



# Applying Hadoop Cloud Computing Technique to Optimal Design of Optical Networks

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# Outline



- **Background**
- **Hadoop cloud computing system**
- **Optimization for optical networks with Hadoop system**
- **Simulation and performance analyses**
- **Conclusions**

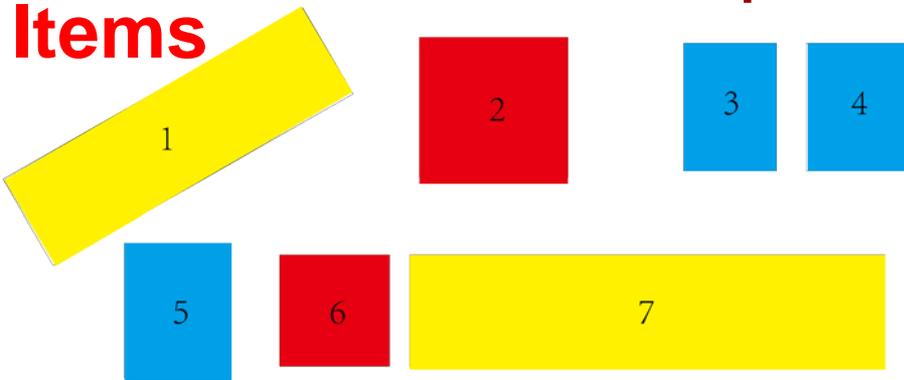
# Background



- Bin-packing problem

→ A classical NP-hard problem that targets to find a minimal number of bins to pack all the (irregular) items

Items



Strategy	Sequence
#1	7,2,6,5,1,3,4
#2	3,4,5,2,6,1,7

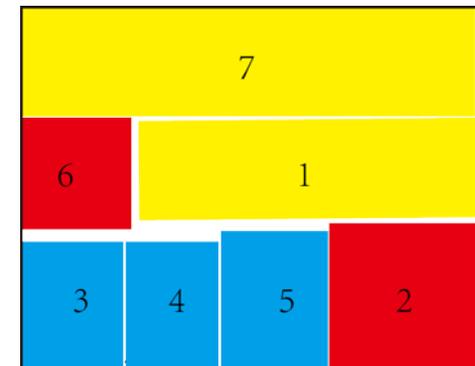
## Sequence 1



Bin 1

Bin 2

## Sequence 2



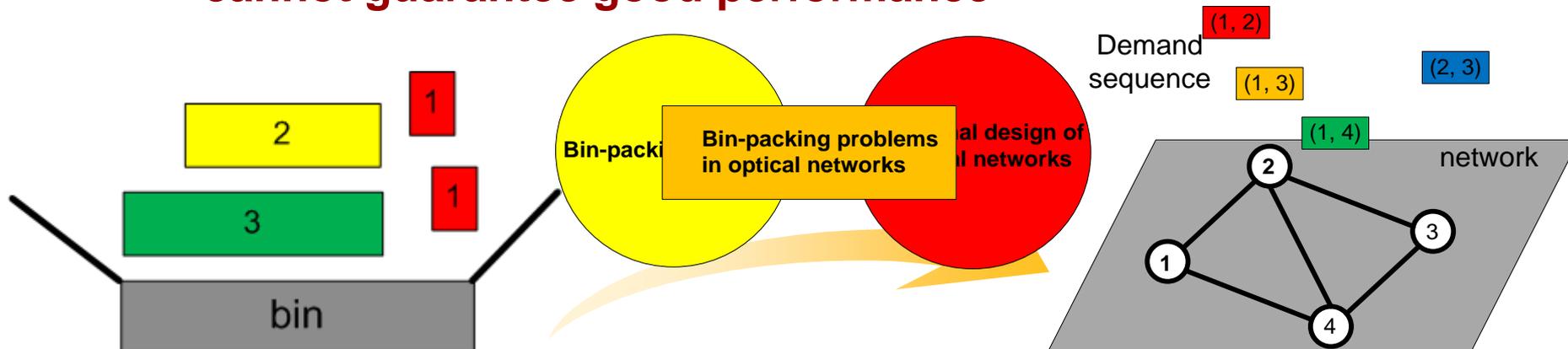
Bin 1

# Background (cont's)



- **Bin-packing problems in optical networks**

- Energy minimization problem, shared backup path protection (SBPP) planning problem, adaptive FEC assignment problem, etc.
- For a traffic demand, in addition to its **size** (i.e., traffic volume), we also need to consider its **shape** (i.e., the route of the demand) when packing (i.e., serving) them
- For these research problems like serving network traffic demands, the simple **largest-smallest** ordering strategy cannot guarantee good performance



