... and here is the very stone that finally enabled us to figure out what all those dogs and cats were saying. Accelerated AI Algorithms for Data-Driven Discovery 000 200 mao .00 E Streeter 100004 11 Xxxx

#### ML@MIT

P. Harris, E. Katsavounidis, M. Coughlin









## Despite differences in language, there is a common theme MLOMIT

P. Harris, E. Katsavounidis, M. Coughlin

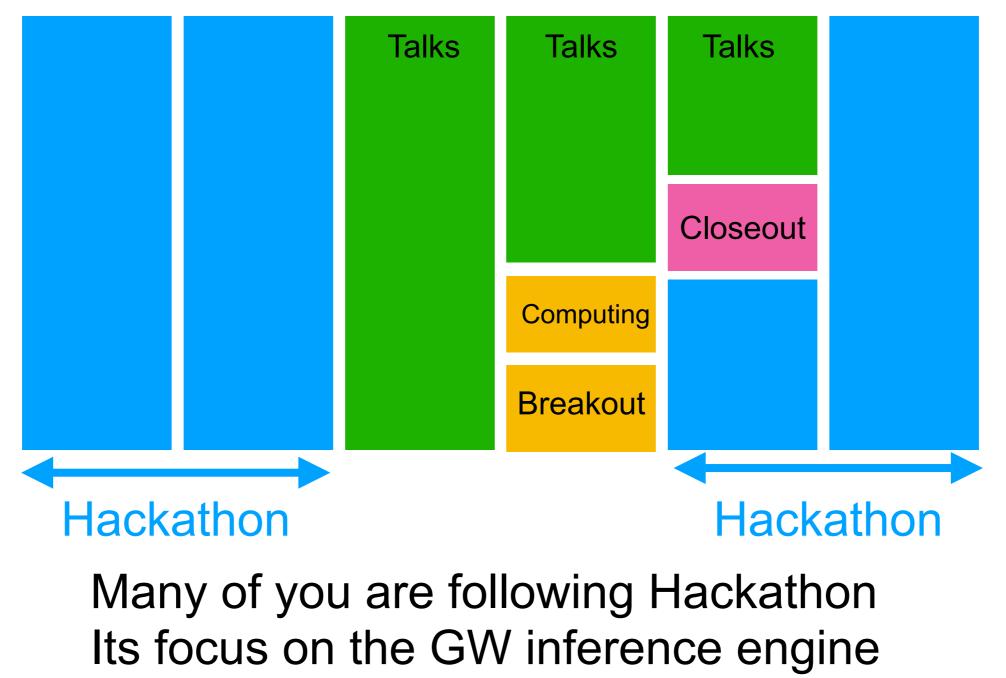
#### Welcome!

- We are excited that you have all come to the meeting
  - We hope to have a lot of interesting discussions
- We thank you for taking the time to come to this meeting
- This meeting is the result of ongoing discussions
  - Many of you are involved in our efforts
    - Astronomy, Gravitational Waves, Particle Physics
- Despite coming from many different venues
  - We all share a common goal and vision
  - We often share computers and code

