R

HOPFIELD MODEL FOR SHORTEST PATH COMPUTATION AND ROUTING IN ATM NETWORK

LEE CHEE WENG

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HOPFIELD MODEL FOR SHORTEST PATH COMPUTATION AND ROUTING IN ATM NETWORK

by

LEE CHEE WENG

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I

Table of Contents

| ACKN | OWLEDGN | MENT | I |
|-------------------|----------|--|------|
| TABLE OF CONTENTS | | | |
| LIST | F FIGURE | S | v |
| LIST | F TABLES | 3 | VI |
| ABSTI | RACT | | VIII |
| Силь | TED 1 | | 1 |
| | | Y | |
| 1.1 | | GROUND | |
| 1.2 | | TIVES | |
| 1.3 | | | |
| 1.4 | | ORGANIZATION | |
| 1.5 | | ER SUMMARY | |
| | | | |
| | | | |
| LITE | RATURE F | REVIEW | 6 |
| 2.1 | ATM. | | 6 |
| | 2.1.1 | ATM CELL | 7 |
| | 2.1.2 | VPC AND VCC | 9 |
| | 2.1.3 | ATM SWITCHING OPERATIONS | 9 |
| | 2.1.4 | ATM PROTOCOL ARCHITECTURE AND ATM SERVICE CATEGORIES | 10 |
| | 2.1.5 | ATM TRAFFIC MANAGEMENT | 13 |
| 2.2 | ROUTING | | 14 |
| | 2.2.1 | SOURCE ROUTING AND HOP-BY-HOP ROUTING | 14 |
| | 2.2.2 | ROUTING ALGORITHM TYPES | 15 |
| | 2.2.3 | ROUTING METRICS | 16 |
| | 2.2.4 | DIJKSTRA'S SHORTEST PATH ROUTING ALGORITHM | 17 |
| | 2.2.5 | PNNI AND ROUTING IN ATM NETWORK | 20 |
| | 2.2.6 | TOPOLOGY OF PNNI NETWORK | 21 |
| | 2.2.7 | TOPOLOGY STATE PARAMETER | 22 |

| | 2.2.8 | THE FLOODING MECHANISM | 24 |
|--------|----------|---|----|
| | 2.2.9 | PATH SELECTION | 24 |
| | 2.2.10 | RELATIONSHIP BETWEEN ROUTING, CAC, NETWORK RESOURCE MANAGEMENT AND NETWORK PROVISIONING | 27 |
| | 2.2.11 | DYNAMIC ROUTING | 29 |
| | 2.2.12 | DYNAMIC ROUTING FOR HANDLING LINKS OR NODES FAILURE | 32 |
| 2.3 | ARTIFI | CIAL NEURAL NETWORK | 33 |
| | 2.3.1 | BIOLOGICAL NEURON | 34 |
| | 2.3.2 | ARTIFICIAL NEURONS | 36 |
| | 2.3.3 | BASIC NEURON (McCulloch-Pitts) Model | 37 |
| | 2.3.4 | THE HOPFIELD MODEL OF ARTIFICIAL NEURON | 39 |
| | 2.3.5 | ARCHITECTURE OF ARTIFICIAL NEURAL NETWORKS | 40 |
| | 2.3.6 | HOPFIELD NEURAL NETWORK | 42 |
| | 2.3.7 | BOLTZMANN MACHINE AND SIMULATED ANNEALING | 44 |
| | 2.3.8 | MEAN-FIELD ANNEALING ALGORITHM | 47 |
| | 2.3.9 | HOPFIELD NETWORK AND ITS MODIFICATION FOR COMBINATORIAL OPTIMIZATION | 48 |
| | 2.3.10 | USING HOPFIELD MODEL FOR SHORTEST PATH ROUTING IN COMPUTER NETWORK | 50 |
| 2.4 | Снарт | ER SUMMARY | 58 |
| Снар | TER 3 | | 59 |
| | | Υ | |
| 3.1 | | RUCTING A NEURAL NETWORK FOR SOLVING COP | |
| 3.2 | | /SIS | |
| 3.3 | | EPRESENTATION OF THE SHORTEST PATH ROUTING PROBLEM | |
| 3.4 | | OPFIELD NETWORK | |
| | 3.4.1 | HOPFIELD MODEL ALGORITHM FOR SOLVING SHORTEST PATH ROUTING | |
| 3.5 | | OLTZMANN MACHINE | |
| - *= | 3.5.1 | BOLTZMANN MACHINE ALGORITHM FOR SOLVING SHORTEST PATH ROUTI | |
| 3.6 CF | HAPTER S | UMMARY | |

