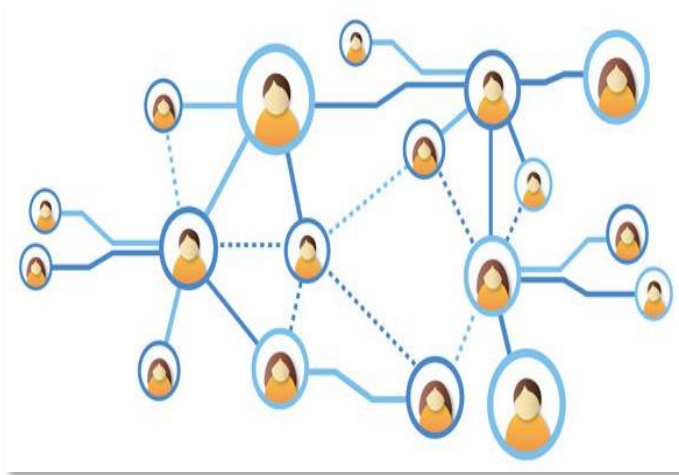


Stad: Stateful Diffusion for Linear Time Community Detection. ICDCS'18

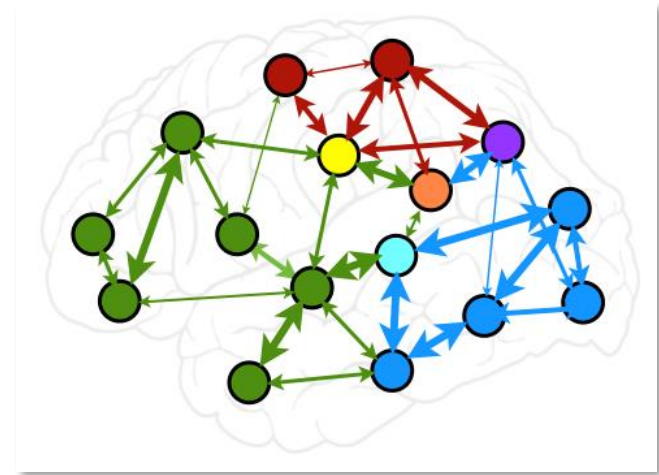
Amira Soliman, Fatemeh Rahimian,
and Sarunas Girdzijauskas

Network Analysis

DISCAN'18, Stockholm, Sweden.



Social networks



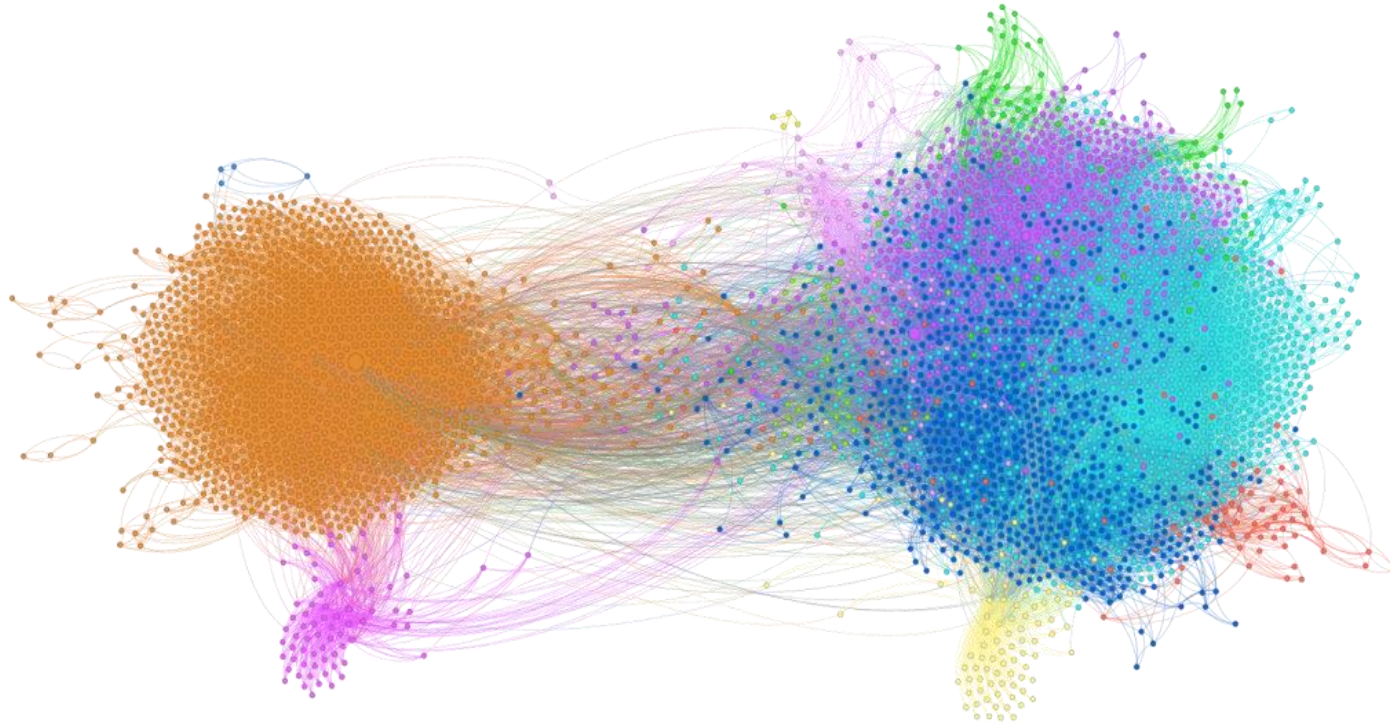
Biological networks

Community Analysis



- **Community represents a group of nodes that are densely connected to each other and sparsely connected to the rest of the graph.**

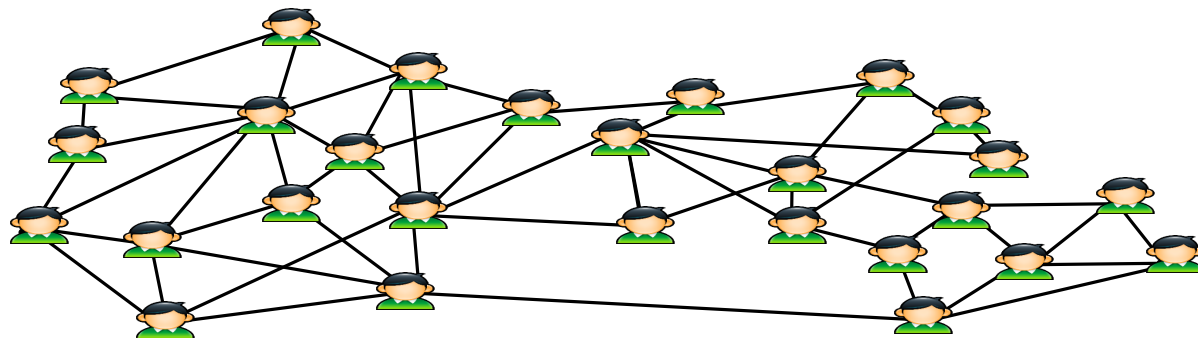
Community Analysis



- Community represents a group of nodes that are densely connected to each other and sparsely connected to the rest of the graph.
- Communities vary in **size** and **internal cohesion**.

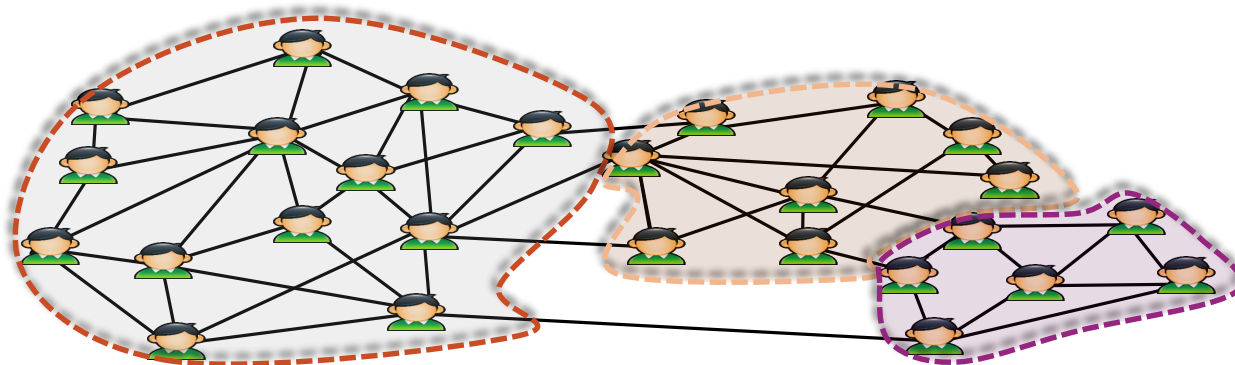
Community-aware Learning

Integrating community structure with analytic tasks enhances the data analytic insights and improves the results (e.g., recommender systems and spam detection).



Community-aware Learning

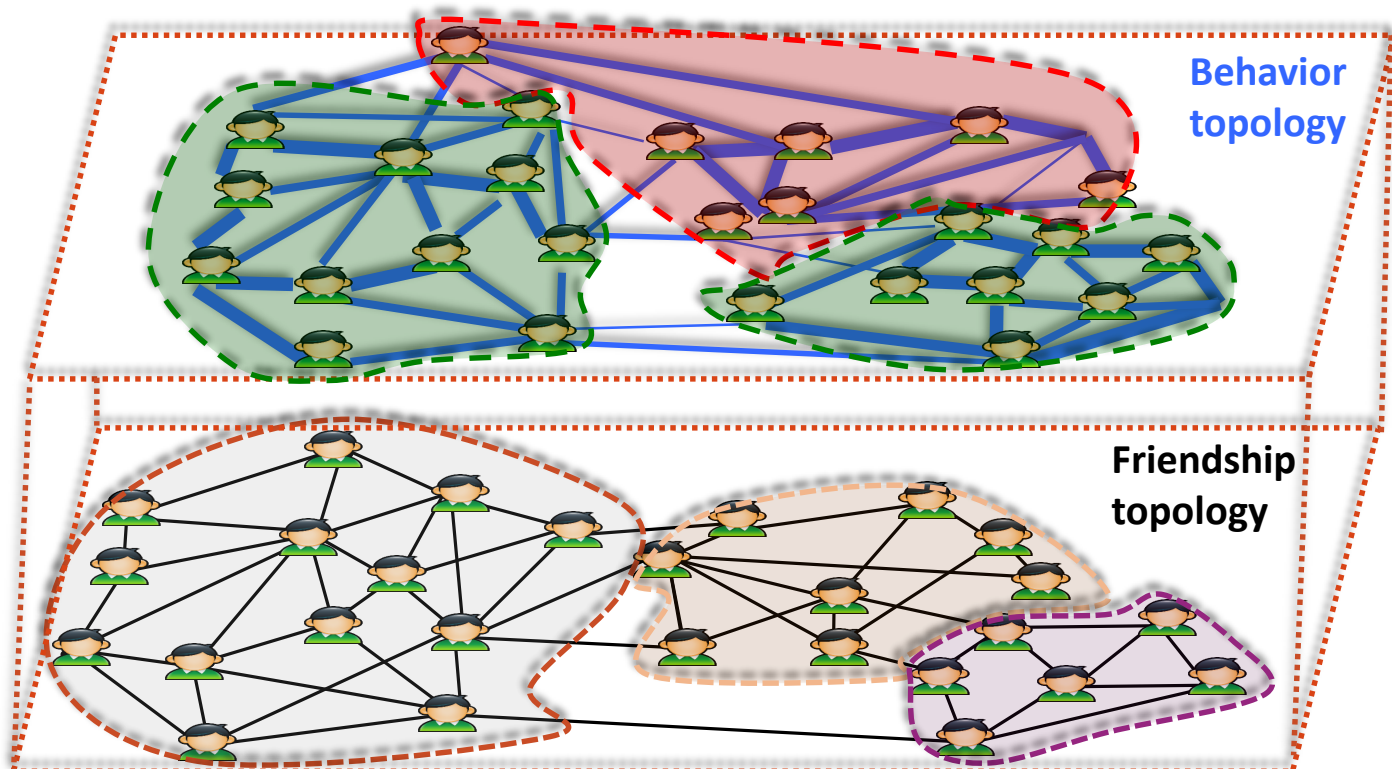
Integrating community structure with analytic tasks enhances the data analytic insights and improves the results (e.g., recommender systems and spam detection).



Community-aware Learning

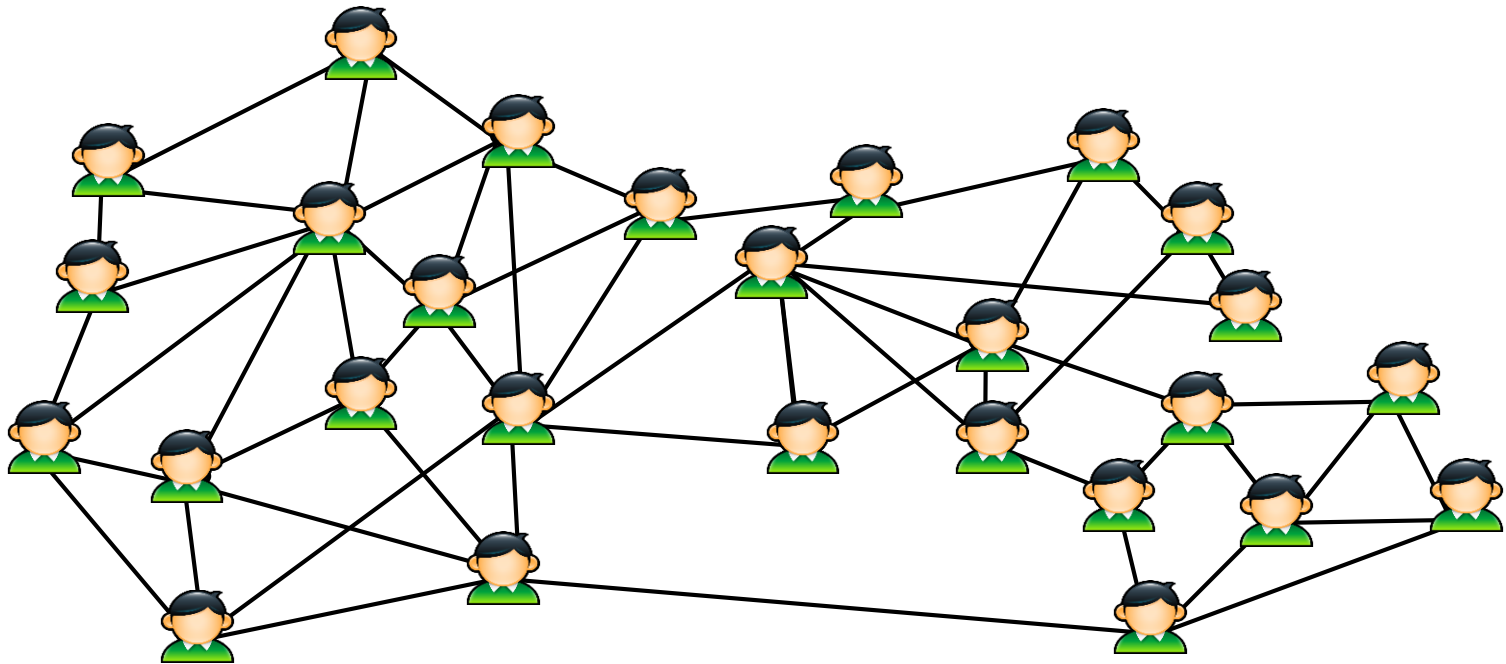
DISCAN'18, Stockholm, Sweden.

