

# Optimizing Asynchronous Successive Halving (ASH) for Hyperparameter Tuning

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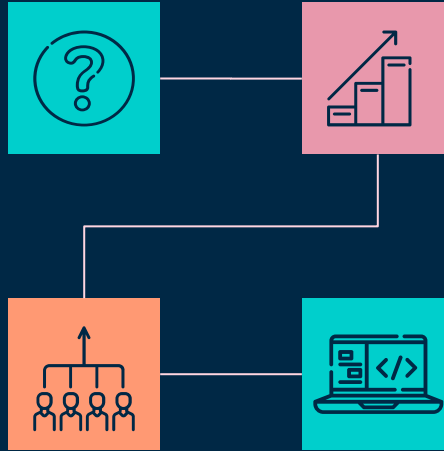
# OUTLINE

## PROBLEM

What are hyperparameters and their relevance

## WHY PARALLEL?

Motivation for parallelization and optimization



## MODEL AND DATA

Examining ASH and the Iris dataset

## IMPLEMENTATION

How we plan on parallelizing the app

# THE PROBLEM

## What are hyperparameters?

- ML models use them to influence **predictive accuracy** and **output**
- Examples include:
  - Learning rate
  - Dropout rate
  - Activation function

## Why are they important?

- Key to **performance improvement** and catering to **hypersensitive** ML models
- Current techniques include:
  - **Manual Tuning**
  - **Brute Force:** Grid Search and Random Search



# MODEL AND DATA

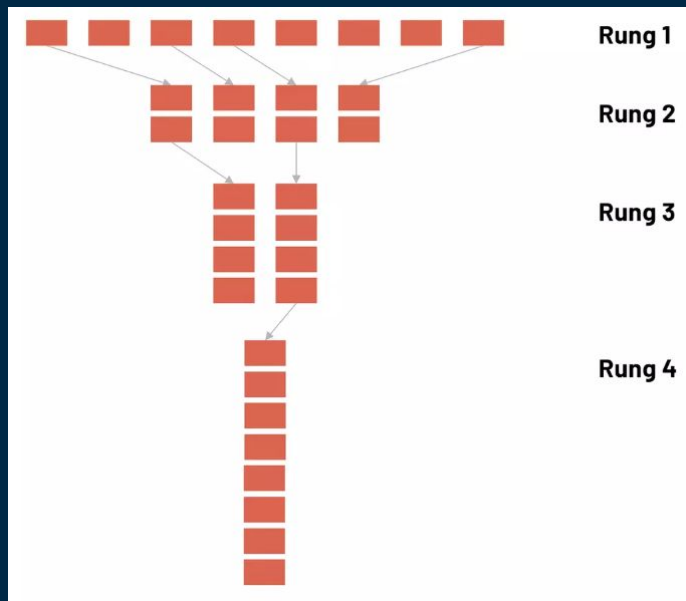


Figure 1. Job allocation and promotion in Successive Halving

## SEQUENTIAL BASELINE: SUCCESSIVE HALVING (SH)

- Uniform resource allocation across candidates per rung

## PARALLELIZATION MECHANISM (ASH)

- Training multiple candidate models simultaneously

## CONTEXT OF THE MODEL

- Optimizing an ML model's hyperparameters

## LIMITATIONS TO THE MODEL

- Robustness, computational cost

## DATA SOURCE AND TRANSFORMATION

- Iris Dataset – transformed from API to C++ vectors

# WHY PARALLEL?

High-Dimensional  
Search Spaces

Large, Complex  
Datasets

Scalability

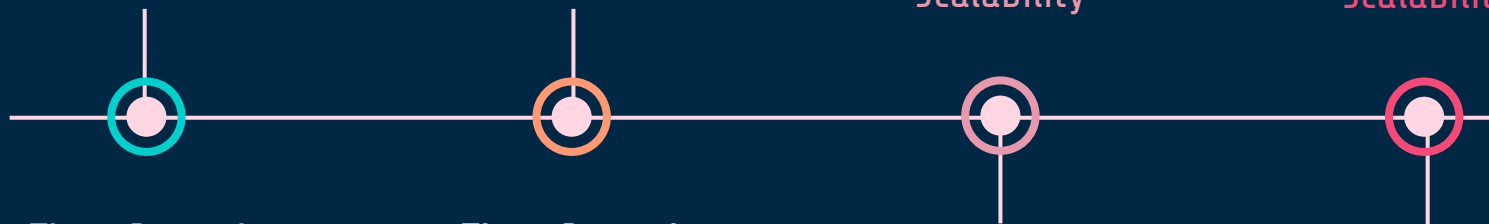
Scalability

Time-Intensity

Time-Intensity

Innovations in Parallel  
Computing

Rise of ML in Industry  
and Production



# IMPLEMENTATION PLAN

Shared memory –  
**multi-threading** the  
training process for  
individual candidate models

openMP

openMPI

**Simultaneously running  
jobs** and enables  
**message-passing** between  
multiple processors



**Concurrently calculate  
gradients** for batch gradient  
descent when training a  
model

Why openMP?

Why openMPI?

**Desynchronize jobs** and  
**communicate** which  
hyperparameter  
configurations are to be  
promoted to next rung

# WORKS CITED

Li, L. (2019, December 19). Massively parallel hyperparameter optimization. ML@CMU | Carnegie Mellon University. Retrieved March 22, 2023, from <https://blog.ml.cmu.edu/2018/12/12/massively-parallel-hyperparameter-optimization/>

Li, L., Jamieson, K., Rostamizadeh, A., Gonina, E., Ben-Tzur, J., Hardt, M., Recht, B., Talwalkar, A. (2020, March 15). A System For Massively Parallel Hyperparameter Tuning. Retrieved March 22, 2023, from <https://arxiv.org/pdf/1810.05934.pdf>