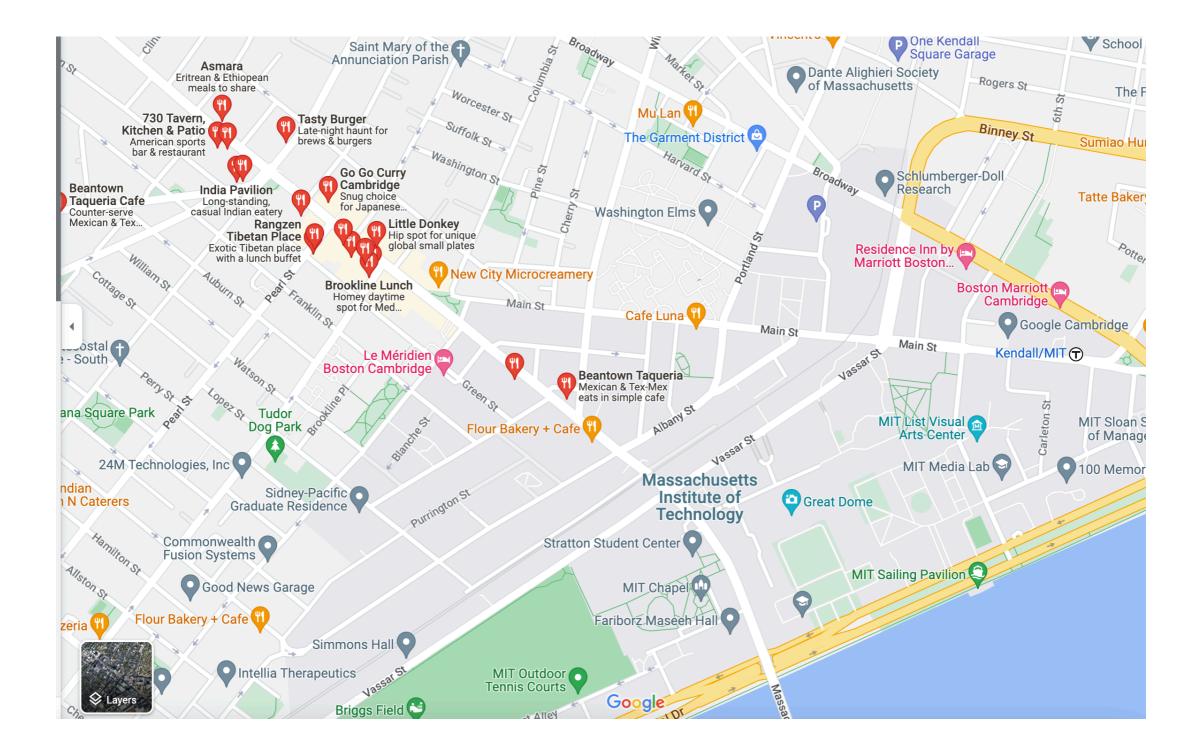
#### Agenda

#### Full Agenda is available here :

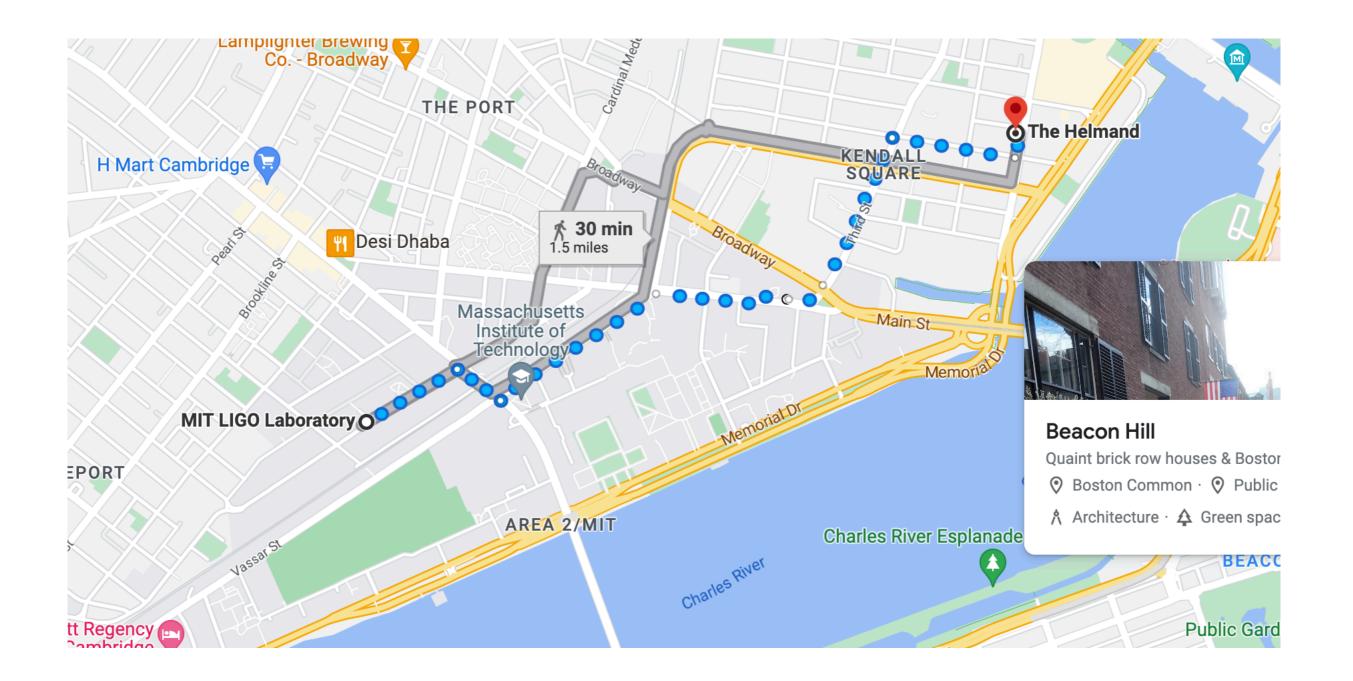
https://indico.cern.ch/event/1224718/timetable/#20230130



#### Cambridge

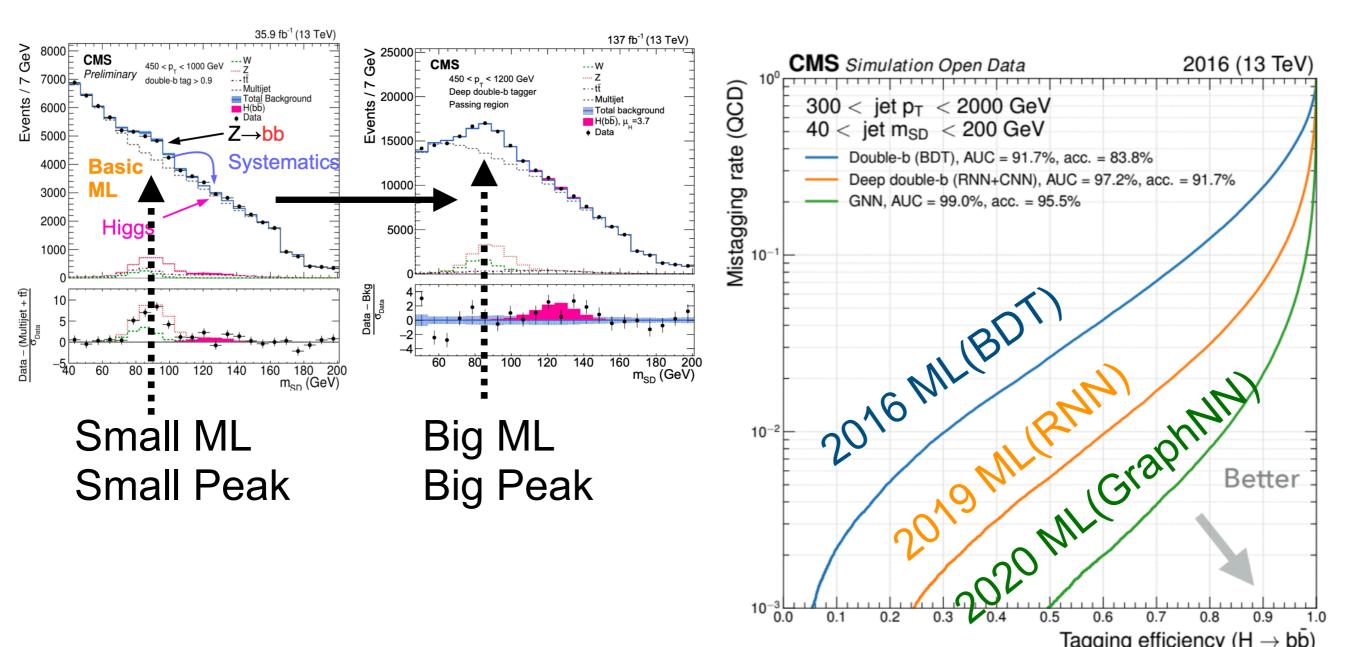


#### **Dinner Location**



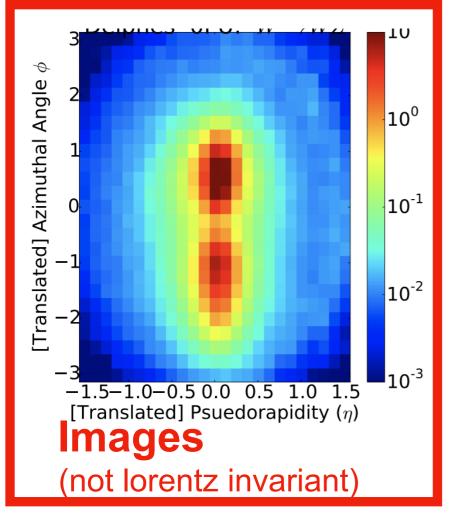
#### An Angle on AI revolution

- Things are starting to change in the way we compute
  - ML algorithms have the ability to go beyond algorithms
    - This is also b/c GPUs have helped to parallelize computation



