

Hi!

CS 744: POLLUX

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Fall 2021

ADMINISTRIVIA

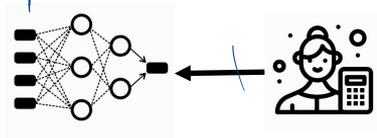
- Course project assignments
 - Emails will go out today (Oct 14th)
 - Introductions due Oct 25th → *LaTeX template*

- Midterm Exam
 - In class on Oct 28th
 - Includes everything from beginning to the end of ML

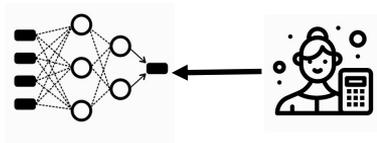
includes Pollex, Nexus

MACHINE LEARNING: TRAINING

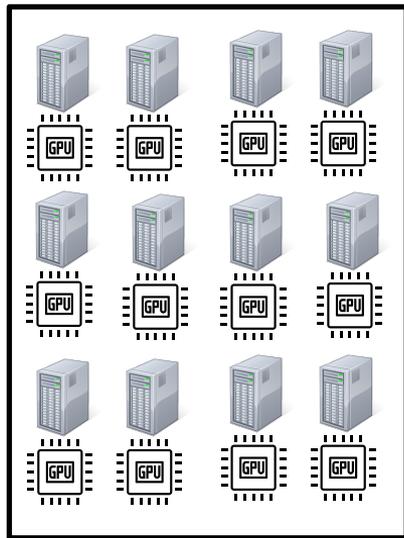
How do I parallelize this
What is the API



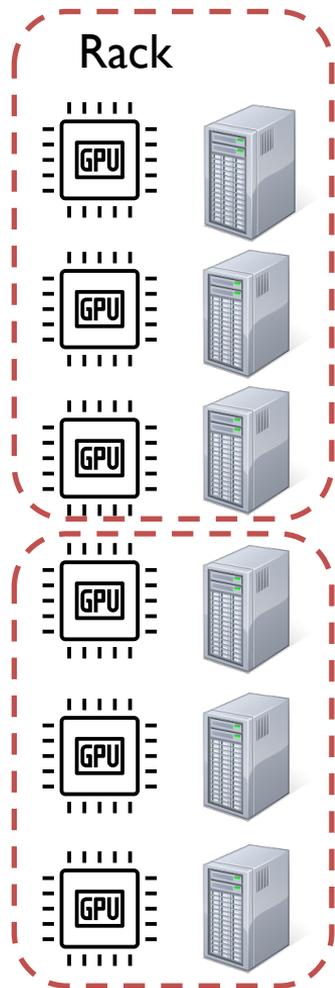
From one job
training



to
multiple jobs!



WORKLOAD CHARACTERISTICS



150 epochs ~ few minutes => many hours

Long running tasks

all the processes for a job need to run at the same time

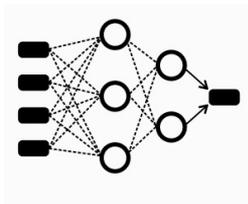
Gang scheduling

GPU Sharing?

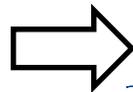
Isolation primitives
like containers to
split cores, memory

EXISTING DL SCHEDULER INTERFACE

Job 1

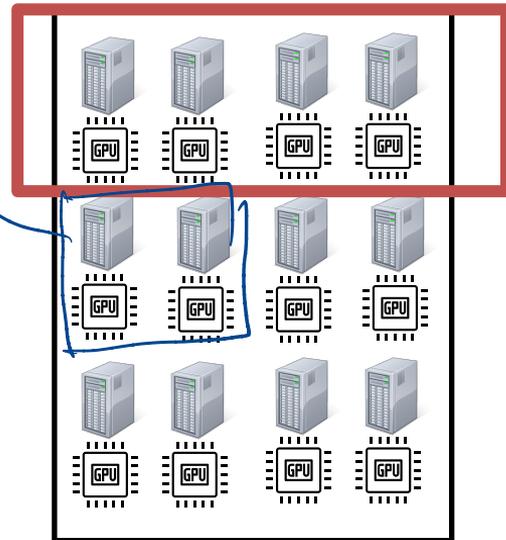
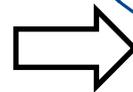


Run job Resnet18
With BatchSize = 64
on Num GPUs = 4



Goals:
Maximize throughput
Fairness
Minimize JCT
...

