

Emerging Needs & Opportunities in Data-Intensive Domains: Astronomy

Joshua Bloom (@profjsb)

Gordon & Betty Moore Foundation Data-Driven Investigator
UC Berkeley, Astronomy



Presented to the National Academy of Sciences Roundtable on the
"Intersection of Domain Expertise and Data Science"

Democratizing Trends in Physical Sciences

Data

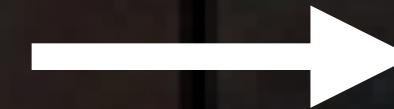
Decreasing cost to obtain,
move, store



open data,
more freely shared

Compute

Decreasing cost,
increasing specialty



more accessible

Technology/ Methodology

Algorithmic innovation,
software tooling



open source

Democratizing Trends in Physical Sciences



Competition for superior inferential capabilities

- ▶ Statistical / machine learning capabilities
- ▶ Computational prowess
- ▶ Ability to innovate methodologies

Outline

Notable data-driven success with domain + methodologies

particle physics & gravitational wave astronomy

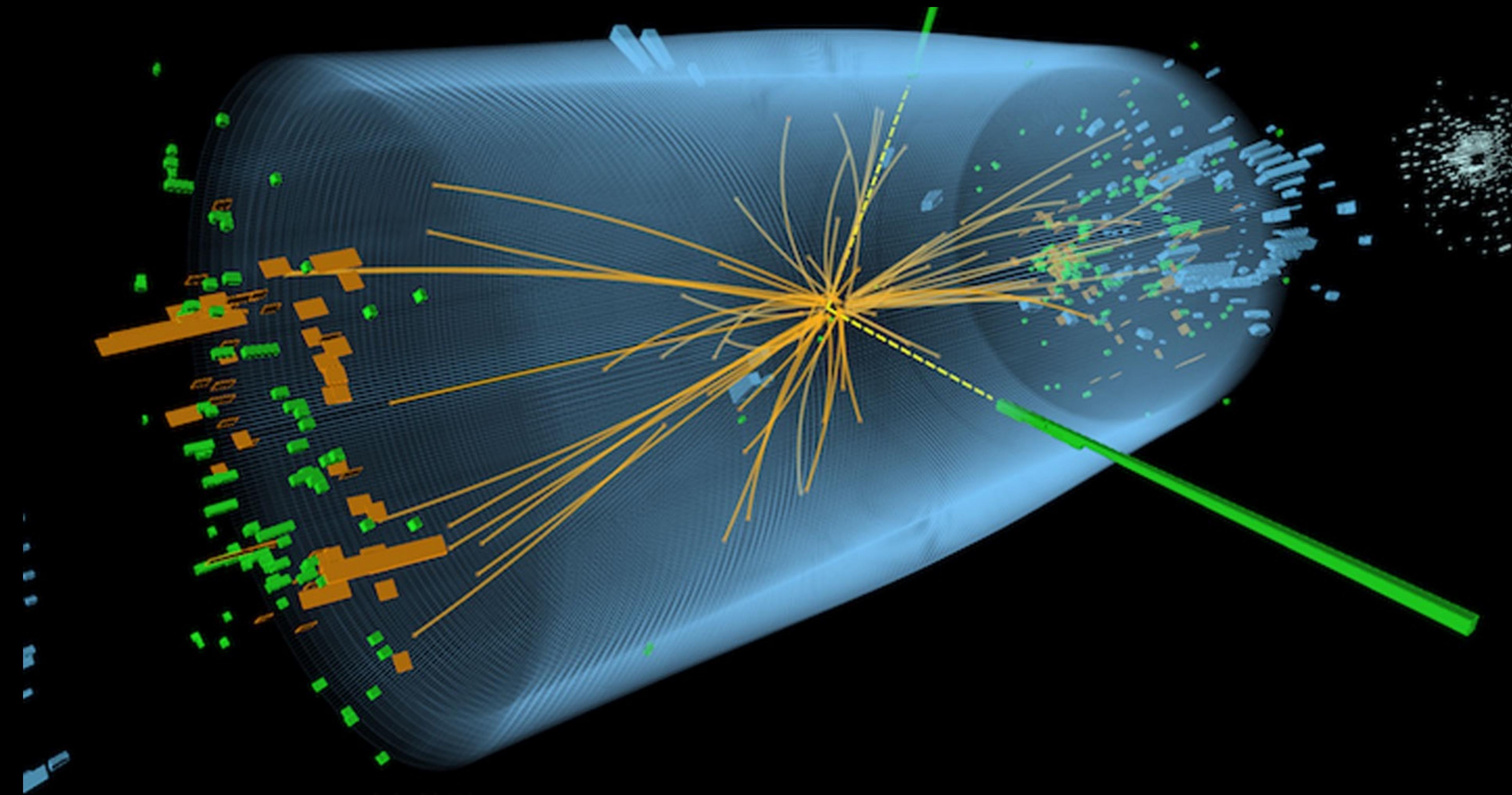
Astronomy's Data deluge

- who wins?
- examples of data science impact in astro

Educational challenges given these trends

Symbiotic relationship between data domains
& methodological sciences

Discovery of the Higgs Boson in 2012



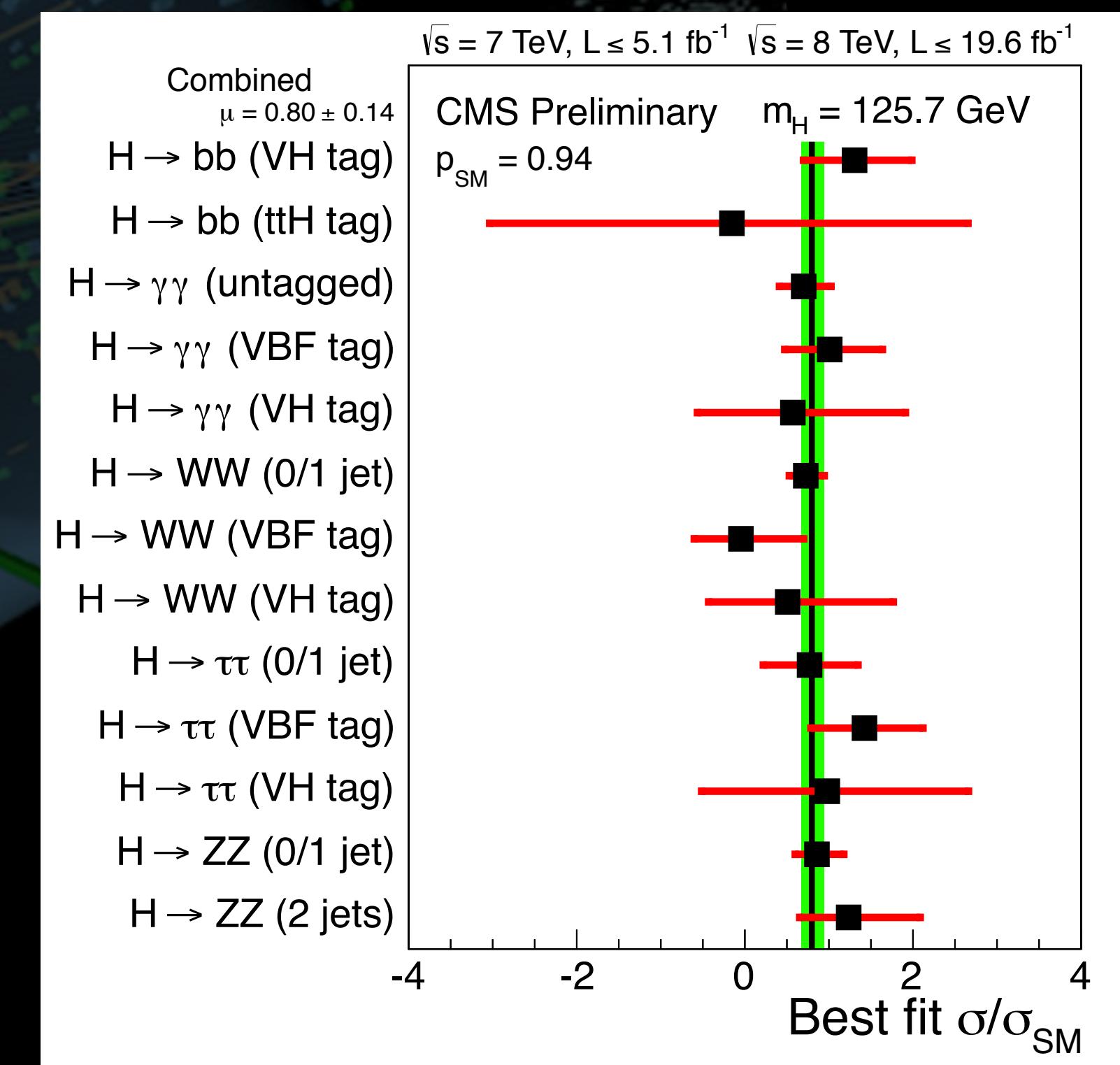
Data & Computation

- ▶ 600M collisions/s; only $\sim 100/s$ are interesting
- ▶ 600 GB/s raw \rightarrow 25 GB/s processed
- ▶ 3.5 MW data centers, Pb/day processing