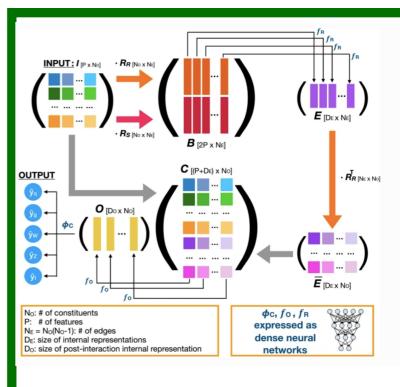
#### **Deep Learning Progression**

#### **2016**



# 2018 Image: product of the second ary vertices Image: product of the second ary vertices

#### 2020



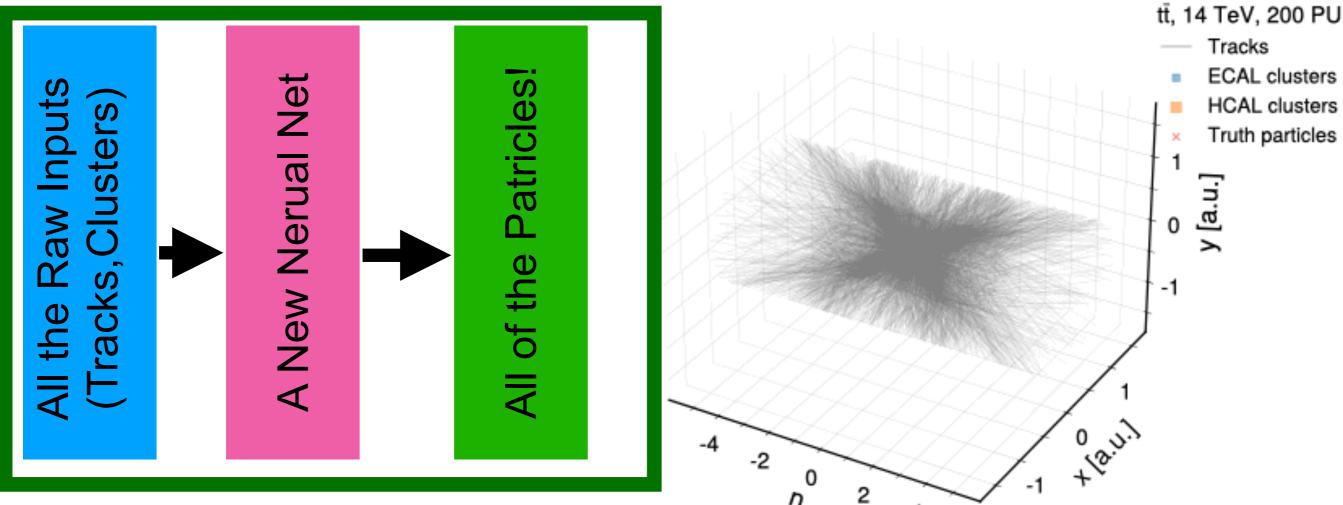
#### **Graphs** (Particles+correlations)

Progressively moving towards use of more info

#### What does this mean?

9

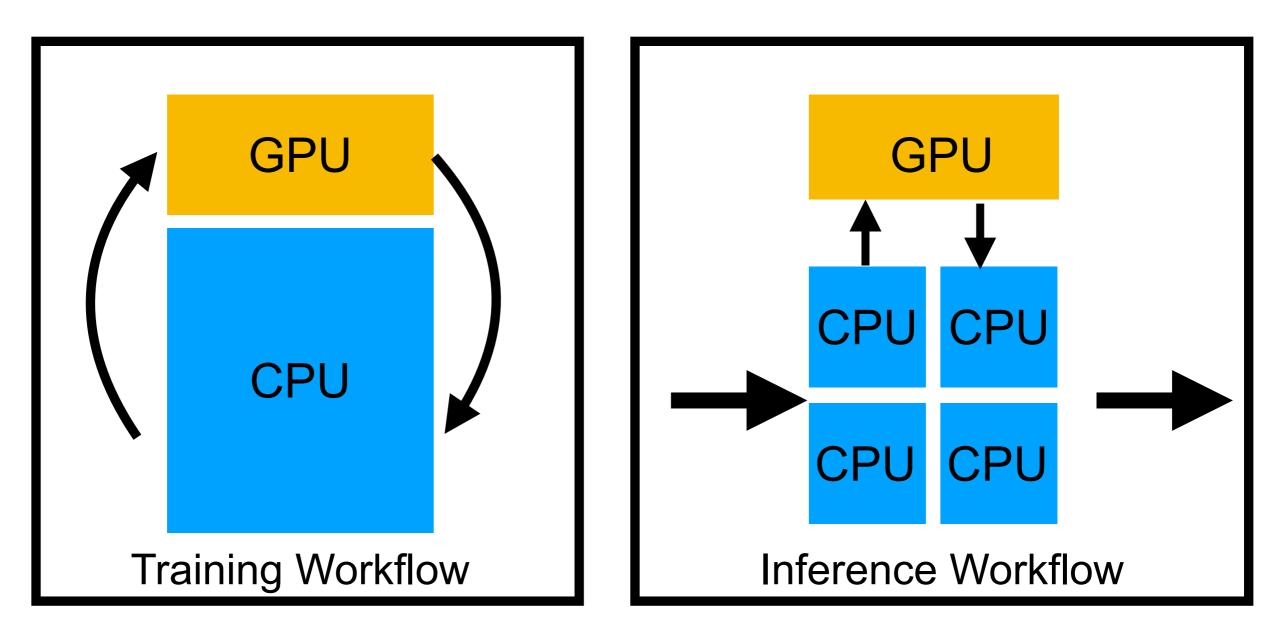
• Inevitable that our algorithms will become progressively larger



