

# Service Management for Stream Processing

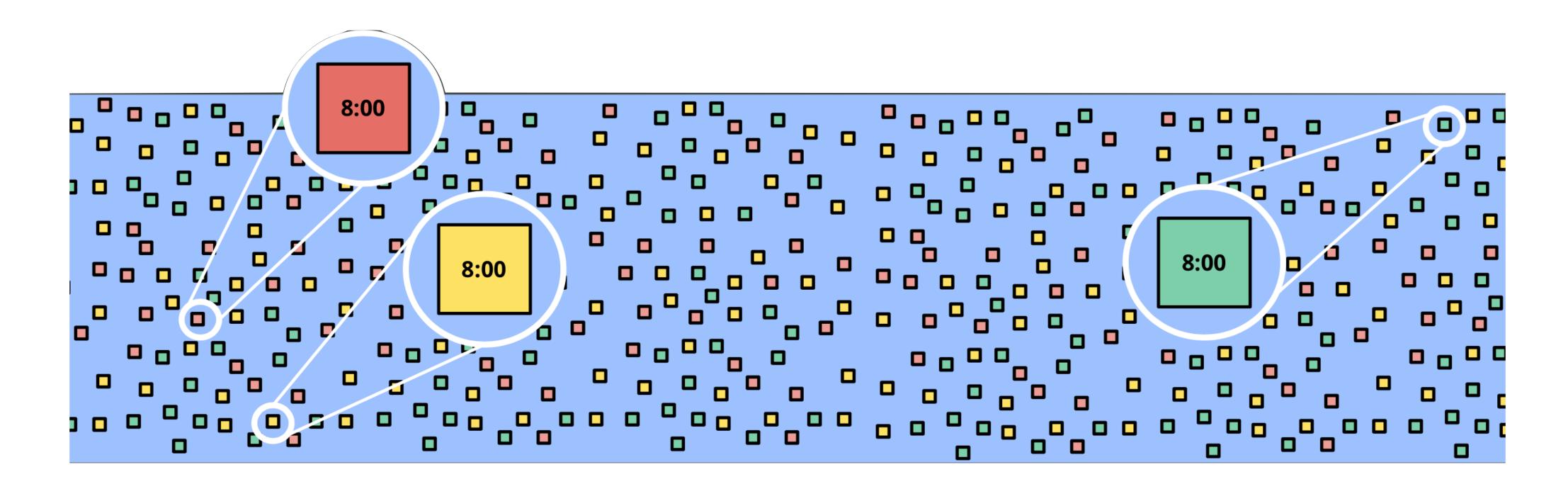
Yalun Lin Hsu



# Turbine

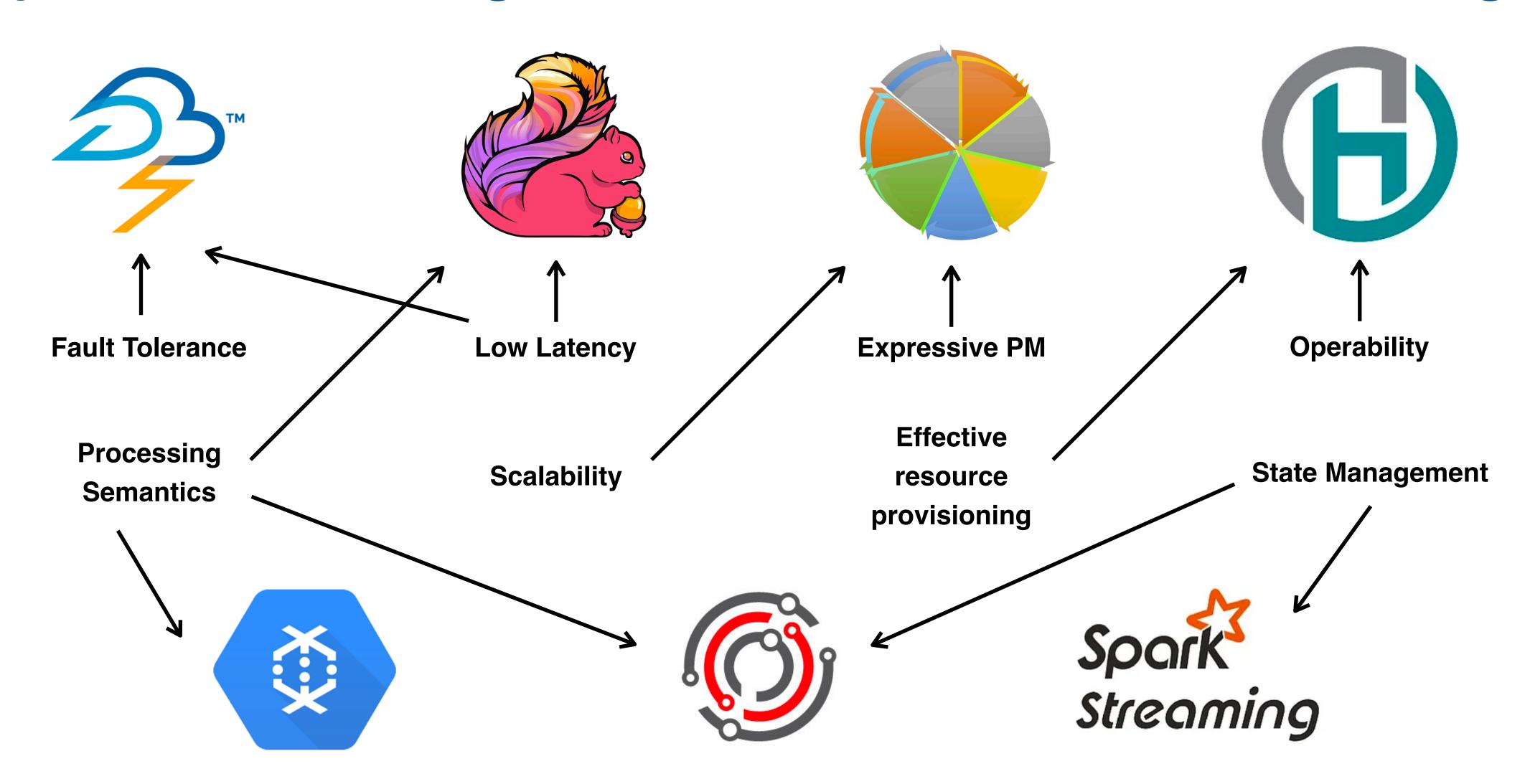
Facebook's Service Management Platform for Stream Processing

#### Review: Streaming

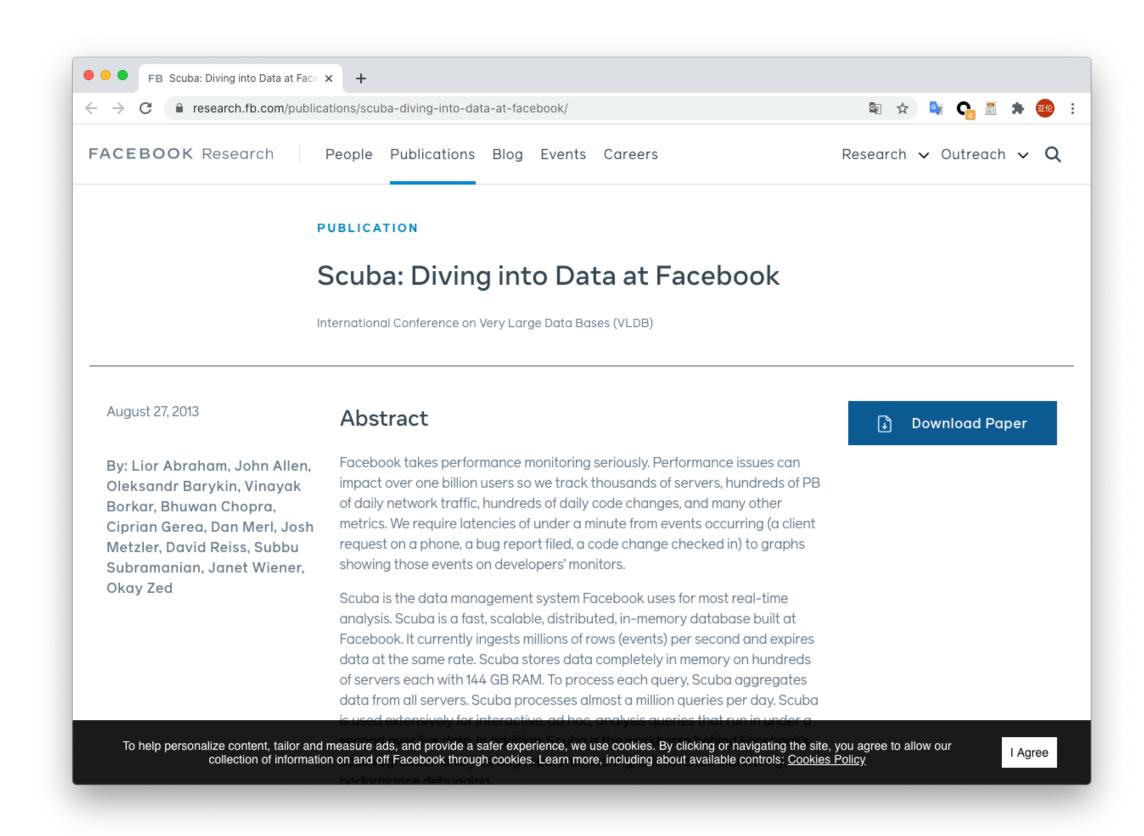


Data can be infinitely big with unknown delays.