Inference

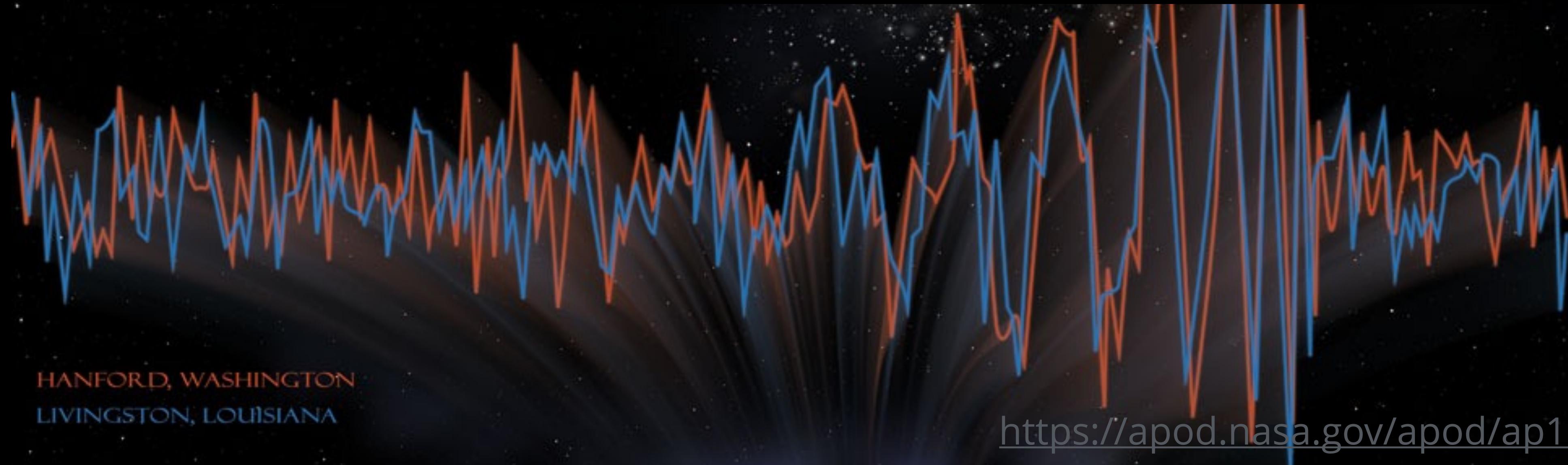
- ▶ "HEP phystatisticians"
- ▶ Statistics combining standard model (theoretical underpinning) & observations

Horvath 1310.6839





Direct Detection of Gravitational Waves in 2016



HANFORD, WASHINGTON
LIVINGSTON, LOUISIANA

<https://apod.nasa.gov/apod/ap160211.html>

Data & Computation

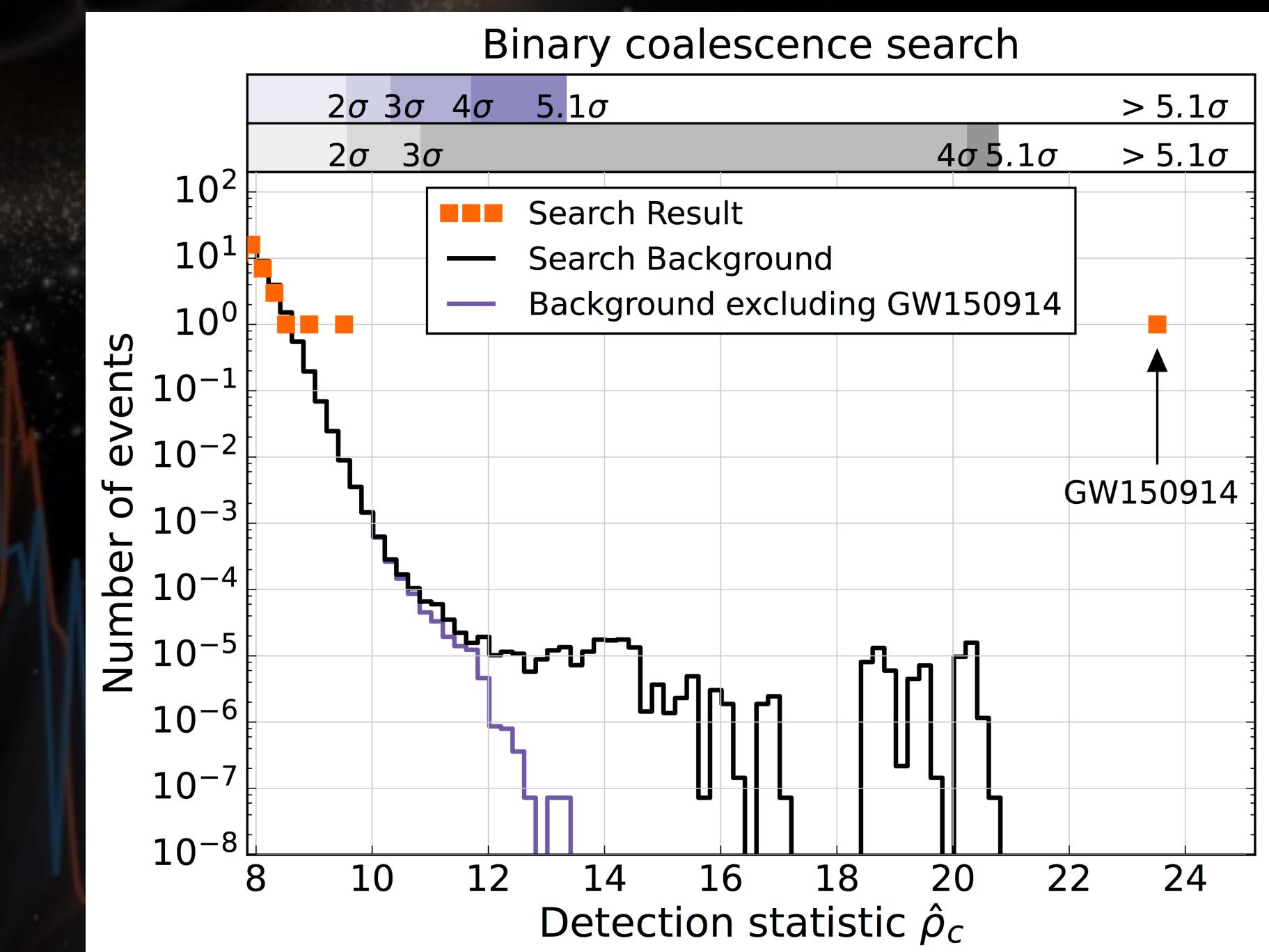
- ▶ aLIGO Data rate: 81 MB/s, 2PB/yr
- ▶ 10^5 templates to continually match (32TB) at 5 TFLOPS
- ▶ 20k CPUs, 12 PB cluster

Shoemaker M060056-01.doc

Inference

- ▶ novel long-tail statistics
- ▶ Need rapid identification for multi wavelength followup

e.g. Cannon et al. 2012



Lessons from Large-Scale Physics Experiments

- Strongly focused on limited number of high-impact discoveries
- Residual inventions (e.g., WWW, grid computing)
- **Diverse Teams:** physicists collaborating with methodological specialists
- Large collaborations producing science results **themselves**

Astronomy's Data Deluge

Large Synoptic Survey Telescope (LSST) - 2020

Light curves for 800M sources every 3 days

10^6 supernovae/yr, 10^5 eclipsing binaries

3.2 gigapixel camera, 20 TB/night

LOFAR & SKA

150 Gps (27 Tflops) → 20 Pps (~100 Pflops)

Many other astronomical surveys are already producing data:

SDSS, iPTF, CRTS, Pan-STARRS, Hipparcos, OGLE, ASAS, Kepler, LINEAR, DES etc.,

- US Open Skies Policy
- Investments in Public Data Dissemination
- Unaffiliated Scientists are expected to reap rewards

If anyone can get the Data, who wins?

domain expertise → She who asks the right questions

+

data science

She who answer questions better & faster than others:

- ▶ computational access
- ▶ methodological inference (e.g., machine learning)
- ▶ better story telling, dissemination of results
- ▶ reproducibility (acceptance)

Some of my work...

