

Distributed Coordination, Consensus, and Coverage in Networked Dynamic Systems

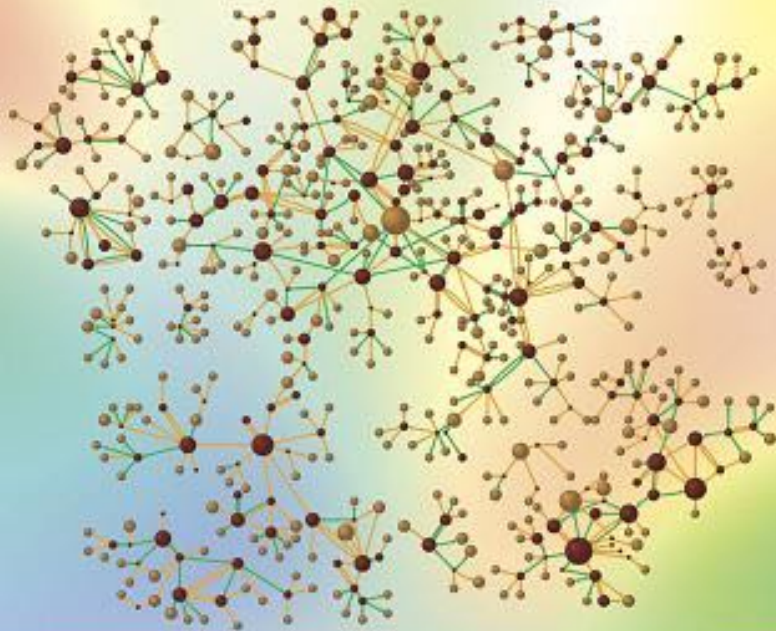
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Electrical & Systems Engineering
and GRASP Laboratory
University of Pennsylvania

Kick-off Meeting, July 28, 2008

ONR MURI: NexGeNetSci



NETWORK SCIENCE



NATIONAL RESEARCH COUNCIL
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**Good news:
Spectacular progress**

ONR MURI: NexGeNetSci



Challenges in the NS report:

- 1. Dynamics, spatial location, and information propagation in networks.*
- 2. Modeling and analysis of very large networks.*
- 3. Design and synthesis of networks.*
- 4. Increasing the level of rigor and mathematical structure.*
- 5. Abstracting common concepts across fields.*
- 6. Better experiments and measurements of network structure.*
- 7. Robustness and security of networks.*

Complexity: dynamics vs. size

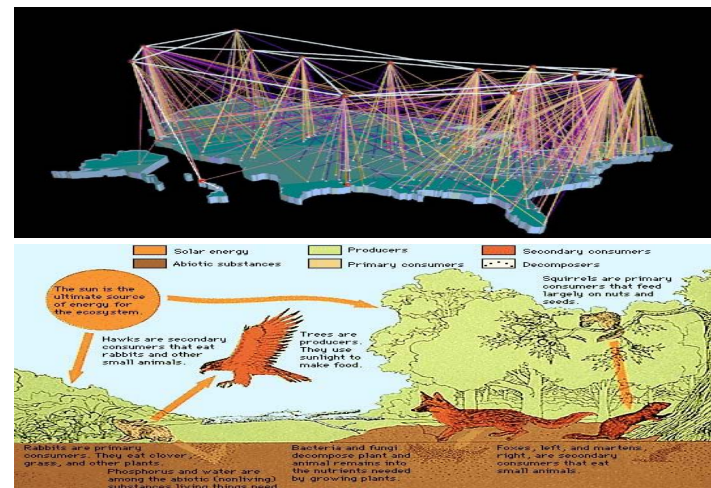
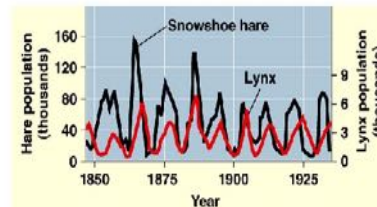
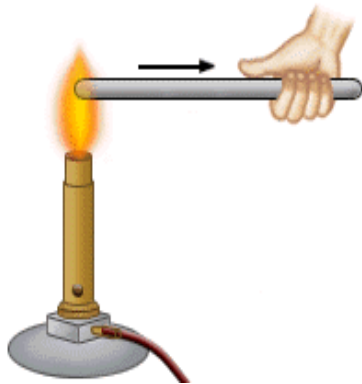
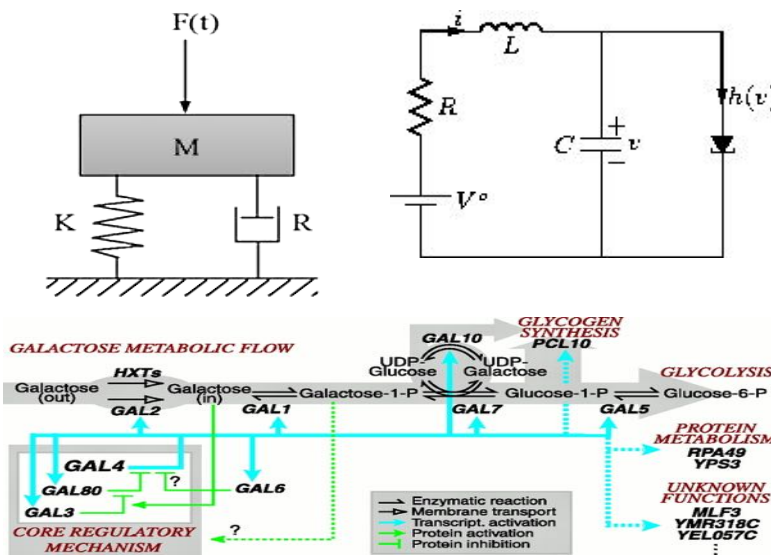
Single agent

