

ACCELERATED, REPRODUCIBLE AND SCALABLE AI-DRIVEN GRAVITATIONAL WAVE DISCOVERY

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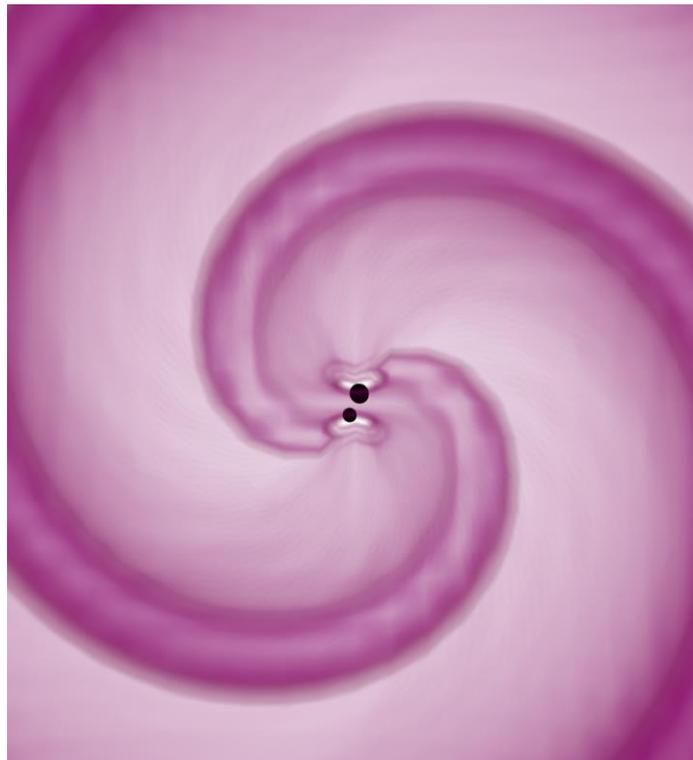


Parsl & funcX Fest
October 27-28, 2021

GRAVITATIONAL WAVE ASTRONOMY

WHAT

Gravitational wave observation of binary black hole mergers

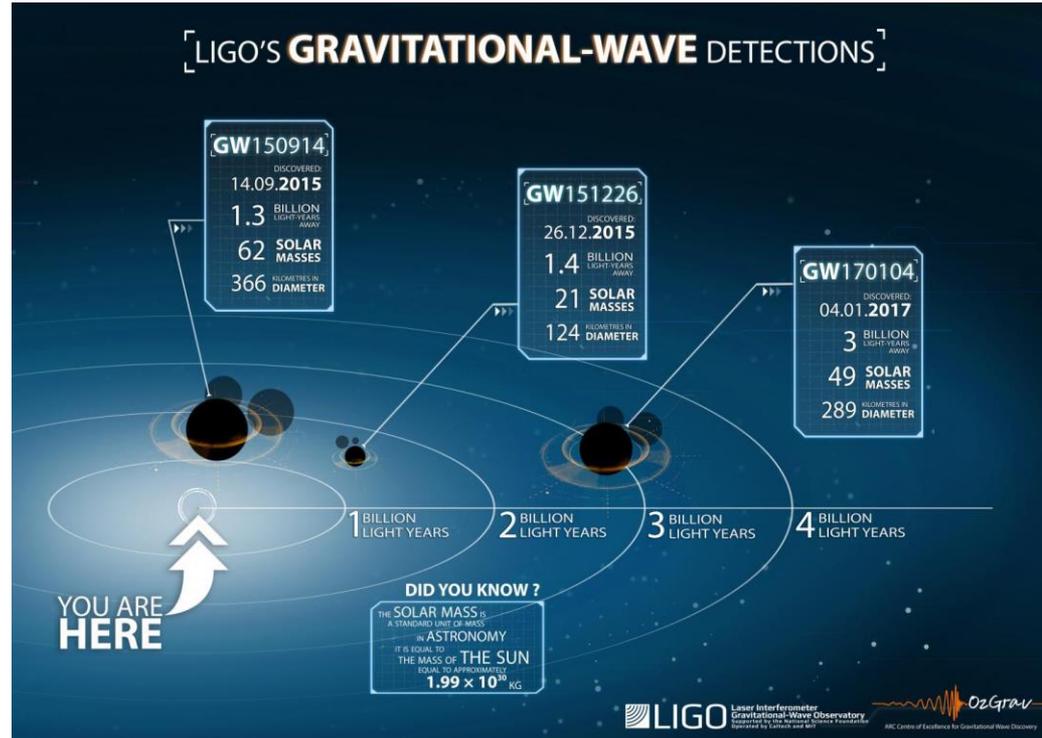


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GRAVITATIONAL WAVE ASTRONOMY