- ▶ Built & Deployed robust, real-time supervised machine learning framework, discovering >10,000 events in > 10 TB of imaging → 75+ journal articles
- ▶ Built probabilistic source classification catalogs on public archives with novel active learning approach

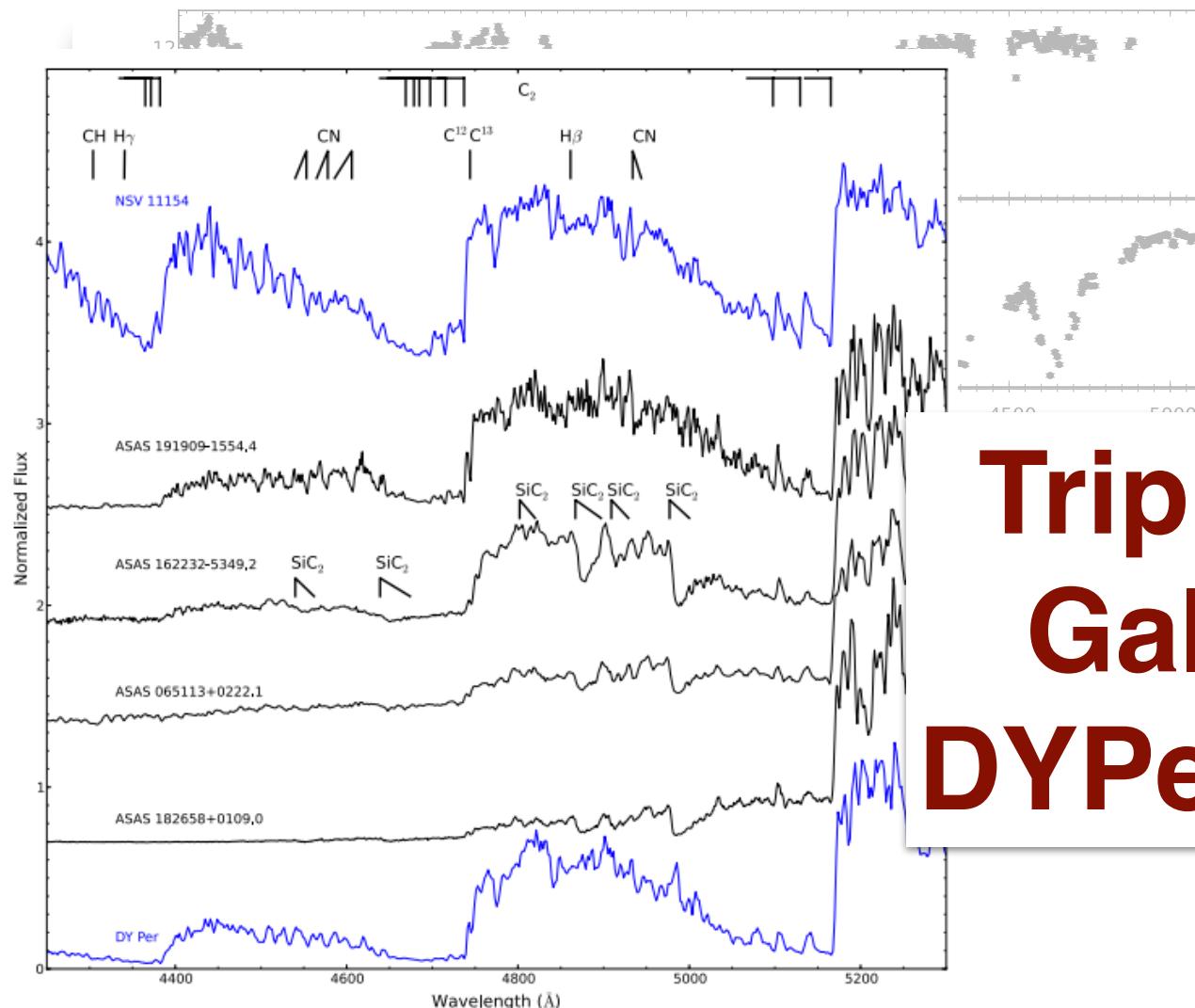
Our ML framework found the Nearest Supernova in 3 Decades ..