# Background (cont's)



- Impact of demand sequence

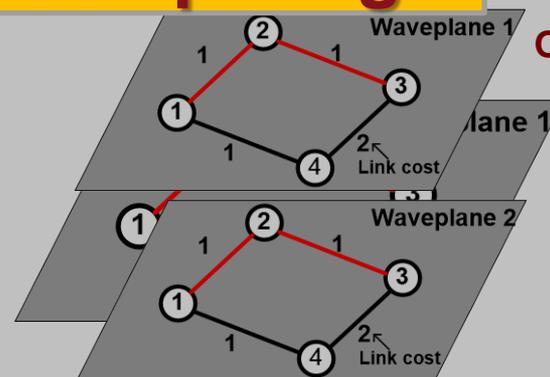
- To achieve good performance, the most straightforward strategy is to randomly shuffle the demand sequences
- Then implement an efficient heuristic algorithm for each of them and finally choose the one with the best performance
- A shuffling strategy has been verified to be successful for the problem [1]



## Parallel computing

### Example

Demand Sequence 2
(1, 2)
(2, 3)
(1, 3)



However, it becomes **computationally intensive and time-consuming** if a huge number of shuffled demand sequences are evaluated

[1] G. Shen, Y. Li, and L. Peng, "Almost-optimal design for optical networks with Hadoop cloud computing: ten ordinary desktops solve 500-node, 1000-link, and 4000-request RWA problem within three hours (Invited)," in Proc. ICTON 2013.

# Hadoop cloud computing system



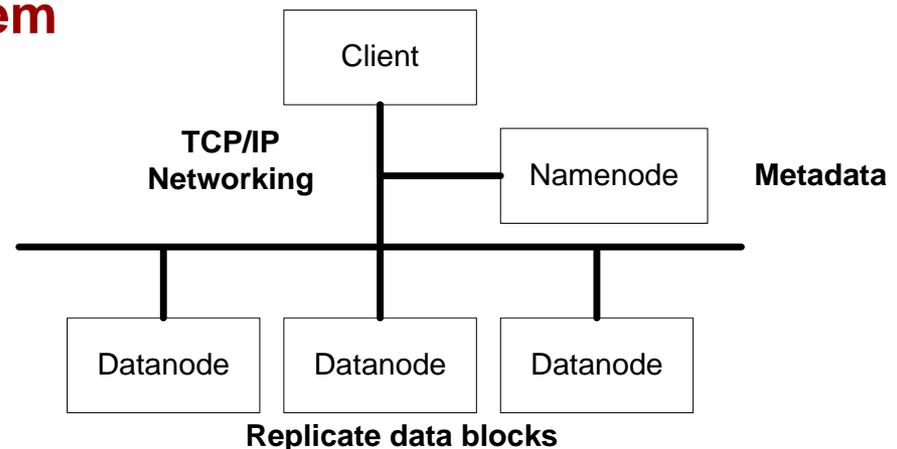
- **Features and advantages**
  - Open source software framework
  - Employ many low-end ordinary machines
  - Provide high parallel computation capacity
  - Cost effective
- **Key functional modules**
  - Hadoop Distributed File System (HDFS)
  - Map/Reduce process

# Hadoop cloud computing system



- **Hadoop Distributed File System (HDFS)**

- A highly fault-tolerant distributed file system designed for commodity hardware
- A HDFS instance typically includes a Namenode and many Datanodes
- Namenode is a coordinator and Datanodes provide storage for HDFS
- TCP/IP communication protocol employed for data transmission in the file system



