

BIS 2019 conference

BSCT 19 workshop

Contagion in Bitcoin networks

27th of June 2019, Universidad de Sevilla, Sevilla,
Spain

***Coquidé Célestin¹, Lages José¹ and
Shepelyansky Dima L.²***

¹Institut UTINAM, university of Bourgogne Franche-Comté,
Besançon, France

²Laboratoire de physique théorique, University of Paul
Sabatier, Toulouse, France.

<http://perso.utinam.cnrs.fr/~lages/apex/>



Outline

- 1) Bitcoin network
- 2) Google matrix (GM) theory and PageRank algorithm
- 3) Ranking of users
- 4) Crisis propagation
- 5) Reduced google matrix method (REGOMAX)
- 6) Conclusion



1) Bitcoin network

*The Bitcoin is a **crypto-currency** created by **Satoshi Nakamoto** in 2009*

- This digital currency is **independent** and very **secured**
- **Every transactions** are **stored** as a chain = **blockchain**
- Usable for everyone who get a Bitcoin **wallet**
- Used for **shopping** and **trading**

*Its **value** has known **skyrocketing** periods*

today (07:38 UTC) 1 BTC = 10,825.88 €



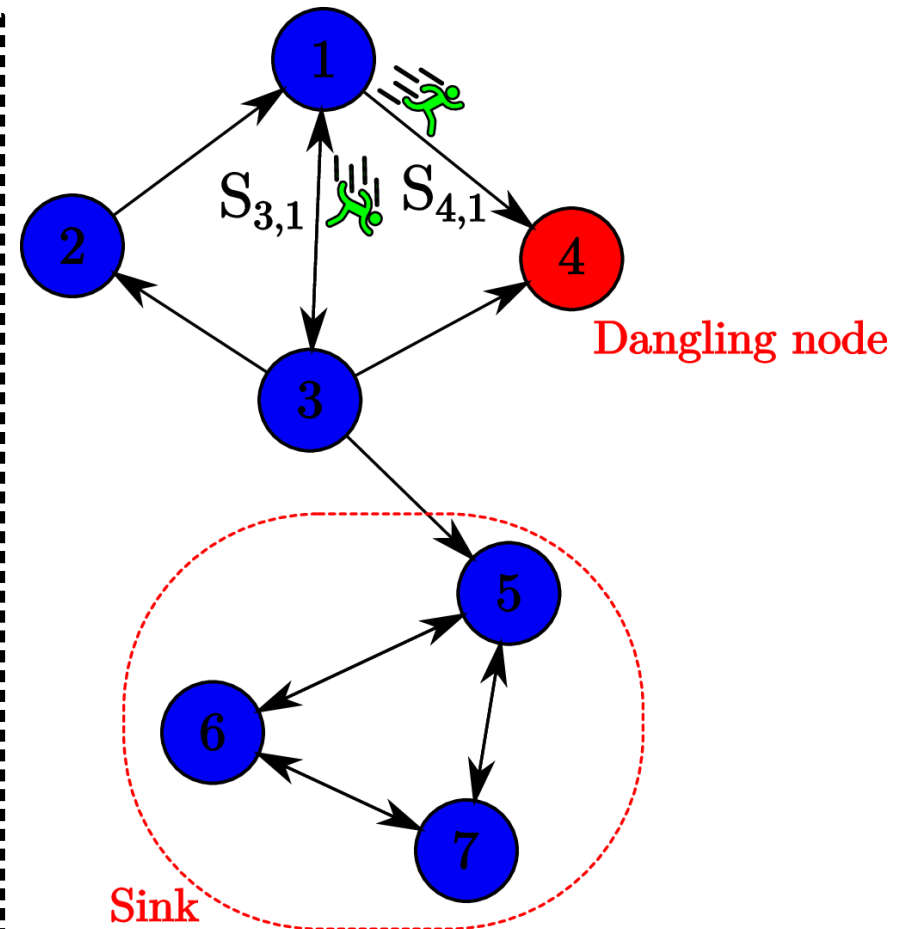
1) Bitcoin network

- About **35 M** of ‘**users**’ by the first quarter (Q1) of **2019**
- Anonymized **transactions** are **accessible** at www.blockchain.com/explorer
- Network construction :
 Node = **Wallet**
 Link = **Transaction** (directed)
- We use data from **2009 Q3** (142 users, 117 Transactions) to **2013 Q2** (6 M, 16 M)

*How can we **analyze efficiently** such a **big** (directed) **network** ?*

2) Google matrix and PageRank

- **Modeling a random walk** through a **directed network**
- **Stochastic matrix (S)**
 - Transition **probabilities**
 - **Dangling node** fixed
- **Google matrix (G)** (Brin and Page 1998)
 - **Sink** effect **avoiding**
 - **Leading eigenvalue** degeneracy = 1





