



# Welcome To Met Connect

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DATASCI 205, Section 1, Group 4*



# Business Case Scenario

We joined the digital innovation team at  
The Met as data scientists!

- 470,000+ artworks
- In-person + online

# Business Vision

Design a data-driven system that...



Powers  
recommendation  
engine



Showcases  
artwork  
connections



Improves web /  
mobile content  
delivery



Generates ideas  
for art exhibits /  
research

# NoSQL Databases We Use



Neo4j

(Graph Data Science Algorithms)



MongoDB



Redis

# Graph Design

## Nodes

- Artist
- Artwork (Paintings only)
- Department
- Time Period

## Edges

### Artwork–Artist

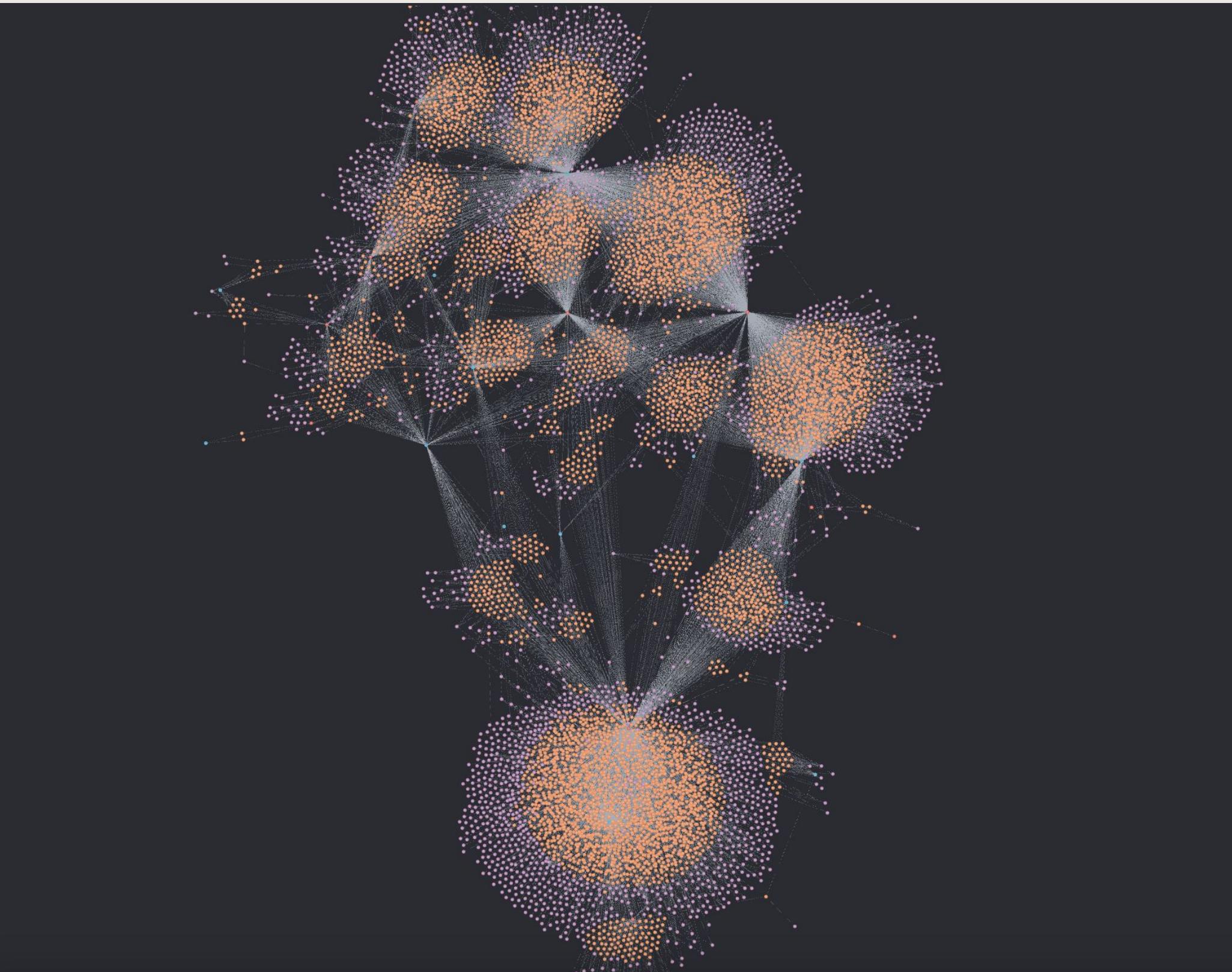
(:Artwork)-[:CREATED\_BY]->(:Artist)

### Artwork–Department

(:Artwork)-[:BELONGS\_TO]->(:Department)