# Hadoop cloud computing system



- **Map step**

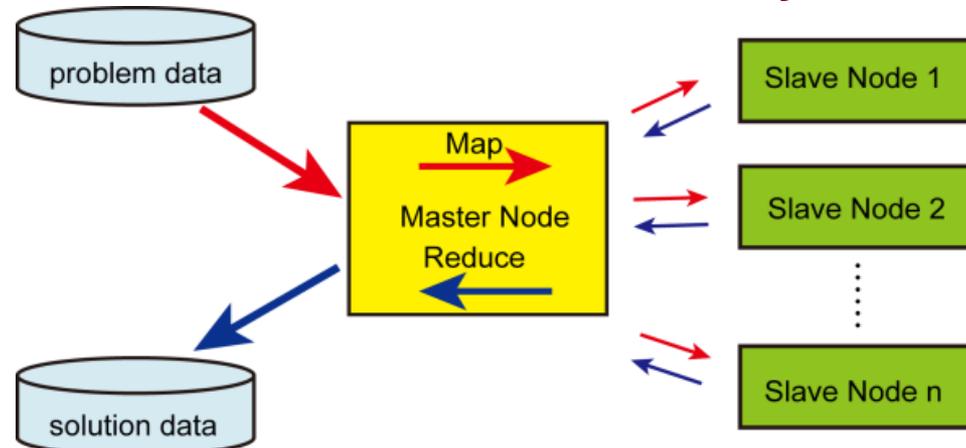
- Receive problem data

- Divide the problem into smaller sub-problems

- Distribute them to **Slave nodes** which will solve the sub-problems and feedback their obtained results to the **Master node**

- **Reduce step**

- Master node is employed to collect the answers to all the sub-problems and combine them in some way to form a final solution



# Hadoop cloud computing system



- **Hardware**

  - Lenovo YangTian T4900d (Dual cores 2.9-GHz CPU)

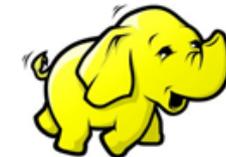
- **Software**

  - Linux (Ubuntu 12.04 LTS)