System size

Multi-agent

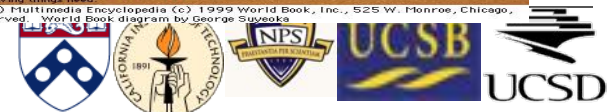
low

State dimensionality

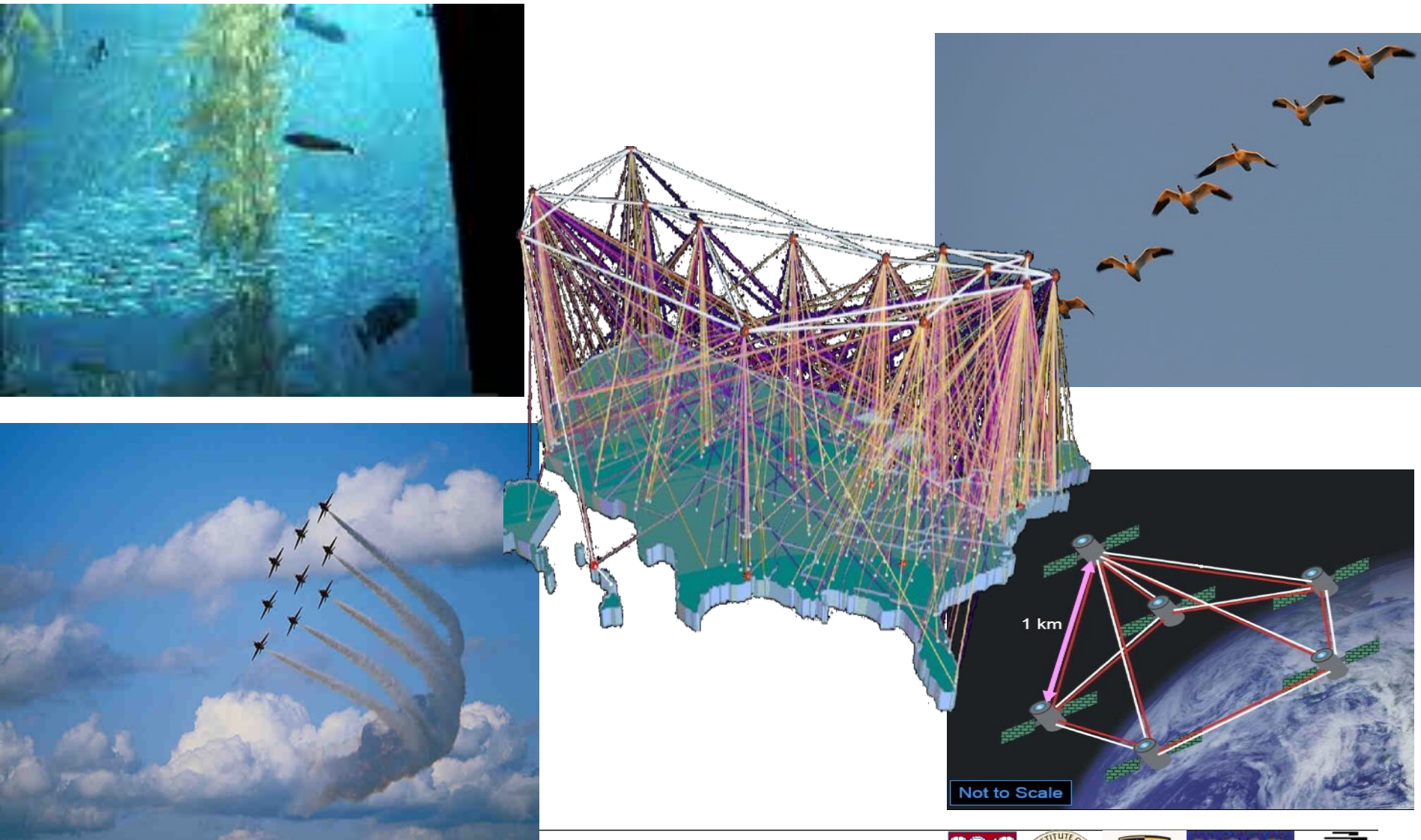


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Substantial Recent Progress



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Case Study: Emergence of Consensus, synchronization, flocking

VOLUME 75, NUMBER 6

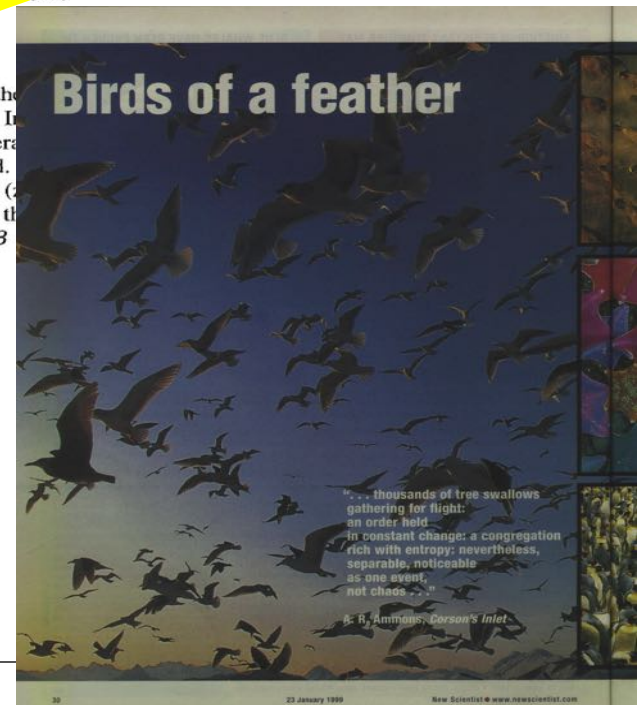
PHYSICAL REVIEW LETTERS

7 AUGUST 1995

Novel Type of Phase Transition in a System of Self-Driven Particles