Job 2



Job 1

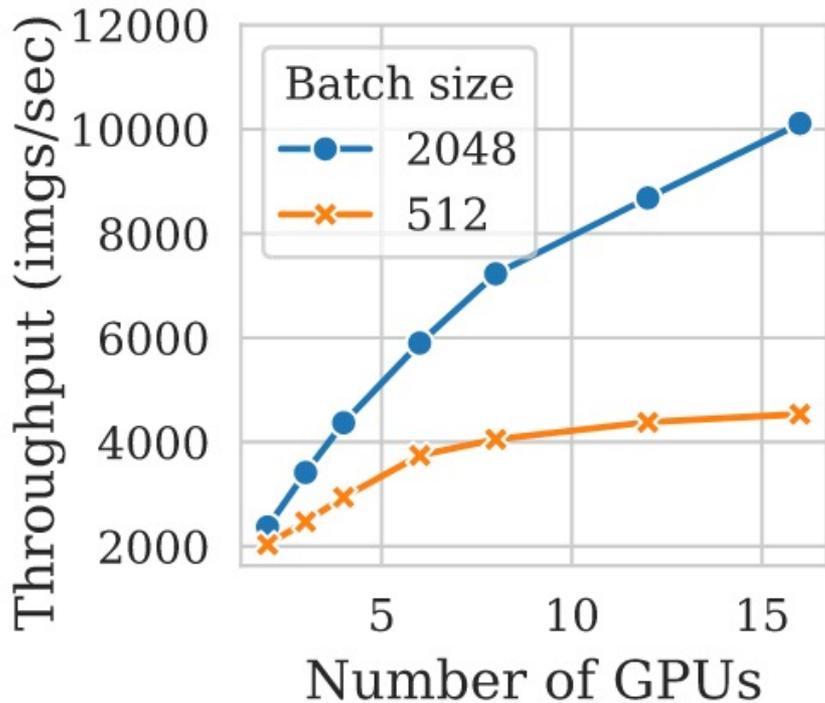
Job 2

num gpus = 2

$$2048 / 16 = 128$$

BATCH SIZE VS THROUGHPUT

ResNet18 on CIFAR-10



$bs = 512$ as you increase GPUs
tput flattens out

$$tput = \frac{BS}{T_{iter}} =$$

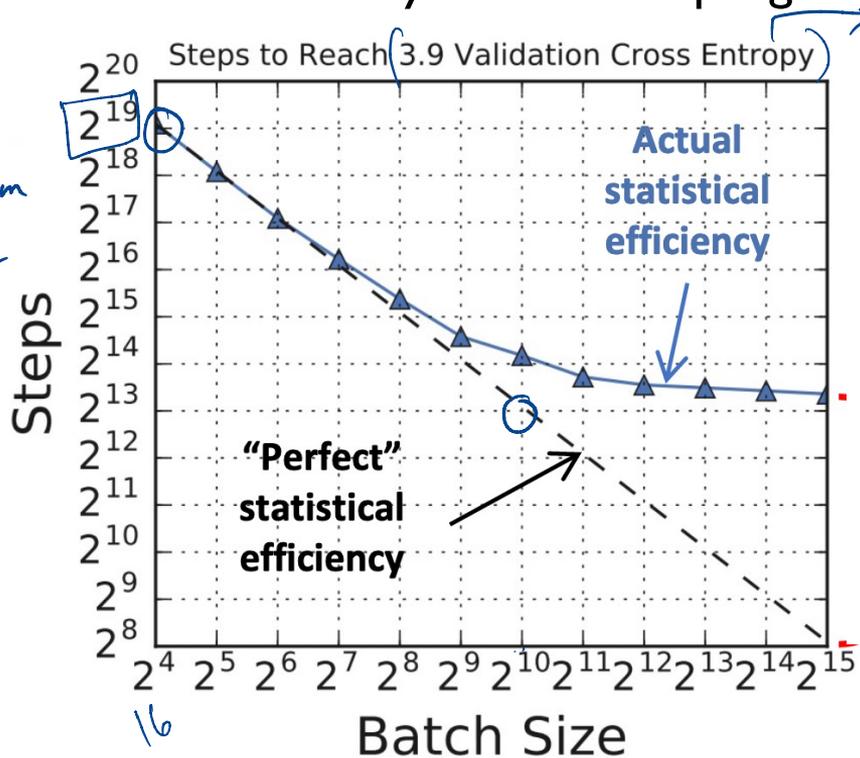
$$Time_{iter} = \frac{T_{grad}}{\quad} + \frac{T_{sync}}{\quad}$$

Inc num GPUs $T_{grad} \downarrow$
 $T_{sync} \uparrow$

The way to get better
tput with more GPUs
= inc. batch size!

BATCH SIZE VS STATISTICAL EFFICIENCY

Statistical Efficiency: Amount of progress made *per training example*



GPU ✓
↓
parallelism
in your
compute

$\boxed{\text{SGD}}$, goal to minimize loss
each iter "progress"