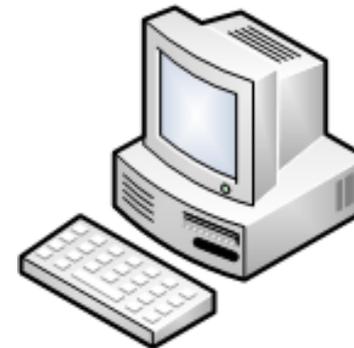
  - Hadoop-1.0.3

  - JDK: Sun-java6-jdk

  - SSH



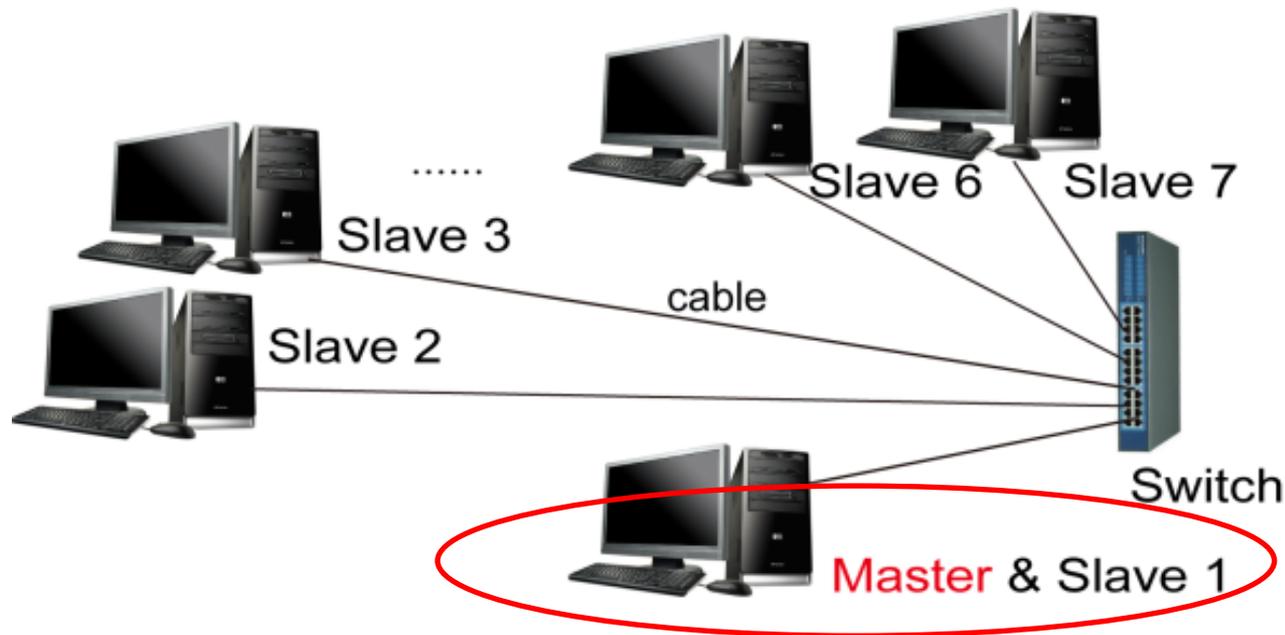
*hadoop*



# Hadoop cloud computing system



- **Seven ordinary desktops Hadoop system**
  - One master and seven slaves interconnected by a **switch**
  - All the computers have **2.9-GHz** dual cores with **2G** memory
  - **Two reduces** on each slave node



# Optimization for optical networks with Hadoop system



- **Energy minimization problem [2]**

- A “Follow the Sun, Follow the Wind” strategy for the IP over WDM network

- An efficient heuristic algorithm to minimize the total non-renewable energy consumption of the network

- **Shuffling strategy**

- Shuffle the node pairs in a list to change their order, and recalculate non-renewable energy consumption

- Compare the results and select the least non-renewable energy as our final result

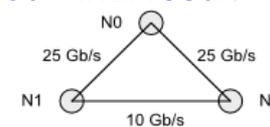
**An example with shuffling strategy:**

**(a) Traffic matrix with serving order 1**

Node pair	Traffic (Gb/s)
(N1, N2)	10
(N0, N1)	25
(N0, N2)	25

(a)

**(b) Established virtual topology for serving order 1**



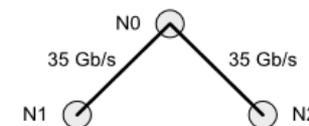
(b)

**(c) Traffic matrix with serving order 2**

Node pair	Traffic (Gb/s)
(N0, N1)	25
(N0, N2)	25
(N1, N2)	10

(c)

