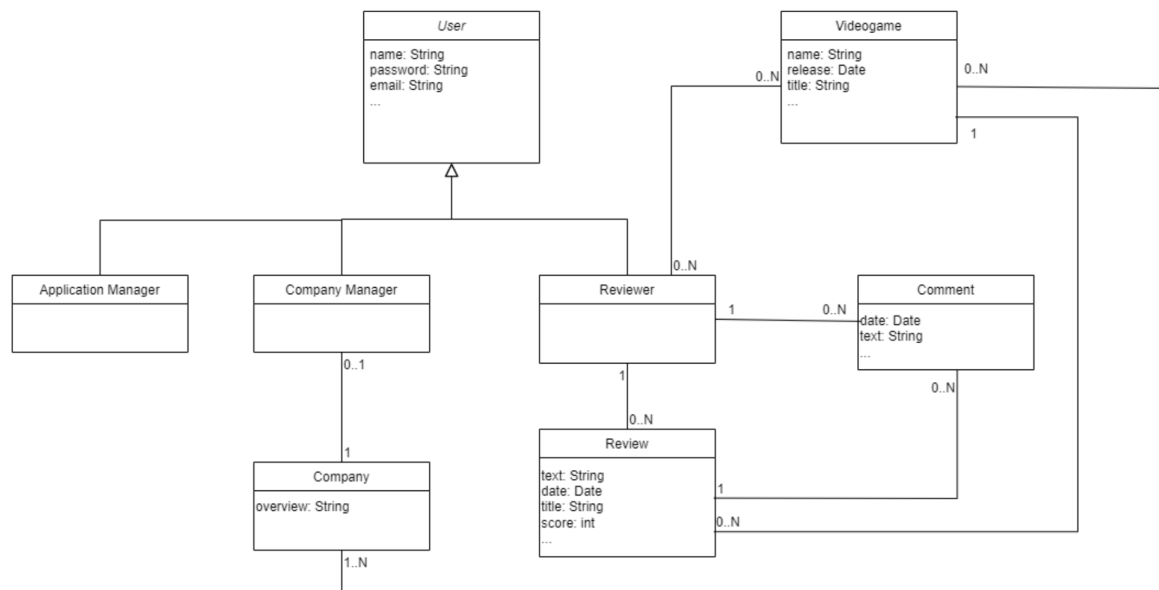


Figure 3.2: UML Class Diagram

- userImages

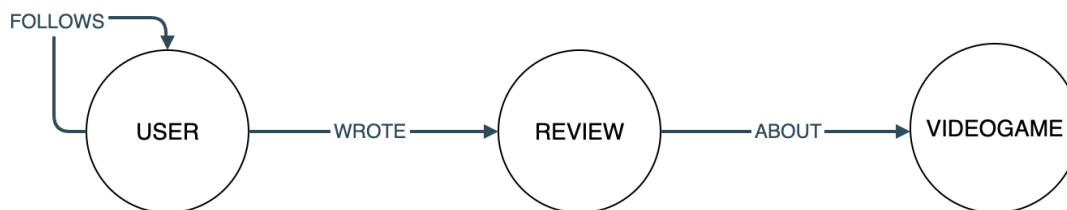
### 3.5.2 Document DB Relationships

### 3.5.3 Document DB Examples

```

1  {
2  "Name": "(Almost) Total Mayhem",
3  "Released": "January 14th, 2011 on Xbox 360",
4  "Publishers": "Peanut Gallery",
5  "Developers": "Peanut Gallery",
6  "Genre": "Action",
7  "Perspective": "Side view",
8  "Gameplay": "Platform",
9  "Setting": "Fantasy",
10 "Media Type": "Download",
11 "Multiplayer Options": "Same/Split-Screen",
12 "Number of Offline Players": "1-2 Players",
13 "Description": "description",
14 "user_review": 6.0,
15 "reviews": [
16 {
17   "score": 8,
18   "quote": "stunning graphhcs!",

```



**Figure 3.3:** Entities handled by *GraphDB* and their relationships

```

19   "author": "Yusoreqa",
20   "date": "2023-10-30",
21   "source": "random",
22   "comments": [
23     {
24       "author": "Hojosu",
25       "quote": "I see that Yusoreqa agrees with me.",
26       "date": "2023-12-06"
27     }
28   ]
29 }
30 ]
31 }

```

### 3.5.4 Graph DB - Neo4j

We handled the following entities via GraphDB (figure ??) :

- Users
- Videogames
- Reviews

## 3.6 Distributed Database Design

Eventual consistency is sufficient, enforcing strong consistency would be too costly.

### 3.6.1 Replicas

We were given the chance to exploit a cluster of three nodes for the project. We deployed one *MongoDB replica* for each node (replicas were not implemented for *Neo4j* since to do