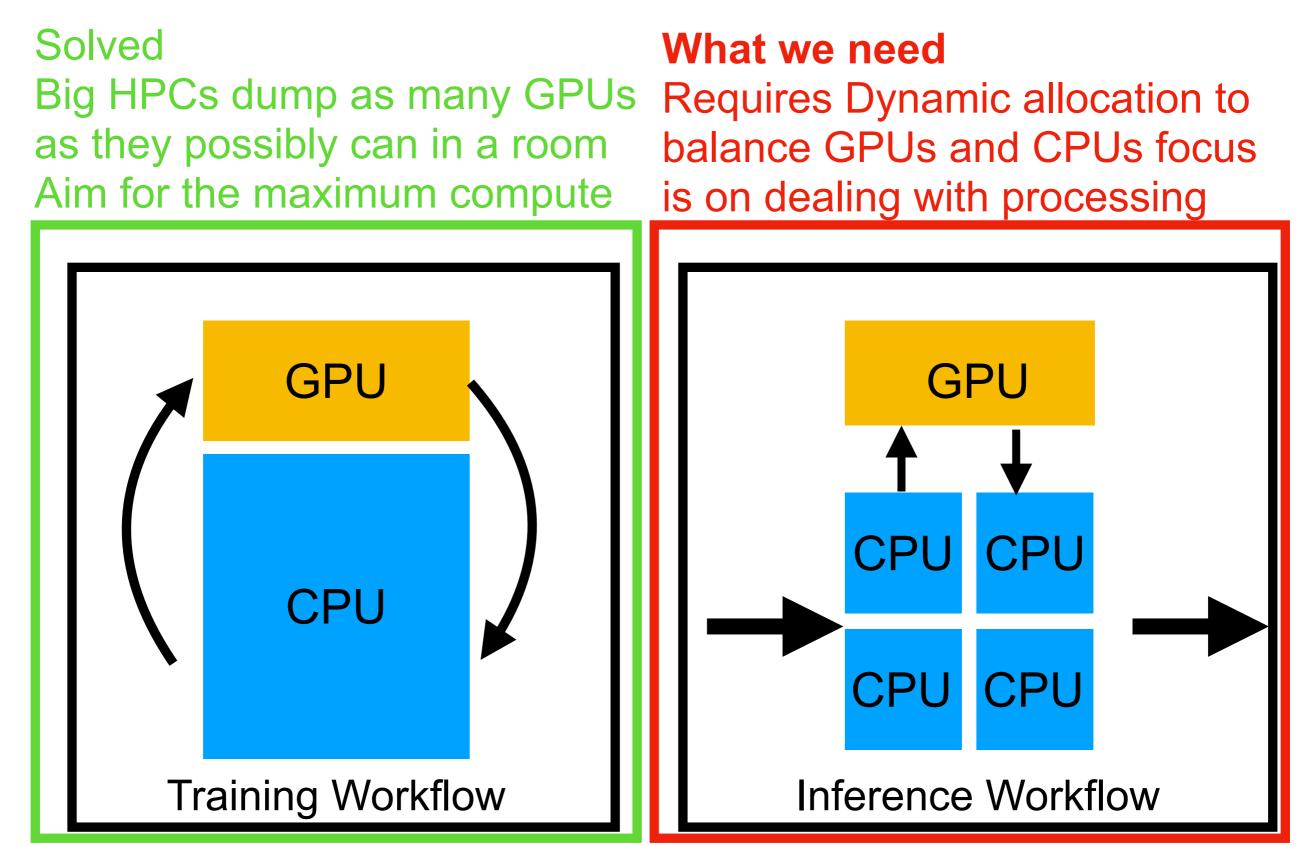
All particles in on fell swoop

## Algorithm Needs

- With the development of AI algorithms we need two things
  - Training and Testing
  - Processing power to run on the data



## Algorithm Needs



## Algorithm Needs

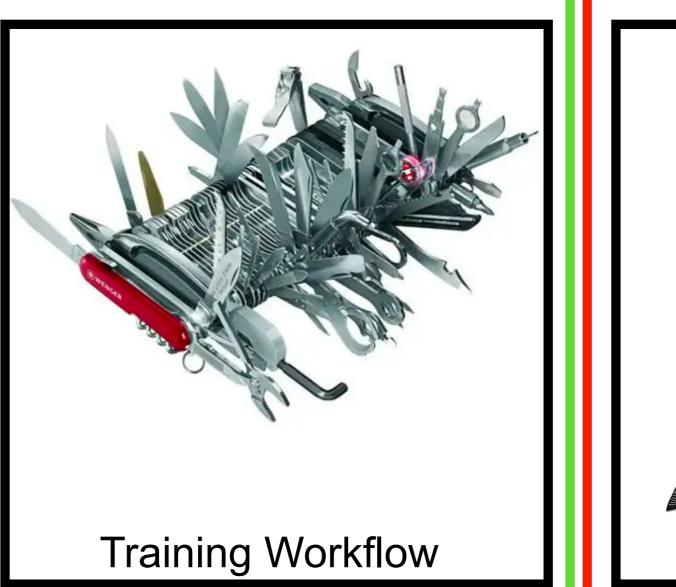
Solved

Big HPCs dump as many GPUs as they possibly can in a room Aim for the maximum compute

#### What we need

Requires Dynamic allocation to balance GPUs and CPUs focus is on dealing with processing

12





## Why are we here?

- We would like to highlight commonalities across domains
  - Computing demands
    - Looking for computing infrastructure for ML science deployment
    - We can assemble a list of common hardware(+tools)
  - Software Stack
    - With all ML algorithms aim for a set of core software tools
    - Need for good tools to validate and deploy algorithms

#### ML Problems

- Across the domains similar ML problems exist
- Highlighting the similarity is critical

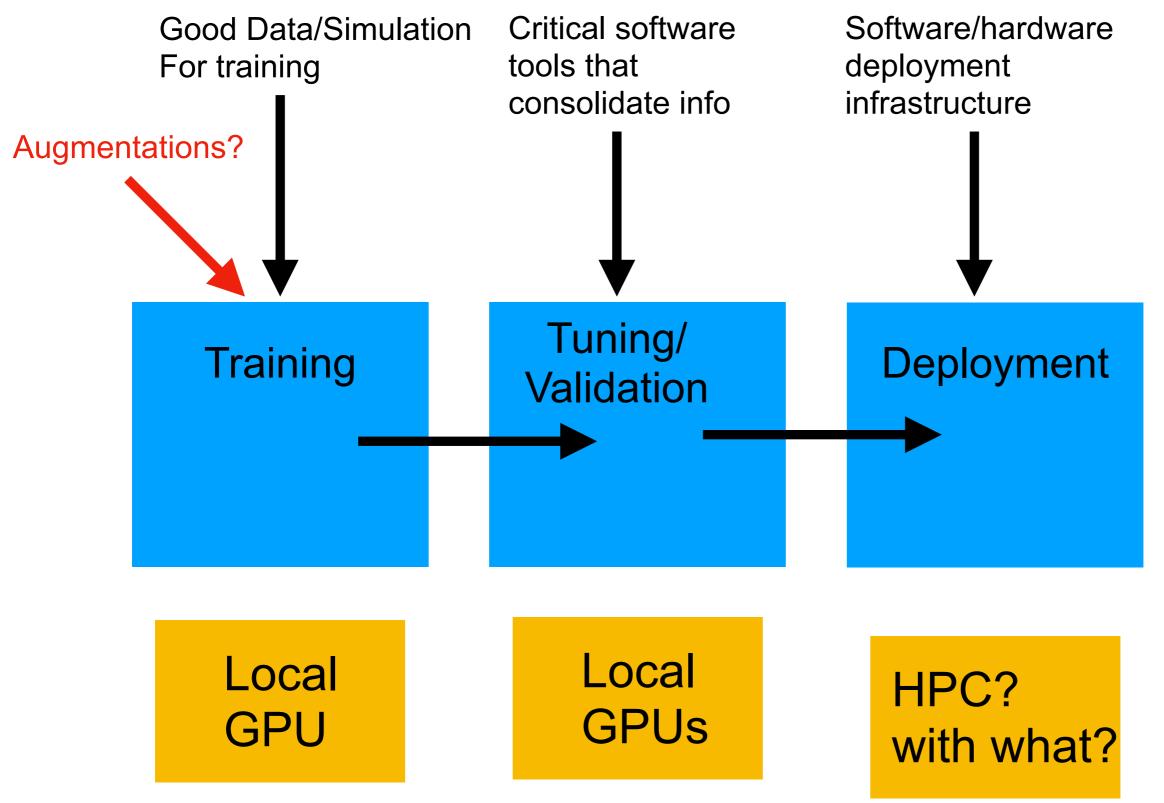
## Why are we here?

