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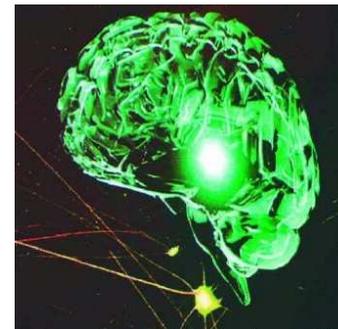
# Control and synchronization in systems coupled via a complex network

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# Synchronization in nonlinear dynamical systems

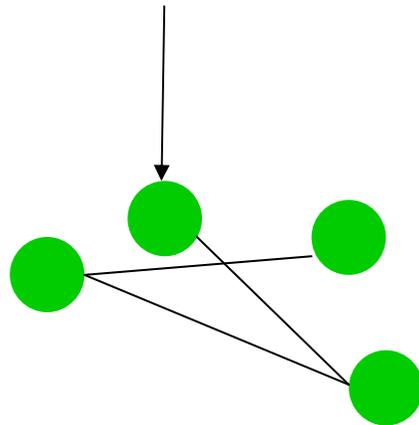
- Synchronization in groups of nonlinear dynamical systems is an active research topic in engineering, biology and systems science.
- Applications:
  - Synchronous firing in groups of fireflies
  - Mobile autonomous robots with limited range of communications
  - Flocking behavior in animals
  - Epileptic behavior



- Coupled system is autonomous, no outside influence to facilitate synchronization

## Synchronization via external control

- Synchronization behavior has also been studied where the synchronization is driven by external forcing (Wang and Chen, 2002).
- Want all systems to synchronize to a desired trajectory  $x(t)$ . Apply  $x(t)$  to a subset of systems. What are the conditions under which the entire systems synchronize to  $x(t)$ ?



## State equations

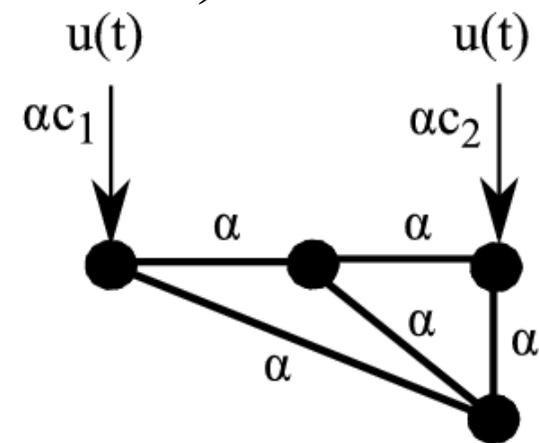
- Unforced network with linear coupling:

$$\frac{dx_i}{dt} = f(x_i, t) - \alpha \sum_j G_{ij} D(t) x_j$$

- Linear control applied to network:

$$\frac{dx_i}{dt} = f(x_i, t) - \alpha \left( \sum_j G_{ij} D(t) x_j + c_i D(t) (x_i - u(t)) \right)$$

- $u(t)$  is the desired trajectory.
- Two sets of parameters describing the coupling:  $\alpha$  describes the coupling between *all* systems,  $c_i$  describes the control coupling.



## Virtual system

- If  $u(t)$  is the trajectory of an uncoupled system, i.e.  $u'(t) = f(u(t), t)$ , then by setting  $x_{n+1}(t) = u(t)$  we get a network of  $n+1$  systems:

$$\frac{dx_i}{dt} = f(x_i, t) - \alpha \sum_j G'_{ij} D(t) x_j$$

- With the new coupling matrix related to  $G$  as:

$$G' = \begin{pmatrix} G + C & -c \\ 0 & 0 \end{pmatrix}$$

- where  $C = \text{diag}(c_1, \dots, c_n)$  and  $c = (c_1, \dots, c_n)^T$ .
- Synchronizing control is then reduced to a problem in synchronization.

## Synchronization

- Various synchronization theorems relate criteria for synchronization with properties of the matrix  $G'$ .
- For a Lyapunov function based approach, we will use the quantity  $\beta_{\min}$ : Under suitable conditions, control is achieved if

$$\beta_{\min} \geq \frac{1}{\alpha}$$

- Precise definition of  $\beta_{\min}$  can be found in the literature.  $\beta_{\min}$  is related to

$$\min_{\lambda} \Re(\lambda(G + C))$$

- In fact, for vertex-balanced networks:

$$\beta_{\min} = \min_{\lambda} \Re(\lambda(G + C)) = \lambda_{\min} \left( \frac{1}{2} (G + G^T) + C \right)$$

## Synchronization condition

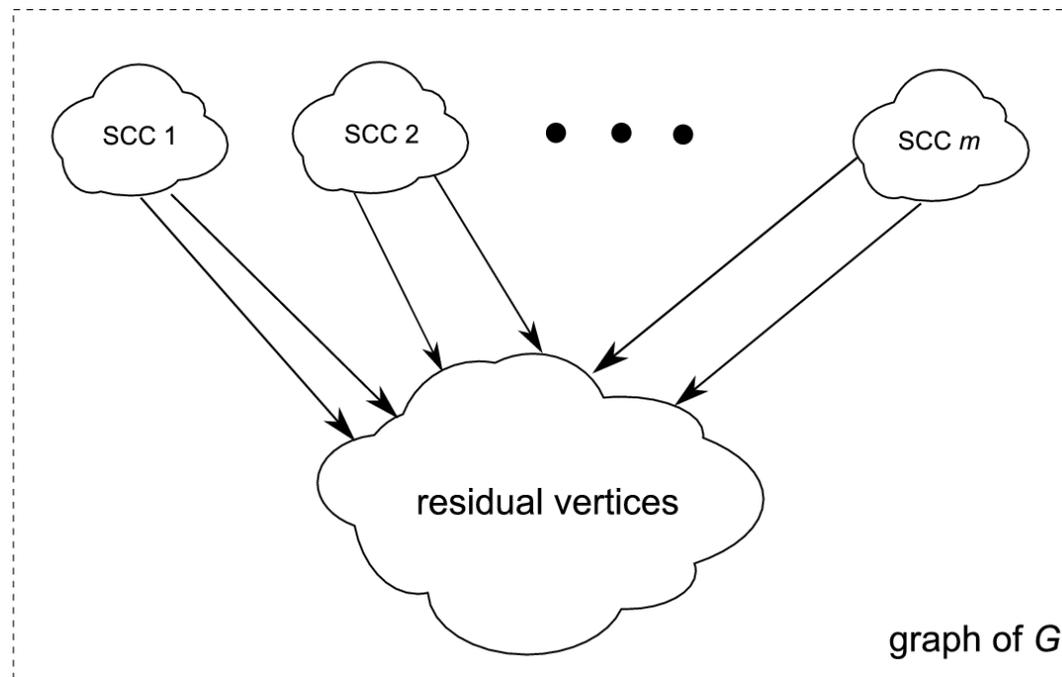
- Recall the synchronization condition:

$$\beta_{\min} \geq \frac{1}{\alpha}$$

- If  $\beta_{\min} > 0$ , then control is achievable for large enough  $\alpha$ .
- Theorem:  $\beta_{\min} > 0$  if and only if there exists a spanning directed forest in the graph of  $G$  such that  $c_i > 0$  whenever the  $i^{\text{th}}$  system is a root of a tree in the forest.
- Thus control can be achieved if and only if forcing is applied to roots of trees in a spanning directed forest of the network.

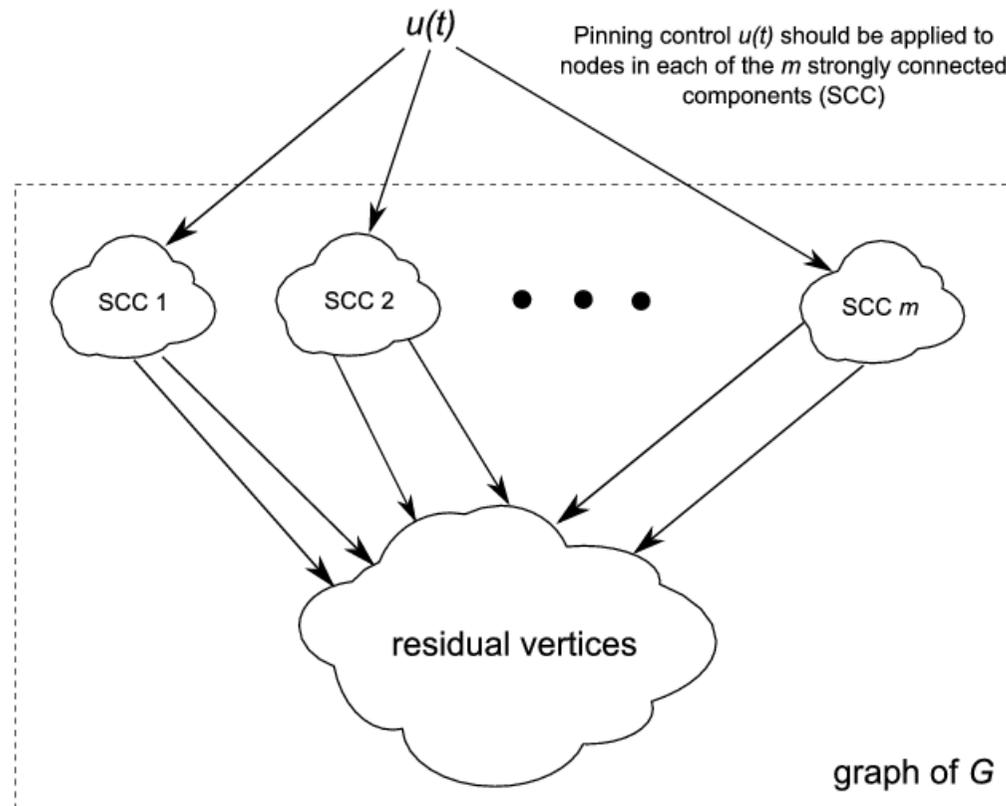
## Spanning directed forest

- By writing the matrix of the graph in Frobenius normal form, we find  $m$  strongly connected components (SCC) which are independent of each other.
- $m$  is the degree of irreducibility of the graph matrix and is the smallest number of trees in a spanning directed forest.
- All spanning forest must have roots in each of these SCCs.



## Applying control at roots of spanning directed forest.

- Synchronizing control can be achieved if and only if forcing is applied to roots of trees in a spanning directed forest of network.
- Thus at least  $m$  systems need to be controlled to achieve synchronizing control.



## How much and where to apply control

- It's clear that how the control is applied influences its effectiveness.
- Let us now consider 2 questions:
  1. How much control to apply?
  2. Where should control be applied to maximize effectiveness?

## How much control to apply?

- Two things to consider here:
  - How many systems to apply control?
  - How much control to apply to each such system?
- Also look at asymptotically behavior as the number of systems  $n$  grows to infinity.
- Consider the case of undirected connected graph, i.e. there is only 1 strongly connected component.

## Upper and lower bounds on $\beta_{\min}$

- For an undirected graph

$$\beta_{\min} = \lambda_{\min}(G + C)$$

- Upper bounds:

$$\lambda_{\min}(G + C) \leq \frac{1}{n} \sum_i c_i$$

$$\lambda_{\min}(G + C) \leq \lambda_{p+1}(G), \forall p < n, p \text{ is number of nonzero elements in } C$$

- Lower bounds:

$$\beta_{\min} \geq \frac{\lambda_2}{\left(1 + \sqrt{1 + \frac{\lambda_2}{\sum_i c_i}}\right)^2} > 0$$

$$\beta_{\min} \geq 2 \min(c, 1) \left(1 - \cos\left(\frac{\pi}{2n+1}\right)\right) > 0$$

## How many systems to apply control?

- Upper bound:

$$\lambda_{\min}(G + C) \leq \lambda_{p+1}(G), \forall p < n, p \text{ is number of systems where control is applied}$$

