



# Structural Group Unfairness

## Measurement and Mitigation by means of the Effective Resistance

Adrian Arnaiz-Rodriguez



Georgina Curto

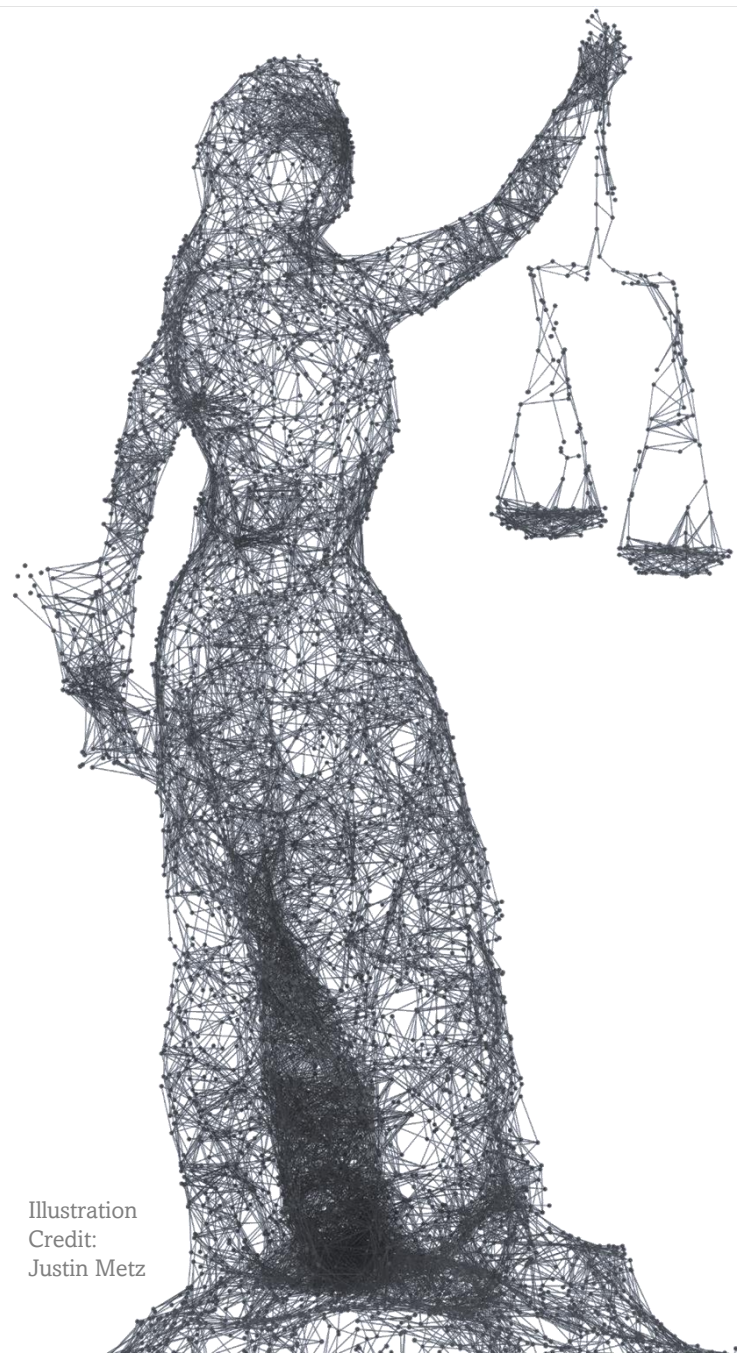


Nuria Oliver



Ethics of Data Analytics and AI - University of Notre Dame – Guest Lecture

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# Main takeaways

Social Networks are very useful to analyze social dynamics or make predictions about people and their relationships

Specific [Arnaiz-Rodriguez, Curto and Oliver 2024]:

Your position on a **social network** defines the access and control of information you have



Social Capital

It is not easy to **quantify it** and intricate to **align the literature** in sociology and graph theory/fairness



Quantitative measures of social capital sociology-aligned based on graph theory

There are some individual and group inequalities and not trivial to fix



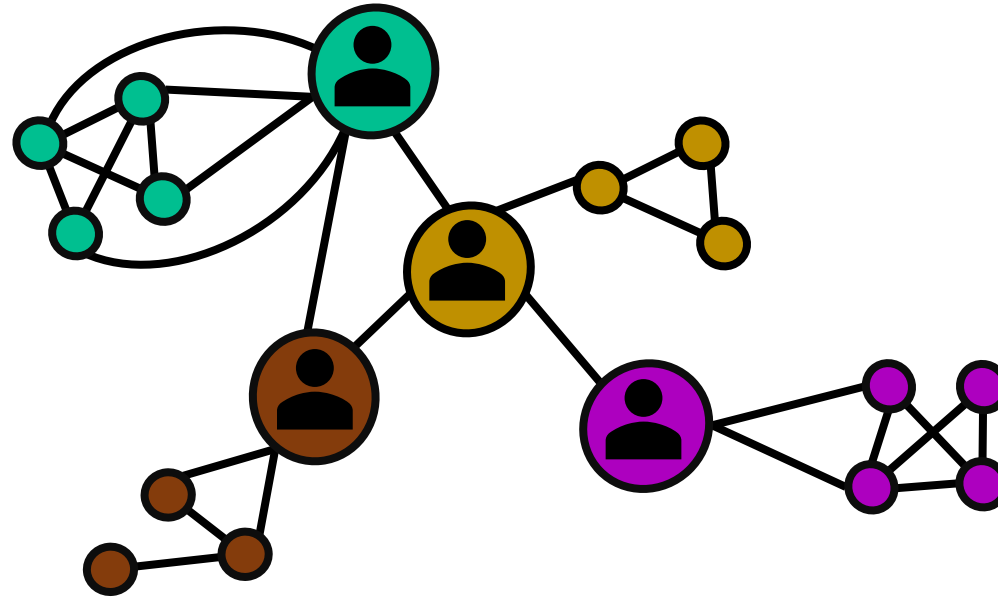
Fix it using an edge augmentation strategy based on the **weakest links**

Arnaiz-Rodriguez, Adrian, Georgina Curto, and Nuria Oliver.

"Structural Group Unfairness: Measurement and Mitigation by means of the Effective Resistance." *Submitted*. (2024)

# Social Networks

Social Networks (or Graphs) are defined by people and interactions between them

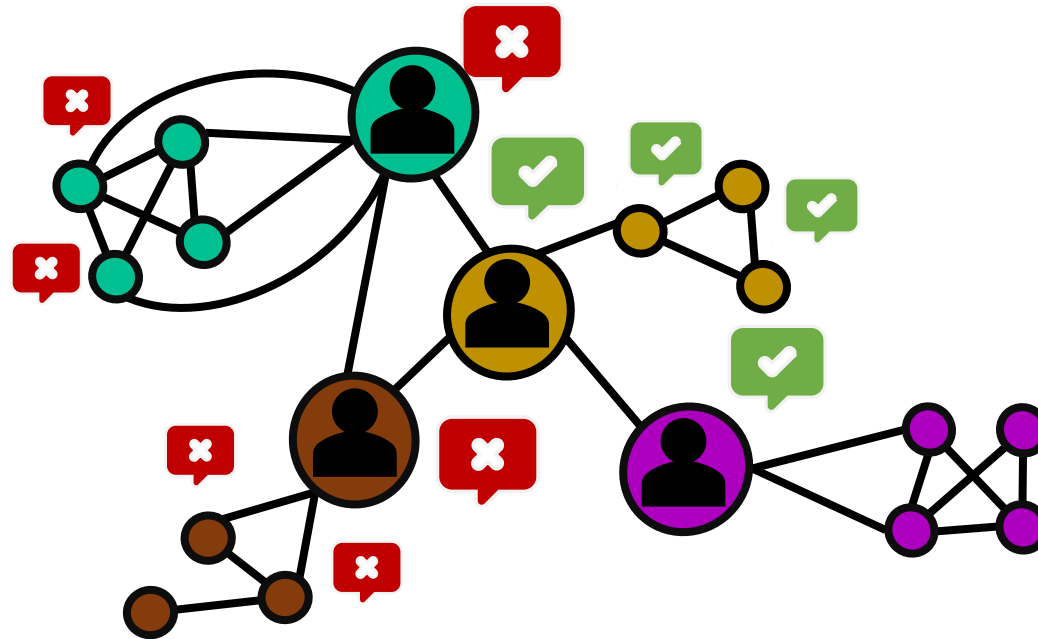


Formally, networks are defined as a set **nodes** (people) connected by **links** (interactions)