2) Google matrix and PageRank

- **PageRank (P) = leading eigenvector** such that **$GP=P$**
 - => Steady-state**
 - => The highest component for the most reachable node**
 - => Efficiency of ingoing link = Importance**
- **Inverting links direction => G^* => CheiRank**
 - => Efficiency of outgoing link = communicability**
- We can **rank** nodes by **PageRank/CheiRank** order



3) Ranking of users

- For each time slice we have **G** , **G^*** , **P** and **P^***
- **PageRank (P)** encodes **information** on Bitcoin **import effectiveness**
- **CheiRank (P^*)** encodes **information** on Bitcoin **export effectiveness**
- **2DRank** brings **informations** contained in **both PR and CR**

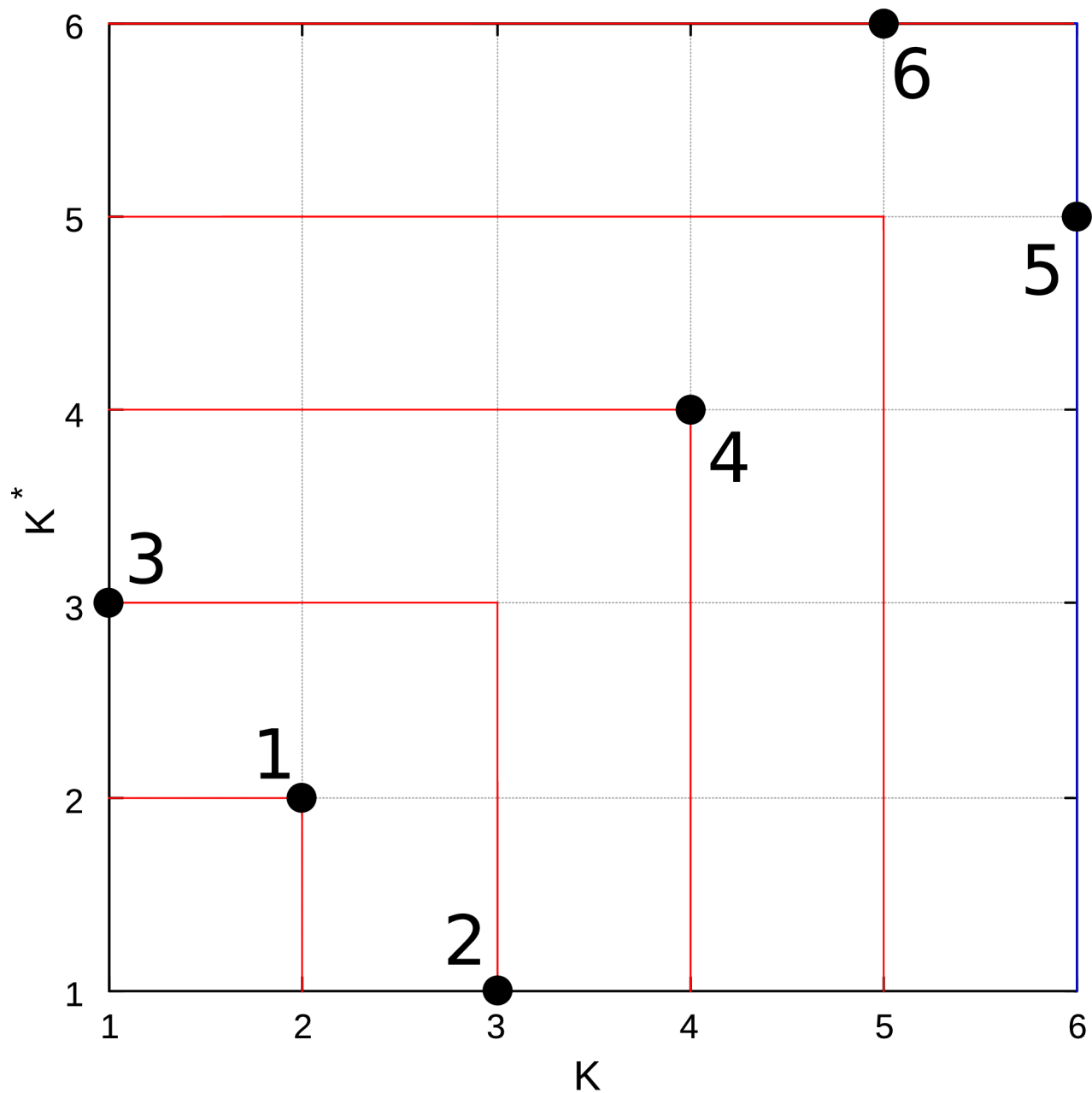
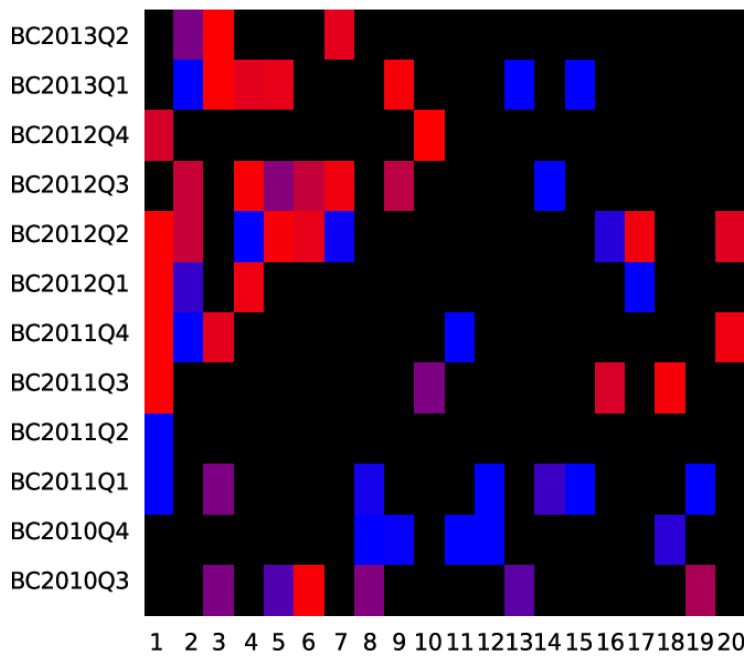