Probabilistic Classification of 50k+ Variable Stars

15 "RCB/DYP" candidates

8 new discoveries



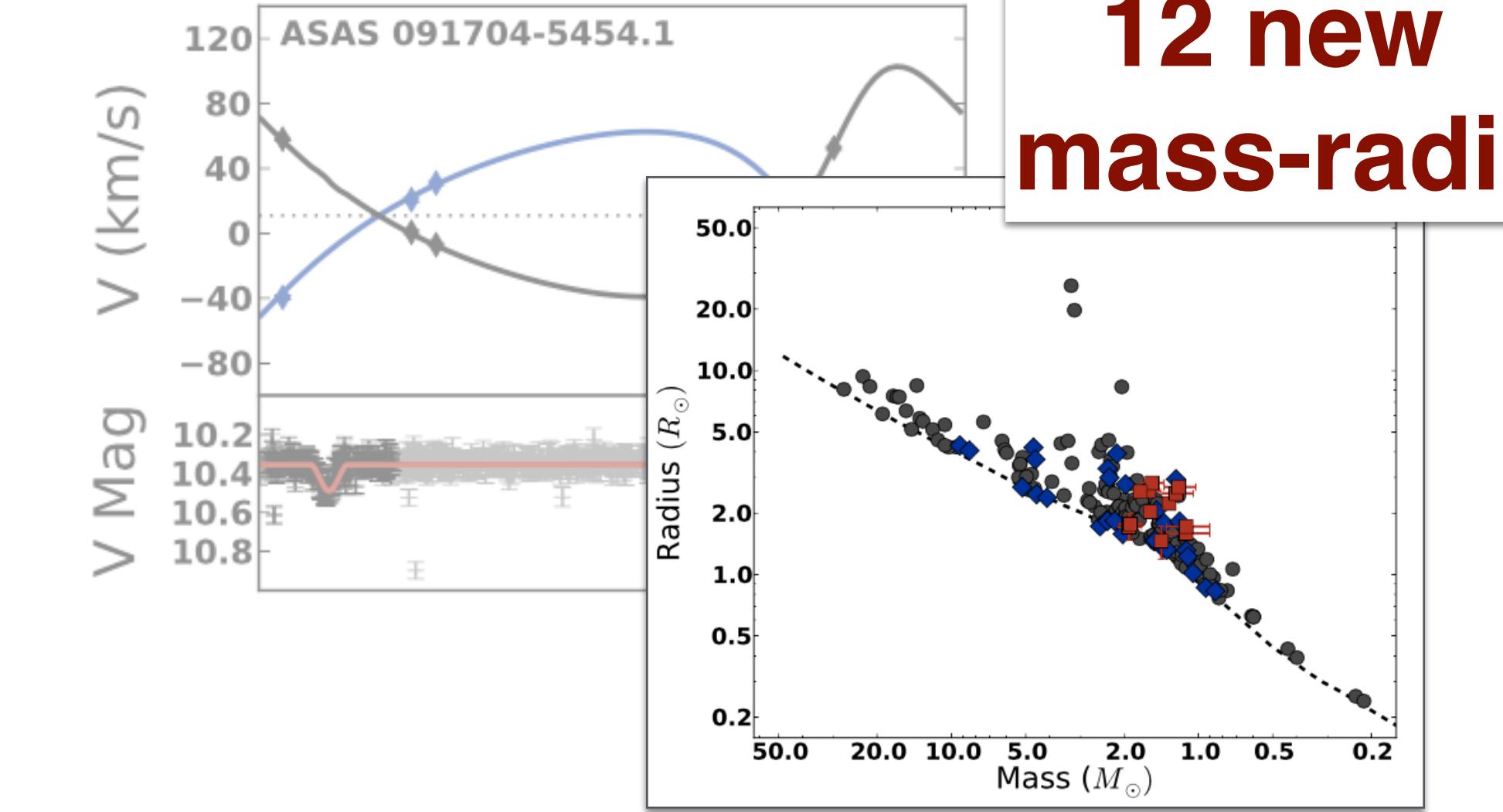
Triple # of Galactic DYPer Stars

DISCOVERY OF BRIGHT GALACTIC R CORONAE BOREALIS AND DY PERSEI VARIABLES: RARE GEMS MINED FROM ACVS

Miller, Richards, JSB,..ApJ 2012

106 "DEB" candidates

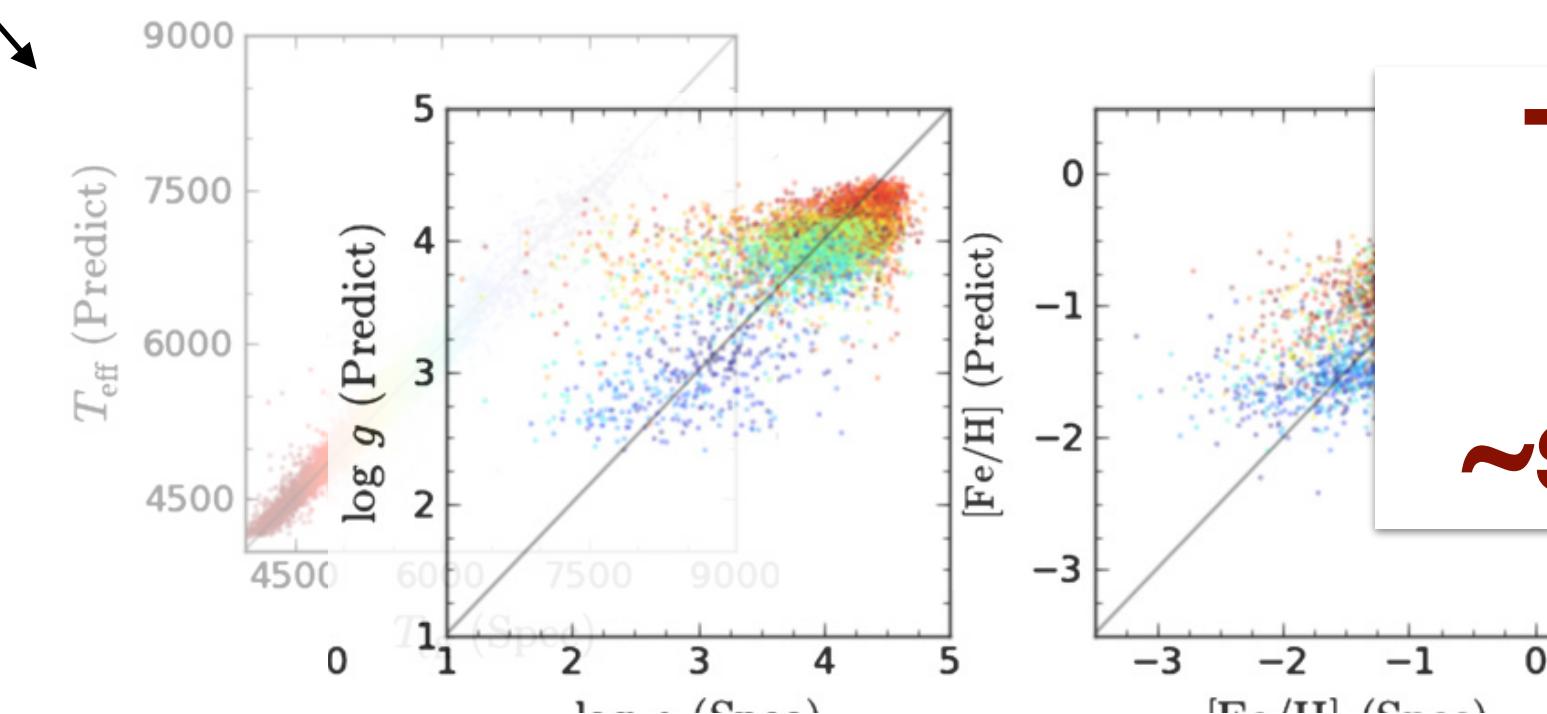
5400 Spectroscopic Targets



12 new mass-radii

The Highly-Eccentric Detached Eclipsing Binaries in ACVS and MACC

Shivvers,JSB,Richards MNRAS,2014



Turn synoptic imagers into ~spectrographs

A MACHINE-LEARNING METHOD TO INFER FUNDAMENTAL STELLAR PARAMETERS FROM PHOTOMETRIC LIGHT CURVES

Miller, JSB, Richards,..ApJ 2015

Some unsupervised work...

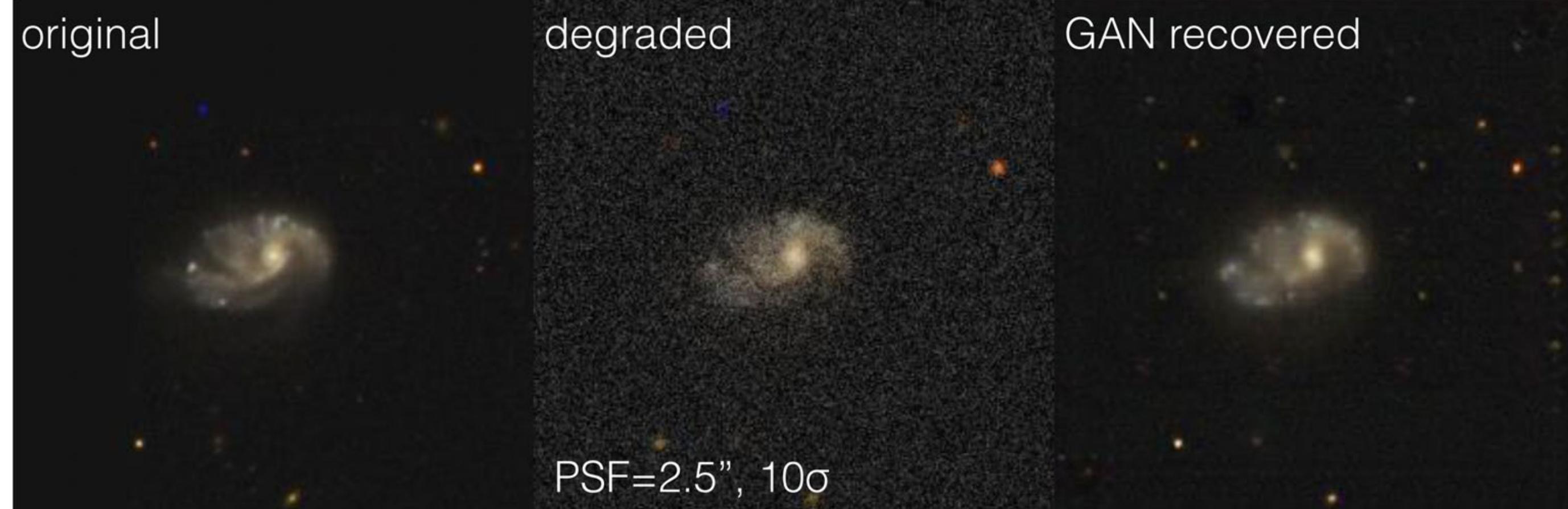


Generative adversarial networks recover features in astrophysical images of galaxies beyond the deconvolution limit

Kevin Schawinski,¹ Ce Zhang,² Hantian Zhang,² Lucas Fowler¹
and Gokula Krishnan Santhanam²

¹*Institute for Astronomy, Department of Physics, ETH Zurich, Wolfgang-Pauli-Strasse 27, CH-8093 Zürich, Switzerland*

²*Systems Group, Department of Computer Science, ETH Zurich, Universitätstrasse 6, CH-8006 Zürich, Switzerland*



Machine Learning Is Bringing the Cosmos Into Focus

The Atlantic

Training neural networks to identify galaxies could forever change humanity's perspective of the universe.

