# Finding Patterns in Complex Social Networks

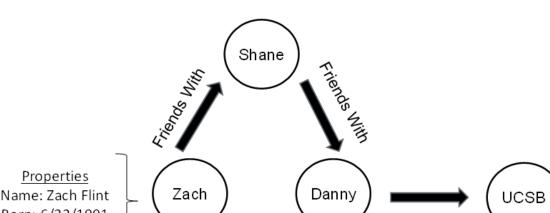
UCCSB UNIVERSITY OF CALIFORNIA SANTA BARBARA	Lennon Ganz Mentor: Dr. Yinghui Wu Faculty Advisor: Dr. Xifeng Yar		INAR(	<section-header><text></text></section-header>
Overview	GLOBAL TERRORISM DATABASE Visualization	Indexing		
<ul> <li>Motivation</li> <li>Networks are all around us</li> <li>Large networks are difficult to analyze and understand</li> </ul>		Inverted Index A mapping of properties to the nodes that contain	Country United States Iraq	Incidents ID=23422, ID=66430, ID=72230, ID=94102,
Challenge         • Networks quickly become large and complex         • Questions (queries) are difficult to write and often		that property. Similar to the index at the back of a book.	Pre	ID=40660, ID=19430, events repeated, expensive earches of entire dataset

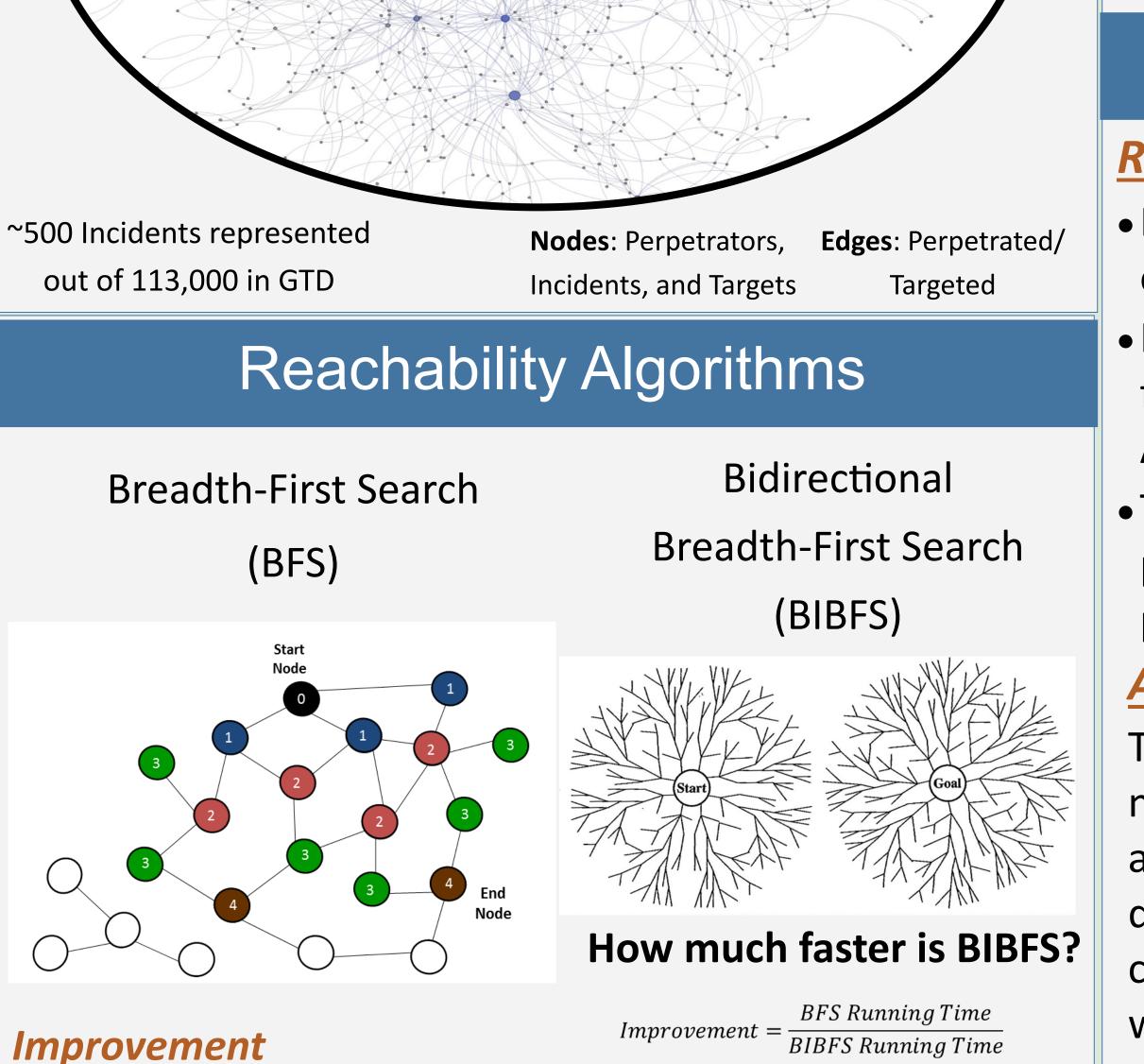
- require learning a special query language
- Algorithms become slow as size of data increases Goals
- Develop a simple, intuitive platform to analyze complex social networks
- Improve the efficiency of existing search algorithms and develop an efficient pattern matching algorithm