### Systems for large-scale distributed streaming



#### Example: Scuba Tailer in Facebook



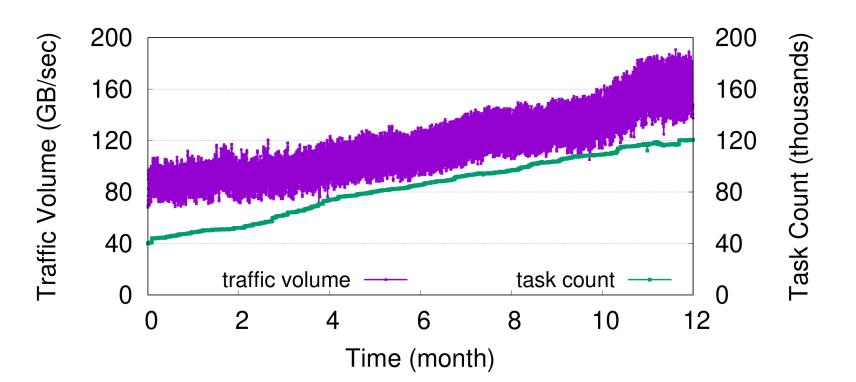


Fig. 1: The growth of Scuba Tailer service in a one-year interval, in terms of task count and input traffic volume.

#### Requirements of Streaming in Facebook

 Low latency analysis of site content interaction, recommendation-related activities ...

- A large number of stateless and stateful stream processing applications;
- Strict Service Level Objectives (SLOs) with low downtime and processing lag

#### Turbine

- A scheduling and lifecycle management framework featuring fast task scheduling and failure recovery;
- An efficient predictive auto scaler that adjusts resource allocation in multiple dimensions;
- An ACIDF application update mechanism.

# Architecture

#### Turbine Architecture

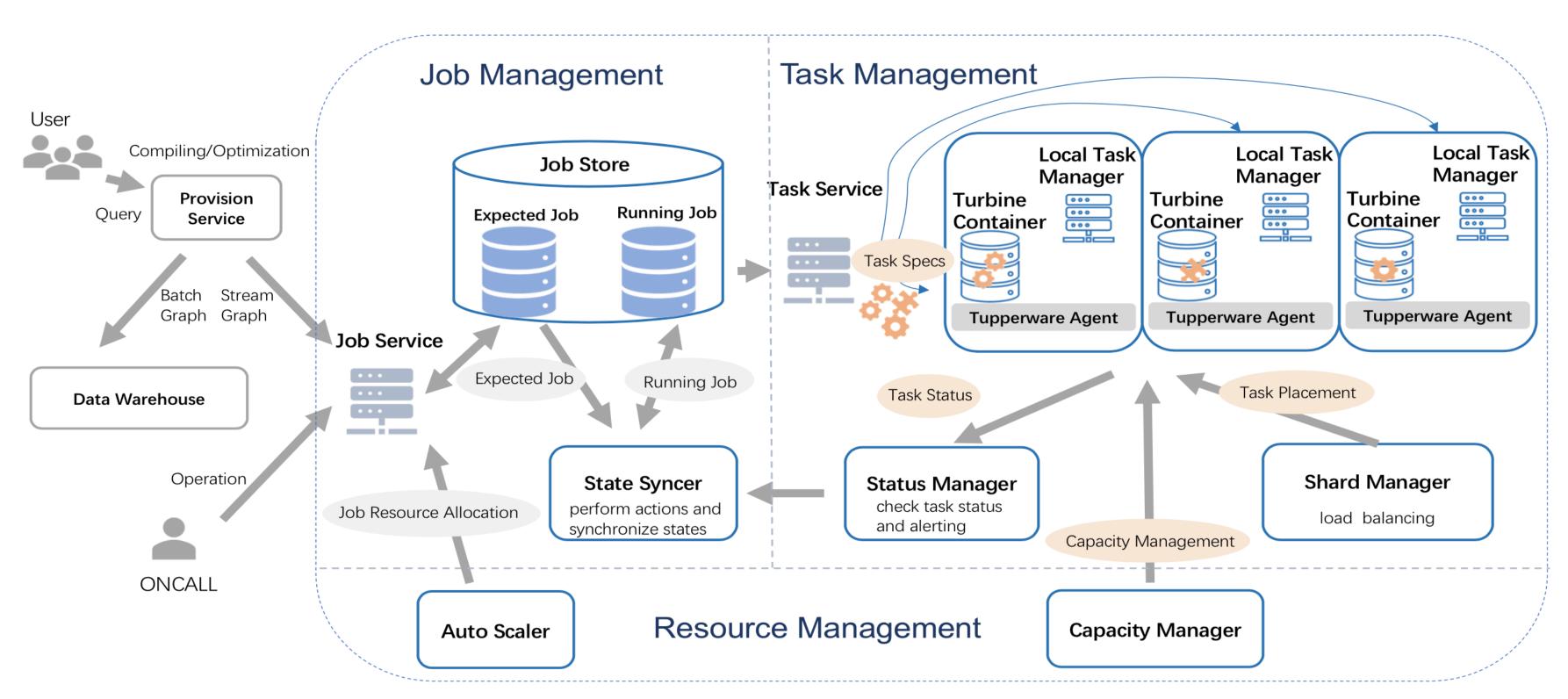
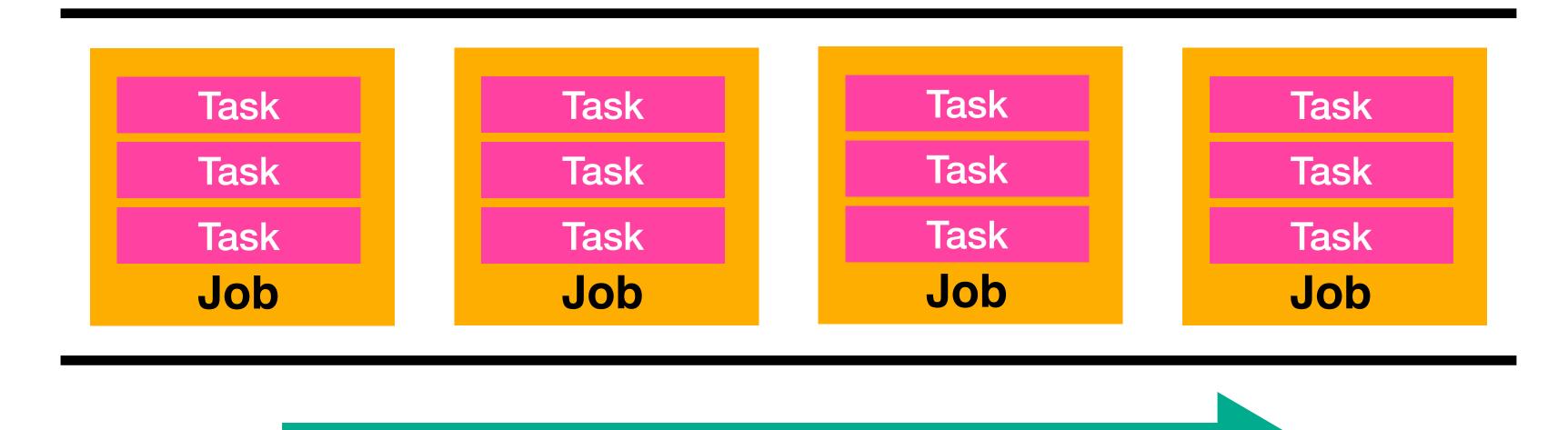


Fig. 2: Turbine's system architecture to manage stream processing services.