Fig1. Construction of the **2DRank**. K is for **PageRank** list, K^* for **CheiRank** list.

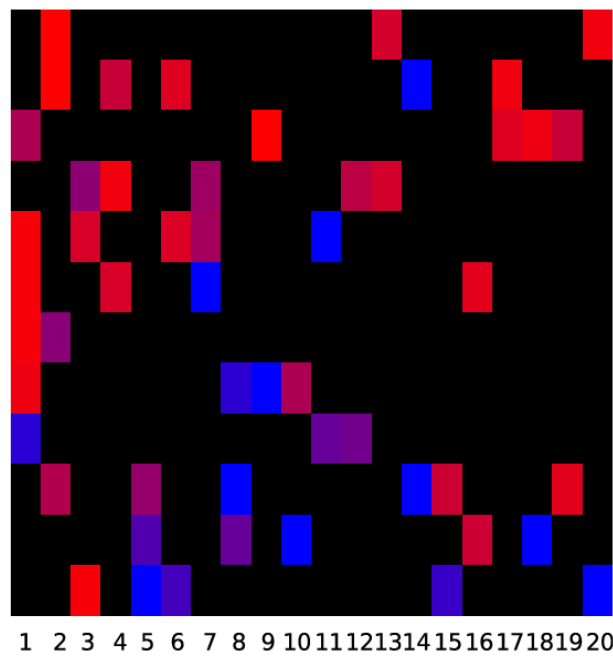
PR

CR

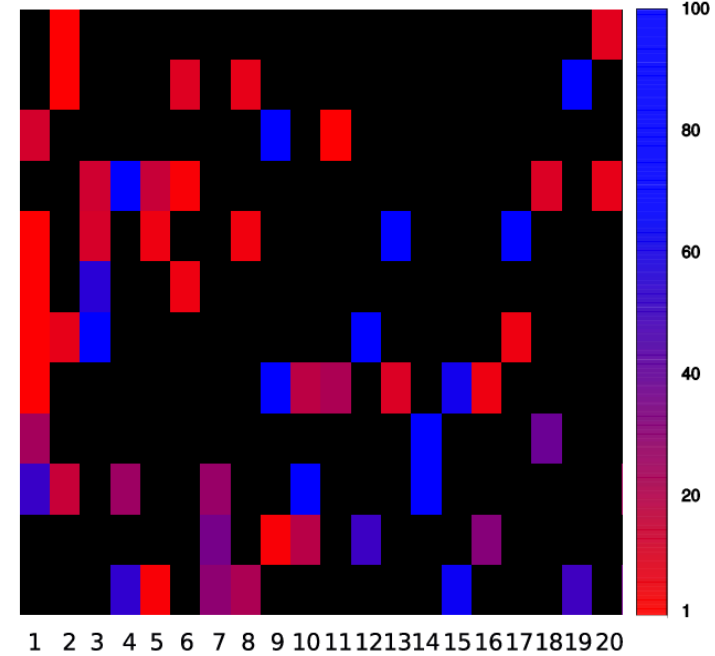
2D



Rank	Node	Occurance
1	6	7
2	4	6
3	11	5
4	24	4
5	18	4
6	26	3
7	15	3
8	2	3
9	20	3
10	30	2
11	443	2
12	220	2
13	23	2
14	515	2
15	89	2
16	49	2
17	1469	2
18	103	2
19	9	2
20	1136	2