- This implies that if  $k$  is the largest integer such that

$$\lambda_{k+1}(G) < \frac{1}{\alpha}$$

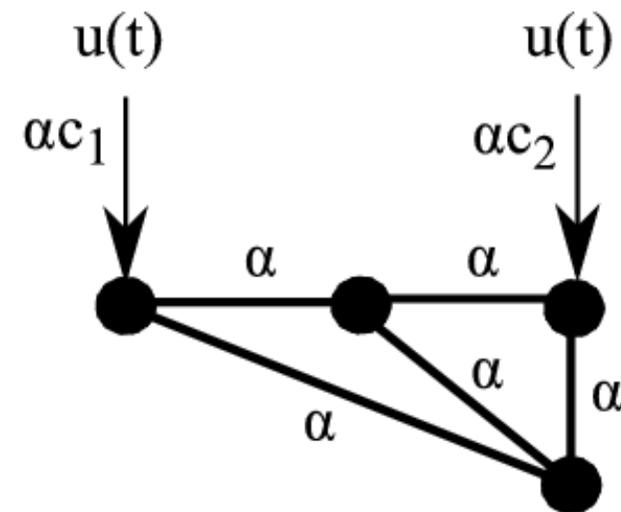
then you need to apply control to at least  $k+1$  systems

## How much control to apply?

- It's clear that making  $\alpha$  and  $c_i$  large enough, we obtained synchronized control.
- The lower bound

$$\beta_{\min} \geq 2 \min(c, 1) \left( 1 - \cos\left(\frac{\pi}{2n+1}\right) \right) > 0$$

provides a measure of how much is needed.

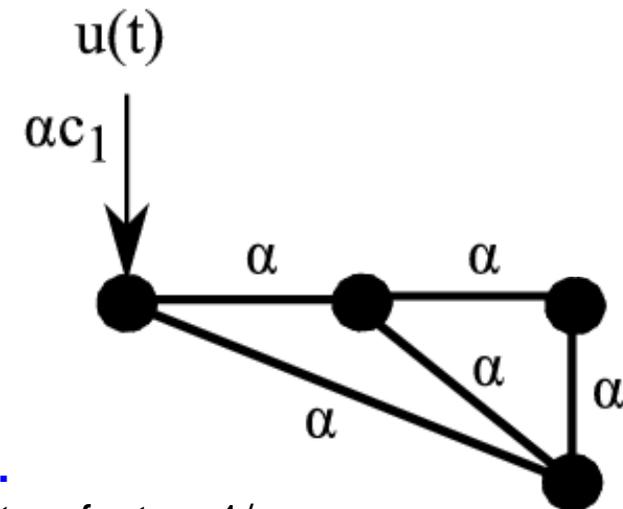


# Asymptotic behavior of $\beta_{\min}$ as function of $n$

- Consider the upper bound

$$\lambda_{\min}(G + C) \leq \frac{1}{n} \sum_i c_i$$

- Assume a single control site ( $c_1 > 0$ ,  $c_j = 0$  for  $j > 1$ ).**
- For fixed control parameter  $c_1$ ,  $\beta_{\min}$  decreases at least as fast as  $1/n$ .
- This means that for bounded control parameter  $c_1$  and  $\alpha$ , control **cannot** be achieved for large  $n$
- In other words, for fixed  $\alpha$ , it is *necessary* for  $c_1$  to grow on the order of  $n$  in order to achieve control.
- Same conclusion apply to vertex-balanced networks, i.e. graphs where the indegree of each vertex is equal to its outdegree.

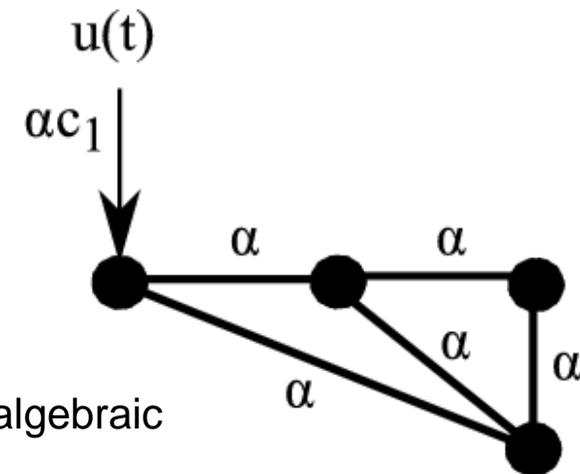


## Asymptotic behavior of $\beta_{\min}$ as function of $n$

- For fixed  $\alpha$ , and  $m=1$ , it is *necessary* for  $c_1$  to grow on the order of  $n$  in order to achieve synchronizing control.
- When is a growth rate of  $o(n)$  *sufficient*?
- Consider the lower bound

$$\beta_{\min} \geq \frac{\lambda_2}{\left(1 + \sqrt{1 + \frac{\lambda_2}{c_1}}\right)^2} > 0$$

- Where  $\lambda_2$  is the second smallest eigenvalue of  $G$ , i.e. the algebraic connectivity of the network.
- For **fully connected** graphs, or **random** graphs,  $\lambda_2$  grows on the order of  $n$  and thus  $\beta_{\min}$  will not vanish for large  $n$  and thus control is achievable for large enough  $c_1$