WHAT

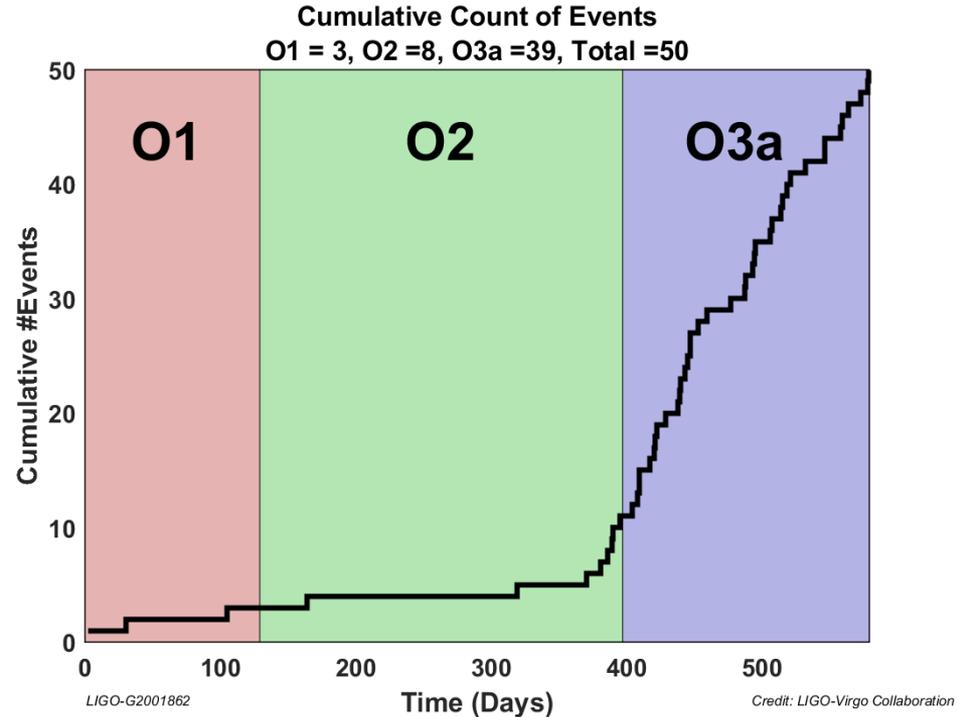
Ground-based detectors continue to improve their sensitivity to gravitational wave sources



GRAVITATIONAL WAVE ASTRONOMY

WHY

Advanced LIGO's enhanced sensitivity boosts detection rate of gravitational wave sources



GRAVITATIONAL WAVE ASTRONOMY

WHY

Number of detections continues to grow

Available computational resources
remain finite and oversubscribed

Radical re-thinking of computational
methods for gravitational wave discovery

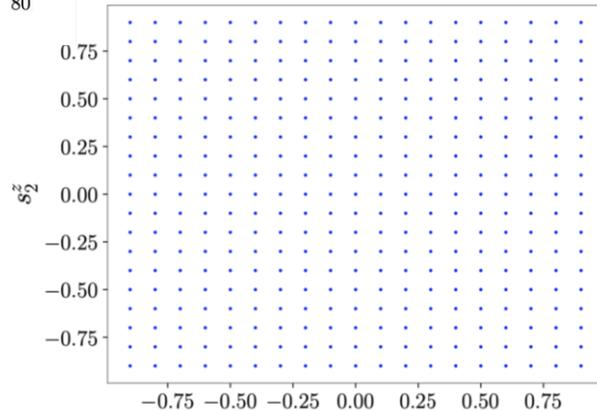
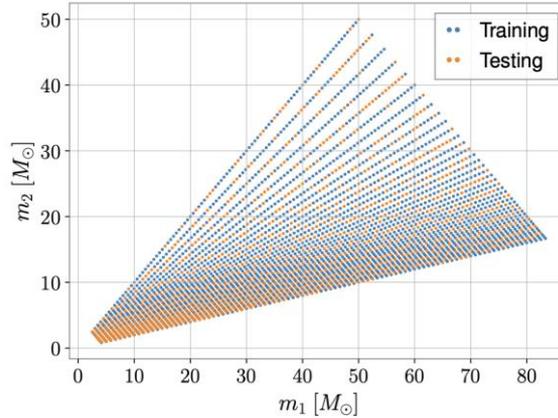


