# Group Project - Neo4j

of Systems and Methods for Big and Unstructured Data Course (SMBUD)

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# Group 78

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### 1 Introduction

The project aims to design and implement a database system to support the management of data related to the film industry. The database will include entities such as Person (actor, director, writer), Title (film, TV series), Episode, ratings, and genre. The goal is to create a comprehensive system that can store and query information about films, TV series, and the people involved in their production. The project will develop with Neo4j, to exploit the relations between the entities and find significant insights from the data, such as collaborations between directors and actors, trend of genres over time, most working actors and the most successful films.

# 2 Assumptions

The project is based on the following assumptions:

- Each person has a unique ID and a name, surname, and date of birth. Some could also have a death date, if passed away.
- Each person can be associated with multiple roles (actor, director, writer, archive footage, music department, producer)
- Each title has a unique identifier and a title type (film, TV series, shortfilm)
- Each title can have multiple episodes
- Each title can have multiple genres
- Each user can rate a title if and only if it has watched the film
- Each title has a unique rating, average rating from the users, and the number of votes
- Each person can be associated with multiple titles
- Each title can have multiple people associated with it

# 3 ER diagram



# POLITECNICO MILANO 1863

Figure 1: E-R Diagram

#### 3.1 Entities

Starting from the considerations previously exposed regarding the implementation hypotheses, we have drawn an ER diagram (**Figure 1**) which includes 5 different entities and 7 many-to-many relationships described below in the logical model:

- **Person**(<u>nconst</u>, PrimaryName, BirthYear, DeathYear, PrimaryProfession, KnownForTitles)
- **Title**(<u>tconst</u>, PrimaryTitle, OriginalTitle, TitleType, StartYear, EndYear, RuntimeMinutes, Genres)
- **Episode**(<u>tconst</u>, ParentTconst, SeasonNumber, EpisodeNumber)
- Ratings(tconst, AverageRating, NumVotes)

#### - Genre(Name)

The **Person** entity describes every possible individual with their own personal data, including their primary profession and titles they are known for. The **Title** entity represents films, TV series or short films with their respective attributes. The **Episode** entity is used to detail episodes of TV series, linked to their parent series. The **Ratings** entity captures the average rating and number of votes for each title. Finally, the **Genre** entity categorizes the titles into different genres.

#### 3.2 Relationships

 $ACTED\_IN (Person) - [: ACTED\_IN] - > (Title)$ 

Relationship between a Person, whose primary profession is actor, and a Title.

**DIRECTED** (Person) - [:DIRECTED] - > (Title)

Relationship between a Person, whose primary profession is director, and a Title.

**WROTE** (Person) - [:WROTE] - > (Title)

Relationship between a Person, whose primary profession is writer, and a Title.

 $PART_OF (Episode) - [: PART_OF] - > (Title)$ 

Relationship between an Episode and its parent Title, whose TitleType is TV series.

 $HAS\_GENRE (Title) - [: HAS\_GENRE] - > (Genre)$ 

Relationship between a Title and a Genre.

 $HAS\_RATING$   $(Title) - [: HAS\_RATING] - > (Rating)$ 

Relationship between a Title and its Rating.

#### 3.3 Constraints:

ciao

## 4 Cypher Queries

#### 4.1 DistributionCenter closest to each User

ORDER BY dist ASC LIMIT 10;

$user\_id$	$nearest\_center$	$\operatorname{dist}$
58210	Chicago IL	1759.9113
48442	Chicago IL	1759.9113
20571	Chicago IL	1759.9113
37767	Philadelphia PA	1897.4164
7543	Philadelphia PA	1897.4164
41572	Philadelphia PA	1897.4164

Table 1: Output della query Neo4j.

## 4.2 Most sold products:

MATCH (:OrderItem)-[:REFERS\_TO]->(p:Product)
RETURN p.name AS product\_name, count(\*) AS sales\_count
ORDER BY sales\_count DESC
LIMIT 10;

Product Name	Quantity
Wrangler Men's Premium Performance Cow-	62
boy Cut Jean	
Puma Men's Socks	48
7 For All Mankind Men's Standard Classic	41
Straight Leg Jean	
True Religion Men's Ricky Straight Jean	37
Kenneth Cole Men's Straight Leg Jean	36
Michael Kors Men's 3 Pack Brief	33
HUGO BOSS Men's Long Pant	31
Lucky Brand Mens Men's 361 Vintage	31
Straight Denim Jean	
Thorlo Unisex Experia Running Sock	31
Diesel Men's Shioner Skinny Straight Leg	31
Jean	

Table 2: Top-selling men's clothing products and their quantities.

## 5 References & Sources

- [1] Course Slides
- [2] https://pysimplegui.readthedocs.io/en/latest/call
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