

Global Emergent Behaviors in Clouds of Agents

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Colonies, Herds, Cybernetworks ...

- **Combined distributed interactions lead to complex behaviors**
- **Agents perform collective tasks & achieve collective decisions**
 - **move colony, distribute specialized tasks, defend colony, forage for food ...**
- **Perform these tasks without clear hierarchical structure nor declared leader**
- **Each agent has narrow spatial sensing, inadequate cognitive ability, access to restricted information, limited ability to motivate or compel others to perform**
- **Still beyond the apparent random behavior of individuals in the colony, coordinated complex behavior arises that exhibits order, possibly inducing the colony to operate in synchrony**

Synchrony

- **Synchrony: coupled dynamical oscillators**
 - Heart rhythms: Peskin's model for the cardiac pacemaker
 - Fireflies flashing in synchrony
 - Pulse-coupled biological oscillators

Kuramoto's model

$$\frac{\partial \theta_i}{\partial t} = \omega_i + \frac{K}{N} \sum_{j=1}^N \sin(\theta_j - \theta_i), \quad i = 1 \dots N$$

- **Synchrony: stochastic networks + long term limits or averages**
 - Abstract from individual behaviors
 - Ergodic limits (SLLN, WLLN) (sample averages converging to ensemble averages)
 - Stoch. (queuing) network models
 - Empirical distributions of states of networked systems
 - Renormalize, thermodynamic (mean field) limits
 - Under appropriate conditions, limiting behaviors described by equilibria of ODEs
 - Synchrony may arise