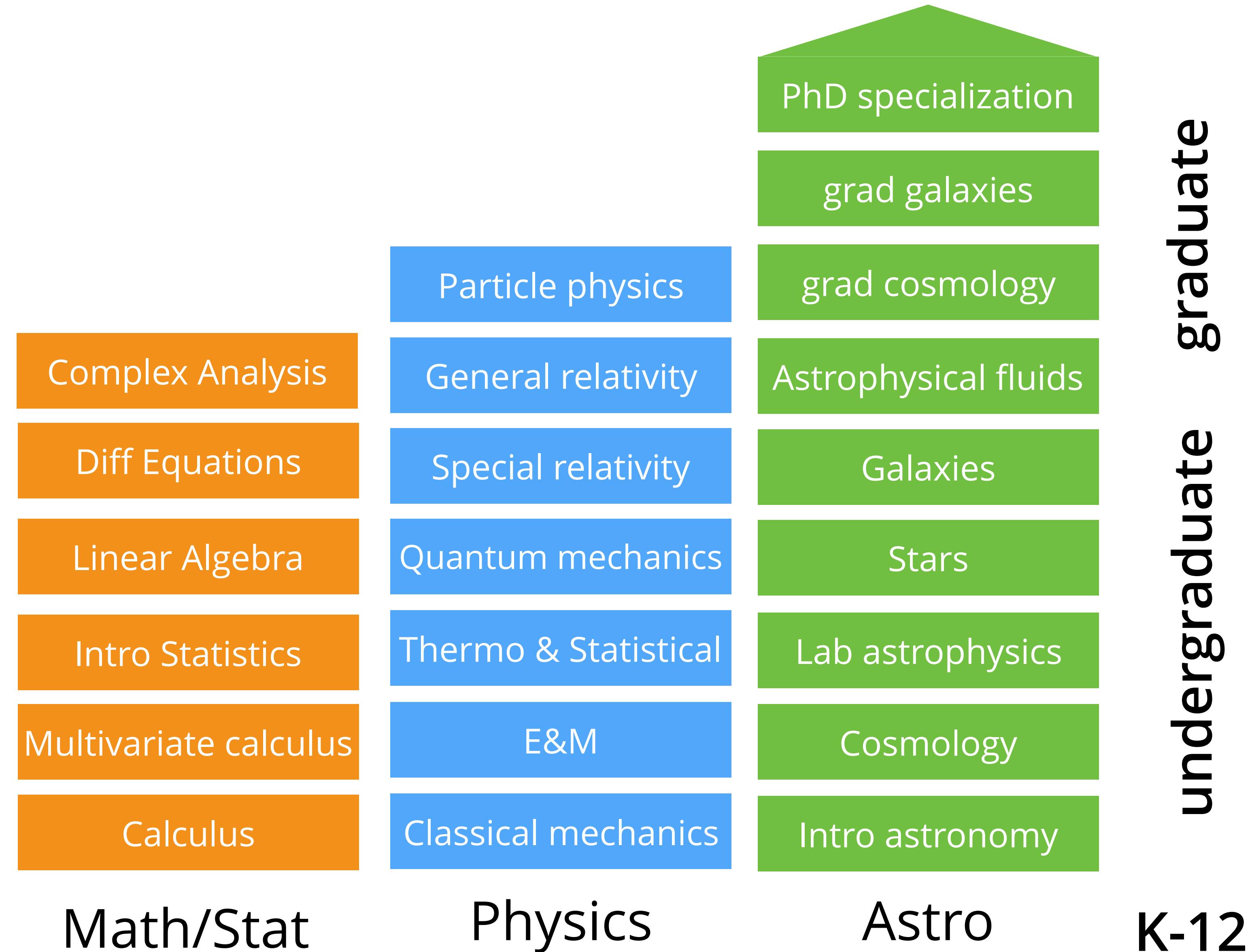
NATURE | NEWS

Astronomers explore uses for AI-generated images

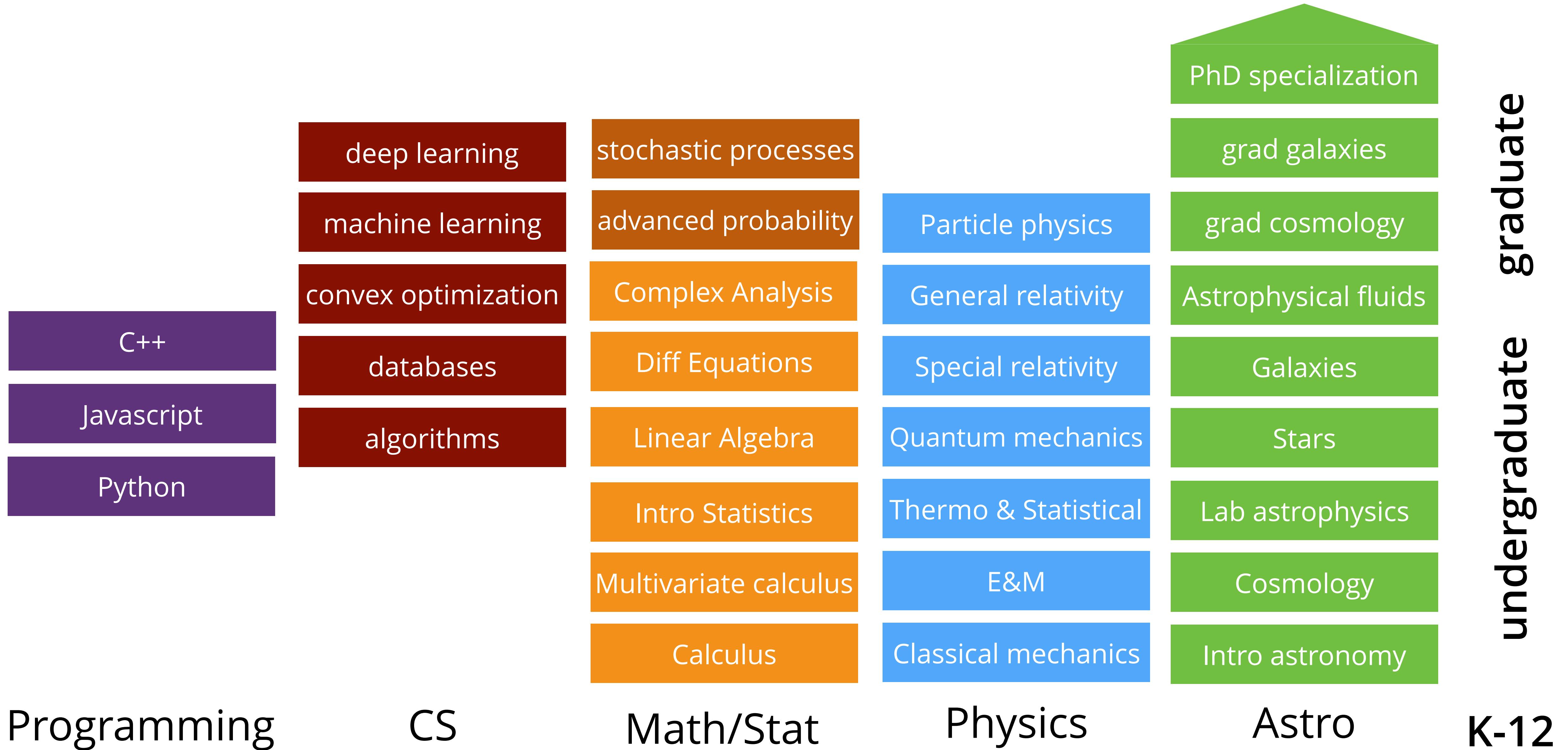
Neural networks produce pictures to train image-recognition programs and scientific software.