Tamás Vicsek,^{1,2} András Czirók,¹ Eshel Ben-Jacob,³ Inon Cohen,³ and David Goldschweig,³

¹Department of Atomic Physics, Eötvös University, Budapest, Hungary



Opinion dynamics, crowd control, synchronization and flocking

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An intuitive model

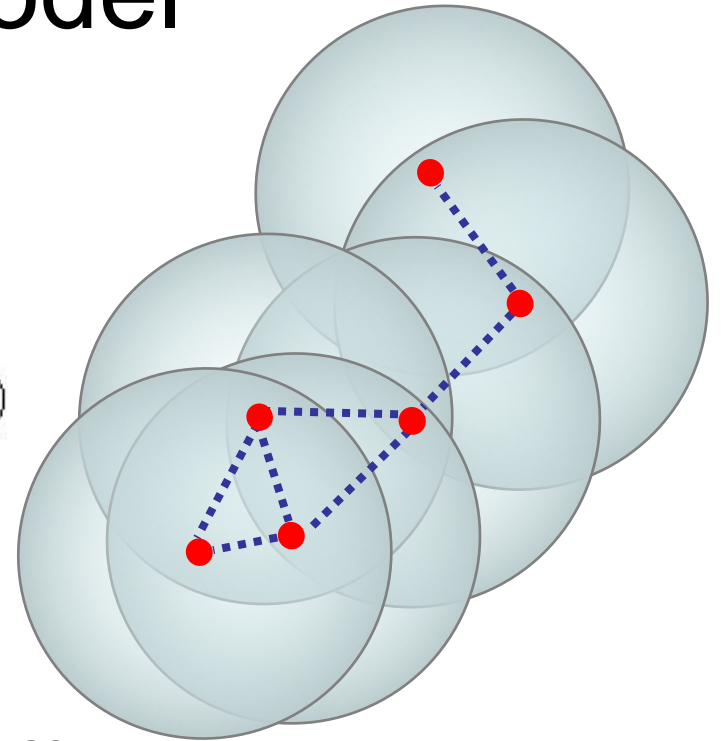
The value of each agent is updated (in discrete time) as a weighted **average** of the value of its neighbors:

$$\theta_i(k+1) = \frac{1}{d_i(k)+1} \left(\sum_{j \in \mathcal{N}_i(k)} w_{ij} \theta_j(k) + w_{ii} \theta_i(k) \right)$$

Neighborhood relation might depend on **actual value, resulting in change in topology**

The neighboring relationship between the agents was represented by a **graph**. $\mathcal{G} = \{\mathcal{V}, \mathcal{E}\}$

The connectivity graph could be **fixed** or **dynamic**.
When do values converge?

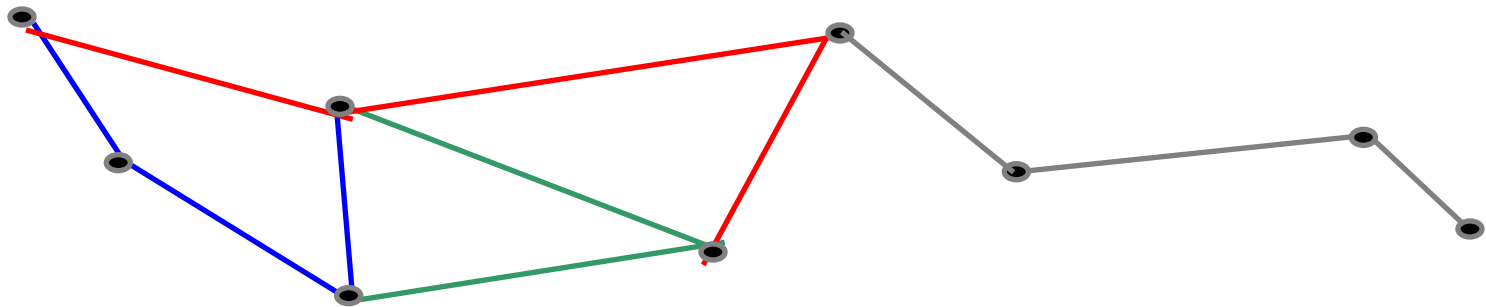


What regimes of topology change are good and which ones are bad?

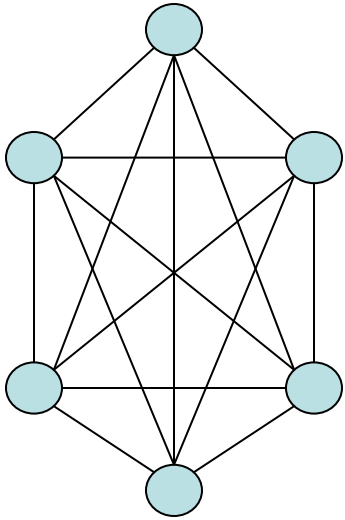
Conditions for reaching consensus

Theorem (Jadbabaie et al. 2003): *If there is a sequence of bounded, non-overlapping time intervals T_k , such that over any interval of length T_k , the network of agents is “jointly connected”, then all agents will reach consensus on their velocity vectors.*

Theorem (Tahbaz Salehi and Jadbabaie '08): when graph process is random, almost sure consensus iff the “average” network reaches deterministic agreement.



Synchronization



$$\frac{d\theta_i}{dt} = \omega_i + \frac{K}{N} \sum_{j=1}^N \sin(\theta_j - \theta_i)$$

N : Number of oscillators

ω_i : Natural frequency of oscillator i , $i = 1, K, N$.