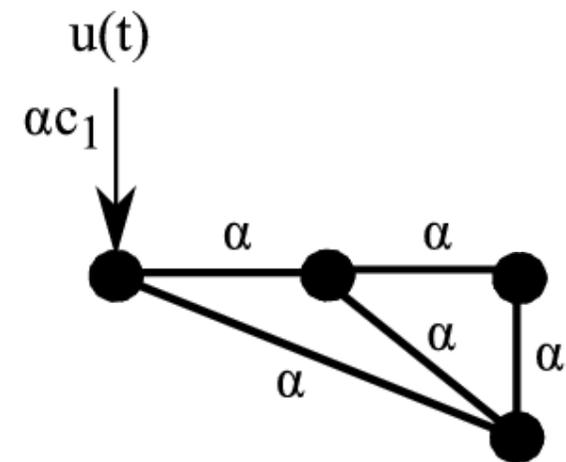


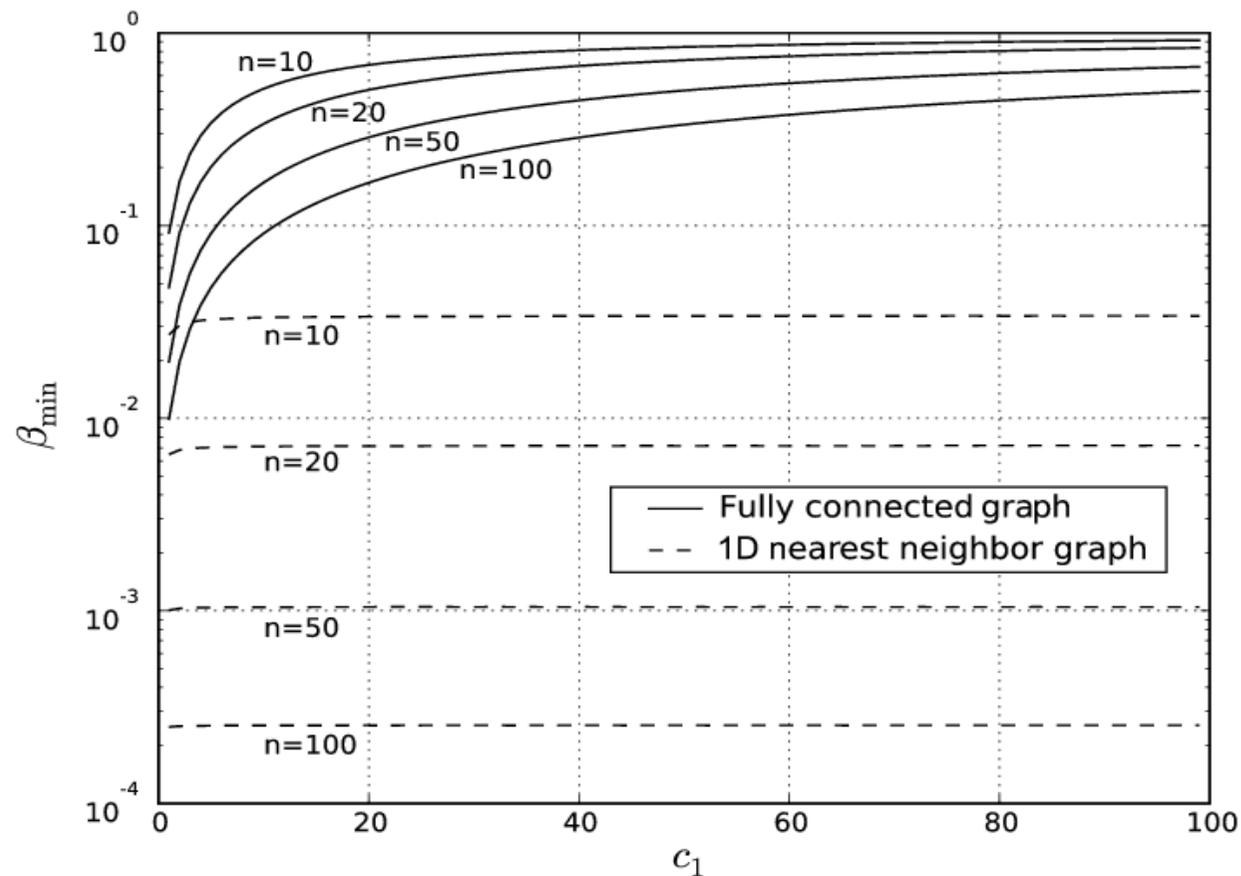
## Locally connected graphs

- Theorem:  $\beta_{\min} \leq \lambda_2(G)$
- Theorem: for locally connected graphs,  $\lambda_2(G) \rightarrow 0$  as  $n \rightarrow \infty$
- Recall the synchronization condition:

$$\beta_{\min} \geq \frac{1}{\alpha}$$

- This means that for locally connected networks, control is **not** achievable for fixed  $\alpha$  as  $n \rightarrow \infty$

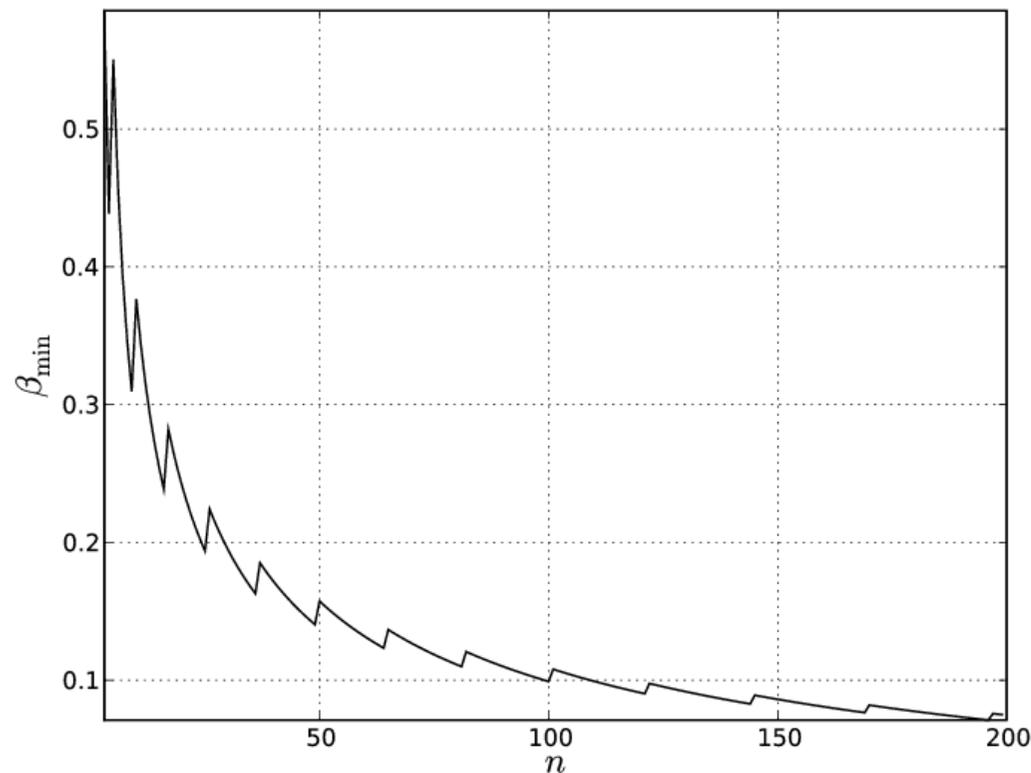




- We see that for the locally connected graph,  $\beta_{\min}$  decreases as  $n$  increases, regardless of  $c_1$ .
- For the fully connected graph, for fixed  $c_1$ ,  $\beta_{\min}$  will decrease as  $n$  decreases, but for large  $c_1$ ,  $\beta_{\min}$  will remain bounded from below.

