Leveraging Deep Learning to Achieve Efficient Resource Allocation with Traffic Evaluation in Datacenter Optical Networks

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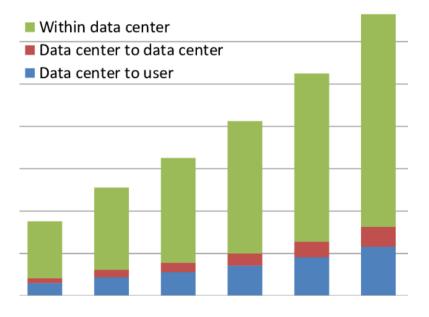
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A. Yu, et al. "Leveraging deep learning to achieve efficient resource allocation with traffic evaluation in datacenter optical networks." 2018 Optical Fiber Communications Conference and Exposition (OFC). IEEE, 2018.

Motivation

- Traffic demand increasing in datacenter networks
 - · Cloud-service, parallel-computing, etc., lead to huge amount of intra datacenter traffic growth.
 - Cisco forecasts 31% increase per year of datacenter traffic by 2021



Datacenter traffic loads is growing



Introduction

- Traditional intra-datacenter traffic detection methods
- Support vector machine (SVM)
- Neural networks (NN)
- Decision tree
- NBD
- Weakness
- Simple prediction model
- Unguaranteed accuracy due to the change of bandwidth explosion and service diversity
- · Unable to detect internal relations for raw data



Deep learning for traffic detection

- Strengths
- Discover deep connections in data
- Accurate prediction in complex networks
- Identify and characterize the complex structural characteristics in huge amounts of raw data

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 Deep neural network-based prediction strategy (DNN-PS)

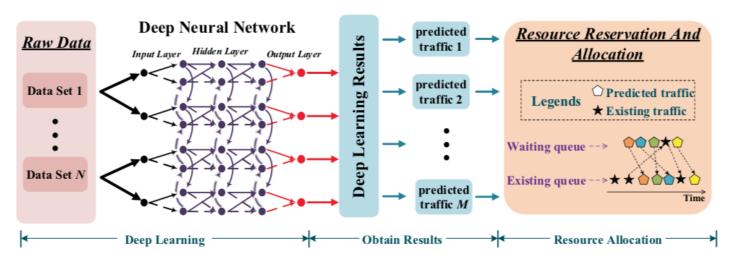


Fig.1 Logic sketch of DNN-PS.

- Build a database containing tens of millions of datacenter traffic information
- Traffic data are collected every 20s from over 200 servers
- Adjusts the weights and bias in DNN during the training process to minimize the objective function.



- Deep learning-based resource allocation algorithm (DL-RA)
- Use DDN-PS model to predict new traffic data
- Allocate resources for the new arrival traffic
- Use α to evaluate the validity of prediction results in terms of accuracy and resources

$$\alpha = \frac{E\left[T_{pj}^{2}(t \cdot t')\right] - \mu_{pj}(t)\mu_{pj}(t')}{\sqrt{D\left[T_{pj}(t)\right]}}\beta + \frac{\sum_{j=1}^{M} \int_{t_{0}}^{t_{N}} R_{pj}(t)dt}{M\int_{t_{0}}^{t_{N}} R(t)dt} (1 - \beta), \quad E\left[T_{pj}^{2}(t \cdot t')\right] - \mu_{pj}(t)\mu_{pj}(t') > 0$$



- Deep learning-based resource allocation algorithm (DL-RA)
- α considers both accuracy of traffic prediction and global resource utilization

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 T_{pi} : arrival time of j^{th} predicted traffic (P-traffic)

 \mathbf{u}_p : value center of T_p at different amounts of arrival time

t': predicted arrival time

t: actual arrival time

 $R_{pj}(t)$: resources required for the j^{th} arrival predicted traffic at t time

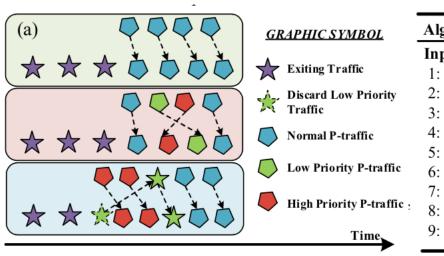
M: total amount of traffic that will arrive during time (t_0 , t_N)

N: arrival time of M_{th} traffic

 β : is the adjustable weight between traffic and resource parameters with different user requirements



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Schematic of traffic queue reordering in DL-RA

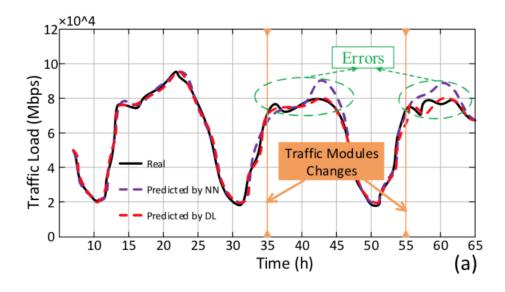
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Algorithms 1: DL-RA(b)Input: training results T_{pj}; expansion parameter Priority = k1: for each T_{pj}2: if R_p \le R_j then3: else if there will be no low priority traffic in queue4: else if k=low then5: T will be added to traffic queue6: else if \alpha > threshold then7: Allocate resources after discarding8: if R_p \le R_j after resource arrangement then9: Allocate resources after arrangement
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The pseudocode of DL-RA algorithm



Simulation results

 Deep learning-based resource allocation algorithm (DL-RA)

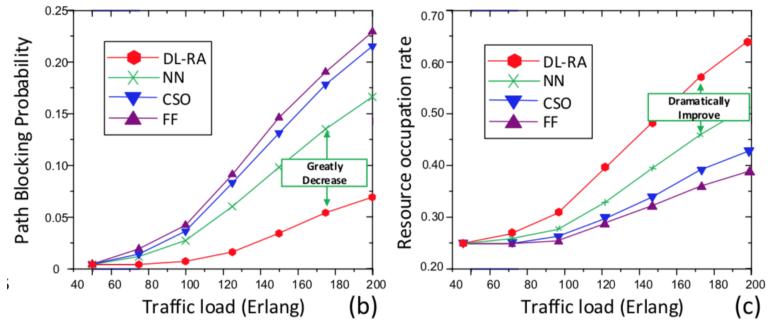


Compare real traffic load, deep learning-based prediction and neural network-based prediction.



Simulation results

 Deep learning-based resource allocation algorithm (DL-RA)



path blocking probability among different resource allocation strategies

resource occupation rate among different resource allocation strategies