### Artwork–Time Period

(:Artwork)-[:ASSOCIATED\_WITH]->(:TimePeriod)



Nodes: 8607  
Relationships: 17844

# Neo4j: Pagerank

## Goals:

- Identify the most influential artists
- Identify the centuries with the most influential artists

## Result:

- Famous artists such as John Singer Sargent, Claude Monet, and Pablo Picasso had the highest Pagerank scores
- Artists from the 18th century had the highest average Pagerank scores

## Business Case Applications:

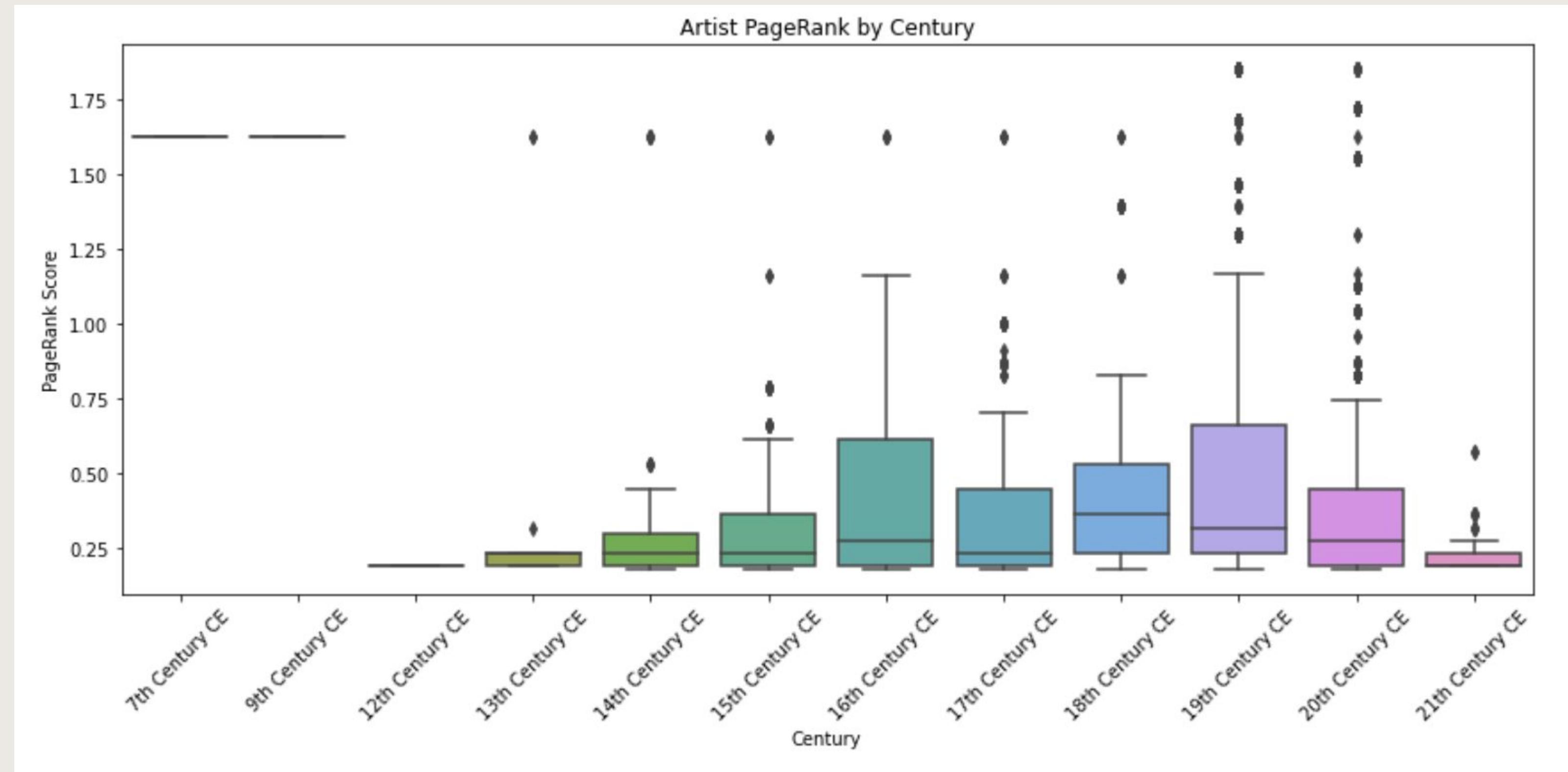
- Identifies current artist rankings to determine future art acquisitions