Rank	Node	Occurance
1	6	6
2	11	4
3	89	3
4	15	3
5	24	3
6	18	3
7	4	3
8	2	3
9	30	2
10	220	2
11	1469	2
12	23	2
13	239	2
14	857	2
15	213	2
16	19618	2
17	159	2
18	3553	2
19	10	2
20	2149	2



Rank	Node	Occurance
1	6	7
2	11	4
3	4	4
4	115	3
5	15	3
6	18	3
7	2	3
8	24	3
9	30	3
10	89	3
11	10	2
12	103	2
13	12	2
14	1	2
15	1302	2
16	1469	2
17	16	2
18	19618	2
19	199	2
20	20	2

Tab.1 List of the **20 most present users** among **12 top 100s**. From **PageRank** algorithm (left), **CheiRank** (center) and **2DRank** (right).



4) Crisis propagation

- Crisis model :

1) B_u is the CheiRank/PageRank **balance** for a given **user**

$$B_u = \frac{P_u^* - P_u}{P_u^* + P_u}$$

2) If $B_u \leq -\kappa$ (**crisis threshold**) then we consider a **bankrupt event** for the user 'u'

3) 'u' needs to **close all** its Bitcoin **import** transactions

Do it for all nodes => **new network**

- We do it for different step τ

Fig2a. **Probability** to find a **bankrupt user** at 4 different iterations τ and crisis threshold κ . Using time slice **BC2013Q1**.