Davide Castelvecchi

Challenge of Data-Driven Domain Education Stack



Challenge of Data-Driven Domain Education Stack



Python Computing for Data Science

Graduate course at UC Berkeley

parallelism

interfacing to other languages

Bayesian inference & MCMC

visualization

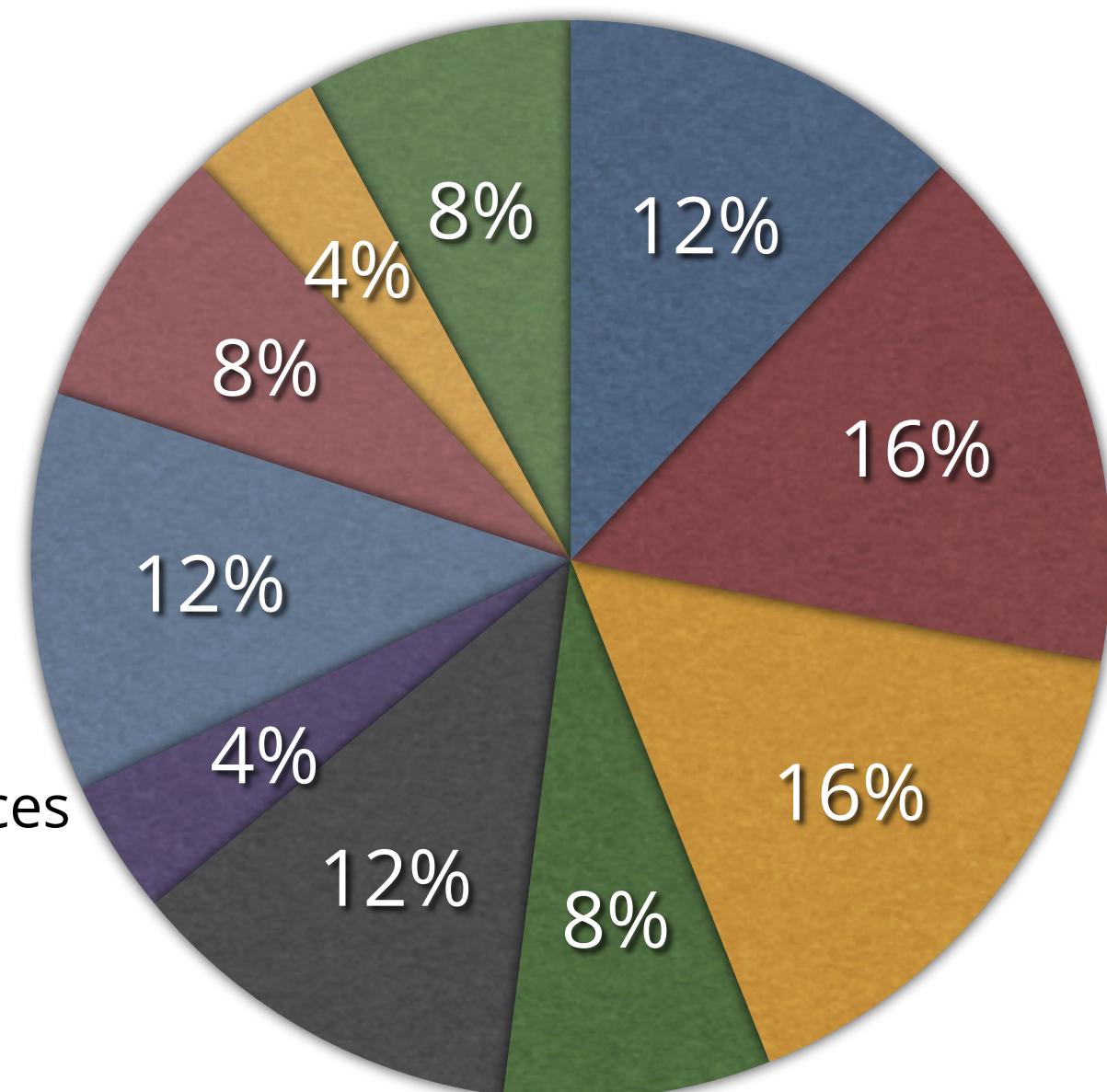
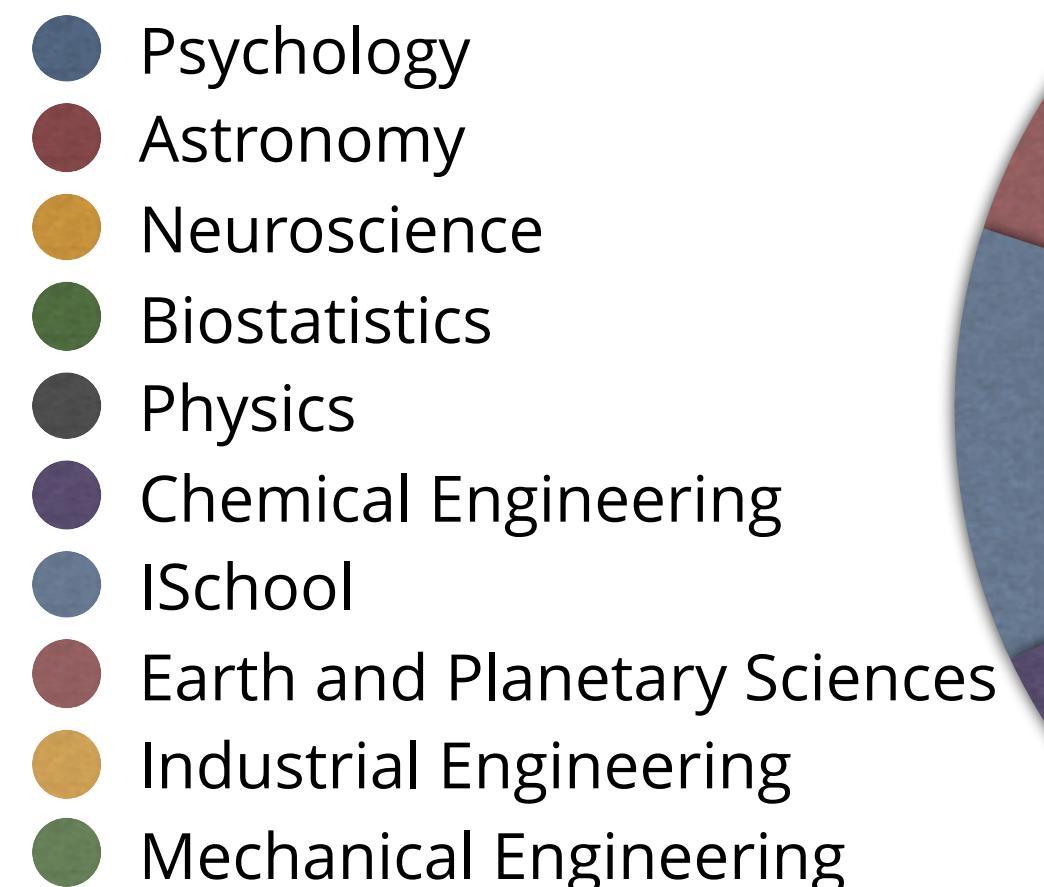
hardware control