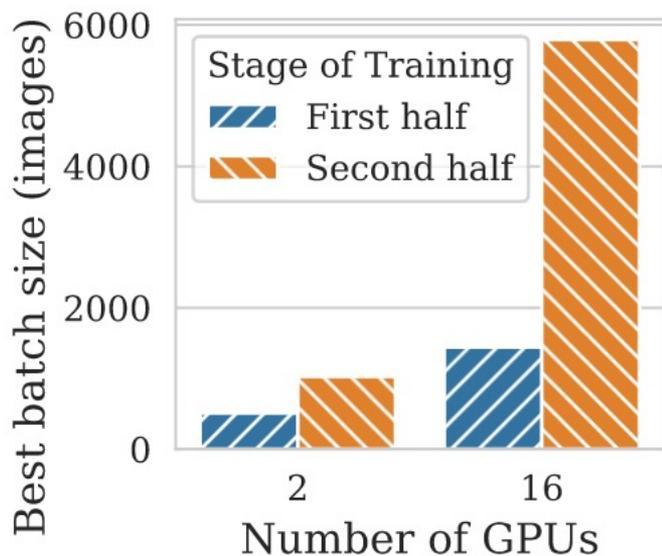
ideal scaling

= double batch size
then we take $\frac{1}{2}$ num steps

larger batch sizes
lead to lower statistical
efficiency.

GRADIENT NOISE SCALE (GNS)

Scale batch size *during* training based on gradient noise



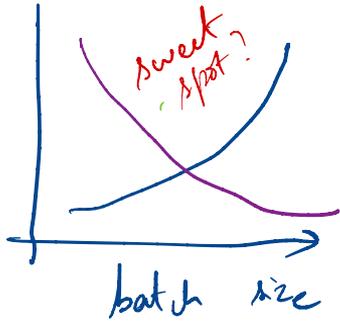
The optimal batch size can vary during training

GNS is correlated with optimal batch size (for stat. efficiency)

Initial iterations → smaller batch size

towards end of training → larger batch size

POLLUX: GOAL, APPROACH



Maximize goodput across all jobs in the cluster

by the scheduler

$$\text{GOODPUT}_t(\star) = \text{THROUGHPUT}(\star) \times \text{EFFICIENCY}_t(M(\star)),$$