GRAVITATIONAL WAVE ASTRONOMY

WHAT

Demonstrate that AI + HPC provide a novel solution for production scale AI-driven gravitational wave detection

Consider 4-D signal manifold of real-time gravitational wave detection algorithms

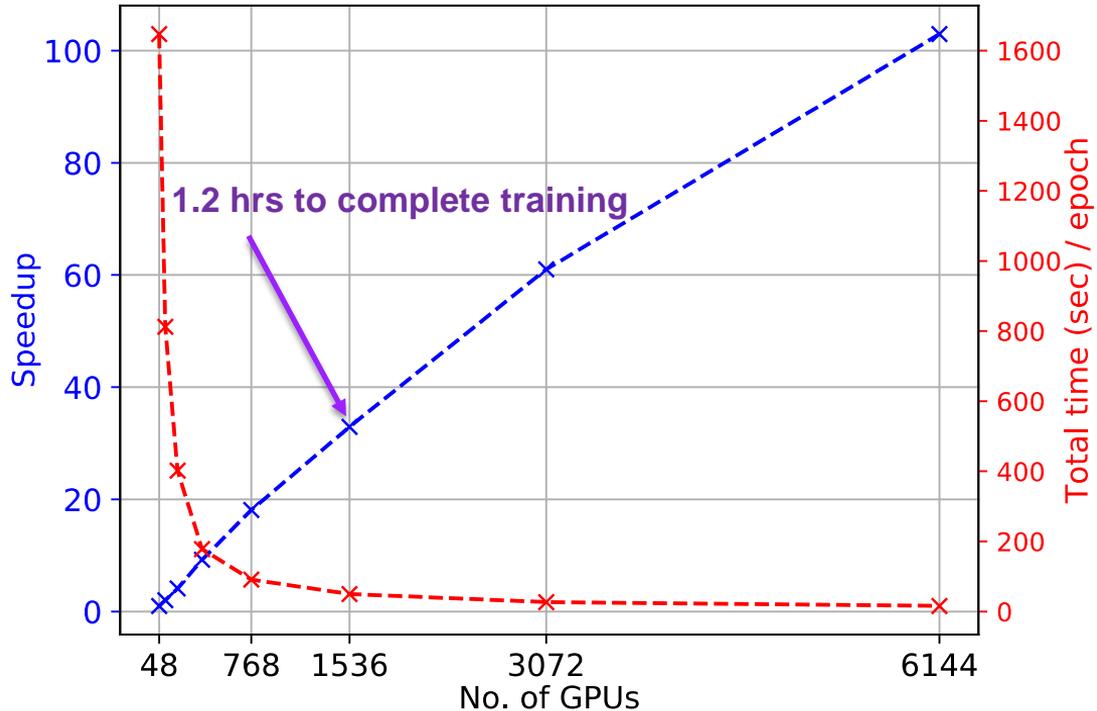


GRAVITATIONAL WAVE ASTRONOMY