## Why Not Relational Database:

- Computationally intensive for a relational database



# Neo4j: PageRank



# Neo4j: Closeness Centrality

**Goal:** Identify the most centralized art departments

**Result:** European Paintings are the most centralized with a closeness centrality score of 0.3446

## Business Case Applications:

- Identifies possible relationships for art history research
- Prompts collaboration on cross-departmental exhibits

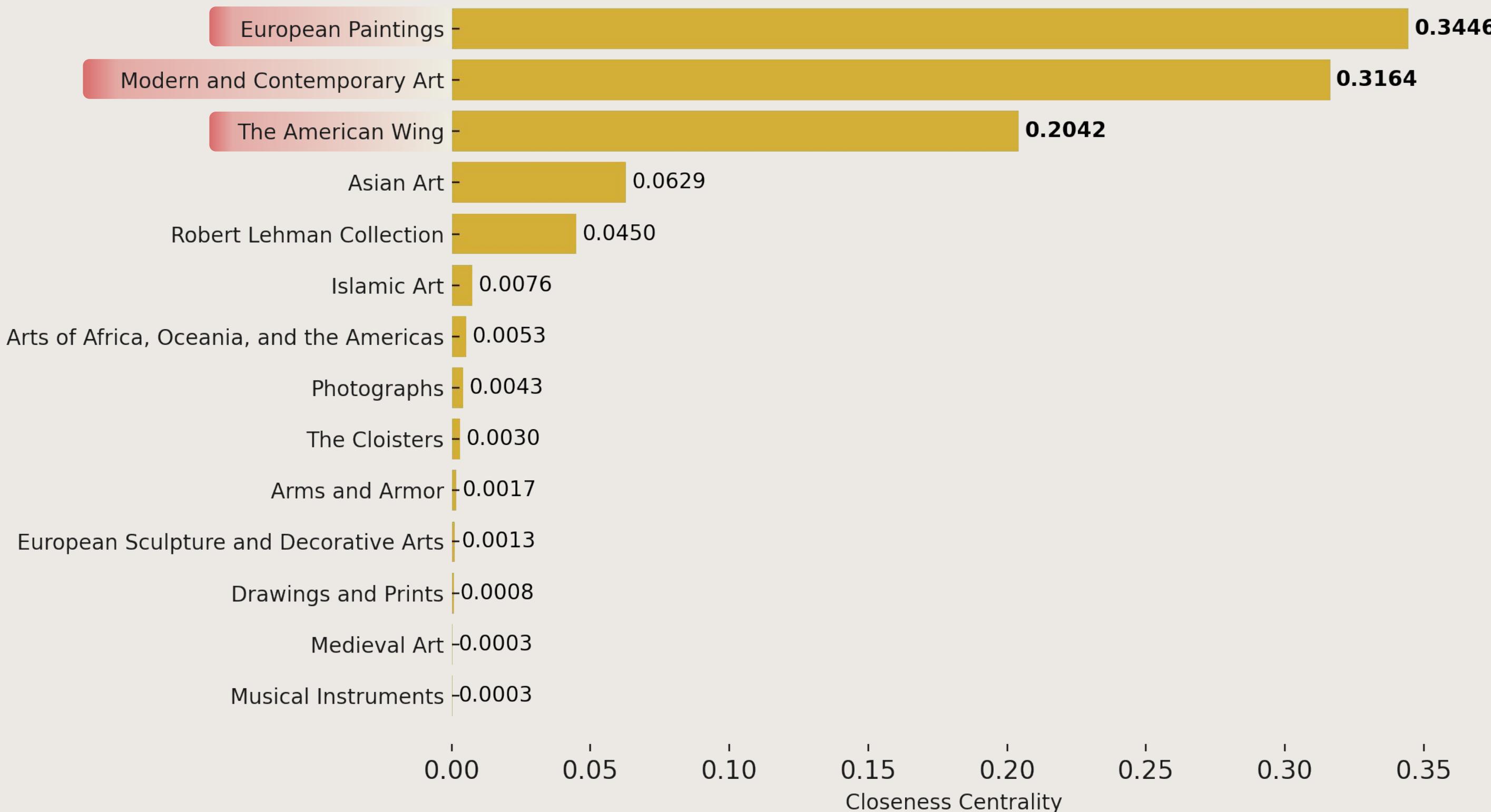
## Why Not Relational Database:

- Shortest Path calculation is computationally expensive & memory-intensive
- Less scalable for complex datasets