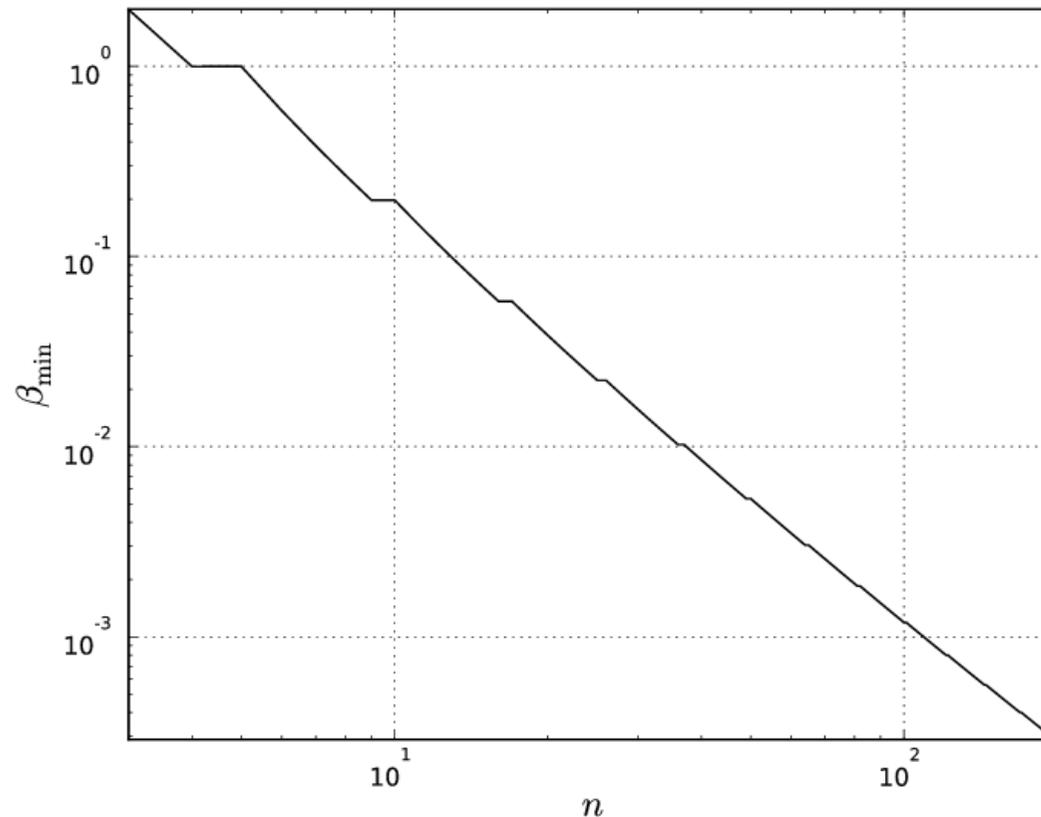
## Small number of control sites

- Similar conclusions as for  $m=1$ .
- If the number of control sites  $m$  grows slower than  $n$ , then  $\alpha$  or  $c_i$  need to grow to maintain synchronization.
- Example: fully connected graph, with  $c_i = \alpha = 1, m = \lfloor \sqrt{n} \rfloor$



## Small number of control sites

- For a bounded parameter  $\alpha$  and a locally connected network, control is not possible if  $m$  grows slower than  $n$ .
- Example: cycle graph with  $c_i = 100n$ ,  $\alpha = 1$ ,  $m = \lfloor \sqrt{n} \rfloor$



## Where to apply control?

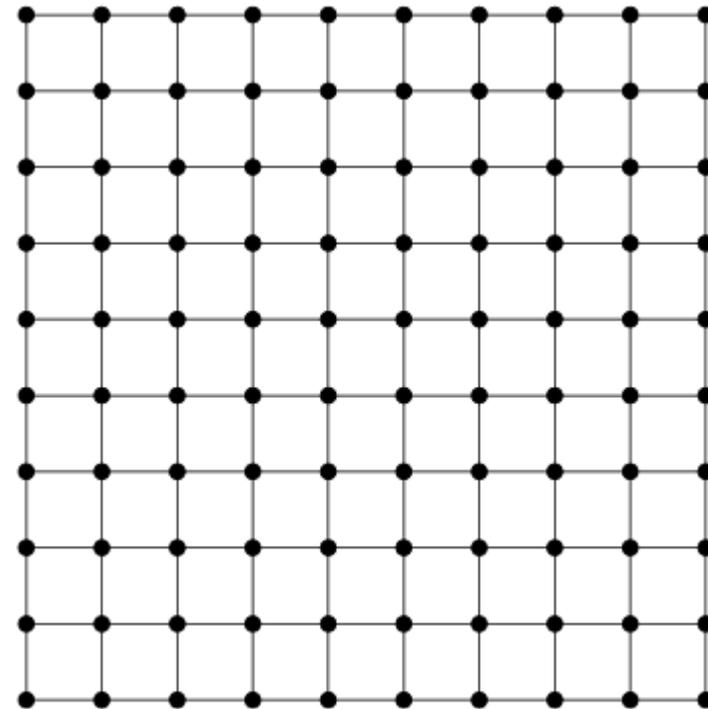
- So far we look at the number of forced systems and the amount of control to apply to each such system.
- The choice of the set of forced system is also important and depends on the underlying network.
- **Where should control be applied to maximize effectiveness?**

## Localization to maximize control effectiveness

- Give a control budget  $\sum_i c_i \leq \gamma$ , how and where should the control signal be applied in order to maximize control effectiveness, i.e. maximize  $\beta_{\min}$ ?
- Recall that  $\beta_{\min} \leq \sum c_i / n$
- Answer is simple if we are allowed to apply control to every system ( $m = n$ ).
- By applying control  $c_i = \gamma/n$  to every system, we get  $C = \gamma/n I$  and  $\beta_{\min} = \gamma/n$ . Thus  $\beta_{\min}$  is maximized subject to the control budget.
- What happens if  $m < n$ ?