WHAT

Densely sampling this 4-D signal manifold requires millions of modeled waveforms

Training stage: 1 month with a single NVIDIA V100 GPU



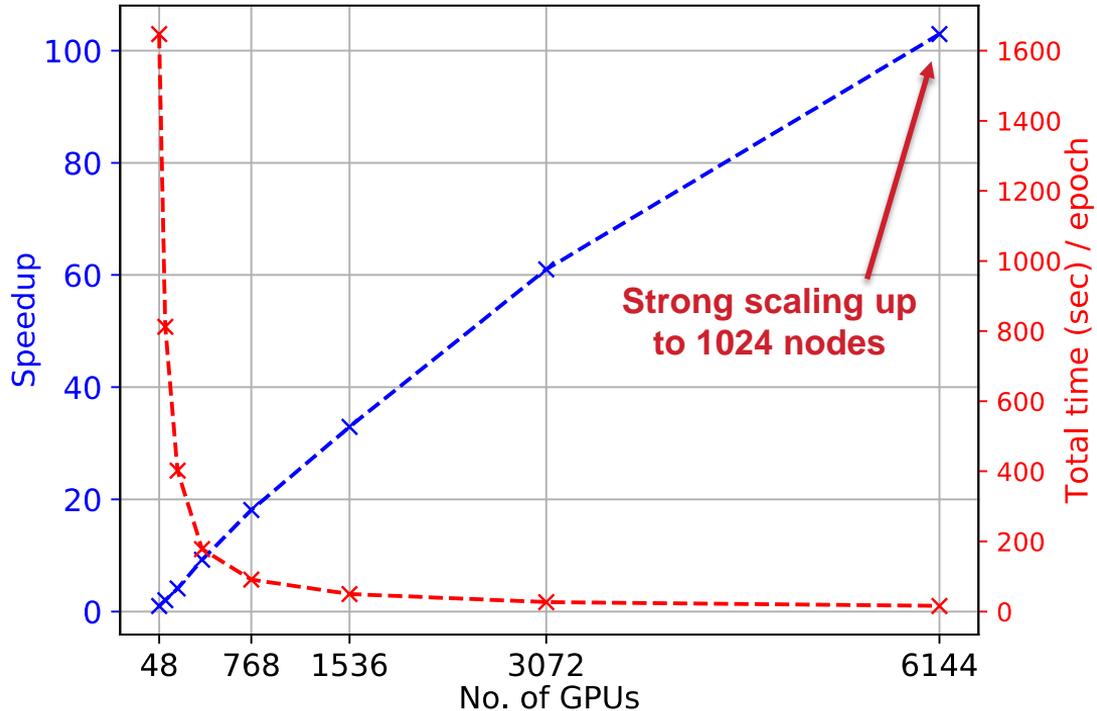
GRAVITATIONAL WAVE ASTRONOMY

HOW

Deployed and used new optimizers in Summit to reach optimal classification performance