# Neo4j: Closeness Centrality

Department Closeness Centrality in Artwork Graph



# Neo4j: Louvain Modularity

**Goal:** Detect communities connected by departments/time periods

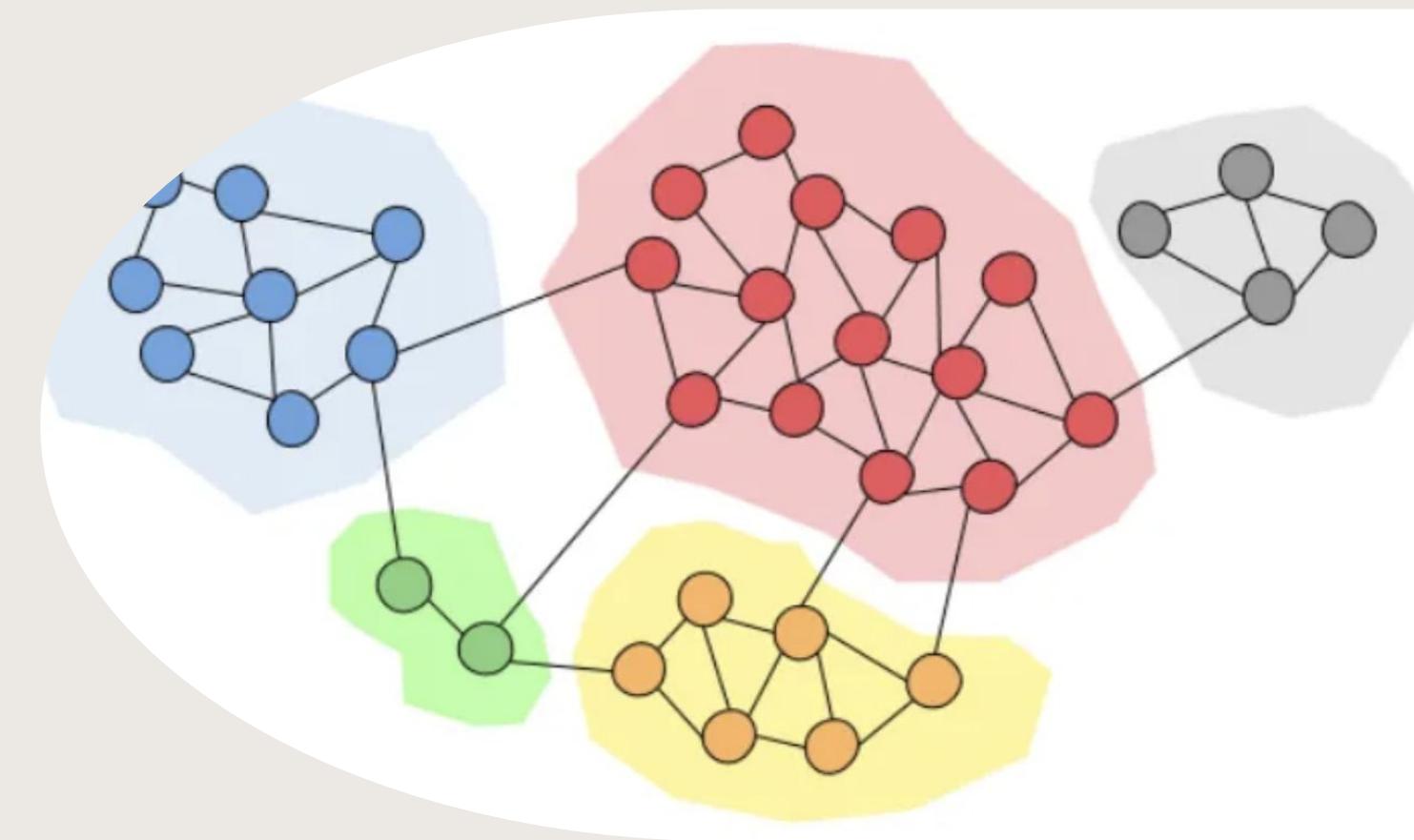
**Result:** 9 communities, 0.61 modularity score