# Social Networks

Formally defined, we can analyze different social aspects:

- **Predictions about people**

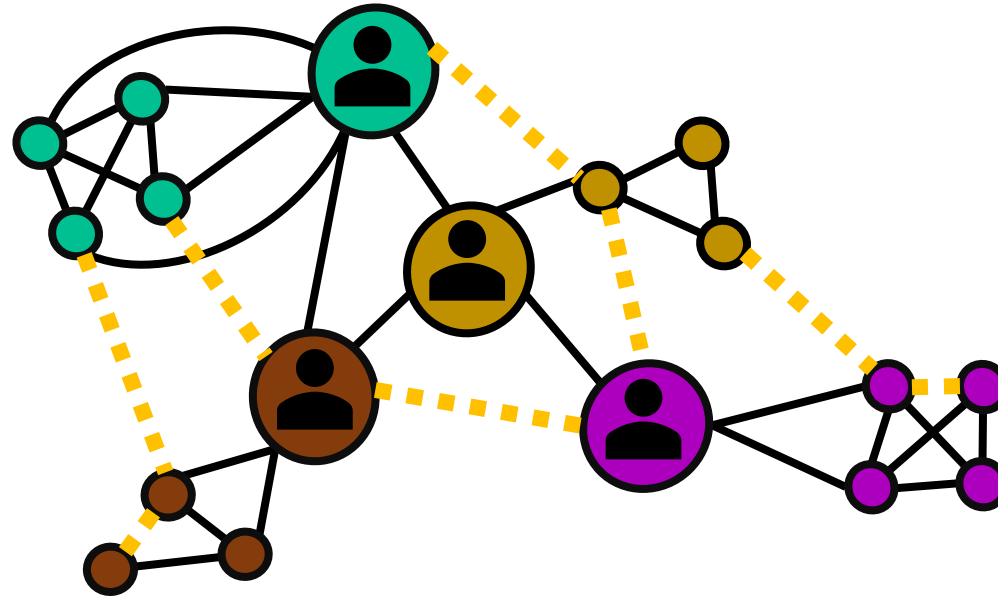


**Risk:** if there are segregated or discriminated communities might lead to biased decisions

# Social Networks

Formally defined, we can analyze different social aspects:

- **Predictions** about people or about their **relations** (Recommenders Systems)

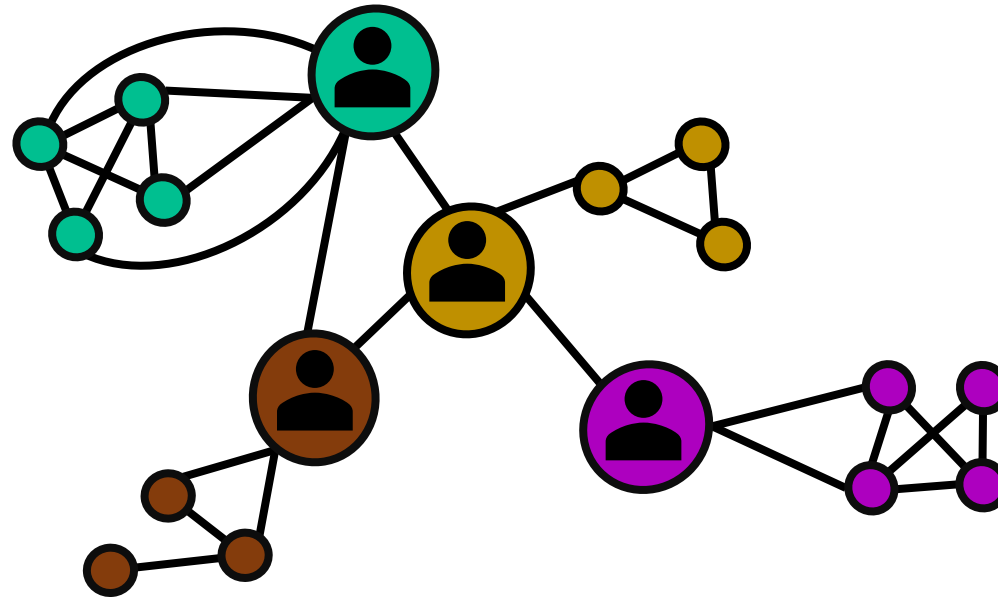


**Risk:** it might lead to polarization or discrimination (disparate information access) due to homophilic behavior

# Social Networks

Formally defined, we can analyze different social aspects:

- **Predictions** about people or about their relations
- **Analyze importance** of people, the relations, the overall flow of information, the social dynamics,...



Source: [Google](#)

Importance of  
people / links

Spread of  
information / virus

Communities /  
Polarization

Influence  
maximization

Social Dynamics

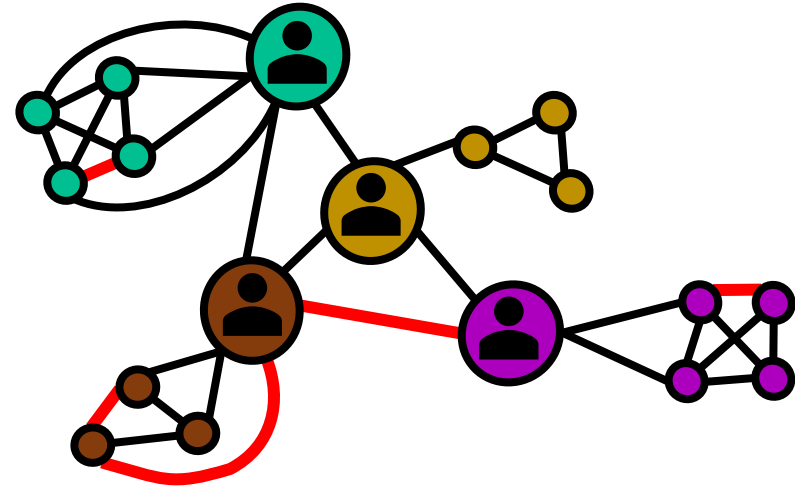
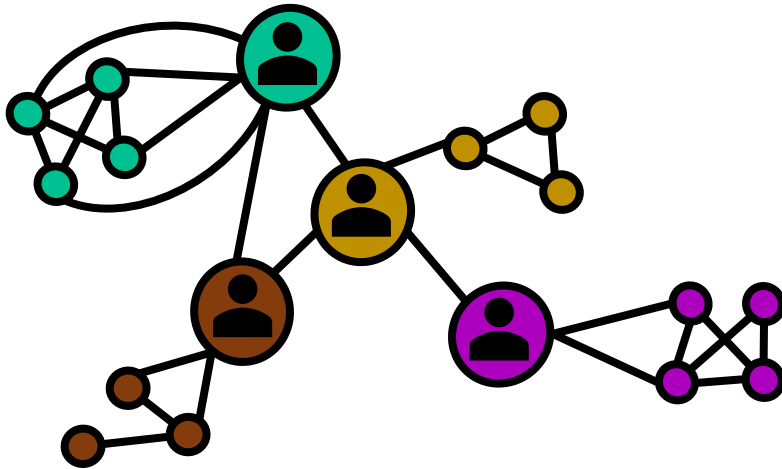
Prediction of  
properties or links

Information Flow

# Recommendation Systems

Naturally, people connect with each other by some kind of similarity [McPherson2001]

Same interests, ideas, hobbies...



