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A Distributed Control of Virtual Network Topologies by Using Attractor Selection Model

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VNT (Virtual Network Topology) Control

- Wavelength-routed optical network**
 - Establish lightpaths between IP routers via OXCs (optical cross connects)

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 - Is a virtual network that consists of a set of lightpaths and IP routers
 - IP network uses the VNT as its network infrastructure and transports traffic on the VNT

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VNT control

- Constructs VNTs by configuring lightpaths according to its control objective
 - To avoid congestion
 - To transport IP traffic efficiently

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Research Background and Objective

- Problems of existing VNT control methods**
 - Not adaptive
 - They have to collect traffic demand matrix information
 - To retrieve traffic demand matrix information generally takes long time
 - They assume that no big traffic changes happen
- VNT control based on attractor selection[5]**
 - More adaptive to traffic changes than existing methods
 - Utilize adaptability of biological systems (attractor selection)
 - Use only a small amount of information (link utilization information)

But, high computational complexity

We extend our previous method and reduce its computational complexity

[5] Y. Koizumi, T. Miyamura, S. Arakawa, K. Shioto, and M. Murata, "Adaptive virtual network topology control based on attractor selection," IEEE/OSA Journal of Lightwave Technology, vol. 28, pp. 1720-1731, June 2010.

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Attractor Selection

Dynamics of system

$$\frac{dx_i}{dt} = \alpha \left[f(x_1, x_2, \dots, x_k) + \eta \right]$$

activity: α , deterministic term: $f(x_1, x_2, \dots, x_k)$, noise: η

Potential

State of system

Current state

Attractors: a part of the equilibrium points in the phase space

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Attractor Selection

Dynamics of system

$$\frac{dx_i}{dt} = \alpha \left[f(x_1, x_2, \dots, x_k) + \eta \right]$$

activity: α , deterministic term: $f(x_1, x_2, \dots, x_k)$, noise: η

Potential

State of system

Converge to an attractor

Control the system

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Attractor Selection

Dynamics of system

$$\frac{dx_i}{dt} = \alpha \left[f(x_1, x_2, \dots, x_k) + \eta \right]$$

activity: α , deterministic term: $f(x_1, x_2, \dots, x_k)$, noise: η

Potential

State of system

Degrade due to traffic change

Control the system

Big effect of noise

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Attractor Selection

Dynamics of system

$$\frac{dx_i}{dt} = \alpha \left[f(x_1, x_2, \dots, x_k) + \eta \right]$$

activity: α , deterministic term: $f(x_1, x_2, \dots, x_k)$, noise: η

Potential

State of system

Search for a suitable attractor

Control the system

Big effect of noise

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Attractor Selection

Dynamics of system

$$\frac{dx_i}{dt} = \alpha \left[f(x_1, x_2, \dots, x_k) + \eta \right]$$

activity: α , deterministic term: $f(x_1, x_2, \dots, x_k)$, noise: η

Potential

State of system

Find a suitable attractor

Converge to the attractor

Control the system

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VNT Control Based on Attractor Selection

• Dynamics of candidates of lightpaths

$$\frac{dx_i}{dt} = \alpha \cdot \left(\zeta \left(\sum_{j=1}^{n-1} W_{ij} x_j \right) - x_i \right) + \eta$$

activity: α , deterministic term: $\zeta \left(\sum_{j=1}^{n-1} W_{ij} x_j \right) - x_i$, noise: η

Condition: α

Maximum link utilization

- x_i : State of candidate of i -th lightpath (l_i)
 - $0 \leq x_i < 0.5$: tear down l_i
 - $0.5 \leq x_i \leq 1$: establish l_i
- α : Condition of IP network determined by the maximum link utilization
- η : Noise (Gaussian noise)
- $\zeta \left(\sum_{j=1}^{n-1} W_{ij} x_j \right) - x_i$: Deterministic term
 - Define W_{ij} by embedding attractors by using knowledge about Hopfield network[10]
 - Size of W_{ij} : $n^2 + n^2 (= n^4)$
 - The number of candidates of lightpaths: n^2

⇒ The computational complexity: $O(n^4)$

[10] Y. Baram, "Orthogonal patterns in binary neural networks," tech.rep., NASA, Mar. 1988.

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Decentralizing Our Previous Method

- Each node constructs sub-VNT
 - Each node has VNT controller
 - Sub-VNT consists of lightpaths that originate from the node
 - Set of sub-VNTs forms a VNT

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Reduction of Computational Complexity

- Reduce computational complexity $O(n^4)$ to $O(n^2)$
 - New method
 - Each controller manages $(n-1)$ lightpaths
 - Previous method
 - A central controller manages $n(n-1)$ lightpaths
- Adaptability to changes in traffic
 - New method must inherit adaptability to changes in traffic from our previous method
 - However, decentralization may cause selfish behavior of each controller
 - New method may configure unsuitable VNTs for traffic demands

We validate the adaptability of our method by simulation

In 1000-node network, our new method requires only 0.032 sec while our previous method requires about 1000 sec

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Simulation Conditions (1)

- Physical network
 - 50 nodes and 200 optical fibers
 - Each node has 16 transmitters and 16 receivers
- Traffic demand matrices
 - Generate randomly
 - Elements follow a log-normal distribution with a mean of β and with a deviation of σ
- Initial attractors
 - Randomly generated

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Simulation Conditions (2)

- Control objective
 - Construct a VNT with a maximum link utilization of less than the target link utilization (0.5)
- Metrics: Success ratio
 - Ratio of the simulation trials where a method successfully achieves the control objective to total number of simulation trials
- Comparison methods
 - Our previous method
 - Heuristic methods (MLDA and I-MLTDA)
 - Assume that heuristic methods can use traffic demand matrices and then these methods configure sub-optimal VNT
 - Use the heuristic methods as benchmarks

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Success Ratio vs. Deviation of Traffic

Traffic Condition
• Traffic demand matrices with mean of 1 and deviation of σ

Our new method achieves the same adaptability as the previous method

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Success Ratio vs. Total Amount of Traffic

Traffic Condition
• Traffic demand matrices with mean of β and deviation of 1

Our new method achieves the same adaptability as the previous method

Conclusion and Future Work

- **Conclusion**

- We propose a distributed VNT control method based on attractor selection
- Our method reduces computational complexity compared to our previous method
- Our method achieves the same adaptability to traffic changes as our previous method while reducing computational complexity

- **Future work**

- Evaluate our method in large networks (1000-node network)