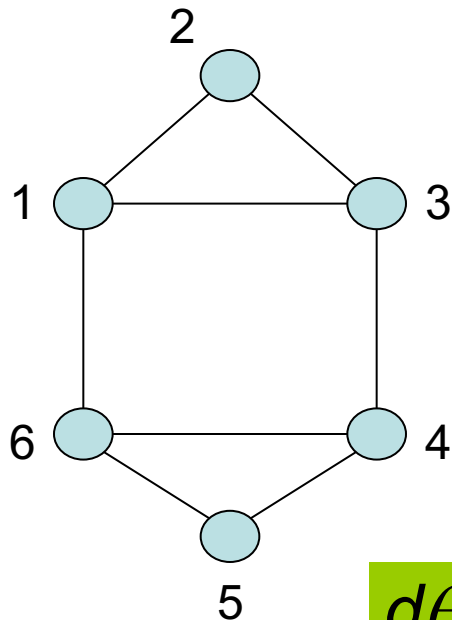
θ_i : Phase of oscillator i , $i = 1, K, N$.

K : Coupling strength

- Model for *pacemaker cells in the heart and nervous system, collective synchronization of pancreatic beta cells, synchronously flashing fire flies, rhythmic applause, gait generation for bipedal robots, ...*
- Benchmark problem in physics
- Not very well understood over arbitrary networks



Kuramoto model & graph topology



$$A = \begin{bmatrix} 0 & 1 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}$$

$$\frac{d\theta_i}{dt} = \omega_i + \frac{K}{N} \sum_{j=1}^N A_{ij} \sin(\theta_j - \theta_i)$$

● B is the incidence matrix of the graph

$$\dot{\theta} = \omega - \frac{K}{N} B \sin(B^T \theta)$$

Kuramoto model, dual decomposition and nonlinear utility minimization

Minimize the misalignment

$$\begin{aligned} \min \quad & \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N A_{ij} (1 - \cos(\theta_i - \theta_j)), \\ \text{s.t.} \quad & \sum_{j=1}^N A_{ij} \sin(\theta_i - \theta_j) = \frac{N\omega_i}{K} \end{aligned}$$

$$L = \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N A_{ij} (1 - \cos(\theta_i - \theta_j)) - \sum_{i=1}^N \sum_{j=1}^N A_{ij} v_i \sin(\theta_i - \theta_j) + \sum_{i=1}^N \frac{Nv_i \omega_i}{K}$$

$$\frac{\partial L}{\partial (\theta_i - \theta_j)} = \sin(\theta_i - \theta_j) - (v_i - v_j) \cos(\theta_i - \theta_j)$$

$$\frac{\partial L}{\partial v_i} = \sum_{j=1}^N A_{ij} \sin(\theta_i - \theta_j) - \frac{N\omega_i}{K}$$

$$\dot{\theta}_i = -\frac{K}{N} \frac{\partial L}{\partial v_i} = \omega_i + \frac{K}{N} \sum_{j=1}^N A_{ij} \frac{(v_i - v_j)}{\sqrt{1 + (v_i - v_j)^2}}$$

Kuramoto model is the just a gradient algorithm for minimization of a global utility which measures misalignment between phasors (exactly like TCP!)

Internet congestion control

S sources (Users):

