

UNIVERSITAT POLITÈCNICA DE  
CATALUNYA

BACHELORS IN COMPUTER SCIENCE AND  
ENGINEERING

MINOR HEADING

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**Graph and matrix algorithms  
for visualizing high dimensional  
data**

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*To my Mother, Father, Professors and Friends. I owe a lot to My  
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## **Abstract**

Motivated by the problem of understanding data from the medical domain, we consider algorithms for visually representing highly dimensional data so that "similar" entities appear close together. We will study, implement and compare several algorithms based on graph and on matrix representation of the data. The first kind are known as "community detection" algorithms, the second kind as "clustering" algorithms. The implementations should be robust, scalable, and provide a visually appealing representation of the main structures in the data.

## Acknowledgement

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# 1 Introduction

In this section we provide an overview of the entire work. We mention the context of the project we have studied, approaches that we have used, goal of the project. We also provide the intended planning, economic estimate and sustainability of the work that has been done.

## 1.1 Context Of the Project

In the present day scenario, the modern science of algorithms and graph theory has brought significant advances to our understanding of complex data. Many complex systems are representable in the form of graphs. Graphs have time and again been used to represent real world networks. One of the most pertinent feature of graphs representing real system is community structures or otherwise known as clusters. Community can be defined as the organization of vertices in groups or clusters, with many edges joining the vertices of the same cluster and comparatively fewer vertices joining the vertices in another neighbouring cluster. Such communities form an independent compartment of a graph exhibiting similar role. Thus, Community detection is the key for understanding the structure of complex graphs, and ultimately educe information from them.

we have complex data today, often representable in form of graphs. we need to visualize, find structure, etc.

## 1.2 Approaches

Virtually in every scientific field dealing with empirical data, primary approach to get a first impression on the data is by trying to identify groups having "similar" behaviour in data.

community finding and clustering. separately, visualization tools



### **1.3 Goal of the Project**

### **1.4 Planning and Budget**

### **1.5 Sustainability**

## **2 Background**

### **2.1 Graph Notation**

Graph ,G , is construct consisting of two finite sets, the set  $V = \{ v_1, v_2, \dots, v_n \}$  of vertices and the set  $E = \{ e_1, e_2, \dots, e_n \}$  of edges where each edge is a pair of vertices from V, for instance,

$$e_i = (v_j, v_k)$$

is an edge from  $v_j$  to  $v_k$  represented as  $G=(V,E)$ .

### **2.2 Matrix Notation**

### **2.3 Equivalence between Graph and Matrix Representation**

### **2.4 State-of-the-art in Community finding**

### **2.5 State-of-the-art in Clustering**

### **2.6 State-of-the-art in Graph Visualization**

explain technical concepts in more detail. for exampleequivalence of graph and matrix representations. state-of-the art in community finding, and clustering state-of-the art in graph visualization

## **3 Community Finding Algorithm**

### **3.1 Louvain Algorithm**

#### **3.1.1 Introduction**

#### **3.1.2 Reasoning**

#### **3.1.3 Description**

#### **3.1.4 Implementation**

#### **3.1.5 Experiments**

#### **3.1.6 Result**

## **4 Matrix Based Algorithm**

### **4.1 Matrix Algorithm**

#### **4.1.1 Introduction**

#### **4.1.2 Reasoning**

#### **4.1.3 Description**

#### **4.1.4 Implementation**

#### **4.1.5 Experiments**

#### **4.1.6 Result**

## **5 Visualization**

### **5.1 Alchemy.js**

#### **5.1.1 Introduction**

#### **5.1.2 Reasoning**

#### **5.1.3 Description**

#### **5.1.4 Methods and Library**

#### **5.1.5 Result**

## **6 Overall System Description**

### **6.1 Alchemy.js**

#### **6.1.1 Introduction**

#### **6.1.2 Implementation Benefits**

#### **6.1.3 Description**