# History-Based Harvesting of Spare Cycles and Storage in Large-Scale Datacenters

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### Outline

- Introduction
- 2 Behavior Patterns
- 3 Co-location Techniques
- 4 Experiment
- Conclusion

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### Introduction

- Overprovision resources in datacenters
  - Low tail latency requirement
  - Provisioned for peak load
  - Unexpected load spikes and failures
- A way to increase utilization and reduce costs in datacenters is to co-locate their latency-critical services and batch workloads.
- Harvest spare compute cycles and storage space for co-location purpose.

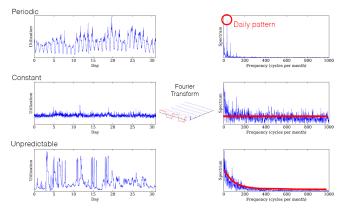
# Challenges

- Interactive services "own" the servers
- Resource availability
  - Interactive services performance
  - Resource availability dynamics task killing
- Data storage co-location
  - Data availability
  - Data durability

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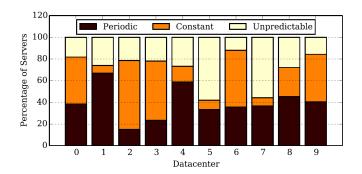
### Resource Utilization

• Identify three main classes of primary tenants.



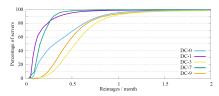
### Resource Utilization

Percentages of servers per class.

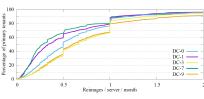


# Disk Reimaging

• Per-server number of reimages in three years.

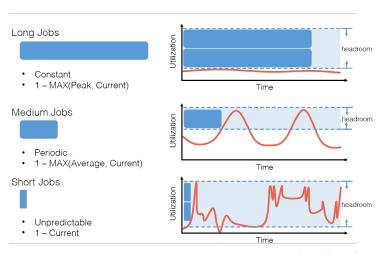


• Per-tenant number of reimages in three years.



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# History-Based Task Scheduling



# Task Scheduling

- Fast Fourier Transform (FFT)
  - Get 3 patterns.
- Clustering algorithm
  - K-means algorithm
  - Average and peak utilizations
- Class selection algorithm

## Class Selection Algorithm

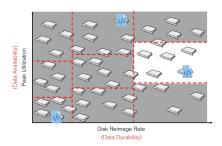
#### Algorithm 1 Class selection algorithm.

```
    Given: Classes C, Headroom(type,c), Ranking Weights W

function SCHEDULE(Batch job J)
3:
        J.type = Length (short, medium, or long) from its last run
4:
        J.req = Max amount of concurrent resources from DAG
        for each c \in C do
6:
            c.weightedroom=Headroom(J.type,c) \times W[J.type,c.class]
7:
        end for
8.
        F = \{ \forall c \in C | \text{Headroom}(J.\text{type}, c) \ge J.\text{req} \}
g.
        if F \neq \emptyset then
10:
            Pick 1 class c \in F probabilistically \propto c.weightedroom
11:
            return \{c\}
12:
        else if Job J can fit in multiple classes combined then
13:
            Pick \{c_0, \ldots, c_k\} \subseteq C probabilistically \propto c.weightedroom
14:
            return \{c_0,\ldots,c_k\}
15:
        else
16:
            Do not pick classes
17:
            return {0}
18.
        end if
19: end function
```

# History-Based Data Placement

- Data availability
  - Diverse in utilization pattern.
- Data durability
  - Diverse in reimaging pattern.



# Replica Placement Algorithm

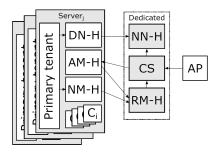
#### **Algorithm 2** Replica placement algorithm.

```
1: Given: Storage space available in each server, Primary reimaging
2:
           stats, Primary peak CPU util stats, Desired replication R
   function PLACE REPLICAS(Block B)
4:
       Cluster primary tenants wrt reimaging and peak CPU util
5:
           into 9 classes, each with the same total space
6:
       Select the class of the server creating the block
7.
       Select the server creating the block for one replica
8.
       for r = 2: r < R: r = r + 1 do
9:
           Select the next class randomly under two constraints:
10:
              No class in the same row has been picked
11:
              No class in the same column has been picked
12:
           Pick a random primary tenant of this class as long as
13:
              its environment has not received a replica
14:
           Pick a server in this primary tenant for the next replica
15:
           if (r \mod 3) == 0 then
16:
               Forget rows and columns that have been selected so far
17:
           end if
18.
       end for
19: end function
```

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## System Implementation

 Overview of YARN-H, Tez-H, and HDFS-H in a co-location scenario.

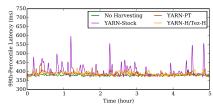


### **Evaluation**

- Environment:
  - 102-server cluster
- Primary tenant (interactive service):
  - Apache Lucene search engine with utilization trace
- Batch task:
  - TPC-DS benchmark

# Batch Task Scheduling

Interactive services performance

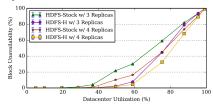


• Job duration - reducing task killing

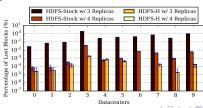


### Data Placement

• Data availability



Data durability



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### Conclusion

- Embody knowledge of the existing primary workloads, and leverage historical utilization.
- Improve batch job performance while protecting primary workloads.
- Eliminate data loss and unavailability in many scenarios.

### Future Work

- Paper
  - Isolation and security in public cloud.
- Project
  - Network utilization.
  - Constraint of Ethernet bandwidth.