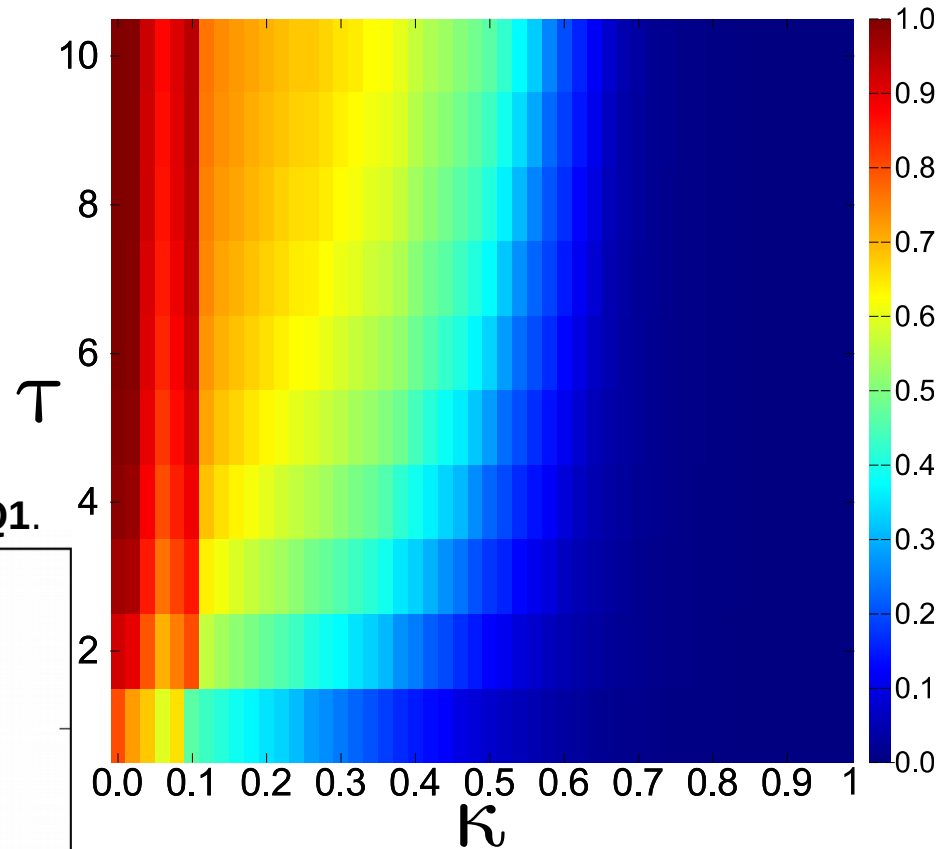
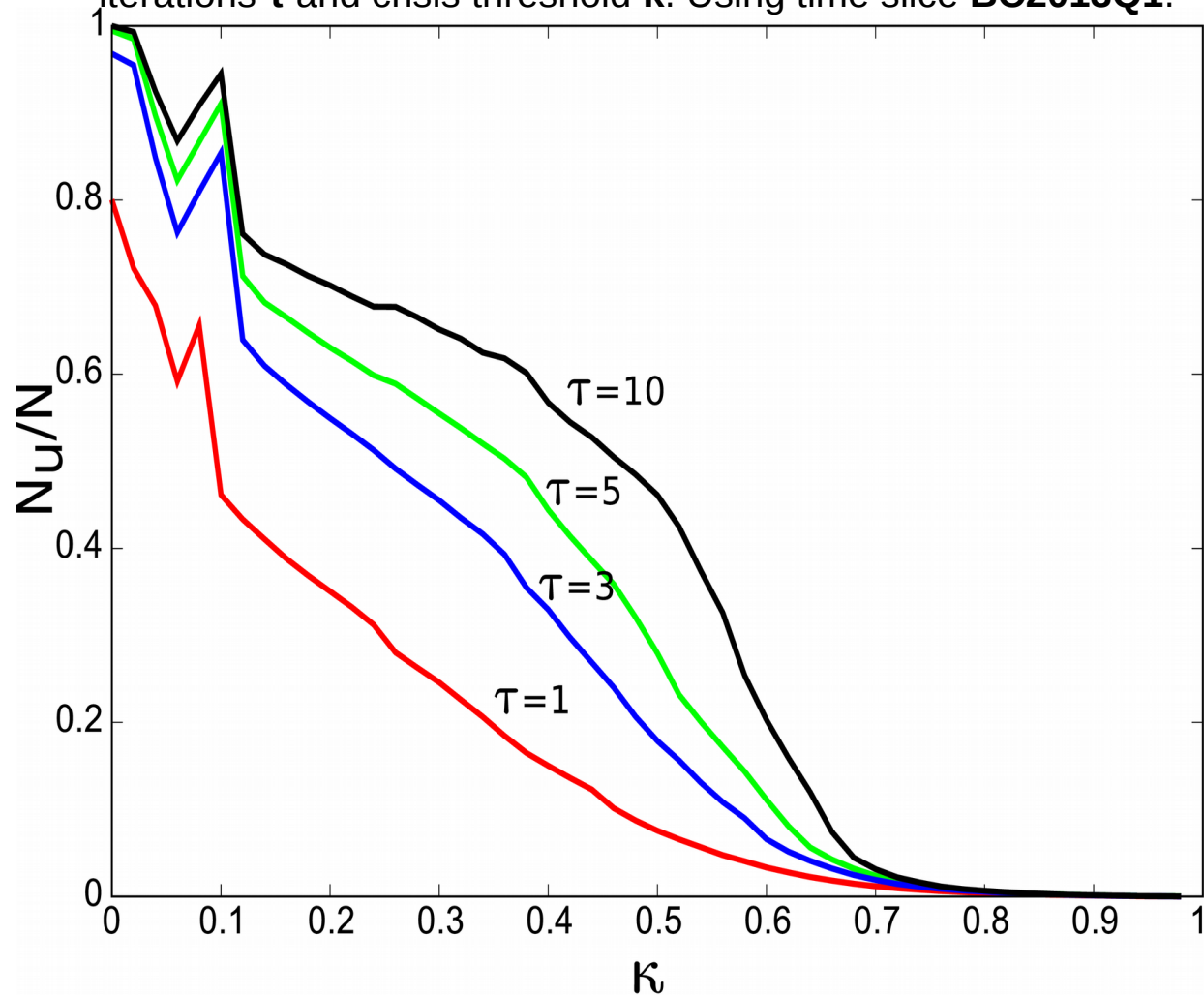


Fig2b. **Probability** to find a **bankrupt user** at several iterations τ and crisis threshold κ . Using the time slice **BC2013Q1**.

