

# On Stability of Banking Networks

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## Abstract

Threats on the stability of a financial system may severely affect the functioning of the entire economy, and thus considerable emphasis is placed on the analyzing the cause and effect of such threats. The financial crisis in the current and past decade has shown that one important cause of instability in global markets is the so-called *financial contagion*, namely the spreadings of instabilities or failures of *individual* components of the network to other, perhaps healthier, components. This leads to a natural question of whether the regulatory authorities could have predicted and perhaps mitigated the current economic crisis by effective computations of some stability measure of the banking networks. Motivated by such observations, we consider the problem of defining and evaluating stabilities of both homogeneous and heterogeneous banking networks against propagation of *synchronous idiosyncratic shocks* given to a subset of banks. We formalize the homogeneous banking network model of Nier *et al.* [38] and its corresponding heterogeneous version, formalize the synchronous shock propagation procedures outlined in [19, 38], define two appropriate stability measures and investigate the computational complexities of evaluating these measures for various network topologies and parameters of interest. Our results and proofs also shed some light on the properties of topologies and parameters of the network that may lead to higher or lower stabilities.

## 1 Introduction and Motivation

In market-based economies, financial systems perform important financial intermediation functions of borrowing from surplus units and lending to deficit units. Financial stability is the ability of the financial systems to absorb shocks and perform its key functions, even in stressful situations. Threats on the stability of a financial system may severely affect the functioning of the entire economy, and thus considerable emphasis is placed on the analyzing the cause and effect of such threats. The concept of instability of a market-based financial system due to factors such as debt financing of investments can be traced back to earlier works of the economists such as Irving Fisher [23] and John Keynes [30] during the 1930's Great Depression era. Subsequently, some economists such as Hyman Minsky [37] have argued that:

such instabilities are inherent in many modern capitalist economies.

In this paper, we investigate systemic instabilities of the banking networks, an important component of modern capitalist economies of many countries. The financial crisis in the current and past decade has shown that an important component of instability in global financial markets is the so-called *financial contagion*, namely the spreadings of instabilities or failures of *individual* components of the network to other, perhaps healthier, components. The general topic of interest in this paper, motivated by the global economic crisis in

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the current and the past decade, is the phenomenon of financial contagion in the context of *banking networks*, and is philosophically related to the following natural extension of the question posed by Minsky and others:

Are the instabilities of the banking networks systemic? Could one have predicted, assuming access to all necessary data, the current economic crisis by effective computations of the stability of the relevant banking networks ?

To investigate these types of questions, one must first settle the following issues:

- What is the *precise* model of the banking network that is studied?
- How *exactly* failures of individual banks propagated through the network to other banks?
- What is an *appropriate stability measure* and what are the computational properties of such a measure?

As prior researchers such as Allen and Babus [1] pointed out, graph-theoretic concepts provide a conceptual framework within which various patterns of connections between banks can be described and analyzed in a meaningful way by modeling banking networks as a *directed* network in which nodes represent the banks and the links represent the direct exposures between banks. Such a network-based approach to studying financial systems is particularly important for assessing financial stability, and in capturing the externalities that the risk associated with a single or small group of institutions may create for the entire system. Conceptually, links between banks have two *opposing* effects on contagion:

- More interbank links increase the opportunity for spreading failures to other banks [25]: when one region of the network suffers from a crisis, another region also incurs a loss because their claims on the troubled region fall in value and, if this spillover effect is strong enough, it can cause a crisis in adjacent regions.
- More interbank links provide banks with a form of *coinsurance* against uncertain liquidity flows [2], *i.e.*, banks can insure against the liquidity shocks by exchanging deposits through links in the network.

## 2 The Banking Network Model

### Homogeneous Networks: Balance Sheets and Parameters for Banks

We provide a precise abstraction of the model as outlined in [38] which builds up on the works of Eboli [19]. The network is modeled by a weighted directed graph  $G = (V, F)$  of  $n$  nodes and  $m$  directed edges, where each node  $v \in V$  corresponds to a bank ( $\text{Bank}_v$ ) and each directed edge  $(v, v') \in F$  indicates that  $\text{Bank}_v$  has an agreement to lend money to  $\text{Bank}_{v'}$ . Let  $\deg_{\text{in}}(v)$  and  $\deg_{\text{out}}(v)$  denote the in-degree and the out-degree of node  $v$ . The model has the following parameters:

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$E$ = total external asset	$I$ = total inter-bank exposure
$A = I + E$ = total asset	$\beta = E/A$ = percent of total external asset to total asset (thus, $I = (1 - \beta)A$ and $E = \beta A$ )
$w = w(e) = \frac{I}{m}$ = weight of any edge $e \in F$	$[0, 1] \ni \gamma$ = percentage of equity to asset
$\Phi$ = severity of shock ( $\Phi \in (0, 1]$ , $\Phi > \gamma$ )	

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Now, we describe the balance sheet for a node  $v \in V$  (i.e., for  $\text{Bank}_v$ ):

Assets		Liabilities	
$\iota_v$	$= \deg_{\text{out}}(v) \times w = \deg_{\text{out}}(v) \times \frac{I}{m}$ $= \text{interbank asset (money lent)}$		
$b_v$	$= \deg_{\text{in}}(v) \times w = \deg_{\text{in}}(v) \times \frac{I}{m}$ $= \text{interbank borrowing (money borrowed)}$	$b_v$	$= \deg_{\text{in}}(v) \times w = \deg_{\text{in}}(v) \times \frac{I}{m}$ $= \text{interbank borrowing (money borrowed)}$
$e_v$	$= \text{share of total external asset } E$ $= (b_v - \iota_v) + \frac{E - \sum_{v \in V} (b_v - \iota_v)}{n}$ $= (b_v - \iota_v) + \frac{E}{n}$ $\text{since } \sum_{v \in V} (b_v - \iota_v) = w \times \sum_{v \in V} (\deg_{\text{in}}(v) - \deg_{\text{out}}(v)) = 0$	$c_v = \gamma \times a_v$	$= \text{net worth (equity)}$
		$d_v$	$= \text{customer deposits}$
$a_v = e_v + \iota_v$	$= b_v + \frac{E}{n}$	$\ell_v = b_v + c_v + d_v$	$= \text{total liability}$
	$= \text{total asset}$		
$a_v = \ell_v$ (balance sheet equation)			

Note that the homogeneous model is completely described by the 4-tuple of parameters  $\langle G, \gamma, \beta, E \rangle$ .

### Balance Sheets and Parameters for Heterogeneous Networks

The heterogeneous version of the model is the same as its' homogeneous counterpart as described above, except that the shares of interbank exposures and external assets for different banks may be different. Formally, the following modifications are done in the homogeneous model:

- $w(e) > 0$  denotes the weight of the edge  $e \in E$  along with the constraint that  $\sum_{e \in F} w(e) = I$ .
- $\iota_v = \sum_{e=(v,v') \in F} w(e)$ , and  $b_v = \sum_{e=(v',v) \in F} w(e)$ .
- $e_v = (b_v - \iota_v) + \alpha_v \times (E - \sum_{v \in V} (b_v - \iota_v))$  for some  $\alpha_v > 0$  along with the constraint  $\sum_{v \in V} \alpha_v = 1$ . Since  $\sum_{v \in V} (b_v - \iota_v) = 0$ , this gives  $e_v = (b_v - \iota_v) + \alpha_v E$ . Consequently,  $a_v$  now equals  $b_v + \alpha_v E$ .

Denoting the  $m$ -dimensional vector of  $w(e)$ 's by  $\mathbf{w}$  and the  $n$ -dimensional vector of  $\alpha_v$ 's by  $\boldsymbol{\alpha}$ , the heterogeneous model is completely described by the 6-tuple of parameters  $\langle G, \gamma, \beta, E, \mathbf{w}, \boldsymbol{\alpha} \rangle$ .

### Idiosyncratic Shock [19, 38]

As in [38], our initial failures are caused by *idiosyncratic shocks* which can occur due to *operations risks* (frauds) or *credit risks*, and has the effect of reducing the external assets of a selected subset of banks perhaps causing them to default. While *aggregated* or *correlated* shocks affecting all banks simultaneously is relevant in practice, idiosyncratic shocks are a cleaner way to study the *stability* of the topology of the banking network. Formally, we select a non-empty subset of nodes (banks)  $\emptyset \subset V_{\text{shock}} \subseteq V$ . For all nodes  $v \in V_{\text{shock}}$ , we simultaneously decrease their external assets from  $e_v$  by  $s_v = \Phi e_v$ , where the parameter  $\Phi \in (0, 1]$  determines the “severity” of the shock. As a result, the new net worth of  $\text{Bank}_v$  becomes  $c'_v = c_v - s_v$ . The effect of this shock is as follows:

- If  $c'_v \geq 0$ ,  $\text{Bank}_v$  continues to operate but with a lower net worth of  $c'_v$ .
- If  $c'_v < 0$ ,  $\text{Bank}_v$  *defaults* (i.e., stops functioning).

## Propagation of an Idiosyncratic Shock [19, 38]

We will add “(t)” to the model parameter  $c_v$  defined in this section to show their dependences on time  $t$ , *i.e.*,  $c_v(t)$  denotes  $c_v$  at time  $t$ , and we use the notation  $t_0^+$  to denote any  $t$  from the set  $\{x | x > t_0\}$ . Let  $V_{\text{alive}}(t) \subseteq V$  be the set of nodes that have not failed at time  $t$  and let  $G_{\text{alive}}(t) = (V_{\text{alive}}(t), E_{\text{alive}}(t))$  be the corresponding node-induced subgraph of  $G$  at time  $t$  with  $\deg_{\text{in}}(v, t)$  and  $\deg_{\text{out}}(v, t)$  denote the in-degree and out-degree of a node  $v \in V_{\text{alive}}(t)$  in the graph  $G_{\text{alive}}(t)$  at time  $t$ . In the continuous-time model, the shock propagates as follows:

- $V_{\text{alive}}(1) = V$ , and  $c_v(1) = \begin{cases} c_v - s_v, & \text{if } v \in V_{\text{shock}} \\ c_v, & \text{otherwise} \end{cases}$
- If a bank's equity ever becomes negative, then the bank fails subsequently. That is, for any  $t_0 \geq 1$ , if  $c_v(t_0) < 0$  then  $v \notin V_{\text{alive}}(t_0^+)$ .

- A failed bank  $\text{Bank}_v$  at time  $t = t_0$  ( $c_v(t_0) < 0$ ) affects the net worth (equity) of all banks that gave loan to  $\text{Bank}_v$  in the following manner. For each edge  $(u, v) \in E_{\text{alive}}(t_0)$  in the network at time  $t_0$ , the equity  $c_u(t_0)$  is decreased by an amount<sup>1</sup> of  $\frac{\min\{|c_v(t_0)|, b_v\}}{\deg_{\text{in}}(v, t_0)}$ . Thus, the shock propagation is defined by the following partial differential equation:

$$\forall t \geq 1 \quad \forall u \in V_{\text{alive}}(t): \quad \frac{\partial c_u(t)}{\partial t} = - \sum_{\substack{v: c_v(t) < 0 \\ (u, v) \in E_{\text{alive}}(t)}} \frac{\min\{|c_v(t)|, b_v\}}{\deg_{\text{in}}(v, t)}$$

A *discrete-time* version of the above can be obtained by the obvious method of quantizing time and replacing the partial differential equations by “difference equations”. With appropriate normalizations, the discrete-time model for shock propagation is described by a synchronous iterative procedure shown in Table 1 where  $t = 1, 2, \dots, T$  denotes the discrete time step at which the synchronous update is done ( $T \leq n$ ).

**Parameter Simplification** We can assume without loss of generality that in the homogeneous shock propagation model  $w = 1$ . To observe this, if  $w = I/m \neq 1$ , then we can divide each of the quantities  $t_v$ ,  $b_v$ ,  $E$  and

<sup>1</sup>If  $|c_v(t_0)| > b_v$ , then the depositors also incur a loss of  $b_v - |c_v(t_0)|$ , but we ignore this for our problem since we are only interested in the demise of banks. In other words, this model assumes that all the depositors are insured for their deposits via the bank, *e.g.*, in United States the Federal Deposit Insurance Corporation provides such an insurance up to a maximum level.

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(* time starts at $t = 1$ and increments in steps of 1 *)
$t = 1 ; V_{\text{alive}}(1) = V$
(* start the shock at $t = 1$ on nodes in $V_{\text{shock}}$ *)
<b>for</b> each bank $v \in V$ <b>do</b>
<b>if</b> $v \in V_{\text{shock}}$ <b>then</b> $c_v(1) = c_v - \Phi e_v$ <b>else</b> $c_v(1) = c_v$
<b>endfor</b>
(* shock propagation at times $t = 2, 3, \dots, T$ *)
<b>while</b> $(t \leq T) \wedge (V_{\text{alive}}(t) \neq \emptyset)$ <b>do</b>
(* transmit loss to next time step *)
<b>for</b> every $u \in V_{\text{alive}}(t)$ <b>do</b>
$c_u(t+1) = c_u(t) - \sum_{\substack{v: c_v(t) < 0 \\ (u, v) \in E_{\text{alive}}(t)}} \frac{\min\{ c_v(t) , b_v\}}{\deg_{\text{in}}(v, t)}$
<b>endfor</b>
(* remove $\text{Bank}_v$ from the network if it is supposed to fail at this step *)
$V_{\text{alive}}(t+1) = V_{\text{alive}}(t)$
<b>for</b> every $v \in V_{\text{alive}}(t)$ <b>do</b>
<b>if</b> $c_v(t) < 0$ <b>then</b> $V_{\text{alive}}(t+1) = V_{\text{alive}}(t+1) \setminus \{v\}$
<b>endfor</b>
$t = t + 1$
<b>endwhile</b>

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Table 1: Discrete-time idiosyncratic shock propagation from a subset  $V_{\text{shock}}$  of nodes up to  $T$  steps.

$d_v$  by  $w$ ; it is easy to see that the outcome of the shock propagation procedure in Table 1 remains the same. Moreover, we will ignore the balance sheet equation since  $d_v$  has no effect in shock propagation.

### 3 Related Prior Works on Financial Networks

Although there is a large amount of literature on stability of financial systems in general and banking systems in particular, much of the prior research is on the empirical side or applicable to small-size networks. Two main categories of prior researches can be summarized as follows.

#### Network formation

Babus [6] proposed a model in which banks form links with each other as an insurance mechanism to reduce the risk of contagion. In contrast, Castiglionesi and Navarro [11] studied decentralization of the network of banks that is optimal from the perspective of a social planner. In a setting in which banks invest on behalf of depositors and there are positive network externalities on the investment returns, fragility arises when “not sufficiently capitalized” banks gamble with depositors’ money. When the probability of bankruptcy is low, the decentralized solution well-approximates the first objective of Babus.

#### Contagion spread in a given network

Although ordinarily one would expect the risk of contagion to be larger in a highly interconnected banking system, some empirical simulations indicate that shocks may have an *extremely complex* effect on the stability of the network in the sense that, higher connectivity between the banks may sometimes lead to *lower risk of contagion*.

Allen and Gale [2] studied how a banking system may respond to contagion when banks are connected under different network structures. In a setting where consumers have the liquidity preferences as introduced by Diamond and Dybvig [17], and have random liquidity needs, banks perfectly insure against liquidity fluctuations by exchanging interbank deposits. The connections created by swapping deposits, however, expose the entire system to contagion. Allen and Gale concluded that incomplete networks are *more prone* to contagion than networks with maximum connectivity since better-connected networks are more resilient because the proportion of the losses in one bank’s portfolio is transferred to more banks through interbank agreements. Freixas, Parigi and Rochet [24] similarly explored the case of banks that face liquidity fluctuations due to the uncertainty about when the consumers will withdraw funds. Gai and Kapadia [25] argued that the higher is the connectivity among banks the more will be the contagion effect during crisis. Haldane [27] suggested that contagion should be measured based on the interconnectedness of each institution within the financial system. Liedorp *et al.* [35] investigated if interconnectedness in the interbank market is a channel through which banks affect each others “riskiness” and showed that both large lending and borrowing shares in interbank markets increase the riskiness of banks active in the dutch banking market.

Dasgupta [16] explored how linkages between banks, represented by cross-holding of deposits, can be a source of contagious breakdowns. His study examined how depositors, who receive a private signal about fundamentals of banks, may want to withdraw their deposits if they believe that enough other depositors will do the same. Lagunoff and Schreft [34] considered a model in which agents are linked in the sense that the return on an agents’ portfolio depends on the portfolio allocations of other agents. Iazzetta and Manna [28] used network topology analysis on monthly data on deposits exchange to gain more insight into the way a liquidity crisis spreads. Nier *et al.* [38] analyzed how systemic risk depends on the structure of the banking system via network theoretic approach to construct banking systems and then to analyze the resilience of the system to contagious defaults. Kleindorfer, Wind and Gunther [32] argued that network analyses can play a crucial role in understanding many important phenomena in finance. Corbo and Demange [15] investigated

how the structure of interbank connections is related to the contagion risk of defaults, given the exogenous default of a bank or set of banks. Babus [7] studied how the trade-off between the benefits and the costs of being linked changes depending on the network structure, and observed that, when the network is maximal, the liquidity can be redistributed in the system such that the risk of contagion is minimal.

The particular model used in this paper for studying stability is the homogeneous model formulated by Nier *et al.* [38] and its heterogeneous version. As we have already stated below, definition of a precise stability measure and analysis of its computational complexity issues for stability calculation were not provided for these models before.

## 4 The Stability and Dual Stability Indices

A banking network is called *dead* if all the banks in the network have failed. Consider a given (homogeneous or heterogeneous) banking network  $\langle G, \gamma, \beta, E, \Phi \rangle$  or  $\langle G, \gamma, \beta, E, \Phi, \mathbf{w}, \alpha \rangle$ . For a subset of nodes  $V'$ , let

$$\text{influenced}(V') = \{v \in V \mid v \text{ fails if all nodes in } V' \text{ are shocked}\}, \text{ and } \text{SI}(G, V', T) = \begin{cases} \frac{|V'|}{n}, & \text{if } \text{influenced}(V') = V \\ \infty, & \text{otherwise} \end{cases}$$

**The Stability Index** The optimal *stability index* of a network  $G$  can then be defined as

$$\text{SI}^*(G, T) = \text{SI}(G, V_{\text{shock}}, T) = \min_{V'} \{ \text{SI}(G, V', T) \}$$

For estimation of this measure, we assume that it is possible for the network to fail, *i.e.*,  $\text{SI}^*(G, T) < \infty$ . Thus,  $0 < \text{SI}^*(G, T) \leq 1$ , and the higher the stability index is, the better is the stability of the network against an idiosyncratic shock. We thus arrive at the natural computational problem.

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**Problem name:** Stability of banking networks ( $\text{STABILITY}_{T, \Phi}$ ).

**Input:** a banking network with  $\Phi$  is the shocking parameter, and an integer  $T > 1$ .

**Valid solution:** A subset  $V' \subseteq V$  such that  $\text{SI}(G, V', T) < \infty$ .

**Objective:** *minimize*  $|V'|$ .

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We denote an optimal subset of nodes that is a solution of Problem  $\text{STABILITY}_{T, \Phi}$  by  $V_{\text{shock}}$ , *i.e.*,  $\text{SI}^*(G, T) = \text{SI}(G, V_{\text{shock}}, T)$ . Note that if  $T \geq n$  then the  $\text{STABILITY}_{T, \Phi}$  finds a minimum subset of nodes which, when shocked, will *eventually* cause the death of the network in an arbitrary number of time steps.

**The Dual Stability Index** Many covering-type minimization problems in combinatorics have a natural maximization dual in which one fixes a-priori the number of covering sets and then finds a maximum number of elements that can be covered with these many sets. For example, the usual dual of the minimum set covering problem is the so-called maximum coverage problem [31]. Analogously, we can define a dual of the stability measure:

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**Problem name:** Dual Stability of Financial Network ( $\text{DUAL-STABILITY}_{T, \Phi, \kappa}$ ).

**Input:** a banking network with  $\Phi$  as the shocking parameter, and two integers  $T, \kappa > 1$ .

**Valid solution:** A subset  $V' \subseteq V$  such that  $|V'| = \kappa$ .

**Objective:** *maximize*  $|\text{influenced}(V')|/\kappa$ .

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The *dual stability index* of a network  $G$  can then be defined as  $\text{DSI}^*(G, T, \kappa) = \max_{V' \subseteq V: |V'| = \kappa} \left| \frac{\text{influenced}(V')}{\kappa} \right|$ .

The dual stability measure is of particular interest when  $\text{SI}^*(G, T) = \infty$ , *i.e.*, the entire network cannot be

made to fail. In this case, a natural goal is to find out if a significant portion of the nodes in the network can be failed by shocking a limited number of nodes of  $G$ ; this is captured by the definition of  $\text{DSI}^*(G, T, \kappa)$ .