database interaction

user interface & web frameworks

timeseries & numerical computing

machine learning

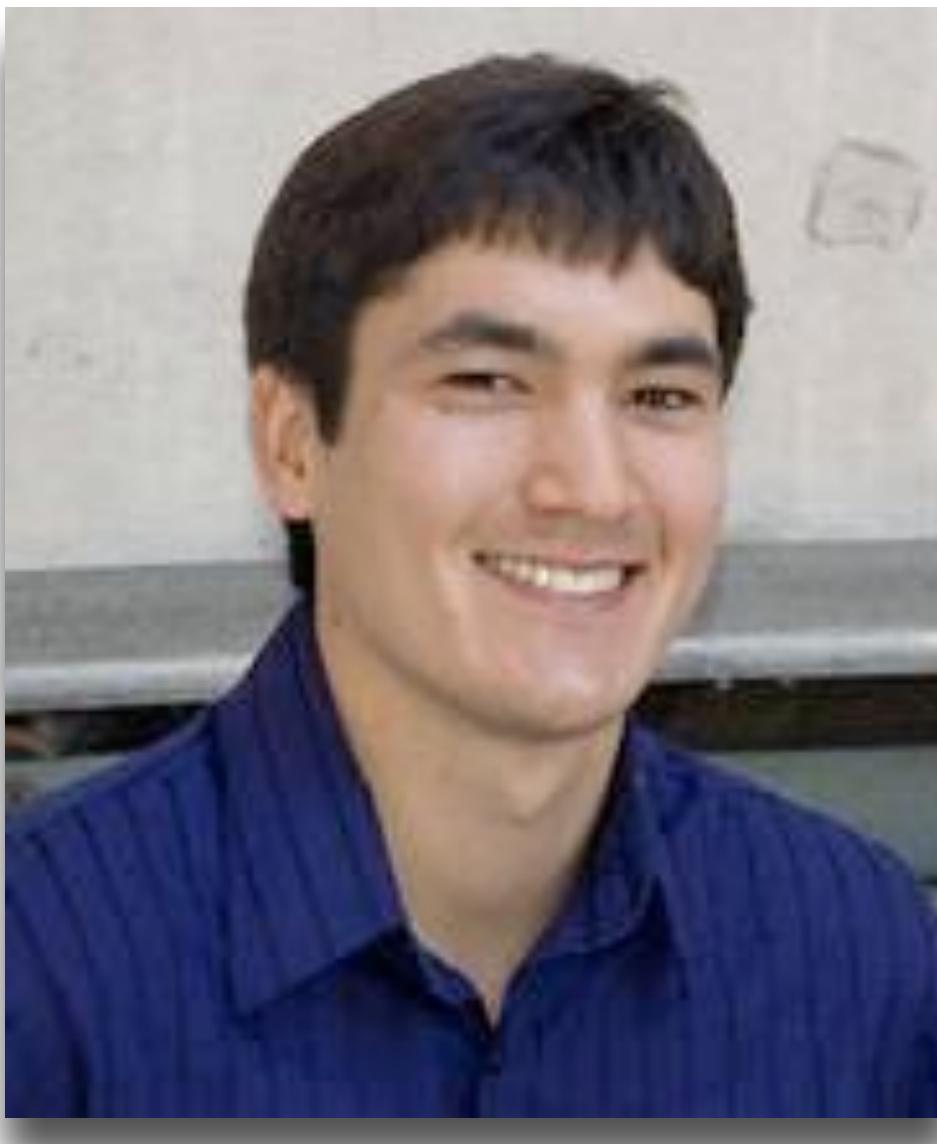


Prevalence of Earth-size planets orbiting Sun-like stars

Erik A. Petigura^{a,b,1}, Andrew W. Howard^b, and Geoffrey W. Marcy^a

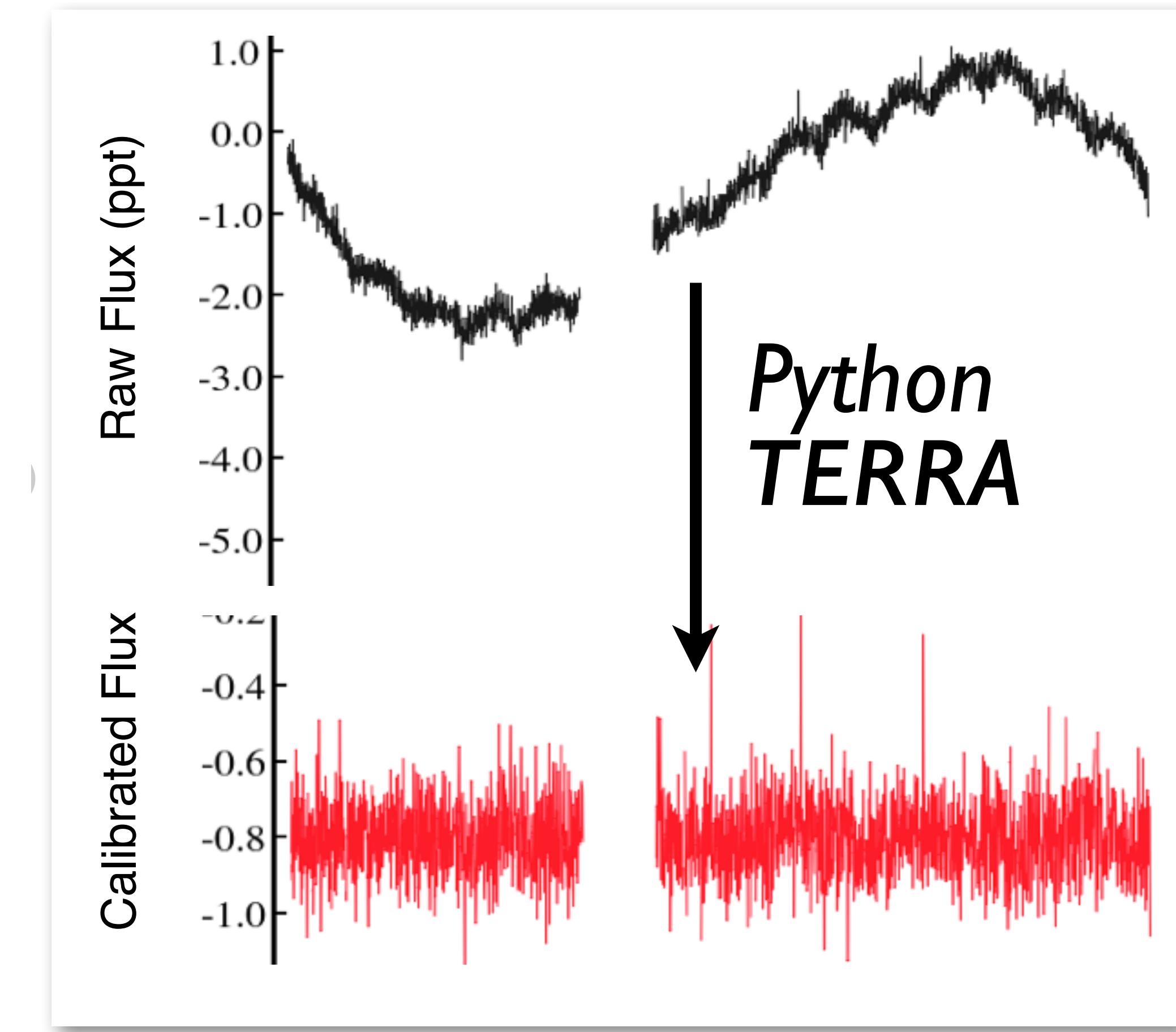
PNAS [2014]

^aAstronomy Department, University of California, Berkeley, CA 94720; and ^bInstitute for Astronomy, University of Hawaii at Manoa, Honolulu, HI 96822



Erik Petigura
Berkeley Astro
Grad Student

Bootcamp/Seminar Alum



DOE/NERSC computation

21st Century Education Mission: Produce Gamma-shaped people

Π

vs.

Γ

deep domain skill/knowledge/training
deep methodological knowledge/skill

deep domain or methodological skill/knowledge/training
strong methodological or domain knowledge/skill

Goal: empower teams of gamma's to excel

A Rich History of Symbiosis

Astronomers co-opt tools...



"Fleming had constructed a spyglass, by means of which visible objects, though very distant from the eye of the observer, were distinctly seen as if nearby."

- Galileo Galilei (1610)

...while astro serves as a testbed for computational exploration



"I love working with astronomers, since their data is worthless."