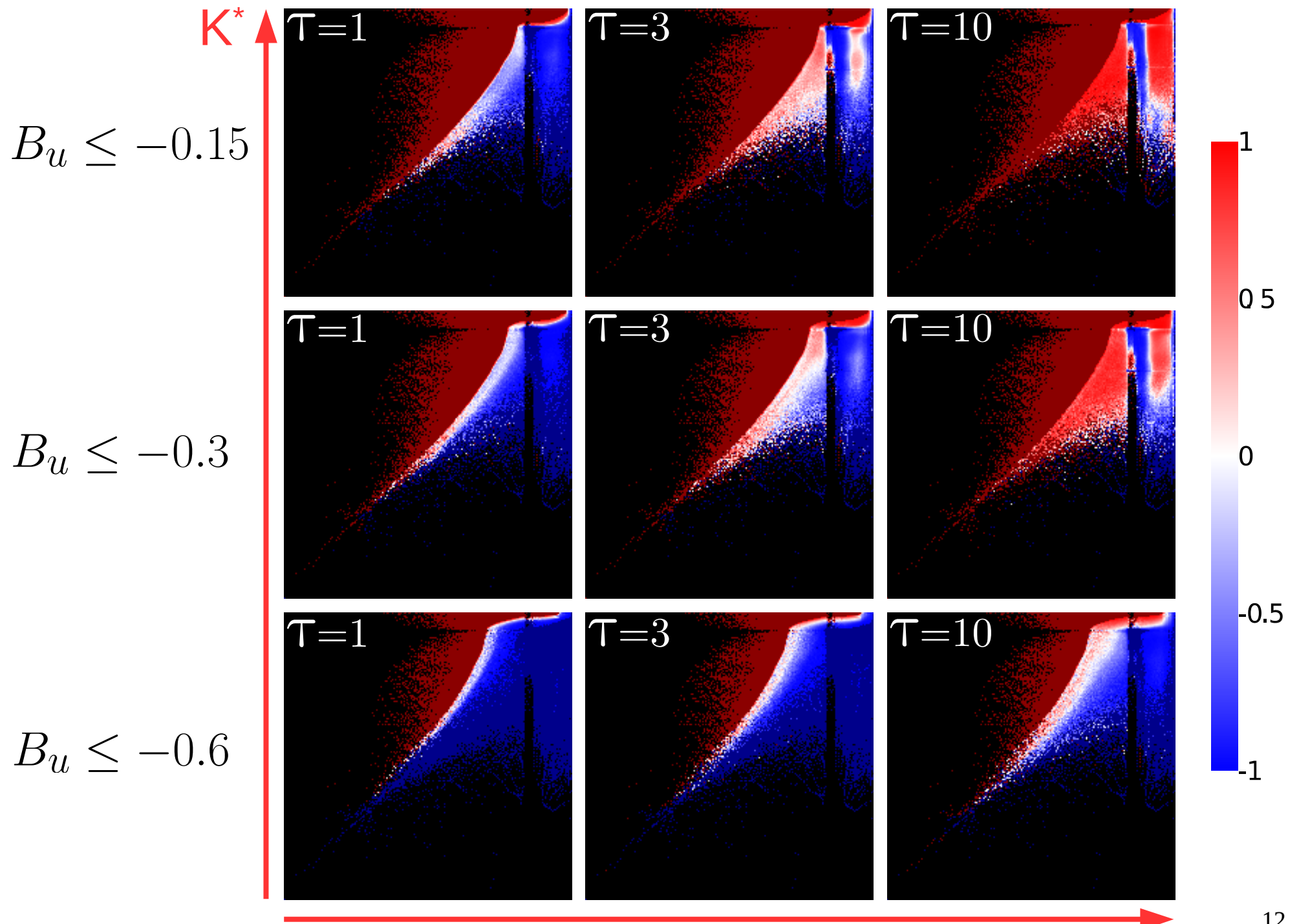
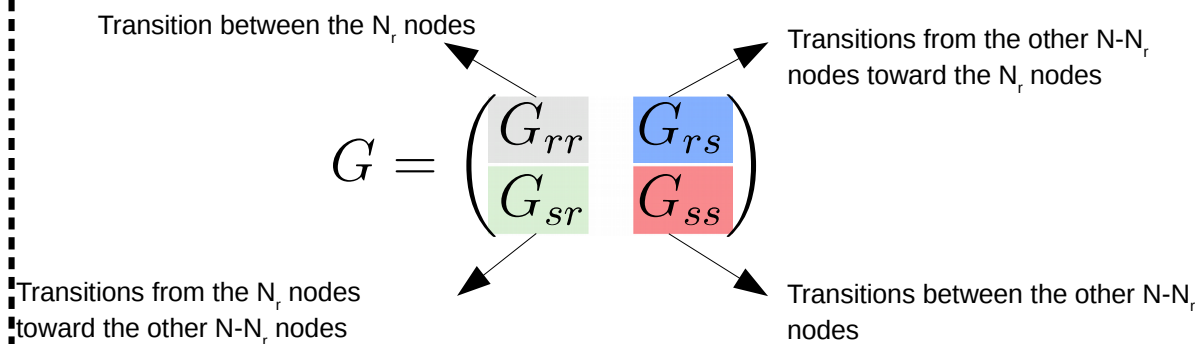


Fig3. **BC13Q1** users in the (K, K^*) plane represented in 200 X 200 log cells. **Bankruptcy** and **Safety** evolution with iteration τ for different κ .