Integrating community structure with analytic tasks enhances the data analytic insights and improves the results (e.g., recommender systems and spam detection).



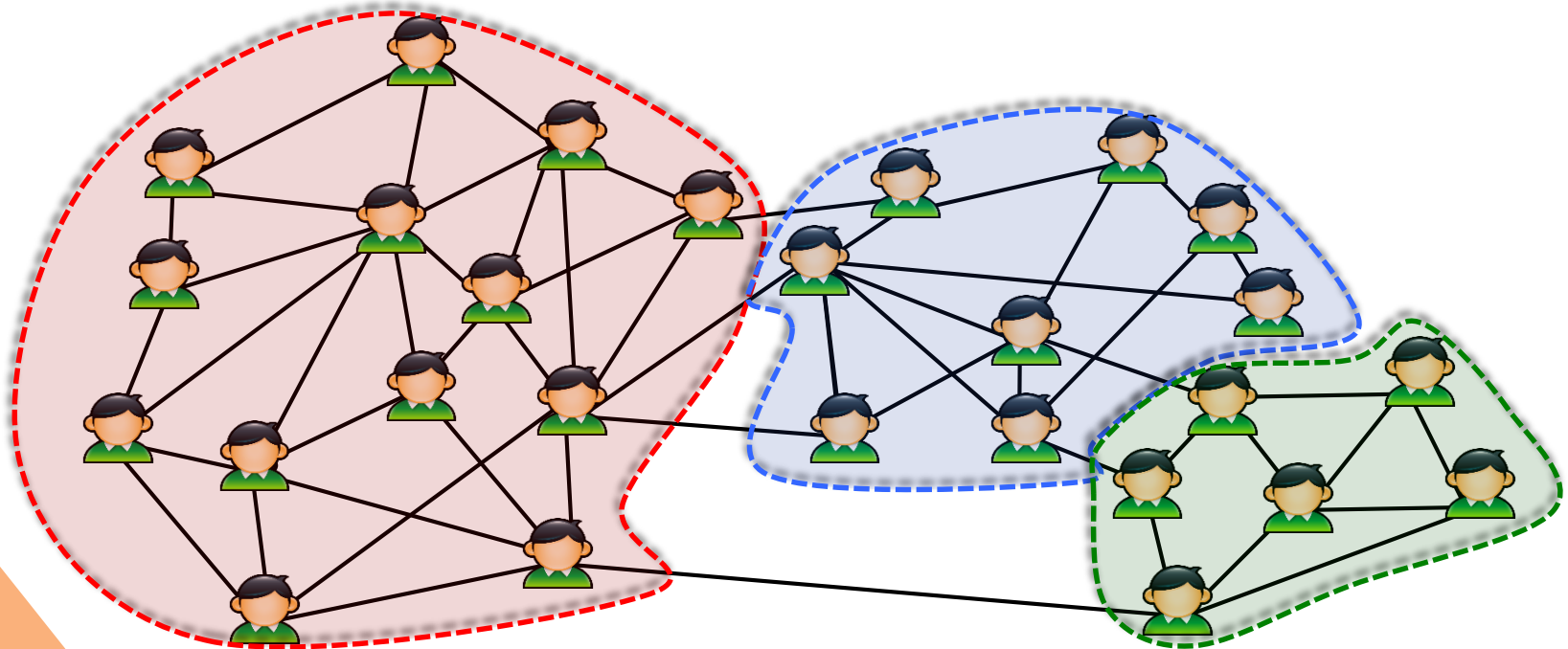
Community Detection

A community corresponds to a group of nodes with more **intra-cluster** than **inter-clusters** edges.



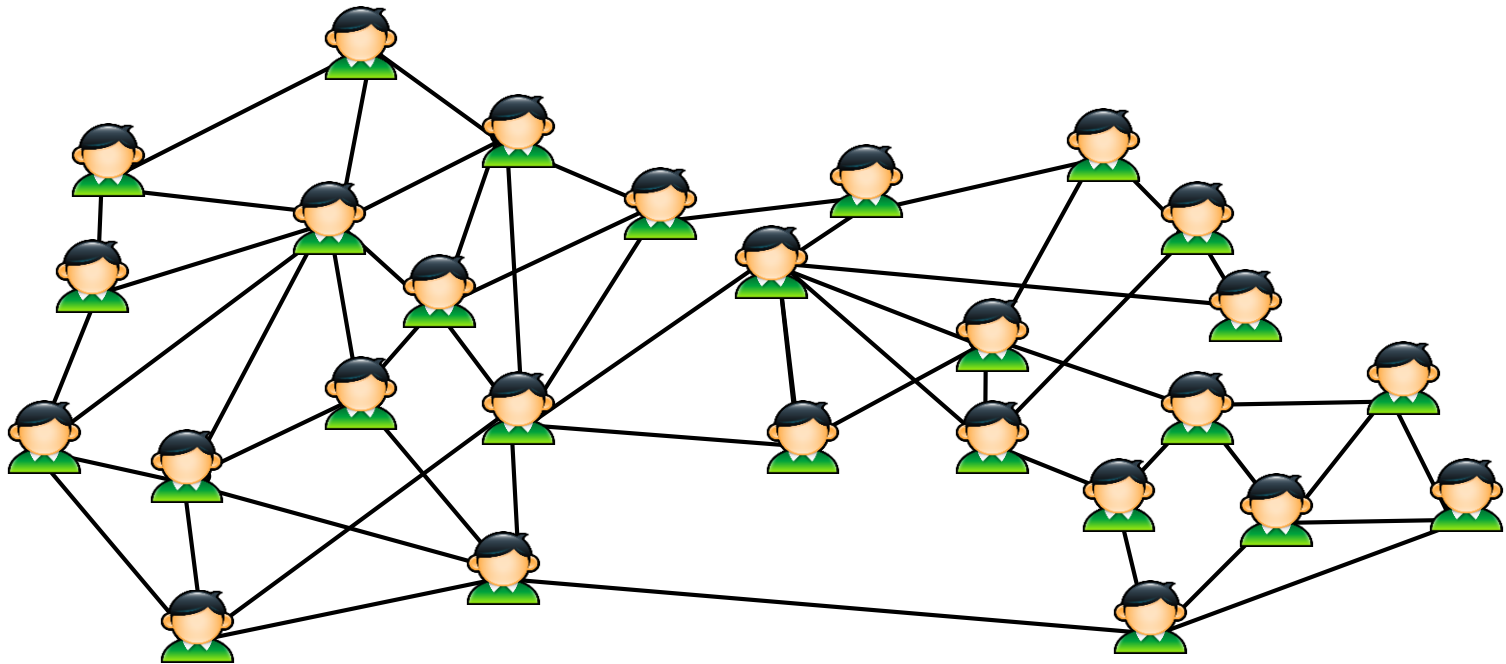
Community Detection

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Community Detection

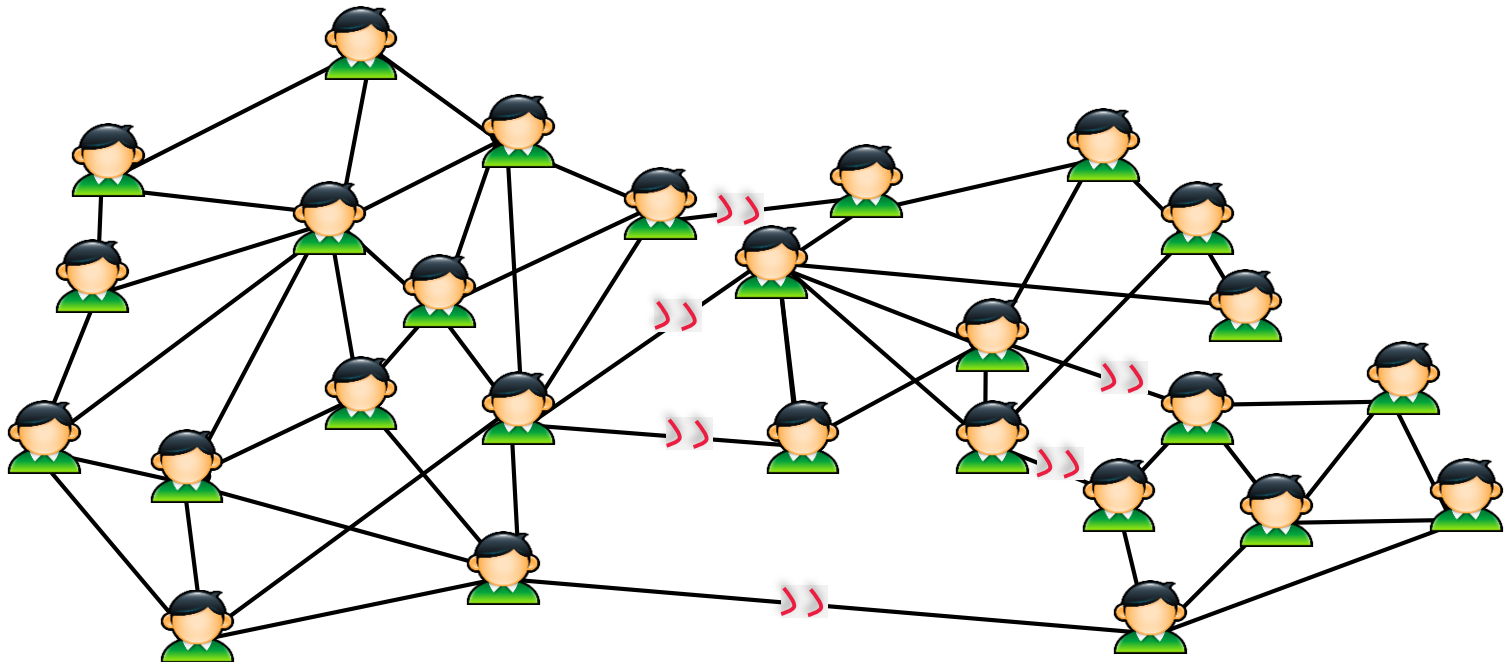
A community corresponds to a group of nodes with more **intra-cluster** than **inter-clusters** edges.



Community Detection

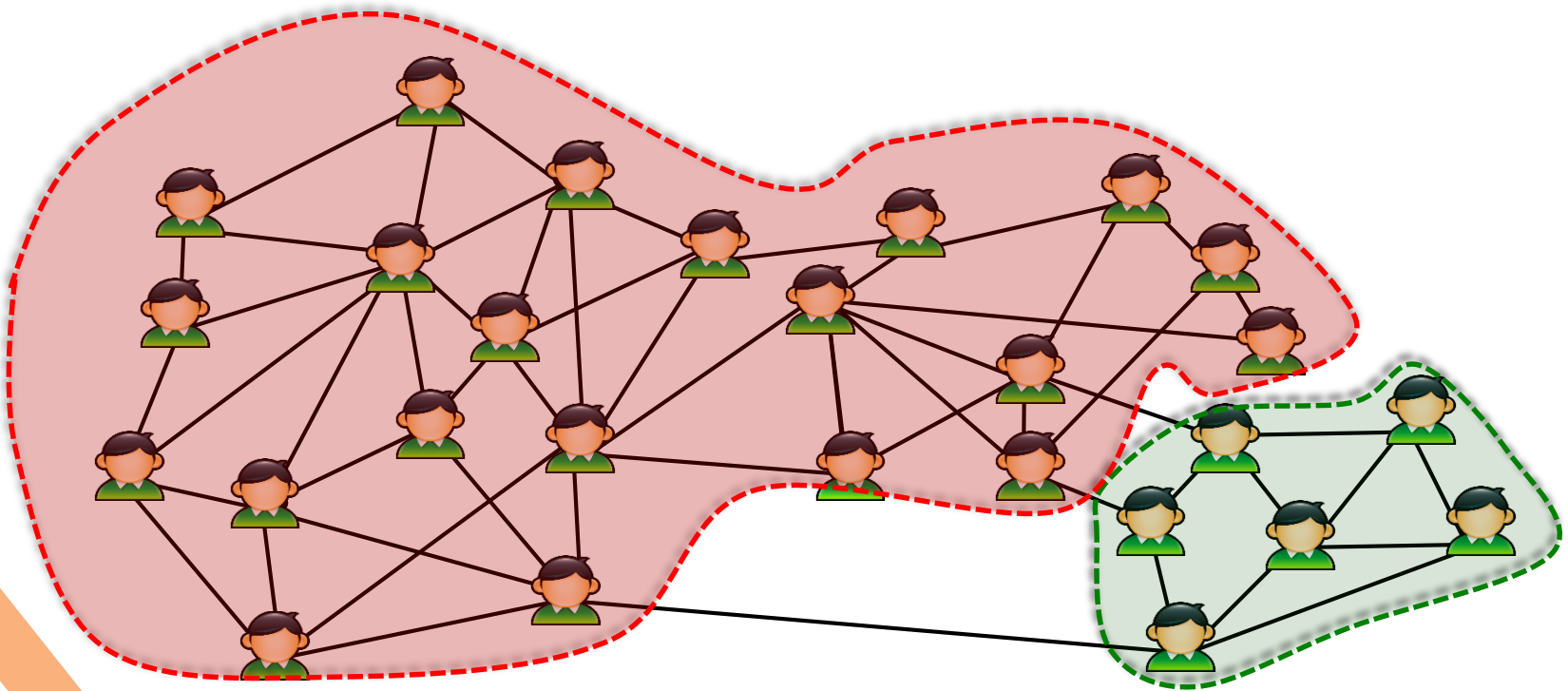
DISCAN'18, Stockholm, Sweden.

A community corresponds to a group of nodes with more **intra-cluster** than **inter-clusters** edges.



Community Detection

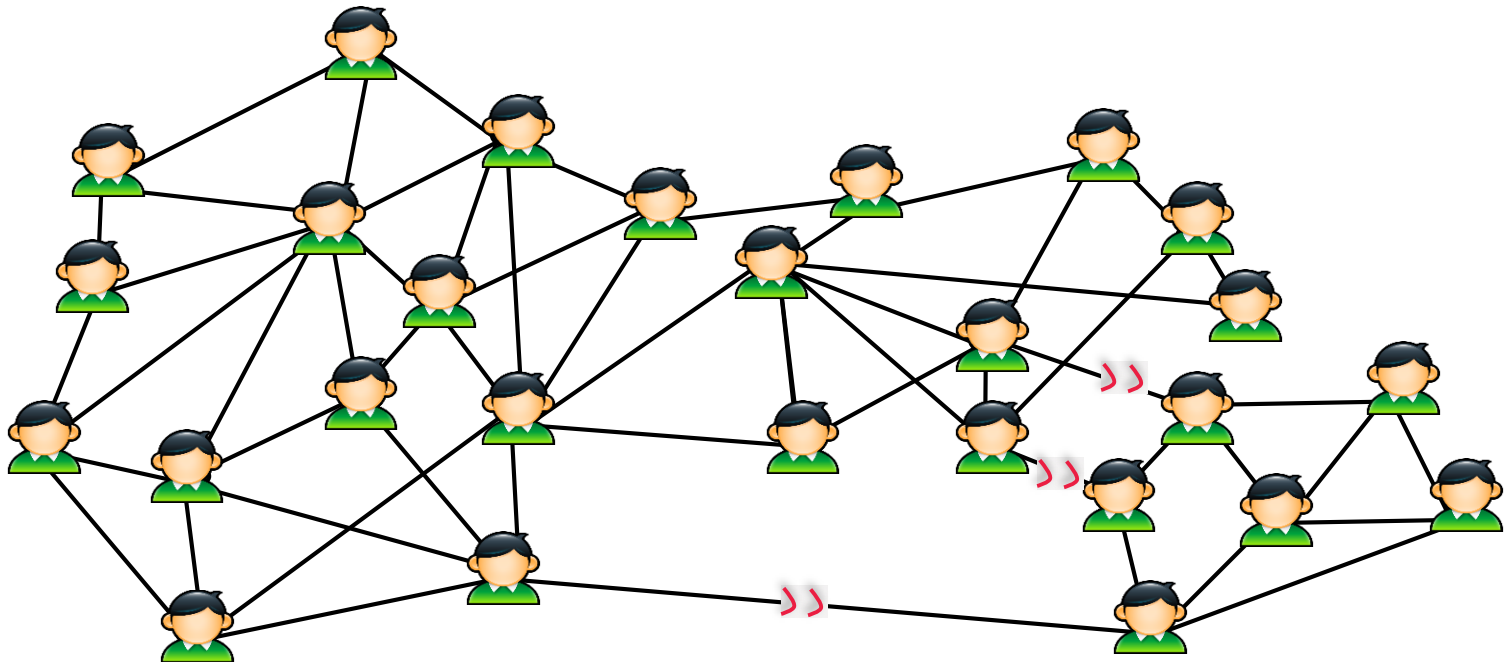
A community corresponds to a group of nodes with more **intra-cluster** than **inter-clusters** edges.