**Violent Death Vs. Slow Poisoning** In our results, we distinguish two cases of death of a network:

**Violent Death** ( $T = 2$ ): The network is dead by the very next step after the shock.

**Slow Poisoning** (any  $T \geq 2$ ): The network may not be dead immediately, but does die eventually.

## 5 Comparison with Other Models for Attribute Propagation in Networks

Models for propagation of beneficial or harmful attributes have been investigated in the past in several other contexts such as influence maximization in social networks [10, 12, 13, 29], disease spreading in urban networks [14, 20, 21], and percolation models in physics and mathematics [40]. However, the model for shock propagation in financial network discussed in this paper is *fundamentally* very different from all these models. Some distinguishing features of our model include:

(a) Almost all of these models include a trivial solution in which the attribute spreads to the entire network if we inject each node individually with the attribute. This is not the case with our model: *a node may not fail when shocked, and the network may not be dead if all nodes are shocked*. For example, consider the network in Fig. 1(i).

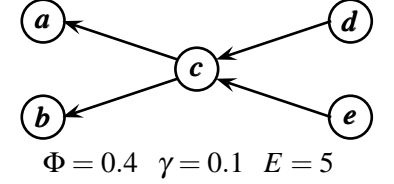


Figure 1: A homogeneous network used in the discussion in Section 5.

- Suppose that all the nodes are shocked. Then, the following events happen.
  - Node  $a$  (and similarly node  $b$ ) fails at  $t = 1$  since  $\Phi(\deg_{\text{in}}(a) + \frac{E}{5}) > \gamma(\deg_{\text{in}}(a) + \frac{E}{5})$ .
  - Node  $c$  also fails at  $t = 1$  since  $\Phi(\deg_{\text{in}}(c) - \deg_{\text{out}}(c) + \frac{E}{5}) = 0.4 > \gamma(\deg_{\text{in}}(c) + \frac{E}{5}) = 0.3$ .
  - Node  $d$  (and similarly node  $e$ ) do not fail at  $t = 1$  since  $\Phi(-\deg_{\text{out}}(d) + \frac{E}{5}) = 0 < \gamma \times \frac{E}{5} = 0.1$  and its equity stays at  $0.1 - 0 = 0.1$ .
  - At  $t = 2$ , node  $d$  (and similarly node  $e$ ) receives a shock from node  $c$  of the amount  $\frac{0.4-0.3}{2} = 0.05 < 0.1$ . Thus, nodes  $d$  and  $e$  do not fail. Since no new nodes fail during  $t > 2$ , the network does not become dead.
- However, suppose that only nodes  $a$  and  $b$  are shocked. Then, the following events happen.
  - Node  $a$  (and similarly node  $b$ ) fails at  $t = 1$  since  $\Phi(\deg_{\text{in}}(a) + \frac{E}{5}) = 0.8 > \gamma(\deg_{\text{in}}(a) + \frac{E}{5}) = 0.2$ .
  - At  $t = 2$ , node  $c$  receives a shock of the amount  $2 \times (0.8 - 0.2) = 1.2 > \gamma(\deg_{\text{in}}(c) + \frac{E}{5}) = 0.3$ . Thus, node  $c$  fails at  $t = 2$ .
  - At  $t = 3$ , node  $d$  (and similarly node  $e$ ) receives a shock of the amount  $\frac{1.2-0.3}{2} = 0.45 > \gamma \times \frac{E}{5} = 0.1$ . Thus, both these nodes fail at  $t = 3$  and the entire network is dead.

As the above example shows, if shocking a subset of nodes makes a network dead, adding more nodes to this subset may *not* necessarily lead to the death of the network, and the stability measure is *neither monotone nor sub-modular*. On the other hand, it is not difficult to construct examples of banking networks such that to make the entire network fail:

- it may be necessary to shock a node even if it does not fail since shocking such a node “weakens” it by decreasing its equity, and

- it may be necessary to shock a node even if it fails due to shocks given to other nodes.

(b) The complexity of the computational aspects of many previous attribute propagation models arise due to the presence of cycles in the graph; for example, see [12] for polynomial-time solutions of some of these problems when the underlying graph does not have a cycle. In contrast, our computational problems are may be hard *even when the given graph is acyclic*; instead, a key component of computational complexity arises due to two or more directed paths sharing a node.

Network and result types		Stability $\text{SI}^*(G, T)$ & assumptions (if any)	Dual Stability $\text{DSI}^*(G, T, \kappa)$ & assumptions (if any)
Homogeneous	Acyclic, $T = 2$ approximation-hardness	$(1 - \varepsilon) \ln n$ , $\text{NP} \not\subseteq \text{DTIME}(n^{\log \log n})$	
	Acyclic, $T = 2$ , approximation ratio	$O\left(\log n + \log \frac{\Phi}{\gamma} + \log \frac{1}{\Phi - \gamma} + \log \frac{E}{ E - \Phi }\right)$	
	Acyclic, any $T > 1$ , approximation-hardness	APX-hard	$(1 - e^{-1} + \varepsilon)^{-1}$ , $\text{P} \neq \text{NP}$
	In-arborescence, any $T > 1$ , exact solution	$O(n^2)$ time, every node fails when shocked	$O(n^3)$ time, every node fails when shocked
Heterogeneous	Acyclic, any $T > 1$ , approximation-hardness	$(1 - \varepsilon) \ln n$ , $\text{NP} \not\subseteq \text{DTIME}(n^{\log \log n})$	$(1 - e^{-1} + \varepsilon)^{-1}$ , $\text{P} \neq \text{NP}$
	Acyclic, $T = 2$ , approximation-hardness		$n^\delta$ , under assumption $(\star)^\dagger$
	Acyclic, any $T > 3$ , approximation-hardness	$2^{\log^{1-\varepsilon} n}$ , $\text{NP} \not\subseteq \text{DTIME}(n^{\text{poly}(\log n)})$	
	Acyclic, $T = 2$ , approximation ratio $^\ddagger$	$O\left(\log n + \log\left(\frac{1}{\Phi}\right) + \log\left(\frac{1}{\gamma}\right) + \log\left(\frac{1}{\Phi - \gamma}\right) + \log\left(\frac{1}{\underline{E}}\right) + \log \bar{E} + \log \bar{w}_{\max} + \log \bar{w}_{\min} + \log \bar{\alpha}_{\max} + \log\left(\frac{1}{\bar{w}_{\min}}\right) + \log\left(\frac{1}{\bar{\alpha}_{\min}}\right) + \log\left(\frac{1}{\bar{w}_{\max}}\right)\right)$	

$^\ddagger$ See Theorem 9.2 for definitions of some parameters in the approximation ratio.

$^\dagger$ See page 37 for statement of assumption  $(\star)$ , which is weaker than the assumption  $\text{P} \neq \text{NP}$ .

Table 2: A summary of our results;  $\varepsilon > 0$  is any arbitrary constant and  $0 < \delta < 1$  is some constant.

## 6 Overview of Our Results and Implications on Banking Systems

We group our results based on whether the network is homogeneous or heterogeneous. For each type of network, we investigate two categories of the stability measure, namely  $T = 2$  vs.  $T > 2$ . A summary of the results proved in this paper appear below in Table 2. *All of our hardness results hold even if the given directed graph is acyclic.* In the sequel, the notation  $\text{poly}(x_1, x_2, \dots, x_k)$  denotes a constant-degree polynomial in the variables  $x_1, x_2, \dots, x_k$ .



## 6.1 Homogeneous Networks

**$T = 2$**  We first show in Theorem 8.1 that, assuming  $\text{NP} \not\subseteq \text{DTIME}(n^{\log \log n})$ ,  $\text{SI}^*(G, 2)$  cannot be approximated to within a factor of  $(1 - \varepsilon) \ln n$  for any constant  $\varepsilon > 0$ . This is done by reducing a corresponding inapproximability result for the dominating set problem for general graphs to our problem. We complement this result by showing, in Theorem 8.2, a logarithmic approximation for computing  $\text{SI}^*(G, 2)$  that almost matches our lower bound. Even though our algorithmic problem can be cast as some type of set covering problem, we *cannot* explicitly enumerate all the necessary sets since there are *exponentially many* such sets. Thus instead we reformulate our problem to that of computing an optimal solution of an appropriate polynomial-size integer linear programming (ILP) problem with non-negative coefficients and then use the greedy approach of [18] to approximate the optimal solution of this reformulation. A careful calculation of the size of the coefficients of the ILP ensures that we have the desired approximation bound.

**Arbitrary  $T$**  We first show in Theorem 8.3 that computing  $\text{SI}^*(G, T)$  is APX-hard for any  $T \geq 2$ , even if the in- and out-degrees of all nodes are small constants, by giving an  $L$ -reduction from the node cover problem for 3-regular graphs to our problem. Technical complications arise from making sure that the transformed graph instance of  $\text{STABILITY}_{T, \Phi}$  has no cycles, small diameter (to ensure that no new nodes fail after a small number of time steps), but yet all nodes must fail within a few time steps without each node being individually shocked.

We next turn our attention to designing efficient algorithms for computing  $\text{SI}^*(G, T)$  for arbitrary  $T$ . If the given graph is a rooted in-arborescence and assuming every node can be individually shocked to fail, we show how to design an  $O(n^2)$  time exact algorithm via dynamic programming. As a by product of this approach, we also show that the stability of this type of network with bounded degrees is large.

For the dual stability measure, we show in Theorem 10.1(a) that, assuming  $\text{P} \neq \text{NP}$ ,  $\text{DSI}^*(G, T, \kappa)$  cannot be approximated to within a factor of  $(1 - (1/e) + \delta)^{-1}$  for any  $\delta > 0$ . This is obtained by translating a lower bound of Feige for the maximum coverage problem [22] to our problem. The reduction takes advantage of the fact that, in dual stability measure, every node of the network need not fail. If the given graph is a rooted in-arborescence and assuming every node can be individually shocked to fail, we show how to design an  $O(n^3)$  time exact algorithm via dynamic programming.

## 6.2 Heterogeneous Networks

Our results for heterogeneous networks show that the problem of computing stability indices is harder than in homogeneous networks, as one would naturally expect.

**Any  $T \geq 2$**  We show in Theorem 9.1 that, assuming  $\text{NP} \not\subseteq \text{DTIME}(n^{\log \log n})$ ,  $\text{SI}^*(G, T)$  for any  $T \geq 2$  cannot be approximated to within a factor of  $(1 - \varepsilon) \ln n$  for any constant  $\varepsilon > 0$ . This is done by reducing a corresponding inapproximability result for the minimum set covering problem to our problem. The inapproximability result is stronger than the corresponding result for homogeneous networks in Theorem 8.3. Unlike homogeneous networks, unequal shares of the total external assets by various banks allows us to “equalize” weighted degrees of nodes and thus an arbitrary instance of set cover can be encoded. For  $T = 2$ , Theorem 9.2, using the same linear program as in Theorem 8.2, provides an algorithm whose approximation ratio is logarithmic in the parameters of the network.

For the dual stability measure, we show in Theorem 11.1 that, under a complexity-theoretic assumption,  $\text{SI}^*(G, 2, \kappa)$  cannot be approximated within an approximation ratio of  $n^\delta$  for some constant  $0 < \delta < 1$ .

**Any  $T > 3$**  For this case, we prove in Theorem 9.3 an even stronger poly-logarithmic inapproximability result than that in Theorem 9.1, namely that, under the assumption of  $\text{NP} \not\subseteq \text{DTIME}(n^{\text{poly}(\log n)})$ , it is

impossible to approximate  $SI^*(G, T)$  within a factor of  $2^{\log^{1-\varepsilon} n}$  in polynomial time for any constant  $0 < \varepsilon < 1$ . This is achieved by translating MINREP, a graph-theoretic abstraction of two prover multi-round protocol for any problem in NP, to our problem preserving the approximation ratio. Many technical complications arise during the reduction procedure, leading to a set of 22 linear equations between the parameters that we must satisfy for our reduction to go through. Intuitively, the two provers in MINREP correspond to two super-nodes in the network that cooperate to fail to a specified set of nodes in the network.

### 6.3 Implications of Our Results on Banking Systems

**Effects of Topological Connectivity** Though many researchers agree that the connectivity of banking networks is related to its stability, the conclusions drawn are somewhat contradictory, namely some researchers conclude that incomplete networks are more prone to contagion than networks with maximum connectivity whereas some other researchers conclude that the higher is the connectivity among banks the more will be the contagion effect during crisis. Based on our results and their proofs, we found that topological connectivity does play a significant role in stability of the network in the following manner.

**Even acyclic networks display complex stability behavior:** Sometimes a cause of the instability of a banking network is attributed to *cyclical* dependencies of borrowing and lending mechanisms among major banks, *e.g.*, banks  $a$ ,  $b$  and  $c$  borrowing from banks  $b$ ,  $c$  and  $a$ , respectively. As our results show, computing the stability measures could be difficult even without the presence of such cycles. Indeed, larger inapproximability results, especially for heterogeneous results, are possible because slight change in network parameters can cause a large change in the stability measure. On the other hand, acyclic small-degree rooted in-arborescence networks exhibit higher values of the stability measure, *e.g.*, if the maximum in-degree of any node in a rooted in-arborescence is 5 and the shock parameter  $\Phi$  is no more than twice the value of the percentage of equity to assets  $\gamma$ , then by Theorem 8.4  $SI^*(G, T) > 0.1$ .

**Intersection of borrowing chains may cause lower stability:** By a *borrowing chain* we mean a directed path from a node  $a$  to another node  $b$ , indicating that bank  $b$  effectively borrowed from bank  $a$  through a sequence of successive intermediaries. Now, assume that there is another directed path from  $a$  to another node  $c$ . Then, failure of  $b$  and  $c$  propagates the resulting shock to  $a$  and, if the shocks arrive at the same step, then the total shock received by bank  $a$  is the addition of these two shocks, which in turn passes to other nodes in the network through  $a$ . For example, in Fig. 1,  $c$  receives the addition of two shocks transmitted by  $a$  and  $b$ , which in turn suffices to make  $c$ ,  $d$  and  $e$  fail.

**Effects of Ratio of External to Internal Assets  $E/I$  and percentage of equity to assets  $\gamma$  for Homogeneous Networks** As our relevant results and their proofs show, lower values of  $E/I$  and  $\gamma$  may cause the network stability to be extremely sensitive with respect to variations of other parameters of the network. For example, in the proof of Theorem 8.1 we have  $\lim_{n \rightarrow \infty} E/I = 0$  and  $\lim_{n \rightarrow \infty} \gamma = 0$ , leading to variation of the stability index by a logarithmic factor; however, in the proof of Theorem 8.3 we have  $E/I = 0.25$  and  $\gamma = 0.23$  leading to much smaller variation of the stability index (by a constant factor only).

**Homogeneous vs. Heterogeneous Networks** Our results and proofs show that heterogeneous networks containing banks with diverse equities tend to exhibit a wider fluctuations of the stability index with respect to parameters, *e.g.*, Theorem 9.3 shows a super-polylogarithmic change of the stability index even if the ratio  $E/I$  is large.

## 7 Preliminary Observations on Shock Propagation

**Proposition 7.1.** *Let  $\langle G = (V, F), \gamma, \beta, E \rangle$  be the given (homogeneous or heterogeneous) banking network. Then, the following are true:*

- (a) *If  $\deg_{\text{out}}(v) = 0$  for some  $v \in V$ , then node  $v$  must be given a shock (and, must fail due to this shock) for the entire network to fail.*
- (b) *Let  $\alpha$  be the number of edges in the longest directed simple path in  $G$ . Then, no new node fails at any time  $t > \alpha$ .*
- (c) *We can assume without loss of generality that  $G$  is weakly connected, i.e., the un-oriented version of  $G$  is connected.*

*Proof.*

(a) Since  $\deg_{\text{out}}(v) = 0$ , no part of any shock given to any other nodes in the network can reach  $v$ . Thus, the network of  $v$ , namely  $c_v = \gamma a_v$  stays strictly positive (since  $\gamma > 0$ ) and node  $v$  never fails.

(b) Let  $t_{\text{last}}$  be the latest time a node of  $G$  failed, and let  $V(t)$  be the set of nodes that failed at time  $t = 1, 2, \dots, t_{\text{last}}$ . Then,  $V(1), V(2), \dots, V(t_{\text{last}})$  is a partition of  $V$ . For every  $i = 1, 2, \dots, t_{\text{last}} - 1$ , add directed edges  $(u, v)$  from a node  $u \in V(i)$  to a node  $v \in V(i+1)$  if  $u$  was last node that transmitted any part of the shock to  $v$  before  $v$  failed. Note that  $(u, v)$  is also an edge of  $G$  and for every node  $v \in V(i+1)$  there must be an edge  $(u, v)$  for some node  $u \in V(i)$ . Thus,  $G$  has a path of length at least  $t_{\text{last}}$ .

(c) This holds since otherwise the stability measures can be computed separately on each weakly connected component.  $\square$

## 8 Our Results on the Stability Index for Homogeneous Networks

### 8.1 Case of $T = 2$ : Violent Demise of the Network

#### 8.1.1 Logarithmic Inapproximability

**Theorem 8.1.**  *$\text{SI}^*(G, 2)$  cannot be approximated to within a factor of  $(1 - \epsilon) \ln n$ , for any constant  $\epsilon > 0$ , unless  $\text{NP} \subseteq \text{DTIME}(n^{\log \log n})$ .*

*Proof.* The dominating set problem for an undirected graph (DOMIN-SET) is defined as follows: given an undirected graph  $G = (V, F)$  with  $n = |V|$  nodes, find a minimum cardinality subset of nodes  $V' \subset V$  such that every node in  $V \setminus V'$  is incident on at least one edge whose other end-point is in  $V'$ . It is known that DOMIN-SAT is equivalent to the minimum set-cover problem under L-reduction [8], and thus cannot be approximated within a factor of  $(1 - \epsilon) \ln n$  unless  $\text{NP} \subseteq \text{DTIME}(n^{\log \log n})$  [22].

Consider an instance  $G = (V, F)$  of DOMIN-SET with  $n$  nodes and  $m$  edges, and let  $\text{OPT}$  denote the size of an optimal solution for this instance. Our (directed) banking network  $\vec{G} = (\vec{V}, \vec{F})$  is obtained from  $G$  by replacing each undirected edge  $\{u, v\}$  by two directed edges  $(u, v)$  and  $(v, u)$ . Thus we have  $0 < \deg_{\text{in}}(v) = \deg_{\text{out}}(v) < n$  for every node  $v \in V$ . We set the global parameters as follows:  $E = 10n$ ,  $\gamma = n^{-2}$  and  $\Phi = 1$ .

For a node  $v$ , let  $\text{Nbr}(v) = \{u \mid \{u, v\} \in E\}$  be the set of neighbors of  $v$  in  $G$ . We claim that if a node  $v$  is shocked at time  $t = 1$ , then all nodes in  $\{v\} \cup \text{Nbr}(v)$  fail at time  $t = 2$ . Indeed, suppose that  $v$  is shocked at  $t = 1$ . Then,  $v$  surely fails because

$$\Phi e_v = \deg_{\text{in}}(v) - \deg_{\text{out}}(v) + \frac{E}{n} = 10 > \frac{2}{n} > \frac{\deg_{\text{in}}(v) + \frac{E}{n}}{n^2} = \gamma a_v$$

Now, consider  $t = 2$  and consider a node  $v$  such that  $v$  has not failed but a node  $u \in \text{Nbr}(v)$  failed at time  $t = 1$ . Then, node  $v$  surely fails because

$$s_{v,2} \geq \frac{\min\{s_{u,1} - c_u, b_u\}}{\deg_{\text{in}}(u, 2)} = \frac{\min\{\Phi e_u - \gamma a_u, \deg_{\text{in}}(u)\}}{\deg_{\text{in}}(u)} > \min\left\{\frac{10 - \frac{2}{n}}{\deg_{\text{in}}(u)}, 1\right\} > \frac{2}{n} > \frac{\deg_{\text{in}}(v) + \frac{E}{n}}{n^2} = \gamma a_v$$

Thus, we have a 1-1 correspondence between the solutions of DOMIN-SET and death of  $\vec{G}$ , namely  $V' \subset V$  is a solution of DOMIN-SET if and only if shocking the nodes in  $V'$  makes  $\vec{G}$  fail at time  $t = 2$ .  $\square$

### 8.1.2 Logarithmic Approximation

The following theorem provides a polynomial-time algorithm with an approximation ratio that is logarithmic in the parameters of the banking network when  $T = 2$  that almost matches the lower bound in Theorem 8.1. We assume  $\Phi > \gamma$  since otherwise there is no solution to our problem.

**Theorem 8.2.** *There is a polynomial-time greedy algorithm for  $T = 2$  that has an approximation ratio of  $O\left(\log n + \log \Phi/\gamma + \log(1/(\Phi-\gamma)) + \log \frac{E}{|E-\Phi|}\right)$ .*

*Proof.* Suppose that  $\Phi e_u < 0$  for some node  $u \in V$ . Then, there exists an optimal solution in which we do not shock the node  $u$ . Indeed, if  $u$  was shocked, the equity of  $u$  increases from  $c_u$  to  $c_u + |\Phi e_u|$  and  $u$  does not propagate any shock to other nodes. Thus, if  $u$  still fails at  $t = 2$ , then it also fails at  $t = 2$  if it was not shocked.