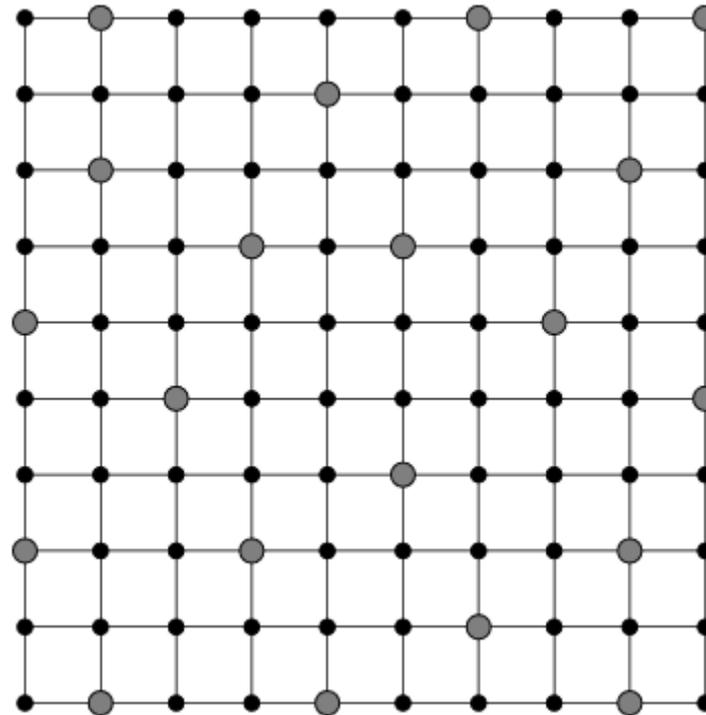
## Example: 2D grid graph

- Consider a 2D grid graph where every node is connected to its 4 nearest neighbours on a rectangular grid.
- Employ the following heuristic to maximize  $\beta_{\min}$ 
  1. Randomly assign  $m$  locations to be the control locations
  2. Randomly move one of these locations if such a move increases  $\beta_{\min}$
  3. Repeat step 2



# Maximize $\beta_{\min}$

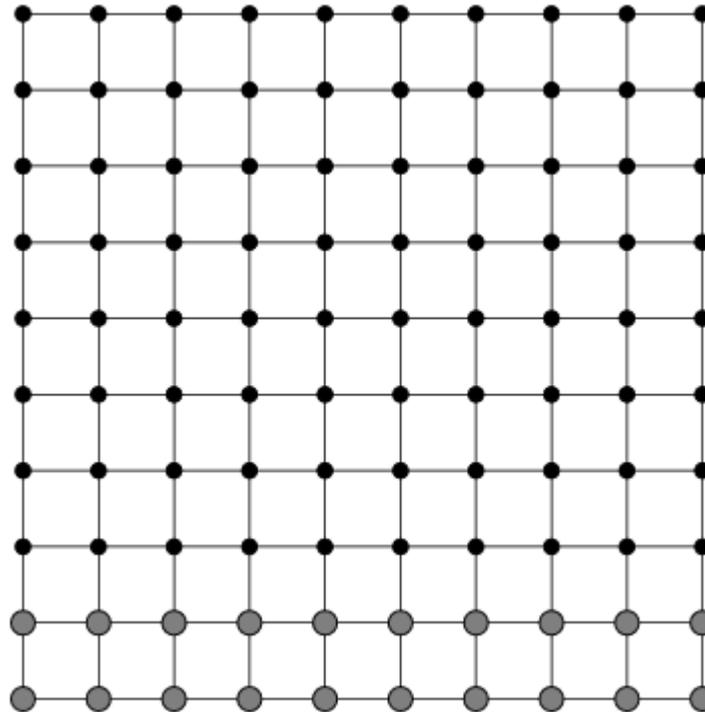
- Apply heuristic for the following parameters:  $n = 100$ ,  $m = 20$ ,  $c = 100$ ,  $\alpha = 1$ .
- The resulting control locations are:



- We note that the control locations are spread out.

# Minimize $\beta_{\min}$

- Repeat same heuristic, but move control locations only if it decreases  $\beta_{\min}$
- Result for  $n = 100$ ,  $m = 20$ ,  $c = 100$ ,  $\alpha = 1$ .



- We note that the control locations are clustered.

## Control locations

- It seems intuitive that control should be applied to locations such that any node is reachable by some control signal via a short path.
- Two characterizations to quantify this idea.
- Let  $P$  be the set of control locations