Community Detection

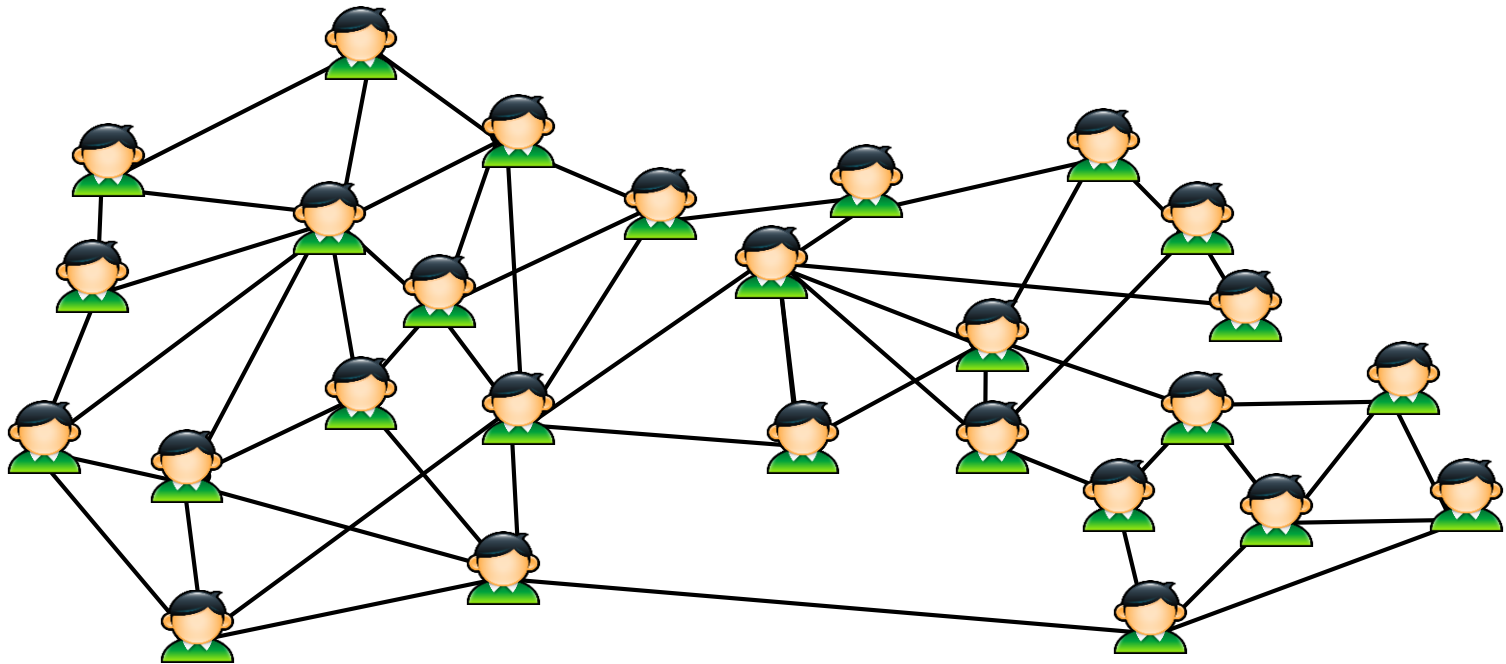
DISCAN'18, Stockholm, Sweden.

A community corresponds to a group of nodes with more **intra-cluster** than **inter-clusters** edges.



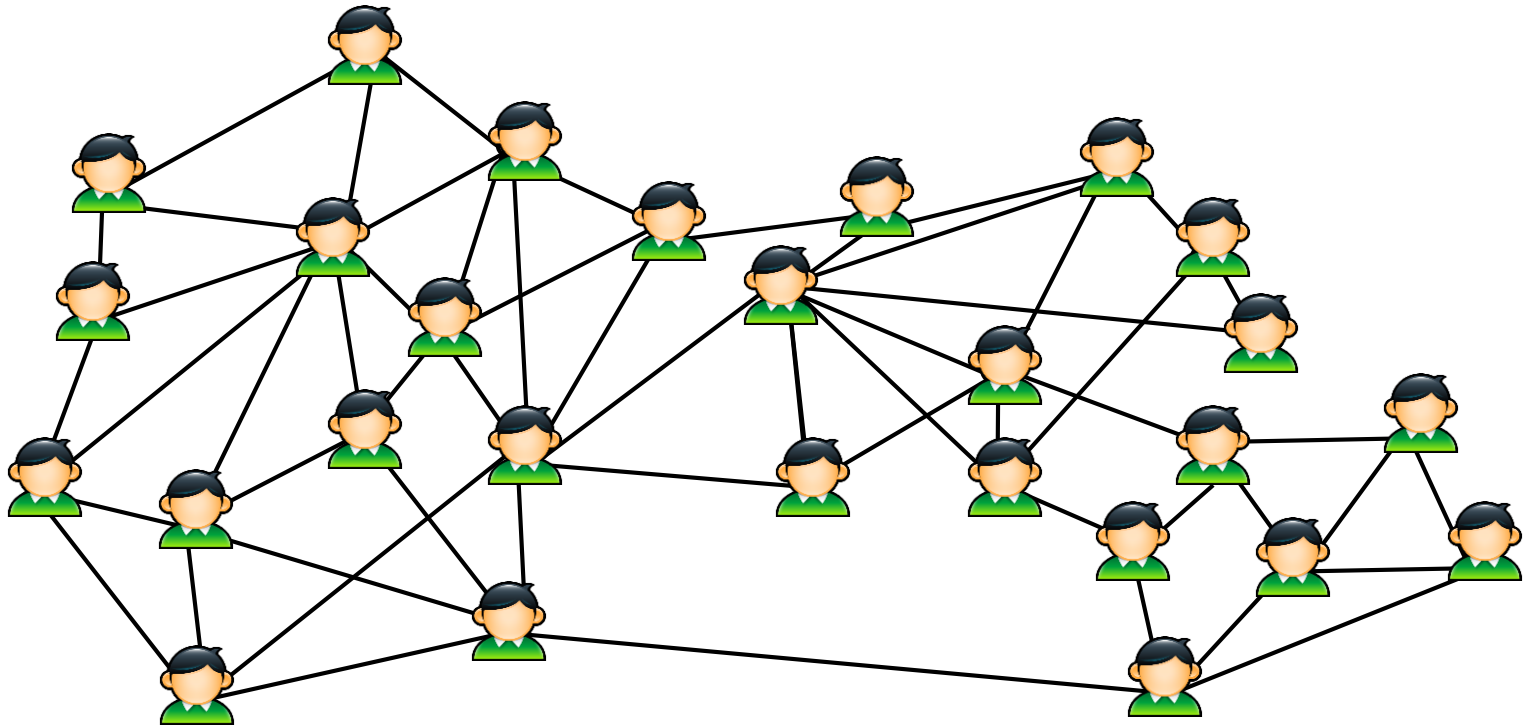
Community Detection

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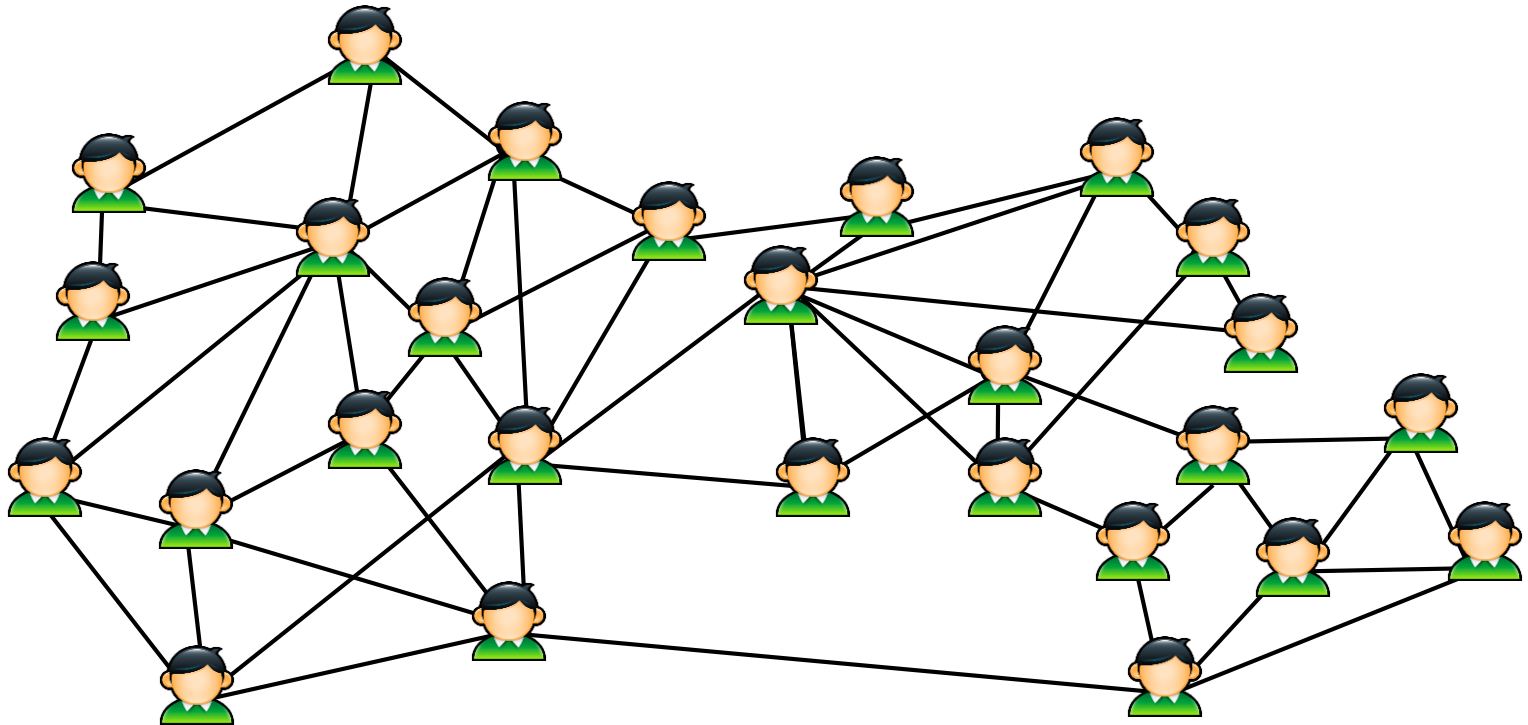
Localized-methods for Community Detection

DISCAN'18, Stockholm, Sweden.



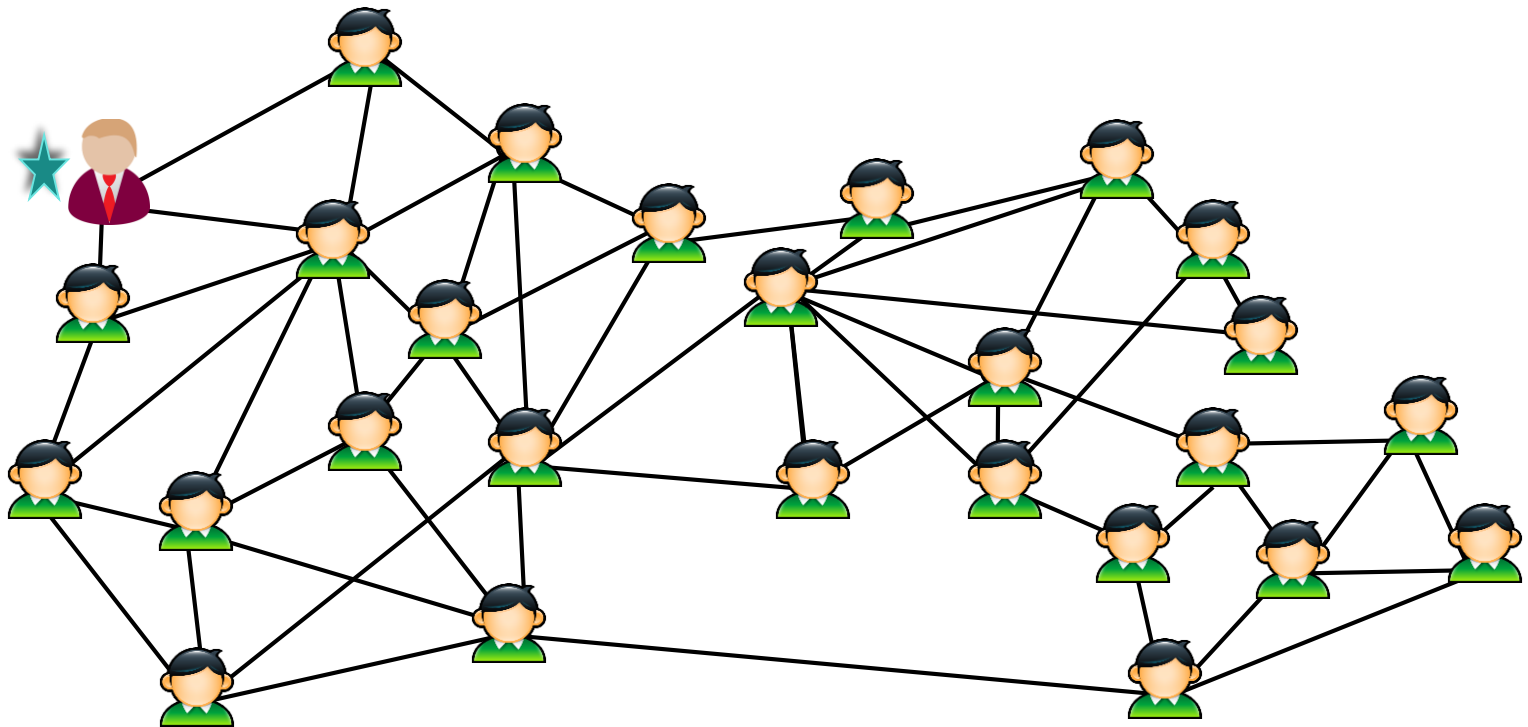
Localized-methods for Community Detection

Random Walks (RWs):