#### Turbine Stream Pipeline

#### **Stream Pipeline**



#### Turbine Architecture

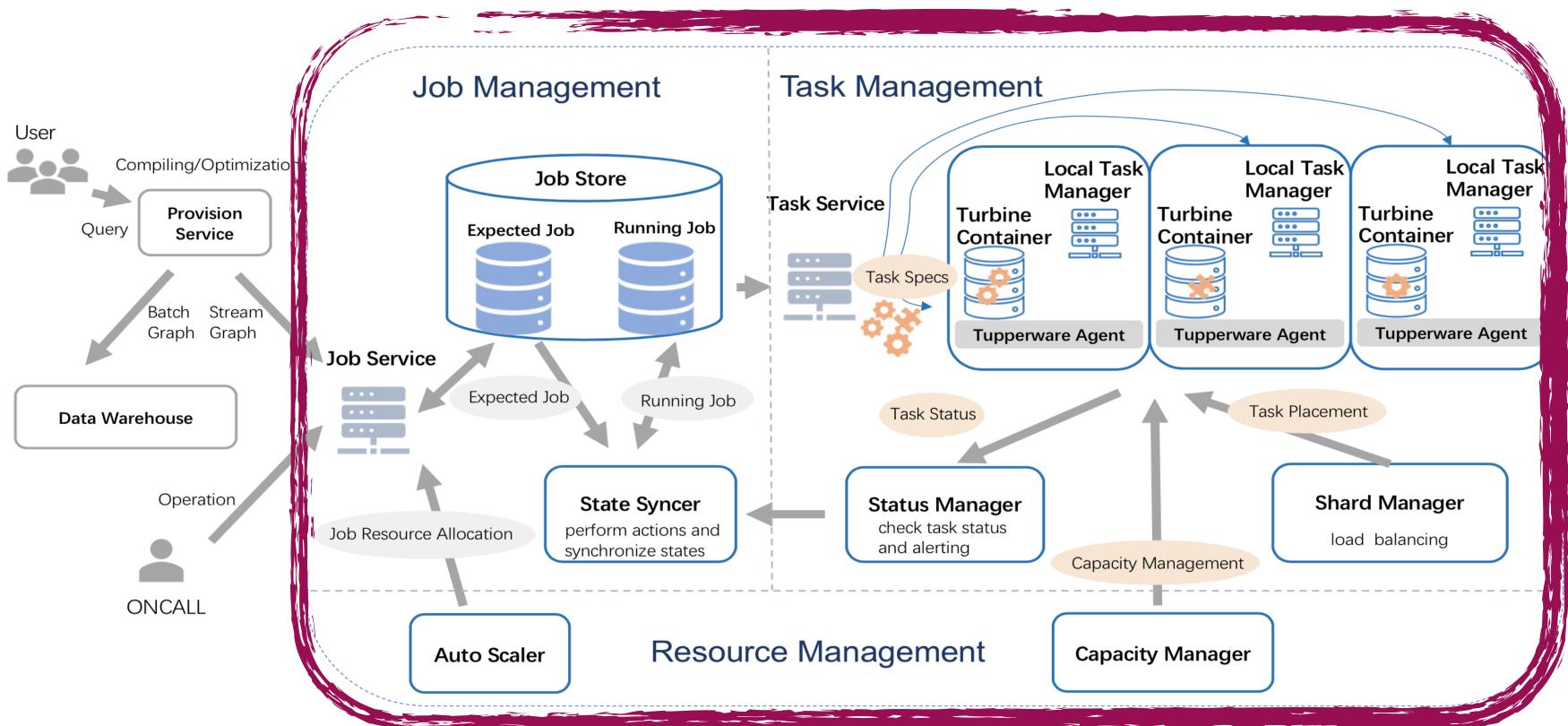
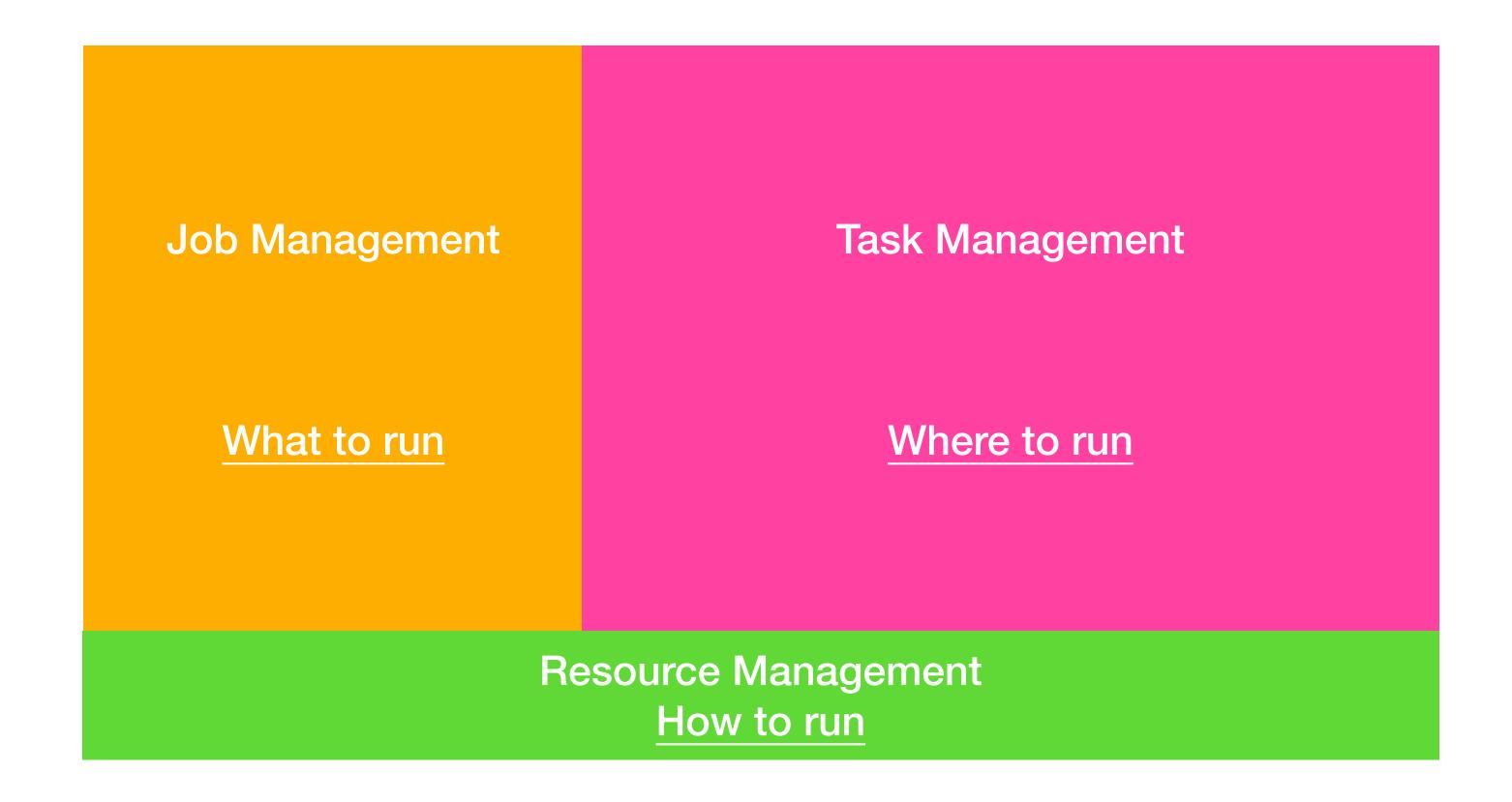
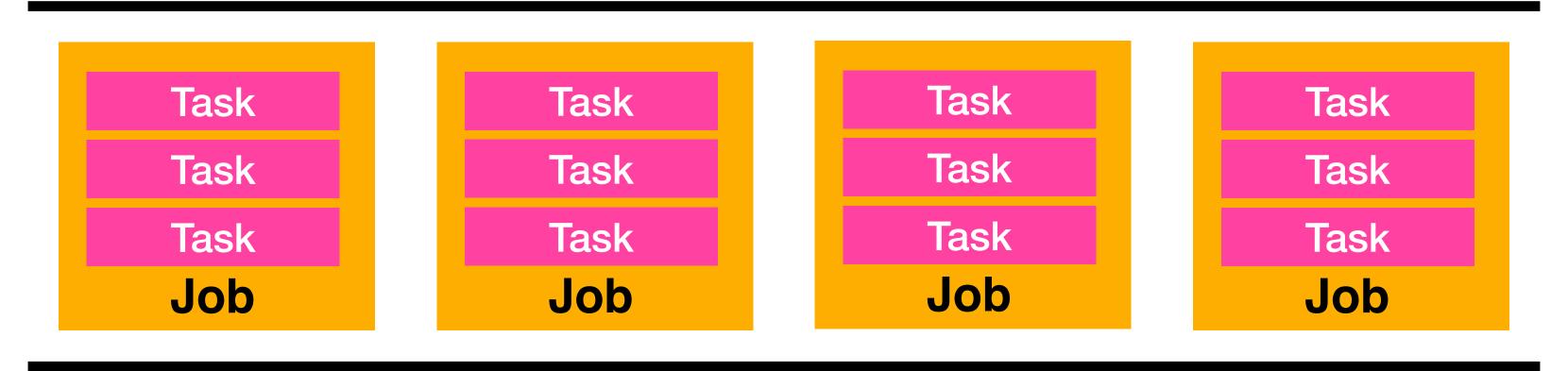


Fig. 2: Turbine's system architecture to manage stream processing services.