## Business Case Applications:

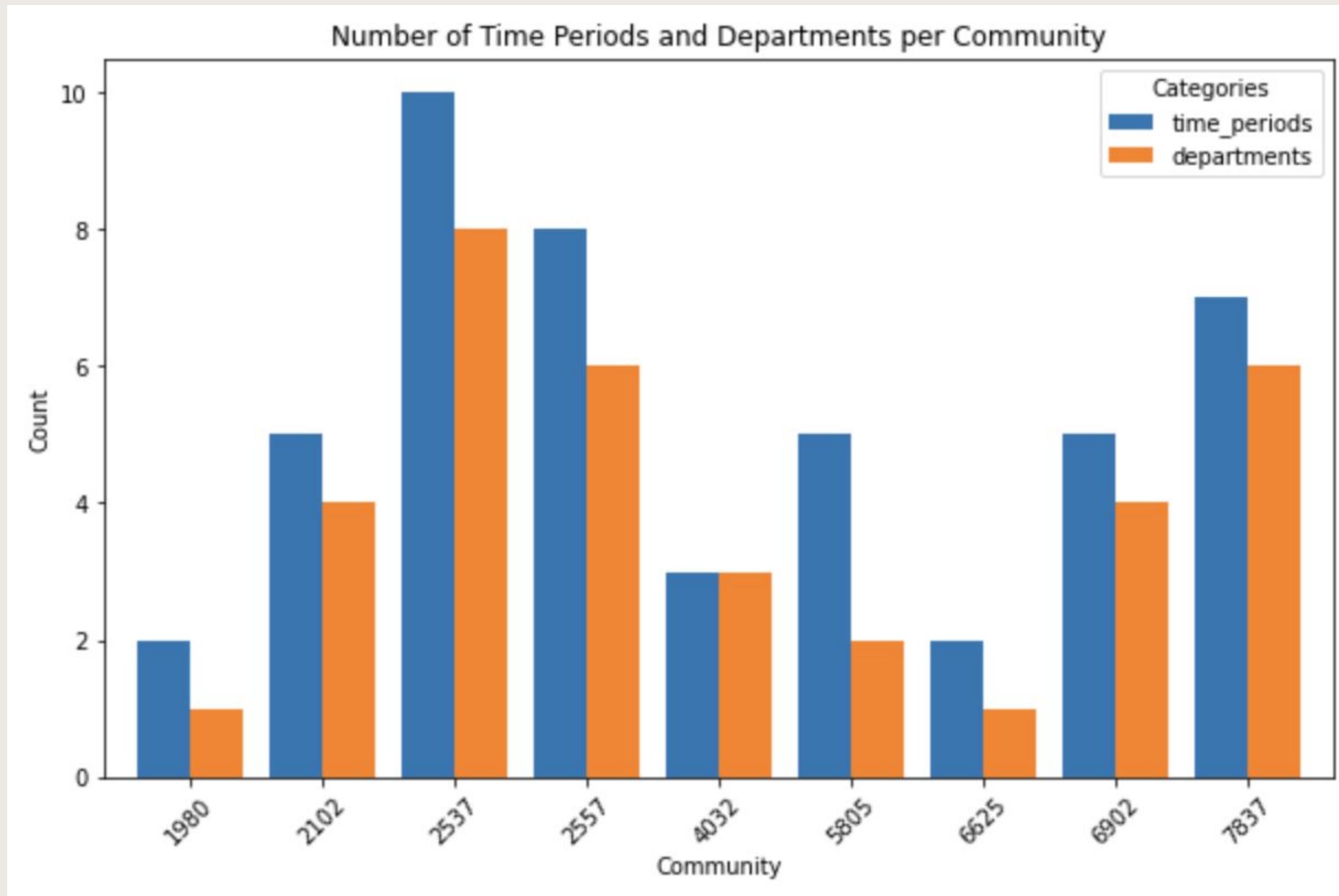
- Reveals relationships ideal for storytelling in tours
- Helps recommend related artworks/artists
- Supports exhibits on movements within a community