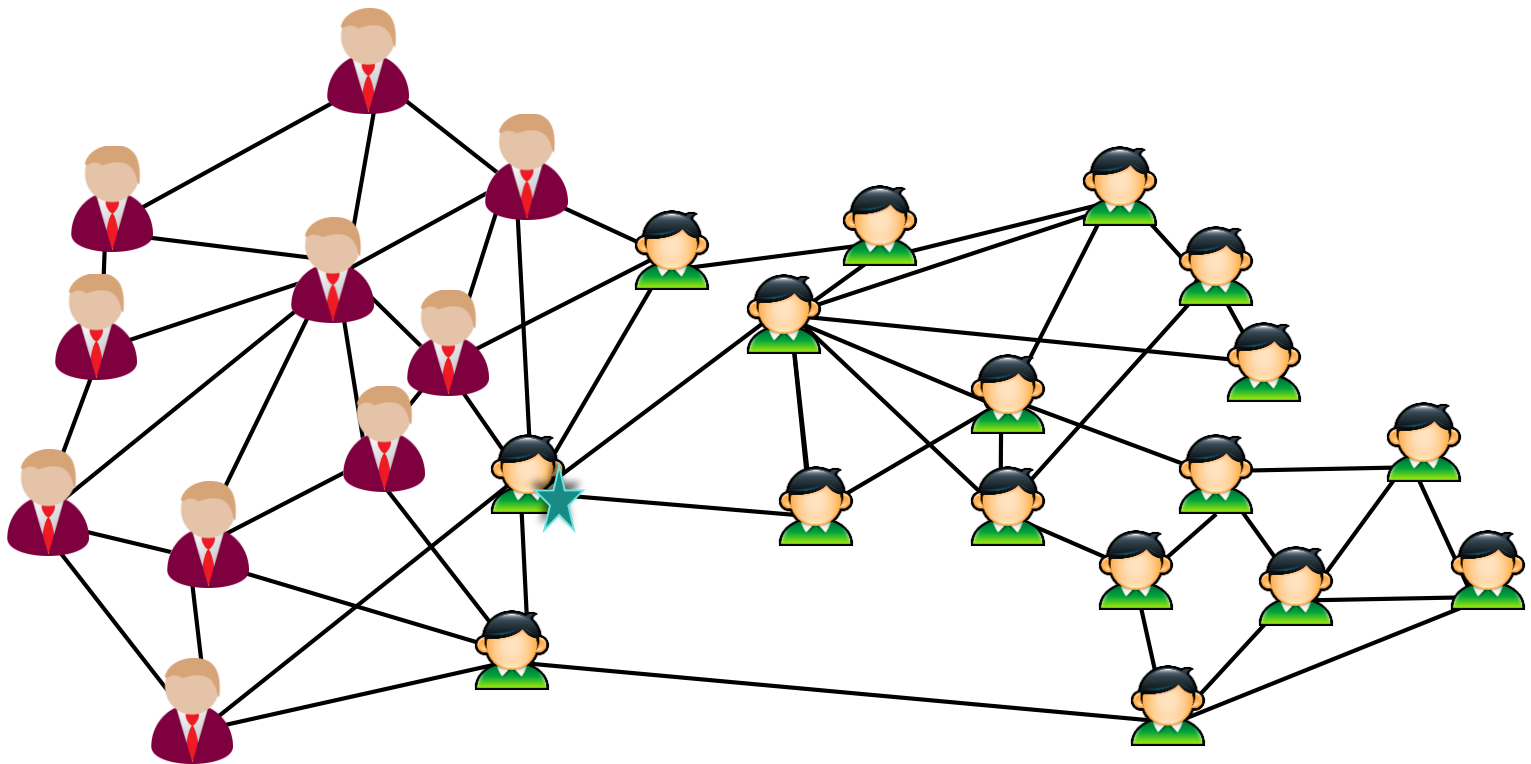
Localized-methods for Community Detection

Random Walks (RWs):



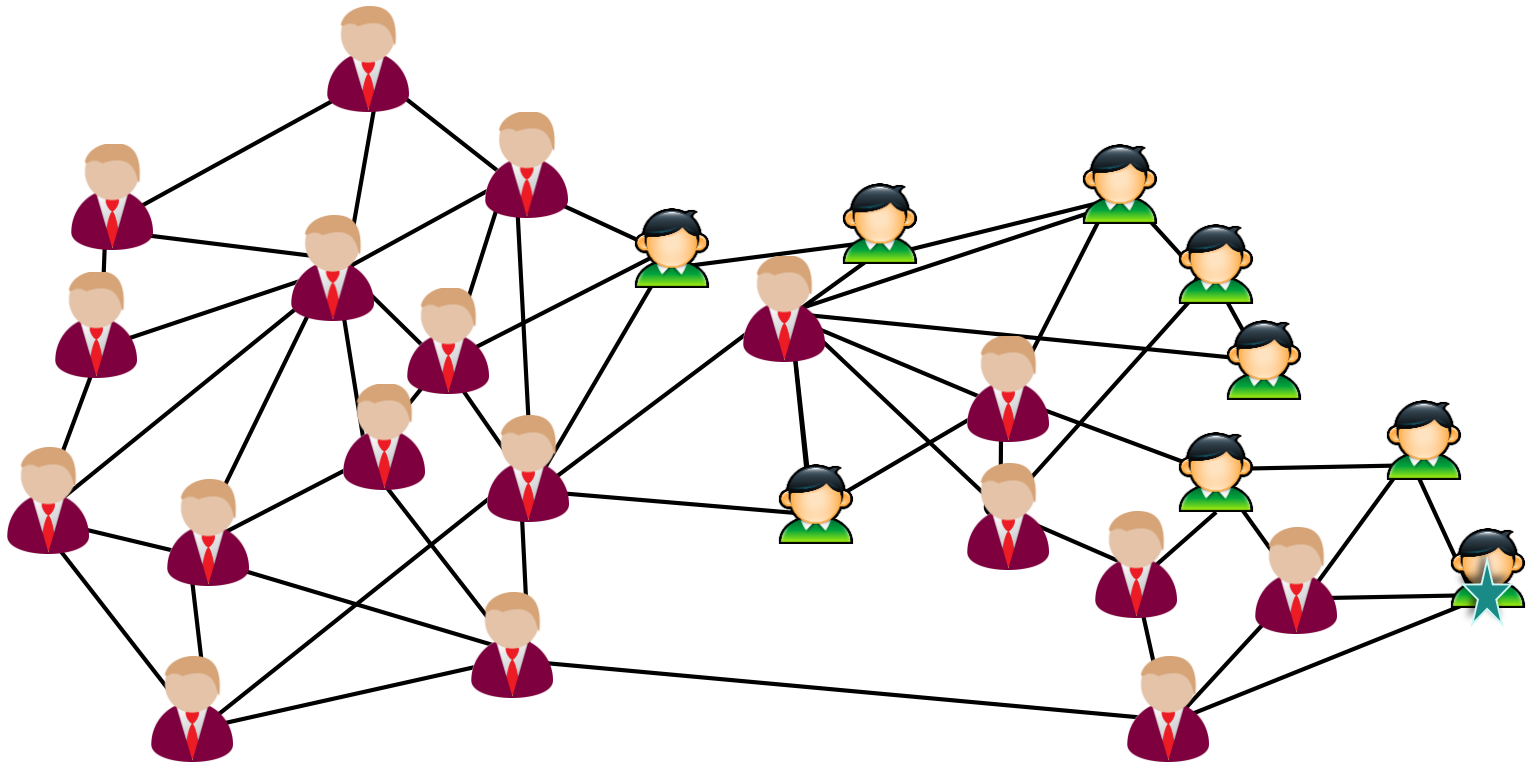
Localized-methods for Community Detection

Random Walks (RWs):



Localized-methods for Community Detection

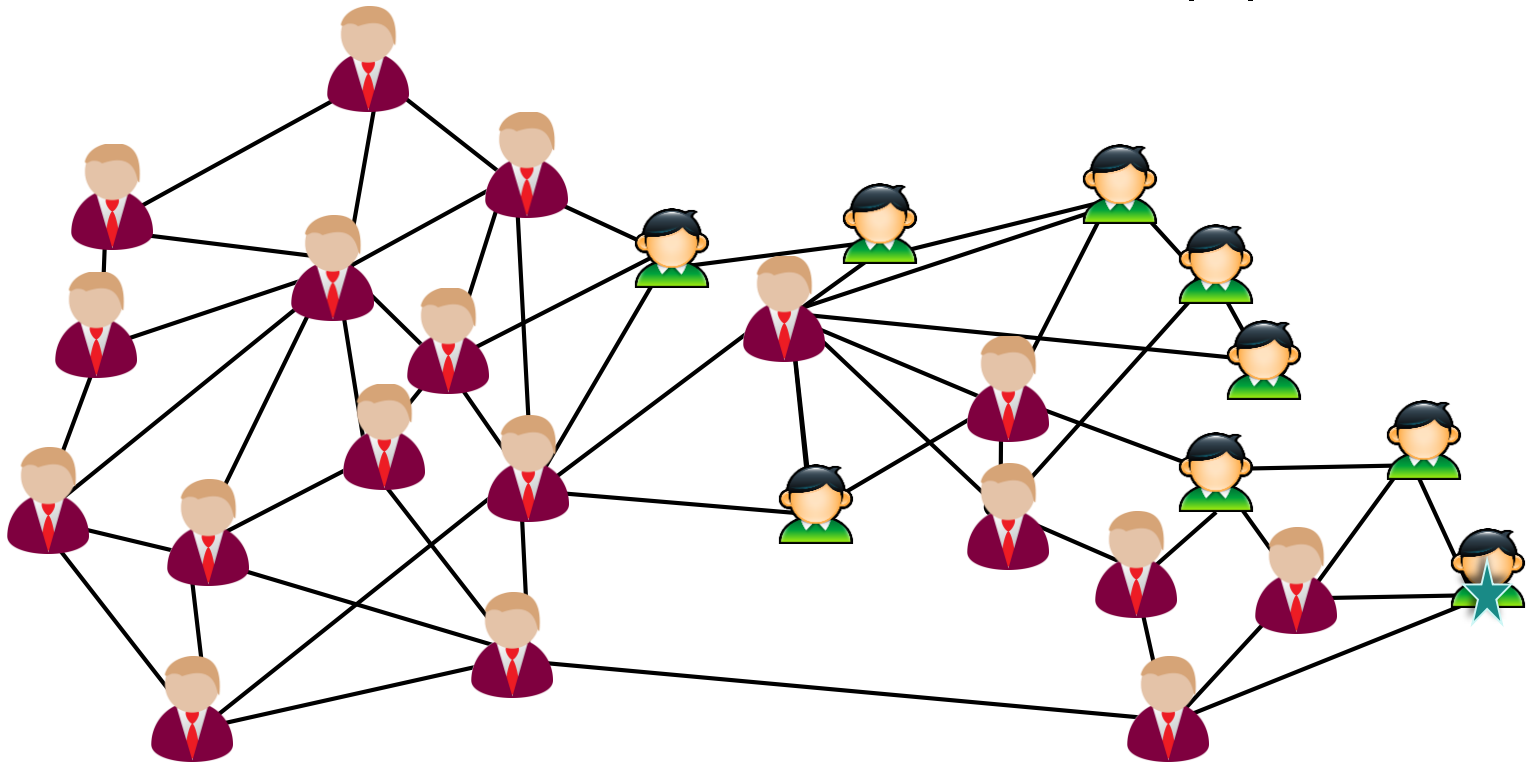
Random Walks (RWs):



Localized-methods for Community Detection

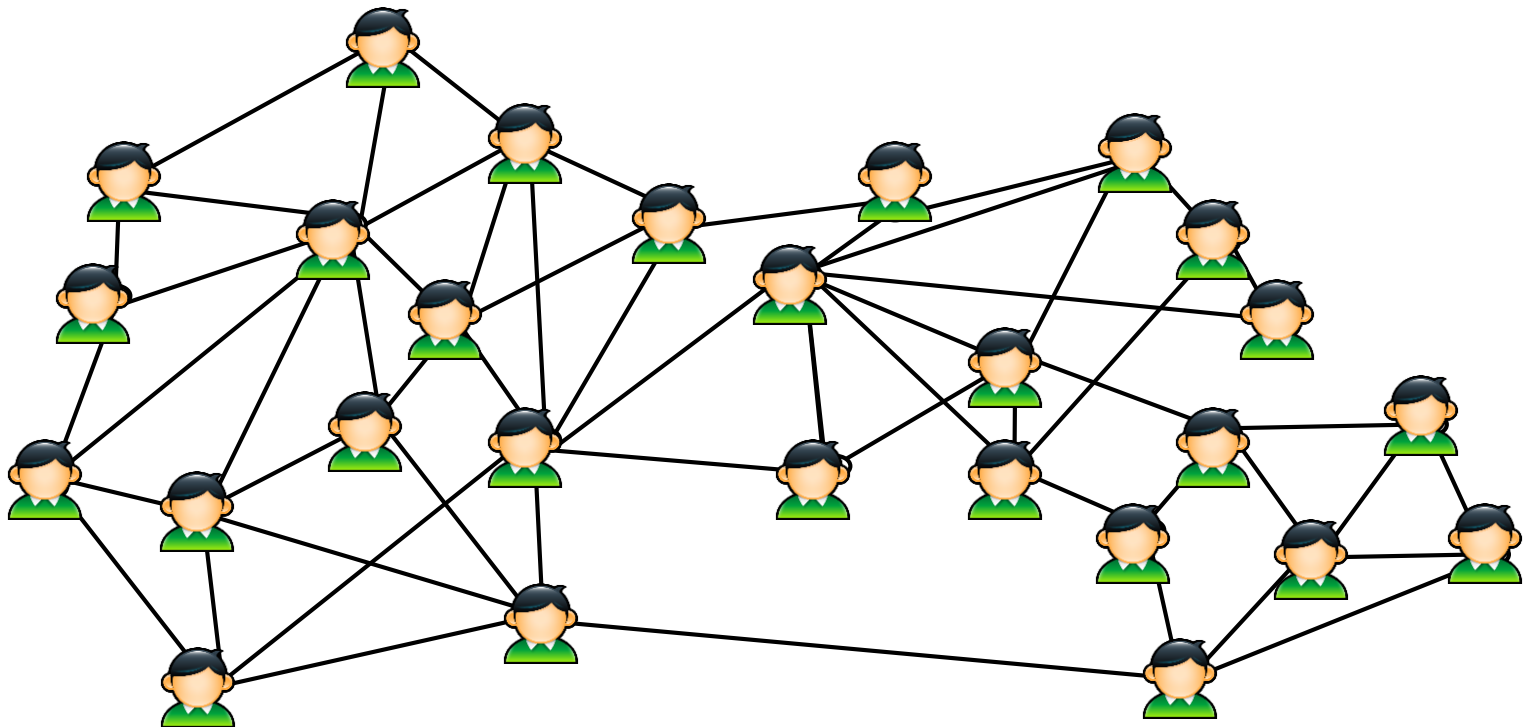
Random Walks (RWs):