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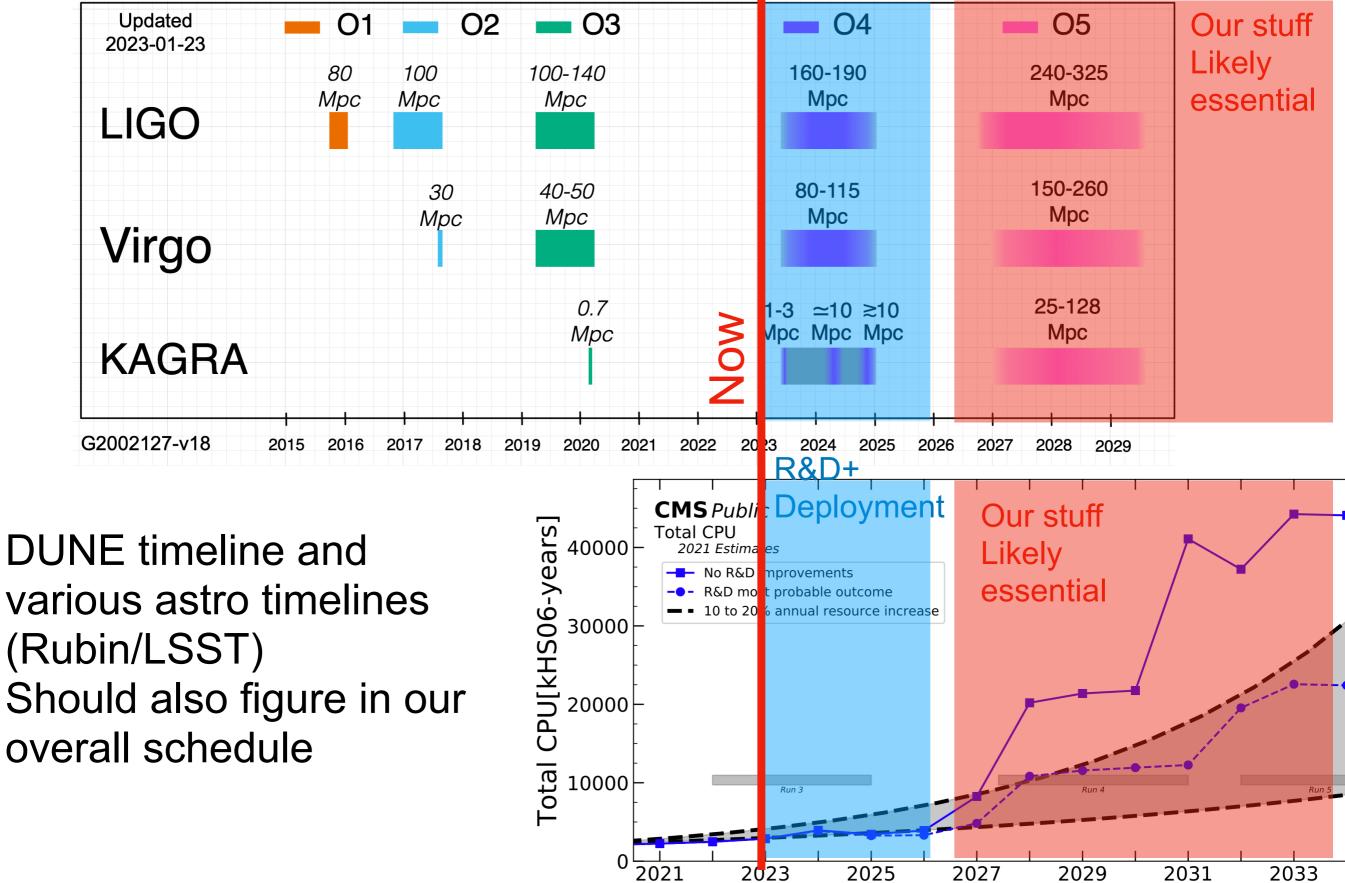
Inlighting the similarity is critical

ince deployment

#### Anatomy of an Algo



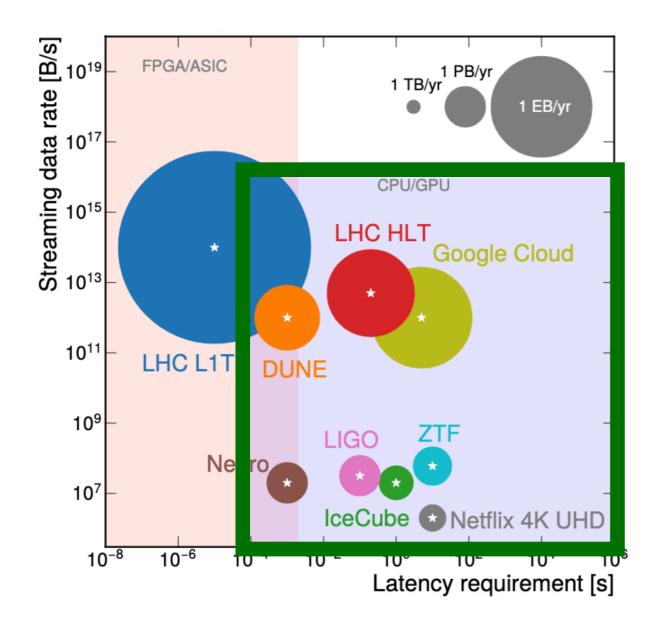
#### Timelines



Year

#### What computes are here?

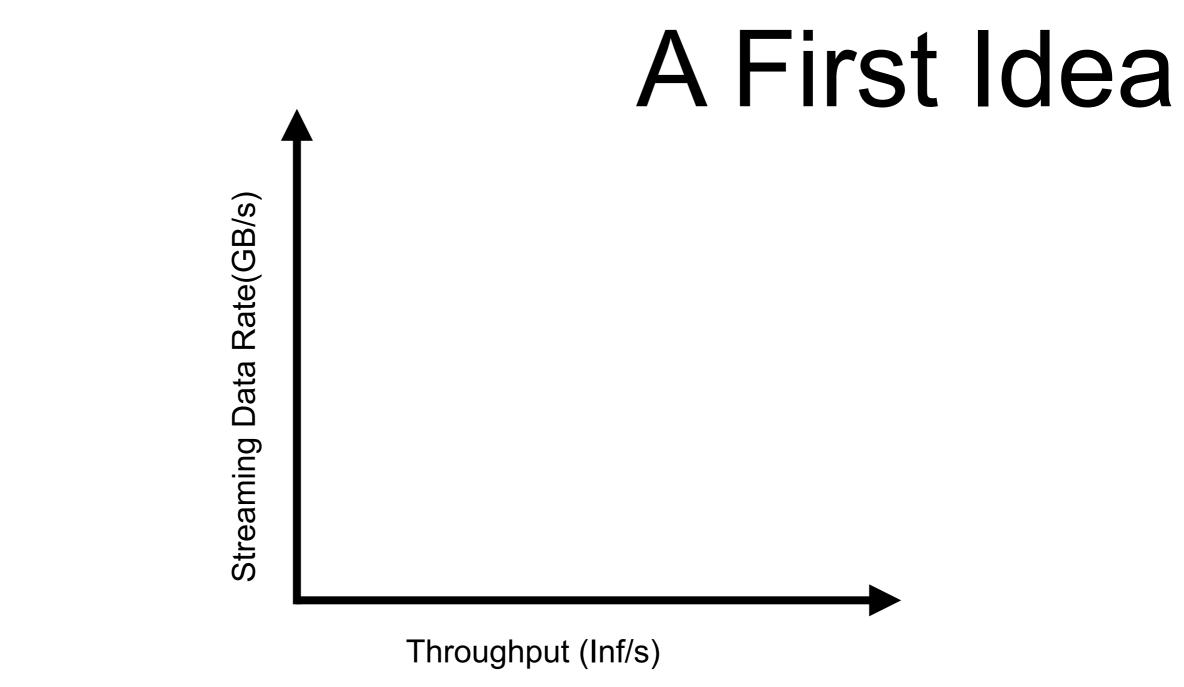
- Within the FastML Community there is a broad range
  - We often try to characterize this range by customization
  - Low Latency and Low Power need more customization



