

Hystax OptScale

MLOps open source platform

Multiply a number of ML/AI experiments
with minimal cloud costs



Hystax



Founded in 2016,
customers in 48 countries



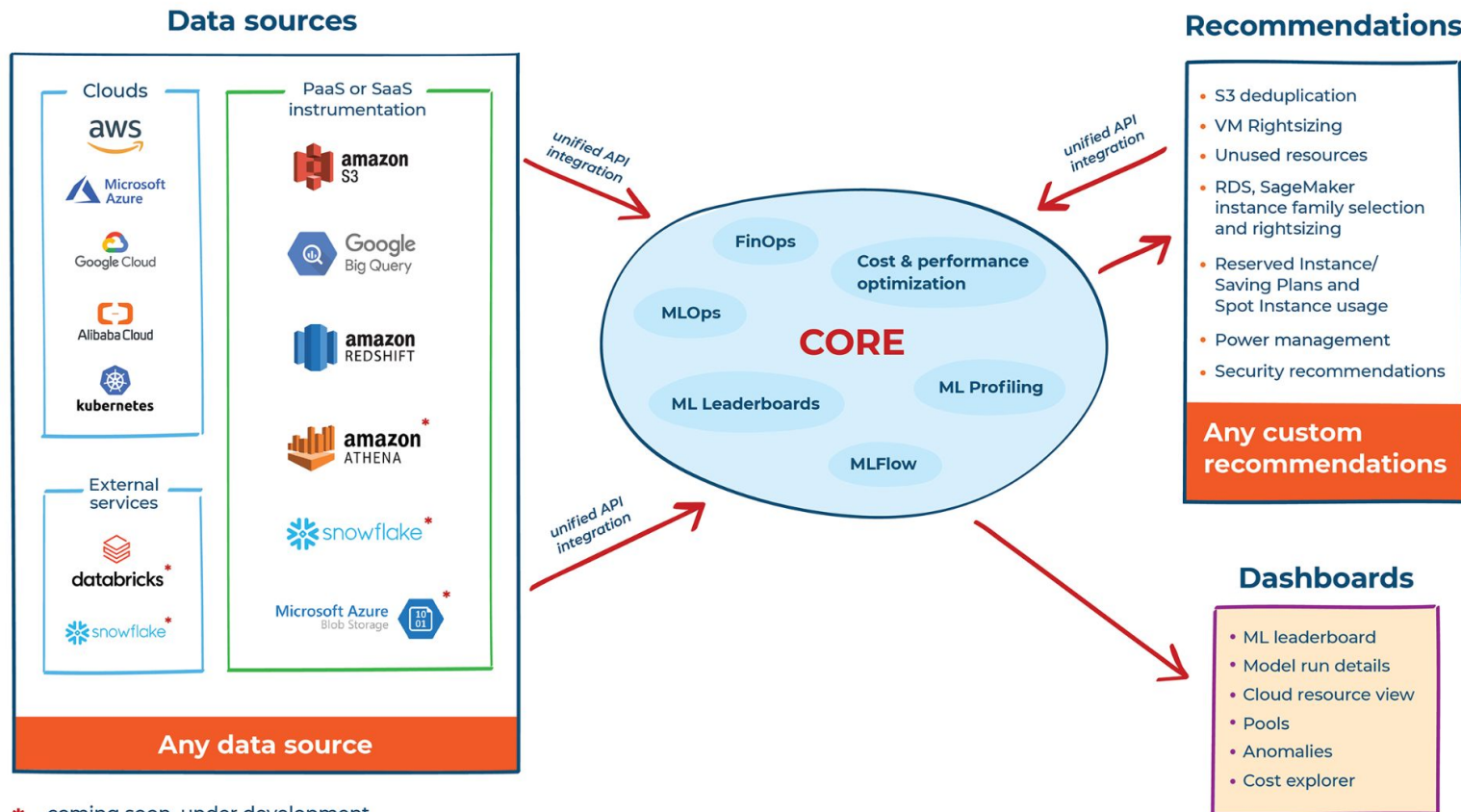
Customers: Airbus, Nutanix,
Orange, Nokia, DHL, Burger King