$$\pi(v) = \frac{d_v}{2|E|}$$



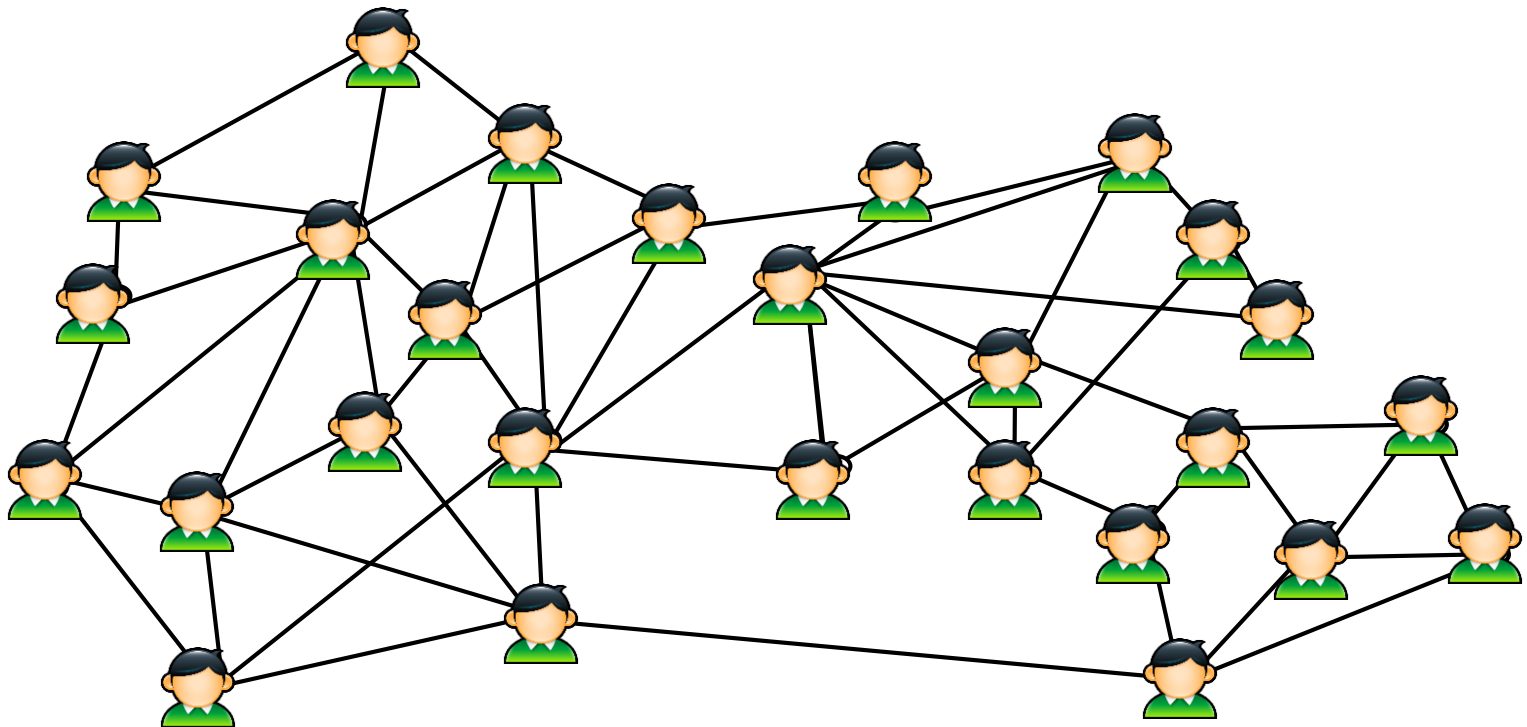
Localized-methods for Community Detection

DISCAN'18, Stockholm, Sweden.



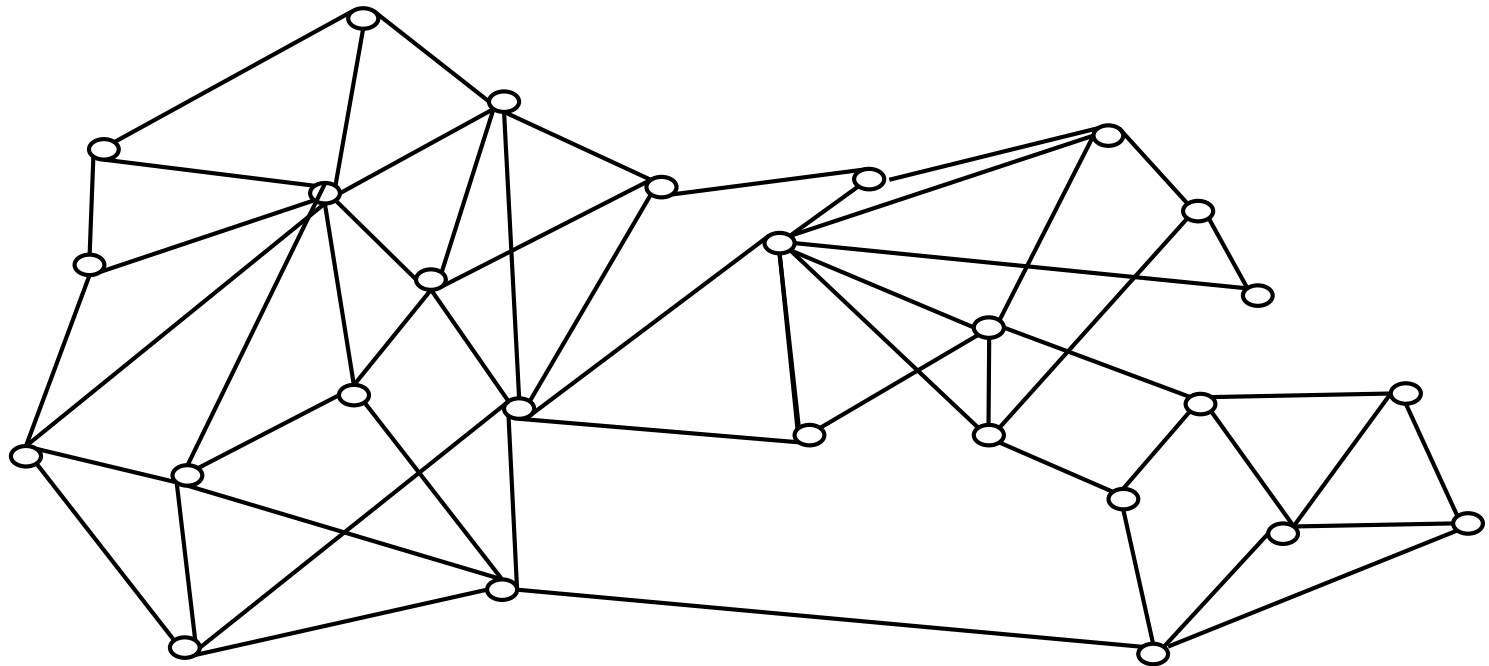
Localized-methods for Community Detection

Diffusion:



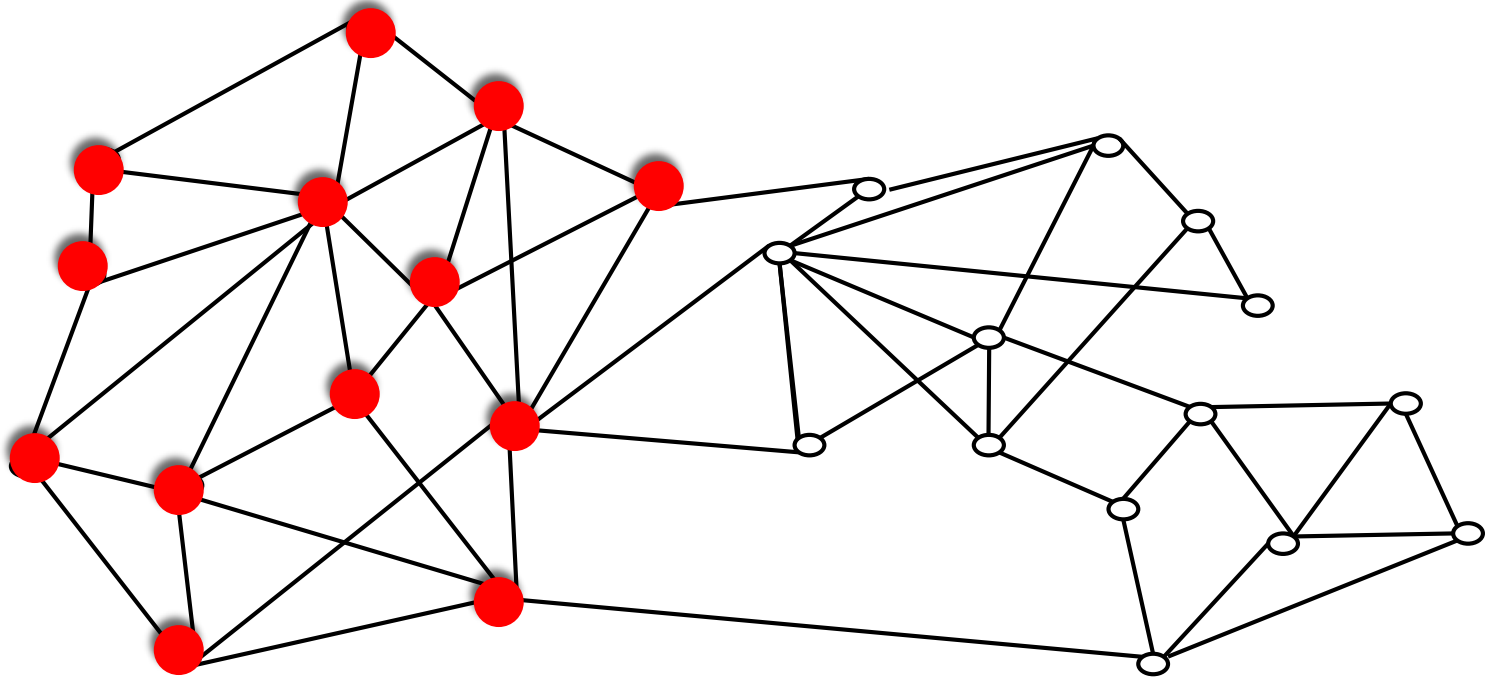
Localized-methods for Community Detection

Diffusion:



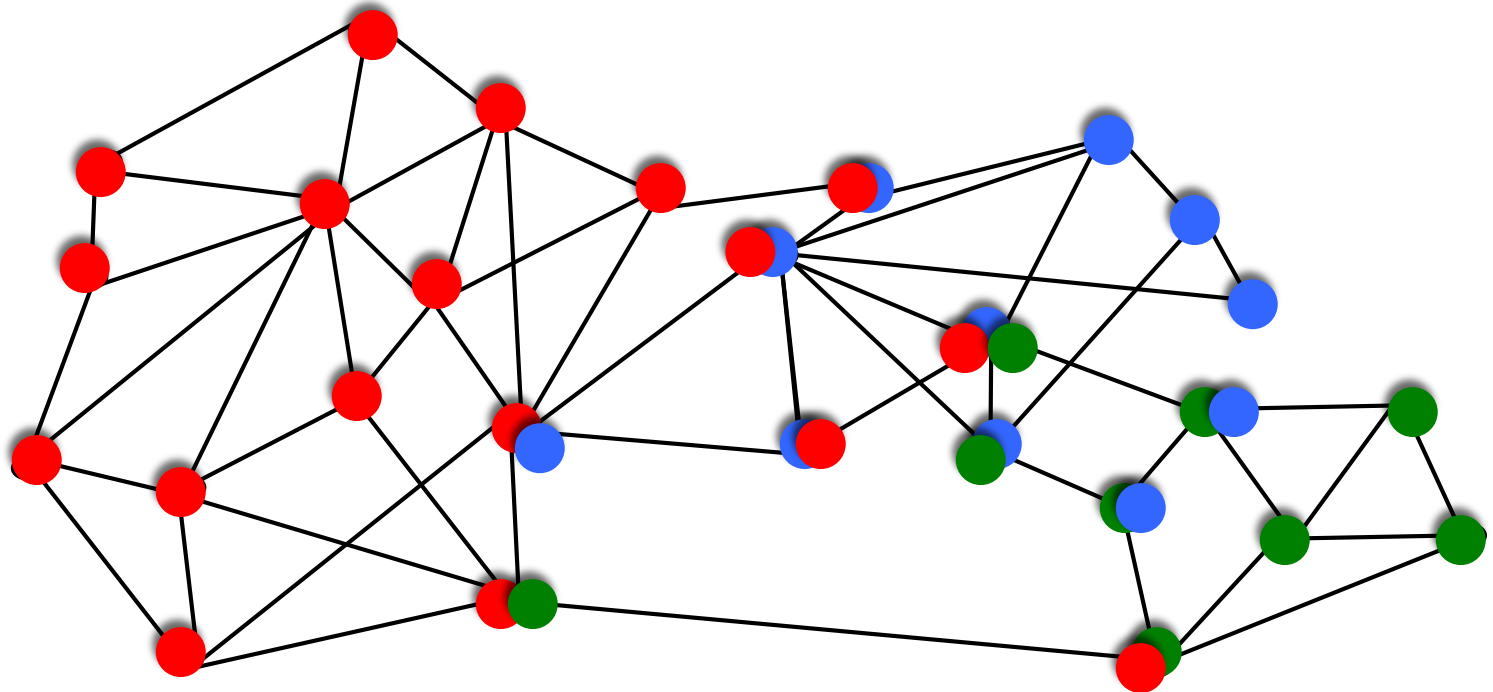
Localized-methods for Community Detection

Diffusion:



Localized-methods for Community Detection

Diffusion:

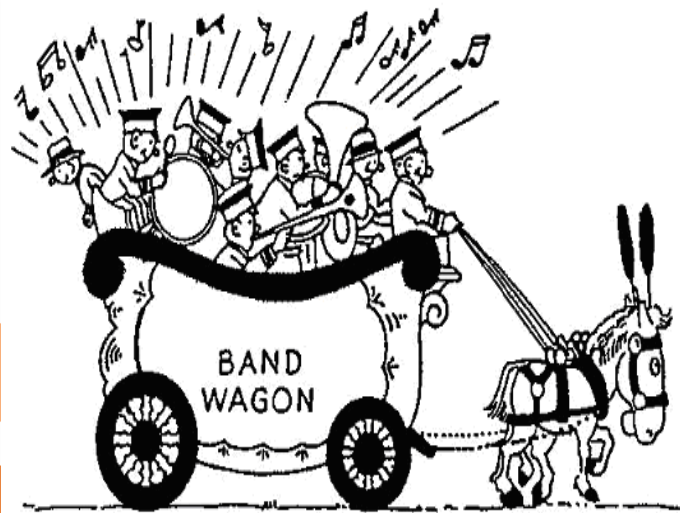


Bandwagon Effect?

DISCAN'18, Stockholm, Sweden.

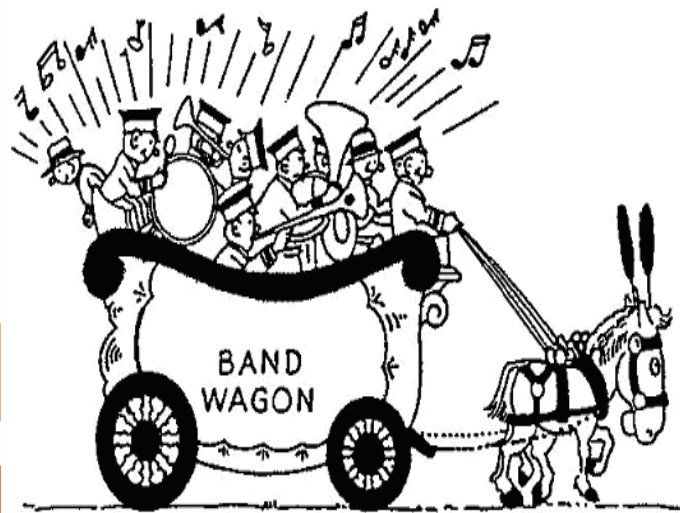
Bandwagon Effect?

DISCAN'18, Stockholm, Sweden.