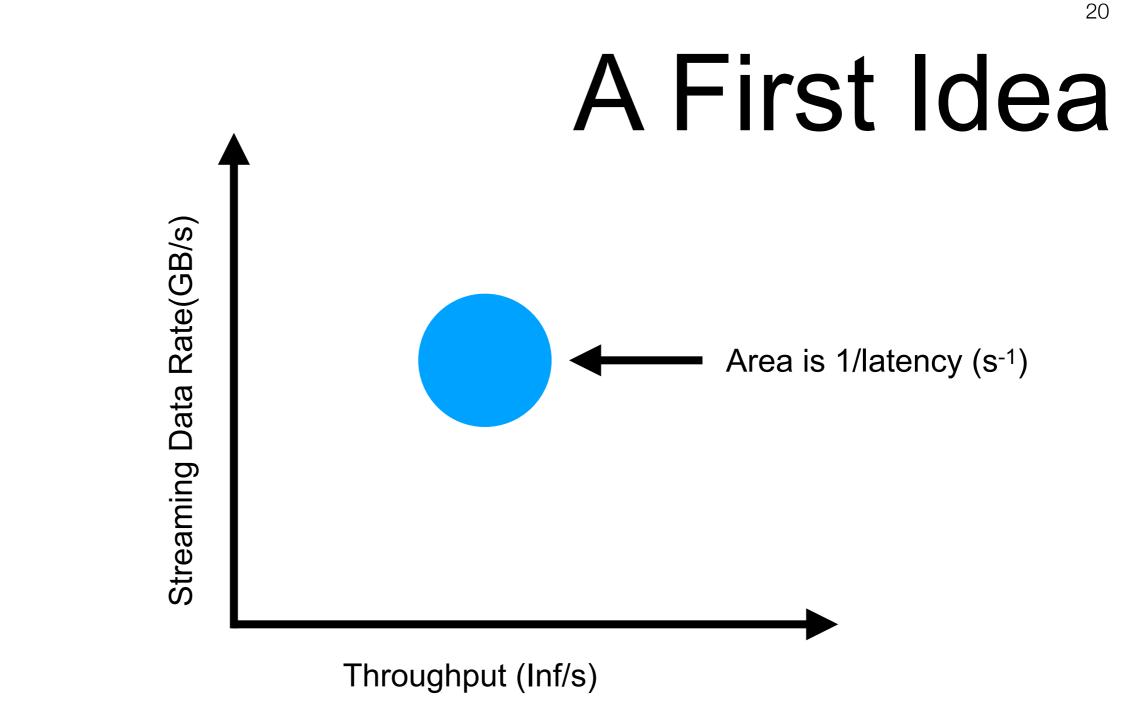
This is our focus here We want to understand the high throughput component

## Visualizing Computing

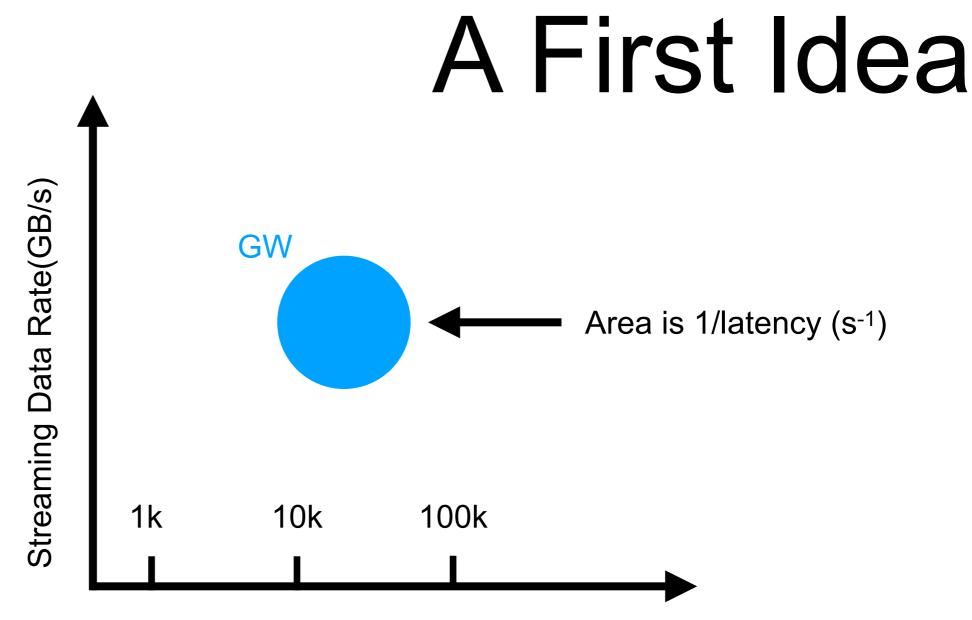
- All of us in the room require at least one thing in common
  - Computers
  - Also, with GPUs/Coprocessors to accelerate things
- As part of this workshop we would like to create a graphic
  - This graph illustrates the computing demands
  - We hope this graphic can be used as a motivator
- The A3D3 graphic has gotten a lot of traction
  - Highlighting the specific challenges for this conference helps
  - Would like to share this with HPCs as a motivator



- This work is not at the cutting edge of latency
  - But speed and overall processing time is important
- We are typically concerned about throughput

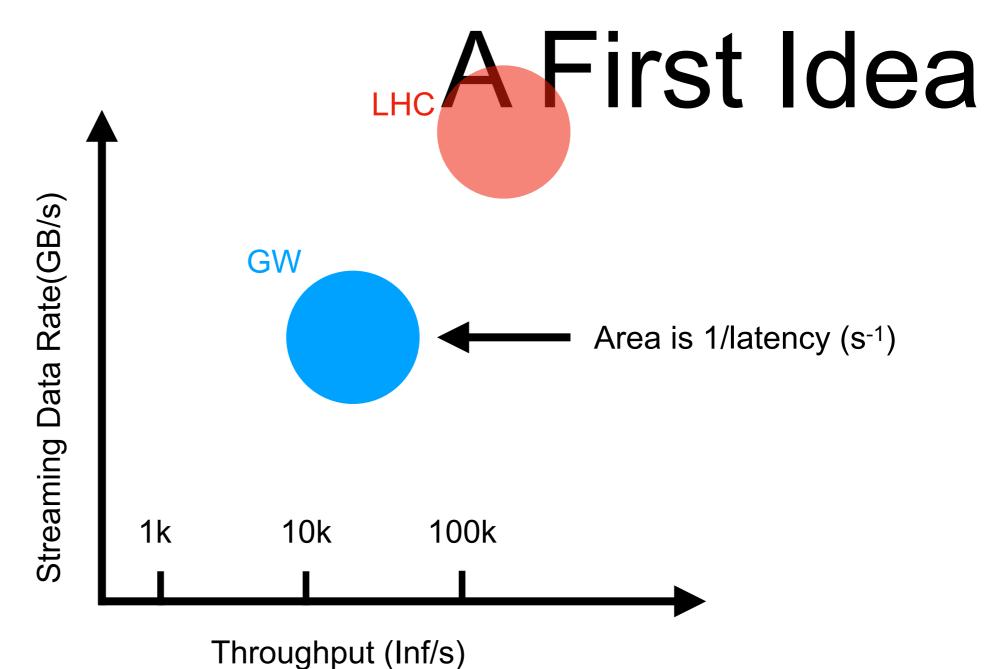


- This work is not at the cutting edge of latency
  - But speed and overall processing time is important
- We are typically concerned about throughput