**(d) Established virtual topology for serving order 2**



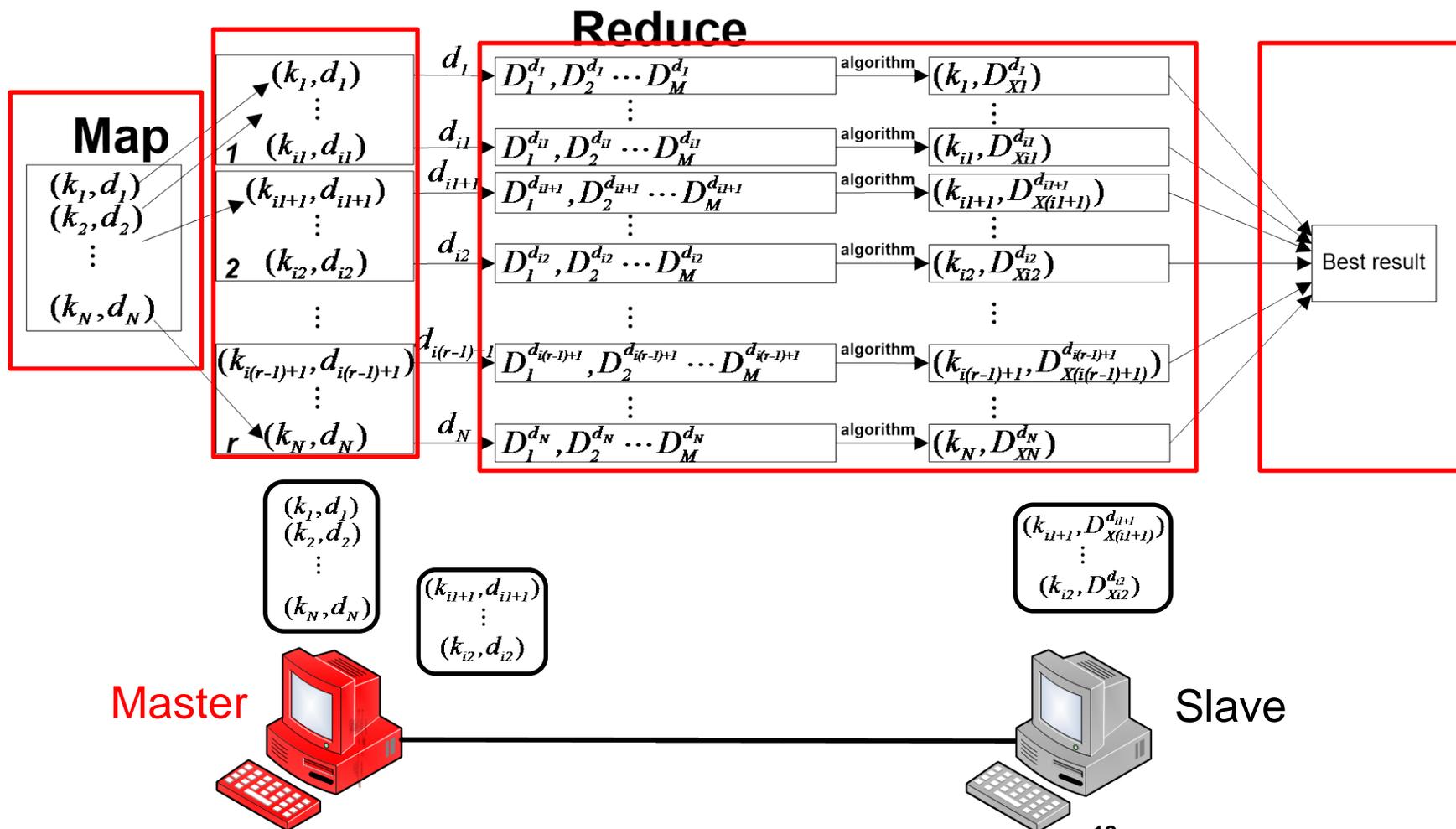
(d)

[2] G. Shen, Y. Lui, and S. K. Bose, ““Follow the Sun, Follow the Wind” lightpath virtual topology reconfiguration in IP over WDM network,” *IEEE/OSA Journal of Lightwave Technology*, vol. 32, no. 11, pp. 2094-2105, Jun. 2014.

# Optimization for optical networks with Hadoop system



- The diagram of Map/Reduce process



# Optimization for optical networks with Hadoop system



- **Map/Reduce process**

- **Map:**

- Generates a total of  $N$  key-value pairs  $(k_i, d_i)$ , for each of which  $k_i$  and  $d_i$  are both set to be  $i$ , i.e.,  $k_i=d_i=i$
    - The key-value pairs are forwarded to  $r$  Reduce functions for parallel computing

- **Reduce:**

- Each of the Reduce functions has  $N/r$  key-value pairs
      - For each key-value pair  $(k_i, d_i)$ , we randomly shuffle the demand list  $M$  times with value  $d_i$  as a random seed.
      - Each of the shuffled demand sequences is then served by the heuristic “algorithm” to generate  $M$  results for each key-value pair  $(k_i, d_i)$
    - We compare these results to choose the one with the best performance for each Reduce function
  - **Compare the best results of each Reduce function to obtain a final global optimum, i.e., “*best result*”**

The whole process evaluates a total of  $N \cdot M$  shuffled demand sequences to obtain the final best solution