OptScale use cases



Open source under Apache 2.0: <https://github.com/hystax/optscale>

OptScale schema



* - coming soon, under development

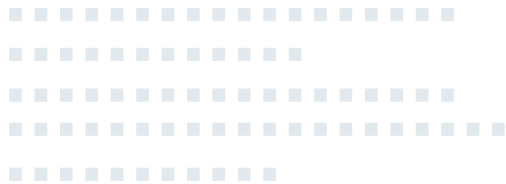
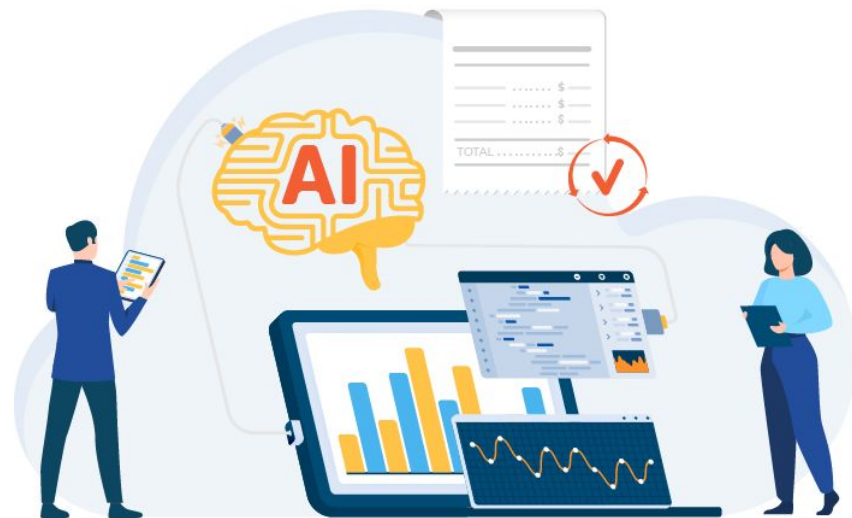
MLOps

- Team and individual ML engineer progress observability
- ML/AI task profiling, bottleneck identification
- PaaS or any external service instrumentation
- Optimization recommendations
- Runsets to automatically scale a number of experiments



ML R&D status observability

- List of models with goals status and active recommendations
- Tracking the number and quality of experiments run by a team
- Cost of an overall model and individual experiments



ML R&D status observability

optscale

Organization
Sunflower Inc.

- Home
- IT Environments
- Pools
- Resources
- OPTIMIZATION
- FINOPS
- PROFILING
- Applications
- Executors
- POLICIES
- SYSTEM

Applications

+ ADD
Filters: None
Owner ▾
Status ▾
Goals ▾

Name	Owner	Last run	Last run duration	Goals ?	Expenses
Shoes categorizer	Sally Wong	✔ Completed 12 hours ago	5 minutes, 59 seconds	Accuracy: ● 0.897 out of 0.999 ▼ 12% Data processed: ● 165 out of 150 ▲ 5% Inference time: ● 0.1 out of 0.2 ▲ 3.8% Data Loss: ● 15 out of 10 ▼ 7%	Total: \$1,278.47 Last 30 days: \$185.47
Image recognition	Geely Wong	✘ Failed 10 hours ago	3 seconds	Accuracy: ● 0.981 out of 0.999 ▲ 1.3% Data processed: ● 190 out of 150 0% Inference time: ● 0.22 out of 0.2 ▼ 10% Data Loss: ● 10 out of 10 ▲ 7%	Total: \$3,270.2 Last 30 days: \$205.7
Behavior prediction	Andy Well	✘ Failed 20 hours ago	3 seconds	Accuracy: ● 0.897 out of 0.999 ▲ 11% Data processed: ● 170 out of 150 ▲ 3.2% Inference time: ● 0.199 out of 0.2 ▲ 5% Data Loss: ● 5 out of 10 ▼ 9% Data corrupted: ● 2 out of 0 ▲ 1%	Total: \$5,111 Last 30 days: \$259.1
Goals met	Lucky Men	✔ Completed 6 hours ago	55 seconds	Accuracy: ● 1.1 out of 0.999 0% Data processed: ● 110 out of 150 ▲ 13% Inference time: ● 0.199 out of 0.2 ▼ 3%	Total: \$1,111 Last 30 days: \$601.5

MANAGE PARAMETERS

ML/AI profiling & optimization

- ML/AI model training tracking and profiling, inside and outside metrics collection
- CPU/RAM/GPU/Disk IO correlation tracking
- Minimal cloud cost for ML/AI experiments and development by utilizing Reserved Instances/Savings Plans and dozens of optimization scenarios

Supported platforms:



ML/AI optimization recommendations

- Utilizing Reserved/Spot Instances and Savings Plans
- Rightsizing and instance family migration
- Detecting CPU, GPU, RAM, and IO bottlenecks
- Cross-regional traffic
- Experiment/run comparison