Let  $V_{\text{shock}}$  denote the set of nodes that we will select for shocking, and, for every node  $v \in V$ , let  $\delta_{v,u}$  be defined as:  $\delta_{v,u} = \begin{cases} \max\{0, \Phi e_v\}, & \text{if } u = v \\ \frac{\min\{\Phi e_v - c_v, b_v\}}{\deg_{\text{in}}(v)}, & \text{if } \Phi e_v > c_v \text{ and } (u, v) \in F. \\ 0, & \text{otherwise} \end{cases}$ . Then, our problem reduces to a covering problem of the following type:

find a minimum cardinality subset  $V_{\text{shock}} \subseteq V$  such that, for every node  $u$ ,  $\sum_{v \in V_{\text{shock}}} \delta_{v,u} > c_u$ .

Note that we cannot even explicitly enumerate, for a node  $u \in V$ , all subsets  $V' \subseteq V \setminus \{u\}$  such that  $\sum_{v \in V'} \delta_{v,u} > c_u$ , since there are exponentially many such subsets. Let the binary variable  $x_v \in \{0, 1\}$  be the indicator variable for a node  $v \in V$  for inclusion in  $V_{\text{shock}}$ . However, we can reformulate our problem as the following integer linear programming problem:

$$\begin{aligned} & \text{minimize } \sum_{v \in V} x_v \\ & \text{subject to } \forall u \in V: \sum_{v \in V} \delta_{v,u} x_v > c_u \\ & \quad x_u \in \{0, 1\} \end{aligned} \tag{1}$$

Let  $\zeta = \min_{u \in V} \left\{ \min_{v \in V} \{\delta_{u,v}\}, c_u \right\}$ . We can rewrite each constraint  $\sum_{v \in V} \delta_{v,u} x_v > c_u$  as  $\sum_{v \in V} \frac{\delta_{v,u}}{\zeta} x_v > \frac{c_u}{\zeta}$  to ensure that every non-zero entry is at least 1. Since the coefficients of the constraints and the objective function are all positive real numbers, (1) can be approximated by the greedy algorithm described in [18, Theorem 4.1] with an approximation ratio of  $2 + \ln n + \ln \left( \max_{v \in V} \left\{ \sum_{u \in V} \frac{\delta_{v,u}}{\zeta} \right\} \right)$ . Now, observe that:

$$\min_{\substack{u \in V \\ \delta_{u,u} > 0}} \{\delta_{u,u}\} = \min_{\substack{u \in V \\ \delta_{u,u} > 0}} \left\{ \Phi \left( \deg_{\text{in}}(u) - \deg_{\text{out}}(u) + \frac{E}{n} \right) \right\} = \Omega \left( \frac{|E - \Phi|}{n} \right)$$

$$\min_{u \in V} \min_{\substack{v \in V \\ \delta_{u,v} > 0}} \{\delta_{u,v}\} = \min_{u \in V} \min_{\substack{v \in V \\ \Phi_{e_v} > c_v}} \left\{ (\Phi - \gamma) \left( 1 + \frac{E}{\deg_{\text{in}}(v)} \right) - \Phi \frac{\deg_{\text{out}}(v)}{\deg_{\text{in}}(v)} \right\} = \Omega \left( \frac{(\Phi - \gamma)E}{n} \right)$$

$$\min_{u \in V} \{c_u\} = \min_{u \in V} \left\{ \gamma \left( \deg_{\text{in}}(u) + \frac{E}{n} \right) \right\} = \Omega \left( \frac{\gamma E}{n} \right)$$

$$\zeta = \min \left\{ \min_{u \in V} \min_{v \in V} \{\delta_{u,v}\}, \min_{u \in V} \{c_u\} \right\} = \Omega \left( \min \left\{ \frac{|E - \Phi|}{n}, \frac{(\Phi - \gamma)E}{n}, \frac{\gamma E}{n} \right\} \right)$$

$$\max_{v \in V} \sum_{u \in V} \delta_{v,u} \leq n \max_{u \in V} \left\{ (\Phi - \gamma) \left( 1 + \frac{E}{\deg_{\text{in}}(u)} \right) - \Phi \frac{\deg_{\text{out}}(u)}{\deg_{\text{in}}(u)} \right\} = O(n(\Phi - \gamma)E)$$

and thus,  $\max_{v \in V} \left\{ \sum_{u \in V} \frac{\delta_{v,u}}{\zeta} \right\} = O \left( \text{poly} \left( n, \frac{\Phi}{\gamma}, \frac{1}{\Phi - \gamma}, \frac{E}{|E - \Phi|} \right) \right)$ , giving the approximation bound.  $\square$

## 8.2 Arbitrary $T$ : Slow Poisoning of the Network

### 8.2.1 APX-hardness for any $T \geq 2$

#### Acyclic Network

We show that  $\text{STABILITY}_{T,\Phi}$  is hard for any  $T \geq 2$  even for restricted types of directed acyclic graphs (DAG).

**Theorem 8.3.** *For any  $T \geq 2$ , computing  $\text{SI}^*(G)$  is APX-hard even if the banking network satisfies  $\vec{G}$  all the following conditions:*

- $\vec{G}$  is a DAG, and
- $\deg_{\text{in}}(v) \leq 3$  and  $\deg_{\text{out}}(v) \leq 2$  for every node  $v$ .

*Proof.* We reduce the 3-MIN-NODE-COVER problem to  $\text{STABILITY}_{T,\Phi}$ . 3-MIN-NODE-COVER is defined as follows. We are given an undirected 3-regular graph  $G$ , i.e., an undirected graph  $G = (V, F)$  in which the degree of every node is exactly 3 (and thus  $|F| = 1.5|V|$ ). A valid solution (node cover) is a subset of nodes  $V' \subseteq V$  such that every edge is incident to at least one node in  $V'$ . The goal is then to find a node cover  $V' \subseteq V$  such that  $|V'|$  is *minimized*. This problem is known to be APX-hard [9].

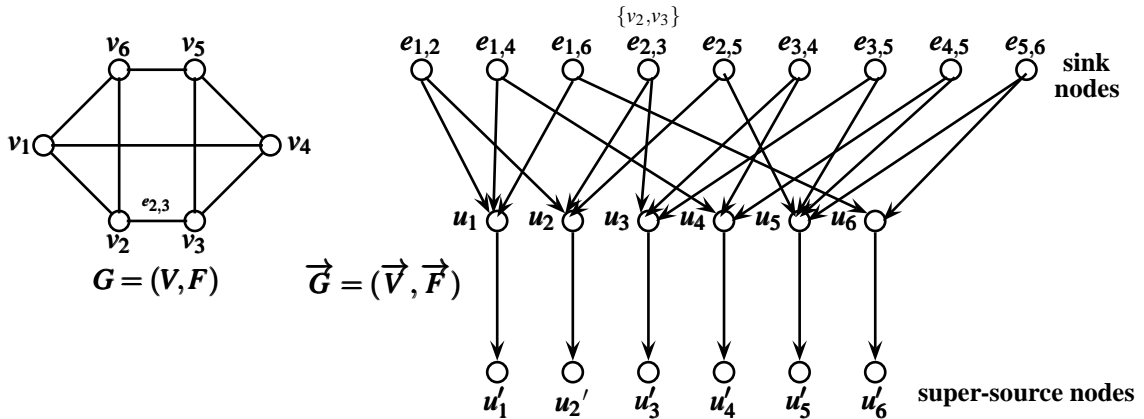


Figure 2: A 3-regular graph  $G = (V, F)$  and its corresponding banking network  $\vec{G} = (\vec{V}, \vec{F})$ .

Given such an instance  $G = (V, F)$  of 3-MIN-NODE-COVER, we construct an instance of the banking network  $\vec{G} = (\vec{V}, \vec{F})$  as follows:

- For every node  $v_i \in V$ , we have two nodes  $u_i, u'_i$  in  $\vec{V}$ , and a directed edge  $(u_i, u'_i)$ . We refer to  $u'_i$  as a “super-source” node.
- For every edge  $\{v_i, v_j\} \in F$  with  $i < j$ , we have a (“sink”) node  $e_{i,j}$  in  $\vec{V}$  and two directed edges  $(e_{i,j}, u_i)$  and  $(e_{i,j}, u_j)$  in  $\vec{F}$ . For notational convenience, the node  $e_{i,j}$  is also sometimes referred to as the node  $e_{j,i}$ .

Thus,  $|\vec{V}| = 3.5|V|$ , and  $|\vec{F}| = 4|V|$ . See Fig. 2 for an illustration. Observe that:

- $\deg_{\text{in}}(u_i) = 3$  and  $\deg_{\text{out}}(u_i) = 1$  for all  $i = 1, 2, \dots, |V|$ .
- $\deg_{\text{in}}(u'_i) = 1$  and  $\deg_{\text{out}}(u'_i) = 0$  for all  $i = 1, 2, \dots, |V|$ . Thus, by Proposition 7.1(a), every node  $u'_i$  must be shocked to make the network fail.
- $\deg_{\text{in}}(e_{i,j}) = 0$  and  $\deg_{\text{out}}(e_{i,j}) = 2$  for all  $i$  and  $j$ . Since  $\deg_{\text{in}}(e_{i,j}) = 0$ , if a node  $e_{i,j}$  is shocked, no part of the shock is propagated to any other node in the network.
- Since the longest path in  $\vec{G}$  has 2 edges, by Proposition 7.1(b) no new node fails at any  $t > 3$ .

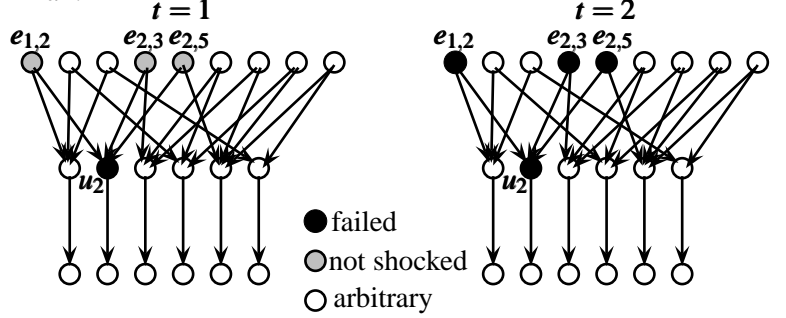


Figure 3: Case (III): if node  $u_2$  is shocked then the nodes  $e_{1,2}, e_{2,3}$  and  $e_{2,5}$  must fail at  $t = 2$ .

For notational convenience, let  $n = |V|$ ,  $\mathcal{E} = E/n$ , and  $e_{i,j_1}, e_{i,j_2}$  and  $e_{i,j_3}$  be the three edges  $\{v_i, v_{j_1}\}$ ,  $\{v_i, v_{j_2}\}$  and  $\{v_i, v_{j_3}\}$  in  $G$  that are incident on the node  $v_i$ . We will select the remaining network parameters, namely  $\gamma, \Phi$  and  $\mathcal{E}$ , based on the following desirable properties.

(I) If a node  $u'_i$  is shocked at  $t = 1$ , it fails:

$$\Phi (\deg_{\text{in}}(u'_i) - \deg_{\text{out}}(u'_i) + \mathcal{E}) > \gamma (\deg_{\text{in}}(u'_i) + \mathcal{E}) \equiv \Phi (1 + \mathcal{E}) > \gamma (1 + \mathcal{E}) \equiv \Phi > \gamma \quad (2)$$

(II) If a node  $e_{i,j}$  is shocked, it does not fail:

$$\deg_{\text{in}}(e_{i,j}) - \deg_{\text{out}}(e_{i,j}) + \mathcal{E} < 0 \equiv \mathcal{E} < 2 \quad (3)$$

(III) If a node  $u_i$  is shocked at  $t = 1$ , then  $u_i$  fails at  $t = 1$ , and the nodes  $e_{i,j_1}, e_{i,j_2}$  and  $e_{i,j_3}$  fail at time  $t = 2$  if they were not shocked (see Fig. 3 for an illustration):

$$\begin{aligned} & \frac{\min \{ \Phi (\deg_{\text{in}}(u_i) - \deg_{\text{out}}(u_i) + \mathcal{E}) - \gamma (\deg_{\text{in}}(u_i) + \mathcal{E}), \deg_{\text{in}}(u_i) \}}{\deg_{\text{in}}(u_i)} > \gamma (\deg_{\text{in}}(e_{i,j_1}) + \mathcal{E}) \\ & \equiv \frac{\min \{ \Phi (2 + \mathcal{E}) - \gamma (3 + \mathcal{E}), 3 \}}{3} > \gamma \mathcal{E} \end{aligned}$$

The above inequality is satisfied provided:

$$\Phi (2 + \mathcal{E}) > \gamma (3 + 4\mathcal{E}) \quad (4)$$

$$1 > \gamma \mathcal{E} \equiv \gamma < \frac{1}{\mathcal{E}} \quad (5)$$

(IV) Consider a sink node  $e_{i,j}$ . Then, we require that if one or both of the super-source node  $u'_i$  and  $u'_j$  are shocked at  $t = 1$  but the none of the nodes  $u_i, u_j$  and  $e_{i,j}$  were shocked, then we require that one or both of the corresponding nodes  $u_i$  and  $u_j$  fail at  $t = 2$ , but the node  $e_{i,j}$  never fails. Pictorially, we want a situation as depicted in Fig. 4. This is satisfied provided the following inequalities hold:

**(IV-1)**  $u_i$  fails at  $t = 2$  if  $u'_i$  was shocked (the case of  $u_j$  and  $u'_j$  is similar):

$$\begin{aligned} \frac{\min \{ \Phi (\deg_{\text{in}}(u'_i) - \deg_{\text{out}}(u'_i) + \mathcal{E}) - \gamma (\deg_{\text{in}}(u'_i) + \mathcal{E}), \deg_{\text{in}}(u'_i) \}}{\deg_{\text{in}}(u'_i)} &> \gamma (\deg_{\text{in}}(u_i) + \mathcal{E}) \\ &\equiv \frac{\min \{ (\Phi - \gamma)(1 + \mathcal{E}), 1 \}}{1} > \gamma(3 + \mathcal{E}) \end{aligned}$$

The above inequality is satisfied provided:

$$(\Phi - \gamma)(1 + \mathcal{E}) > \gamma(3 + \mathcal{E}) \equiv \Phi(1 + \mathcal{E}) > \gamma(4 + 2\mathcal{E}) \quad (6)$$

$$1 > \gamma(3 + \mathcal{E}) \equiv \gamma < \frac{1}{3 + \mathcal{E}} \quad (7)$$

**(IV-2)**  $e_{i,j}$  never fails even if both  $u_i$  and  $u_j$  have failed:

$$\frac{\min \{ (\Phi - \gamma)(1 + \mathcal{E}), 1 \}}{1} - \gamma(3 + \mathcal{E}) \leq \frac{\gamma \mathcal{E}}{2} \equiv \min \{ (\Phi - \gamma)(1 + \mathcal{E}), 1 \} \leq 3\gamma \left(1 + \frac{\mathcal{E}}{2}\right)$$

The above inequality is satisfied provided:

$$(\Phi - \gamma)(1 + \mathcal{E}) \leq 3\gamma \left(1 + \frac{\mathcal{E}}{2}\right) \equiv \Phi(1 + \mathcal{E}) \leq \gamma \left(4 + \frac{5\mathcal{E}}{2}\right) \quad (8)$$

$$1 \leq 3\gamma \left(1 + \frac{\mathcal{E}}{2}\right) \equiv \gamma \geq \frac{2}{6 + 3\mathcal{E}} \quad (9)$$

There are obviously many choices

of parameters  $\gamma$ ,  $\Phi$  and  $\mathcal{E}$  that satisfy Equations (2)–(9); here we exhibit just one. Let  $\mathcal{E} = 1$  which satisfied Equation (3). Choosing  $\gamma = 0.23$  satisfies Equations (5), (7) and (9). Letting  $\Phi = 0.7$  satisfies Equations (2), (4), (6) and (8).

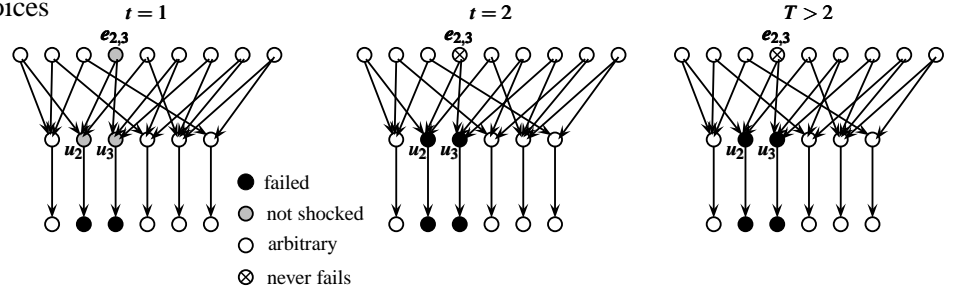


Figure 4: Case **(IV)**: to make  $e_{2,3}$  fail, at least one of  $u_2$  or  $u_3$  must be shocked.

Suppose that  $V' \subset V$  is a solution of 3-MIN-NODE-COVER. Then, we shock all the super-nodes, and the nodes in  $V'$ . By **(I)** and **(III)** all the super-nodes and the nodes in  $(\cup_{v_i \in V \setminus V'} \{v_i\})$  fails at  $t = 1$ , and by **(III)** the nodes in  $\cup_{\substack{\{v_i, v_j\} \in E \\ i < j}} \{e_{i,j}\}$  fails  $t = 2$ . Thus, we obtain a solution of  $\vec{G}$  by shocking  $|V'| + n$  nodes.

Conversely, consider a solution of the  $\text{STABILITY}_{T, \Phi}$  problem on  $\vec{G}$ . Remember that all the super-nodes must be shocked, which ensures that we need to shock  $n + a$  nodes for some integer  $a \geq 0$ , and that any node  $v_i$  that is not shocked will fail at  $t = 2$ . By **(II)** it is of no use to shock the sink nodes. Thus, the shocked nodes consist of all super-nodes and a subset  $V'$  of cardinality  $a$  of the nodes  $u_1, u_2, \dots, u_n$ . By **(IV)** for every node  $e_{i,j}$  at least one of the nodes  $u_i$  or  $u_j$  must be in  $U$ . Thus, the set of nodes  $\{v_i | u_i \in U\}$  form a node cover of  $G$  of size  $a$ .

That the reduction is an L-reduction follows from the observation that any locally improvable solution of 3-MIN-NODE-COVER has between  $n/3$  and  $n$  nodes.  $\square$

### 8.2.2 Algorithmic Results for any $T > 1$

#### Restricted Acyclic Network: Polynomial-Time Solution for arbitrary $T$

Note that the APX-hardness result of Theorem 8.3 has constant values for both  $\Phi$  and  $\gamma$ , and requires  $\deg_{\text{out}}(v) = 2$  for some nodes  $v$ . It is natural to ask if the problem is computationally hard even if  $\deg_{\text{out}}(v) \leq 1$  for every node  $v$ . We show that with this and some mild additional restriction, the network may be highly stable (i.e.,  $\text{SI}^*(G, T)$  is large), and  $\text{SI}^*(G, T)$  can be computed in polynomial time for any  $T > 1$ . Recall that an in-arborescence rooted at node  $r$  ( $r$ -in-arborescence) is a digraph with  $\deg_{\text{out}}(r) = 0$ ,  $\deg_{\text{out}}(u) = 1$  for any other node  $u \neq r$ , and whose underlying undirected graph is a spanning tree.

**Theorem 8.4.** *If the banking network  $G$  is a rooted in-arborescence then  $\text{SI}^*(G, T) > \frac{\gamma}{\Phi \deg_{\text{in}}^{\max}}$ , where  $\deg_{\text{in}}^{\max} = \max_{v \in V} \{\deg_{\text{in}}(v)\}$  is the maximum in-degree over all nodes of  $G$ . Moreover, under the assumption that any individual node of the network can be failed by shocking,  $\text{SI}^*(G, T)$  can be computed exactly in  $O(n^2)$  time.*

**Remark 8.5.** *Thus, for example, when  $\deg_{\text{in}}^{\max} = 3$ ,  $\gamma = 0.1$  and  $\Phi = 0.15$ , we get  $\text{SI}^*(G, T) > 0.22$  and the network cannot be put to death without shocking more than 22% of the nodes. The proof gives an example for which the lower bound is tight.*

In the rest of this section, we prove the above theorem. Let  $G = (V, F)$  be the given in-arborescence rooted at node  $r$ . We will use the following notations and terminologies:

- $u \rightarrow v$  and  $u \rightsquigarrow v$  denote a directed edge and a directed path of one or more edges, respectively, from node  $u$  to node  $v$ .
- If  $(u, v) \in F$  then  $v$  is the *parent* of  $u$  and  $u$  is a *child* of  $v$ . Similarly, if  $u \rightsquigarrow v$  exists in  $G$  then  $v$  is an *ancestor* of  $u$  and  $u$  is a *descendent* of  $v$ .
- Let  $\nabla(u) = \{v \mid u \rightsquigarrow v \text{ exists in } G\}$  denote the set of all proper ancestors of  $u$ , and  $\Delta(u) = \{v \mid v \rightsquigarrow u \text{ exists in } G\} \cup \{u\}$  denote the set of all descendents of  $u$  (including the node  $u$  itself). Note that for the network  $G$  to fail, at least one node in  $\nabla(u) \cup \{u\}$  must be shocked for every node  $u$ .