Approach

1. Job submitted with initial batch size, num GPUs

2. Profile jobs during execution to model throughput, efficiency

3. Tune batch size/GPU and num GPUs based on resource availability

MODELING STATISTICAL EFFICIENCY

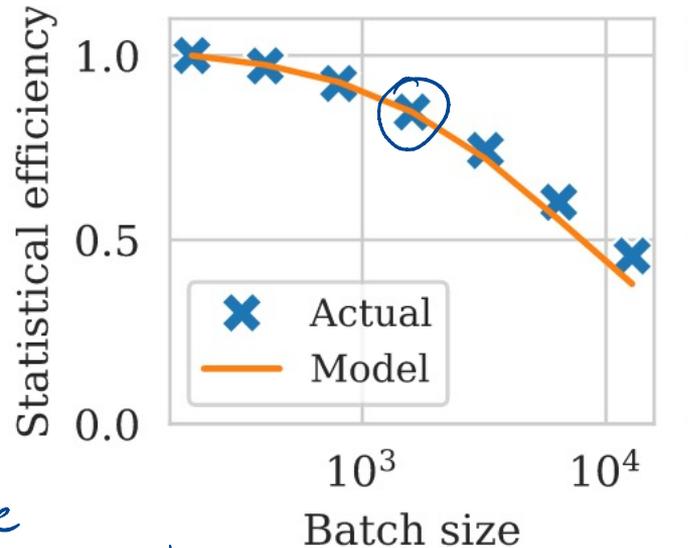
For a batch size M , initial batch size M_0

$$\text{EFFICIENCY}_t(M) = \frac{\varphi_t + M_0}{\varphi_t + M}$$

iteration number

φ_t : Computed using gradients noise
(GNS) using a **specific batch size**

*Run the job using M' as batch size
Measure φ_t at iteration t using m'
Use formula above to infer eff. at other batch sizes*



MODELING SYSTEM THROUGHPUT

num GPUs ← batch size GPU grad. back steps batch size time per iter

$$\text{THROUGHPUT}(a, m, s) = M(a, m, s) / T_{\text{iter}}(a, m, s).$$

fixed cost → per example cost

$$T_{\text{grad}}(m) = \alpha_{\text{grad}} + \beta_{\text{grad}} \cdot m,$$

$$T_{\text{iter}}(a, m, 0) = (T_{\text{grad}}(a, m)^\gamma + \underline{T_{\text{sync}}}(a)^\gamma)^{1/\gamma};$$

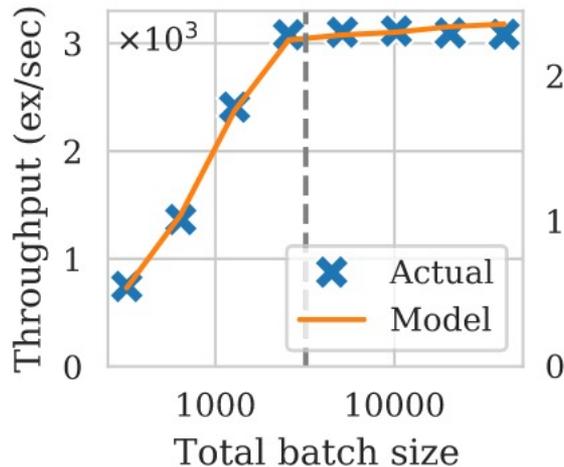
Account for overlap

$\gamma = 1 \rightarrow$ no overlap $\Rightarrow T_{\text{grad}} + T_{\text{sync}}$