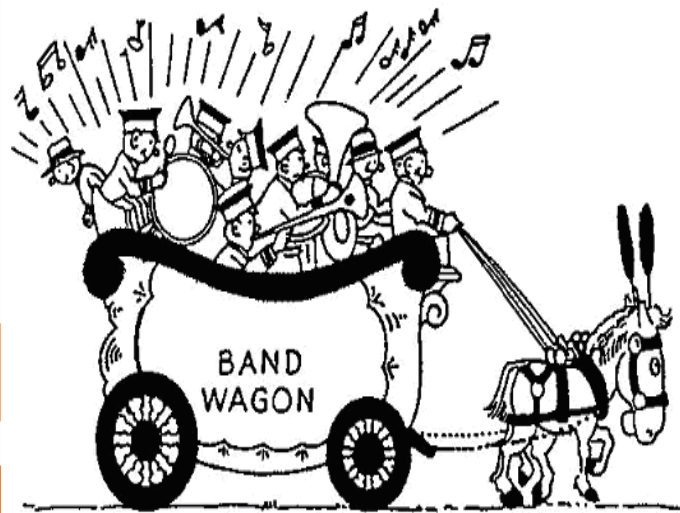
Bandwagon Effect?

- It is a **cognitive bias** that refers to the tendency to follow other people just because they are the **majority**.



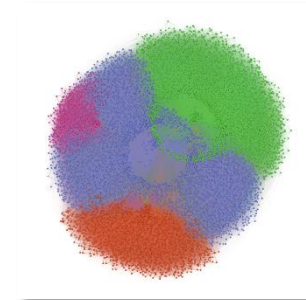
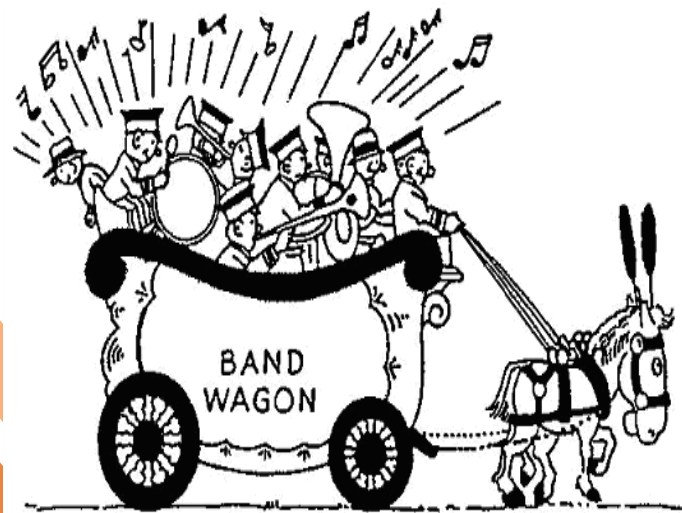
Bandwagon Effect?

- It is a **cognitive bias** that refers to the tendency to follow other people just because they are the **majority**.
- Heterogenous Community Size!



Bandwagon Effect?

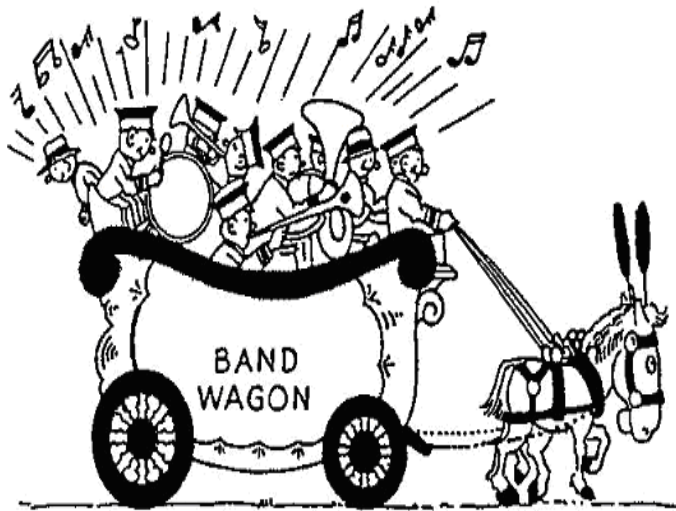
- It is a **cognitive bias** that refers to the tendency to follow other people just because they are the **majority**.
- Heterogenous Community Size!



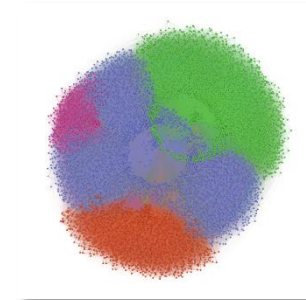
Ground-truth

Bandwagon Effect?

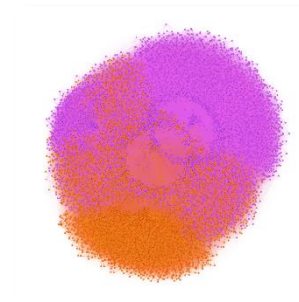
- It is a **cognitive bias** that refers to the tendency to follow other people just because they are the **majority**.



- Heterogenous Community Size!



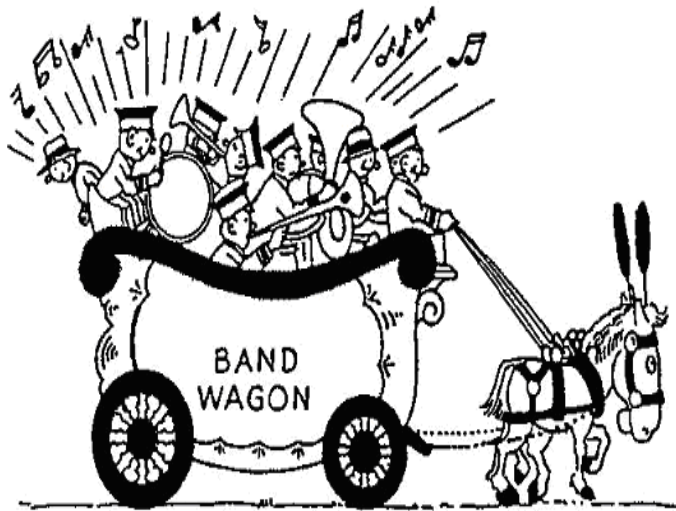
Ground-truth



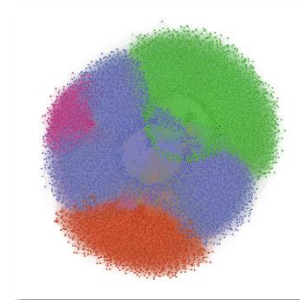
Resolution limit

Bandwagon Effect?

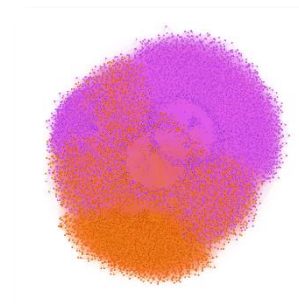
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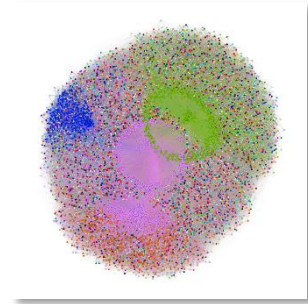
- Heterogenous Community Size!



Ground-truth

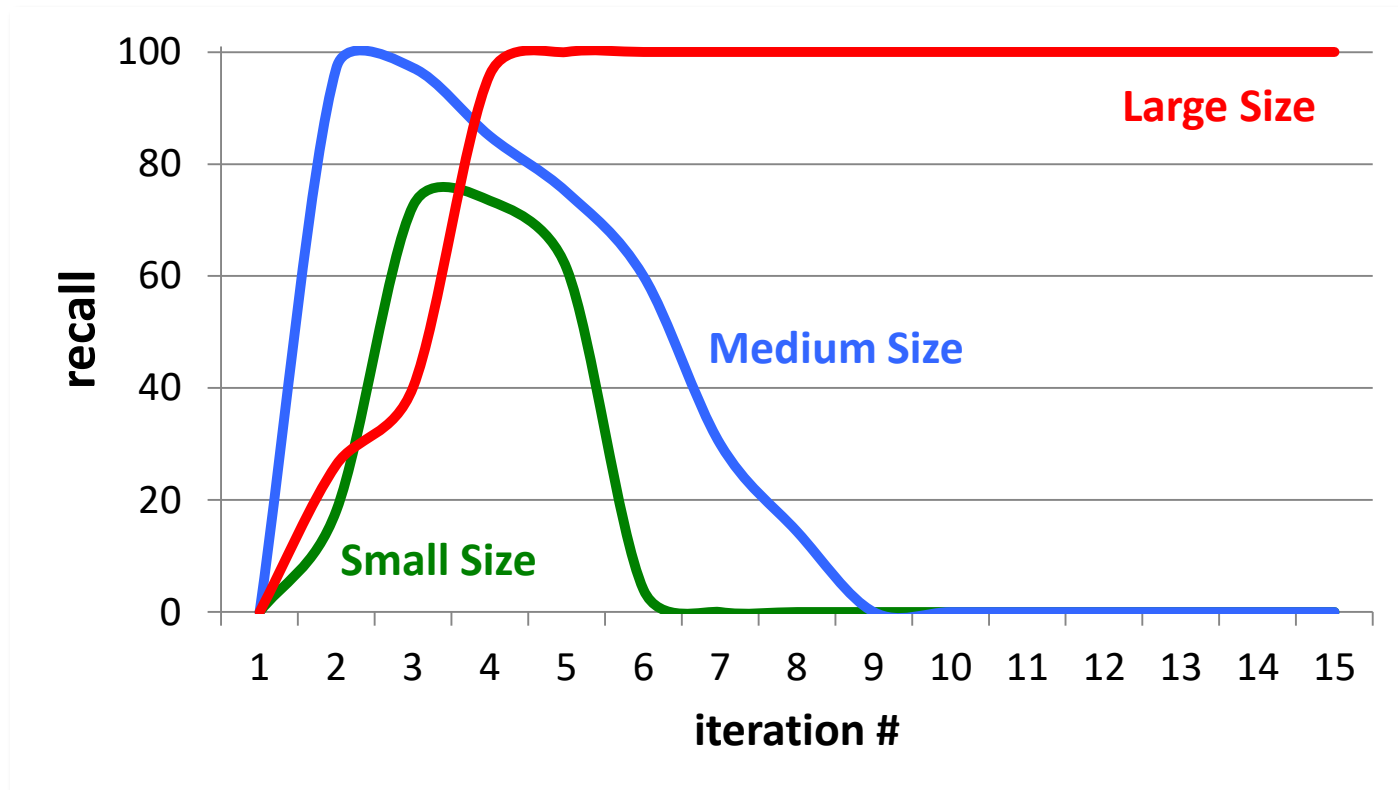


Resolution limit



Field-of-view limit

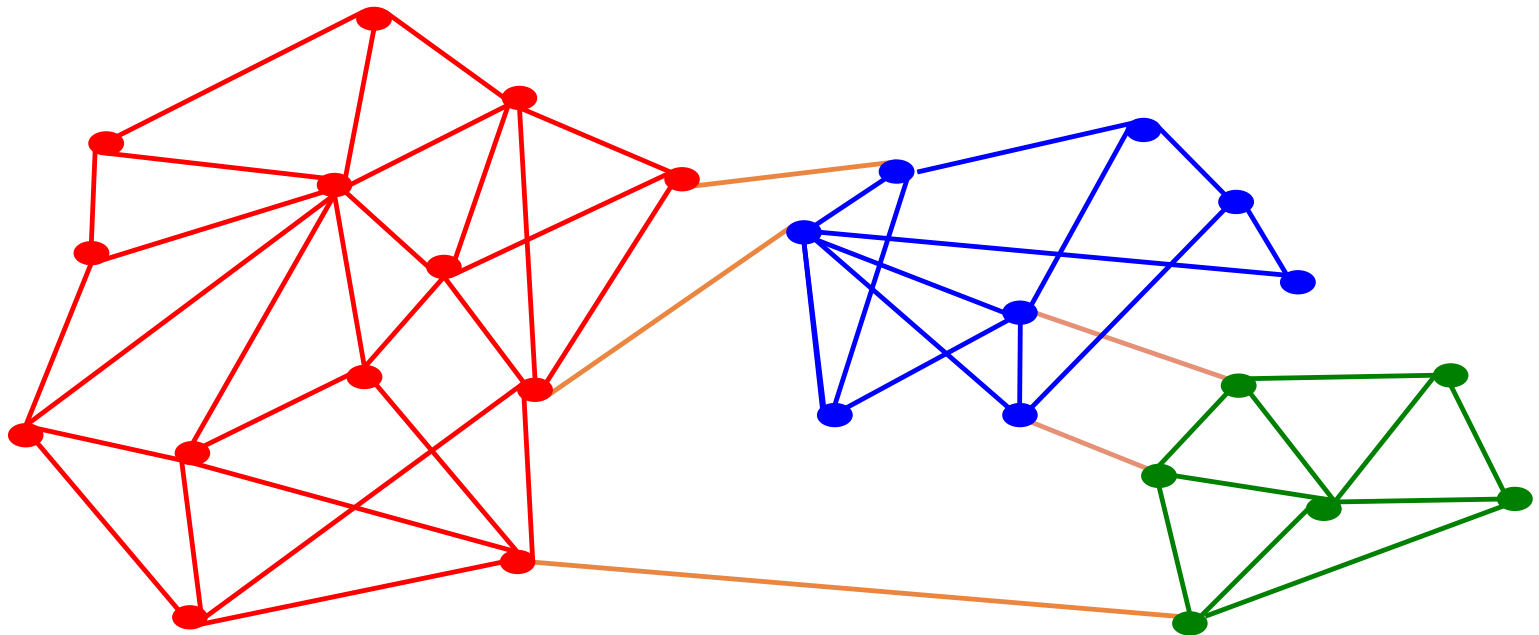
Heterogenous Community Size



Random Walks and Diffusion Performance!

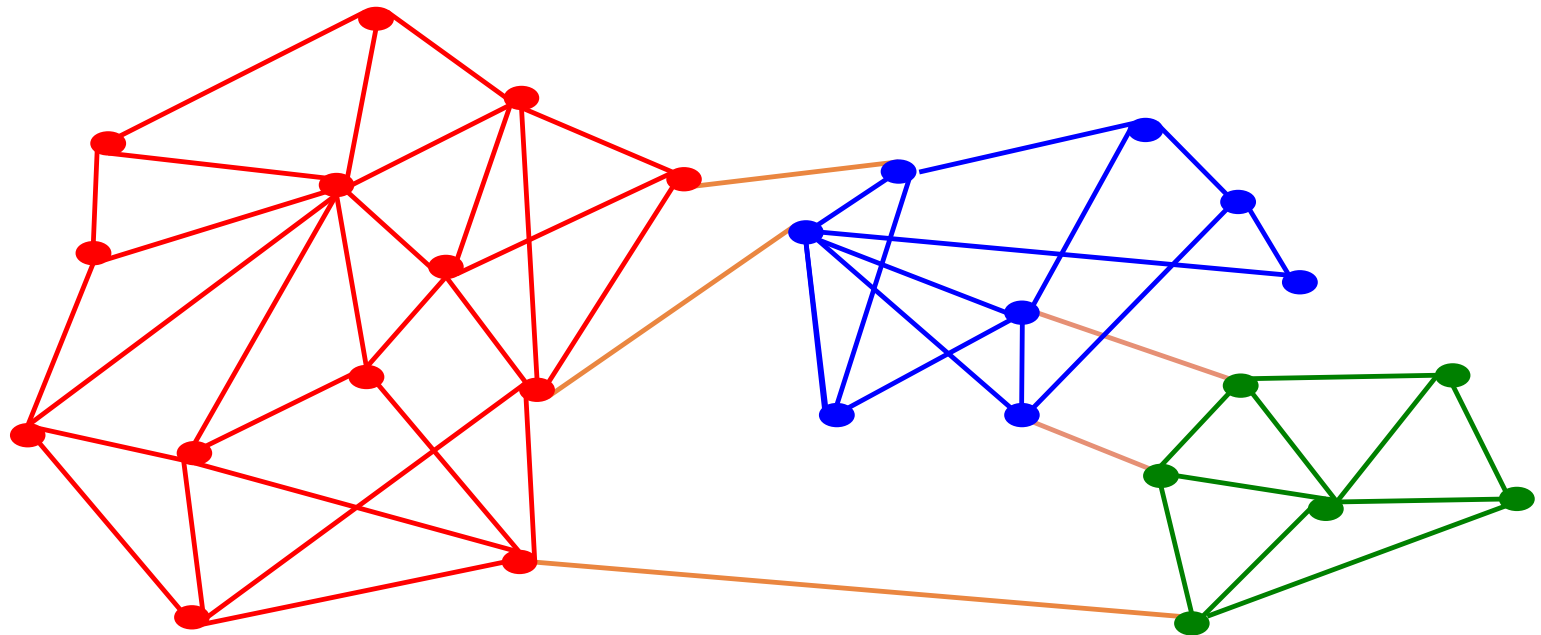
Adaptive Diffusion

DISCAN'18, Stockholm, Sweden.