Recommender Systems (RecSys) are based on the same

Create connections by user similarity based on features and connections

e.g.: if we share lot of friends but we are not already connected → suggest connection

# Challenges:

- Accurately and efficiently quantify polarization [Hohmann, 2023]
- Identify and quantify causes in the generation of the polarization [Santos, 2021]
- Identify new links to enforce diversity (destroy filter bubbles) [Masrour, 2020]
- Identify links that maximize the information flow in the network [Arnaiz-Rodriguez et al, 2024]

Hohmann, Marilena, et al. "Quantifying ideological polarization on a network using generalized Euclidean distance." *Science Advances* (2023)

Santos, Fernando P., et al. "Link recommendation algorithms and dynamics of polarization in online social networks." *PNAS* (2021)

Masrour, Farzan, et al. "Bursting the filter bubble: Fairness-aware network link prediction." *AAAI* (2020)

**Arnaiz-Rodriguez, Adrian, et al. "Structural Group Unfairness: Measurement and Mitigation by means of the Effective Resistance." *Submitted* (2024)**





# Structural Social Capital

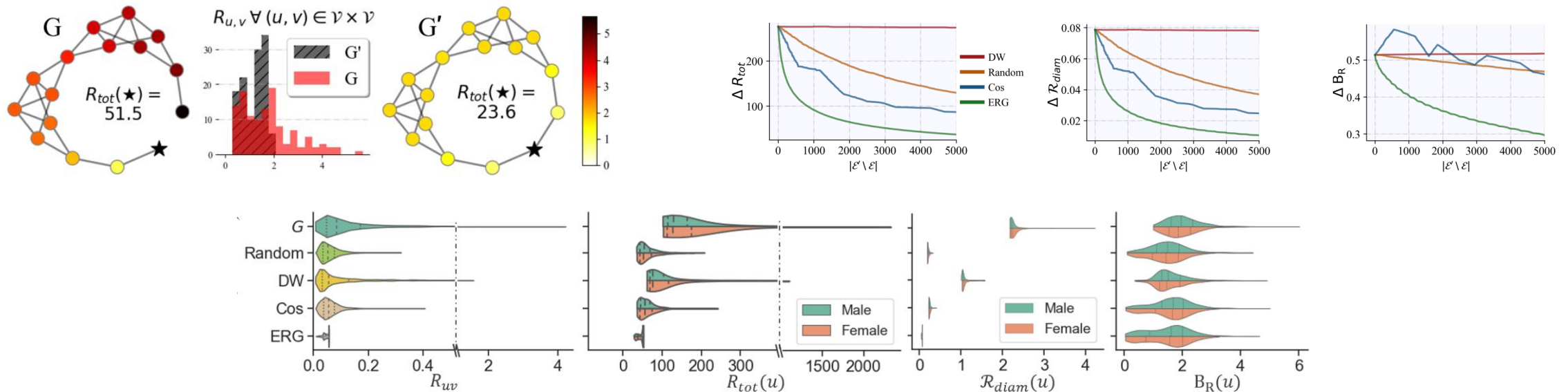
Effective Measurement and Mitigation of Information Access and Control disparities in Social Networks using Spectral Graph Theory

# Structural Group Unfairness: Measurement and Mitigation by means of the Effective Resistance

Adrian Arnaiz-Rodriguez  
adrian@ellisalicante.org  
ELLIS Alicante  
Spain

Georgina Curto  
gcurtore@nd.edu  
University of Notre Dame  
Notre Dame, USA

Nuria Oliver  
nuria@ellisalicante.org  
ELLIS Alicante  
Spain



<https://arxiv.org/pdf/2305.03223.pdf>

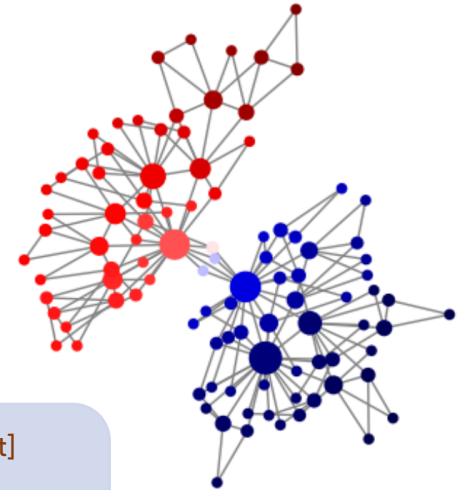
# What is Social Capital?

## Social Capital

- Power derived from your **connections** on the network
- Networks, relationships, and **norms of trust and reciprocity** within a community or society that facilitate cooperation and collective action

## Structural Social Capital

- Social capital derived from the **position** of a person in a network [Burt]
- **Position determines the control and access of information**  
*Job and financial opportunities, education information, health information  
power to spread out the word or influence others,...*



The flow of information in the network determines the social capital of individuals  
Some of them will be more central and important for the flow

# Unfairness in Social Capital

Different people have different structural role → **Individual** disparities wrt information access and control  
[Bashardoust 2023]

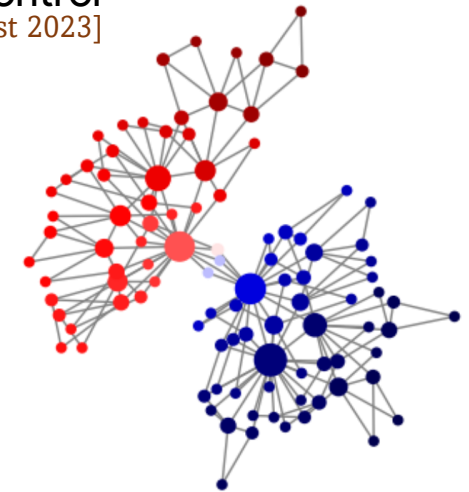
[Arnaiz-Rodriguez et al, 2024]

**Groups** are distributed in the network in different manners

→ different flow of information on different groups

→ different social capital (Unfairness in Social Capital)

→ **Structural Group Unfairness**



Accurate  
Measurement

Effective Mitigation

Burt, Ronald S . The network structure of social capital. *Research in organizational behavior* (2000)

Bashardoust, A. Reducing Access Disparities in Networks using Edge Augmentation. *FAccT* (2023)

**Arnaiz-Rodriguez, Adrian, et al. Structural Group Unfairness: Measurement and Mitigation by means of the Effective Resistance. (2024)**

# Measure Group Social Capital

**Information flow** in the network determines the **node's information access and control**

- IF usually based as **pairwise distances** between nodes

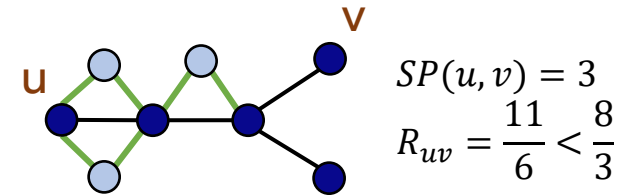
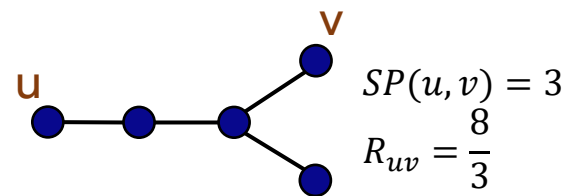
Previous work use metrics that

- **Fail to capture the global properties of the topology** → Shortest-path
- Using measures that lack theoretical guarantees → Independent Cascade

**We propose to use Effective Resistance ( $R_{uv}$ ) as the pairwise distance metric**

- Theoretical measure of information distance between nodes and hence of structural social capital
- Captures **global behavior** and **long-range dependencies**
- $R_{uv}$  is the *expected* time to reach **v** from **u** and come back to **u**
  - i.e., average of all existing paths between u and v!
  - Captures the number and length of paths between nodes
  - The smaller ER is the more and shorter paths between u and v are

$$R_{uv} = (\mathbf{e}_u - \mathbf{e}_v) \mathbf{L}^\dagger (\mathbf{e}_u - \mathbf{e}_v)^\top$$