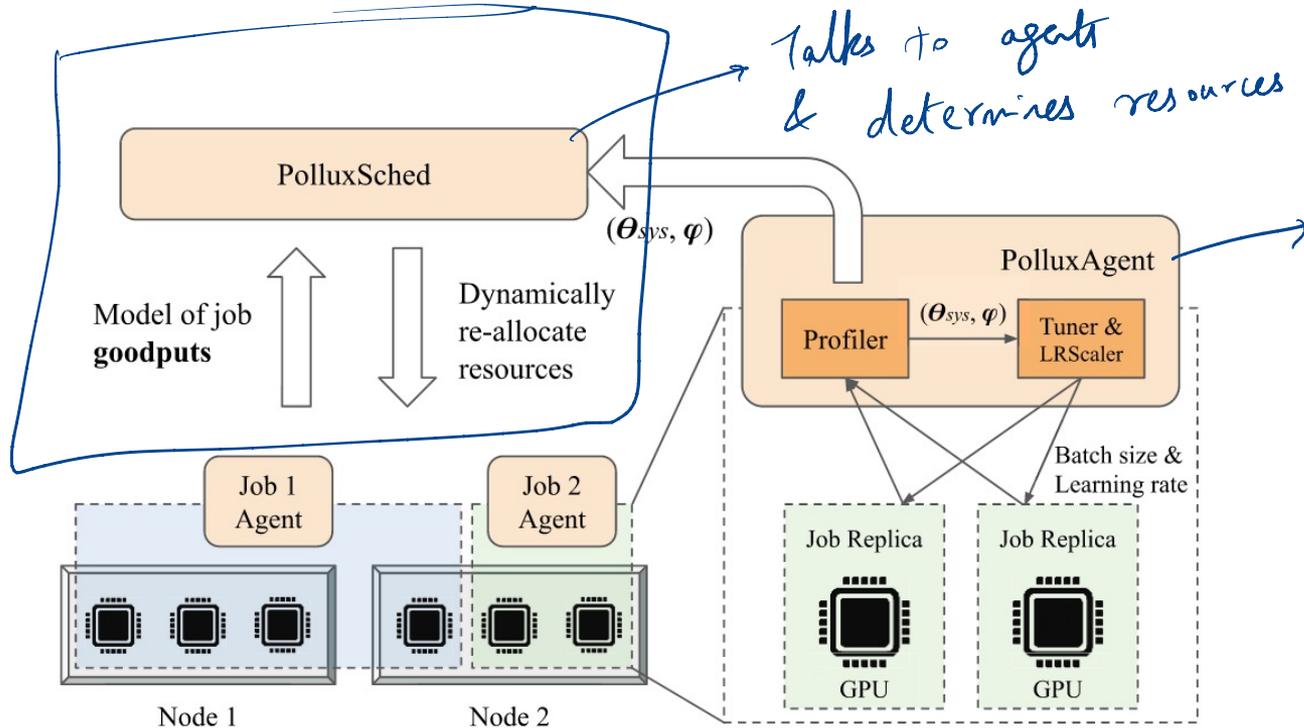
$\gamma = \infty \rightarrow$ perfect overlap $\Rightarrow \max(T_{\text{grad}}, T_{\text{sync}})$

Takeaway : Learn gamma, alpha, beta etc. by running the job

Predict throughputs!



POLLUX: ARCHITECTURE



POLLUX: CLUSTER WIDE OPTIMIZATION

→ optimization problem solved by the scheduler

Maximize fitness across all jobs

$$\text{FITNESS}_p(A) = \left(\frac{1}{J} \sum_{j=1}^J \text{SPEEDUP}_j(A_j)^p \right)^{1/p}$$

J jobs in my queue
 A_j = Allocation for this job

$p \rightarrow$ fairness knob

$p = 1$ then fitness is just the sum
 $p = -\infty$ minimum speedup \Rightarrow max-min fairness on SPEED UP
 $p = -1$ used in eval.