Global Behavior Analysis

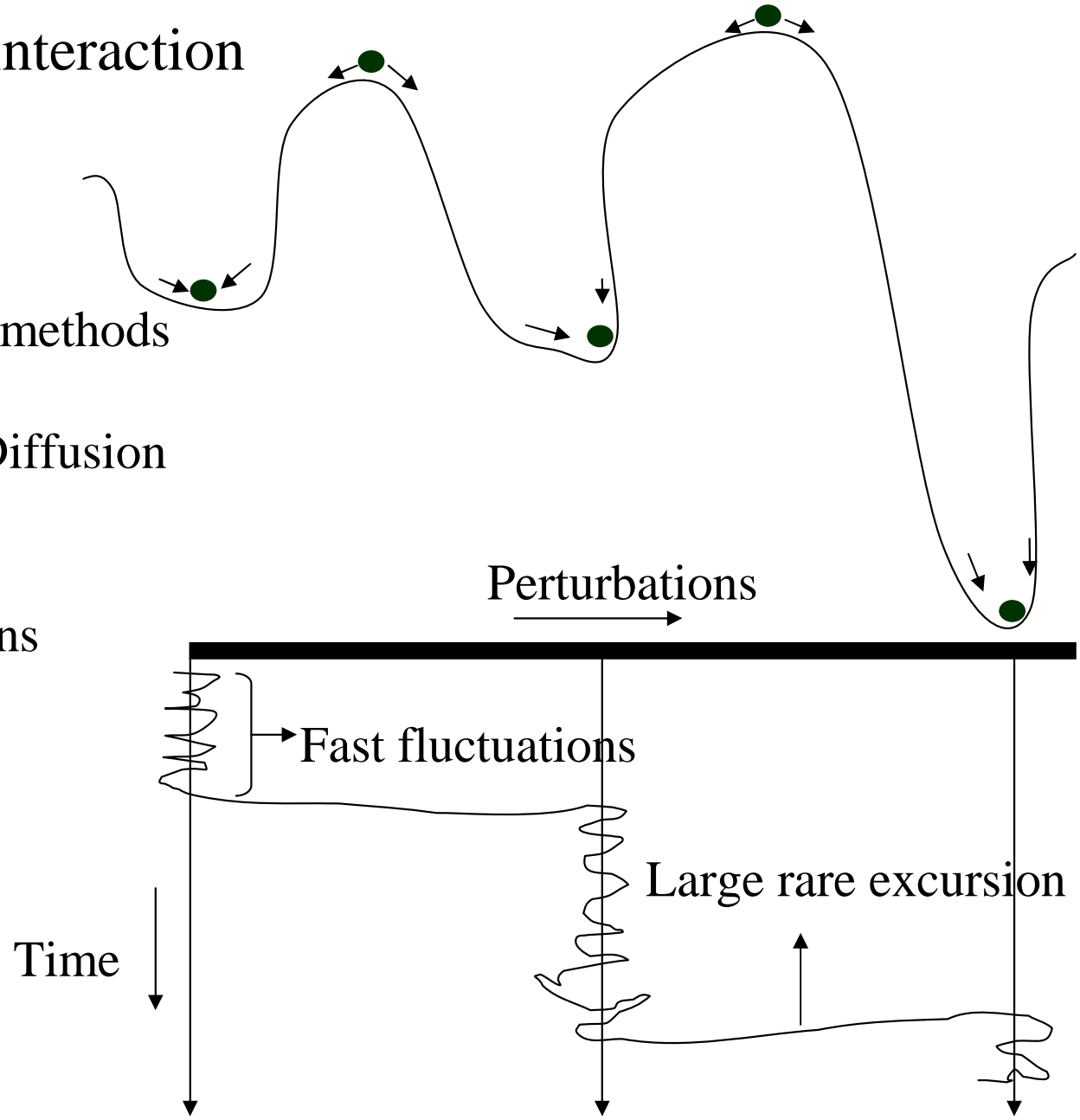
An Event-based Networked Abstraction

Global Behavior: Mathematical Abstraction

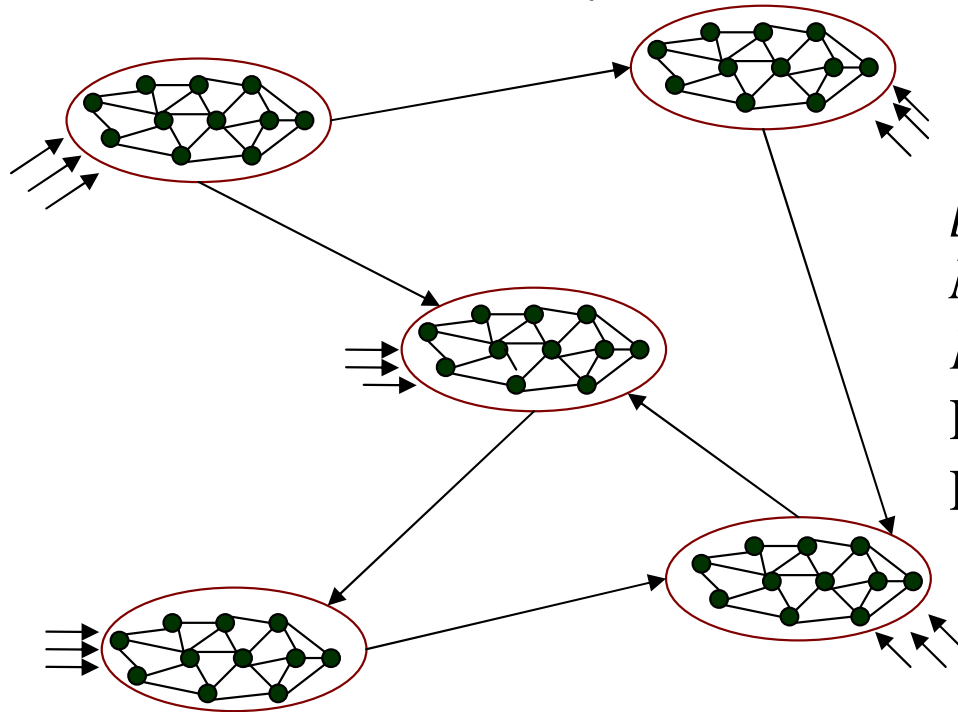
Goal: Model for complex interaction

Mathematical abstraction tools

- Global equilibria- Mean field methods
- Faster scale perturbations – Diffusion (Brownian) Approximations
- Rare Events – Large Deviations



Event-based Network Model



[Antunes, 2006]

M supernodes

K classes local events

Each node C units capacity

Events occur, interact, random times

For each class k event at each agent in each supernode i :

- External occurrences: Poisson, rate λ_{ik}
- Event influence time: exponential, mean $1/\mu_{ik}$
- Intra-(super)node interaction: Poisson, rate γ_{ik}
- Inter-(super)node interaction: Poisson, rate γ^{ij}
- Node with zero capacity rejects other events

Local Configuration Space

- *Local state* of agent n at supernode i : # of \neq event types affecting n

$$X_{in}^N(t) = \mathbf{n} = (n_1, \dots, n_K)$$

- Space of local interactions:

$$\mathcal{X} = \left\{ \mathbf{n} = (n_1, \dots, n_K) \left| \sum_{k=1}^K A_k n_k \leq C_k \right. \right\}$$