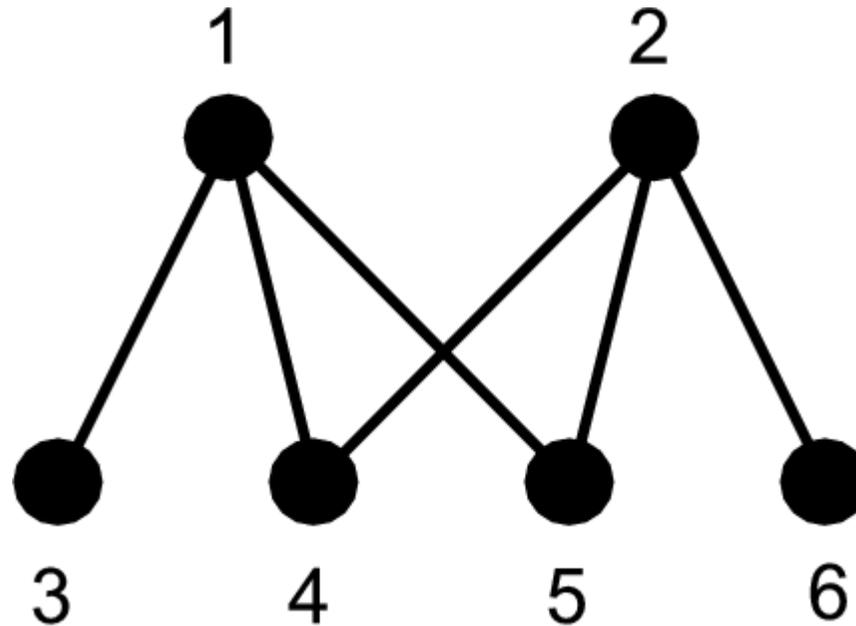
- $D_P = \max_{v \in V} d(v, P)$

- $D_P^a = \frac{\sum_{v \in V \setminus P} d(v, P)}{|V \setminus P|}$

- $D_P$  describes the maximal distance between vertices in  $P$  and any other vertex, where  $D_P^a$  describes the average distance from  $P$  to vertices outside of  $P$ .
- $D_P$  provides a lower bound to  $\beta_{\min}$

$$\beta_{\min} \geq \frac{c}{2} \left( 2 \left( r + \frac{1}{2} (2r)^{-D_P} \right) \right)^{-D_P} > 0$$

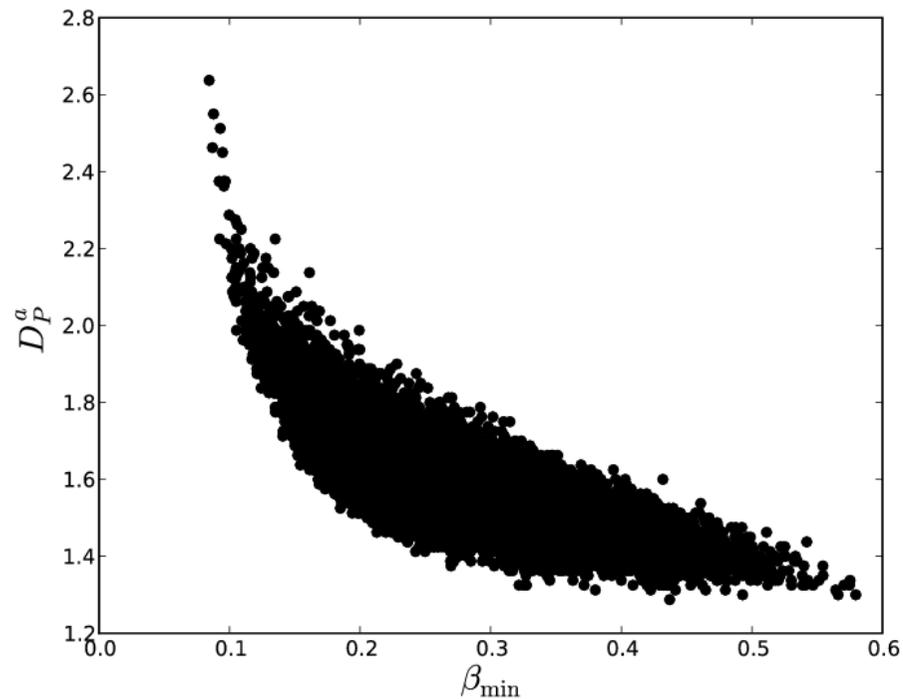
# Minimizing $D_p$ does not maximize $\beta_{\min}$



- Example: single control site ( $m=1$ ),  $c = 10$ .
- Applying control at vertices 4 or 5 minimizes  $D_p$  whereas applying control at vertices 1 or 2 maximizes  $\beta_{\min}$ .

## How about $D_p^a$ ?

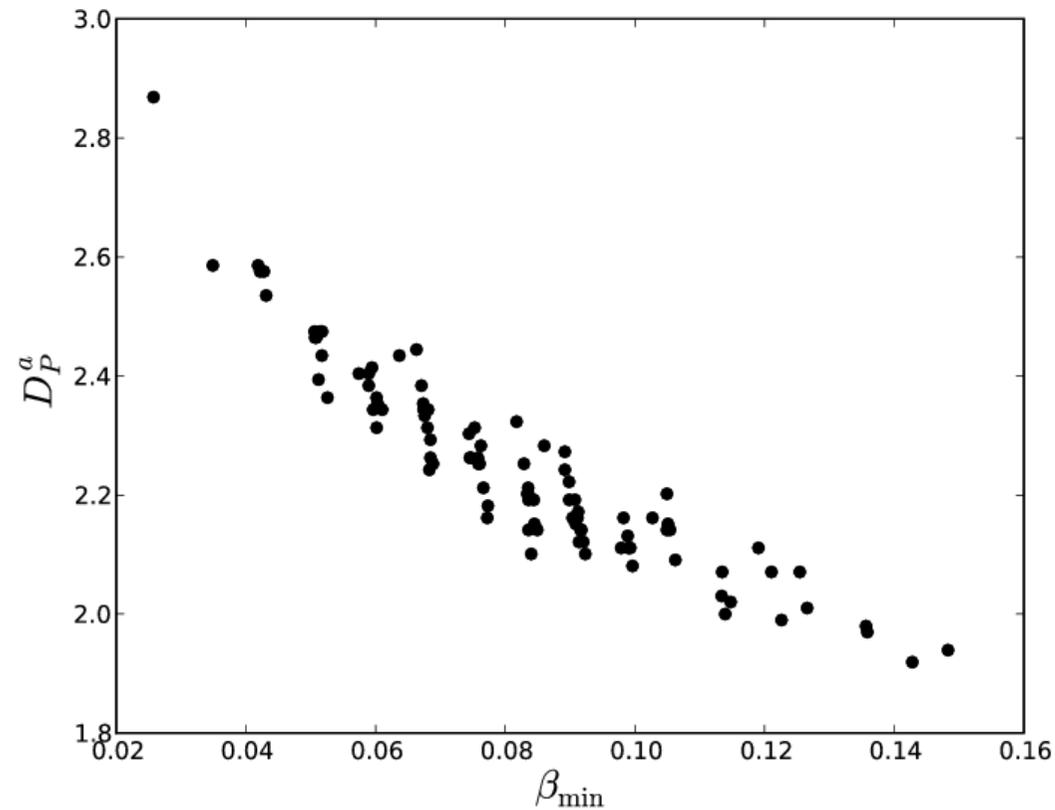
- For a fixed number of control sites  $m$ , is  $\beta_{\min}$  maximized by a configuration that minimize  $D_p^a$ ?
- Experiment 1: 20000 random sets of 20 control locations on the grid graph.