Suppose that we shock a node  $u$  of  $G$  (and shock no other nodes in  $\Delta(u)$ ). If  $u$  fails, then the shock splits and propagates to a subset of nodes in  $\Delta(u)$  until each split part of the shock terminates because of one of the following reasons:

- the component of the shock reaches a “leaf” node  $v$  with  $\deg_{\text{in}}(v) = 0$ , or
- the component of the shock reaches a node  $v$  with a sufficiently high  $c_v$  such that  $v$  does not fail.

Based on the above observations, we define the following quantities.

**Definition 8.6** (see Fig. 5 for illustrations). *The influence zone of a shock on  $u$ , denoted by  $\text{iz}(u)$ , is the set of all failed nodes  $v \in \Delta(u)$  within time  $T$  when  $u$  is shocked (and, no other node in  $\Delta(u)$  is shocked). Note that  $u \in \text{iz}(u)$ .*

Note that, for any node  $u$ ,  $\text{iz}(u)$  can be computed in  $O(n)$  time.

**Lemma 8.7.** *For any node  $u$ ,  $|\text{iz}(u)| < 1 + \deg_{\text{in}}(u) \left( \frac{\Phi}{\gamma} - 1 \right)$ .*



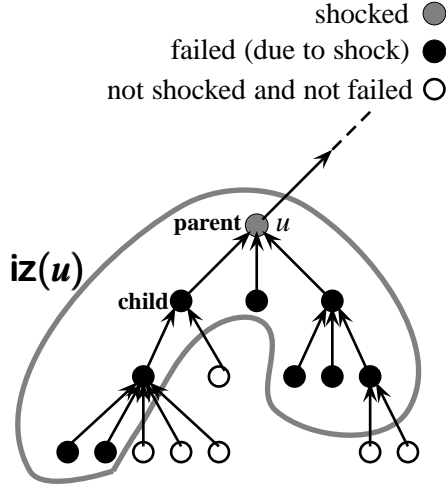


Figure 5: Influence zone of a shock on  $u$ .

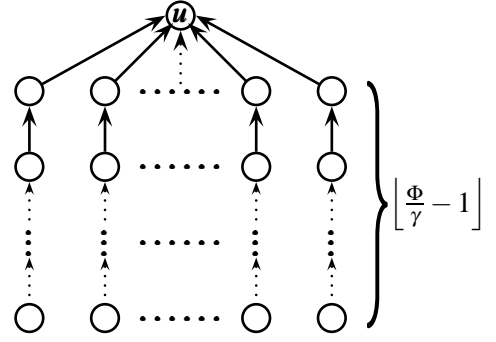


Figure 6: A tight example for the bound in Lemma 8.7 ( $\mathcal{E} = 0$ ).

*Proof.* For notational simplicity, let  $\mathcal{E} = E/n$ . If the node  $u$  does not fail when shocked, or  $u$  fails but it has no child, then  $|iz(u)| \leq 1$  and our claim holds since  $\Phi > \gamma$ . Otherwise,  $u$  fails and each of its  $\deg_{in}(u)$  children at level 2 receives a part of the shock given by

$$\begin{aligned} \mathfrak{D} &= \min \left\{ \frac{\Phi(\deg_{in}(u) - 1 + \mathcal{E}) - \gamma(\deg_{in}(u) + \mathcal{E})}{\deg_{in}(u)}, 1 \right\} \\ &< \Phi \left( 1 + \frac{\mathcal{E}}{\deg_{in}(u)} \right) - \gamma \left( 1 + \frac{\mathcal{E}}{\deg_{in}(u)} \right) \leq \Phi(1 + \mathcal{E}) - \gamma(1 + \mathcal{E}) \end{aligned}$$

Consider a child  $v$  of  $u$ . Each node  $v' \in \Delta(v)$  that fails due to the shock subtracts an amount of  $\gamma(\deg_{in}(v') + \mathcal{E}) \geq \gamma(1 + \mathcal{E})$  from  $\mathfrak{D}$  provided this subtraction does not result in a negative value. Thus, the total number of failed nodes is strictly less than  $1 + \deg_{in}(u) \frac{\Phi(1 + \mathcal{E}) - \gamma(1 + \mathcal{E})}{\gamma(1 + \mathcal{E})} = 1 + \deg_{in}(u) \left( \frac{\Phi}{\gamma} - 1 \right)$ .  $\square$

**Remark 8.8.** The bound in Lemma 8.7 is tight as shown in Fig. 6.

Lemma 8.7 immediately implies that  $SI^*(G, T) > \frac{n / \left( 1 + \deg_{in}^{\max} \left( \frac{\Phi}{\gamma} - 1 \right) \right)}{n} \geq \frac{\gamma}{\Phi \deg_{in}^{\max}}$ . We now provide a polynomial time algorithm to compute  $SI^*(G, T)$  exactly assuming each node can be shocked to fail individually. For a node  $u$ , define the following:

- For every node  $u' \in \nabla(u)$ ,  $SI_{SAS}^*(G, T, u, u')$  is the number of nodes in an optimal solution of  $STABILITY_{T, \Phi}$  for the subgraph induced by the nodes in  $\Delta(u)$  (or  $\infty$ , if there is no feasible solution of  $STABILITY_{T, \Phi}$  for this subgraph under the stated conditions) assuming the following:
  - $u'$  was shocked,
  - $u$  was *not* shocked, and
  - no node in the path  $u' \rightsquigarrow u$  excluding  $u'$  was shocked.
- $SI_{SAS}^*(G, T, u)$  is the number of nodes in an optimal solution of  $STABILITY_{T, \Phi}$  for the subgraph induced by the nodes in  $\Delta(u)$  (or  $\infty$ , if there is no feasible solution of  $STABILITY_{T, \Phi}$  under the stated conditions)<sup>2</sup> assuming that the node  $u$  was shocked (and therefore failed).

<sup>2</sup>Intuitively, a value of  $\infty$  signifies that the corresponding quantity is undefined.

We consider the usual partition of the nodes of  $G$  into *levels*:  $\text{level}(r) = 1$  and  $\text{level}(u) = \text{level}(v) + 1$  if  $u$  is a child of  $v$ . We will compute  $\text{Sl}_{\text{SAS}}^*(G, T, u)$  and  $\text{Sl}_{\text{SANS}}^*(G, T, u, v)$  for the nodes  $u$  level by level, starting with the highest level and proceeding to successive lower levels. By Observation 7.1(a), the root  $r$  must be shocked to fail for the entire network to fail, and thus  $\text{Sl}_{\text{SAS}}^*(G, T, r)$  will provide us with our required optimal solution.

Every node  $u$  at the highest level has  $\deg_{\text{in}}(u) = 0$ . In general,  $\text{Sl}_{\text{SAS}}^*(G, T, u)$  and  $\text{Sl}_{\text{SANS}}^*(G, T, u, u')$  can be computed for any node  $u$  with  $\deg_{\text{in}}(u) = 0$  as follows:

**Computing  $\text{Sl}_{\text{SAS}}^*(G, T, u)$  when  $\deg_{\text{in}}(u) = 0$ :**  $\text{Sl}_{\text{SAS}}^*(G, T, u) = 1$  by our assumption that every node can be shocked to fail.

**Computing  $\text{Sl}_{\text{SANS}}^*(G, T, u, u')$  when  $\deg_{\text{in}}(u) = 0$ :**

- If  $u \in \text{iz}(u')$  then shocking node  $v$  makes node  $u$  fail. Since node  $u$  fails without being shocked, we have  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = 0$ .
- Otherwise, node  $u$  does not fail. Thus, there is no feasible solution and  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = \infty$ .

Note that we only count the number of nodes in  $\Delta(u)$  in the calculations of  $\text{Sl}_{\text{SANS}}^*(G, T, u, u')$  and  $\text{Sl}_{\text{SAS}}^*(G, T, u)$ .

Now, consider a node  $u$  at some level  $\ell$  with  $\deg_{\text{in}}(u) > 0$ . Let  $v_1, v_2, \dots, v_{\deg_{\text{in}}(u)}$  be the children of  $u$  at level  $\ell + 1$ . Note that  $\nabla(v_1) = \nabla(v_2) = \dots = \nabla(v_{\deg_{\text{in}}(u)})$ .

**Computing  $\text{Sl}_{\text{SAS}}^*(G, T, u)$  when  $\deg_{\text{in}}(u) > 0$ :** By our assumption,  $u$  fails when shocked. Note that no node in  $\Delta(u) \setminus \{u\}$  can receive any component of a shock given to a node in  $V \setminus \Delta(u)$  since  $u$  failed. For each child  $v_i$  of  $u$  we have two choices:  $v_i$  is shocked and (and, therefore, fails), or  $v_i$  is not shocked. Thus, in this case we have  $\text{Sl}_{\text{SAS}}^*(G, T, u) = 1 + \sum_{i=1}^{\deg_{\text{in}}(u)} \min \left\{ \text{Sl}_{\text{SAS}}^*(G, T, v_i), \text{Sl}_{\text{SANS}}^*(G, T, v_i, u) \right\}$ .

**Computing  $\text{Sl}_{\text{SANS}}^*(G, T, u, u')$  when  $\deg_{\text{in}}(u) > 0$ :** Since  $u'$  is shocked and  $u$  is not shocked, the following cases arise:

- If  $u \notin \text{iz}(u')$  then  $u$  does not fail. Thus, there is no feasible solution for the subgraph induced by the nodes in  $\Delta(u)$  under this condition, and  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = \infty$ .
- Otherwise,  $u \in \text{iz}(u')$ , and therefore  $u$  fails when  $u'$  is shocked. For each child  $v_i$  of  $u$ , there are two options:  $v_i$  is shocked and fails, or  $v_i$  is not shocked. Thus, in this case we have  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = \sum_{i=1}^{\deg_{\text{in}}(u)} \min \left\{ \text{Sl}_{\text{SAS}}^*(G, T, v_i), \text{Sl}_{\text{SANS}}^*(G, T, v_i, u') \right\}$ .

Let  $\ell_{\text{max}}$  be the maximum level number of any node in  $G$ . Based on the above observations, we can design the dynamic programming algorithm as shown in Fig. 7 to compute an optimal solution of  $\text{STABILITY}_{T, \Phi}$  on  $G$ . It is easy to check that the running time of our algorithm is  $O(n^2)$ .

## 9 Our Results on the Stability Index for Heterogeneous Networks

### 9.1 Logarithmic Inapproximability for any $T > 1$

**Theorem 9.1.** *Assuming  $\text{NP} \not\subseteq \text{DTIME}(n^{\log \log n})$ , for any constant  $0 < \varepsilon < 1$  and any  $T \geq 2$ , it is impossible to approximate  $\text{STABILITY}_{T, \Phi}$  within a factor of  $(1 - \varepsilon) \ln n$  in polynomial time even if  $G$  is a DAG.*

*Proof.* The (unweighted) SET-COVER problem is defined as follows. We have an universe  $\mathcal{U}$  of  $n$  elements, a collection of  $m$  sets  $\mathcal{S}$  over  $\mathcal{U}$ . The goal is to pick a sub-collection  $\mathcal{S}' \subseteq \mathcal{S}$  containing a *minimum* number of sets such that these sets “cover”  $\mathcal{U}$ , i.e.,  $\cup_{S \in \mathcal{S}'} S = \mathcal{U}$ . It is known that there exists instances of

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(* preprocessing *)
 $\forall u \in V$ : compute  $\text{iz}(u)$ 
(* dynamic programming *)
for  $\ell = \ell_{\max}, \ell_{\max}-1, \dots, 1$  do
  for each node  $u$  at level  $\ell$  do
    if  $\deg_{\text{in}}(u) = 0$  then
       $\text{Sl}_{\text{SAS}}^*(G, T, u) = 1$ 
       $\forall u' \in \nabla(u)$ : if  $u \in \text{iz}(u')$  then  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = 0$  else  $\text{Sl}_{\text{SANS}}^a(G, T, u, u') = \infty$ 
    else (*  $\deg_{\text{in}}(u) > 0$  *)
       $\text{Sl}_{\text{SAS}}^*(G, T, u) = 1 + \sum_{i=1}^{\deg_{\text{in}}(u)} \min \{ \text{Sl}_{\text{SAS}}^*(G, T, v_i), \text{Sl}_{\text{SANS}}^*(G, T, v_i, u) \}$ 
       $\forall u' \in \nabla(u)$ : if  $u \notin \text{iz}(u')$  then  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = \infty$ 
      else  $\text{Sl}_{\text{SANS}}^*(G, T, u, u') = \sum_{i=1}^{\deg_{\text{in}}(u)} \min \{ \text{Sl}_{\text{SAS}}^*(G, T, v_i), \text{Sl}_{\text{SANS}}^*(G, T, v_i, u') \}$ 
    endif
  endif
endfor
return  $\text{Sl}_{\text{SAS}}^*(G, T, r)$  as the solution

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Figure 7: A polynomial time algorithm to compute  $\text{Sl}^*(G, T)$  when  $G$  is a rooted in-arborescence and each node of  $G$  fails individually when shocked.

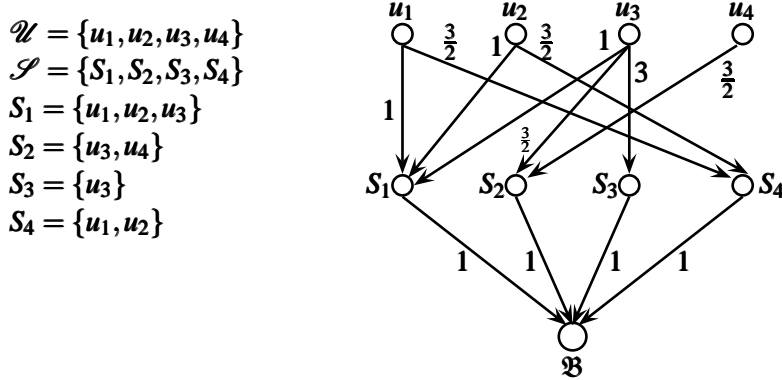


Figure 8: An instance  $\langle \mathcal{U}, \mathcal{S} \rangle$  of SET-COVER and its corresponding banking network  $G = (V, F)$ .

SET-COVER that cannot be approximated within a factor of  $(1 - \delta) \ln n$ , for any constant  $0 < \delta < 1$ , unless  $\text{NP} \subseteq \text{DTIME}(n^{\log \log n})$  [22]. Without any loss of generality, one may assume that every element  $u \in \mathcal{U}$  belongs to at least two sets in  $\mathcal{S}$  since otherwise the only set containing  $u$  must be selected in any solution.

Given such an instance  $\langle \mathcal{U}, \mathcal{S} \rangle$  of SET-COVER, we now construct an instance of the banking network  $G = (V, F)$  as follows:

- We have a special node  $\mathcal{B}$ .
- For every set  $S \in \mathcal{S}$ , we have a node  $S$ , and a directed edge  $(S, \mathcal{B})$ .
- For every element  $u \in \mathcal{U}$ , we have a node  $u$ , and directed edges  $(u, S)$  for every set  $S$  that contains  $u$ .

Thus,  $|V| = n + m + 1$ , and  $|F| < nm + m$ . See Fig. 8 for an illustration. We set the shares of internal assets for each bank as follows:

- For each set  $S \in \mathcal{S}$ , if  $S$  contains  $k > 1$  elements then, for each element  $u \in S$ , we set the weight of the edge  $e = (u, S)$  as  $w(e) = \frac{3}{k}$ .

- For each set  $S \in \mathcal{S}$ , we set the weight of the edge  $(S, \mathfrak{B})$  as 1.

Thus,  $I = 4m$ . Also, observe that:

- For any  $S \in \mathcal{S}$ ,  $b_S = 3$ , and  $\iota_S = 1$ .
- For any  $u \in \mathcal{U}$ ,  $b_u = 0$ . Also, since  $u$  belongs to at least two sets in  $\mathcal{S}$  and any set has at most  $n - 1$  elements,  $\frac{2}{n} \leq \iota_u < \frac{3n}{2}$ .
- $b_{\mathfrak{B}} = m$  and  $\iota_{\mathfrak{B}} = 0$ .
- Since  $\deg_{\text{in}}(u) = 0$  for any element  $u \in \mathcal{U}$ , if a node  $u$  is shocked, no part of the shock is propagated to any other node in the network.
- Since the longest path in  $G$  has 2 edges, by Proposition 7.1(b) no new node in  $G$  fails for  $T > 3$ .

Let the share of external assets for a node (bank)  $y$  be denoted by  $E_y$  (thus,  $\sum_{y \in V} E_y = E$ ). We will select the remaining network parameters, namely  $\gamma$ ,  $\Phi$  and the  $E_y$  values, based on the following properties.

(I) If the node  $\mathfrak{B}$  is shocked at  $t = 1$ , it fails:

$$\Phi(b_{\mathfrak{B}} - \iota_{\mathfrak{B}} + E_{\mathfrak{B}}) > \gamma(b_{\mathfrak{B}} + E_{\mathfrak{B}}) \equiv \Phi(m + E_{\mathfrak{B}}) > \gamma(m + E_{\mathfrak{B}}) \equiv \Phi > \gamma \quad (10)$$

(II) For any  $S \in \mathcal{S}$ , if node  $S$  is shocked at  $t = 1$ , then  $S$  fails at  $t = 1$ , and, for every  $u \in S$ , node  $u$  fails at time  $t = 2$ :

$$\begin{aligned} & \frac{\min \{ \Phi(b_S - \iota_S + E_S) - \gamma(b_S + E_S), b_S \}}{\deg_{\text{in}}(S)} > \gamma(b_u + E_u) \\ & \equiv \frac{\min \{ \Phi(2 + E_S) - \gamma(3 + E_S), 3 \}}{|S|} > \gamma E_u \end{aligned}$$

The above inequality is satisfied if:

$$\Phi(2 + E_S) > \gamma(3 + E_S + |S|E_u) \quad (11)$$

$$\Phi(2 + E_S) - \gamma(3 + E_S) \leq 3 \quad (12)$$

(III) For any  $u \in \mathcal{U}$ , consider the node  $u$ , and let  $S_1, S_2, \dots, S_p \in \mathcal{S}$  be the  $p$  sets that contain  $u$ . Then, we require that if the node  $\mathfrak{B}$  is shocked at  $t = 1$  then  $\mathfrak{B}$  fails at  $t = 1$ , every node among the set of nodes  $\{S_1, S_2, \dots, S_p\}$  that was not shocked at  $t = 1$  fails at  $t = 2$ , but the node  $u$  does not fail if the none of the nodes  $u, S_1, S_2, \dots, S_p$  were shocked, This is satisfied provided the following inequalities hold:

(III-1) Any node among the set of nodes  $\{S_1, S_2, \dots, S_p\}$  that was not shocked at  $t = 1$  fails at  $t = 2$ . This is satisfies provided for any set  $S \in \mathcal{S}$  the following holds:

$$\frac{\min \{ \Phi(b_{\mathfrak{B}} - \iota_{\mathfrak{B}} + E_{\mathfrak{B}}) - \gamma(b_{\mathfrak{B}} + E_{\mathfrak{B}}), b_{\mathfrak{B}} \}}{\deg_{\text{in}}(\mathfrak{B})} > \gamma(b_S + E_S) \equiv \min \left\{ (\Phi - \gamma) \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right), 1 \right\} > \gamma(3 + E_S)$$

The above inequality is satisfied provided:

$$(\Phi - \gamma) \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right) > \gamma(3 + E_S) \equiv \Phi \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right) > \gamma \left( 4 + E_S + \frac{E_{\mathfrak{B}}}{m} \right) \quad (13)$$

$$1 > \gamma(3 + E_S) \equiv \gamma < \frac{1}{3 + E_S} \quad (14)$$

(III-2)  $u$  does not fail if the none of the nodes  $u, S_1, S_2, \dots, S_p$  were shocked:

$$\min \left\{ (\Phi - \gamma) \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right), 1 \right\} - \gamma(3 + E_S) \leq \frac{\gamma E_u}{n} \equiv \min \left\{ (\Phi - \gamma) \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right), 1 \right\} \leq \gamma \left( 3 + E_S + \frac{E_u}{n} \right)$$

The above inequality is satisfied provided:

$$(\Phi - \gamma) \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right) \leq \gamma \left( 3 + E_S + \frac{E_u}{n} \right) \equiv \Phi \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right) \leq \gamma \left( 4 + E_S + \frac{E_{\mathfrak{B}}}{m} + \frac{E_u}{n} \right) \quad (15)$$

$$(\Phi - \gamma) \left( 1 + \frac{E_{\mathfrak{B}}}{m} \right) \leq 1 \equiv \gamma \geq \Phi - \frac{1}{1 + \frac{E_{\mathfrak{B}}}{m}} \quad (16)$$

There are many choices of parameters  $\gamma, \Phi$  and  $E_y$ 's satisfying Equations (10)–(16); we exhibit just one:

$$\forall S \in \mathcal{S}: E_S = 0 \quad E_{\mathfrak{B}} = 0 \quad \forall u \in \mathcal{U}: E_u = \frac{1}{100n} \quad \gamma = 0.1 \quad \Phi = 0.4 + \frac{1}{n^{10000}}$$

Suppose that  $\mathcal{S}' \subset \mathcal{S}$  is a solution of SET-COVER. Then, we shock the node  $\mathfrak{B}$  and the nodes  $S$  for each  $S \in \mathcal{S}'$ . By (I) and (II) the node  $\mathfrak{B}$  and the nodes  $S$  for each  $S \in \mathcal{S}'$  fails at  $t = 1$ , and by (III) the nodes  $u$  for every  $u \in \mathcal{U}$  fails  $t = 2$ . Thus, we obtain a solution of  $G$  by shocking  $|\mathcal{S}'| + 1$  nodes.