# Simulation conditions

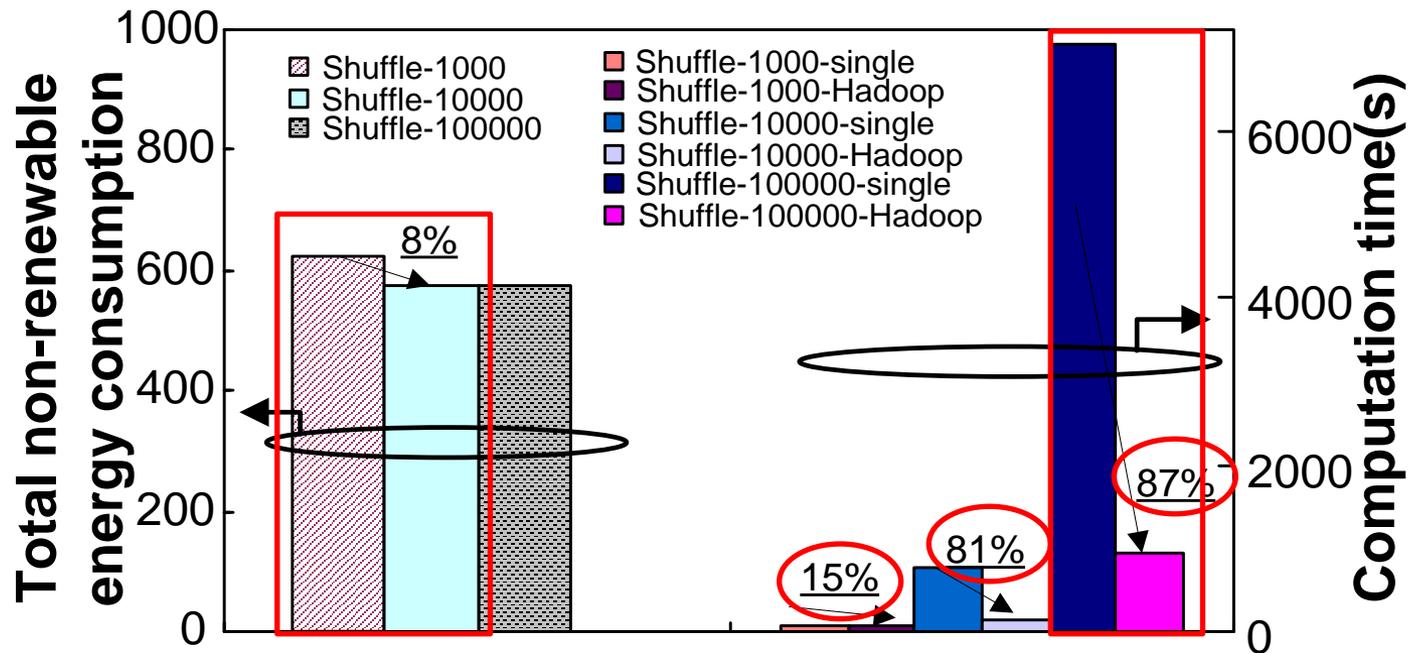


- **Map/Reduce process**
  - **M: 10**
  - **N: [100, 10,000]**
  - **Shuffled demand sequences: [1,000, 100,000]**
- **Energy minimization problem**
  - **Test network case: 43-link US backbone network (USNET)**
  - **Traffic demand between each node pair is randomly generated within the range of [10, 30] Gb/s**
  - **The total non-renewable energy consumption is shown for USNET on May 2<sup>th</sup>, 2014**

# Performance analyses



- **Total non-renewable energy consumption and computation time (USNET)**



- When the number of shuffled demand sequences increases from 1,000 to 10,000, the total non-renewable energy consumption decreases 8%.
- Hadoop system can significantly shorten computation time for the same number of shuffled demand sequences
- Hadoop system is more efficient for a larger optimization problem

# Optimization for optical networks with Hadoop system



- **SBPP network planning problem [3]**
  - Distance adaptive routing and spectrum assignment (RSA) for SBPP-based elastic optical networks
  - An efficient heuristic algorithm based on spectrum window planes (SWPs) to minimize the maximal number of required frequency slots (FSs) in the entire network
  - **Shuffling strategy**
    - Shuffle the node pair list many times
    - Run the heuristic algorithm for each shuffled node pair list to obtain the required number of FSs,  $FS_i$
    - Finally compare the obtained  $FS_i$  to choose the one with the smallest number, i.e.,  $FS_{min} = \min \{FS_i\}$ .

[3] C. Wang, G. Shen, and S. K. Bose, "Distance adaptive dynamic routing and spectrum allocation in elastic optical networks with shared backup path protection," *IEEE/OSA Journal of Lightwave Technology*, vol. 33, no. 14, pp. 2955-2964, Apr. 2015.