# Key Terms

- **Query** A question or request for information
- Algorithm A set of instructions to answer a question or complete a process
- "Social" Network A network of social interactions

**Graph -** Data structure that represents entities (nodes) and the relationships (edges) between





20

15

10

25

20

15

10

Performance

mprovement

(x faster)

#### Performance vs. Graph Size

1500

Performance vs. Edge Probability

0.4

0.5

0.6

1000

Number of Nodes

2500

2000

# Summary

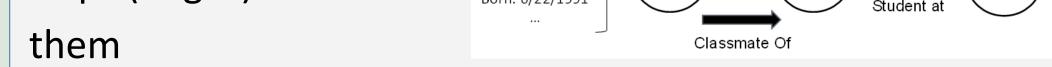
## Results

- Bidirectional searching is almost always faster than one-way searching, upwards of 30x or more
- Inverted indexing greatly improves performance for finding specific nodes at the cost of memory/storage. Access time is reduced from seconds to milliseconds
- The pattern matching algorithm successfully matches patterns of nodes and edges each with specified properties

## Accomplishment

This system makes the process of analyzing large networks and graphs faster and more intuitive. This allows for a better understanding of each specific dataset and ultimately a better understanding of the complex connections and relationships present in the world around us

# Future Development and Goals



**Node -** An entity in a graph (e.g. person, place event)

Edge - A connection between two nodes in a graph, representing a relationship between them

**Density** - # of edges divided by # of nodes

Types of Queries

Туре	Example (Dataset = Global Terrorism Database)
Aggregation	"How many incidents were there in 2005?" "What was the most common type of weapon used?"
Reachability	"Have these two terrorist groups ever collaborated on an attack?"
	"Find instances where a terrorist group targeted a

vincingly outperforms the standard BFS approach by over 30x

• As graph size increases with

constant density, BIBFS con-

• As edge density increases with constant number of nodes, BIBFS remains faster but approaches the speed of BFS

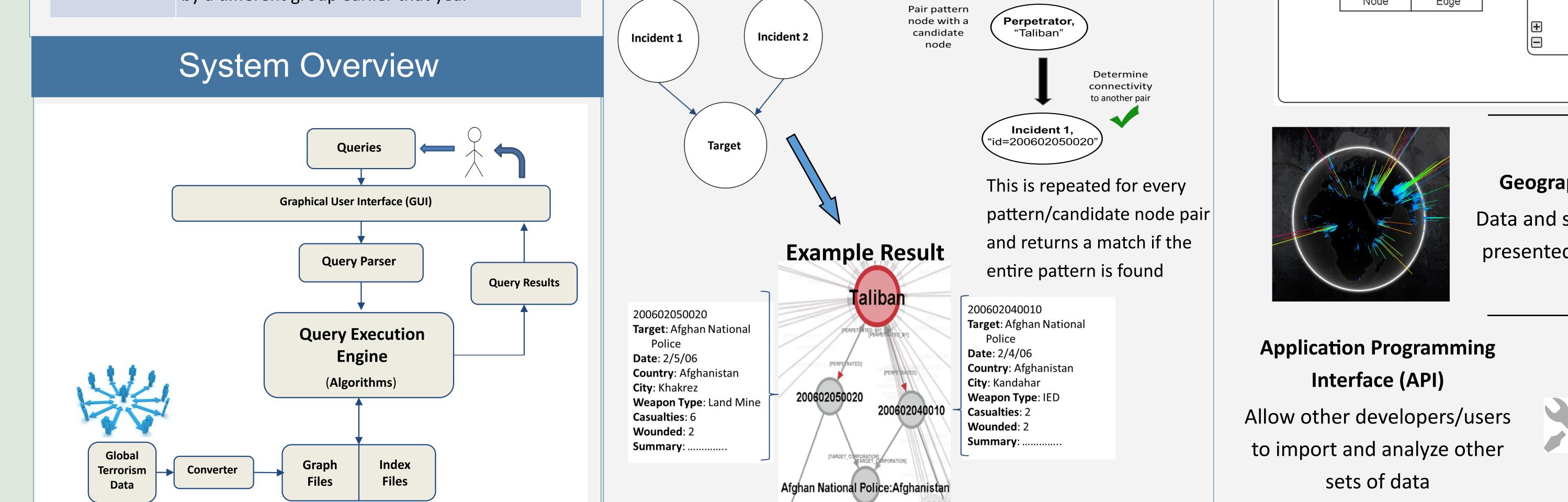
## Drawbacks

- Larger overhead (resourceheavy)
- Must know starting and ending nodes (unable to freely explore graph)