## Why Not Relational Database:

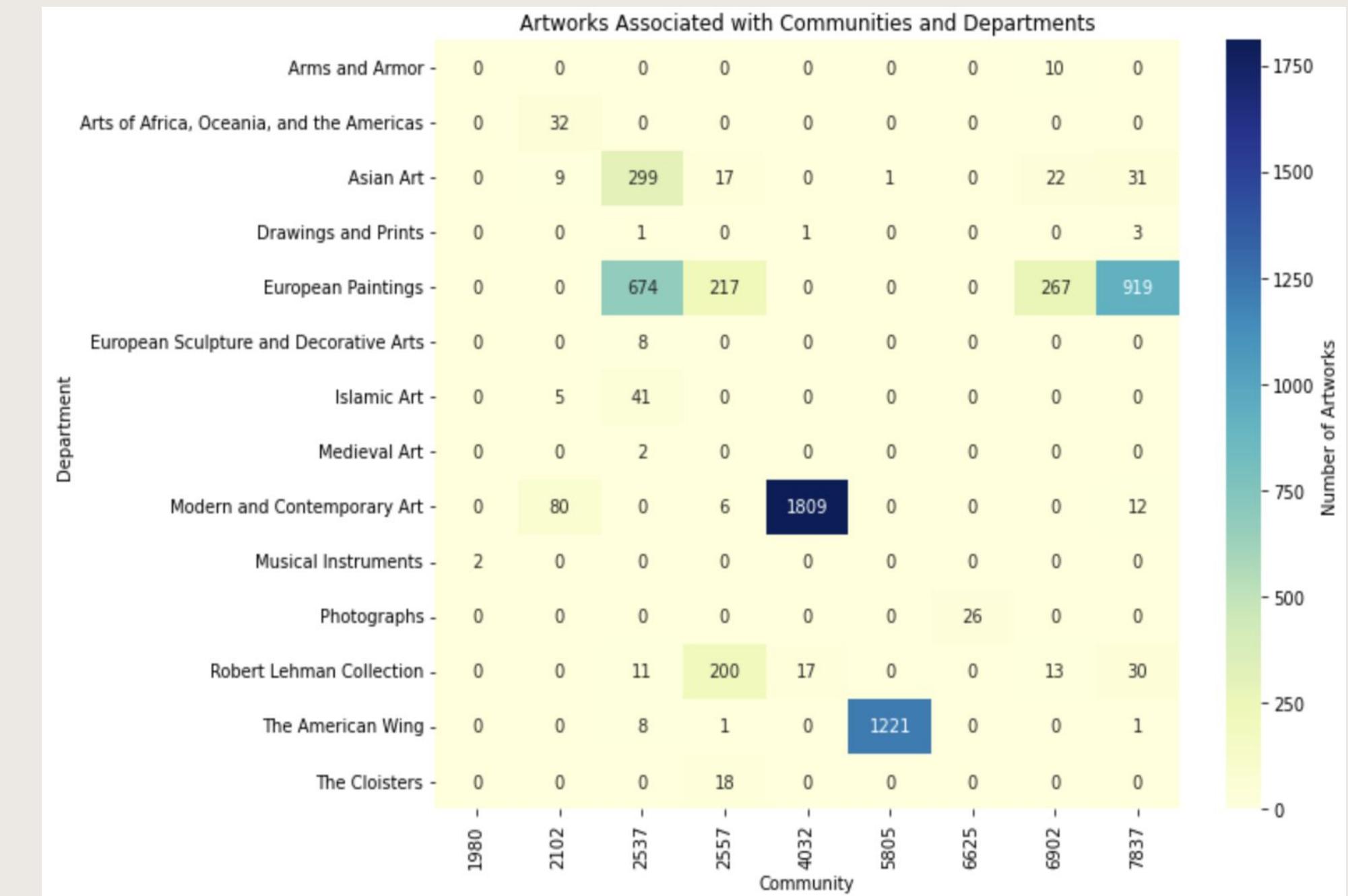
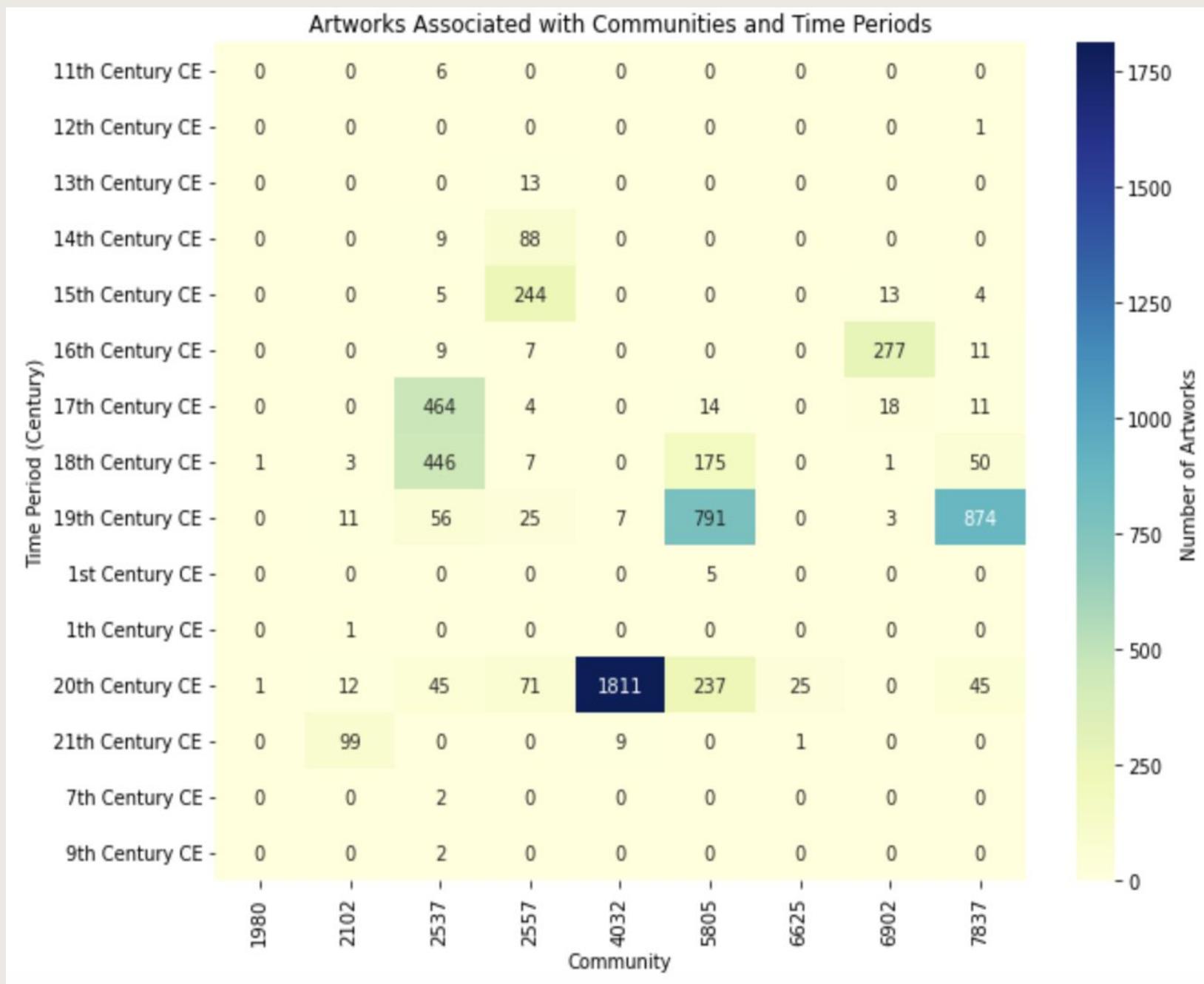
- Struggle with deeply interconnected data
- Require expensive joins