ML/AI profiling & optimization

Application overview

Applications / Shoes categorizer

[PROFILING INTEGRATION](#) [CONFIGURE](#)

OVERVIEW EXECUTORS

Aborted

Status

1 minute, 5 seconds

Last run duration

\$284.25

Lifetime cost

\$8.48k

Summary savings

19

Recommendations count

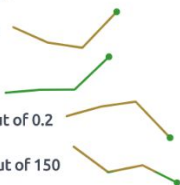
Tracked parameters

Accuracy: ● 1 out of 0.98

Data Loss: ● 5 out of 10

Inference time: ● 0.15 out of 0.2

Data processed: ● 132 out of 150



Application summary

Key: shoes_categorizer

Runs: 9

Last successful run: 3 months ago

Last run cost: \$14

Owner: Charlie Fisher

Last run executor

i-3920fedd

Cloud: AWS HQ

Name: sunflower-eu-fra-1

Region: us-west-2

Size: t3a.medium

Recommendations

Executors upgrade

[See details](#)

11

Count

Cross-region traffic

[See details](#)

\$8.48k

Savings

Spot/Preemptible instances usage

[See details](#)

3

Count

Local storage bottleneck

[See details](#)

1

Count

GPU memory

[See details](#)

1

Count

ML/AI profiling & optimization

IT Environments

Pools

Resources

OPTIMIZATION ▾

FINOPS ▾

PROFILING ▾

Applications

Executors

POLICIES ▾

SYSTEM ▾

Shoes categorizer / #2_relaxed_antonelli

OVERVIEW EXECUTORS

Completed

Status

16 minutes

Duration

\$872

Expenses

Executors summary

Data read: 109 B

Data written: 0 B

Tasks CPU: 2 hours, 13 minutes

CPU uptime: 2 hours, 13 minutes

Go to [executors list](#) for this run.

Goals

Accuracy: ● 0.9332 out of 0.94

Loss: ● 0.1221 out of 0.11

Inference time: ● 0.12 out of 0.2

Data len: ● 20329 out of 100

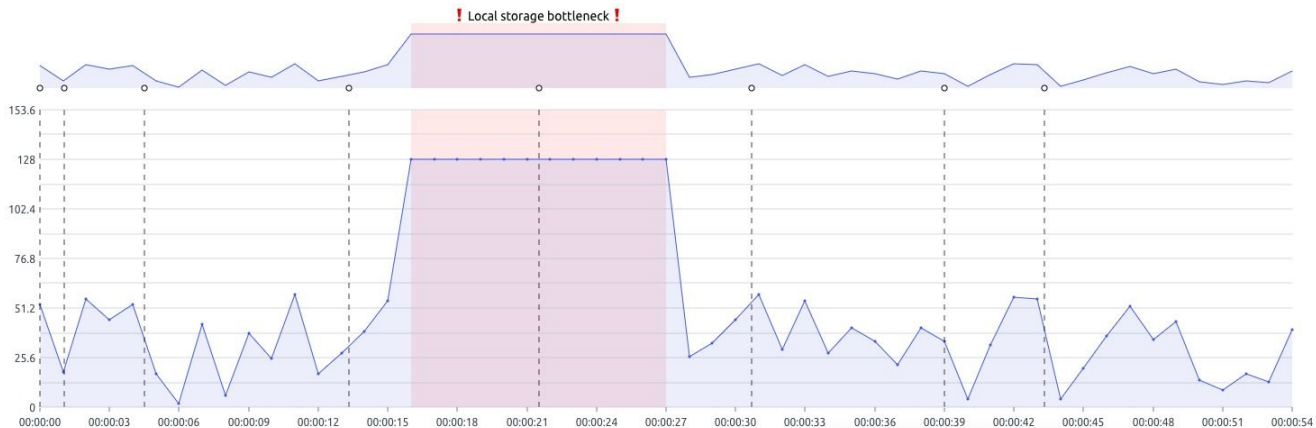
Image size: ● 1080 out of 300

Tags

Mode: Development

Execution

Host CPU
 Process CPU
 Host RAM
 Process RAM
 Disk read
 Disk write
 Network receive
 Network send
 GPU
 GPU memory
 loss
 accuracy