- Vector of local node states

$$\mathbf{X}^N(t) = \{X_{in}^N(t), i \leq M, \mathbf{n} \in \mathcal{X}\}$$

- Detailed network dynamics given by $\mathbf{X}^N(t)$, embeds *global* behavior, intractable analysis for large N

Global Configuration Space

- Let $\mathbf{Y}_{i\mathbf{n}}^N(t)$ % nodes in supernode i with configuration \mathbf{n} at time t
- *Global network state* set of *empirical distribution* processes

$$\mathbf{Y}^N(t) = \{\mathbf{Y}_{i\mathbf{n}}^N(t), i \leq M, \mathbf{n} \in \mathcal{X}\} \in \mathbb{R}^{M\mathcal{X}+1}$$

- $\mathbf{Y}^N(t)$ represents emerging global behaviors in terms of empirical averages
- Global concerns: what % nodes failed, not which nodes failed
- Dimension $\mathbf{Y}^N(t)$ (probability simplex-valued process) independent of N
- $\mathbf{Y}^N(t)$ is jump Markov process

Mean-Field Limit

Theorem 1 As $N \rightarrow \infty$, the processes $\{\mathbf{Y}^N(t)\}$ converge weakly (in distribution w.r.t. the Skorokhod topology) to a deterministic trajectory $\{\mathbf{Y}(t)\}$. The limiting (mean-field) trajectory $\mathbf{Y}(t)$ is absolutely continuous and satisfies the ODE,

$$\dot{Y}_{i,\mathbf{n}}(t) = f_{i,\mathbf{n}}(\mathbf{Y}(t)), \quad i \leq M, \mathbf{n} \in \mathcal{X}$$

where the vector fields $f_{i,\mathbf{n}}(\cdot)$ are given by

$$f_{i,\mathbf{n}}(\mathbf{y}) = \sum_{k=1}^K \left[\lambda_{i,k} + \gamma_{i,k} \sum_{\mathbf{m} \in \mathcal{X}} m_k y_{i,\mathbf{m}} \right. \\ \left. + \sum_{j \neq i} \gamma_k^{j,i} \sum_{\mathbf{m} \in \mathcal{X}} m_k y_{j,\mathbf{m}} \right] [y_{i,\mathbf{n}-\mathbf{e}_k} \mathbb{I}_{(n_k \geq 1)} - y_{i,\mathbf{n}} \mathbb{I}_{(\mathbf{n}+\mathbf{e}_k \in \mathcal{X})}] \\ + \sum_{k=1}^K \left[\gamma_{i,k} + \mu_{i,k} + \sum_{j \neq i} \gamma_k^{i,j} \right] [(n_k + 1) y_{i,\mathbf{n}+\mathbf{e}_k} \mathbb{I}_{(\mathbf{n}+\mathbf{e}_k \in \mathcal{X})} - n_k y_{i,\mathbf{n}}]$$

Theorem 1: Discussion

- The proof is based on a martingale representation of the jump Markov processes $\mathbf{Y}^N(t)$.
- The trajectory $\mathbf{Y}(t)$ essentially is a probability measure valued process, taking values in the $M\mathcal{X}$ dimensional probability simplex.
- The dynamical equations (fluid limit) may be viewed intuitively as a coupled collection of the limiting differential equations obtained in [Antunes-2006], the coupling being attributed to the presence of multiple supernodes interacting with each other.
- The fluid limit provides qualitative characterization of the network model, for example, the stability, attractors of the asymptotic dynamical system.
- The fixed points of the ODE, i.e., $\mathbf{f}_{i,\mathbf{n}}(\mathbf{y}^*) = 0$, correspond to the *global equilibria*.
- Such an equilibrium always exists.

Global Equilibria: Existence

Some Notation

For a vector $\rho = (\rho_1, \dots, \rho_K) \in \mathbb{R}^K$, consider the function $v_\rho : \mathcal{X} \mapsto \mathbb{R}_+$,

$$v_\rho(\mathbf{n}) = \frac{1}{Z(\rho)} \prod_{l=1}^K \frac{(\rho_l)^{n_l}}{n_l!}, \quad \mathbf{n} \in \mathcal{X}$$

where $Z(\cdot)$ is the partition function given by

$$Z(\rho) = \sum_{\mathbf{n} \in \mathcal{X}} \prod_{l=1}^K \frac{(\rho_l)^{n_l}}{n_l!}$$

Thus v_ρ represents a family of probability measures on the configuration space \mathcal{X} , being parameterized by the K dimensional vector ρ .