# Neo4j: Louvain Modularity



# Neo4j: Louvain Modularity



# MongoDB

## Goal

- Generate exhibition ideas based on visitor interests

## Method

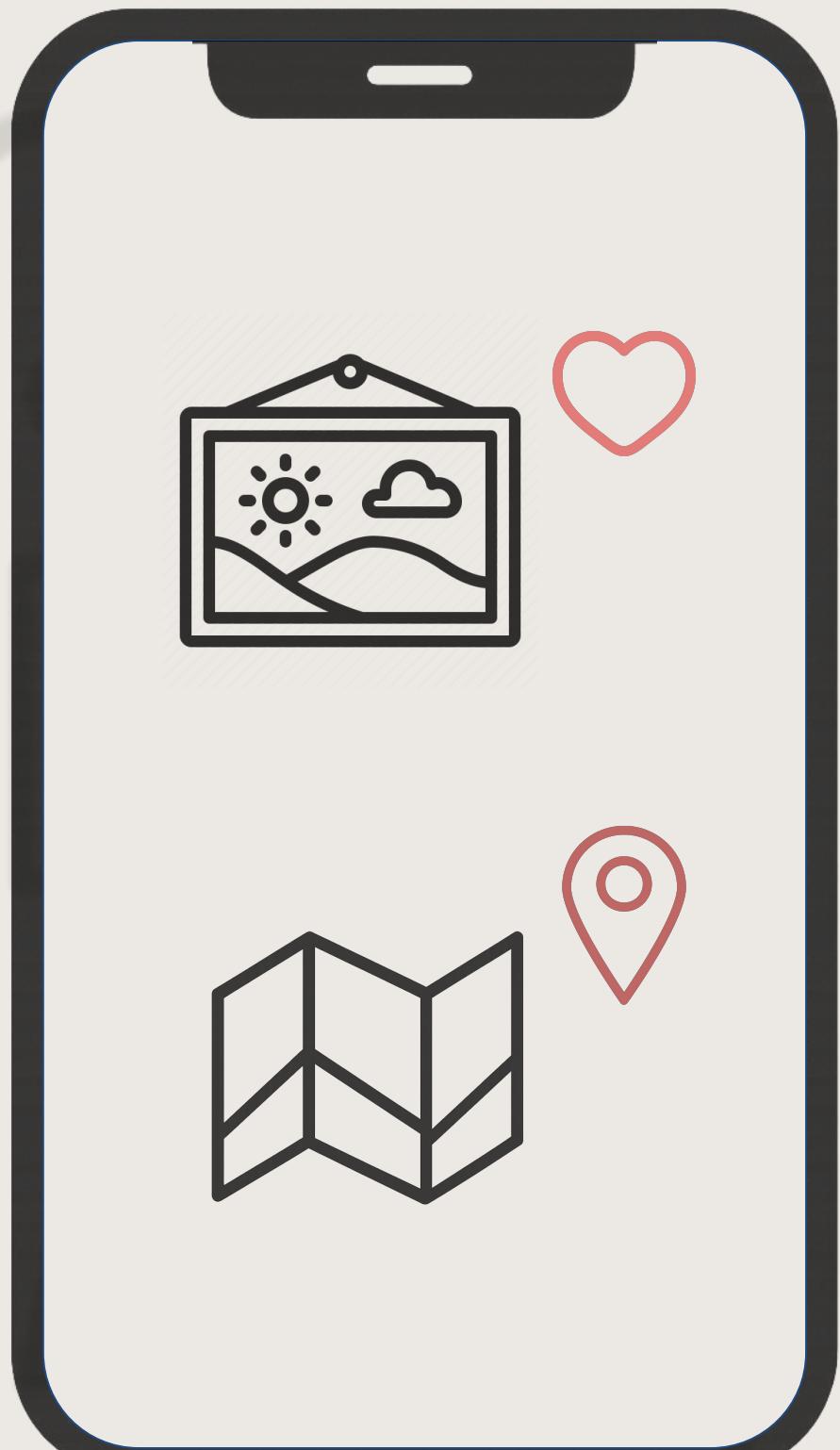
- Visitors can participate in a survey at the end of their visit to rate their favorite departments, artworks, and artists that they saw that day.
- Create points of view (POVs) for artists, artwork, departments, and time periods.
- Store and query this data with Mongo DB for future analysis when showcasing new exhibits.