Conversely, consider a solution of the STABILITY $_{T,\Phi}$  problem on  $G$ . If a node  $u$  for some  $u \in \mathcal{U}$  was shocked, we can instead shock the node  $S$  for any set  $S$  that contains  $u$ , which by (II) still fails all the nodes in the network and does not increase the number of shocked nodes. Thus, after such normalizations, we may assume that the shocked nodes consist of  $\mathfrak{B}$  and a subset  $\mathcal{S}' \subseteq \mathcal{S}$  of nodes. By (II) and (III) for every node  $u \in \mathcal{U}$  at least one set that contains  $u$  must be in  $\mathcal{S}'$ . Thus, the collection of sets in  $\mathcal{S}'$  form a cover of  $\mathcal{U}$  of size  $|\mathcal{S}'|$ .  $\square$

## 9.2 Logarithmic Approximation for $T = 2$

For any positive number  $x > 0$ , let  $\bar{x} = \max \{x, 1/x\}$  and  $\underline{x} = \min \{x, 1/x\}$ . Let  $w_{\min} = \min_{e: w(e) > 0} \{w(e)\}$ ,  $w_{\max} = \max_e \{w(e)\}$ ,  $\alpha_{\min} = \min_{v: \alpha_v > 0} \{\alpha_v\}$ ,  $\alpha_{\max} = \max_v \{\alpha_v\}$  and  $s = \min_{\substack{E_1, E_2 \subseteq E \\ \sum_{e \in E_1} w(e) \neq \sum_{e \in E_2} w(e)}} \left\{ \left| \sum_{e \in E_1} w(e) - \sum_{e \in E_2} w(e) \right| \right\}$ . The following theorem provides a polynomial-time algorithm with a logarithmic approximation ratio when  $T = 2$ .

**Theorem 9.2.** *There is a polynomial-time greedy algorithm for  $T = 2$  that has an approximation ratio of  $O\left(\log n + \log(1/\Phi) + \log(1/\gamma) + \log(1/(\Phi - \gamma)) + \log \bar{E} + \log(1/\underline{E}) + \log \bar{w}_{\max} + \log \bar{w}_{\min} + \log \bar{\alpha}_{\max} + \log(1/\underline{w}_{\min}) + \log(1/\underline{\alpha}_{\min}) + \log(1/\underline{w}_{\max})\right)$ .*

*Proof.* We can reuse the proof of the corresponding approximation for homogeneous networks in Theorem 8.2 to obtain an approximation ratio of  $2 + \ln n + \ln \left( \max_{v \in V} \left\{ \sum_{u \in V} \frac{\delta_{v,u}}{\zeta} \right\} \right)$ , where  $\zeta = \min_{u \in V} \left\{ \min_{v \in V} \{\delta_{u,v}\}, c_u \right\}$ ,

provided we recalculate  $\max_{v \in V} \left\{ \sum_{u \in V} \frac{\delta_{vu}}{\zeta} \right\}$ . Then,

$$\begin{aligned}
\min_{\substack{u \in V \\ \delta_{u,u} > 0}} \{ \delta_{u,u} \} &= \min_{\substack{u \in V \\ \delta_{u,u} > 0}} \left\{ \Phi \left( \sum_{e=(v',u) \in F} w(e) - \sum_{e=(u,v') \in F} w(e) + \alpha_v E \right) \right\} = \Omega \left( \text{poly} \left( s, \Phi, \underline{E}, \underline{\alpha_{\min}} \right) \right) \\
\min_{u \in V} \min_{\substack{v \in V \\ \delta_{u,v} > 0}} \{ \delta_{u,v} \} &= \min_{u \in V} \min_{\substack{v \in V \\ \Phi_{e_v > c_v}}} \left\{ (\Phi - \gamma) \left( 1 + \frac{\alpha_v E}{\sum_{e=(v',v) \in F} w(e)} \right) - \Phi \frac{\sum_{e=(v,v') \in F} w(e)}{\sum_{e=(v',v) \in F} w(e)} \right\} = \Omega \left( \text{poly} \left( n^{-1}, \Phi - \gamma, \Phi, \underline{E}, \underline{w_{\max}}, \underline{w_{\min}}, \underline{\alpha_{\min}} \right) \right) \\
\min_{u \in V} \{ c_u \} &= \min_{u \in V} \left\{ \gamma \left( \sum_{e=(v',u) \in F} w(e) + \alpha_u E \right) \right\} = \Omega \left( \text{poly} \left( n^{-1}, \gamma, \underline{E}, \underline{\alpha_{\min}}, \underline{w_{\min}} \right) \right) \\
\zeta &= \min \left\{ \min_{u \in V} \min_{v \in V} \{ \delta_{u,v} \}, \min_{u \in V} \{ c_u \} \right\} = \Omega \left( \text{poly} \left( n^{-1}, \Phi - \gamma, \Phi, \gamma, \underline{E}, \underline{w_{\min}}, \underline{\alpha_{\min}}, \underline{w_{\max}} \right) \right) \\
\max_{v \in V} \sum_{u \in V} \delta_{v,u} &\leq n \max_{u \in V} \left\{ (\Phi - \gamma) \left( 1 + \frac{\alpha_v E}{\sum_{e=(v',v) \in F} w(e)} \right) - \Phi \frac{\sum_{e=(v,v') \in F} w(e)}{\sum_{e=(v',v) \in F} w(e)} \right\} = O \left( \text{poly} \left( n, \overline{E}, \overline{w_{\max}}, \overline{w_{\min}}, \overline{\alpha_{\max}} \right) \right)
\end{aligned}$$

and thus,  $\max_{v \in V} \left\{ \sum_{u \in V} \frac{\delta_{vu}}{\zeta} \right\} = O \left( \text{poly} \left( n, \Phi^{-1}, \gamma^{-1}, (\Phi - \gamma)^{-1}, \overline{E}, \underline{E}^{-1}, \overline{w_{\max}}, \overline{w_{\min}}, \overline{\alpha_{\max}}, \underline{w_{\min}}^{-1}, \underline{\alpha_{\min}}^{-1}, \underline{w_{\max}}^{-1} \right) \right)$ , giving the desired approximation bound.  $\square$

### 9.3 Stronger Poly-logarithmic Inapproximability for any $T > 3$

In this section, we prove a stronger lower bound result for  $\text{STABILITY}_{T,\Phi}$  for heterogeneous networks provided we are interested in the stability of the network in four or more time steps.

**Theorem 9.3.** *Assuming  $\text{NP} \not\subseteq \text{DTIME} \left( n^{\text{poly}(\log n)} \right)$ , for any constant  $0 < \varepsilon < 1$  and any  $T > 3$ , it is impossible to approximate  $\text{STABILITY}_{T,\Phi}$  within a factor of  $2^{\log^{1-\varepsilon} n}$  in polynomial time even if  $G$  is a DAG.*

*Proof.* The MINREP problem (with minor modifications from the original setup) is defined as follows. We are given a bipartite graph  $G = (V^{\text{left}}, V^{\text{right}}, F)$  such that the degree of every node of  $G$  is at least 10, a partition of  $V^{\text{left}}$  into  $\frac{|V^{\text{left}}|}{\alpha}$  equal-size subsets  $V_1^{\text{left}}, V_2^{\text{left}}, \dots, V_\alpha^{\text{left}}$ , and a partition of  $V^{\text{right}}$  into  $\frac{|V^{\text{right}}|}{\beta}$  equal-size subsets  $V_1^{\text{right}}, V_2^{\text{right}}, \dots, V_\beta^{\text{right}}$ .

These partitions define a natural “bipartite super-graph”  $G_{\text{super}} = (V_{\text{super}}, F_{\text{super}})$  in the following manner.  $G_{\text{super}}$  has a “super-node” for every  $V_i^{\text{left}}$  (for  $i = 1, 2, \dots, \alpha$ ) and for every  $V_j^{\text{right}}$  (for  $j = 1, 2, \dots, \beta$ ). There exists an “super-edge”  $h_{i,j}$  between the super-node for  $V_i^{\text{left}}$  and the super-node for  $V_j^{\text{right}}$  if and only if there exists  $u \in V_i^{\text{left}}$  and  $v \in V_j^{\text{right}}$  such that  $\{u, v\}$  is an edge of  $G$ . A pair of nodes  $u$  and  $v$  of  $G$  “witnesses” the super-edge  $h_{i,j}$  of  $H$  provided  $u$  is in  $V_i^{\text{left}}$ ,  $v$  is in  $V_j^{\text{right}}$  and the edge  $\{u, v\}$  exists in  $G$ , and a set of nodes  $V' \subseteq V$  of  $G$  witnesses a super-edge if and only if there exists at least one pair of nodes in  $S$  that witnesses the super-edge.

The goal of MINREP is to find  $V_1 \subseteq V^{\text{left}}$  and  $V_2 \subseteq V^{\text{right}}$  such that  $V_1 \cup V_2$  witnesses every super-edge of  $H$  and the size of the solution, namely  $|V_1| + |V_2|$ , is minimum. For notational simplicity, let  $n = |V^{\text{left}}| + |V^{\text{right}}|$ . The following result is a consequence of Raz’s parallel repetition theorem [33, 39].

**Theorem 9.4.** [33] *Let  $L$  be any language in NP and  $0 < \delta < 1$  be any constant. Then, there exists a reduction running in  $n^{\text{poly}(\log n)}$  time that, given an input instance  $x$  of  $L$ , produces an instance of MINREP such that:*

- if  $x \in L$  then MINREP has a solution of size  $\alpha + \beta$ ;
- if  $x \notin L$  then MINREP has a solution of size at least  $(\alpha + \beta) \cdot 2^{\log^{1-\delta} n}$ .

Thus, the above theorem provides a  $2^{\log^{1-\delta} n}$ -inapproximability for MINREP under the complexity-theoretic assumption of  $\text{NP} \not\subseteq \text{DTIME}(n^{\text{polylog}(n)})$ .

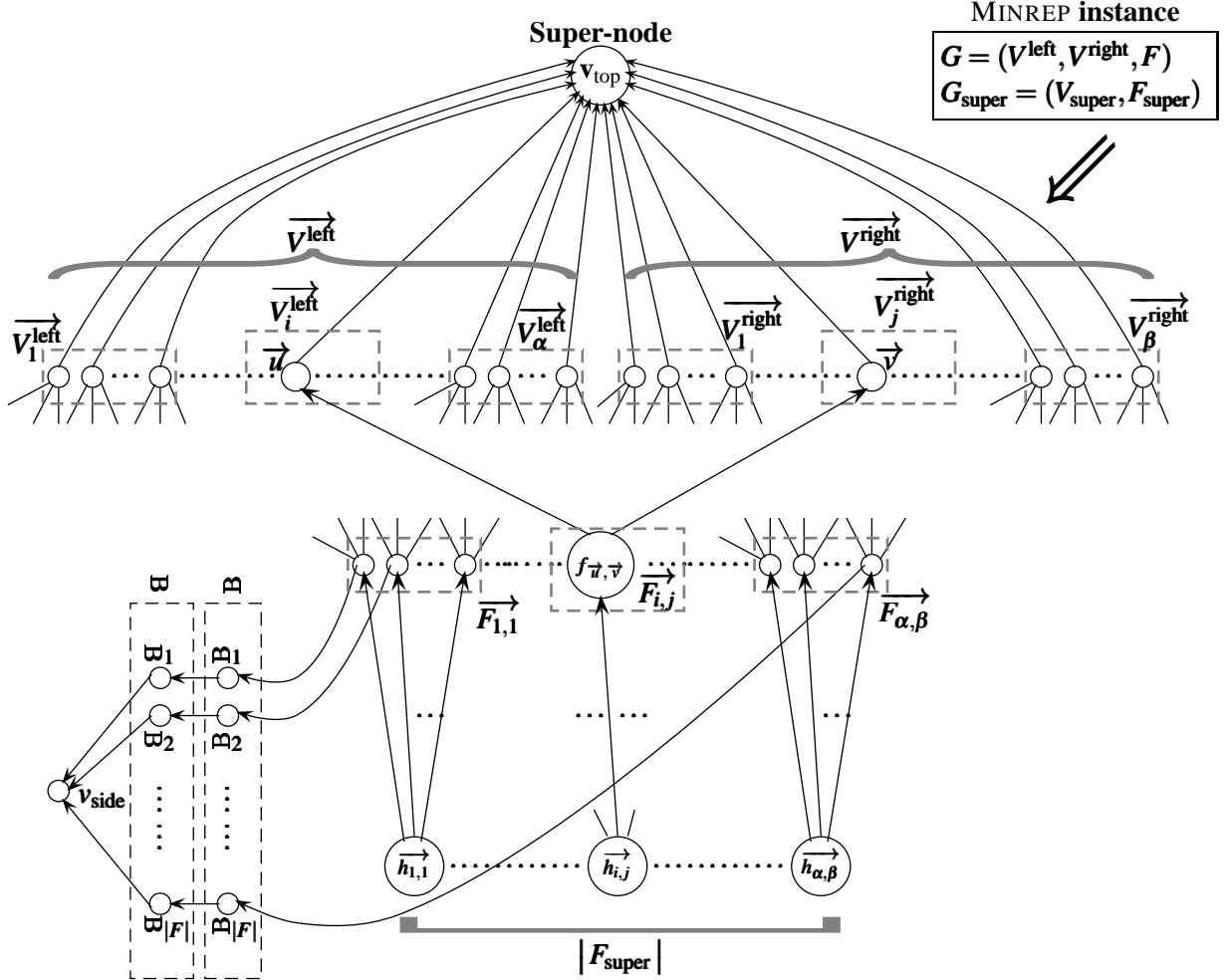


Figure 9: Reduction of an instance of MINREP to  $\text{STABILITY}_{T,\Phi}$  for heterogeneous networks.

Let  $F_{i,j} = \left\{ \{u, v\} \mid u \in V_i^{\text{left}}, v \in V_j^{\text{right}}, \{u, v\} \in F \right\}$ . We now show our construction of an instance of  $\text{STABILITY}_{T,\Phi}$  from an instance of MINREP. Our directed graph  $\vec{G} = (\vec{V}, \vec{F})$  for  $\text{STABILITY}_{T,\Phi}$  is constructed as follows (see Fig. 9 for an illustration):

**Nodes:**

- For every node  $u \in V_i^{\text{left}}$  of  $G$  we have a corresponding node  $\vec{u}$  in the set of nodes  $\vec{V}_i^{\text{left}}$  in  $\vec{G}$ , and for every node  $v \in V_j^{\text{right}}$  of  $G$  we have a corresponding node  $\vec{v}$  in the set of nodes  $\vec{V}_j^{\text{right}}$  in  $\vec{G}$ . The total number of such nodes is  $n$ .
- For every edge  $\{u, v\}$  of  $G$  with  $u \in V_i^{\text{left}}$  and  $v \in V_j^{\text{right}}$ , we have a corresponding node  $f_{\vec{u}, \vec{v}}$  in the set of nodes  $\vec{F}_{i,j}$  in  $\vec{G}$ . There are  $|F|$  such nodes.

- For every super-edge  $h_{i,j}$  of  $G_{\text{super}}$ , we have a node  $\overrightarrow{h_{i,j}}$  in  $\overrightarrow{G}$ . There are  $|F_{\text{super}}|$  such nodes.
- We have one “top super-node”  $v_{\text{top}}$ , one “side super-node”  $v_{\text{side}}$ , and  $2|F|$  additional nodes  $\varpi_1, \varpi_2, \dots, \varpi_{|F|}$ ,  $\mathfrak{m}_1, \mathfrak{m}_2, \dots, \mathfrak{m}_{|F|}$ . Let  $\varpi = \bigcup_{j=1}^{|F|} \varpi_j$  and  $\mathfrak{m} = \bigcup_{j=1}^{|F|} \mathfrak{m}_j$ .

Thus,  $n + 3|F| + 2 < |\overrightarrow{V}| = n + |F| + |F_{\text{super}}| + 2 + 2|F| < n + 4|F| + 2$ .

Edges:

- For every node  $u$  of  $G$ , we have an edge  $(u, v_{\text{top}})$  in  $\overrightarrow{G}$ . There are  $n$  such edges.
- For every edge  $\{u, v\}$  of  $G$ , we have two edges  $(f_{\overrightarrow{u}, \overrightarrow{v}}, \overrightarrow{u})$  and  $(f_{\overrightarrow{u}, \overrightarrow{v}}, \overrightarrow{v})$  in  $\overrightarrow{G}$ . There are  $2|F|$  such edges.
- For every super-edge  $h_{i,j}$  of  $G_{\text{super}}$  and for every edge  $f_{u,v}$  in  $F_{i,j}$ , we have an edge  $(\overrightarrow{h_{i,j}}, f_{\overrightarrow{u}, \overrightarrow{v}})$  in  $\overrightarrow{G}$ . There are  $|F|$  such edges.
- Let  $p_1, p_2, \dots, p_{|F|}$  be any arbitrary ordering of the edges in  $F$ . Then, for every  $j = 1, 2, \dots, |F|$ , we have the edges  $(v_{\text{side}}, \varpi_j)$ ,  $(\varpi_j, \mathfrak{m}_j)$  and  $(\mathfrak{m}_j, p_j)$ . The total number of such edges is  $3|F|$ .

Thus,  $|\overrightarrow{E}| = n + 6|F|$ .

Distribution of internal assets: We set the weight of every edge to 1. Thus,  $I = n + \sum_{u \in V^{\text{left}} \cup V^{\text{right}}} \deg(u) + 4|F| = n + 6|F|$ .

Let  $\deg(u) \geq 10$  be the degree of node  $u \in V^{\text{left}} \cup V^{\text{right}}$ . Observe that:

- $b_{v_{\text{top}}} = n$ , and  $\iota_{v_{\text{top}}} = 0$ . Since  $\deg_{\text{out}}(v_{\text{top}}) = 0$ , by Proposition 7.1(a) the node  $v_{\text{top}}$  must be shocked to make the network fail.
- $b_{v_{\text{side}}} = |F|$ , and  $\iota_{v_{\text{side}}} = 0$ . Since  $\deg_{\text{out}}(v_{\text{side}}) = 0$ , by Proposition 7.1(a) the node  $v_{\text{side}}$  must be shocked to make the network fail.
- For any  $u \in V^{\text{left}} \cup V^{\text{right}}$ ,  $b_{\overrightarrow{u}} = \deg(u)$  and  $\iota_{\overrightarrow{u}} = 1$ .
- For any node  $f_{\overrightarrow{u}, \overrightarrow{v}}$ ,  $b_{f_{\overrightarrow{u}, \overrightarrow{v}}}} = 1$  and  $\iota_{f_{\overrightarrow{u}, \overrightarrow{v}}}} = 2$ .
- For every node  $\overrightarrow{h_{i,j}}$ ,  $b_{\overrightarrow{h_{i,j}}} = 0$  and  $\iota_{\overrightarrow{h_{i,j}}} = |F_{i,j}|$ . Since  $\deg_{\text{in}}(\overrightarrow{h_{i,j}}) = 0$  for any node  $\overrightarrow{h_{i,j}}$ , if such a node is shocked, no part of the shock is propagated to any other node in the network.
- For every  $j$ ,  $b_{\mathfrak{m}_j} = \iota_{\mathfrak{m}_j} = b_{\varpi_j} = \iota_{\varpi_j} = 1$ .
- Since the longest directed path in  $G$  has 4 edges, by Proposition 7.1(b) no new node in  $G$  fails for  $t > 4$ .

Let the share of external assets for a node (bank)  $y$  be denoted by  $E_y$  (thus,  $\sum_{y \in V} E_y = E$ ). We will select the remaining network parameters, namely  $\gamma$ ,  $\Phi$  and the set of  $E_y$  values, based on the following desirable properties and events. For the convenience of the readers, all the relevant constraints are also summarized in Table 3. Assume that no nodes in  $(\bigcup_{i,j} \overrightarrow{F_{i,j}}) \cup (\bigcup_{i,j} \{\overrightarrow{h_{i,j}}\})$  were shocked at  $t = 1$ .