Adaptive Diffusion

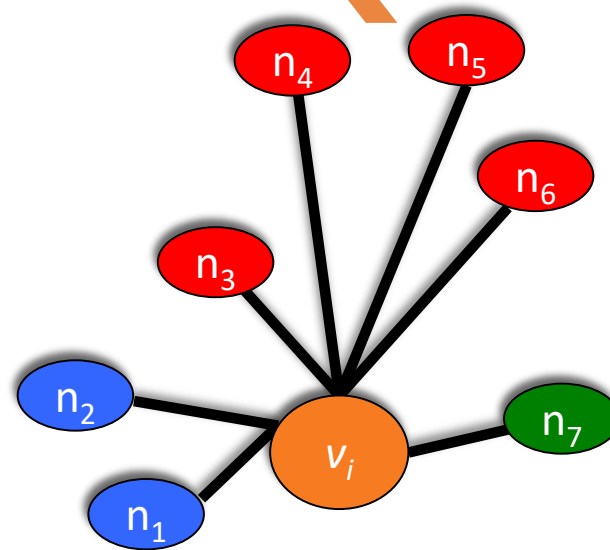
DISCAN'18, Stockholm, Sweden.



Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.



Colors

Units

Red

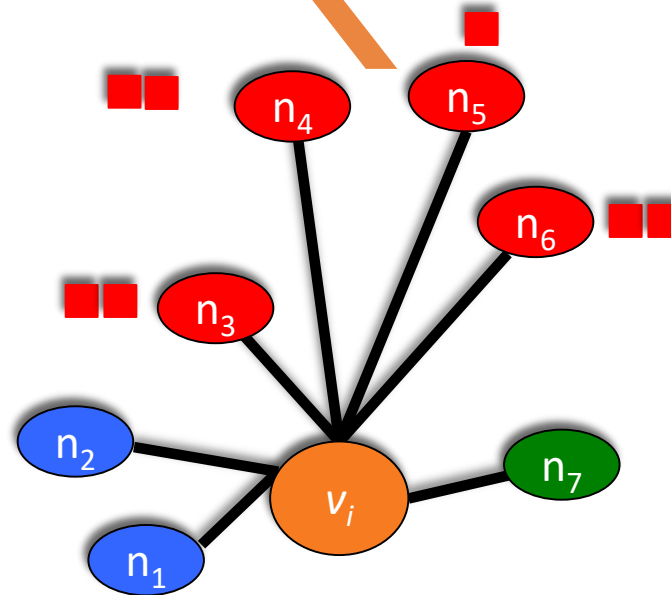
Green

Blue

Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.



Colors

Units

Red

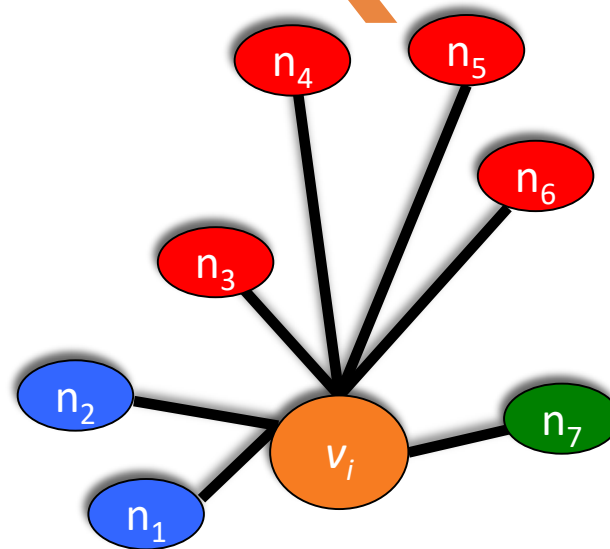
Green

Blue

Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.



Colors

Units

Red



Green



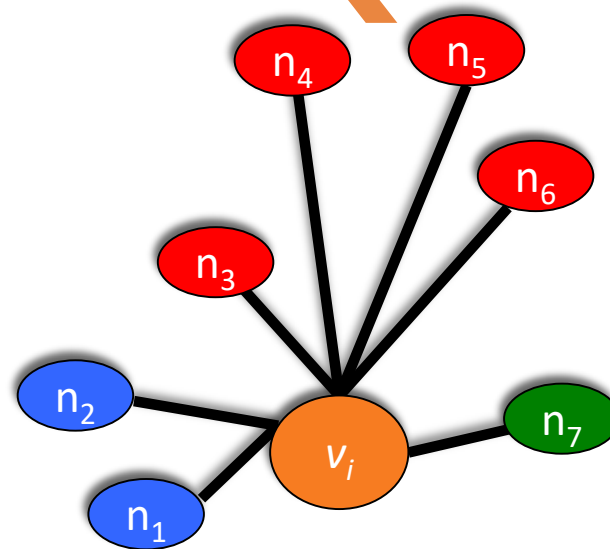
Blue



Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.



Colors

Units

Red



Green



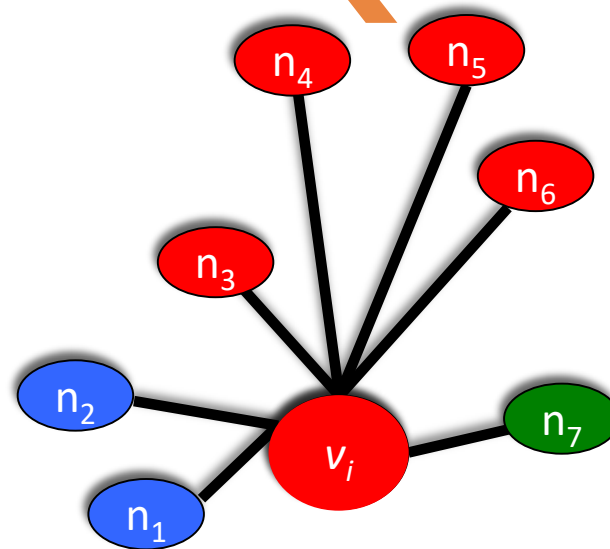
Blue



Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.
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Colors

Units

Red



Green



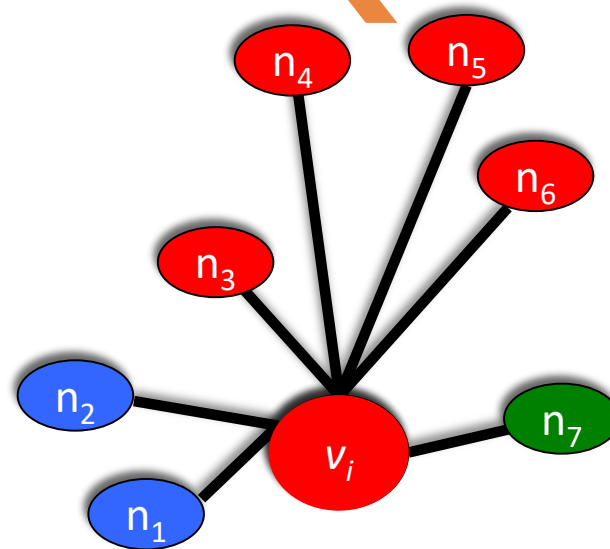
Blue






Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.
3. Send “all” of received color amount back to the neighbors.

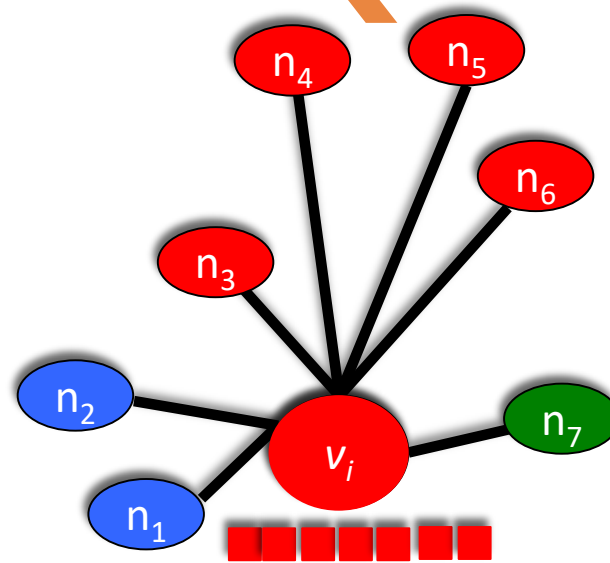


Colors	Units
Red	
Green	
Blue	

Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.
3. Send “all” of received color amount back to the neighbors.

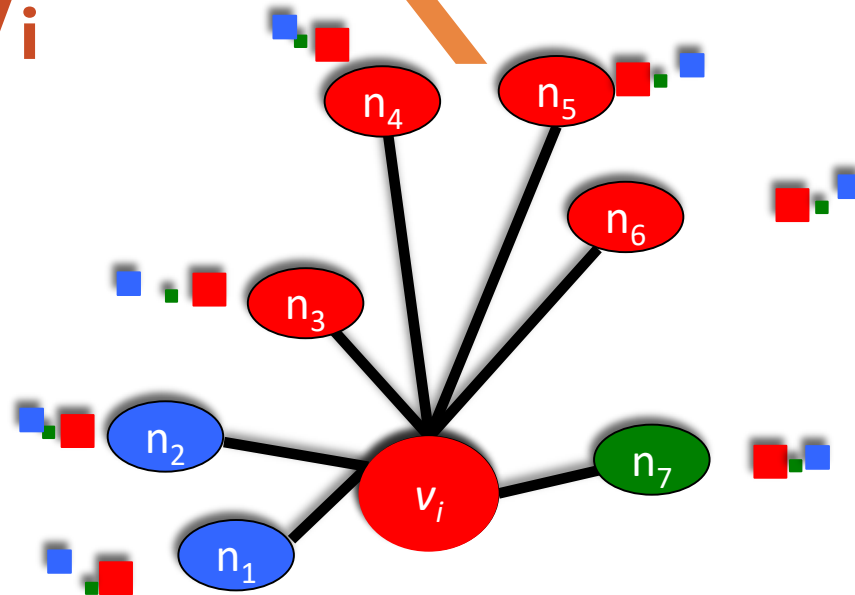


Colors	Units
Red	
Green	■
Blue	■ ■

Diffusion @ v_i

Regular diffusion steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.
3. Send **“all”** of received color amount back to the neighbors.

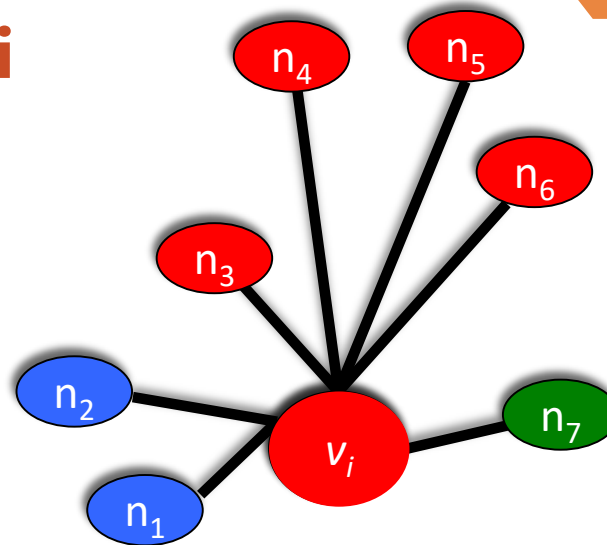





Colors	Units
Red	
Green	
Blue	

Stateful Diffusion (Stad) @ v_i

Stad steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.
3. Slow down diffusion speed: send “**some**” of received color amount back to the neighbors.

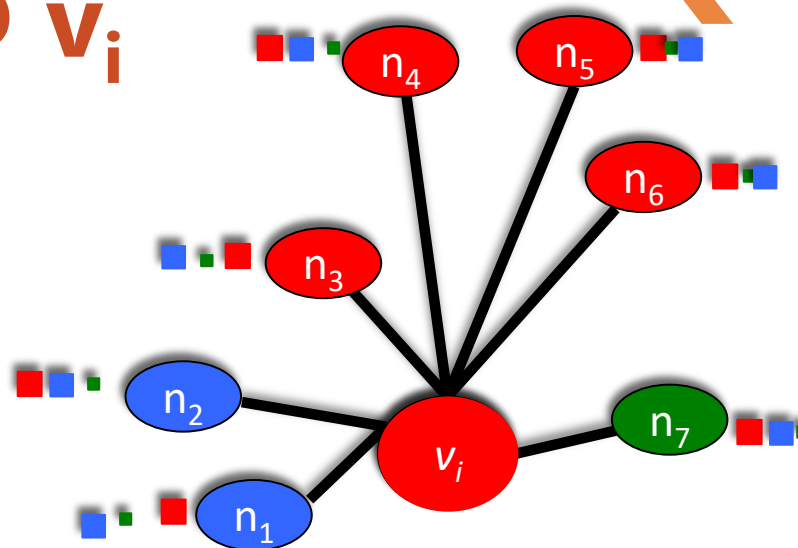


Colors	Units
Red	
Green	
Blue	

Stateful Diffusion (Stad) @ v_i

Stad steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.
3. Slow down diffusion speed: send “**some**” of received color amount back to the neighbors.

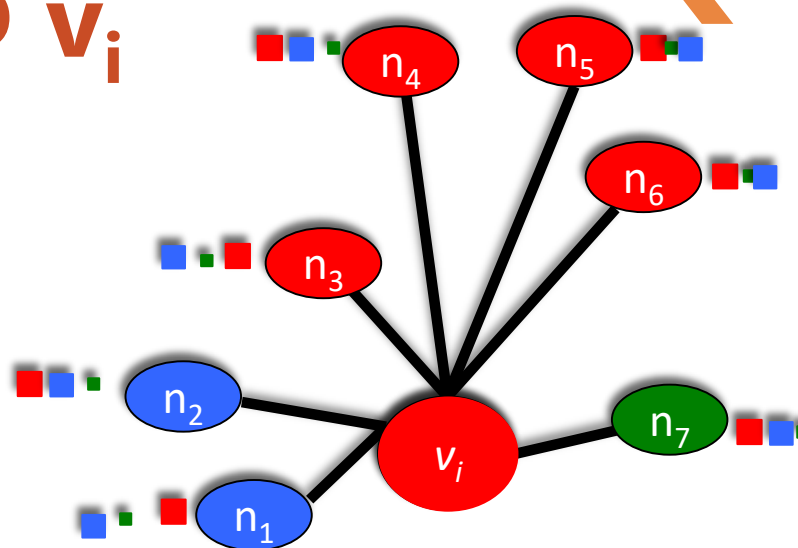


Colors	Units
Red	■ ■
Green	
Blue	

Stateful Diffusion (Stad) @ v_i

Stad steps:

1. Receive colors from neighbors.
2. Decide membership by selecting the dominant color.
3. Slow down diffusion speed: send “**some**” of received color amount back to the neighbors.