## Why not a relational database

- Requires flexibility to analyze our data from multiple POVs



# Redis



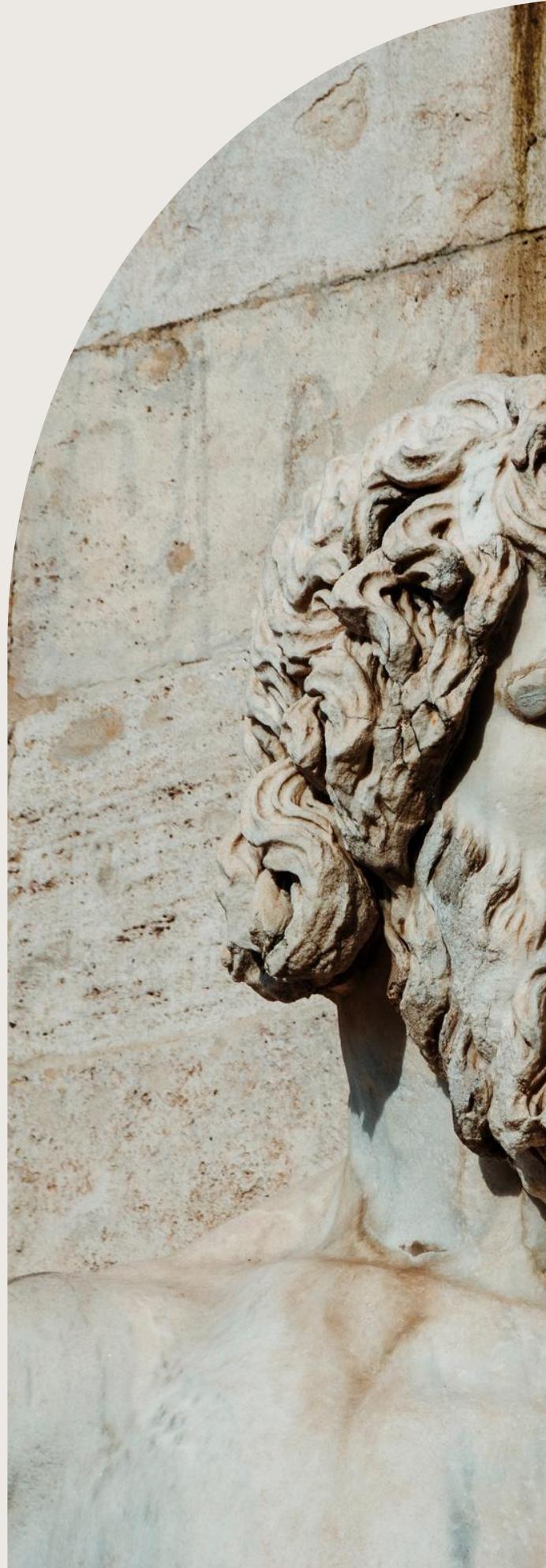
**Goal:** Create a mobile, real-time recommendation system for visitors

**Method:**

- Visitors favorite artworks on mobile app
- Visitors' location tracked by mobile app
- App uses favorites and/or live location in museum (key) to provide recommendations of artwork currently on view in nearby galleries (value)

**Why Not Relational Database:**

- Requires immediate access to *subsets* of the database
- Cannot provide *dynamic* updates as needed



# References

[The Met Dataset](#)

[Our GitHub Repo](#)



# Q&A

## Any Questions?