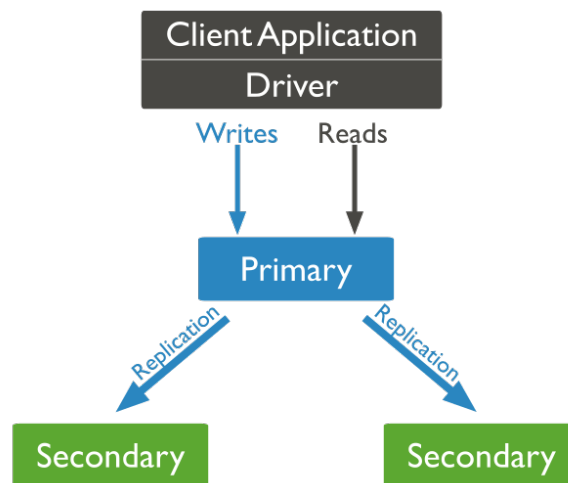
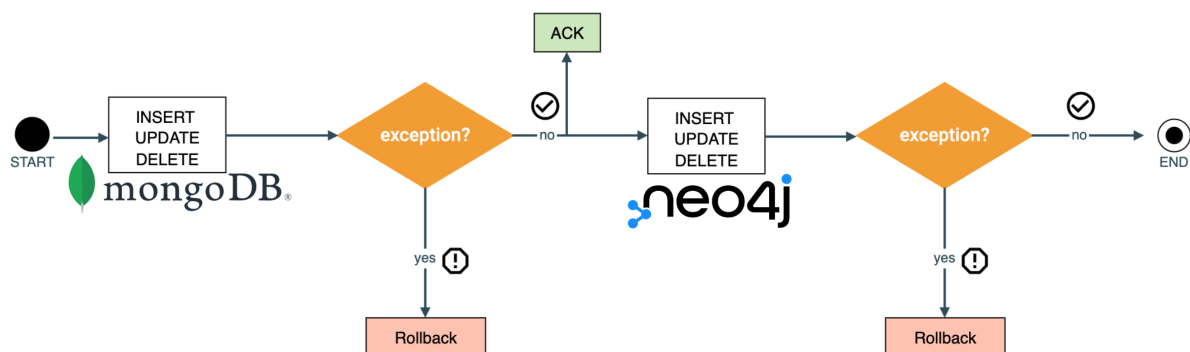


Figure 3.4: Replica Set

Figure 3.5: Handling consistency between *MongoDB* and *Neo4j*

so we would have needed the *Enterprise edition*). The *primary* is the only member in the replica set that receives write operations (figure ??). MongoDB applies write operations on the primary and then records the operations on the primary's *oplog*. *Secondary* members replicate this log and apply the operations to their data sets.

## Write Concern

### 3.6.2 Sharding

*mongos* instances will pass the *write concern* on to the shards

### 3.6.3 Handling inter-databases consistency

Having to deal with two different architectures (*DocumentDB* and *GraphDB*), we need to face the problem of redundancy and handle the consistency of data that are stored

within both databases. Any rollback triggered by a failure of an insert, an update or a delete operation, can lead to inconsistencies. To address this problem, whenever a write operation is performed successfully on *MongoDB*, an ACK (*acknowledgement*) is sent to the user. Data on *GraphDB* will be eventually consistent: if any exception occurs when the updates over data on *Neo4j* are attempted, than a rollback operation will be executed in order to bring back both the databases in a consistent state (just as shown in figure ??).

### Eventual Consistency

Eventual consistency is dealt with by exploiting the *asynchronous execution support* in *Spring* and the `@Async` annotation. The caller will not have to wait for the complete execution of the called method: annotating a method of a bean with `@Async` will make it execute in a separate thread.

# Chapter 4

## Implementation

Application:

- *Java with Maven*
- *SpringBoot*
- *Javascript, HTML, CSS*

### 4.1 Spring Boot

### 4.2 Java Entities

#### 4.2.1 Users

Simple User - Reviewer

Company Manager

Admin

#### 4.2.2 Videogames

#### 4.2.3 Reviews and Comments

# Chapter 5

## Queries

### 5.1 Mongo DB

### 5.2 Neo4j

#### 5.2.1 Get Followers

#### 5.2.2 Get Following

#### 5.2.3 Friends' suggestions

From users' relationships

From common interests

### 5.3 Indexes

#### 5.3.1 MongoDB

MongoDB					
Query	Collection	Index Keys	Keys examined	Documents examined	ms
Query1	videogames	-	numKeys	numDoc	<i>ms</i>
		Name	numKeys	numDoc	<i>ms</i>
Query2	reviews	-	numKeys	numDoc	<i>ms</i>
		author	numKeys	numDoc	<i>ms</i>
Query3	reviews	-	numKeys	numDoc	<i>ms</i>
		{author, game}	numKeys	numDoc	<i>ms</i>
Query4	reviews	-	numKeys	numDoc	<i>ms</i>
		game	numKeys	numDoc	<i>ms</i>
Query5	comments	-	numKeys	numDoc	<i>ms</i>
		reviewId	numKeys	numDoc	<i>ms</i>
Query6	companies	-	numKeys	numDoc	<i>ms</i>
		Name	numKeys	numDoc	<i>ms</i>
Query7	users	-	numKeys	numDoc	<i>ms</i>
		username	numKeys	numDoc	<i>ms</i>
Query8	userImages	-	numKeys	numDoc	<i>ms</i>
		username	numKeys	numDoc	<i>ms</i>

### 5.3.2 Neo4j

Neo4j					
Query	Index Name	Nodes' Label	Property	Hits	(ms)
Query1	-	-	-	numHits	<i>ms</i>
	<i>username_index</i>	User	username	numHits	<i>ms</i>
Query2	-	-	-	numHits	<i>ms</i>
	<i>game_index</i>	Game	name	numHits	<i>ms</i>
Query3	-	-	-	numHits	<i>ms</i>
	<i>review_index</i>	Review	reviewId	numHits	<i>ms</i>

## Chapter 6

## Chapter