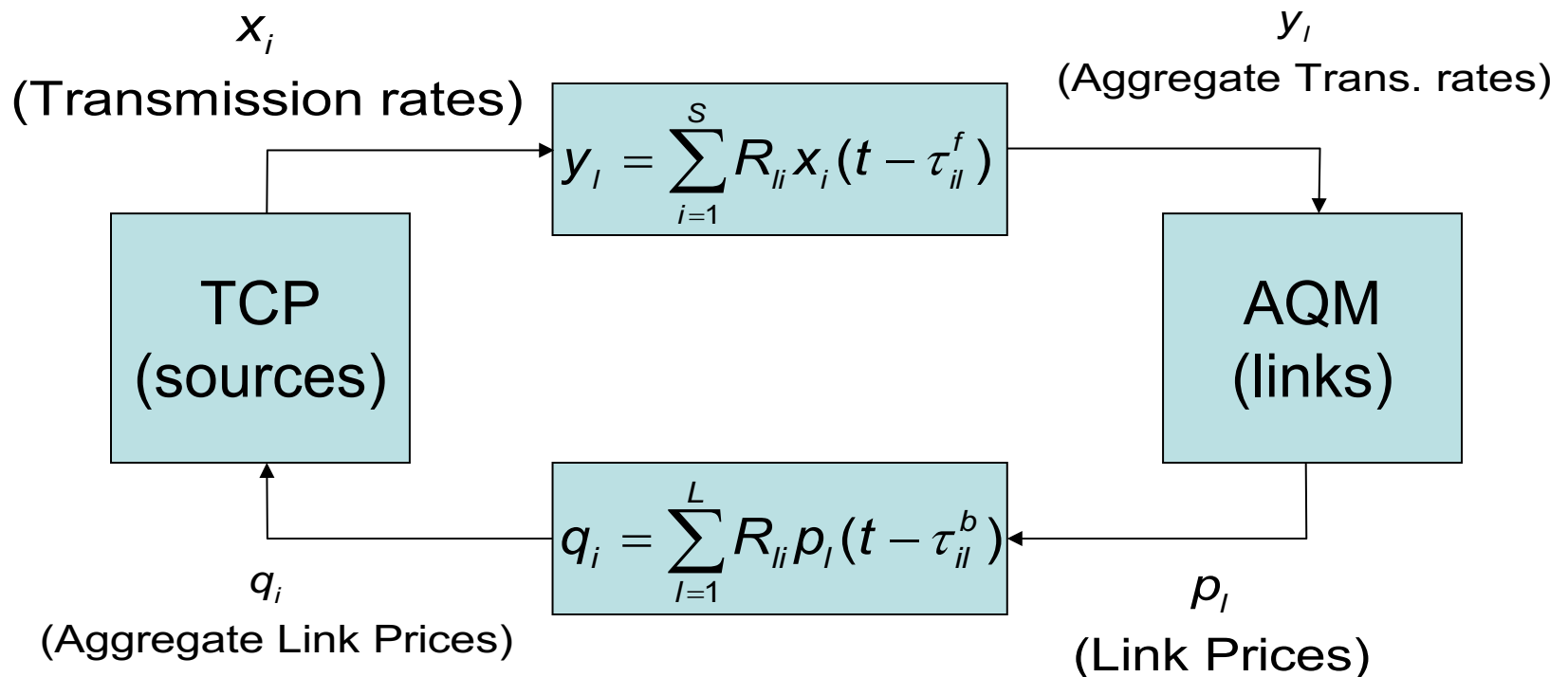
Round Trip time τ_i , $i = 1, \dots, S$

L links (Routers):

Capacity c_l , $l = 1, \dots, L$

R Routing Matrix:

$$R_{li} = \begin{cases} 1 & \text{if source } i \text{ uses link } l \\ 0 & \text{otherwise} \end{cases}$$



Consensus, synchronization, and congestion control

- All 3 involve dynamics over networks
- All can be posed as nonlinear utility maximization /network flow problems
- Common algorithms in all 3 are implementation of a distributed gradient algorithm for solving an implicit optimization
- Proofs scale to arbitrary dimensions, arbitrary topologies, with arbitrary delays
- Success in analysis due to interplay of dynamics with the combinatorial interconnection structure (eg. graphs)

Beyond graphs: Higher order combinatorial specs

Given a set of points: $V = \{v_1, \dots, v_n\}$

k -simplex: An unordered subset of elements of V of size $k + 1$.

Face of a k -simplex: Any size k subset of a simplex

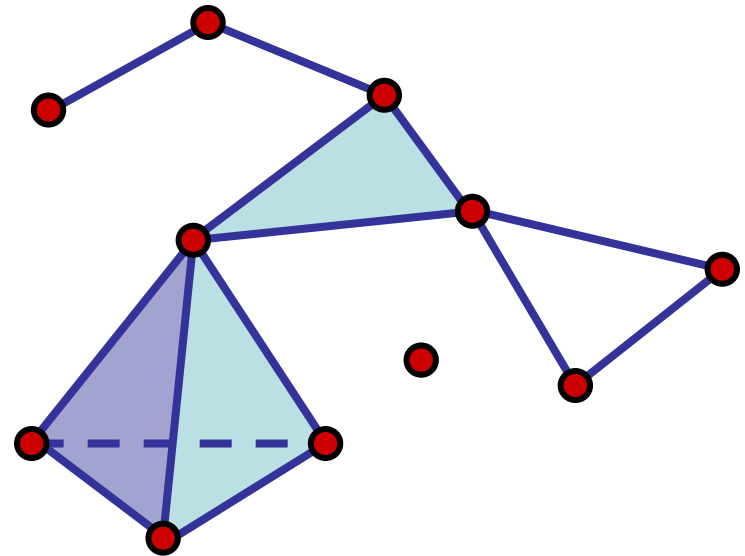
Simplicial Complex:

A collection of simplices closed under the inclusion of faces

Higher order simplices are like higher order terms in a combinatorial “Taylor Expansion”

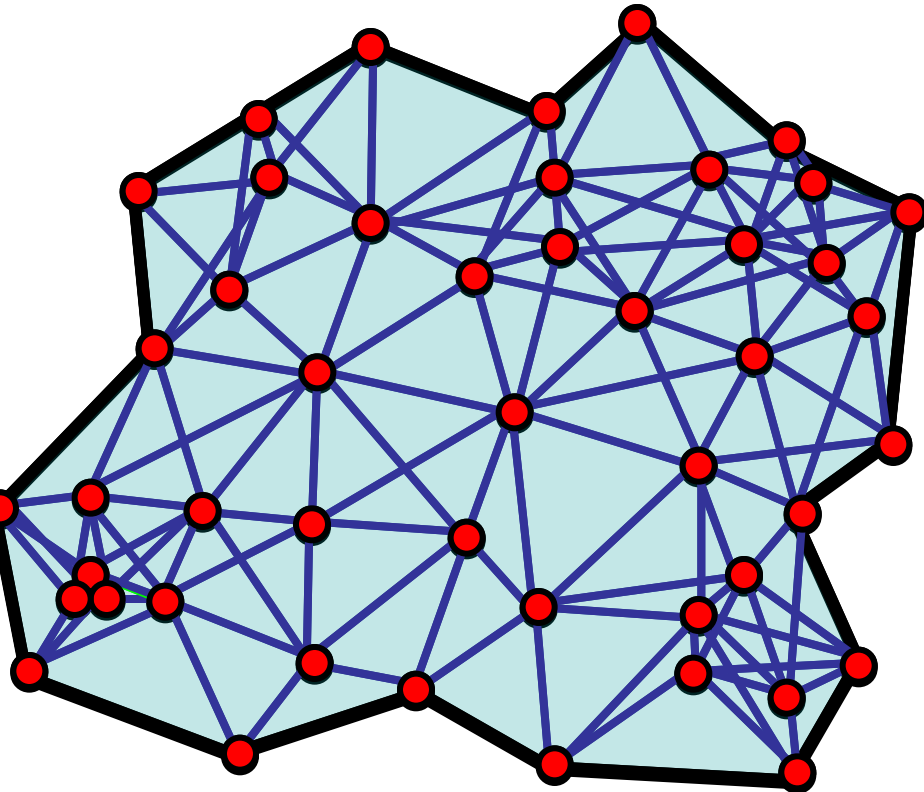
Graphs are just the “linear term”