Throughput (Inf/s)

- LIGO basic specs: 100k channels at 16000 Hz
  - Data Rate is 6.4 GB/s
  - Aiming for a 16000 inferences/s
  - Latency of one sedond



• LHC High Level Trigger :

- Data Rate is 1.2 TB/s
- Aiming for a 100000 events per second
- Latency of one second

# Standardization of Software

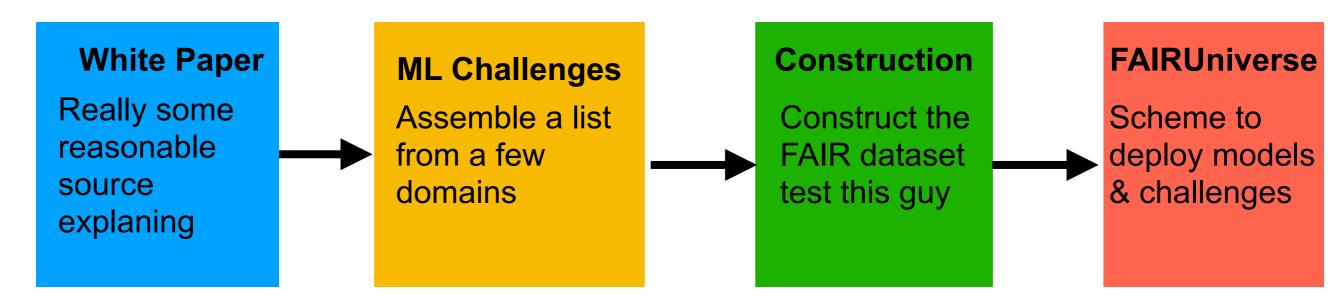
- There is an opportunity to standardize software infrastructure
  - A computing need that is common amongst domains
    - Astro: Event brokering
    - LIGO : Low latency alerts
    - Neutrino/LHC : Data Reconstruction
  - Some domains require different computing to others
- There is also a need for common ML problems and strategies
  - Deployment of effective deep learning problems
  - Anomaly detection
  - Data Augmentation/Generation for effective learning

## Getting the right models

- Last part of the document, we would like to highlight ML
  - Classifying the style of problems helps alignment
    - Supervised learning & its challenges
    - Data Generation/augmentation with ML
    - Anomaly detection
    - Semi-supervision
- For these range of models we would like to highlight a few
  - Ultimately, we would like to elevate to an ML challenge

## ML Challenges

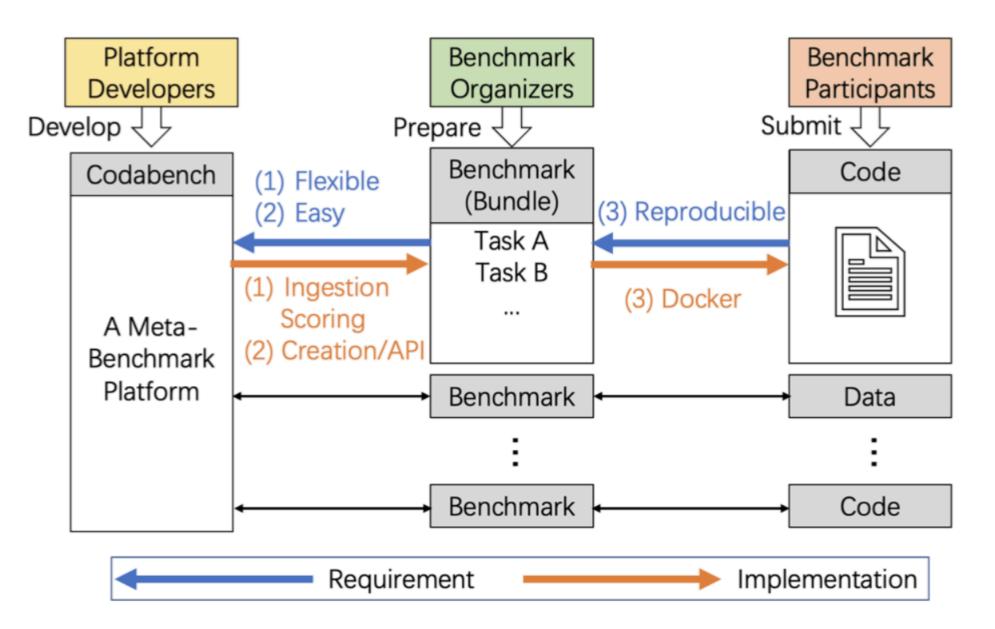
- Through the HDR community
  - We are working to organize a set of ML Challenges
  - Aiming to align this work with two other communities
  - MLCommons scientific (through ML tiny)
  - FAIRUniverse grant aimed at supporting