(I) Suppose that the node  $v_{\text{top}}$  is shocked at  $t = 1$ . Then, the following happens.



$\Phi > \gamma \quad (17)$	$\left  \Phi > \gamma \left( 1 + \frac{\deg(u) + E_{\vec{u}}}{1 + \frac{E_{v_{\text{top}}}}{n}} \right) \quad (18) \right $	$\Phi \leq \gamma + \frac{1}{1 + \frac{E_{v_{\text{top}}}}{n}} \quad (19)$	$\left  \Phi > \gamma \left( 1 + \frac{1}{\deg(u) - 1 + E_{\vec{u}}} \right) \quad (20) \right $
$\frac{\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}})}{\deg(u)} + \frac{\Phi(\deg(v) - 1 + E_{\vec{v}}) - \gamma(\deg(v) + E_{\vec{v}})}{\deg(v)} > \gamma \left( 1 + E_{f_{\vec{u}, \vec{v}}} \right) \quad (21)$			
$\Phi \leq \gamma \left( \frac{\deg(u) + E_{\vec{u}}}{\deg(u) - 1 + E_{\vec{u}}} \right) + \frac{\deg(u)}{\deg(u) - 1 + E_{\vec{u}}} \quad (22)$			
$\frac{\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}})}{\deg(u)} + \frac{\Phi(\deg(v) - 1 + E_{\vec{v}}) - \gamma(\deg(v) + E_{\vec{v}})}{\deg(v)} - \gamma \left( 1 + E_{f_{\vec{u}, \vec{v}}} \right) > \gamma E_{h_{i,j}}^{\rightarrow} \quad (23)$			
$\gamma E_{h_{i,j}}^{\rightarrow} < 1 \quad (24)$	$\left  \Phi \leq \gamma \left( 1 + \frac{1}{E_{\varpi_j}} \right) \quad (25) \right $	$\Phi > \frac{\gamma \left( 2 + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} \right)}{\left( 1 + \frac{E_{v_{\text{side}}}}{ F } \right)} \quad (26)$	
$\Phi > \gamma \left( \frac{( F  + E_{v_{\text{side}}})}{3 F  +  F E_{\varpi_j} +  F E_{\varpi_j} + E_{v_{\text{side}}}} \right) \quad (29)$		$\left  \Phi \leq \gamma + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{ F }} \quad (27) \right $	$\Phi \leq \gamma \left( 1 + \frac{1}{E_{\varpi_j}} \right) \quad (28)$
$\Phi \leq \gamma \left( 1 + \frac{1 + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{ F }} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{ F }} \quad (30)$		$\Phi \leq \gamma \left( \frac{3 + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{ F }} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{ F }} \quad (31)$	
$\Phi > \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{ F } + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)}} \right) \quad (32)$			
$\Phi \leq \gamma \left( \frac{1 + \frac{E_{v_{\text{top}}}}{n} + \deg(u) + E_{\vec{u}}}{1 + \frac{E_{v_{\text{top}}}}{n}} \right) + \frac{\deg(u)}{1 + \frac{E_{v_{\text{top}}}}{n}} \quad (33)$		$\Phi > \gamma \left( \frac{6 + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{\varpi_j} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}} + 1}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}}}{2 + \frac{E_{v_{\text{side}}}}{ F } + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{v}} - 1}{\deg(v)}} \right) \quad (34)$	
$\Phi \leq \gamma \left( \frac{3 + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \quad (35)$		$\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{\varpi_j}}{1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \quad (36)$	
$\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{\varpi_j} + \frac{E_{h_{i,j}}^{\rightarrow}}{ F_{i,j} }}{1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \right) \quad (37)$			
$\Phi \leq \gamma \left( \frac{2 + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j}} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j}} \quad (38)$		$\Phi \leq \gamma \left( \frac{2 + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + 1 + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \quad (39)$	
$\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{f_{\vec{u}, \vec{v}}} + E_{\varpi_j} + \frac{\gamma E_{h_{i,j}}^{\rightarrow}}{ F_{i,j} }}{2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \right) \quad (40)$			
$\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{ F } + E_{\varpi_j} + E_{f_{\vec{u}, \vec{v}}} + E_{\varpi_j}}{2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{ F } - E_{\varpi_j} + E_{\varpi_j}} \quad (41)$			

Table 3: List of all inequalities to be satisfied in the proof of Theorem 9.3.

(I-a)  $v_{\text{top}}$  fails at  $t = 1$ :

$$\Phi(b_{v_{\text{top}}} - \iota_{v_{\text{top}}} + E_{v_{\text{top}}}) > \gamma(b_{v_{\text{top}}} + E_{v_{\text{top}}}) \equiv \Phi(n + E_{v_{\text{top}}}) > \gamma(n + E_{v_{\text{top}}}) \equiv \boxed{\Phi > \gamma} \quad (17)$$

(I-b) Each node  $\vec{u} \in \overrightarrow{V^{\text{left}}} \cup \overrightarrow{V^{\text{right}}}$  that was not shocked at  $t = 1$  fails at  $t = 2$ :

$$\begin{aligned} \frac{\min\{\Phi(b_{v_{\text{top}}} - \iota_{v_{\text{top}}} + E_{v_{\text{top}}}) - \gamma(b_{v_{\text{top}}} + E_{v_{\text{top}}}), b_{v_{\text{top}}}\}}{\deg_{\text{in}}(v_{\text{top}})} &> \gamma(b_{\vec{u}} + E_{\vec{u}}) \\ &\equiv \frac{\min\{\Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n\}}{n} > \gamma(\deg(u) + E_{\vec{u}}) \end{aligned}$$

These constraints are satisfied provided:

$$\frac{\Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}})}{n} > \gamma(\deg(u) + E_{\vec{u}}) \equiv \boxed{\Phi > \gamma \left( 1 + \frac{\deg(u) + E_{\vec{u}}}{1 + \frac{E_{v_{\text{top}}}}{n}} \right)} \quad (18)$$

$$\Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}) \leq n \equiv \boxed{\Phi \leq \gamma + \frac{1}{1 + \frac{E_{v_{\text{top}}}}{n}}} \quad (19)$$

(I-c) If the nodes  $\vec{u}$ ,  $\vec{v}$  and  $f_{\vec{u}, \vec{v}}$  were *not* shocked at  $t = 1$ , then the part of the shock, say  $\sigma_1$ , given to  $v_{\text{top}}$  that is received by node  $f_{\vec{u}, \vec{v}}$  at  $t = 3$  is:

$$\begin{aligned} \sigma_1 &= \frac{\min\left\{\frac{\min\{\Phi(b_{v_{\text{top}}} - \iota_{v_{\text{top}}} + E_{v_{\text{top}}}) - \gamma(b_{v_{\text{top}}} + E_{v_{\text{top}}}), b_{v_{\text{top}}}\}}{\deg_{\text{in}}(v_{\text{top}})} - \gamma(b_{\vec{u}} + E_{\vec{u}}), b_{\vec{u}}\right\}}{\deg_{\text{in}}(\vec{u})} + \frac{\min\left\{\frac{\min\{\Phi(b_{v_{\text{top}}} - \iota_{v_{\text{top}}} + E_{v_{\text{top}}}) - \gamma(b_{v_{\text{top}}} + E_{v_{\text{top}}}), b_{v_{\text{top}}}\}}{\deg_{\text{in}}(v_{\text{top}})} - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}}\right\}}{\deg_{\text{in}}(\vec{v})} \\ &= \frac{\min\left\{\frac{\min\{\Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n\}}{n} - \gamma(\deg(u) + E_{\vec{u}}), \deg(u)\right\}}{\deg(u)} \\ &\quad + \frac{\min\left\{\frac{\min\{\Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n\}}{n} - \gamma(\deg(v) + E_{\vec{v}}), \deg(v)\right\}}{\deg(v)} \end{aligned}$$

On the other hand, if the node  $f_{\vec{u}, \vec{v}}$  and *exactly* one of the nodes  $\vec{u}$  and  $\vec{v}$ , say  $\vec{u}$ , were *not* shocked at  $t = 1$ , then the part of the shock, say  $\sigma'_1$ , given to  $v_{\text{top}}$  that is received by node  $f_{\vec{u}, \vec{v}}$  at  $t = 3$  is:

$$\sigma'_1 = \frac{\min\left\{\frac{\min\{\Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n\}}{n} - \gamma(\deg(u) + E_{\vec{u}}), \deg(u)\right\}}{\deg(u)}$$

(II) Suppose that some node  $\vec{u}$  is shocked at  $t = 1$ . Then, the following happens.

(II-a) Node  $\vec{u}$  fails at  $t = 1$ :

$$\Phi(b_{\vec{u}} - \iota_{\vec{u}} + E_{\vec{u}}) > \gamma(b_{\vec{u}} + E_{\vec{u}}) \equiv \boxed{\Phi > \gamma \left( 1 + \frac{1}{\deg(u) - 1 + E_{\vec{u}}} \right)} \quad (20)$$

**(II-b)** Node  $f_{\vec{u}, \vec{v}} \in \vec{F}_{i,j}$  fails at  $t = 2$  and node  $\vec{h}_{i,j}$  fails at  $t = 3$  if both  $\vec{u}$  and  $\vec{v}$  were shocked at  $t = 1$ :

$$\begin{aligned}
& \frac{\min\{\Phi(b_{\vec{u}} - \iota_{\vec{u}} + E_{\vec{u}}) - \gamma(b_{\vec{u}} + E_{\vec{u}}), b_{\vec{u}}\}}{\deg_{\text{in}}(\vec{u})} + \frac{\min\{\Phi(b_{\vec{v}} - \iota_{\vec{v}} + E_{\vec{v}}) - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}}\}}{\deg_{\text{in}}(\vec{v})} > \gamma(b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}}) \\
& \equiv \\
& \frac{\min\{\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}}), \deg(u)\}}{\deg(u)} \\
& + \frac{\min\{\Phi(\deg(v) - 1 + E_{\vec{v}}) - \gamma(\deg(v) + E_{\vec{v}}), \deg(v)\}}{\deg(v)} > \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) \\
& \frac{\min\left\{\frac{\min\{\Phi(b_{\vec{u}} - \iota_{\vec{u}} + E_{\vec{u}}) - \gamma(b_{\vec{u}} + E_{\vec{u}}), b_{\vec{u}}\}}{\deg_{\text{in}}(\vec{u})} + \frac{\min\{\Phi(b_{\vec{v}} - \iota_{\vec{v}} + E_{\vec{v}}) - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}}\}}{\deg_{\text{in}}(\vec{v})} - \gamma(b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}}), b_{f_{\vec{u}, \vec{v}}}\right\}}{\deg_{\text{in}}(f_{\vec{u}, \vec{v}})} \\
& > \gamma(b_{\vec{h}_{i,j}} + E_{\vec{h}_{i,j}}) \\
& \equiv \\
& \min\left\{\frac{\min\{\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}}), \deg(u)\}}{\deg(u)} + \frac{\min\{\Phi(\deg(v) - 1 + E_{\vec{v}}) - \gamma(\deg(v) + E_{\vec{v}}), \deg(v)\}}{\deg(v)} - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}), 1\right\} > \gamma E_{\vec{h}_{i,j}}
\end{aligned}$$

These constraints are satisfied provided the inequalities (17)–(20) are satisfied, and the following holds:

$$\boxed{\frac{\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}})}{\deg(u)} + \frac{\Phi(\deg(v) - 1 + E_{\vec{v}}) - \gamma(\deg(v) + E_{\vec{v}})}{\deg(v)} > \gamma(1 + E_{f_{\vec{u}, \vec{v}}})} \quad (21)$$

$$\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}}) \leq \deg(u) \equiv \boxed{\Phi \leq \gamma\left(\frac{\deg(u) + E_{\vec{u}}}{\deg(u) - 1 + E_{\vec{u}}}\right) + \frac{\deg(u)}{\deg(u) - 1 + E_{\vec{u}}}} \quad (22)$$

$$\boxed{\frac{\Phi(\deg(u) - 1 + E_{\vec{u}}) - \gamma(\deg(u) + E_{\vec{u}})}{\deg(u)} + \frac{\Phi(\deg(v) - 1 + E_{\vec{v}}) - \gamma(\deg(v) + E_{\vec{v}})}{\deg(v)} - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) > \gamma E_{\vec{h}_{i,j}}} \quad (23)$$

$$\boxed{\gamma E_{\vec{h}_{i,j}} < 1} \quad (24)$$

**(III)** When the node  $v_{\text{side}}$  is shocked at  $t = 1$ , the following happens.

**(III-a)**  $v_{\text{side}}$  fails at  $t = 1$ :

$$\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) > \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}}) \equiv \Phi(|F| + E_{v_{\text{side}}}) > \gamma(|F| + E_{v_{\text{side}}}) \equiv \Phi > \gamma$$

which is same as (17).

**(III-b)** If a node  $\varpi_j \in \varpi$  is shocked at  $t = 1$ , it does not fail:

$$\Phi(b_{\varpi_j} - \iota_{\varpi_j} + E_{\varpi_j}) \leq \gamma(b_{\varpi_j} + E_{\varpi_j}) \equiv \boxed{\Phi \leq \gamma\left(1 + \frac{1}{E_{\varpi_j}}\right)} \quad (25)$$

**(III-c)** Any node  $\varpi_j \in \varpi$  fails at  $t = 2$  irrespective of whether  $\varpi_j$  was shocked or not:

$$\frac{\min\{\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) - \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}}), b_{v_{\text{side}}}\}}{\deg_{\text{in}}(v_{\text{side}})} > \gamma(b_{\varpi_j} + E_{\varpi_j})$$

These constraints are satisfied provided:

$$\frac{\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) - \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}})}{\deg_{\text{in}}(v_{\text{side}})} > \gamma(b_{\varpi_j} + E_{\varpi_j}) \equiv \boxed{\Phi > \frac{\gamma\left(2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}\right)}{\left(1 + \frac{E_{v_{\text{side}}}}{|F|}\right)}} \quad (26)$$

$$\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) - \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}}) \leq b_{v_{\text{side}}} \equiv \boxed{\Phi \leq \gamma + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{|F|}}} \quad (27)$$

**(III-d)** If a node  $\varpi_j \in \varpi$  is shocked at  $t = 1$ , it does not fail (and thus, by **(III-b)**, it does not fail at  $t = 2$  also):

$$\Phi(b_{\varpi_j} - \iota_{\varpi_j} + E_{\varpi_j}) \leq \gamma(b_{\varpi_j} + E_{\varpi_j}) \equiv \boxed{\Phi \leq \gamma\left(1 + \frac{1}{E_{\varpi_j}}\right)} \quad (28)$$

**(III-e)** Any node  $\varpi_j \in \varpi$  fails at  $t = 3$  irrespective of whether  $\varpi_j$  was shocked or not:

$$\frac{\min\left\{\frac{\min\{\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) - \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}}), b_{v_{\text{side}}}\}}{\deg_{\text{in}}(v_{\text{side}})} - \gamma(b_{\varpi_j} + E_{\varpi_j}), b_{\varpi_j}\right\}}{\deg_{\text{in}}(\varpi_j)} > \gamma(b_{\varpi_j} + E_{\varpi_j}) \equiv$$

$$\min\left\{\frac{\min\{\Phi(|F| + E_{v_{\text{side}}}) - \gamma(|F| + E_{v_{\text{side}}}), |F|\}}{|F|} - \gamma(1 + E_{\varpi_j}), 1\right\} > \gamma(1 + E_{\varpi_j})$$

These constraints are satisfied provided all the previous constraints hold and the following holds:

$$\frac{\Phi(|F| + E_{v_{\text{side}}}) - \gamma(|F| + E_{v_{\text{side}}})}{|F|} - \gamma(1 + E_{\varpi_j}) > \gamma(1 + E_{\varpi_j}) \equiv \boxed{\Phi > \gamma\left(\frac{(|F| + E_{v_{\text{side}}})}{3|F| + |F|E_{\varpi_j} + |F|E_{\varpi_j} + E_{v_{\text{side}}}}\right)} \quad (29)$$

$$\frac{\Phi(|F| + E_{v_{\text{side}}}) - \gamma(|F| + E_{v_{\text{side}}})}{|F|} - \gamma(1 + E_{\varpi_j}) \leq 1 \equiv \boxed{\Phi \leq \gamma\left(1 + \frac{1 + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{|F|}}\right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{|F|}}} \quad (30)$$

**(III-f)** Consider a directed path  $(v_{\text{side}} \leftarrow \varpi_j \leftarrow \varpi_j \leftarrow p_j)$  from  $p_j = f_{\vec{u}, \vec{v}}$  to  $v_{\text{side}}$ . The maximum value of its proportion of shock receive by  $p_j$  from this path, say  $\sigma_2$ , is obtained by shocking all the nodes  $v_{\text{side}}, \varpi_j, \varpi_j$  and is given by (assuming all previous inequalities hold):

$$\sigma_2 = \frac{\min\left\{\frac{\min\left\{\frac{\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) - \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}})}{\deg_{\text{in}}(v_{\text{side}})} - \left(\gamma(b_{\varpi_j} + E_{\varpi_j}) - \Phi(b_{\varpi_j} - \iota_{\varpi_j} + E_{\varpi_j})\right), b_{\varpi_j}\right\}}{\deg_{\text{in}}(\varpi_j)} - \left(\gamma(b_{\varpi_j} + E_{\varpi_j}) - \Phi(b_{\varpi_j} - \iota_{\varpi_j} + E_{\varpi_j})\right), b_{\varpi_j}\right\}}{\deg_{\text{in}}(\varpi_j)}$$

$$= \min\left\{\min\left\{\Phi\left(1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j}\right) - \gamma\left(2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}\right), 1\right\} - \left(\gamma(1 + E_{\varpi_j}) - \Phi E_{\varpi_j}\right), 1\right\}$$

Similarly, the minimum value of its proportion of shock receive by  $p_j$  from this path, say  $\sigma_2$ , is obtained by shocking only the node  $v_{\text{side}}$  and is given by (assuming all previous inequalities hold):

$$\begin{aligned}\sigma'_2 &= \frac{\min \left\{ \frac{\Phi(b_{v_{\text{side}}} - \iota_{v_{\text{side}}} + E_{v_{\text{side}}}) - \gamma(b_{v_{\text{side}}} + E_{v_{\text{side}}})}{\frac{\deg_{\text{in}}(v_{\text{side}})}{\deg_{\text{in}}(\varpi_j)}} - \gamma(b_{\varpi_j} + E_{\varpi_j}), b_{\varpi_j} \right\}}{\deg_{\text{in}}(\varpi_j)} \\ &= \min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - \gamma(1 + E_{\varpi_j}), 1 \right\}\end{aligned}$$

We want node  $f_{\vec{u}, \vec{v}}$  to fail at  $t = 4$  assuming it did not fail already. Since  $f_{\vec{u}, \vec{v}}$  did not fail at  $t = 2$ , at most one of the nodes  $\vec{u}$  and  $\vec{v}$  was shocked. There are two cases to consider: when neither  $\vec{u}$  nor  $\vec{v}$  was shocked, or when exactly one of these nodes, say  $\vec{v}$ , was shocked (assuming all previous inequalities hold):

$$\begin{aligned}\sigma'_2 + \sigma_1 &= \min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - \gamma(1 + E_{\varpi_j}), 1 \right\} \\ &+ \frac{\min \left\{ \frac{\min \{ \Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n \}}{n} - \gamma(\deg(u) + E_{\vec{u}}), \deg(u) \right\}}{\deg(u)} \\ &+ \frac{\min \left\{ \frac{\min \{ \Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n \}}{n} - \gamma(\deg(v) + E_{\vec{v}}), \deg(v) \right\}}{\deg(v)} \\ &> \gamma(b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}}) \\ &\equiv \\ &\min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - \gamma(1 + E_{\varpi_j}), 1 \right\} \\ &+ \frac{\min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{top}}}}{n} \right) - \gamma \left( 1 + \frac{E_{v_{\text{top}}}}{n} \right) - \gamma(\deg(u) + E_{\vec{u}}), \deg(u) \right\}}{\deg(u)} \\ &+ \frac{\min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{top}}}}{n} \right) - \gamma \left( 1 + \frac{E_{v_{\text{top}}}}{n} \right) - \gamma(\deg(v) + E_{\vec{v}}), \deg(v) \right\}}{\deg(v)} \\ &> \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) \\ &\sigma'_2 + \sigma'_1 + \frac{\min \{ \Phi(b_{\vec{v}} - \iota_{\vec{v}} + E_{\vec{v}}) - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}} \}}{\deg_{\text{in}}(\vec{v})} > \gamma(b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}}) \\ &\equiv\end{aligned}$$

$$\min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - \gamma(1 + E_{\varpi_j}), 1 \right\} + \frac{\min \left\{ \frac{\min \{ \Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n \}}{n} - \gamma(\deg(u) + E_{\vec{u}}), \deg(u) \right\}}{\deg(u)}$$

$$\begin{aligned}
& + \frac{\min\{\Phi(b_{\vec{v}} - \iota_{\vec{v}} + E_{\vec{v}}) - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}}\}}{\deg_{\text{in}}(\vec{v})} > \gamma(b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}}) \\
& \equiv \\
& \min\left\{\Phi\left(1 + \frac{E_{v_{\text{side}}}}{|F|}\right) - \gamma\left(2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}\right) - \gamma(1 + E_{\mathfrak{m}_j}), 1\right\} \\
& + \frac{\min\left\{\Phi\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma(\deg(u) + E_{\vec{u}}), \deg(u)\right\}}{\deg(u)} \\
& + \frac{\min\{\Phi(\deg(\vec{v}) - 1 + E_{\vec{v}}) - \gamma(\deg(\vec{v}) + E_{\vec{v}}), \deg(\vec{v})\}}{\deg_{\text{in}}(\vec{v})} > \gamma(1 + E_{f_{\vec{u}, \vec{v}}})
\end{aligned}$$