First, candidate nodes (nodes that might fit in indexes

0.1

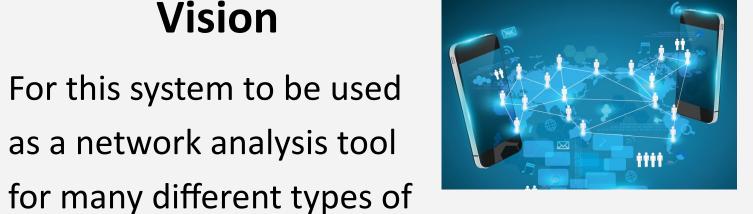


#### Vision

data; social networks,

cyber security networks,

neural networks and more.





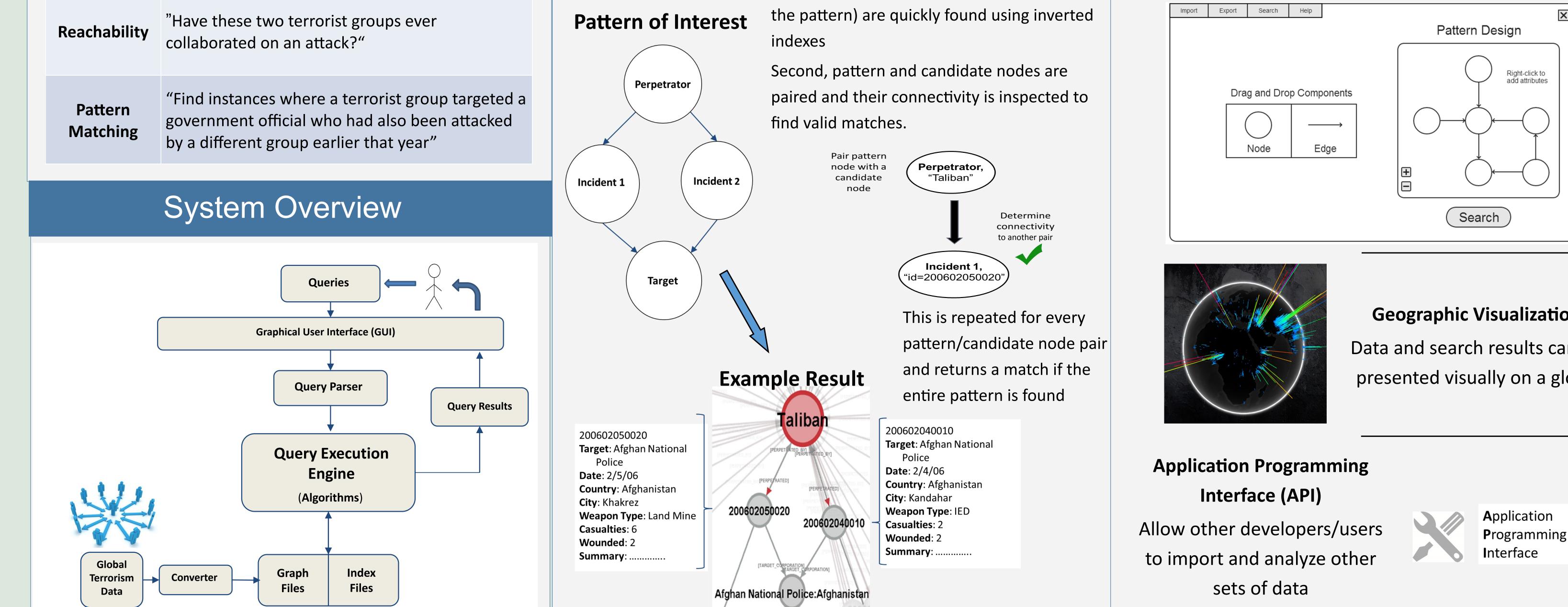




### **Graphical User Interface (GUI)**

Allow users to easily and visually write queries, create search patterns, and view

results



**Geographic Visualization** Data and search results can be presented visually on a globe