Global Equilibria: Existence (*Contd.*)

Theorem 2:

- Let $\mathbf{y}^* \in \mathbb{R}^{M\mathcal{X}}$ be an equilibrium of the mean field dynamical system. Then, there exists vectors $\rho^1, \dots, \rho^M \in \mathbb{R}^K$, such that, $y_{i,\mathbf{n}}^* = v_{\rho^i}(\mathbf{n})$, for all $i \leq M$ and $\mathbf{n} \in \mathcal{X}$ and the parameterizations $\rho^i = (\rho^1, \dots, \rho^K)$ satisfy the fixed point equations,

$$\rho_k^i = \frac{\frac{\lambda_{i,k}}{\gamma_{i,k}} + \langle I_k, v_{\rho^i} \rangle + \sum_{j \neq i} \frac{\gamma_k^{j,i}}{\gamma_{i,k}} \langle I_k, v_{\rho^j} \rangle}{1 + \frac{\mu_{i,k}}{\gamma_{i,k}} + \sum_{j \neq i} \frac{\gamma_k^{i,j}}{\gamma_{i,k}}}$$

and $\langle I_k, v_{\rho^j} \rangle = \sum_{\mathbf{m} \in \mathcal{X}} m_k v_{\rho^j}(\mathbf{m})$ for all k, j .

- There always exists an equilibrium point of the mean-field ODE.
- If, in addition, $K = 1$, i.e., there is only one class of events, the equilibrium of the mean-field ODE is unique.

Global Equilibria: Existence (*Contd.*)

- Theorem 2 presents a parametric representation of the equilibrium points of the fluid dynamical system.
- Provides a characterization of the equilibrium points of the fluid ODE in terms of solutions of the fixed point equations.
- Shows the existence of at least one equilibrium always.
- The equilibria may not be unique, leading to the possibility of complicated system phenomena like metastability.
- However, if the number of classes $K = 1$, then the assertions conclude the existence of a unique equilibrium of the fluid ODE.

Synchronous Global Equilibria

Definition 1 : A fixed point \mathbf{y}^ of the mean-field dynamical equations, i.e., $V(\mathbf{y}^*) = 0$, is said to be a **synchronous** equilibrium of the asymptotic system if there exists $\mathbf{a} \in \mathbb{R}^{\mathcal{X}}$, such that,*

$$\mathbf{y}^* = [\mathbf{a}^T, \mathbf{a}^T, \dots, \mathbf{a}^T]^T$$

Equivalently, by the parametric representation of equilibria, this implies the existence of $\rho \in \mathbb{R}^K$, such that (v_ρ, \dots, v_ρ) is an equilibrium of the asymptotic dynamical system.

- Synchronous equilibria (if they exist) may lead to eventual (w.r.t. time) synchronous behavior of the supernodes.
- The synchrony between supernodes in terms of the global equilibrium configuration is important in many applications including load balancing etc.
- May also be viewed from a system design perspective, where the goal is to design the inter-supernode coupling parameters $\gamma_k^{i,j}$ under network constraints to lead to synchronous supernode operation.

Synchronous Global Equilibria: Existence Cdtns

Theorem 3: Let the network parameters satisfy the following additional conditions:

(i) *The entry rates of events of different classes are same for all the supernodes, i.e., $\lambda_{i,k} = \lambda_k, \forall i, k$.*

(ii) $\gamma_{i,k} + \sum_{j \neq i} \gamma_k^{j,i} = x_k \geq 0, \forall i, k$

(iii) $\gamma_{i,k} + \mu_{i,k} + \sum_{j \neq i} \gamma_k^{i,j} = y_k, \forall i, k$

Then,

(1) *There exists a synchronous fixed point. In other words, there exists $\rho \in \mathbb{R}_+^K$, such that (v_ρ, \dots, v_ρ) is an equilibrium distribution for the asymptotic dynamical system.*

(2) *If, in addition, $K = 1$ (only one class of events), the above synchronous fixed point is the unique equilibrium of the mean-field limit $\mathbf{Y}(t)$.*

Conclusions

- We introduced a queueing stochastic network model to describe the interactions of clouds of agents and studied its asymptotic behavior after suitable normalization.
- The asymptotic behavior of the network is described by an ordinary differential equation (ODE) whose fixed points correspond to the global equilibria.
- Different global behaviors may emerge as different equilibria of this ODE (metastability). Further study is needed to determine the regions of the parameters spaces that lead to metastability.
- Prelimit behavior (finite N) requires a large deviation analysis of the convergence to the mean-field limit and offers an important direction for future research.