600-fold speed up in training

Developed AI ensemble for real-time gravitational wave detection



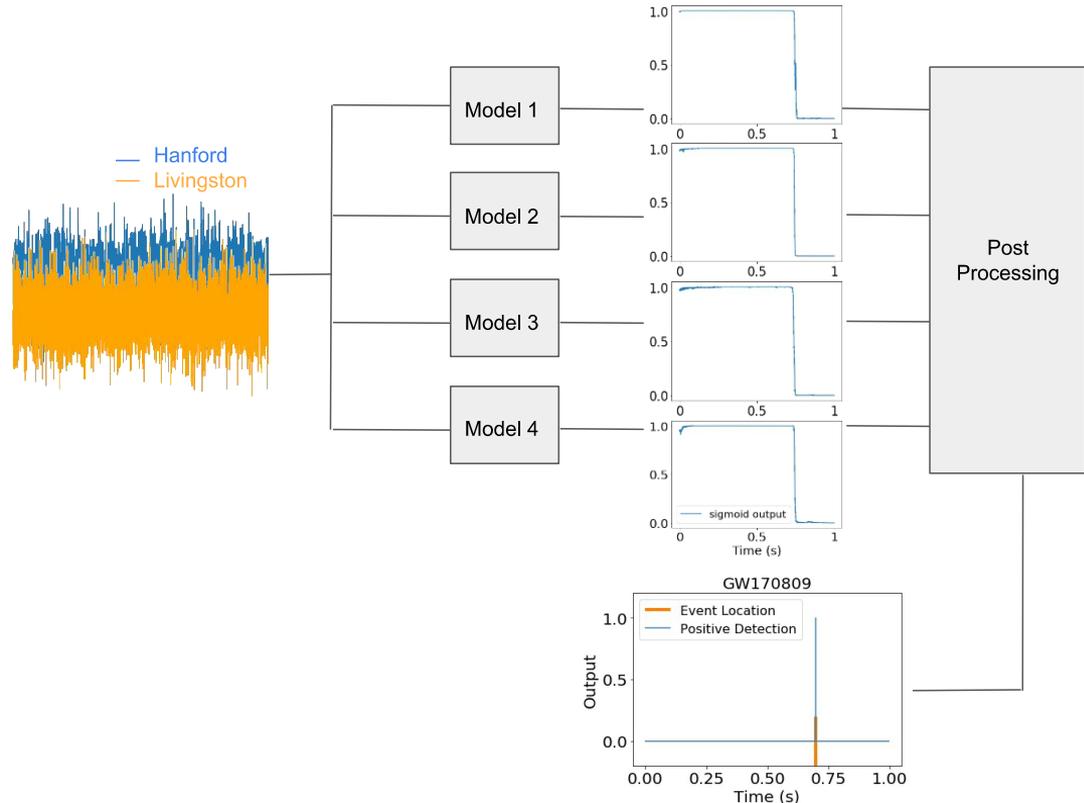
GRAVITATIONAL WAVE ASTRONOMY

HOW

4 AI models processing advanced LIGO data in tandem

Each model processes data faster than real time

Combine output of all models to increase confidence of detection



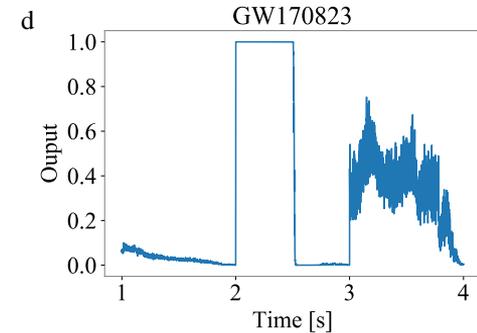
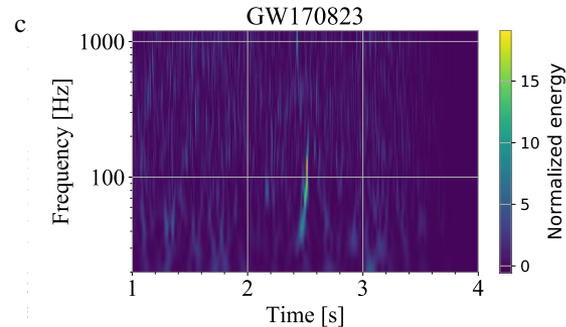
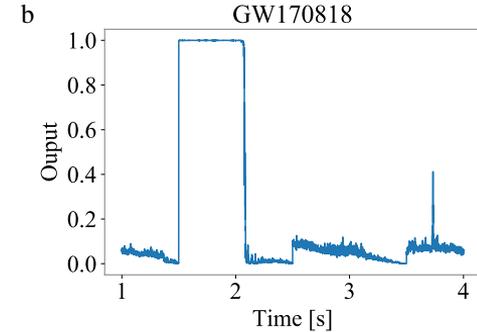
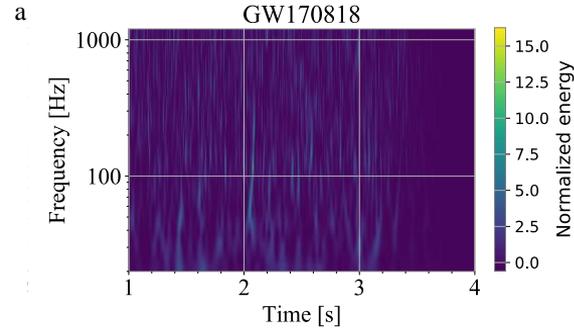
GRAVITATIONAL WAVE ASTRONOMY