# Turbine Core Components



#### Communication between Jobs



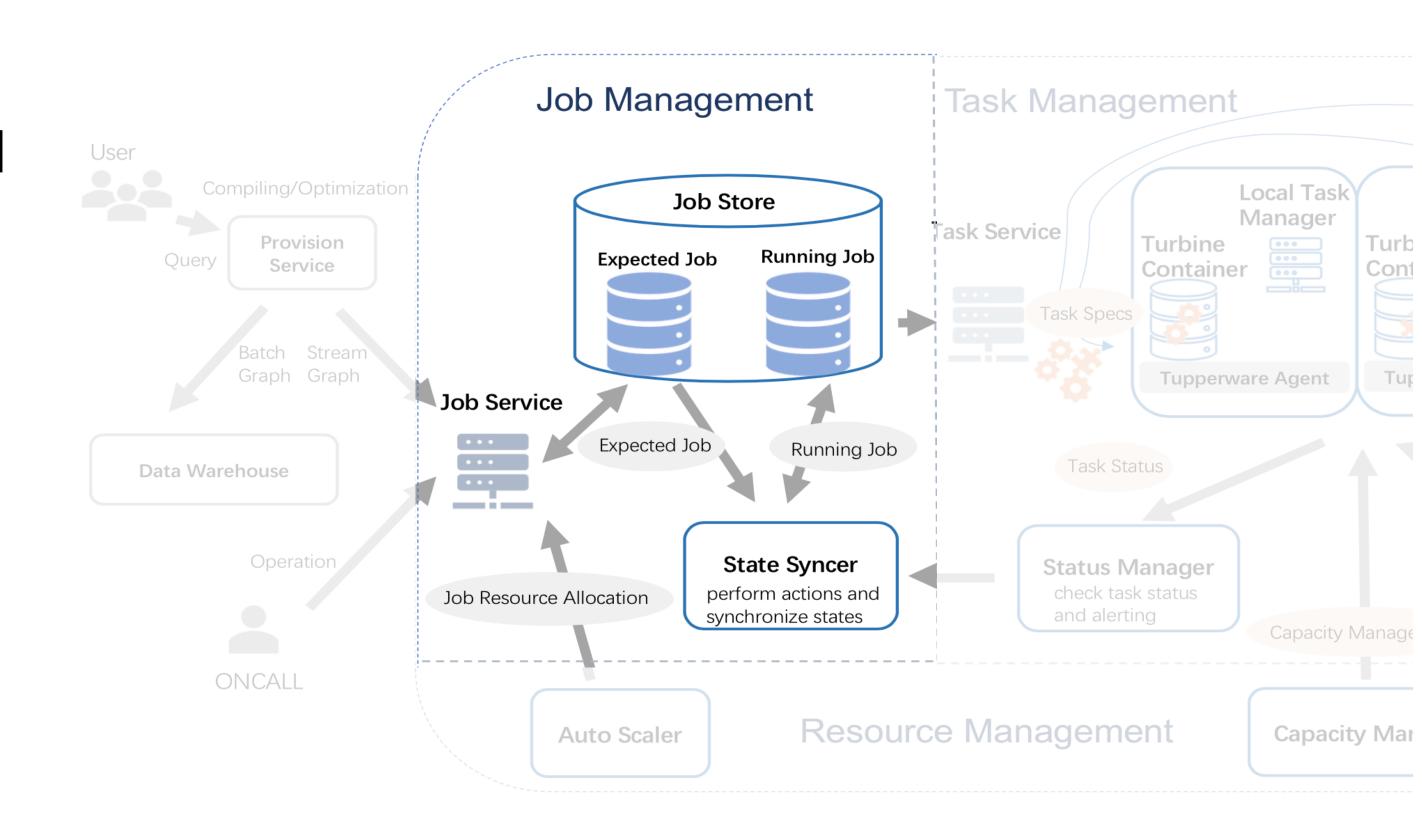


# Job Management

#### Job Management Components

#### To guarantee ACIDF:

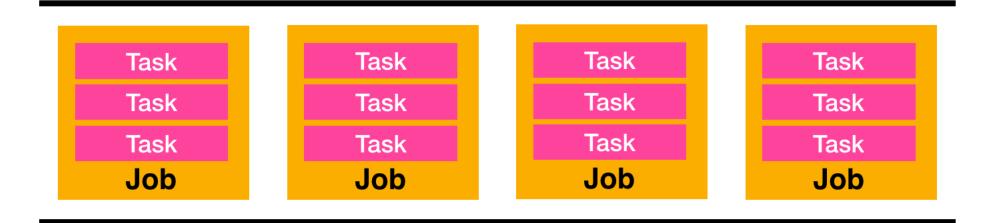
- Job Store: Contains the current and desired config params for each job;
- Job Service: Guarantees job changes are committed;
- State Syncer: Executes job update actions.



#### Job Configurations Update Issue

Isolation

Consistency



T0: Run 10 tasks

T1: Auto Scaler (Internal) - 15 tasks

T1: Oncall1 (External User) - 20 tasks

T1: Oncall2 (External User) - 30 tasks

### Hierarchical Expected Job Configuration

<b>Expected Job Table</b>
Base Configuration
Provisioner Configuration
Scaler Configuration
Oncall Configuration

