



Self-Adaptive Ad-Hoc/Sensor Network Routing with Attractor-Selection

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Outline of Presentation

- Introduction and motivation
- Adaptive response by attractor-selection
- Application to ad-hoc routing
- Numerical examples
- Conclusion and outlook





Introduction



- Requirements in ad-hoc network routing: scalable, robust, adaptive, fully distributed and self-organizing
 - These features can often be found in biological systems (swarm intelligence)
- Main idea:

randomized selection method of next hop using method inspired from biology





Adaptive Response by Attractor-Selection (ARAS)

- Method from cell biology:
 - reaction to lack of nutrient when no signaling pathway exists from environment to DNA
- Description by stochastic differential equation system
- Attractor:
 - region within which the orbit of dynamical system returns regardless of initial conditions and noise
- Activity:
 - mapping of environment to "goodness" of current system state





General Concept of ARAS





Yuragi

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Mathematical Model of ARAS

 Consider a system with M possible choices m_i, i = 1, ..., M with:

$$\frac{dm_i}{dt} = \frac{syn(\alpha)}{1 + \hat{m}^2 - m_i^2} - deg(\alpha)m_i - \eta_i$$
$$\hat{m} = \max_i \{m_i\}$$

• $syn(\alpha)$ and $deg(\alpha)$ are the rate of synthesis and degradation and are functions of the activity α and η_i is white noise.

$$syn(\alpha) = \alpha [\beta \alpha^{\gamma} + \phi^{*}]$$
 and $deg(\alpha) = \alpha$





Mathematical Model (2)

• Define

$$\varphi(\alpha) = \frac{syn(\alpha)}{deg(\alpha)}$$

 In equilibrium there are M solutions with entries



- $x_i^{(k)} = \begin{cases} \varphi(\alpha) & i = k \quad H \text{ value} \\ \frac{1}{2} \left[\sqrt{4 + \varphi(\alpha)^2} \varphi(\alpha) \right] & i \neq k \quad L \text{ value} \end{cases}$
- H and L merge at $\phi^* = \frac{1}{\sqrt{2}}$





Mapping of Activity

- Activity reflects the "goodness" of the system.
- Initialized with 0 and dynamics follow as

$$\frac{d\alpha}{dt} = \delta(\alpha^* - \alpha)$$

$$\alpha^* = 1 - \left(1 - \frac{distance(s,d)}{path_length}\right) \left(1 - \frac{min_hops}{hops}\right)$$

• Objective:

short path lengths and low hop counts





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Ad-Hoc Routing with ARAS



- MARAS routing decision with ARAS
- Geographic information is used for routing
- At certain intervals, all nodes are probed for their relative distance to the destination and stored in sets: neighbor set N_n , candidate set C_n



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Summary of Algorithm

Node n receives packet destined for d

- if n = d, calculate α^* and update all nodes along the path, process packet.
- determine neighbor and candidate set N_n and C_n
- if C_n is empty, set $A_n = N_n$. Otherwise set $A_n = C_n$
- Perform ARAS on set A_n and forward packet according to hop probabilities.



Example Behavior



- Example scenario with source node (24) sending to destination node (29)
- Instant reaction to failure of node 11 at time 500
- Unnecessary detours removed on activity updates



Simple Numerical Results

- Nodes randomly distributed (2-dimensional homogeneous Poisson Process with rate λ) in unit square
- source node and destination node are the ones with smallest/largest x-coordinates
- Results averaged from 500 simulations with 3000 time steps each
- 95% confidence intervals
- Comparison to Greedy selection of next hop
- Performance metric: success rate of packets





Delivery Rate vs. Node Density



- Low node density or range reduce success rate
- MARAS outperforms Greedy due to stochastic selection





Resilience to Topology Changes



- Nodes in "transit area" switch state with probability q
- Improvement of MARAS over Greedy





Density vs. Radius

 Probability of empty candidate set computed over geometry of intersecting circles





Conclusion and Outlook

- Biologically-inspired method for selecting next hop in ad-hoc networks
- Increased resilience through stochastic routing
- Feedback based (reinforcement learning)
- Future work:
 - More in-depth comparison with other routing methods required
 - Definition of more accurate input/activity mapping