HOW

4 AI models processing advanced LIGO data in tandem

Each model processes data faster than real time

Target: identify real events while reducing # of misclassifications



GRAVITATIONAL WAVE ASTRONOMY

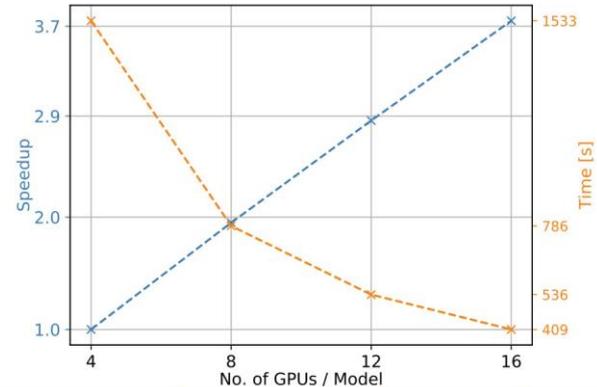
HOW

Use AI ensemble to process one month of advanced LIGO data

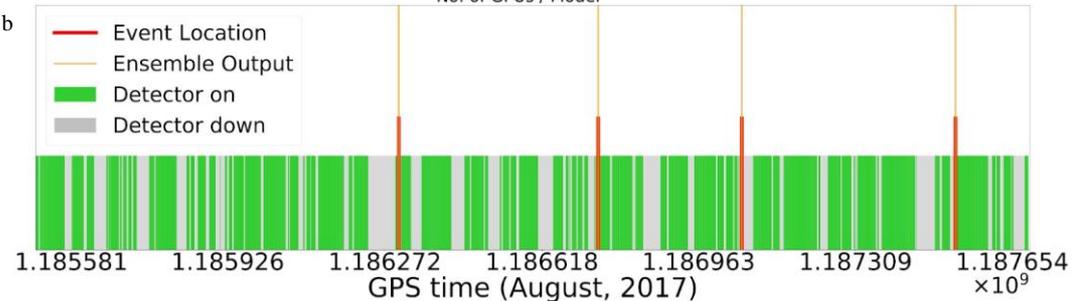
Quantify sensitivity, inference speed and scalability

Distribute inference on Hardware-Accelerated Learning (HAL) GPU cluster at NCSA
[64 NVIDIA V100 GPUs]

a



b



GRAVITATIONAL WAVE ASTRONOMY

REALLY?

Establish reproducibility, scalability and performance of results

Make AI ensemble and post-processing pipeline open source and containerized at the Data and Learning Hub for Science (DLHub)

DLHub

Data and Learning Hub for Science

A simple way to find, share, publish, and run machine learning models and discover training data for science

Documentation

[Read the Docs](#)[Examples](#)[Python SDK](#)[CLI](#)

DLHub Models

[Browse Models](#)

Papers and Presentations

[DLHub Paper](#)[DLHub Slides](#)

GRAVITATIONAL WAVE ASTRONOMY

REALLY?

Open source + containerized is
great

Can we do better than that?

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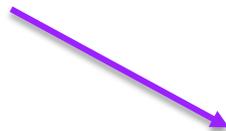
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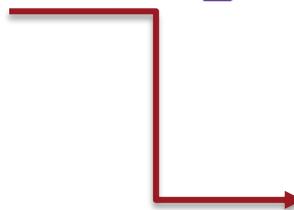


PRESENT – STATIC APPROACH

DLHub & funcX:
reproducible, scalable and
accelerated AI-discovery



Deploy AI
models in DLHub

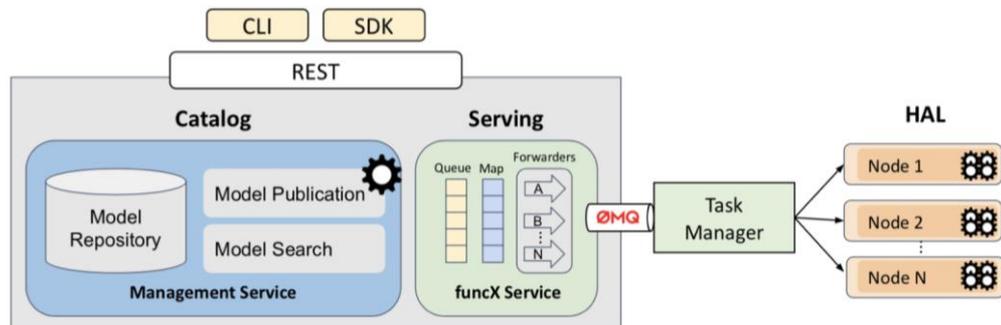


Reduce time-to-insight
with HPC platforms
Optimal distributed
training

Already used
at scale!