These constraints are satisfied provided all the previous constraints hold and the following holds:

$$\begin{aligned}
& \Phi\left(1 + \frac{E_{v_{\text{side}}}}{|F|}\right) - \gamma\left(2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}\right) - \gamma(1 + E_{\mathfrak{m}_j}) + \\
& \frac{\Phi\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma(\deg(u) + E_{\vec{u}})}{\deg(u)} + \frac{\Phi\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma(\deg(v) + E_{\vec{v}})}{\deg(v)} > \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) \\
& \equiv \boxed{\Phi > \gamma\left(\frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\mathfrak{m}_j}}{1 + \frac{E_{v_{\text{side}}}}{|F|} + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)}}\right)} \quad (31)
\end{aligned}$$

$$\Phi\left(1 + \frac{E_{v_{\text{side}}}}{|F|}\right) - \gamma\left(2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}\right) - \gamma(1 + E_{\mathfrak{m}_j}) \leq 1 \equiv \boxed{\Phi \leq \gamma\left(\frac{3 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\mathfrak{m}_j}}{1 + \frac{E_{v_{\text{side}}}}{|F|}}\right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{|F|}}} \quad (32)$$

$$\Phi\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma(\deg(u) + E_{\vec{u}}) \leq \deg(u) \equiv \boxed{\Phi \leq \gamma\left(\frac{1 + \frac{E_{v_{\text{top}}}}{n} + \deg(u) + E_{\vec{u}}}{1 + \frac{E_{v_{\text{top}}}}{n}}\right) + \frac{\deg(u)}{1 + \frac{E_{v_{\text{top}}}}{n}}} \quad (33)$$

$$\begin{aligned}
& \Phi\left(1 + \frac{E_{v_{\text{side}}}}{|F|}\right) - \gamma\left(2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}\right) - \gamma(1 + E_{\mathfrak{m}_j}) + \frac{\Phi\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma\left(1 + \frac{E_{v_{\text{top}}}}{n}\right) - \gamma(\deg(u) + E_{\vec{u}})}{\deg(u)} \\
& + \frac{\Phi(\deg(\vec{v}) - 1 + E_{\vec{v}}) - \gamma(\deg(\vec{v}) + E_{\vec{v}})}{\deg_{\text{in}}(\vec{v})} > \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) \\
& \equiv \\
& \boxed{\Phi > \gamma\left(\frac{6 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\mathfrak{m}_j} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}} + 1}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}}}{2 + \frac{E_{v_{\text{side}}}}{|F|} + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{v}} - 1}{\deg(v)}}\right)} \quad (34)
\end{aligned}$$

(IV) By (II-b) node  $\vec{h}_{i,j}$  fails at  $t = 3$  provided both the nodes  $\vec{u}$  and  $\vec{v}$  were shocked at  $t = 1$ .

Our goal is to make sure that node  $\vec{h}_{i,j}$  does not fail in any other condition (assuming the node itself was not shocked). Assuming the nodes  $\vec{u}$ ,  $\vec{v}$  and  $f_{\vec{u}, \vec{v}}$  were not shocked, the maximum amount of shock that  $f_{\vec{u}, \vec{v}} \in \vec{F}_{i,j}$  can receive is when all the nodes before  $f_{\vec{u}, \vec{v}}$  in the path

$(v_{\text{side}}) \leftarrow (\varpi_j) \leftarrow (\mathfrak{m}_j) \leftarrow (p_j) \leftarrow (f_{\vec{u}, \vec{v}})$  were shocked and no more than one of the nodes  $\vec{u}$  or  $\vec{v}$  was

shocked. Based on this, the following constraints must hold for  $\vec{h}_{i,j}$  not to fail.

$$\begin{aligned}
& \frac{\min \left\{ \sigma_1 + \sigma_2 - \gamma \left( b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}} \right), b_{f_{\vec{u}, \vec{v}}} \right\}}{\deg_{\text{in}}(f_{\vec{u}, \vec{v}})} \leq \frac{\gamma \left( b_{\vec{h}_{i,j}} + E_{\vec{h}_{i,j}} \right)}{|F_{i,j}|} \\
& \equiv \\
& \min \left\{ \frac{\min \left\{ \frac{\min \{ \Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n \}}{n} - \gamma(\deg(u) + E_{\vec{u}}), \deg(u) \right\}}{\deg(u)} \right. \\
& \quad \left. + \frac{\min \left\{ \frac{\min \{ \Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n \}}{n} - \gamma(\deg(v) + E_{\vec{v}}), \deg(v) \right\}}{\deg(v)} \right. \\
& \quad \left. + \min \left\{ \min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right), 1 \right\} - (\gamma(1 + E_{\varpi_j}) - \Phi E_{\varpi_j}), 1 \right\} - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}), 1 \right\} \\
& \leq \frac{\gamma E_{\vec{h}_{i,j}}}{|F_{i,j}|} \\
& \equiv \\
& \min \left\{ \Phi \left( \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} \right) - \gamma \left( 3 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} \right) \right. \\
& \quad \left. + \min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j} \right) - \gamma \left( 3 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\varpi_j} \right), 1 \right\}, 1 \right\} \leq \frac{\gamma E_{\vec{h}_{i,j}}}{|F_{i,j}|}
\end{aligned}$$

These constraints are satisfied provided all the previous constraints hold and the following holds:

$$\Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j} \right) - \gamma \left( 3 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\varpi_j} \right) \leq 1 \equiv \boxed{\Phi \leq \gamma \left( \frac{3 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}}} \quad (35)$$

$$\boxed{\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\varpi_j}}{1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}}} \quad (36)$$

$$\boxed{\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + E_{f_{\vec{u}, \vec{v}}} + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{\varpi_j} + \frac{E_{\vec{h}_{i,j}}}{|F_{i,j}|}}{1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{1}{\deg(v)} + \frac{E_{v_{\text{top}}}}{n \deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}} \right) \quad (37)$$

On the other hand, if exactly one of the nodes  $\vec{u}$  or  $\vec{v}$ , say  $\vec{u}$ , was shocked at  $t = 1$ , then the maximum amount of shock that  $f_{\vec{u}, \vec{v}} \in \vec{F}_{i,j}$  can receive is modified, and the new conditions for our desired goal become as follows.

$$\begin{aligned}
& \frac{\min \left\{ \sigma'_1 + \frac{\min \{ \Phi(b_{\vec{v}} - \iota_{\vec{v}} + E_{\vec{v}}) - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}} \}}{\deg_{\text{in}}(\vec{v})} + \sigma_2 - \gamma(b_{f_{\vec{u}, \vec{v}}} + E_{f_{\vec{u}, \vec{v}}}), b_{f_{\vec{u}, \vec{v}}} \right\}}{\deg_{\text{in}}(f_{\vec{u}, \vec{v}})} \leq \frac{\gamma(b_{h_{i,j}} + E_{h_{i,j}})}{|F_{i,j}|} \\
& \equiv \\
& \min \left\{ \frac{\min \left\{ \frac{\min \{ \Phi(n + E_{v_{\text{top}}}) - \gamma(n + E_{v_{\text{top}}}), n \}}{n} - \gamma(\deg(u) + E_{\vec{u}}), \deg(u) \right\}}{\deg(u)} + \frac{\min \{ \Phi(b_{\vec{v}} - \iota_{\vec{v}} + E_{\vec{v}}) - \gamma(b_{\vec{v}} + E_{\vec{v}}), b_{\vec{v}} \}}{\deg_{\text{in}}(\vec{v})} \right. \\
& \quad \left. + \min \left\{ \min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right), 1 \right\} - (\gamma(1 + E_{\mathfrak{m}_j}) - \Phi E_{\mathfrak{m}_j}), 1 \right\} - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}), 1 \right\} \\
& \leq \frac{\gamma E_{h_{i,j}}}{|F_{i,j}|} \\
& \equiv \\
& \min \left\{ \Phi \left( 1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} \right) - \gamma \left( 2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} \right) \right. \\
& \quad \left. + \min \left\{ \min \left\{ \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right), 1 \right\} - (\gamma(1 + E_{\mathfrak{m}_j}) - \Phi E_{\mathfrak{m}_j}), 1 \right\} - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}), 1 \right\} \\
& \leq \frac{\gamma E_{h_{i,j}}}{|F_{i,j}|}
\end{aligned}$$

These constraints are satisfied provided all the previous constraints hold and the following holds:

$$\Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) \leq 1 \equiv \boxed{\Phi \leq \gamma \left( \frac{2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j}}{1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j}} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j}}} \quad (38)$$

$$\Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - (\gamma(1 + E_{\mathfrak{m}_j}) - \Phi E_{\mathfrak{m}_j}) \leq 1 \quad (39)$$

$$\equiv \boxed{\Phi \leq \gamma \left( \frac{2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + 1 + E_{\mathfrak{m}_j}}{1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\mathfrak{m}_j}} \right) + \frac{1}{1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\mathfrak{m}_j}}} \quad (40)$$



$$\begin{aligned}
& \Phi \left( 1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} \right) - \gamma \left( 2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} \right) \\
& + \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - (\gamma(1 + E_{\varpi_j}) - \Phi E_{\varpi_j}) - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) \leq \frac{\gamma E_{h_{i,j}}}{|F_{i,j}|} \\
& \equiv \\
& \boxed{\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{f_{\vec{u}, \vec{v}}} + E_{\varpi_j} + \frac{\gamma E_{h_{i,j}}}{|F_{i,j}|}}{2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}} \right)} \quad (41)
\end{aligned}$$

$$\begin{aligned}
& \Phi \left( 1 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} \right) - \gamma \left( 2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} \right) \\
& + \Phi \left( 1 + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} \right) - \gamma \left( 2 + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} \right) - (\gamma(1 + E_{\varpi_j}) - \Phi E_{\varpi_j}) - \gamma(1 + E_{f_{\vec{u}, \vec{v}}}) \leq 1 \\
& \equiv \\
& \boxed{\Phi \leq \gamma \left( \frac{6 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} + \frac{E_{\vec{u}}}{\deg(u)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} + E_{\varpi_j} + E_{f_{\vec{u}, \vec{v}}} + E_{\varpi_j}}{2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}} \right) + \frac{1}{2 + \frac{1}{\deg(u)} + \frac{E_{v_{\text{top}}}}{n \deg(u)} - \frac{1}{\deg(v)} + \frac{E_{\vec{v}}}{\deg(v)} + \frac{E_{v_{\text{side}}}}{|F|} - E_{\varpi_j} + E_{\varpi_j}}} \quad (42)
\end{aligned}$$

There are many choices of parameters  $\gamma$ ,  $\Phi$  and  $E_y$ 's satisfying inequalities (17)–(26); we exhibit just one:

$$\begin{aligned}
& \gamma = 4n^{-1000} \quad \Phi = n^{-1000} \quad \forall u \in V^{\text{left}} \cup V^{\text{right}}: E_{\vec{u}} = 1 \quad E_{v_{\text{top}}} = n^3 \quad E_{v_{\text{side}}} = n^2 |F| \\
& \forall u \in V^{\text{left}} \forall v \in V^{\text{right}}: E_{f_{\vec{u}, \vec{v}}} = 1 \quad \forall h_{i,j} \in F_{\text{super}} \forall f_{u,v} \in F_{i,j}: E_{h_{i,j}} = 1 \quad \forall j: E_{\varpi_j} = E_{\varpi_j} = \frac{1}{4}
\end{aligned}$$

Remembering that  $10 \leq \deg(u) < n$  for any node  $u \in V^{\text{left}} \cup V^{\text{right}}$  and  $|F_{i,j}| < |F|$ , it is relatively straightforward to verify that all the inequalities are satisfied for all sufficiently large  $n$ . Note that

$$E = E_{v_{\text{top}}} + E_{v_{\text{side}}} + \sum_{u \in V^{\text{left}} \cup V^{\text{right}}} E_{\vec{u}} + \sum_{\{u,v\} \in F} E_{f_{\vec{u}, \vec{v}}} + \sum_{h_{i,j} \in F_{\text{super}}} E_{h_{i,j}} + \sum_{j=1}^{|F|} (E_{\varpi_j} + \sum E_{\varpi_j}) = n^3 + n^2 |F| + n + \frac{3}{2} |F| + |F_{\text{super}}|$$

and thus the ratio of total external assets to total internal assets  $E/I$  is large. We can now finish our proof by selecting  $\delta$  such that  $\log^{1-\delta} n = \log^{1-\varepsilon} |\vec{V}| - 1$  and showing the following:

**(completeness)** If MINREP has a solution of size  $\alpha + \beta$  on  $G$  then then  $\text{SI}^*(\vec{G}, T) \leq \alpha + \beta + 2$ .

**(soundness)** If every solution of MINREP on  $G$  is of size at least  $(\alpha + \beta)2^{\log^{1-\delta} n}$  then  $\text{SI}^*(\vec{G}, T) \geq \frac{\alpha + \beta}{2} 2^{\log^{1-\delta} n}$ .

### Proof of Completeness (MINREP has a solution of size $\alpha + \beta$ )

Let  $V_1 \subseteq V^{\text{left}}$  and  $V_2 \subseteq V^{\text{right}}$  be a solution of MINREP such that  $|V_1| + |V_2| = \alpha + \beta$ . We shock the nodes  $v_{\text{top}}$  and  $v_{\text{side}}$ , and every node  $\vec{u}$  for every  $u \in V^{\text{left}} \cup V^{\text{right}}$ . By **(I-a)**  $v_{\text{top}}$  fails at  $t = 1$ , and by **(I-b)** and **(II-a)** every node in  $\cup_{i=1}^{\alpha} V_i^{\text{left}} \cup \cup_{j=1}^{\beta} V_j^{\text{right}}$  fails on or before  $t = 2$ . By **(III-a)**, **(III-b)** and **(III-c)** every node in  $\{V_{\text{shock}}\} \cup \varpi \cup \varpi$  fails on or before  $t = 3$ . Since  $V_1$  and  $V_2$  are a valid solution of MINREP, for every

super-edge  $h_{i,j}$  there exists  $u \in V_1$  and  $v \in V_2$  such that  $u \in V_i^{\text{left}}$ ,  $v \in V_j^{\text{right}}$  and  $\{u, v\} \in F$ ; since we shock the nodes  $\vec{u}$  and  $\vec{v}$ , by **(II-a)** both  $\vec{u}$  and  $\vec{v}$  fail at  $t = 1$ , by **(II-b)** the node  $f_{\vec{u}, \vec{v}}$  fails at  $t = 2$ , and by **(II-c)** the node  $\vec{h}_{i,j}$  fails at  $t = 3$ . Thus, the network  $\vec{G}$  fails at  $t = 3$  and  $\text{SI}^*(\vec{G}, T) = \alpha + \beta + 1$  for  $t \geq 4$ .

### Proof of Soundness (every solution of MINREP is of size at least $(\alpha + \beta)2^{\log^{1-\delta} n}$ )

We will prove the logically equivalent contrapositive of our claim, *i.e.*, we will show that if  $\text{SI}^*(\vec{G}, T) < \frac{\alpha + \beta}{2} 2^{\log^{1-\delta} n}$  then MINREP has a solution of size strictly less than  $(\alpha + \beta)2^{\log^{1-\delta} n}$ . Consider a solution of  $\text{STABILITY}_{T, \Phi}$  on  $\vec{G}$  that shocks at most  $z = \frac{\alpha + \beta}{2} 2^{\log^{1-\delta} n}$  nodes. Note that the nodes  $v_{\text{top}}$  and  $v_{\text{side}}$  must be shocked at  $t = 1$  by Proposition 7.1(a). By **(I-a)** and **(III-a)**, the nodes  $v_{\text{top}}$  and  $v_{\text{side}}$  fails at  $t = 1$ , by **(I-b)** and **(III-c)** every node in  $\vec{V}^{\text{left}} \cup \vec{V}^{\text{right}} \cup \vec{\Xi}$  fails at  $t = 2$ , by **(III-e)** every node in  $\vec{\mathfrak{M}}$  fails at  $t = 3$ , by **(III-f)** every node  $f_{\vec{u}, \vec{v}}$  fails at  $t = 4$  unless it was shocked at  $t = 1$  and by **(IV)** a node  $\vec{h}_{i,j}$  fails only if  $\vec{h}_{i,j}$ ,  $f_{\text{ovru}, \text{ovru}} \in \vec{F}_{i,j}$  or both the nodes  $\vec{u}$  and  $\vec{v}$  were shocked at  $t = 1$ . We “normalize” this given solution in the following manner (each step of the normalization assumes that the previous steps have been already carried out):

- If a node from  $\vec{\mathfrak{M}} \cup \vec{\Xi}$  was shocked at  $t = 1$ , we do not shock it. By **(III)** this has no effect on the failure of the network.
- If a node  $f_{\vec{u}, \vec{v}} \in \vec{F}_{i,j}$  was shocked, we do not shock it but instead shock the nodes  $\vec{u}$  and  $\vec{v}$  if they were not already shocked in the given solution. This at most doubles the number of nodes shocked and, by **(II-b)**, the node  $f_{u,v}$  fails at  $t = 2$  and the node  $\vec{h}_{i,j}$  fails at  $t = 3$  if it was not shocked at  $t = 1$ . Thus, after this sequence of normalization steps, we may assume that no  $f_{\vec{u}, \vec{v}}$  node was shocked.
- If a node  $\vec{h}_{i,j}$  was shocked at  $t = 1$ , we do not shock it but instead shock the nodes  $\vec{u}$  and  $\vec{v}$  (for some  $u$  and  $v$  such that  $\{u, v\} \in F_{i,j}$ ) if they were not already shocked in the given solution. This at most doubles the number of nodes shocked and, by **(II-b)**, the node  $f_{u,v}$  fails at  $t = 2$  and the node  $\vec{h}_{i,j}$  fails at  $t = 3$ . Thus, after this sequence of normalization steps, we may assume that no  $\vec{h}_{i,j}$  node was shocked.

These normalizations result in a solution of  $\text{STABILITY}_{T, \Phi}$  of size at most  $2z$  in which the nodes  $v_{\text{top}}$ ,  $v_{\text{side}}$ , a subset  $\vec{V}_1 \subseteq \vec{V}^{\text{left}}$  and a subset  $\vec{V}_2 \subseteq \vec{V}^{\text{right}}$  of nodes. Our solution of MINREP is  $V_1 = \{v \mid \vec{v} \in \vec{V}_1\} \subseteq V^{\text{left}}$  and  $V_2 = \{v \mid \vec{v} \in \vec{V}_2\} \subseteq V^{\text{right}}$  of size  $2z - 2 < 2z$ . Since failure of every  $\vec{h}_{i,j}$  is attributed to shocking two nodes  $\vec{u}$  and  $\vec{v}$  such that  $f_{\vec{u}, \vec{v}} \in \vec{F}_{i,j}$ , every super-edge  $h_{i,j}$  of  $G$  is witnessed by the two nodes  $u$  and  $v$ .  $\square$

## 10 Our Results on the Dual Stability Index for Homogeneous Networks

For any  $T > 1$  our first result in Theorem 10.1(a) provides an inapproximability gap of about  $\frac{e}{e-1} \approx 1.582$ . The second result exhibits a polynomial time algorithm for rooted in-arborescences assuming every node can be individually shocked to fail.

**Theorem 10.1.** *For any  $T > 1$ , the following results hold:*

- (a) *Assuming  $P \neq NP$ ,  $\text{DSI}^*(G, T, \kappa)$  cannot be approximated to within a factor of  $(1 - e^{-1} + \delta)^{-1}$ , for any  $\delta > 0$ , even if  $G$  is a DAG ( $e$  is the base of natural logarithm).*

- (b) If  $G$  is a rooted in-arborescence then  $\text{DSI}^*(G, T, \kappa) < 1 + \deg_{\text{in}}^{\max} \left( \frac{\Phi}{\gamma} - 1 \right)$ , where  $\deg_{\text{in}}^{\max} = \max_{v \in V} \{\deg_{\text{in}}(v)\}$  is the maximum in-degree over all nodes of  $G$ . Moreover, under the assumption that any individual node of the network can be failed by shocking,  $\text{DSI}^*(G, T, \kappa)$  can be computed exactly in  $O(n^3)$  time.