| CHAPTER 4SIMULATION MODEL AND EXPERIMENT RESULTS | | 70 |
|--|--|----|
| | | 70 |
| 4.1 | THE 5-NODE COMPUTER NETWORK | 70 |
| 4.2 | SIMULATION MODEL | |
| | 4.2.1 SIMULATION MODEL FOR HOPFIELD NEURAL NETWORK | 72 |
| | 4.2.2 SIMULATION MODEL FOR BOLTZMANN MACHINE | 77 |
| 4.3 | CHAPTER SUMMARY | 78 |
| Снав | PTER 5 | 79 |
| Disc | USSION AND CONCLUSION | 79 |
| 5.1 | Conclusions | |
| 5.2 | FURTHER STUDY | 81 |
| REFE | RENCES | 82 |

List of Figures

| Figure 2.1 UNI and NNI Signaling | 7 |
|---|----|
| Figure 2.2 Basic format of the UNI and NNI | 8 |
| Figure 2.3 VCC and VPC concepts | 9 |
| Figure 2.4 Virtual Circuit and Virtual Path Switching. | 10 |
| Figure 2.5 ATM protocol architecture | 11 |
| Figure 2.6 Service class relationship. | 12 |
| Figure 2.8 Switching system architectural reference model for PNNI | 21 |
| Figure 2.9 The PNNI network hierarchy model [3] | 22 |
| Figure 2.10 Operation of crankback [3] | 25 |
| Figure 2.11 DTL processing in connection setup [3] | 26 |
| Figure 2.12 Network provisioning, network resources management, CAC and routing | |
| relationship [11] | 27 |
| Figure 2.13 Layer architecture [11] | 28 |
| Figure 2.14 A simple neuron | 34 |
| Figure 2.15 McCulloch-Pitts Model of artificial neuron | |
| Figure 2.16 Electronic model of the basic neuron cell | 39 |
| Figure 2.17 Hopfield model of artificial neuron | 40 |
| Figure 2.18 Continuous Hopfield Network | 43 |
| Figure 2.19 A 5-node network | 51 |
| Figure 2.20 A 5-node network with pseudo link | 52 |
| Figure 3.1 Boltzmann machine architecture | 66 |
| Figure 4.1 A 5-node computer network | 70 |
| | |

List of Tables

| Table 2.1 | Field and description of the ATM cell | 8 |
|-----------|---|----|
| Table 2.2 | Topology State Parameters | 23 |
| Table 2.3 | Cost matrix of a 5-node network. | 50 |
| Table 2.4 | Rauch and Winarske's representation of the shortest path | 51 |
| Table 2.5 | Thomopoulus proposed representation of the shortest path | 52 |
| | Representation of the shortest path | |
| Table 4.1 | Link cost matrix | 73 |
| Table 4.2 | Simulation result 1 of HOPFIELD2 | 74 |
| Table 4.3 | Simulation result 2 of HOPFIELD2. | 74 |
| Table 4.4 | Simulation result 3 of HOPFIELD2. | 75 |
| Table 4.5 | Simulation result 4 of HOPFIELD2 | 75 |
| | Simulation result 5 of HOPFIELD2 | 76 |
| Table 4.7 | Simulation result 6 of HOPFIELD2 | 76 |
| Table 4.8 | Simulation result of the BOLTZMANN1 machine | 77 |
| Table 5.1 | Comparison of simulation results between HOPFIELD1 and BOLTZMANN1 | 80 |

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ABSTRACT

This research focuses on the application of Hopfield neural network and Boltzmann machine in solving the shortest path routing problem in the ATM network environment. Hopfield neural network and Boltzmann machine are two types of neural network which are commonly used for solving optimization problem such as the shortest path routing problem.

The objectives of this research are to construct a Hopfield neural network and a Boltzmann machine for solving the shortest path routing problem in the ATM network. Both of these two types of neural network are built based on a chosen example of an ATM network. The Private Network-Node Interface (PNNI) network is a type of ATM network in which the shortest path routing mechanism can be used.

Simulation of the shortest path computation for an ATM network is done for both the Hopfield neural network and the Boltzmann machine.