GRAVITATIONAL WAVE ASTRONOMY

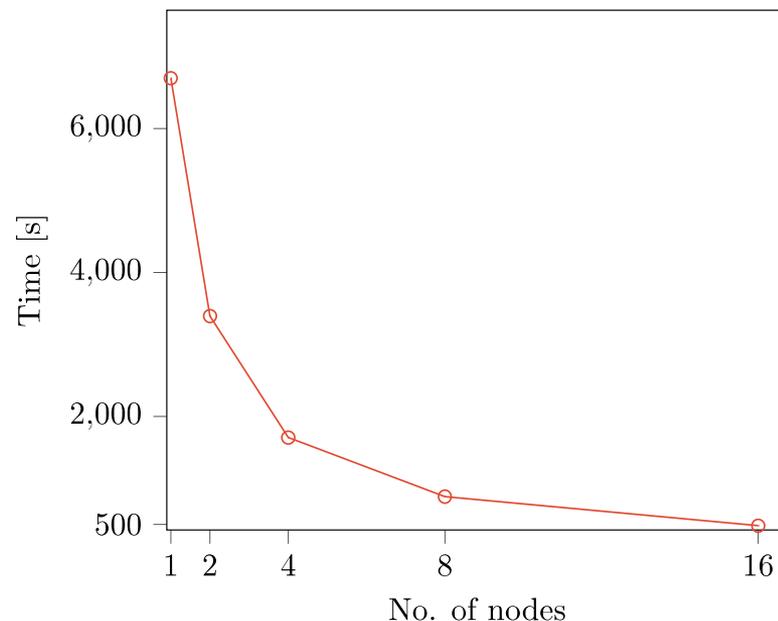
DLHub & funcX



Deploy funcX end-point at HAL

Call AI models hosted at DLHub

Optimal scalability, reproducibility established



PRESENT – STATIC APPROACH

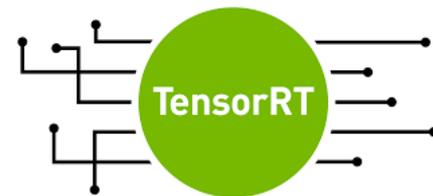


Reduce time-to-insight
with HPC platforms
Optimal distributed
training

Already used
at scale!



Deploy AI
models in DLHub



DLHub & funcX:
reproducible, scalable
and accelerated AI-
discovery at the edge

**TensorRT further reduced the
analysis to just 2 minutes!**

[nature](#) > [nature astronomy](#) > [articles](#) > [article](#)

Article | [Published: 05 July 2021](#)

Accelerated, scalable and reproducible AI-driven gravitational wave detection

[E. A. Huerta](#) , [Asad Khan](#), [Xiaobo Huang](#), [Minyang Tian](#), [Maksim Levental](#), [Ryan Chard](#), [Wei Wei](#), [Maeve Heflin](#), [Daniel S. Katz](#), [Volodymyr Kindratenko](#), [Dawei Mu](#), [Ben Blaiszik](#) & [Ian Foster](#)

[Nature Astronomy](#) **5**, 1062–1068 (2021) | [Cite this article](#)

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NVIDIA DEVELOPER

Aug 04, 2021

English 

AI Detects Gravitational Waves Faster than Real Time

By [Michelle Horton](#)

The team's models are **open-source** and readily available.



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Contributor Nature Astronomy

BEHIND THE PAPER

From Disruption to Sustained Innovation: Artificial Intelligence for Gravitational Wave Astrophysics

 **Eliu Huerta**
Lead for Translational AI, Argonne National Laboratory

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DYNAMIC AI

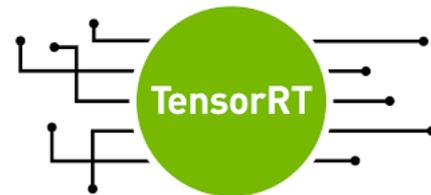
DLHub & funcX:
reproducible, scalable
and accelerated AI-
discovery at the edge

Summit
Theta-G AURORA ...

Reduce time-to-insight
with HPC platforms



Deploy dynamic AI
models in DLHub



Edge Distributed
Computing
TensorRT ...

Major upgrade of
AI models

Active/Transfer/Reinforcement



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This research used resources of the Argonne Leadership Computing Facility, which is a DOE Office of Science User Facility supported under Contract DE-AC02-06CH11357

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