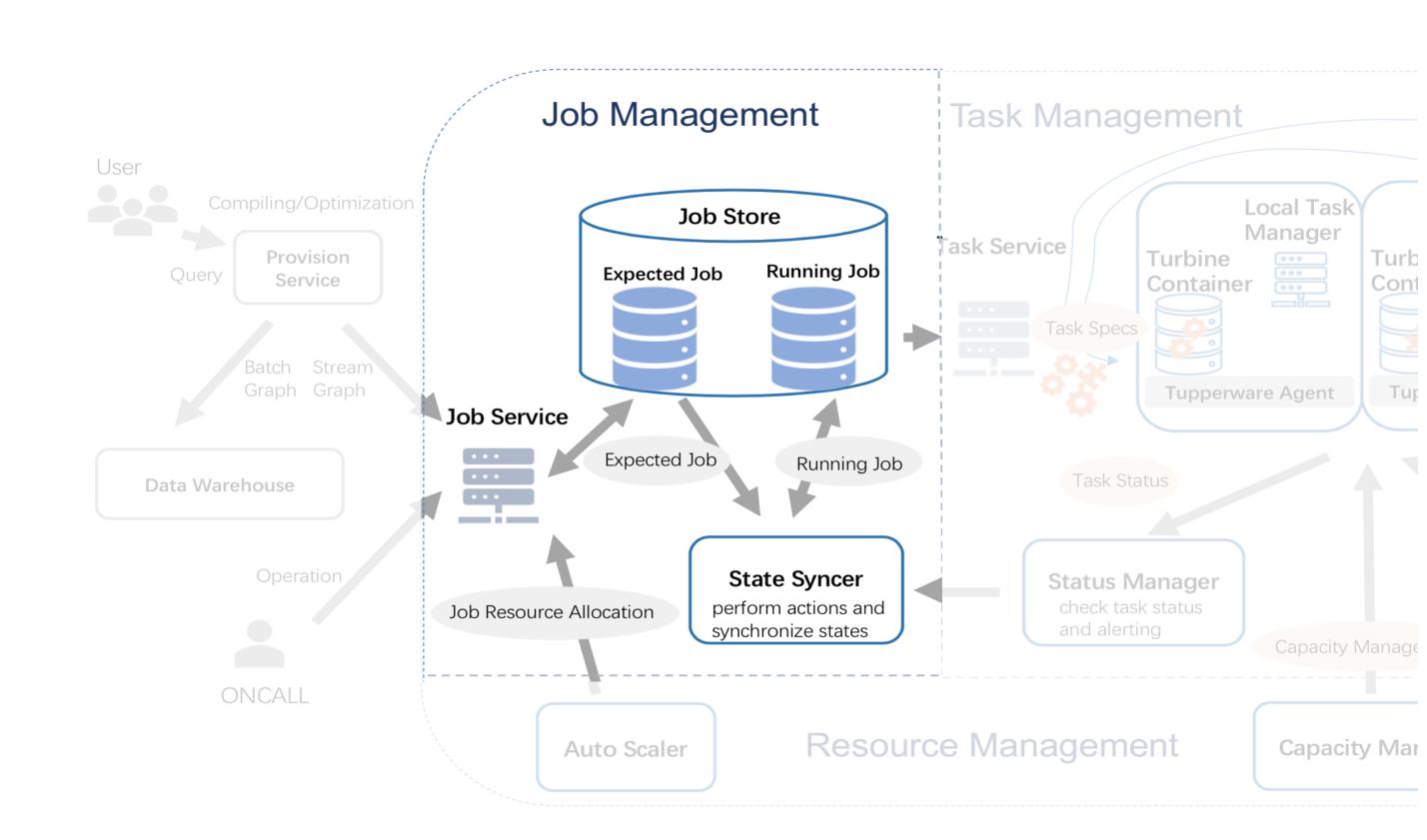
Running Job Table
Running Configuration

#### Algorithm 1 Merge JSON Configs

```
1: procedure LAYERCONFIGS(bottomConfig, topConfig)
2: layeredConfig ← bottomConfig.copy()
3: for each (key, topValue) ∈ topConfig do
4: if isInstance(topValue, JsonMap) && (key in bottomConfig) then
5: layeredValue ← layerConfigs(bottomConfig.get(key), topValue)
6: layeredConfig.update(key, layeredValue)
7: else
8: layerConfig.update(key, topValue)
9: end if
10: end for
11: return layeredConfig
12: end procedure
```

# State Syncer

- Atomicity
- Fault-tolerance
- Durability

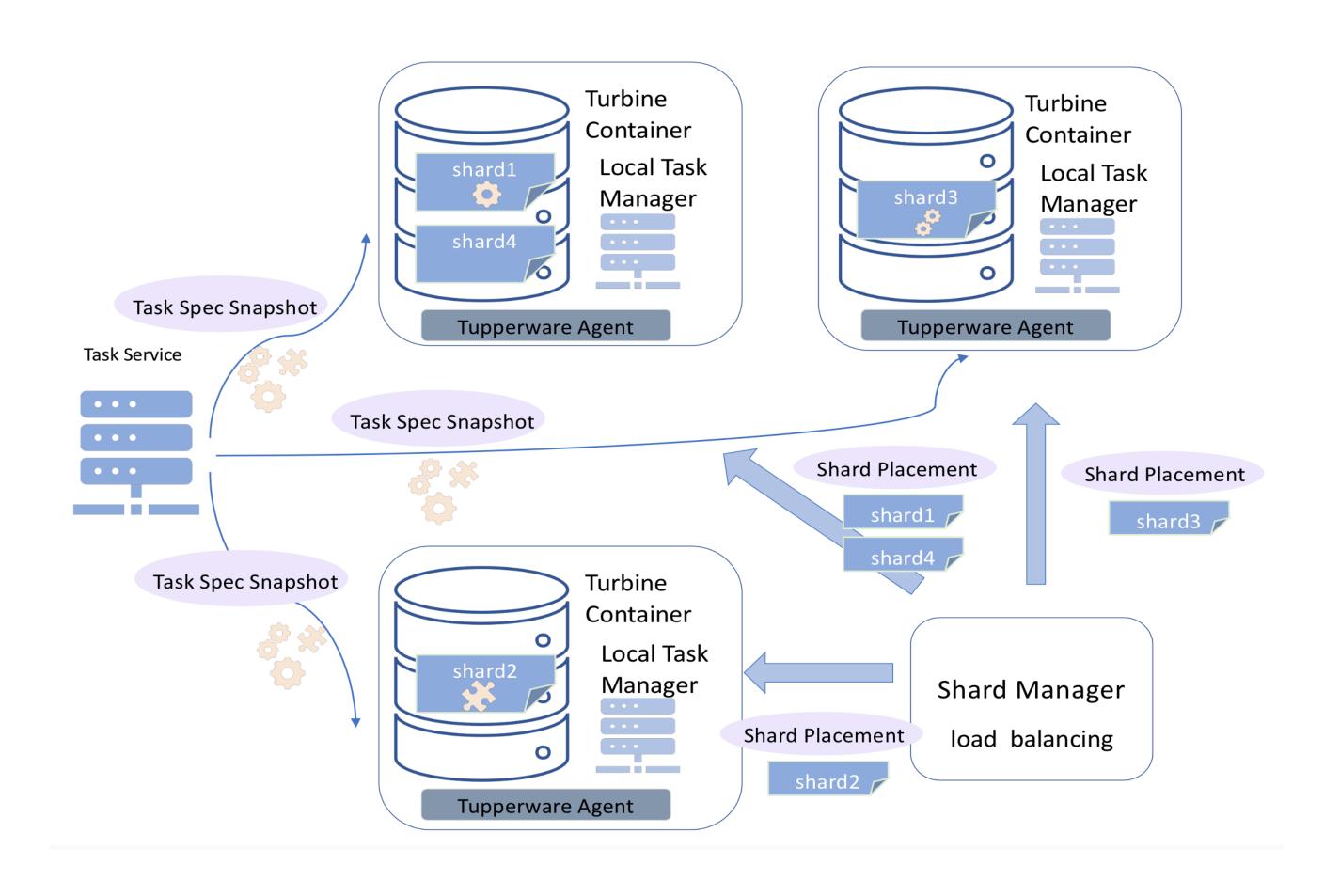


# Task Placement & Load Balancing

### Goal of Task Management

 Schedule tasks without duplication; Task Management Job Management **Local Task** Local Task **Local Task** Manager Manager Manager Fail-over tasks to healthy hosts during host **Task Service Turbine Turbine** Turbine Container Turbine Container Container failures; **Tupperware Agent Tupperware Agent Tupperware Agent** Job Service Restart tasks upon crashes; **Shard Manager** State Syncer **Status Manager** check task status load balancing and alerting Load balance tasks across the cluster for Resource Management even CPU, Memory and IO usage. Auto Scaler **Capacity Manager** 