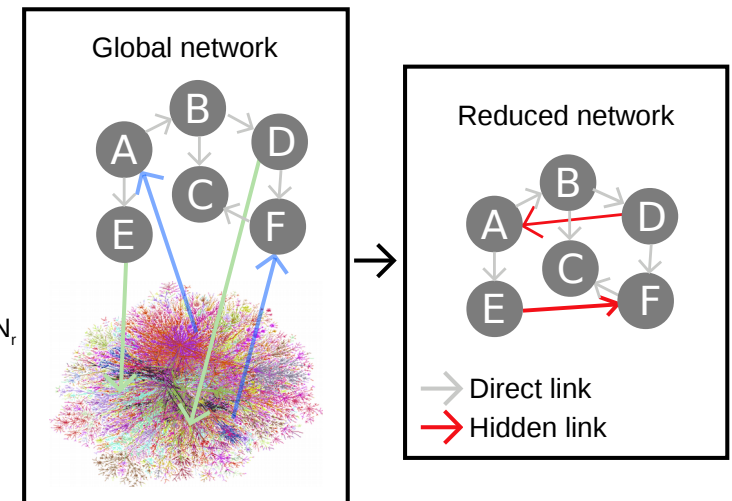
5) Reduced google matrix method (REGOMAX)

- REGOMAX** allows us to **point out hidden links** between nodes of interest (Frahm and Shepelyansky 2016, Frahm et al. 2016)



$$G_R = G_{rr} + G_{pr} + G_{qr}$$

Direct links Contribution from Hidden links
«PageRank»



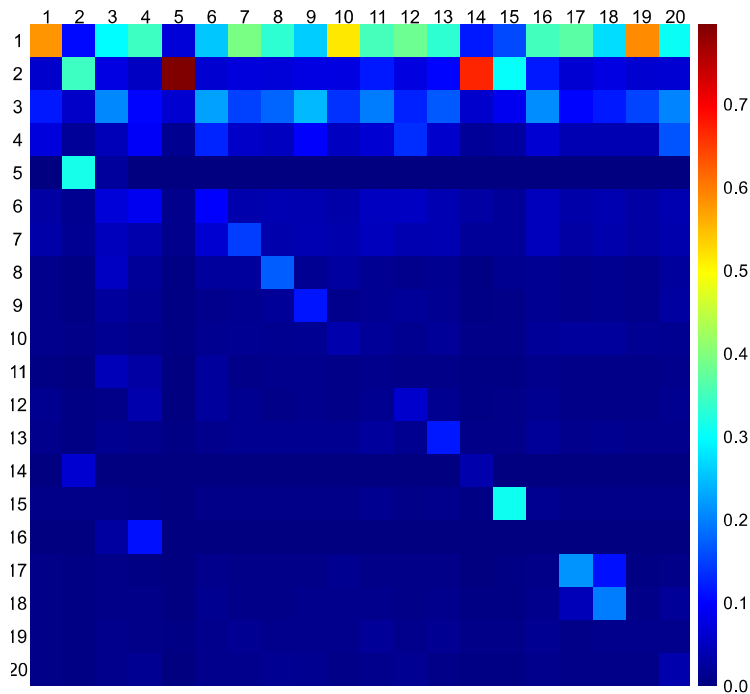
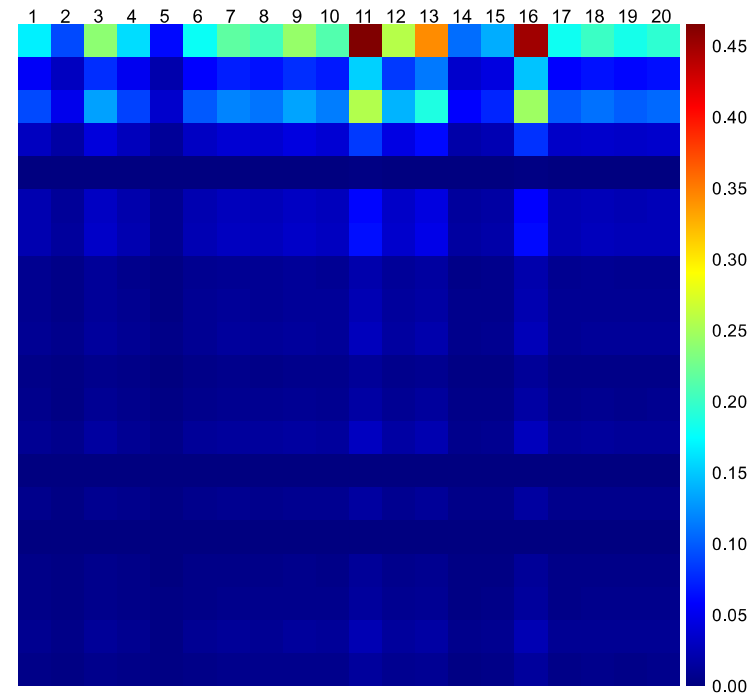
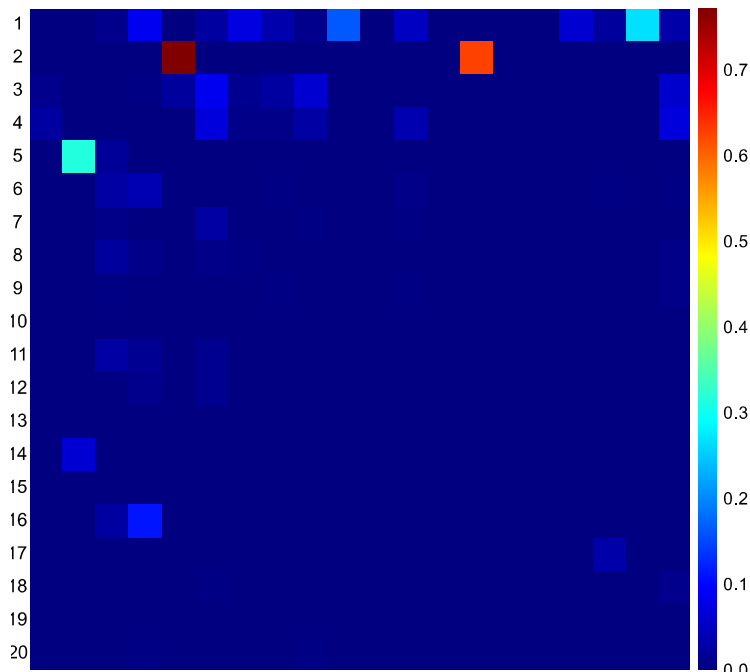
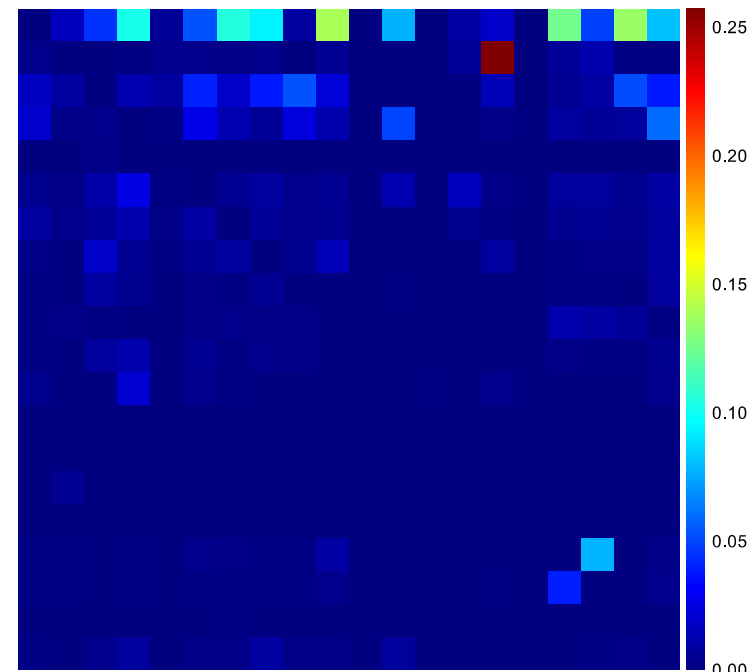
G_R  G_{pr}  G_{rr}  G_{qrnd} 