Solve to find A_j that will lead to max. fitness

Approach: Use a search algorithm (genetic algorithm) to find allocation

SUMMARY

DL Workloads Scheduling: Batch Size, Num GPUs

Pollux: Optimal batch size, num GPUs varies

Across jobs and during a job's execution

Models for system throughput, statistical efficiency

Cluster-wide optimization

DISCUSSION

<https://forms.gle/hQTrk53W3wwkEu9A8>

What are some similarities or differences between Mesos and DL schedulers like Pollux?

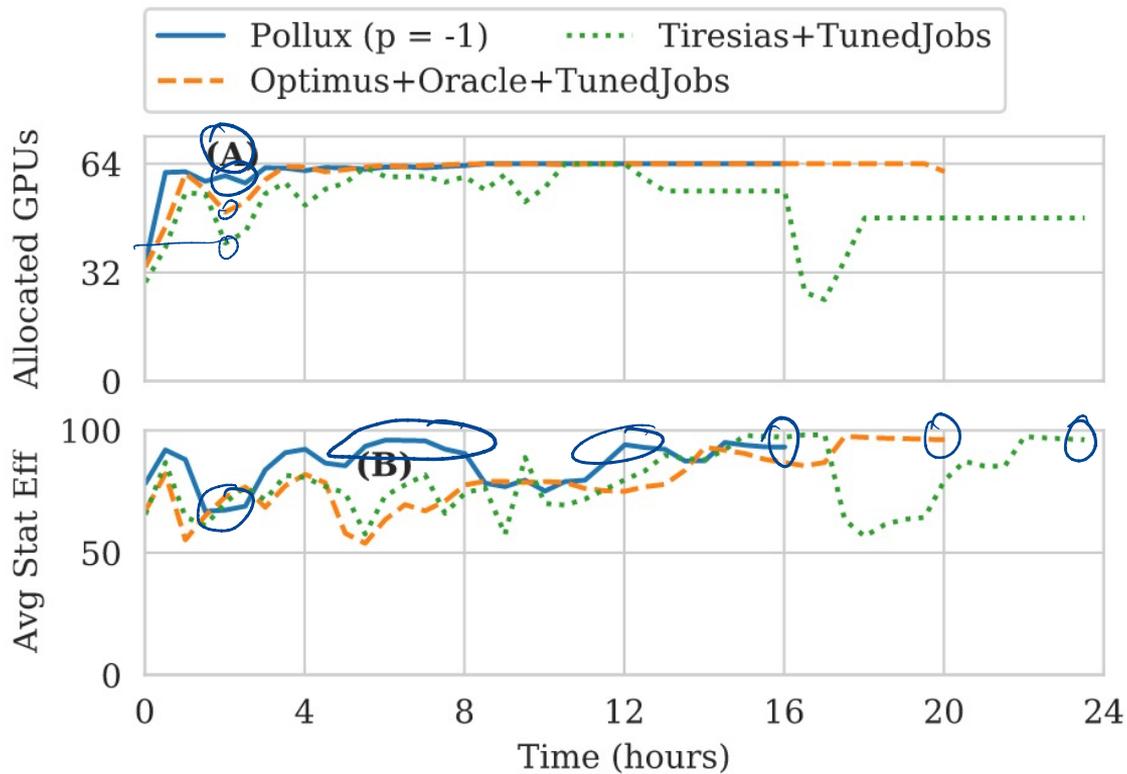
Similarity

- Goals like resource utilization
- Flexibility of workloads
 - ↳ Mesos: MPI, Spark
 - ↳ Pollux: only DL short jobs, long jobs.
- Decomposition
 - ↳ Job-level & Centralized

Diff

- Mesos only does allocation.
- Pollux allocation + Job config (batch size, LR)
- General API Mesos
 - ↳ rejecting offers to the job
- Pollux centralized sched. makes alloc.

At Point (A)
demand is
low \Rightarrow Pollux
can adapt
But statistical
efficiency is
not higher!



Pollux is
finishing
faster
higher
statistical
efficiency

NEXT STEPS

Next Class: Nexus

Course Project Introductions!

Midterm after that