Annual Bootcamp at UW to award results & have a tutorial

#### FAIRUniverse has established Infrastructure

Codabench and "Fair Universe" Platform Based on https://www.codabench.org/

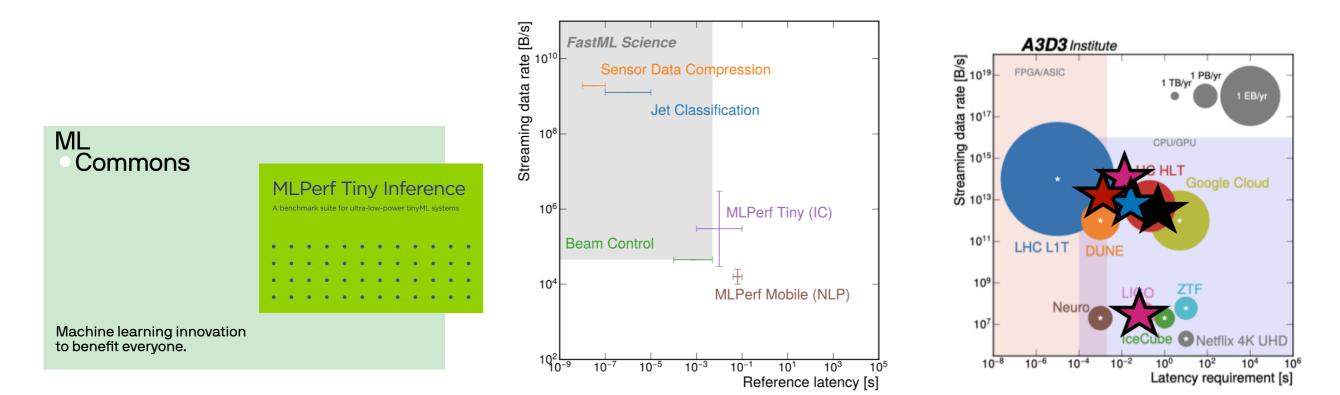


https://docs.google.com/presentation/d/

1hqnlvmMgPgVfm7GzDjb6vJfgafl3PRInd9SX1H0GoFA/edit?usp=sharing

#### Idea for ML Challenges

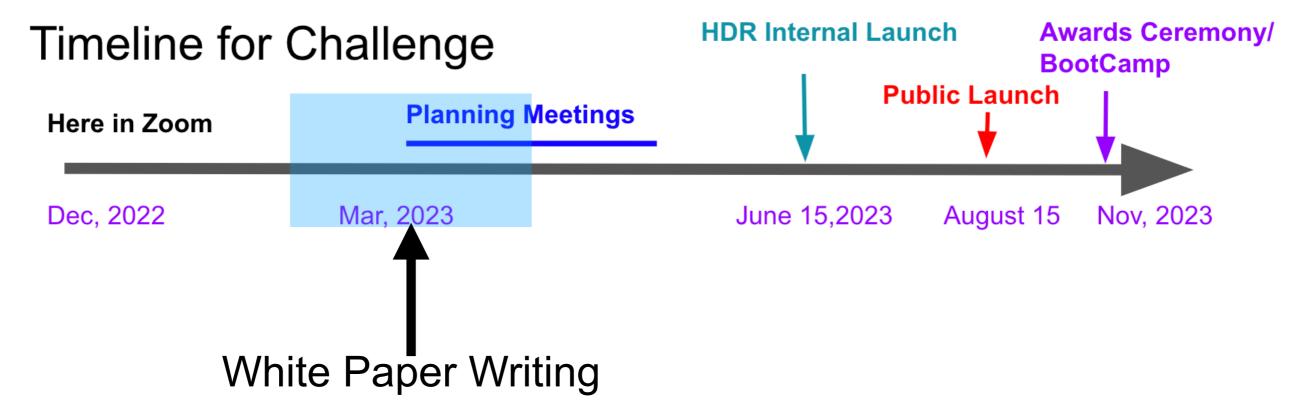
- There is one underway Icecube Kaggle Challenge
- <u>Dylan's talk</u> from FastML lists some HEP benchmark motivations
  - LHC tracking as a new benchmark
  - LIGO DeepClean as another benchmark
- More complicated challenges
  - Can we make a data generation challenge, or scheduling



## A Point to Highlight

- The best way for us to collaborate across domains
  - Making easy-to-use curated datasets or ML problems
  - We have the people in house to really test these datasets
- This is also a way to tie the different domains together
  - We can use this white paper to start testing out our challenges
    - Preparation of datsets
    - Release of models
- Can we get a dataset/model from each scientific domain
  - Also do we have the right benchmarks to do this?

#### Roadmaps



- This white paper can help us to assemble some models
  - We would like to assemble a few pages on these
    - Illustration of the problem (paragraph)
    - Illustration of the dataset preparation
    - Impact on the field

#### Conclusions

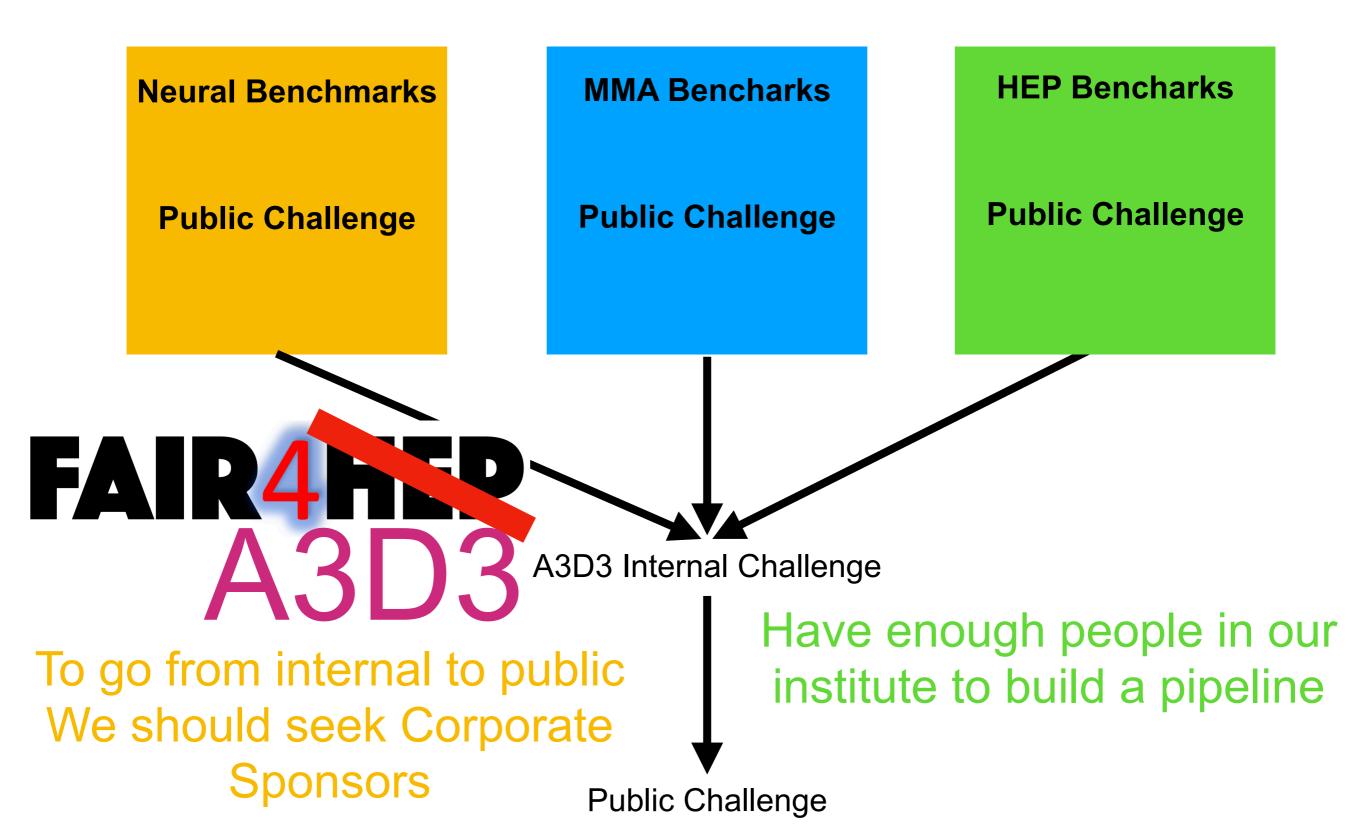
- Welcome! Enjoy your time here in Cambridge
  - We would like to write a white paper
  - We have some discussion time at the end of the conference
- Outline for the White paper (Lets keep it short!)
  - Discussion of computing tools and software
    - Path to aligning these across domains
  - List of critical models in the field
    - What makes these models
  - One plot to rule them all and bind these sections
- A roadmap for future computing can helps us move this forward

#### White Paper

https://www.overleaf.com/3629142192jgpsrnqvzccd

#### Backup

#### Possible Idea



# Standardization of Software

- Alert brokering and Image processing
- ML Problems
- Anomaly detection
- Data Augmentation Generation