# Simulation conditions

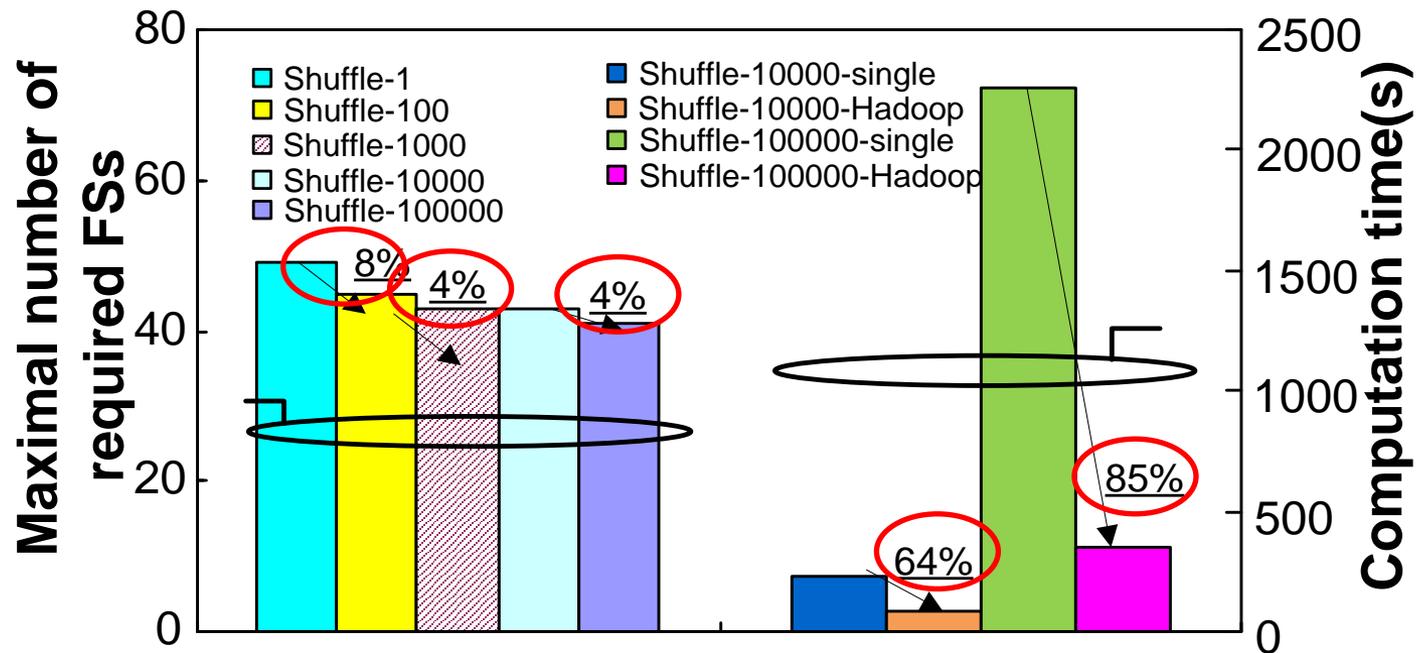


- **SBPP network planning problem**
  - Test network case: **COST239 network**
  - Frequency slots in each link: **400 FSs**
  - Bandwidth granularity of each FS: **12.5 GHz**
  - Modulation format set: **BPSK, QPSK, 8-QAM**
  - The bandwidth request between each node pair is uniformly randomly distributed within a range of **[10, 400] Gb/s**

# Performance analyses



- Maximal number of required FSs and computation time (COST239)



- The shuffling strategy is helpful to significantly reduce the number of required FSs
- The Hadoop system does help to reduce computation time compared to a single machine

# Optimization for optical networks with Hadoop system



- **Adaptive FEC assignment problem [4]**
  - A new adaptive FEC assignment strategy for the elastic optical network
  - A heuristic algorithm based on slot window planes (SWPs) to maximize the total number of FSs used for actual user data transmission
  - **Shuffle strategy**
    - Shuffle the lightpath demand list for many times
    - For each shuffled demand list, the adaptive FEC-based SWP algorithm is applied to find a solution
    - The best solution is selected as the final solution when all the shuffled demand lists have been evaluated and compared.

[4] Y. Li, H. Dai, G. Shen, and S. K. Bose, "Adaptive FEC selection for lightpaths in elastic optical networks," in *Proc. OFC 2014*.