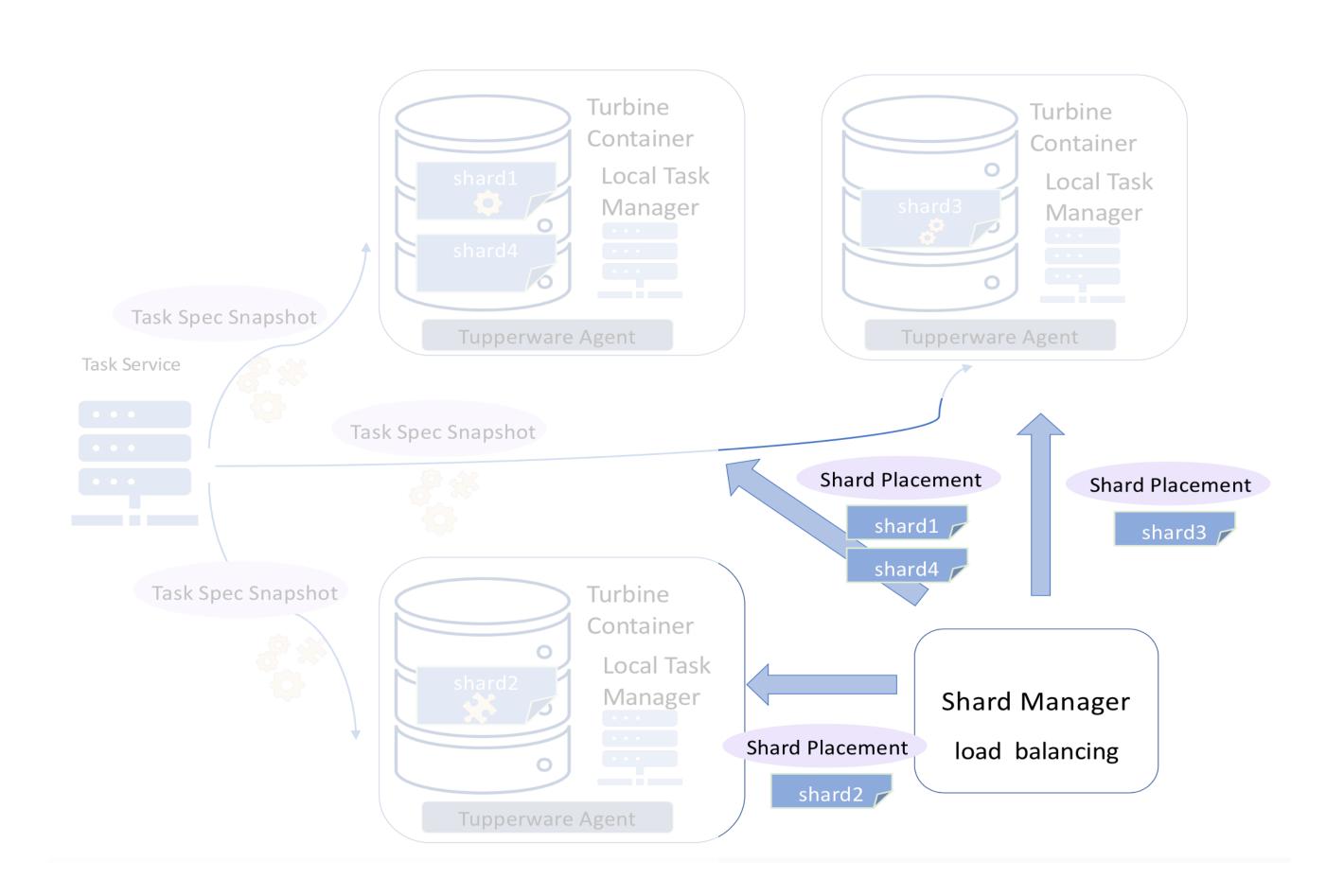
# Task Management Overview



### Scheduling

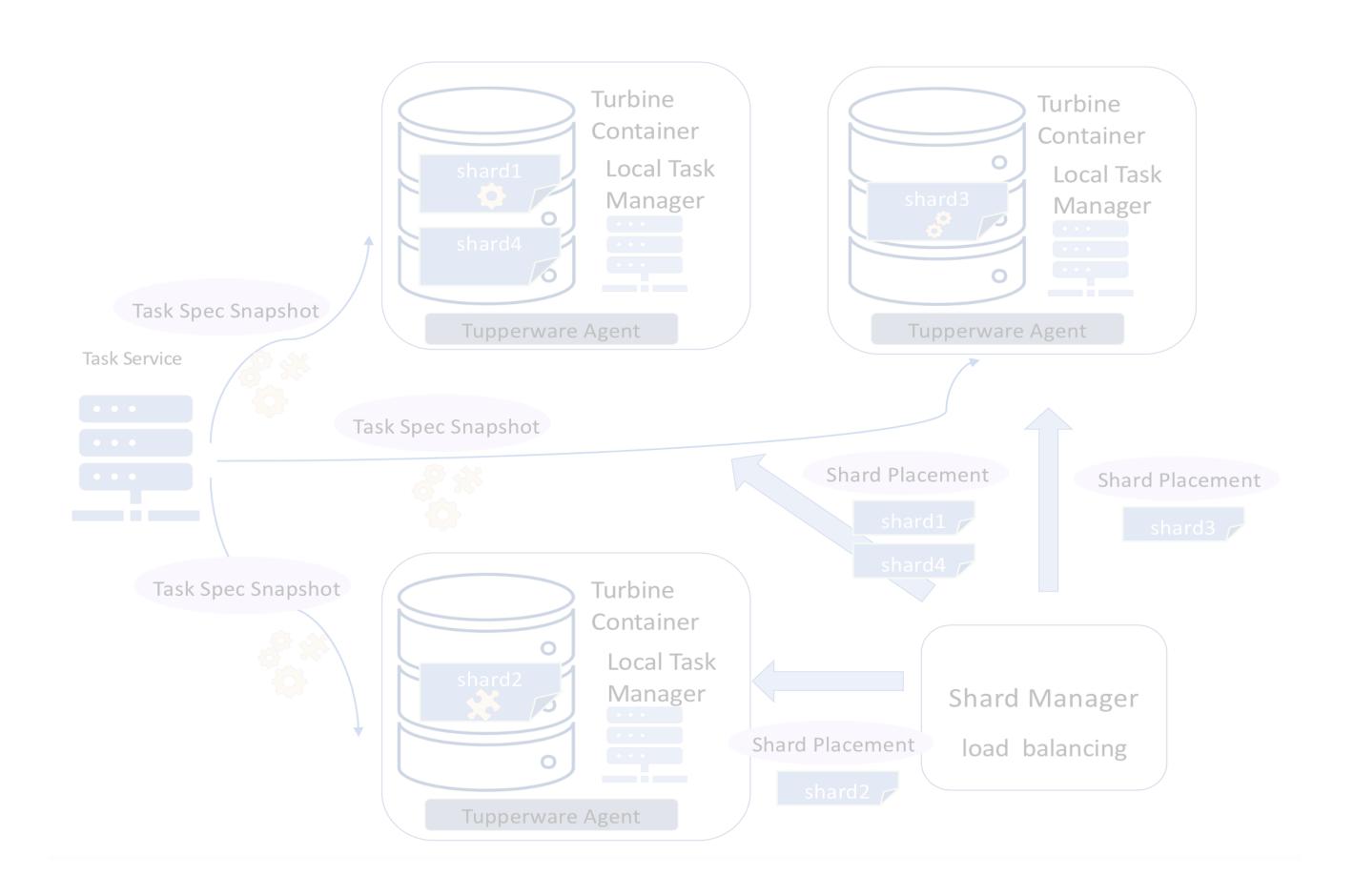
#### Task to Shard Mapping:

- Local Task Manager determines
   tasks <-> shard;
- Fetch Task Service periodically;
- Can handle load balance and failover even when Task Service is down



#### Load Balancing

- Container —> capacity
   Shard —> load
- A bin-packing problem
- Define shard load is IMPORTANT
  - Dynamic resource usage (C, C++)
  - Xmx, cgroup usage (JAVA)
  - A background load aggregator thread in each Task Manager



# Elastic Resource Management

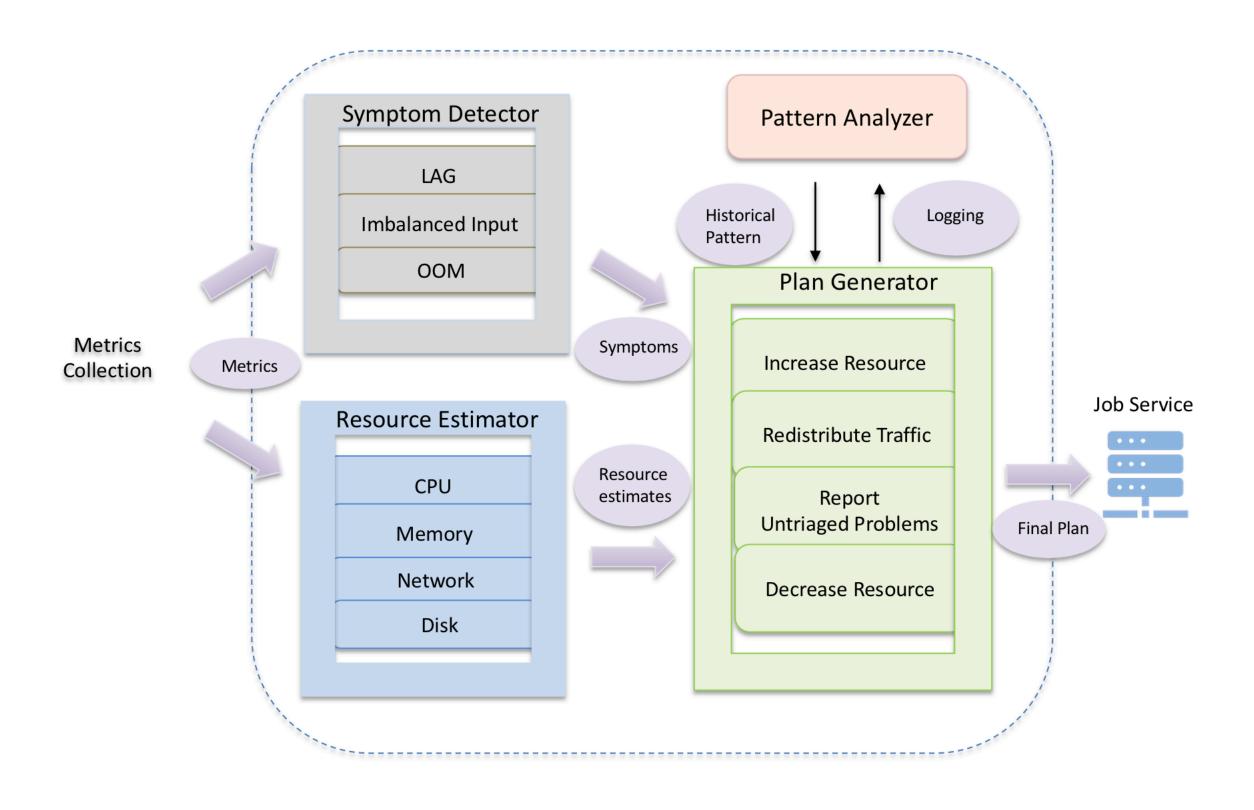
#### Reactive Auto Scaler — 1st Generation

#### Use **Symptom Detectors** & **Diagnosis Resolvers**:

- Lack of accurate estimation of required resources;
- Without knowing the lower bounds on the resource requirements for a given job;
- Making scaling decisions without understanding the root cause of a particular symptom may amplify the original problem.