We define the (inverse) social capital of a single node as the total sum of all it's distances to everyone else

$$R_{\text{tot}}(u) = \sum_{v \in \mathcal{V}} R_{uv}$$

# Measures of Group Social Capital

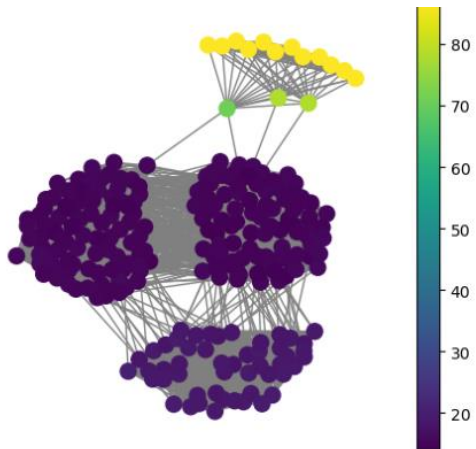
Access

## Group Isolation

$$R_{\text{tot}}(S_i) = \mathbb{E}_{u \sim S_i} [R_{\text{tot}}(u)]$$

Expected  $R_{uv}$  when sampling one node from group  $S_i$  and another node at random

**Intuition:** average distance to everyone for the nodes of the group

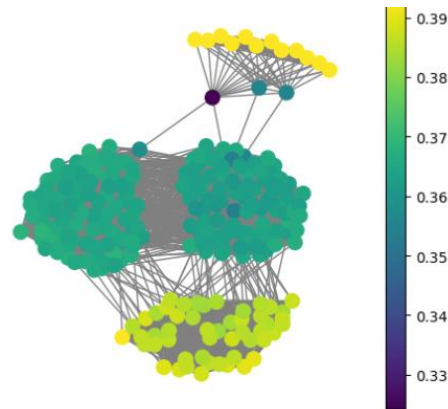


## Group Diameter

$$R_{\text{diam}}(S_i) = \mathbb{E}_{u \sim S_i} [\max_{v \in \mathcal{V}} R_{uv}]$$

Expected **worst-case**  $R_{uv}$  when sampling one node from group  $S_i$  and another node at random

**Intuition:** average maximum distance for the nodes of the group



Control

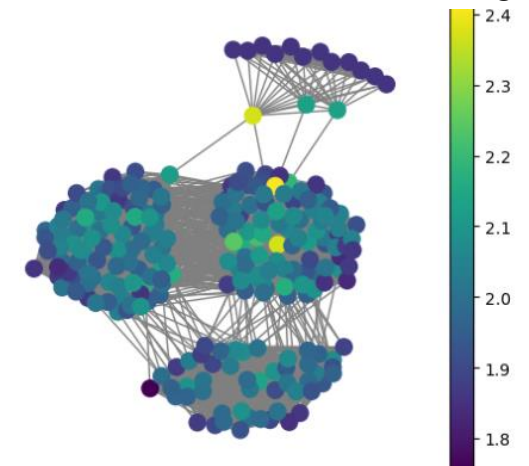
## Group Control

$$B_R(S_i) = \mathbb{E}_{u \sim S_i} [B_R(u)]$$

$$B_R(u) = \sum_{v \in \mathcal{N}(u)} R_{uv},$$

Expected **Information bottleneck** (similar to current-flow **betweenness**)

**Intuition:** average information control for the nodes of the group



# Measures of Structural Group Unfairness

Based on Rawlsian concept of Fairness, SGU measures the social capital gap for the most disadvantaged group

## Isolation Disparity

$$R_{\text{tot}}(S_i) = R_{\text{tot}}(S_j), \forall i, j \in SA.$$

$$\Delta R_{\text{tot}} = \max_{i,j \in SA} (R_{\text{tot}}(S_i) - R_{\text{tot}}(S_j)).$$

## Diameter Disparity

$$\mathcal{R}_{\text{diam}}(S_i) = \mathcal{R}_{\text{diam}}(S_j), \forall i, j \in SA.$$

$$\Delta \mathcal{R}_{\text{diam}} = \max_{i,j \in SA} (\mathcal{R}_{\text{diam}}(S_i) - \mathcal{R}_{\text{diam}}(S_j)).$$

## Control Disparity

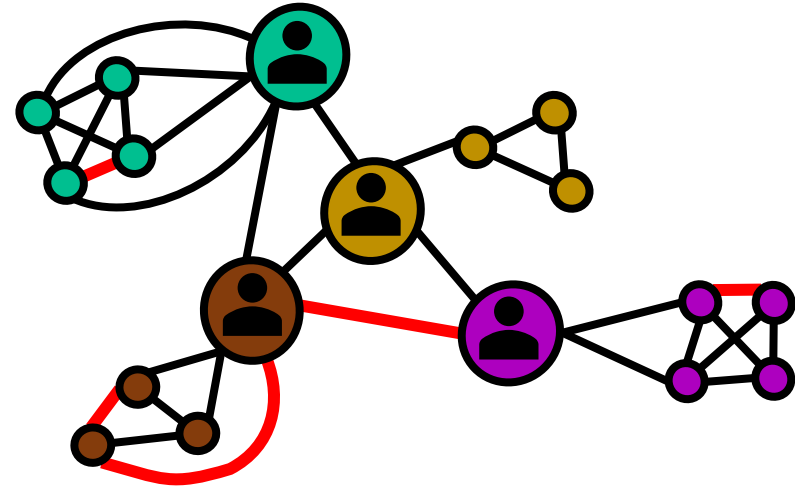
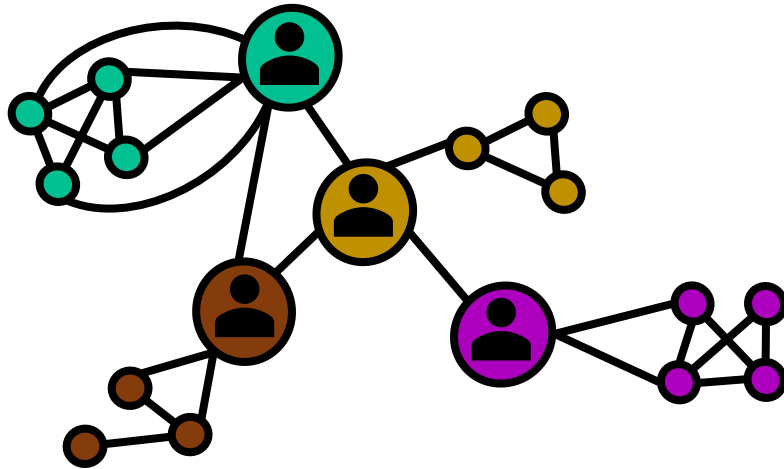
$$B_R(S_i) = B_R(S_j) = 2 - \frac{2}{|\mathcal{V}|}, \forall i, j \in SA.$$

$$\Delta B_R = \max_{i,j \in SA} (B_R(S_i) - B_R(S_j)).$$

$G$	$R_{\text{tot}} \downarrow$		$\mathcal{R}_{\text{diam}} \downarrow$		$B_R \uparrow$	
Facebook (female)	221.4	$\Delta R_{\text{tot}}$	2.29	$\Delta \mathcal{R}_{\text{diam}}$	1.93	$\Delta B_R$
Facebook (male)	<b>179.8</b>	41.62	<b>2.25</b>	0.042	<b>2.03</b>	0.107
UNC28 (female)	608.6	$\Delta R_{\text{tot}}$	2.11	$\Delta \mathcal{R}_{\text{diam}}$	1.99	$\Delta B_R$
UNC28 (male)	<b>586.3</b>	22.4	2.11	0.006	<b>2.00</b>	0.009
Google+ (female)	564.1	$\Delta R_{\text{tot}}$	1.31	$\Delta \mathcal{R}_{\text{diam}}$	1.81	$\Delta B_R$
Google+ (male)	<b>287.7</b>	276.4	<b>1.24</b>	0.078	<b>2.32</b>	0.51

Group social capital and SGU in real-world online social networks. Group with the largest social capital is highlighted in bold.