# Simulation conditions

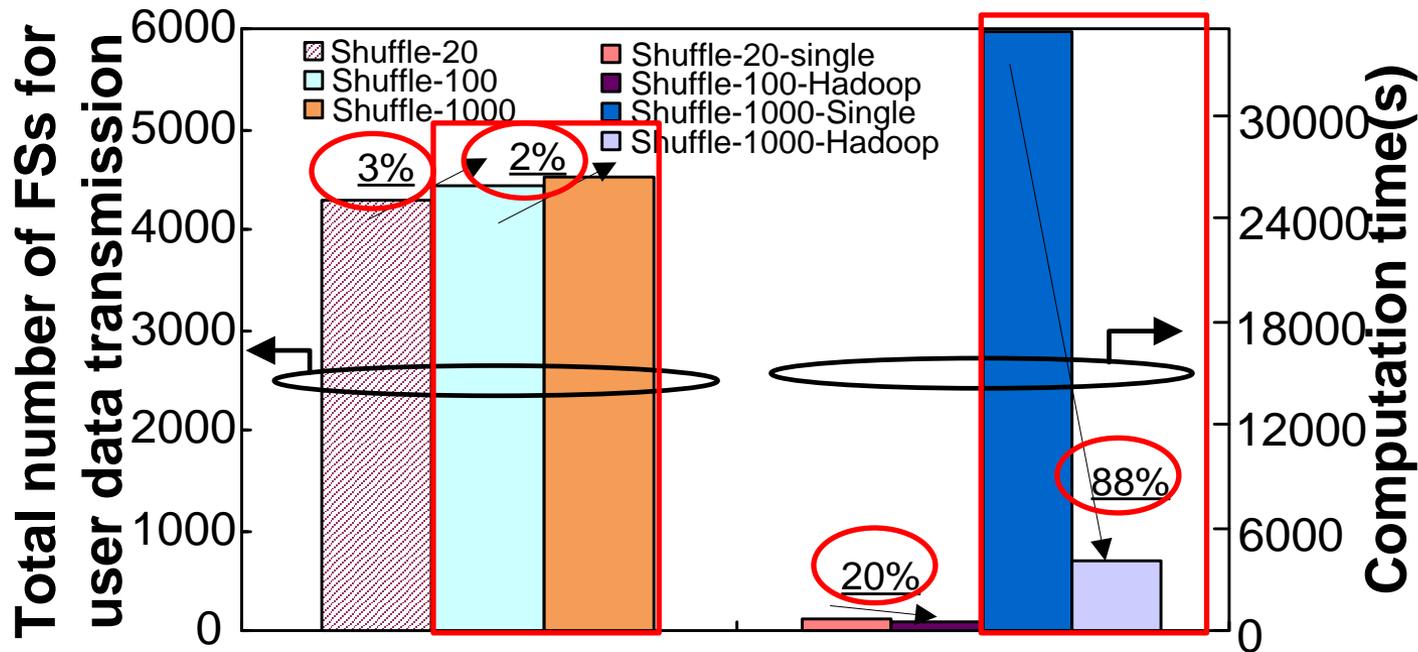


- **Adaptive FEC assignment problem**
  - Test network case: 43-link US backbone network (USNET)
  - Frequency slots in each link: **400 FSs**
  - The number of FSs required by users is uniformly randomly distributed within a range of **[25, 35]**

# Performance analyses



- **Total number of FSs for user data transmission and computation time (USNET)**



- The Hadoop system can perform better than a single machine by **3%** and it can also shorten computation time **20%** even though it evaluates 10 times more shuffled demand sequences
- A further increase of the number of shuffled demand sequences to 1,000 can lead to **2%** further performance improvement, under which the computational efficiency of the Hadoop system is more significant to take only about **1 hour**, while a single desktop took about **9 hours**

# Conclusions



- The bin-packing optimization problems in the optical network can be efficiently solved by evaluating multiple shuffled demand sequences
- To overcome the difficulty of high computational complexity, we built a Hadoop computing system to parallel evaluate multiple shuffled demand sequences
- The approach was efficient to find solutions close to optimality and significantly reduce computation times compared to a single machine



**Thank you!**