- Upper and lower envelope appears to be convex.

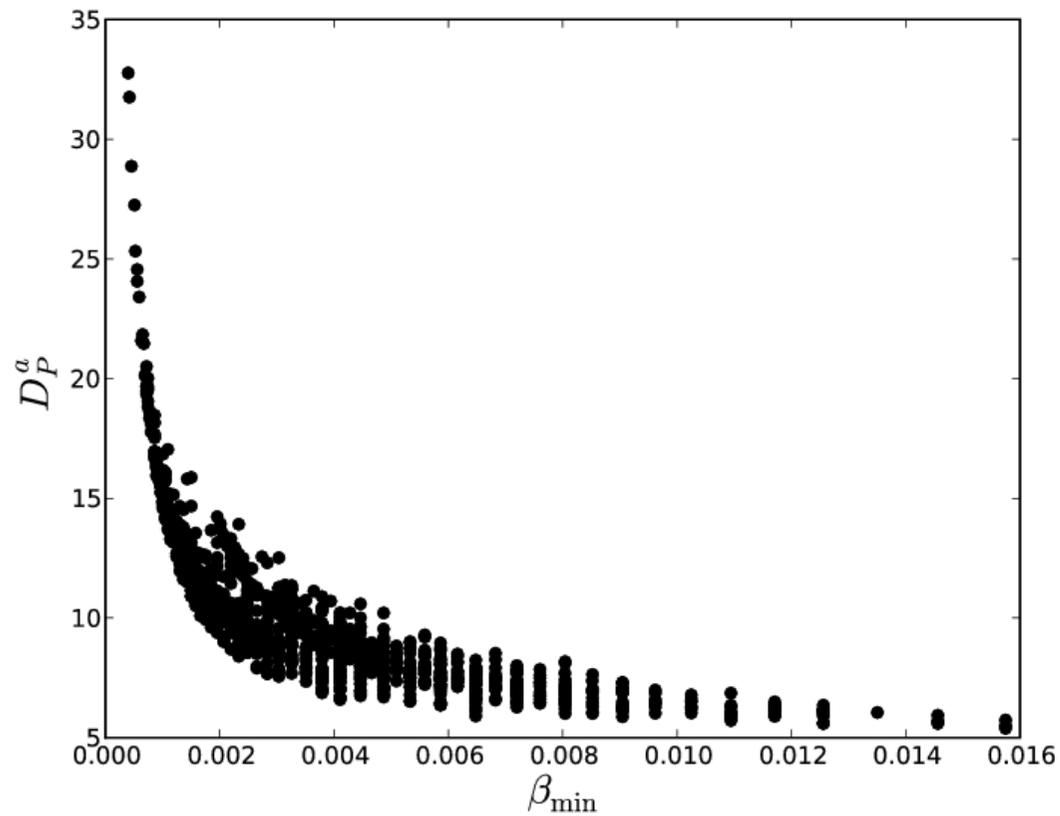
## Experiment 2

- A random graph with 100 vertices and 500 edges,  $c = 100$ . Single control site



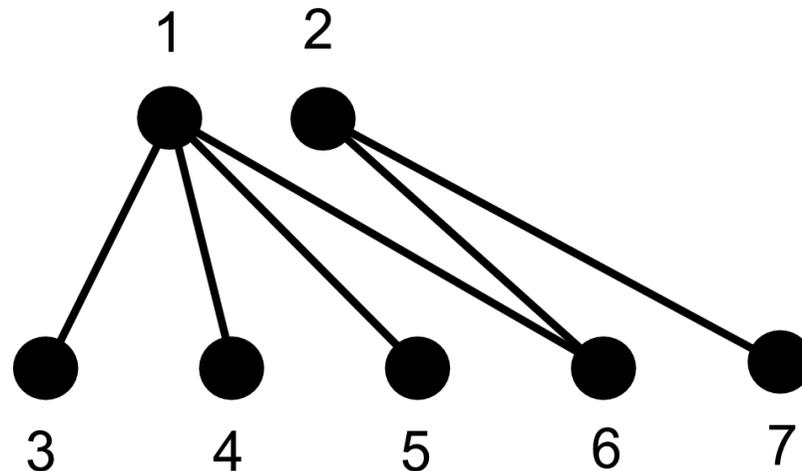
## Experiment 3

- Path graph with 100 vertices and 5 control sites,  $c = 100$ .



## Experiment 4

- Undirected graphs of  $n$  vertices are enumerated
- For each graph, a single control is applied with  $c = 10$ .
- Check whether the location where  $\beta_{\min}$  is maximized corresponds to a minimal  $D_p^a$
- Experiments show that this is true for all graphs of 6 vertices or less, single control site ( $m=1$ ),  $c = 10$ .
- Counterexamples exist for  $n=7$



- $D_p^a$  is minimized at vertex 1, whereas  $\beta_{\min}$  is maximized by applying control at vertex 6.
- $\beta_{\min}$  is second largest at vertex 1.
- Similar results for the question whether the location where  $\beta_{\min}$  is minimized corresponds to a maximal  $D_p^a$  (true for  $n \leq 6$ , false for  $n = 7$ )

## Some provable results

- Theorem: for the cycle graph and large enough forcing, the optimal placement of  $P$  minimizes  $D_P$ , i.e. the forcing systems are spread out on the cycle.
- A similar result can be shown for path graphs.



Thank you very much.  
Any questions?