# How to mitigate SGU?



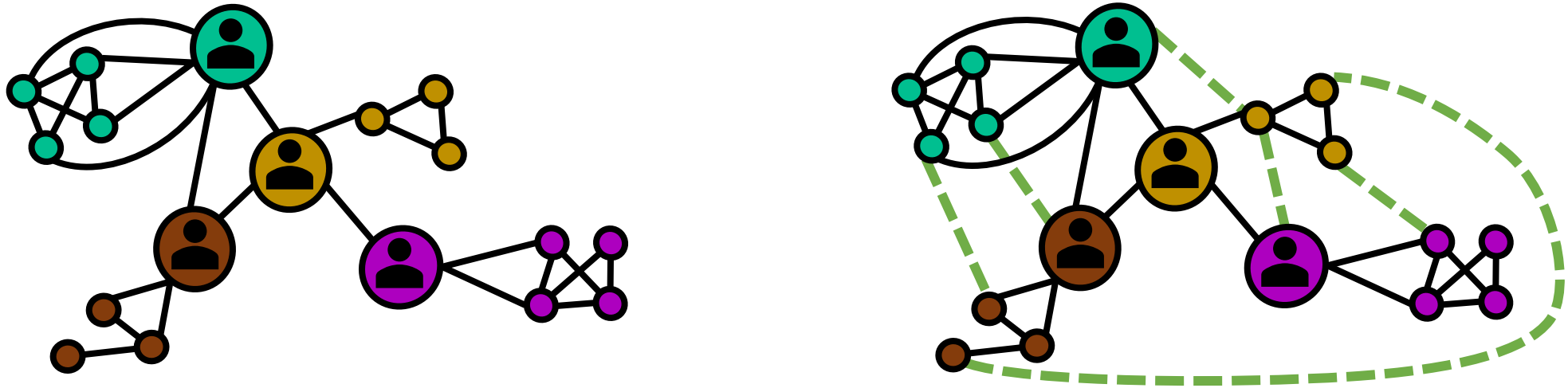
Common RecSys create connections by user similarity leading to same dynamics of segregation, polarization

RecSys do not maximize Information Flow on the graph, but **engagement** → **Same SGU**



# How to mitigate SGU? Weak ties

We propose a greedy Edge Augmentation Strategy based on weak ties



Create weak connections improves the network's information flow  
Reduces polarization, discrimination and **isolation**  
Improves diversity and **optimal flow of information**

Granovetter. The strength of weak ties: A network theory revisited. *Sociological theory* (1983)

Burt. 2004. Structural holes and good ideas. *American journal of sociology* (2004)

Arnaiz-Rodriguez, et al. Structural Group Unfairness: Measurement and Mitigation by means of the Effective Resistance. (2024)

# ERG: Edge Augmentation to mitigate SGU

- Use **Effective Resistance informed edge addition** reduce the disparities in social capital between groups and individuals **while also improve the social capital for all groups**
- Problem defined as a budgeted densifier problem

$$G' = \min_{G'=(\mathcal{V},\mathcal{E}')} \mathbb{E}_{u,v \sim \mathcal{V} \times \mathcal{V}} [R_{uv}] \quad \text{s.t.} \quad |\mathcal{E}' \setminus \mathcal{E}| = B \quad \mathcal{E} \subset \mathcal{E}'$$

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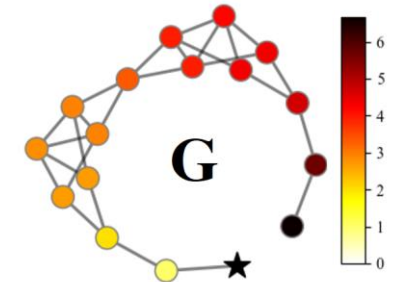
## Algorithm 1: ERG-Link

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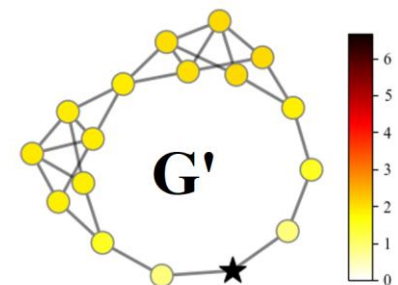
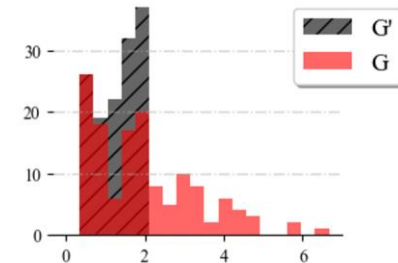
**Data:** Graph  $G = (\mathcal{V}, \mathcal{E})$ , a protected attribute  $SA$ , budget  $B$  of total number of edges to add

**Result:** New Graph  $G' = (\mathcal{V}', \mathcal{E}')$  with  $B$  new edges

- 1  $\mathbf{L} = \mathbf{D} - \mathbf{A}$ ;
  - 2  $S_d = \text{argmax}_{S_i \forall i \in SA} R_{\text{tot}}(S_i)$ ; // Identify the most disadvantaged group
  - 3 **Repeat**
  - 4  $\mathbf{L}^\dagger = \sum_{i>0} \frac{1}{\lambda_i} \phi_i \phi_i^\top = \left( \mathbf{L} + \frac{\mathbf{1}\mathbf{1}^\top}{n} \right)^{-1} - \frac{\mathbf{1}\mathbf{1}^\top}{n}$ ;
  - 5  $\mathbf{R} = \mathbf{1} \text{diag}(\mathbf{L}^\dagger)^\top + \text{diag}(\mathbf{L}^\dagger) \mathbf{1}^\top - 2\mathbf{L}^\dagger$ ; // Compute effective resistance
  - 6  $C = \{(u, v) \mid u \in S_d \text{ or } v \in S_d, (u, v) \notin \mathcal{E}'\}$ ; // Select edge candidates
  - 7  $\mathcal{E}' = \mathcal{E}' \cup \text{arg max}_{(u,v) \in C} R_{uv}$ ; // Add edge with maximum effective resistance
  - 8  $\mathbf{L} = \mathbf{L} + (\mathbf{e}_u - \mathbf{e}_v)(\mathbf{e}_u - \mathbf{e}_v)^\top$ ; // Fast update of  $\mathbf{L}$
  - 9 **Until**  $|\mathcal{E}' \setminus \mathcal{E}| = B$ ;
  - 10 **return**  $G'$ ;
- 



$$R_{u,v} \forall (u, v) \in \mathcal{V} \times \mathcal{V}$$



# ERG: Edge Augmentation to mitigate SGU

- Use **Effective Resistance informed edge addition** reduce the disparities in social capital between groups and individuals **while also improve the social capital for all groups**

Problem defined as a budgeted densifier problem

$$G' = \min_{G'=(V,\mathcal{E}')} \mathbb{E}_{u,v \sim V \times V} [R_{uv}] \quad \text{s.t.} \quad |\mathcal{E}' \setminus \mathcal{E}| = B \quad \mathcal{E} \subset \mathcal{E}'$$

Algorithm 1: ERG-Link

Compute all pairwise ER distances **add the edge with maximum ER**  
 that has -at least- one endpoint on the discriminated group

1  $L = D - A;$

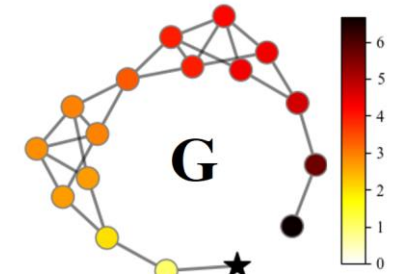
2 We have a budget of  $B$  edges to add, so we have to be very effective