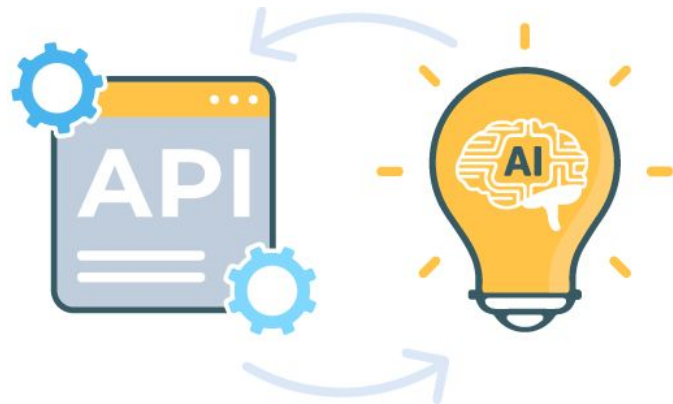
PaaS or any external service instrumentation

- Cost, performance, and output details of any API call to PaaS or an external service
- Metrics tracking and visualization
- Performance and cost optimization of API calls
- Cross-regional traffic
- S3, Redshift, BigQuery - ready, unified way to add more services



Runsets

- Automated run of a number of experiments with configurable datasets, hyperparameter ranges and model versions
- Optimal hardware with cost-efficient usage of Spot, Reserved Instances / Savings Plans
- Configurable experiment goals and success criteria
- Various complete/abort conditions - take first successful, complete all
- Integrated profiling to identify bottlenecks



Runsets

Runset overview

AWS GPU Instances / #3_gentle_sky

6

Configurations tried

1

Runs met goals

\$73.2

Total expenses

Application: **Shoes categorizer**

Parameters

Data source: AWS HQ

Region: us-east-1

Instance type: t3a

Maximum parallel runs: 14

Hyperparameters count

Model path: 2 ⓘ

Dataset path: 3 ⓘ

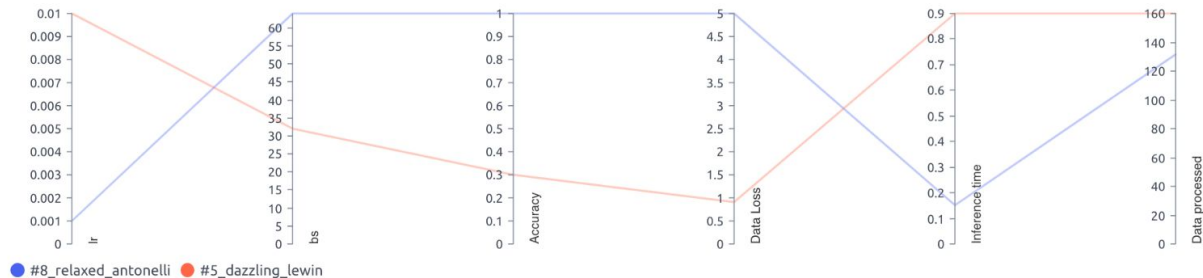
Learning rate: 4 ⓘ

Stop conditions

Stop runset when projected expenses exceeds 20\$

Stop individual run if its duration exceeds 3 minutes

Correlations



RUNS EXECUTORS

Filters: None | Status ▾ | Goals ▾

🔍 Search

# ↓	Status ⓘ	Hyperparameters	Goals ⚙️	Started at	Duration	Executors
#9_whispering_fog	🛑 Aborted Reached plateau for the Accuracy goal with the value of 0.87	lr: 0.01 bs: 32	Accuracy: - Data Loss: - Inference time: - Data processed: -	01/26/2023 05:55 AM	1 minute, 5 seconds	i-3920fedd Size: t3a.medium Expenses: \$14

Roadmap



- Cost plugin for MLflow, WanDB, and neptune.ai
- Integration with Optuna to optimize Reserved Instance and other hardware parameter usage
- Model versioning
- Better hardware selection recommendations based on usage patterns and algorithms

FINOPS & CLOUD COST OPTIMIZATION

FinOps and cost management

- Forecast and monitor an IT infrastructure cost
- Identify wastage and optimize IT expenses
- Bring resource / application / service observability
- IT asset management
- Set TTL and budget constraints
- Establish a long-term FinOps process by engaging engineering teams

Supported platforms:



OptScale vs cloud-native cost explorer

- Cloud resource visibility and filtering across all the clouds, accounts and regions
- Dozens of optimization scenarios not supported by clouds incl. one of the best rightsizing engines
- Cost allocation not just by tags but other properties
- Geo and network traffic map
- TTL rules and budget constraints
- FinOps: OptScale is built for engineers to be responsible for their cloud resources

Contacts

 +1 628 251 1280

 hystax.com

 info@hystax.com

 1250 Borregas Avenue, Sunnyvale, CA, 94089