- Jim Gray, Microsoft

“novelty² problem”

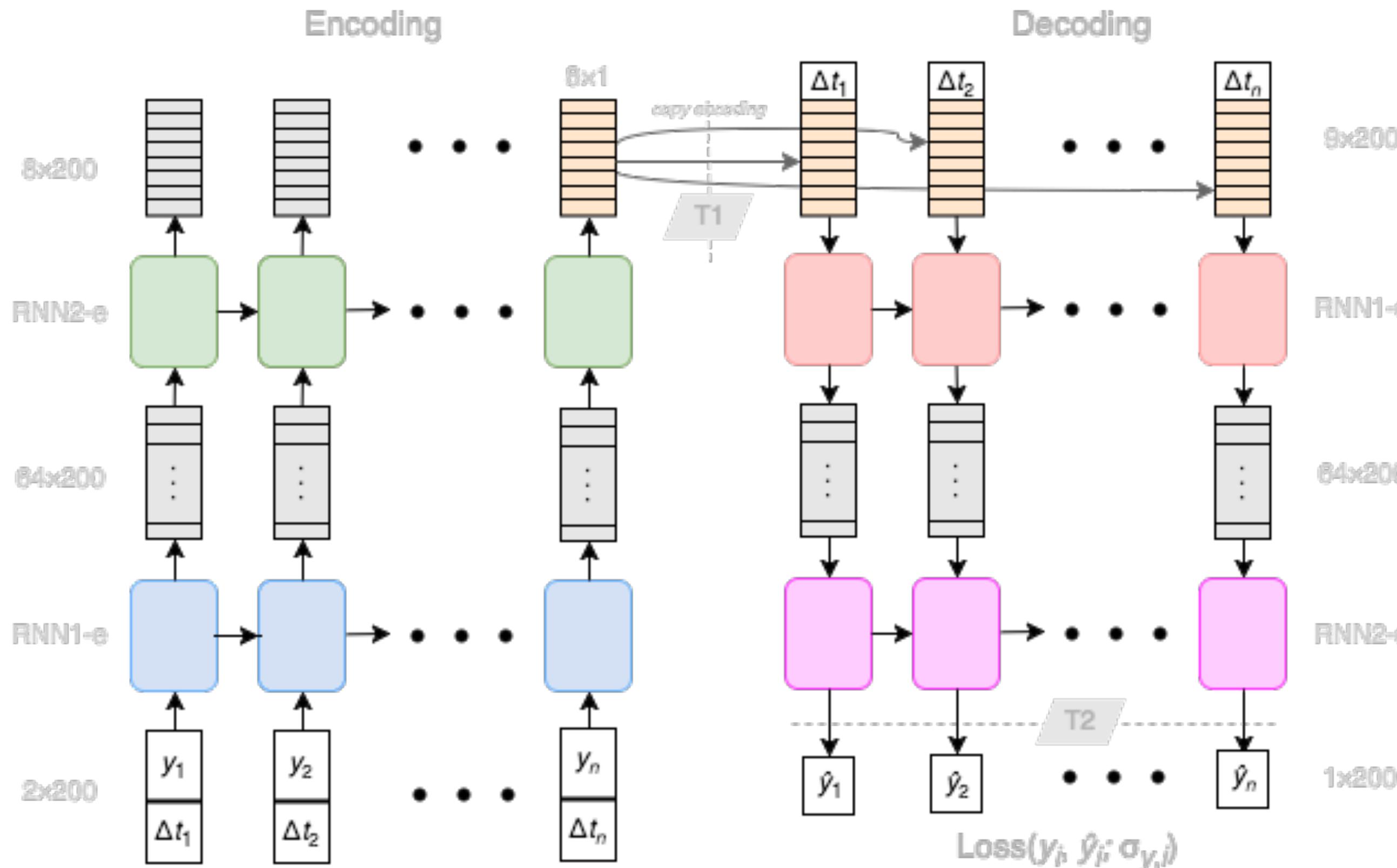
Established CS/Stats/Math *in Service*
of novelty in domain science

VS.

Novelty in domain science driving & informing novelty
in CS/Stats/Math

<https://medium.com/tech-talk/dd88857f662>

Deep Learning for (Astronomical) Timeseries Inference



"Automated feature extraction for irregularly-sampled time series using deep neural networks"

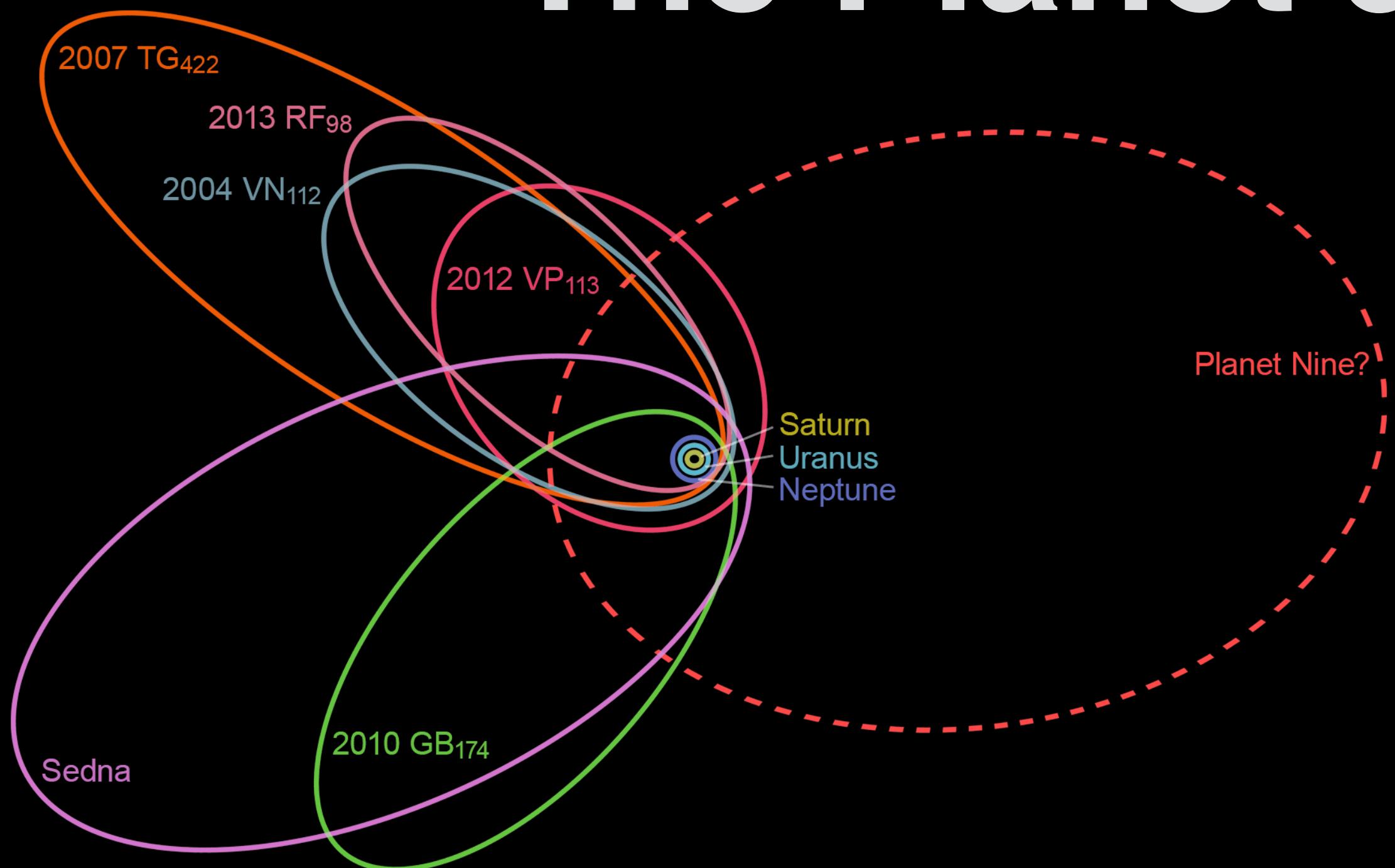
Naul, JSB+, in prep



Network for RNN+GRU

Stats PhD from Stanford

The Planet 9 Opportunity



hypothesized to **exist** using mostly public data about comet orbits + deep domain knowledge + sophisticated computing

Batygin & Brown 1601.05438

My Prediction: the discovery data of Planet X have already been obtained & reside in existing public data archives. It will be a group of clever astronomers & statisticians with a lot of compute resources that will make the retrospective discovery of the century...

Summary

- Democratizing trends in the physical sciences
- Winners will be those domain experts with superior inferential capabilities, not themselves, but in teams
- Data science education for domains should be focused on learning enough to work with those methodological domains
- Symbiosis: Physical domain science data/questions can also foster novel methodological approaches

Emerging Needs & Opportunities in Data-Intensive Domains: Astronomy

Thanks!

Joshua Bloom
@profjsb