Fig4. Reduced google matrix for the top 20 PageRank users of BC13Q1.



6) Conclusion

- **Blockchain** can be **used** for **Bitcoin transactions network analysis**
- With such a complex network **GM** and **PageRank/CheiRank** algorithms are **useful**
- With our simple **crisis contagion model** we can have access to **different scenario** of crisis **propagation** of the **Bitcoin network**
- **Top users** seem to get **bankruptcy rapidly**
- By using **REGOMAX** for inferring **hidden** (indirect) **relationships** we showed that top users have strong hidden + direct transactions and partnerships



6) Conclusion

- We don't have **metadata** about the **nature** of the **transactions**
 - Services, shopping, internal exchange (**same person** but **multiple accounts**)
- We want to **compare** contagion **robustness** of **bitcoin** network with **other net** like **WTN**, **banking exchange** network



6) Conclusion

- **Wikipedia network :**

- Hidden relationship between political leaders (Frahm et al. 2016)

- Terrorist groups (El Zant et al. 2018)

- Application to biomedical topic such as infectious diseases (Rollin et al. 2019)

- World influence and Interactions of Universities (Coquidé et al. 2019)

- **Trade network :**

- Influence of petroleum and gas trade on EU economies (Coquidé et al. 2019)

- Economic activities with WTO (Coquidé et al. 2019)

- Crypto-currency

- **Game network :**

- Game of go (Coquidé et al. 2017)



Thank you