Lots of tools from graph theory extend to Simplicial complexes

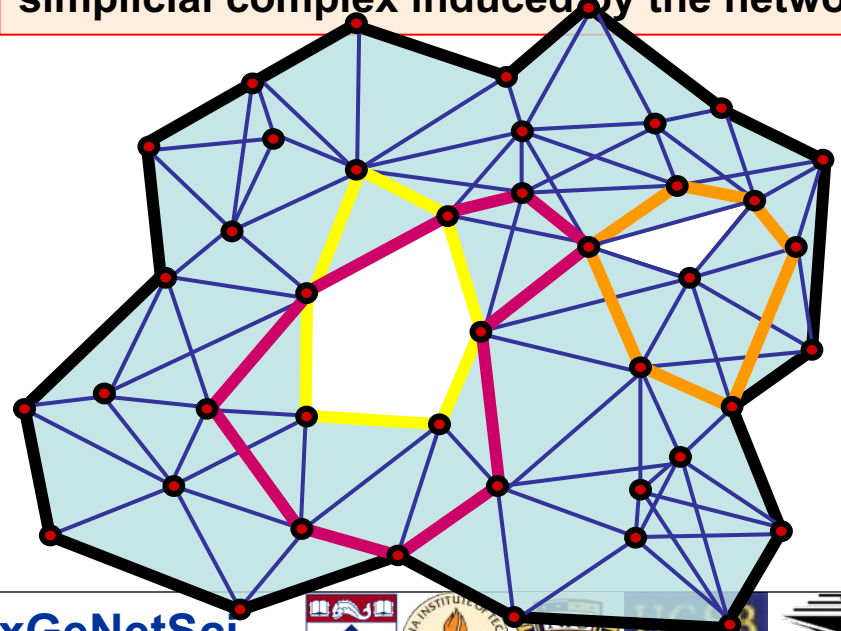


Case study: Blind Coverage

- A bounded domain of interest \mathcal{D} filled with a finite number of sensors
- Each sensor can communicate with other sensors within a distance r
- Any group of sensors in pairwise communication cover their entire convex hull.
- Can we find the minimal set of sensors to cover the entire domain?

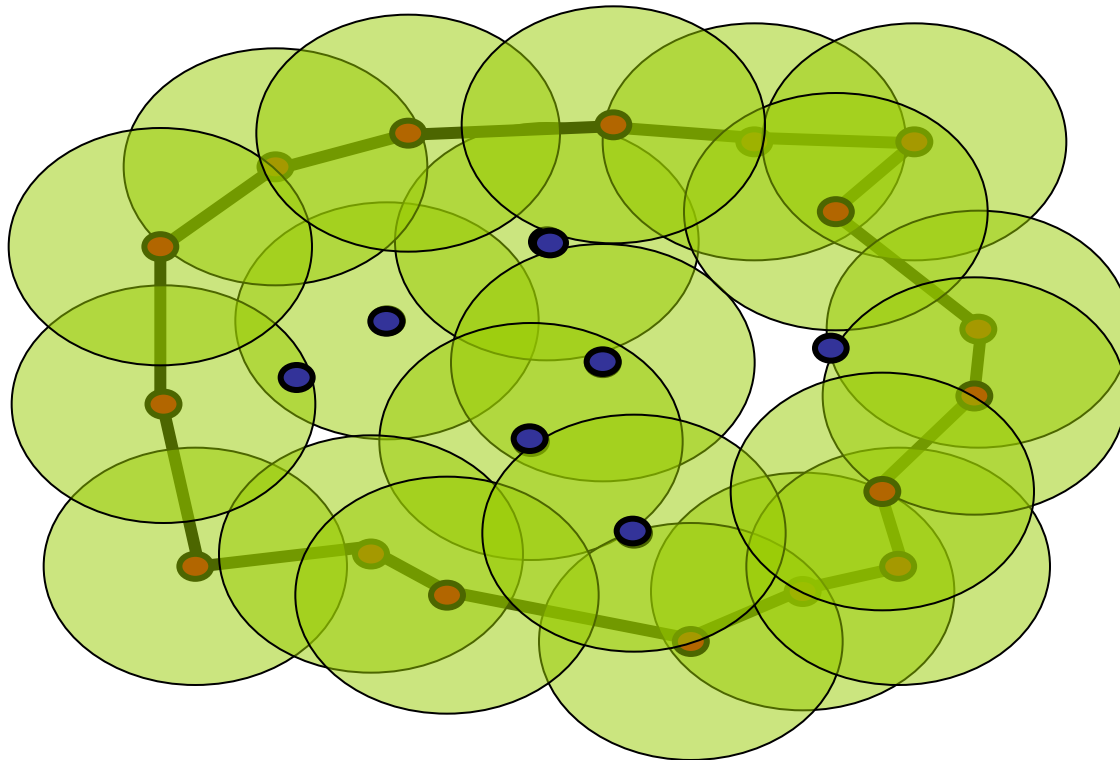


Coverage verification amounts to finding shortest representative cycles in homology classes (e.g non-contractible cycles) of a simplicial complex induced by the network



Example II: Intruder-free coverage

- Given a set of moving sensors, is there an escape strategy for an intruder in \mathcal{D} ?



- Algebraic topology (Hodge theory) + “consensus-like” algorithms give us a decentralized test for intruder detection

Final thoughts

- Need to analyze network dynamics when nodes are dynamical systems, and change in node values changes topology
- “Network Topology” has a precise topological meaning, more than the interconnection structure
- Interplay of algebraic topology, spectral and algebraic graph theory, distributed systems and dynamics and control lead to better understanding of networks
- Need to develop distributed algorithms for computing network topological invariants
- Work presented here partially supported by ARO MURI SWARMS and DARPA DSO StoMP projects