*Proof.*

(a) The max  $\kappa$ -cover problem is defined as follows. An instance of the problem is an universe  $\mathcal{U}$  of  $n$  elements, a collection of  $m$  sets  $\mathcal{S}$  over  $\mathcal{U}$ , and a positive integer  $\kappa$ . The goal is to pick a sub-collection  $\mathcal{S}' \subseteq \mathcal{S}$  of  $\kappa$  sets such that the number of elements covered, namely  $|\cup_{S \in \mathcal{S}'} S|$ , is maximized. Let  $\text{OPT}$  denote the maximum number of elements covered by an optimal solution of the max  $\kappa$ -cover problem. It was shown in [22] that, assuming  $P \neq NP$ , the max  $\kappa$ -cover problem cannot be approximated to within a factor of  $\frac{1}{(1-\frac{1}{e}+\delta)}$  for any constant  $\delta > 0$ . More precisely, [22] provides a polynomial-time reduction for a restricted but still NP-hard version of the Boolean satisfiability problem (3-CNF5) instances of max  $\kappa$ -cover with  $\kappa = |\mathcal{U}|^\alpha$ , for some constant  $0 < \alpha < 1$ , and shows that

- (1) if the CNF formula is satisfiable, then  $\text{OPT} = |\mathcal{U}|$ ;
- (2) if the CNF formula is not satisfiable, then  $\text{OPT} < (1 - \frac{1}{e} + g(\kappa)) |\mathcal{U}|$ , where  $g(\kappa) \rightarrow 0$  and  $\kappa \rightarrow \infty$ .

Our reduction from max  $\kappa$ -cover to  $\text{DUAL-STABILITY}_{T, \kappa}$  is as follows<sup>3</sup>. In our graph  $G = (V, F)$ , we have an element node  $\tilde{u}$  for every element  $u \in \mathcal{U}$ , a set node  $\tilde{S}$  for every set  $S \in \mathcal{S}$ , and directed edges  $(\tilde{u}, \tilde{S})$  for every element  $u \in \mathcal{U}$  and set  $S \in \mathcal{S}$  such that  $u \in S$ . Thus,  $n = |V| = |\mathcal{U}| + |\mathcal{S}|$  and  $|F| = \sum_{S \in \mathcal{S}} |S|$ . We now set the remaining parameters as follows:  $E = n$ ,  $\gamma = n^{-2}$  and  $\Phi = 1$ . Now, we observe the following:

- If an element node  $\tilde{u}$  is shocked, it does not fail since  $\Phi(\deg_{\text{in}}(\tilde{u}) - \deg_{\text{out}}(\tilde{u}) + \frac{E}{n}) \leq 0$  whereas  $\gamma(\deg_{\text{in}}(\tilde{u}) + \frac{E}{n}) = n^{-2} > 0$ .
- If a set node  $\tilde{S}$  is shocked, it fails since  $\Phi(\deg_{\text{in}}(\tilde{S}) - \deg_{\text{out}}(\tilde{S}) + \frac{E}{n}) \geq 2$  whereas  $\gamma(\deg_{\text{in}}(\tilde{S}) + \frac{E}{n}) \leq \frac{n+1}{n^2} < 1$ .
- If a set node  $\tilde{S}$  is shocked, then every element node  $\tilde{u}$  for  $u \in S$  fails at  $t = 2$ . To observe this, note that

$$\frac{\min \{ \Phi(\deg_{\text{in}}(\tilde{S}) - \deg_{\text{out}}(\tilde{S}) + \frac{E}{n}) - \gamma(\deg_{\text{in}}(\tilde{u}) + \frac{E}{n}), \deg_{\text{in}}(\tilde{S}) \}}{\deg_{\text{in}}(\tilde{S})} \geq \frac{2 - \frac{n+1}{n^2}}{n} > \frac{n+1}{n^2} \geq \gamma \left( \deg_{\text{in}}(\tilde{S}) + \frac{E}{n} \right)$$

- Since the longest directed path in  $G$  has one edge, no new nodes fails during  $t > 2$ .

Based on the above observations, one can identify the sets selected in max  $k$ -cover with the set nodes selected for shocking in  $\text{DUAL-STABILITY}_{T, \kappa}$  on  $G$  to conclude that  $\text{DSI}^*(G, T, \kappa) = \text{OPT} + \kappa$ . Thus, using (1) and (2), inapproximability gap is

$$\frac{|\mathcal{U}| + \kappa}{(1 - \frac{1}{e} + g(\kappa)) |\mathcal{U}| + \kappa} = \frac{|\mathcal{U}| + |\mathcal{U}|^\alpha}{(1 - \frac{1}{e} + g(\kappa)) |\mathcal{U}| + |\mathcal{U}|^\alpha} \rightarrow \frac{1}{1 - \frac{1}{e} + \delta} \text{ as } |\mathcal{U}| \rightarrow \infty \text{ for any } \delta > 0$$

(b) The bound  $\text{DSI}^*(G, T, \kappa) < 1 + \deg_{\text{in}}^{\max} \left( \frac{\Phi}{\gamma} - 1 \right)$  follows directly using Lemma 8.7 and the definition of  $\text{DSI}^*(G, T, \kappa)$ . To provide a polynomial time algorithm for  $\text{DSI}^*(G, T, \kappa)$ , we suitably modify the algorithm described in the proof of Theorem 8.4. We redefine  $\text{SI}_{\text{SANS}}^*(G, T, u, v)$  and  $\text{SI}_{\text{SAS}}^*(G, T, u)$  in the following manner:

<sup>3</sup>However, this exact construction will not work in the proof of Theorem 8.1 since the entire network needs to fail in that proof.

- For every node  $u' \in \nabla(u)$  and every integer  $0 \leq k \leq \kappa$ ,  $\text{DSI}_{\text{SANS}}^*(G, T, u, u', k)$  is the number of nodes in an optimal solution of  $\text{DUAL-STABILITY}_{T, \Phi, \kappa}$  (or  $\infty$  if there is no feasible solution of  $\text{DUAL-STABILITY}_{T, \Phi, \kappa}$ ) for the subgraph induced by the nodes in  $\Delta(u)$  assuming the following:
  - $u'$  was shocked,
  - $u$  was not shocked,
  - no node in the path  $u' \rightsquigarrow u$  except  $u'$  was shocked, and
  - total number of shocked nodes in  $\Delta(u)$  is exactly  $k$ .
- For every integer  $0 \leq k \leq \kappa$ ,  $\text{DSI}_{\text{SAS}}^*(G, T, u, k)$  is the number of nodes in an optimal solution of  $\text{DUAL-STABILITY}_{T, \Phi, \kappa}$  for the subgraph induced by the nodes in  $\Delta(u)$  (or  $\infty$ , if there is no feasible solution of  $\text{STABILITY}_{T, \Phi}$  under the stated conditions) assuming that the node  $u$  was shocked (and therefore failed), and the number of shocked nodes in  $\Delta(u)$  is exactly  $k$ .

Computing these quantities becomes slightly more computationally involved as shown below.

**Computing  $\text{DSI}_{\text{SAS}}^*(G, T, u, k)$  when  $\deg_{\text{in}}(u) = 0$ :**

$$\text{DSI}_{\text{SAS}}^*(G, T, u, 1) = 1 \text{ and } \text{DSI}_{\text{SAS}}^*(G, T, u, k) = -\infty \text{ for any } k \neq 1.$$

**Computing  $\text{DSI}_{\text{SANS}}^*(G, T, u, u', k)$  when  $\deg_{\text{in}}(u) = 0$ :**

- If  $u \in \text{iz}(u')$  then shocking node  $v$  makes node  $u$  fail. Thus,  $\text{SI}_{\text{SANS}}^*(G, T, u, u', 1) = 1$  and  $\text{SI}_{\text{SANS}}^*(G, T, u, u', k) = -\infty$  for any  $k \neq 1$ .
- Otherwise, node  $u$  does not fail. Thus,  $\text{DSI}_{\text{SANS}}^*(G, T, u, u') = -\infty$ .

**Computing  $\text{DSI}_{\text{SAS}}^*(G, T, u)$  when  $\deg_{\text{in}}(u) > 0$ :** In this case we have

$$\text{DSI}_{\text{SAS}}^*(G, T, u, k) = 1 + \min_{k_1 + k_2 + \dots + k_{\deg_{\text{in}}(u)} = k-1} \left\{ \sum_{i=1}^k \min \left\{ \text{DSI}_{\text{SAS}}^*(G, T, v_i, k_i), \text{DSI}_{\text{SANS}}^*(G, T, v_i, u, k_i) \right\} \right\}$$

**Computing  $\text{DSI}_{\text{SANS}}^*(G, T, u, u', k)$  when  $\deg_{\text{in}}(u) > 0$ :** Since  $u'$  is shocked and  $u$  is not shocked, the following cases arise:

- If  $u \notin \text{iz}(u')$  then  $u$  does not fail. Then,

$$\text{DSI}_{\text{SANS}}^*(G, T, u, u', k) = \min_{k_1 + k_2 + \dots + k_{\deg_{\text{in}}(u)} = k} \left\{ \sum_{i=1}^{\deg_{\text{in}}(u)} \min \left\{ \text{DSI}_{\text{SAS}}^*(G, T, v_i, k_i), \text{SI}_{\text{SANS}}^*(G, T, v_i, u', k_i) \right\} \right\}$$

- Otherwise,  $u \in \text{iz}(u')$ , and therefore  $u$  fails when  $u'$  is shocked. Then,

$$\text{DSI}_{\text{SANS}}^*(G, T, u, u', k) = 1 + \min_{k_1 + k_2 + \dots + k_{\deg_{\text{in}}(u)} = k} \left\{ \sum_{i=1}^{\deg_{\text{in}}(u)} \min \left\{ \text{DSI}_{\text{SAS}}^*(G, T, v_i, k_i), \text{DSI}_{\text{SANS}}^*(G, T, v_i, u', k_i) \right\} \right\}$$

It only remains to show how we compute  $\min_{k_1 + k_2 + \dots + k_{\deg_{\text{in}}(u)} = F} \left\{ \sum_{i=1}^{\deg_{\text{in}}(u)} \min \left\{ \text{DSI}_{\text{SAS}}^*(G, T, v_i, k_i), \text{DSI}_{\text{SANS}}^*(G, T, v_i, u', k_i) \right\} \right\}$  for  $F \in \{k-1, k\}$  in polynomial time. It is easy to cast this problem as an instance of the unbounded integral knapsack problem in the following manner:

- We have  $\deg_{\text{in}}(u)$  objects  $\mathcal{O}_1, \mathcal{O}_2, \dots, \mathcal{O}_{\deg_{\text{in}}(u)}$ , each of *unlimited* supply and each of *weight* 1.
- The *cost* of selecting  $k_i$  objects of the type  $\mathcal{O}_i$  is  $\min \left\{ \text{DSI}_{\text{SAS}}^*(G, T, v_i, k_i), \text{DSI}_{\text{SANS}}^*(G, T, v_i, u', k_i) \right\}$ .
- The *goal* is to select a total of *exactly*  $F$  objects such that the total cost is *minimum*.

The standard pseudo-polynomial time dynamic programming algorithm for Knapsack can be used to solve the above instance in  $O(k \deg_{\text{in}}(u)) = O(n^2)$  time. Thus, the total running time of our algorithm is  $O(n^3)$ .  $\square$

## 11 Our Results on the Dual Stability Index for Heterogeneous Networks

We show in the theorem below that, for  $T = 2$ , the dual stability measure cannot be approximated within a large approximation factor provided a complexity-theoretic assumption is satisfied. To understand this assumption, we recall the following definitions and terminologies from [5].

- A random  $(m, n, d)$  hyper-graph  $H$  is a random hyper-graph of  $n$  nodes,  $m$  hyper-edges each having exactly  $d$  nodes obtained by choosing each hyper-edge independently and uniformly at random. For our purpose, assume that  $d$  is a constant, and  $m \geq n^c$  for some constant  $c > 3$ . Let  $Q: \{0, 1\}^d \mapsto \{0, 1\}$  denote a  $d$ -ary predicate, and let  $\mathcal{F}_{Q, m}$  be a distribution over  $d$ -local functions from  $\{0, 1\}^n$  to  $\{0, 1\}^m$  by defining the random  $d$ -local function  $f_{H, Q}: \{0, 1\}^n \mapsto \{0, 1\}^m$  to be the function whose  $i^{\text{th}}$  output is computed by applying the predicate  $Q$  to the  $d$  inputs that are indexed by the  $i^{\text{th}}$  hyper-edge of  $H$ .
- The  $\kappa$  densest sub-hypergraph problem ( $\text{DS}_\kappa$ ) is defined as follows: given an hyper-graph  $G = (V, F)$  with  $n = |V|$  and  $m = |F|$  such that every hyper-edge contains *exactly*  $d$  nodes and an integer  $\kappa > 0$ , select a subset  $V' \subseteq V$  of exactly  $\kappa$  nodes which maximizes the number of edges induced by the selected nodes, *i.e.*, maximizes  $|\{ \{u_1, u_2, \dots, u_d\} \in F \mid u_1, u_2, \dots, u_d \in V' \}|$ .

The essence of the complexity-theoretic assumption is that if, for a suitable choice of  $Q$ ,  $\mathcal{F}_{Q, m}$  is a collection of one-way functions, then  $\text{DS}_\kappa$  is hard to approximate within a large approximation factor. More precisely, the technical assumption is as follows:

- ( $\star$ ) If  $\mathcal{F}_{Q, m}$  is  $1/o(1/\sqrt{n} \log n)$ -pseudorandom, then for  $\kappa = n^{1-\frac{c-3}{2d}}$  for some constant  $c > 3$  there exists instances  $G = (V, F)$  of  $\text{DS}_\kappa$  with  $m \geq n^c$  such that it is not possible to decide in polynomial time if there is a solution of  $\text{DS}_\kappa$  with at least  $\frac{(1+o(1))m}{n^{\frac{c-3}{2}(1-\frac{1}{d})}}$  edges (the “yes” instance), or if every solution of  $\text{DS}_\kappa$  has at most  $\frac{(1-o(1))m}{n^{\frac{c-3}{2}}}$  edges (the “no” instance).

**Theorem 11.1.** *Under the technical assumption ( $\star$ ) stated above,  $\text{DSI}^*(G, 2, \kappa)$  cannot be approximated with a ratio of  $n^\delta$  for some constant  $\delta > 0$  even if  $G$  is a DAG.*

*Proof.* Given an instance  $G = (V, F)$  of  $\text{DS}_\kappa$  as stated in ( $\star$ ), we construct an instance graph  $\vec{G} = (\vec{V}, \vec{F})$  as follows:

- For every node  $u \in V$ , we have a node  $\vec{u} \in \vec{V}$ , and for every edge  $e = \{u_1, u_2, \dots, u_d\} \in F$ , we have a node  $\vec{e}$  (also denoted by  $\overrightarrow{\{u_1, u_2, \dots, u_d\}}$ ) in  $\vec{V}$ . Thus, the total number of nodes of  $\vec{G}$  is  $|\vec{V}| = m + n$ .
- For every hyper-edge  $e = (u_1, u_2, \dots, u_d) \in F$ , we have  $d$  edges  $(e, u_1), (e, u_2), \dots, (e, u_d) \in \vec{F}$ . We set the weight (share of internal asset) of every edge  $(e, u_i)$  to 2. Thus,  $|I| = 2dm$ .

Let the share of external assets for a node (bank)  $\vec{y} \in \vec{V}$  be denoted by  $E_{\vec{y}}$  (thus,  $\sum_{\vec{y} \in \vec{V}} E_{\vec{y}} = E$ ). We will select the remaining network parameters as follows. For each  $e \in F$ ,  $E_{\vec{e}} = 1.99d$ , and for each  $u \in V$ ,  $E_{\vec{u}} = 0$ . Thus,  $E = 1.99dm$ . Finally, we set  $\Phi = 1$  and  $\gamma = 1/2$ . We prove the following:

**(completeness)** If  $\text{DS}_\kappa$  has a solution with  $\alpha \geq \frac{(1+o(1))m}{n^{\frac{c-3}{2}(1-\frac{1}{d})}}$  hyper-edges then  $\text{DSI}^*(\vec{G}, 2, \kappa) \geq \kappa + \alpha$ .

**(soundness)** If every solution of  $\text{DS}_\kappa$  has at most  $\beta = \frac{(1-o(1))m}{n^{\frac{c-3}{2}}}$  hyper-edges then  $\text{DSI}^*(\vec{G}, 2, \kappa) \leq \kappa + \beta$ .

Note that with  $c = 5$  (and, thus  $m \geq n^5$ ), and sufficiently large  $d$  and  $n$ , we have

$$\frac{\kappa + \alpha}{\kappa + \beta} = \frac{n^{1-\frac{c-3}{2d}} + \frac{(1+o(1))m}{n^{\frac{c-3}{2}(1-\frac{1}{d})}}}{n^{1-\frac{c-3}{2d}} + \frac{(1-o(1))m}{n^{\frac{c-3}{2}}}} = \frac{n^{1-\frac{1}{d}} + \frac{(1+o(1))m}{n^{1-\frac{1}{d}}}}{n^{1-\frac{1}{d}} + \frac{(1-o(1))m}{n}} \geq (1-o(1))n^{1/d}$$

which proves the theorem with  $\delta = 1/d$ .

### Proof of Completeness ( $\text{DS}_\kappa$ has a solution with $\alpha$ hyper-edges)

Let  $V' \subseteq V$  be a solution of  $\text{DS}_\kappa$  with at least  $\alpha$  hyper-edges. We shock all the nodes in  $V_{\text{shock}} = \{\vec{u} \mid u \in V'\}$ . Every shocked node  $\vec{u}$  fails at  $t = 1$  since  $\Phi(b_{\vec{u}} - \iota_{\vec{u}} + E_{\vec{u}}) = 2\deg_{\text{in}}(\vec{u}) > \deg_{\text{in}}(\vec{u}) = \gamma(b_{\vec{u}} + E_{\vec{u}})$ . Now, consider a hyper-edge  $e = (u_1, u_2, \dots, u_d) \in F$  such that  $u_1, u_2, \dots, u_d \in V'$ . Then, the node  $\vec{e}$  fails at  $t = 2$  since

$$\sum_{i=1}^d \frac{\min\{\Phi(b_{\vec{u}_i} - \iota_{\vec{u}_i} + E_{\vec{u}_i}) - \gamma(b_{\vec{u}_i} + E_{\vec{u}_i}), b_{\vec{u}_i}\}}{\deg_{\text{in}}(\vec{u}_i)} = d > 0.995d = \gamma(b_{\vec{e}} + E_{\vec{e}})$$

### Proof of Soundness (every solution of $\text{DS}_\kappa$ has at most $\beta$ hyper-edges)

We will prove the logically equivalent contrapositive of our claim, *i.e.*, we will show that if  $\text{DSI}^*(\vec{G}, 2, \kappa) > \beta + \kappa$  then  $\text{DS}_\kappa$  has a solution of with strictly more than  $\beta$  hyper-edges. First, note that we can assume without loss of generality that, for any hyper-edge  $e \in F$ , the node  $\vec{e}$  is not shocked. Otherwise, if we shock node  $\vec{e}$ , then it does not fail since at  $t = 1$  since  $\Phi(b_{\vec{e}} - \iota_{\vec{e}} + E_{\vec{e}}) = -0.01d < 0.995d = \gamma(b_{\vec{e}} + E_{\vec{e}})$ , and in fact doing so increases its equity to  $1.005d$ . Since the equity of  $\vec{e}$  increased by shocking it, if this node failed in the given solution then it would also fail if it was not shocked. So, we can instead shock a node  $\vec{u}$  that was not shocked in the given solution; such a node must exist since  $\kappa < n$ .

Note that we have already shown in the proof of the completeness part that, for any  $e = (u_1, u_2, \dots, u_d) \in F$ , if the  $d$  nodes  $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_d$  are shocked then  $\vec{e}$  fails at  $t = 2$ . Thus, our proof is complete provided we show that such a node  $\vec{e}$  does *not* fail at  $t = 2$  if *at least* one of the nodes  $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_d$  is *not* shocked. Let  $S \subset \{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_d\}$  be the set of shocked nodes among these  $d$  nodes. Then,  $\vec{e}$  does not fail at  $t = 2$  since

$$\sum_{u_i \in S} \frac{\min\{\Phi(b_{\vec{u}_i} - \iota_{\vec{u}_i} + E_{\vec{u}_i}) - \gamma(b_{\vec{u}_i} + E_{\vec{u}_i}), b_{\vec{u}_i}\}}{\deg_{\text{in}}(\vec{u}_i)} \leq d - 1 \leq 0.995d = \gamma(b_{\vec{e}} + E_{\vec{e}})$$

for all sufficiently large  $d$ . □

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