The amount of resources needed for a given job is often predictable.

#### 2nd Auto Scaler Overview



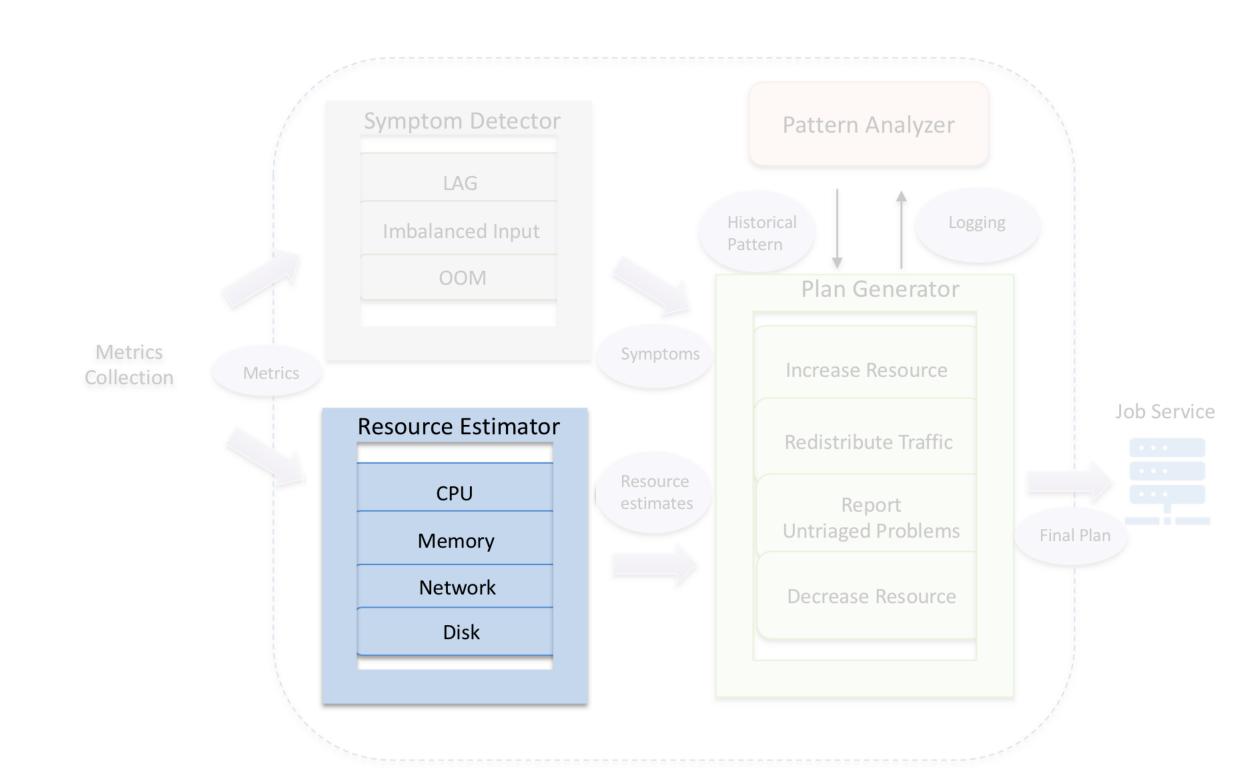
#### Resource Estimator

#### Stateless Job:

- Ofen CPU intensive;
- CPU% = (X + B/t) / (P \* k \* n);

#### Stateful job:

- CPU + memory + disk;
- Aggregation job the memory size is proportional to the input data kept in memory.
- Join operator the memory/disk size is proportional to the join window size, input matching, and input disorder.