3 Repeat

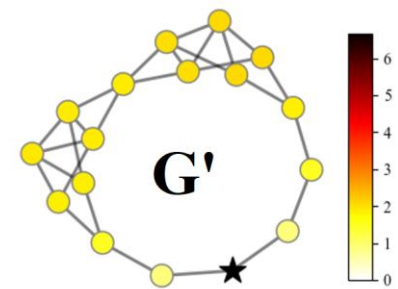
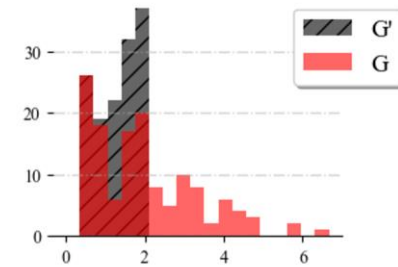
As ER is heavily theoretically grounded and studied in Theoretical CS, this simple approach is effective and provides strong insights

```

5  $R = 1$ 
6  $C = \{(u, v) \mid u \in S_d \text{ or } v \in S_d, (u, v) \notin \mathcal{E}'\};$  // Select edge candidates
7  $\mathcal{E}' = \mathcal{E}' \cup \arg \max_{(u,v) \in C} R_{uv};$  // Add edge with maximum effective resistance
8  $L = L + (e_u - e_v)(e_u - e_v)^T;$  // Fast update of L
9 Until  $|\mathcal{E}' \setminus \mathcal{E}| = B;$ 
10 return  $G';$ 
    
```



$$R_{u,v} \forall (u, v) \in V \times V$$



# Experiments on Real World: Group SC and SGU

Results on 3 real-world online social networks: Facebook, UNC and Google+

(a) Facebook ( $B=50$ )

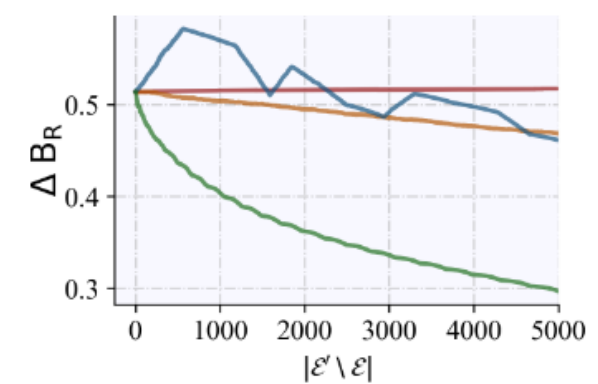
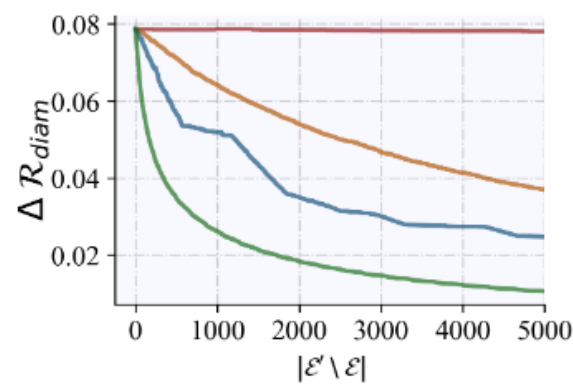
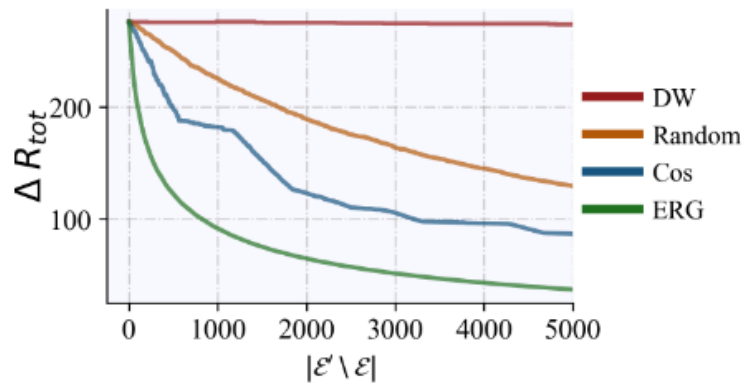
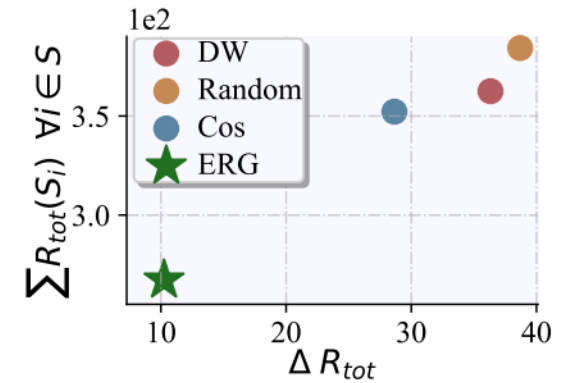
	$\Delta R_{tot}$	$\Delta \mathcal{R}_{diam}$	$\Delta B_R$
$G$ (original)	41.62	0.042	0.107
Random	38.7	0.039	0.108
DW	36.3	0.031	0.104
Cos	28.7	0.029	0.120
ERG	<b>10.3</b>	<b>0.009</b>	<b>0.098</b>

(b) UNC28 ( $B=5000$ )

	$\Delta R_{tot}$	$\Delta \mathcal{R}_{diam}$	$\Delta B_R$
$G$ (original)	22.4	0.006	0.009
Random	19.8	0.005	0.014
DW	22.2	0.006	0.004
Cos	19.1	0.005	0.102
ERG	<b>8.8</b>	<b>0.002</b>	<b>0.003</b>

(c) Google+ ( $B=5000$ )

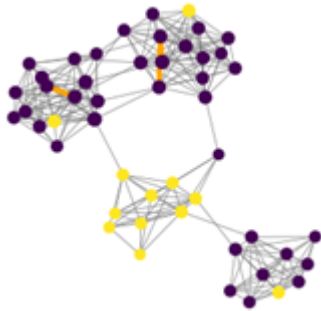
	$\Delta R_{tot}$	$\Delta \mathcal{R}_{diam}$	$\Delta B_R$
$G$ (original)	276.4	0.078	0.51
Random	129.4	0.037	0.47
DW	274.1	0.078	0.51
Cos	86.8	0.025	0.47
ERG	<b>37.1</b>	<b>0.011</b>	<b>0.29</b>



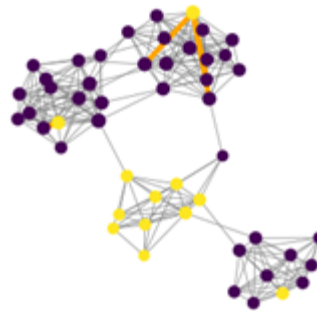
Effectively mitigates the social capital gap for the most disadvantaged group while increasing the social capital of all groups