Colors

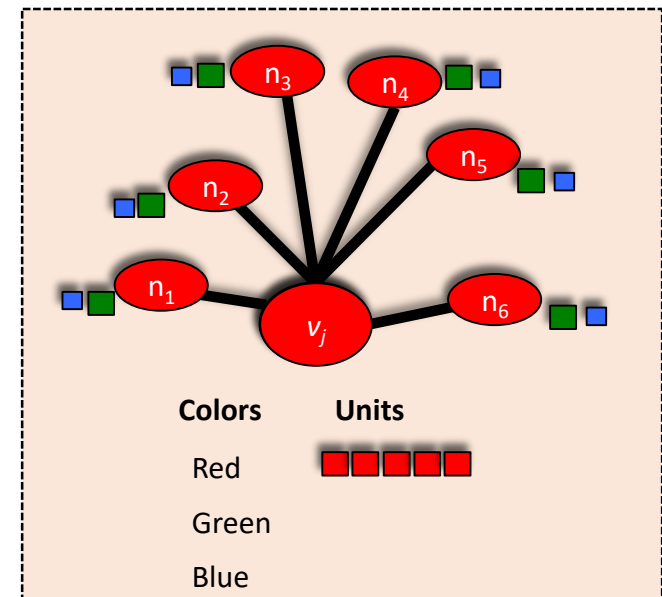
Red

Green

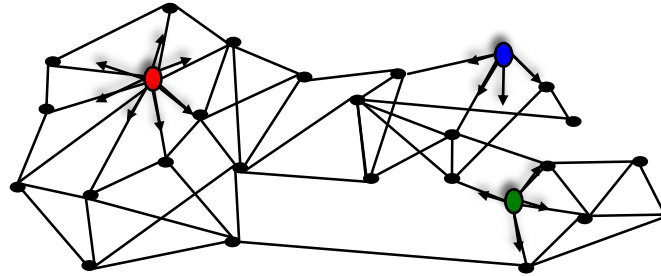
Blue

Units

■ ■



Stad: Fixed vs Adaptive Diffusion



(a) Initialization by selecting a set of random seeds

Iterations:

(1)

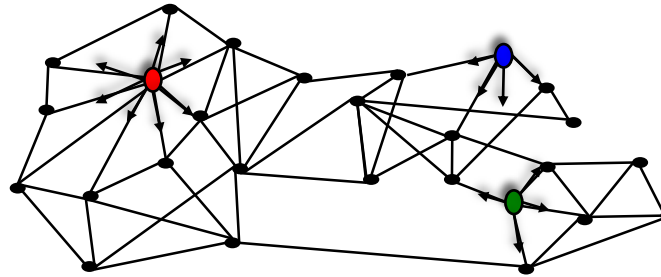
(2)

(3)

(b) Diffusion

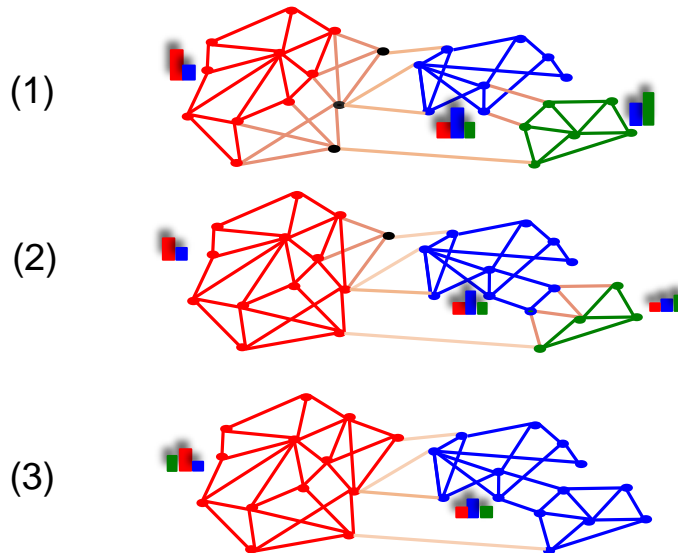
(c) Stad

Stad: Fixed vs Adaptive Diffusion



(a) Initialization by selecting a set of random seeds

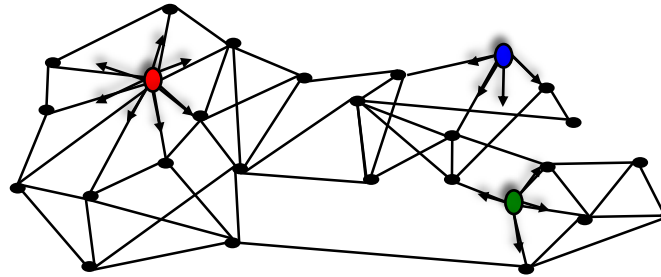
Iterations:



(b) Diffusion

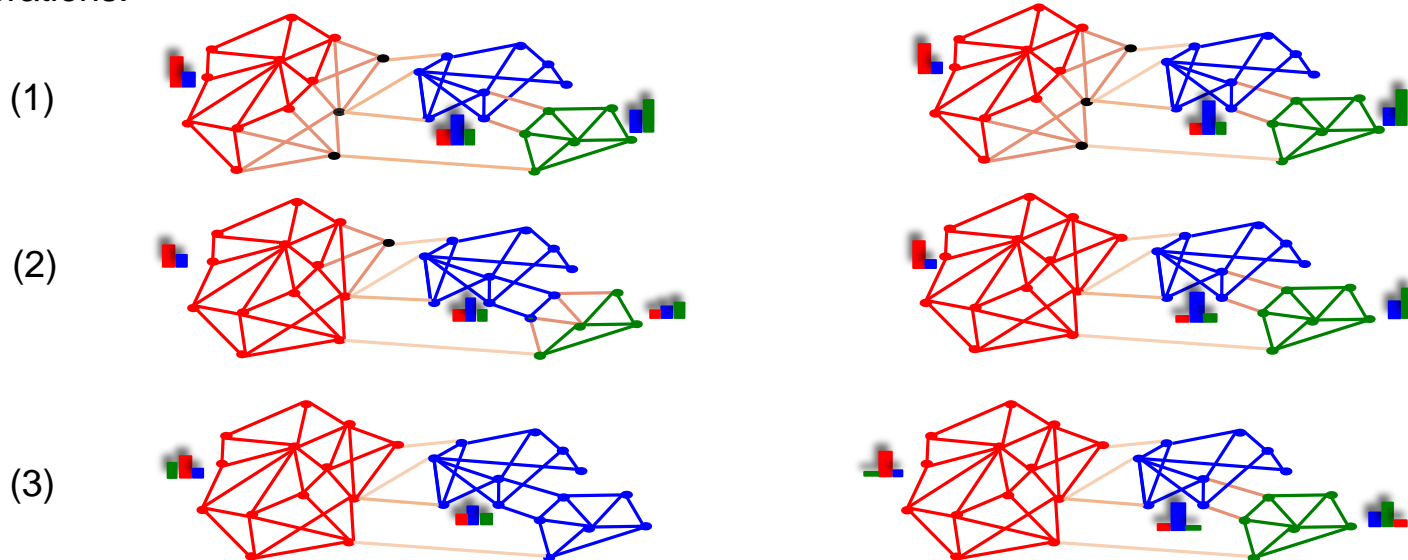
(c) Stad

Stad: Fixed vs Adaptive Diffusion



(a) Initialization by selecting a set of random seeds

Iterations:

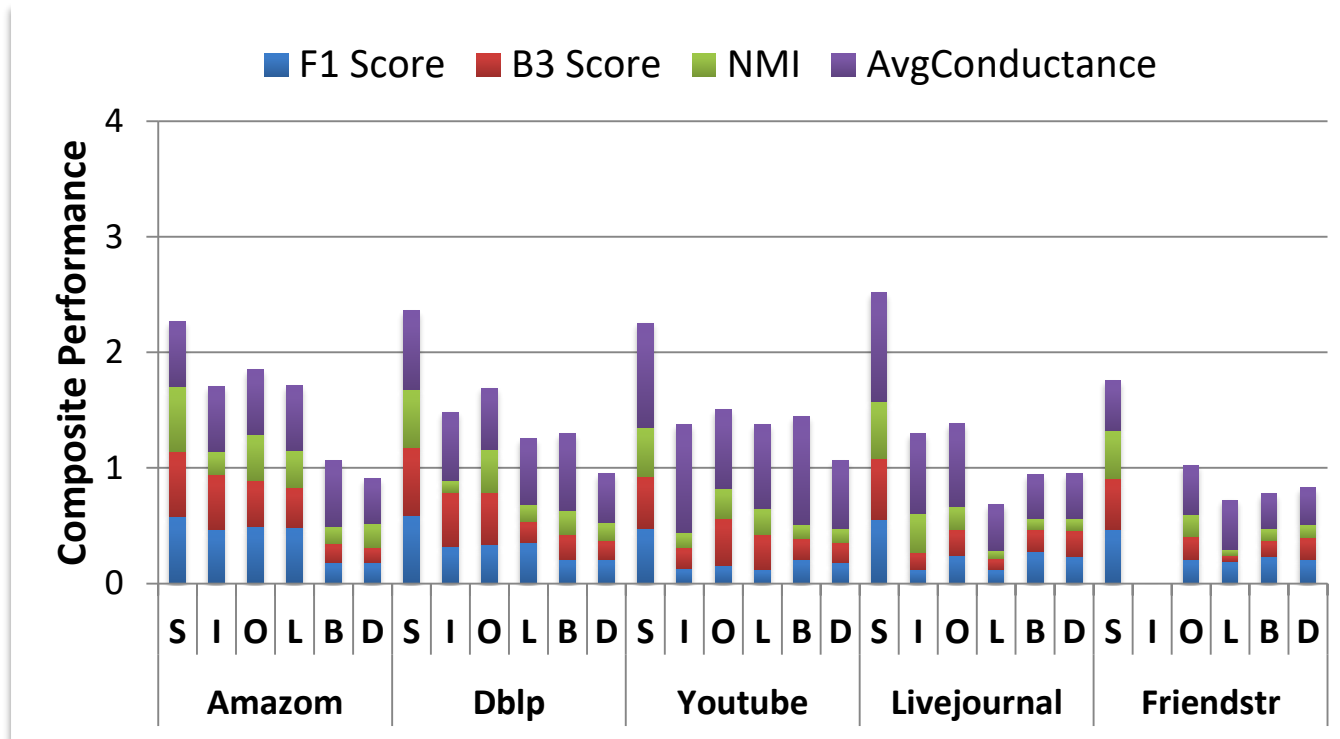


(b) Diffusion

(c) Stad

Experimental Results

Accuracy compared to baseline approaches



Performance comparison with the-state-of-the-art community detection approaches:

Stad (S), Infomap (I), Oslom (O), Louvain (L), Bigclam (B), Diffusion (D).

Conclusion

Decentralized Community Detection

Stateful Diffusion

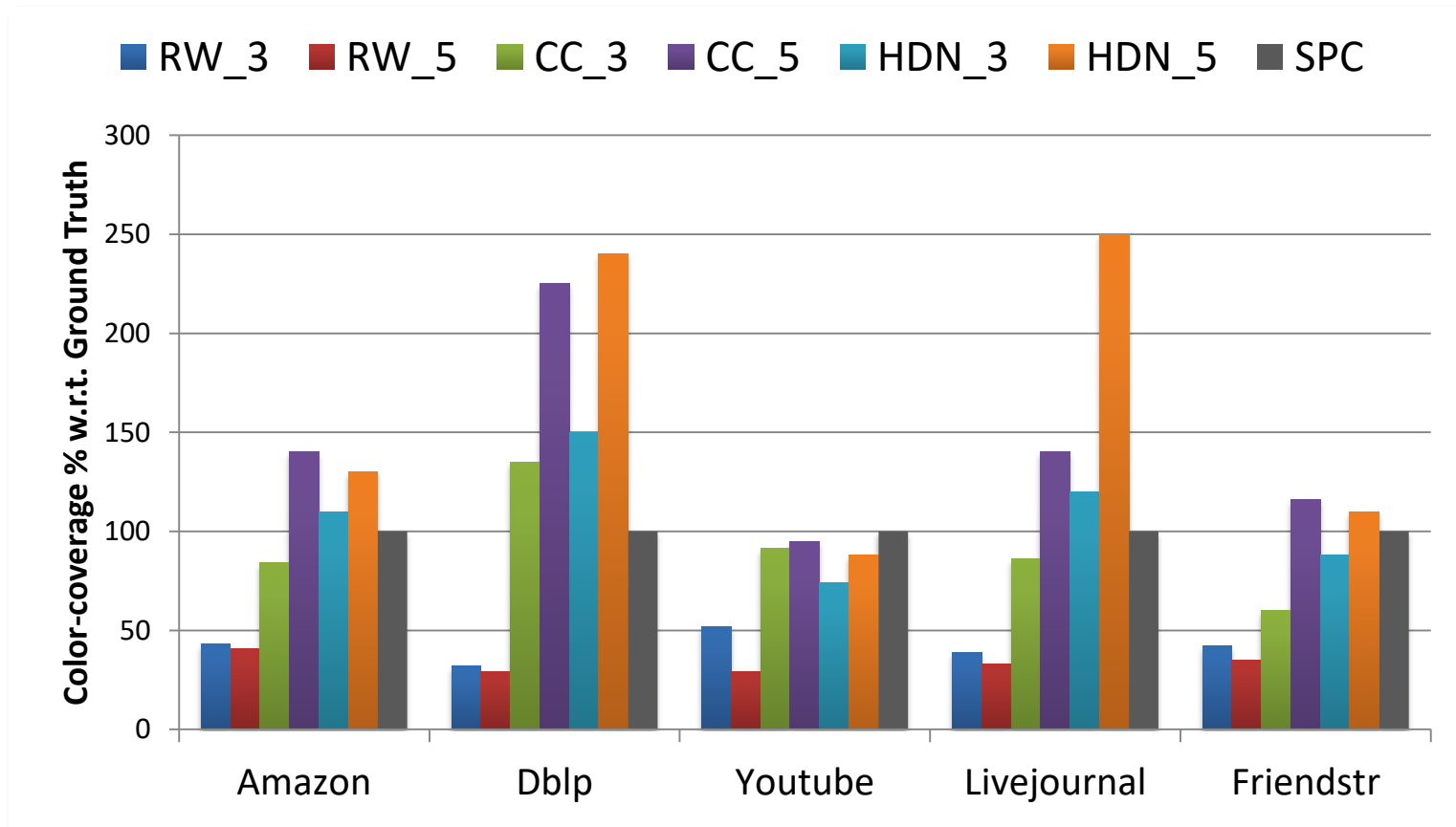
- Extracting communities with heterogenous size distribution.
- Performing optimization at node level while handling each community flow independently.
- Extracting disjoint as well as multiple community membership(s).

Future Work

- Adapting Stad to directed graphs that would require handling the challenges of edge directionality (dead-ends, cycles, etc.)

Experimental Results

Stad: Selecting seeds for color initialization



Thanks and questions are very welcomed 😊