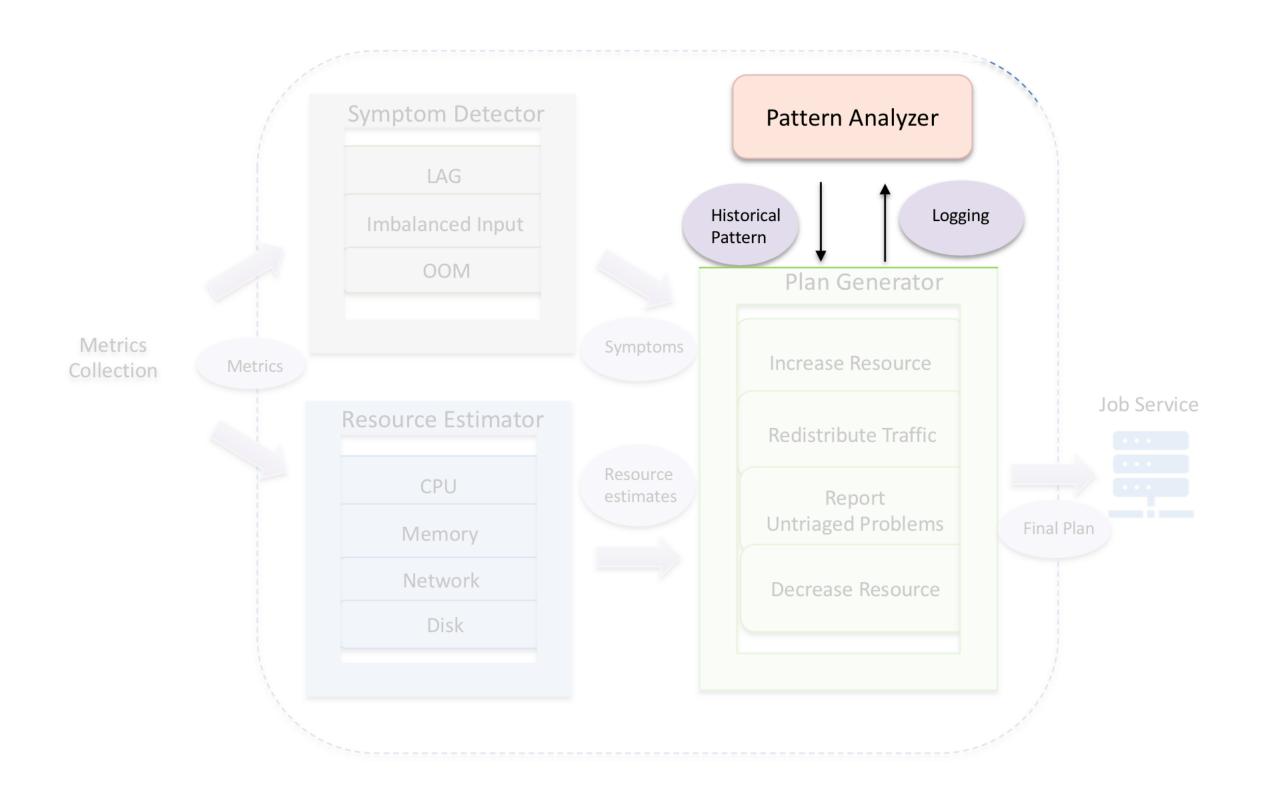


#### More on Paper ...

- Pattern Analyzer
  - Resource Adjustment Data
  - Historical Workload Patterns

Vertical vs Horizontal

Capacity Management



# Production Experience

#### Scuba Application

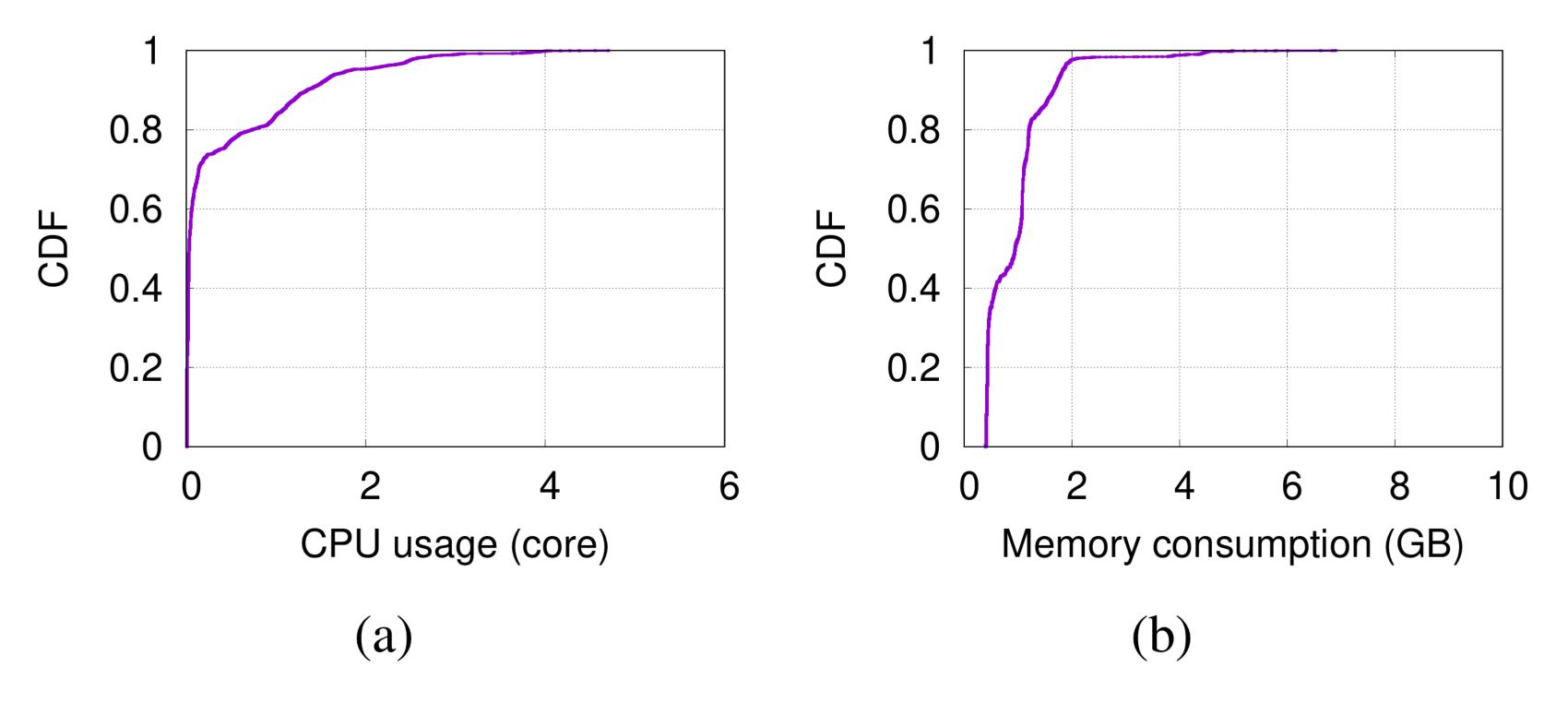


Fig. 5: CPU and memory usage of Scuba Tailer tasks.

#### Load Balance

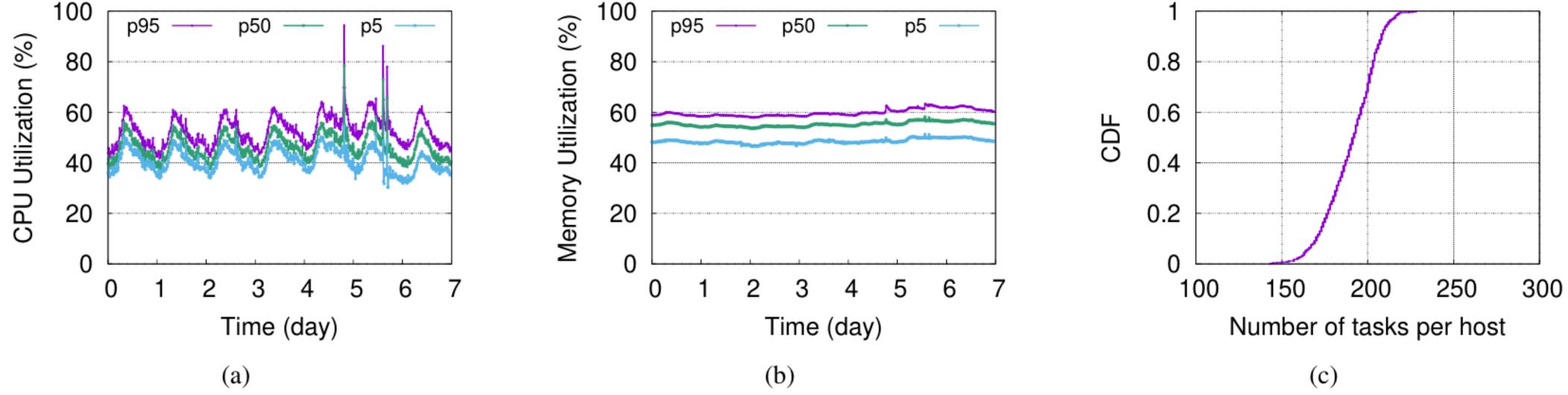


Fig. 6: In one Turbine cluster with > 600 hosts, CPU and memory utilization numbers are very close across hosts. With each host running hundreds of tasks, the reserved per-host headroom can tolerate simultaneous input traffic spike from many tasks.

### Workload Change

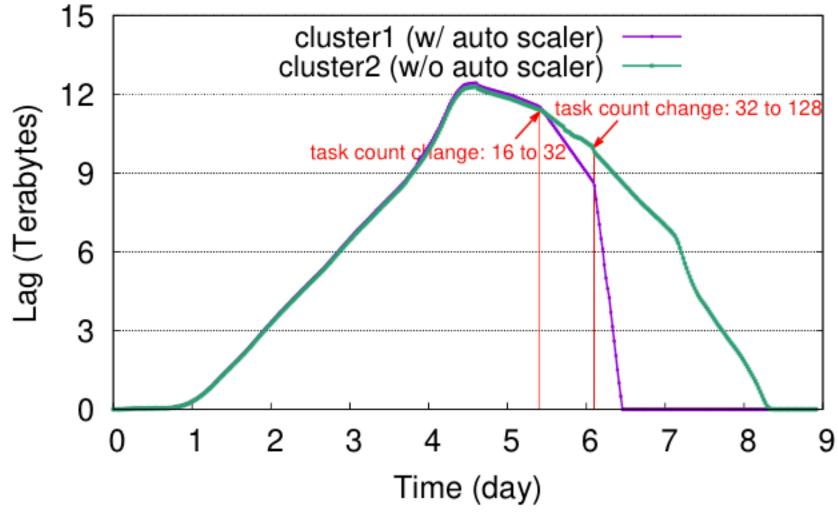


Fig. 8: Turbine's Auto Scaler helped a backlogged Scuba tailer job recover much faster.

### Workload Change

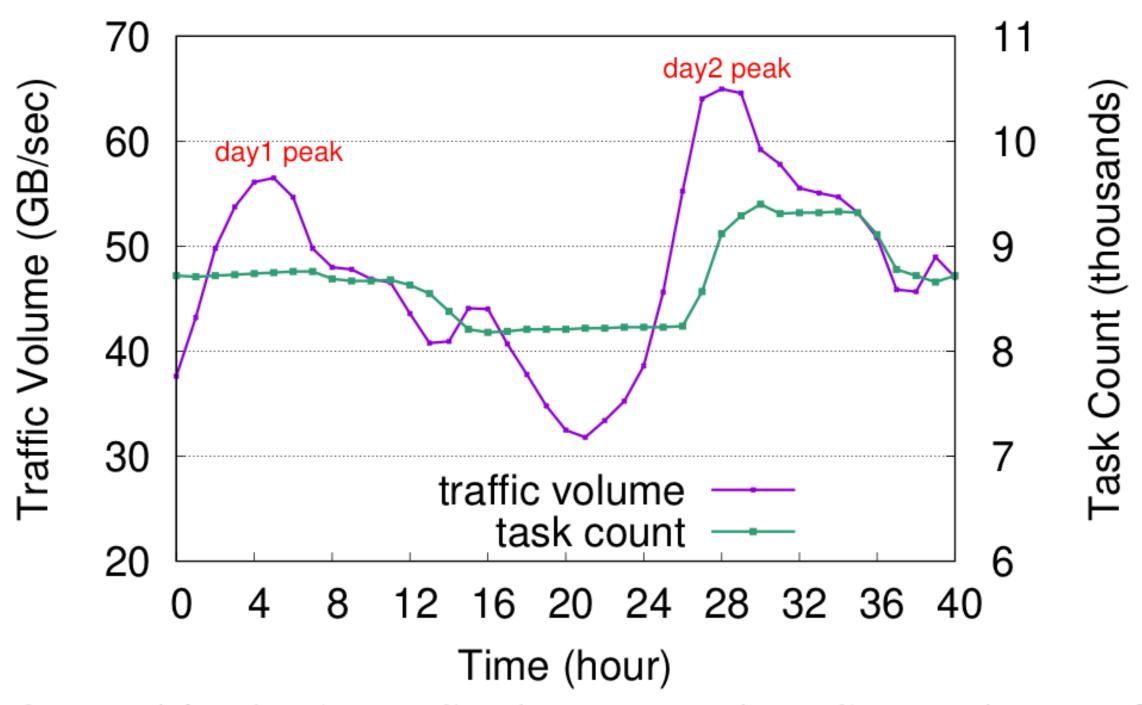


Fig. 9: Turbine's Auto Scaler reacted to Storm by performing horizontal scaling at cluster level.

# Thanks