# Effect on the Network Structure

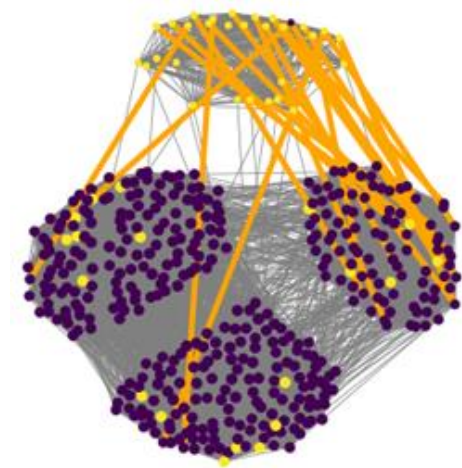
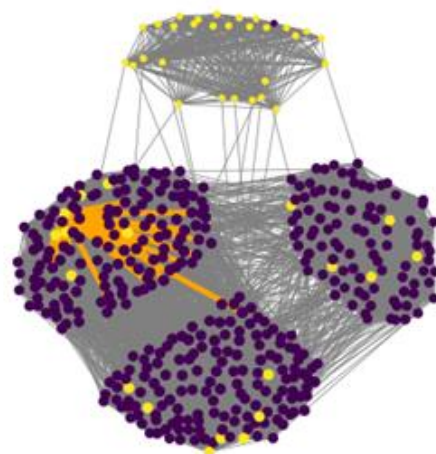
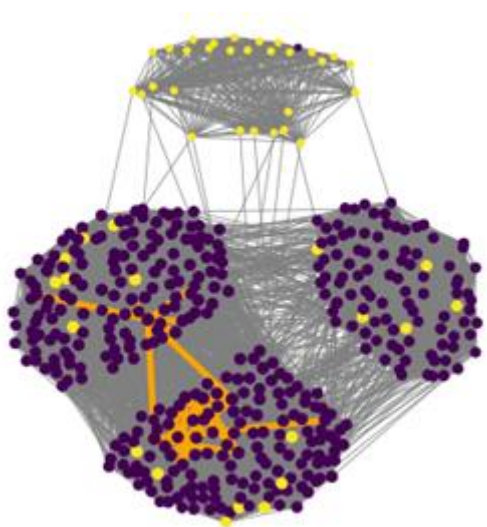
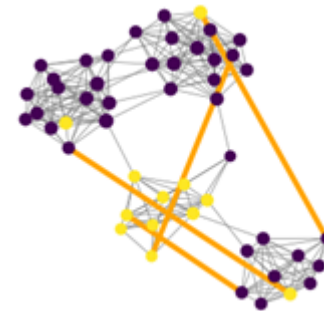
RecSys



RecSys + Prioritization

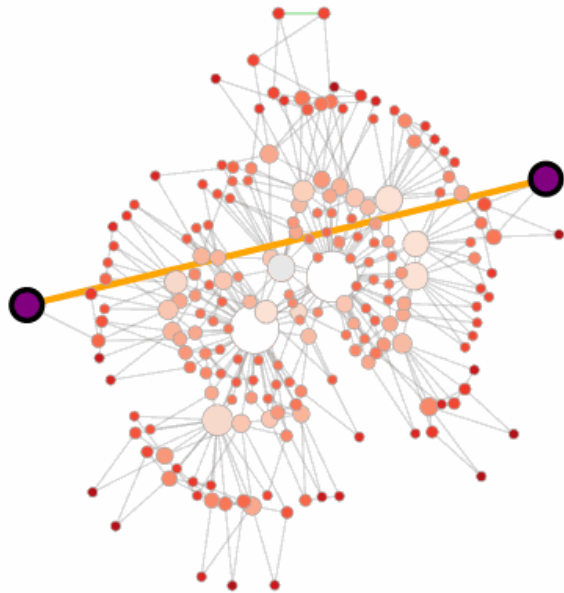


ERG

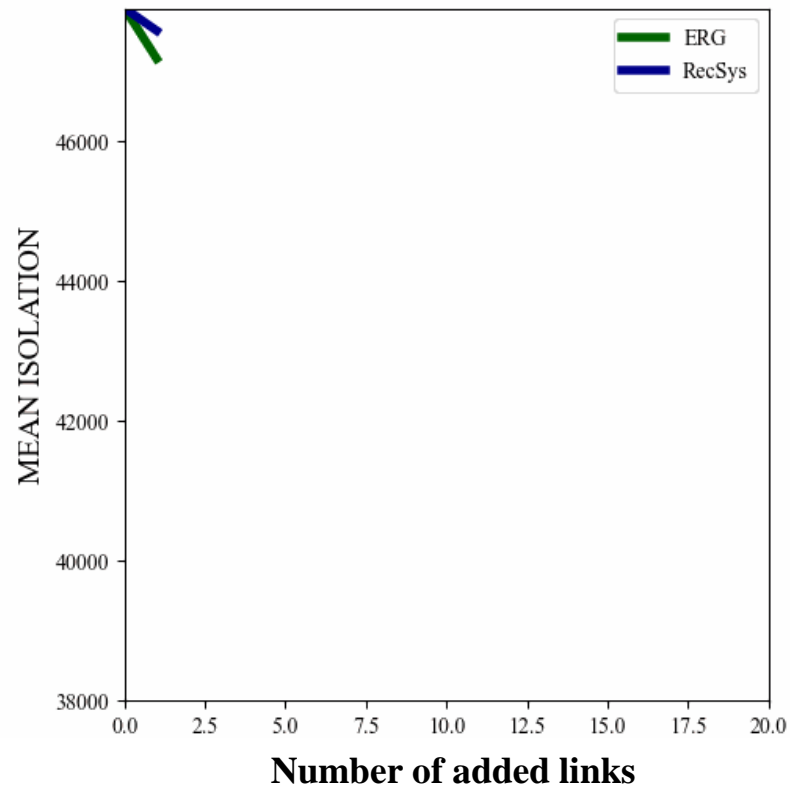


# Effect on Individual Social Capital

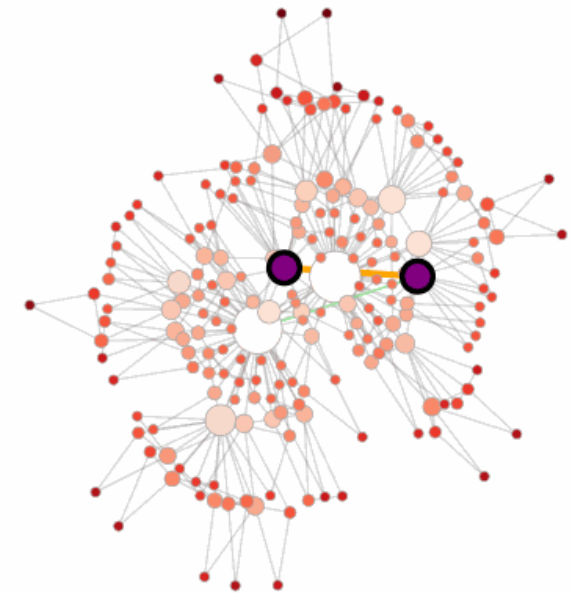
Intervention to maximize the information flow



**Average Individual Isolation**  
The lower the better the info is flows in the network



Classical RecSys



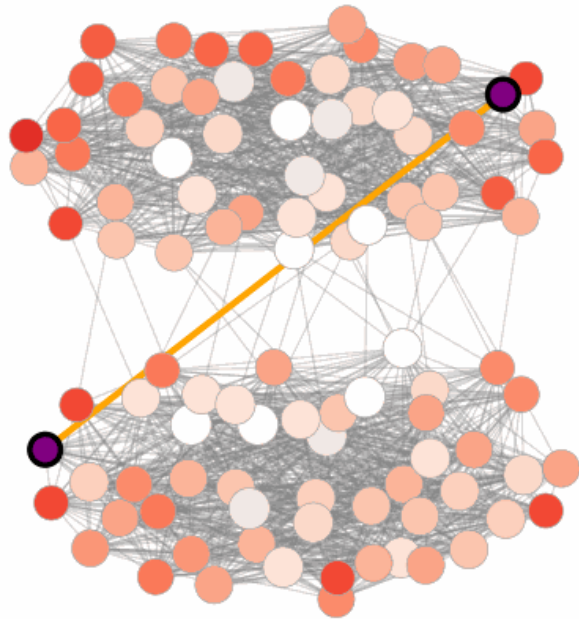
😊 Low  High 😞

Isolation of a person



# Effect on Individual Social Capital

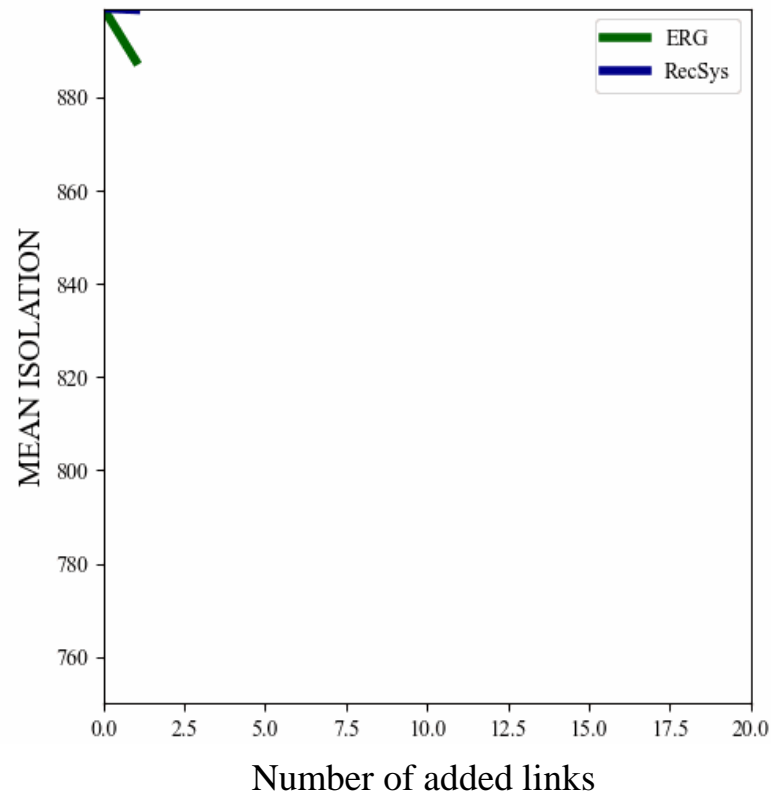
Intervention to maximize the information flow



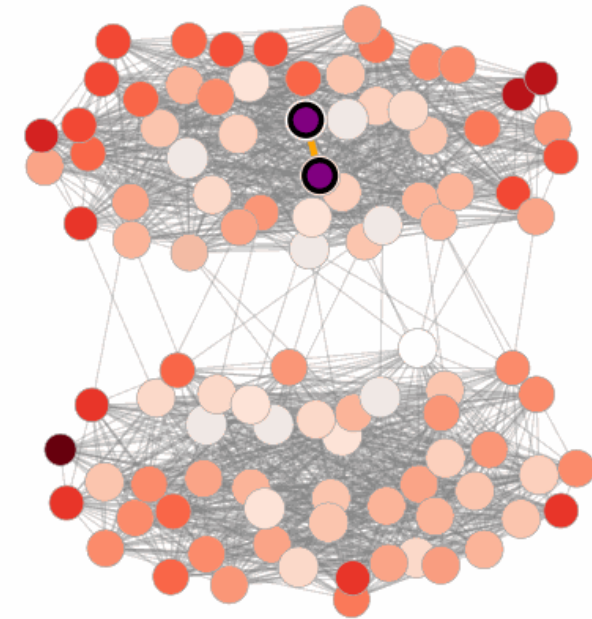
## Polarized Society

### Average Individual Isolation

The lower the better the info is flows in the network



## Classical RecSys



Low

Isolation of a person

High



# Main takeaways

Social Networks are very useful to analyze social dynamics or make predictions about people and their relationships

Specific [Arnaiz-Rodriguez, Curto and Oliver 2024]:

Your position on a **social network** defines the access and control of information you have



Social Capital

It is not easy to **quantify it** and intricate to **align the literature** in sociology and graph theory and fairness



Quantitative measures of social capital sociology-aligned based on graph theory

**Effective resistance:**  
isolation, diameter and control

There are some individual and group inequalities and not trivial to fix



Fixed by an edge augmentation strategy based on the **weakest links**

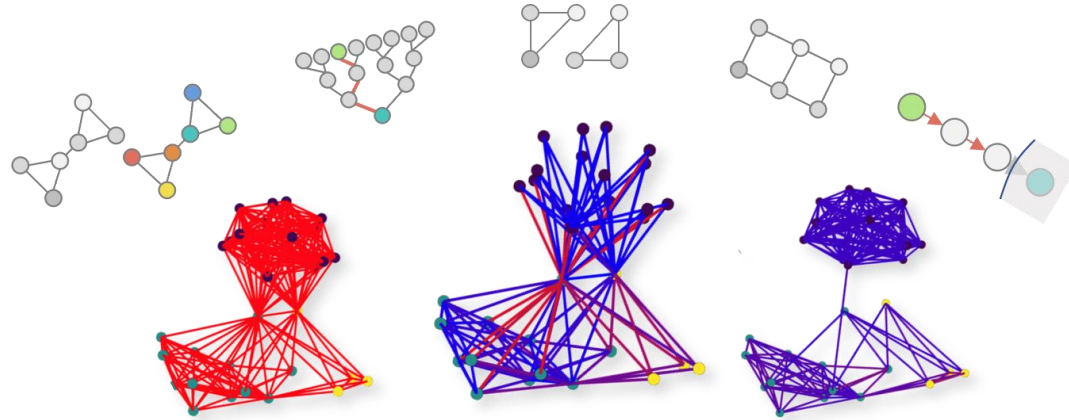
**ERG:**  
Adding the edge that maximizes the information flow on the whole network

Arnaiz-Rodriguez, Adrian, Georgina Curto, and Nuria Oliver.

"Structural Group Unfairness: Measurement and Mitigation by means of the Effective Resistance." *Submitted.* (2024)



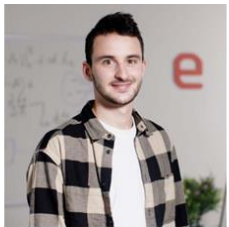
# Rewiring tutorial in LoG conference



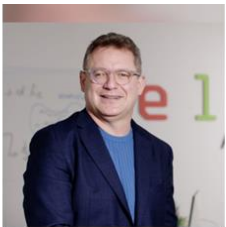
## Graph Rewiring from Theory to Applications in Fairness



<https://ellisalicante.org/tutorials/GraphRewiring>



Adrián Arnaiz Rodríguez  
ELLIS Alicante



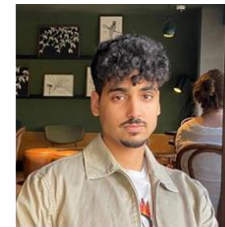
Francisco Escolano  
University of Alicante  
ELLIS Alicante



Nuria Oliver  
ELLIS Alicante



Edwin Hancock  
University of York

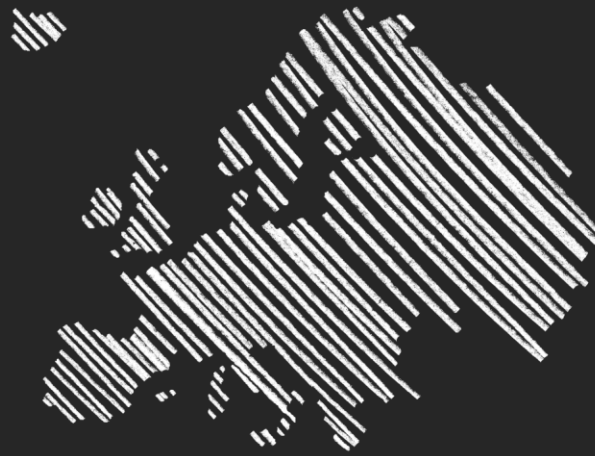


Ahmed Begga  
University of Valencia



Arnaiz-Rodríguez, A. et al. "DiffWire: Inductive Graph Rewiring via the Lovász Bound" Learning on Graphs Conference (2022)

Arnaiz-Rodríguez, A. et al. "Tutorial on Graph Rewiring: From Theory to Applications in Algorithmic Fairness" Learning on Graphs Conference (2022)



e l l i a s

UNIT  
ALICANTE

adrian@ellisalicante.org

@arnaiztech