

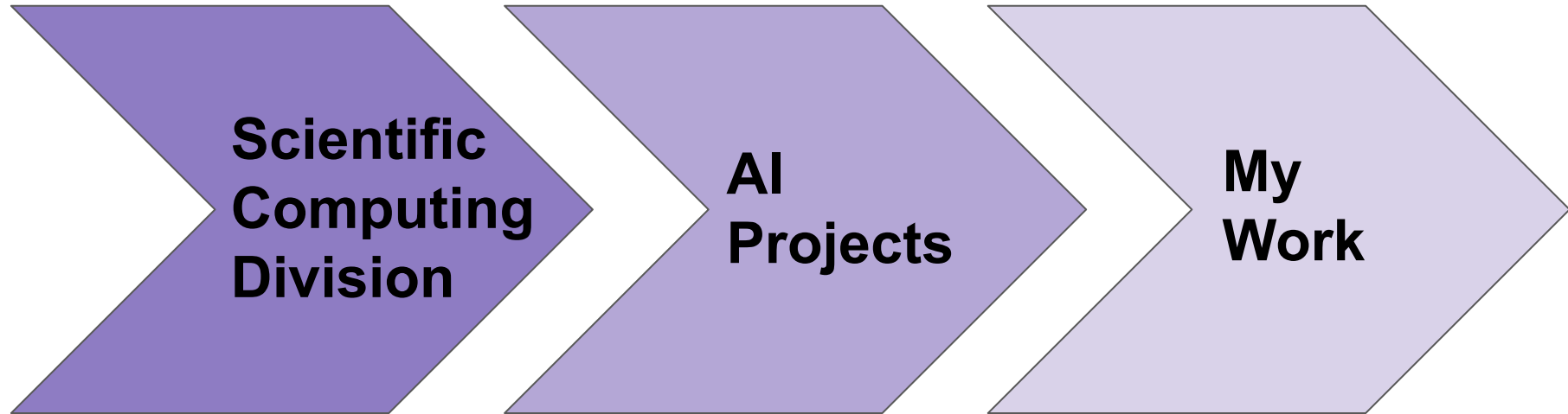
Scientific Computing Division and Me

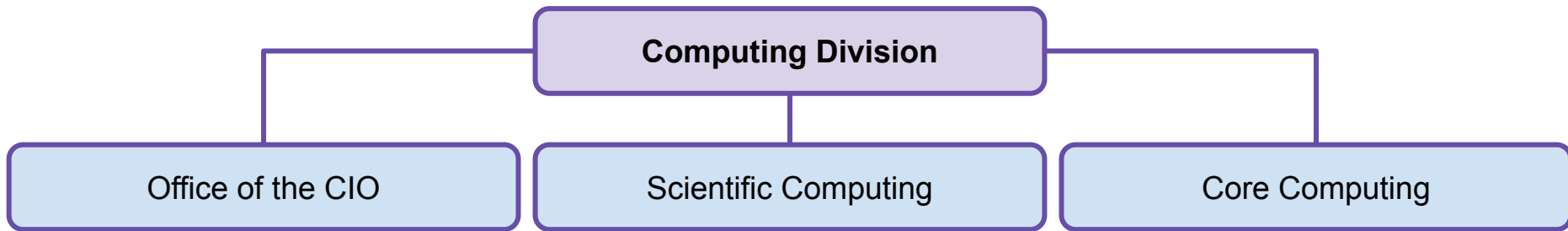


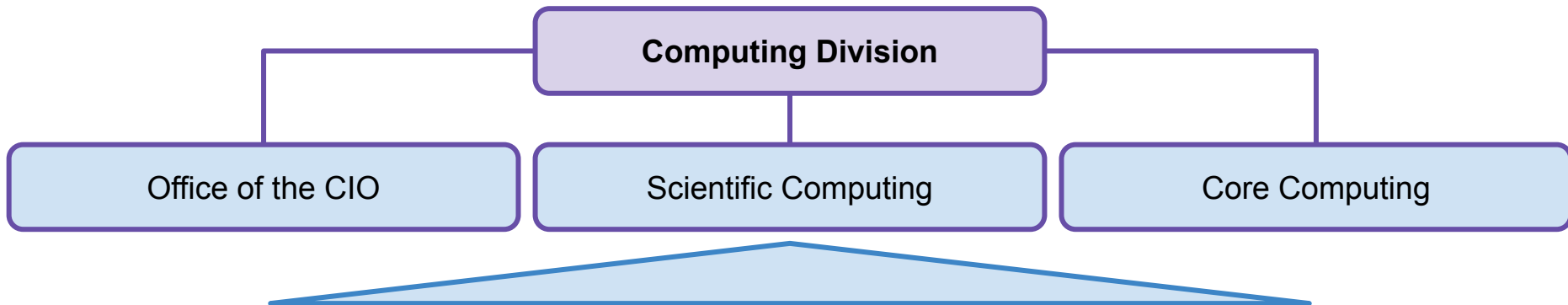
Aleksandra Čiprijanović

New Perspectives 2021
aleksand@fnal.gov

Talk Outline







- Scientific Computing Services
- Scientific Data Services
- Scientific Computing Facilities
- Cross Cutting Projects & Initiatives
- AI & Software for Physics Applications
- Framework, DAQ & Electronics ...

- Deliver world-class **computing services, operations and software** engineering support to Fermilab-based experiments, astronomical surveys, CMS and the high-energy physics community at large.
- Engages in **R&D activities required to maintain or advance capabilities** of Fermilab's physics program.

AI capabilities and focus areas

Uncertainty quantification
Learning on graphs

Theory and
new algorithms

Coprocessors for fast inference
Distributed training

Computing hardware
and infrastructure

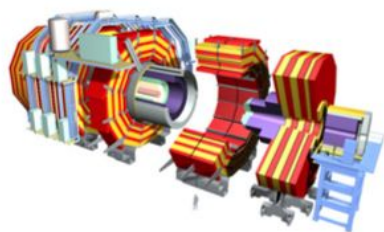
Accelerate Discovery
Science

Operations and
control systems

Real-time AI
at sensor/edge

Self operating telescopes
Experiment controls

hls4ml and FPGAs
ASIC codesign



Accelerators



Neutrinos



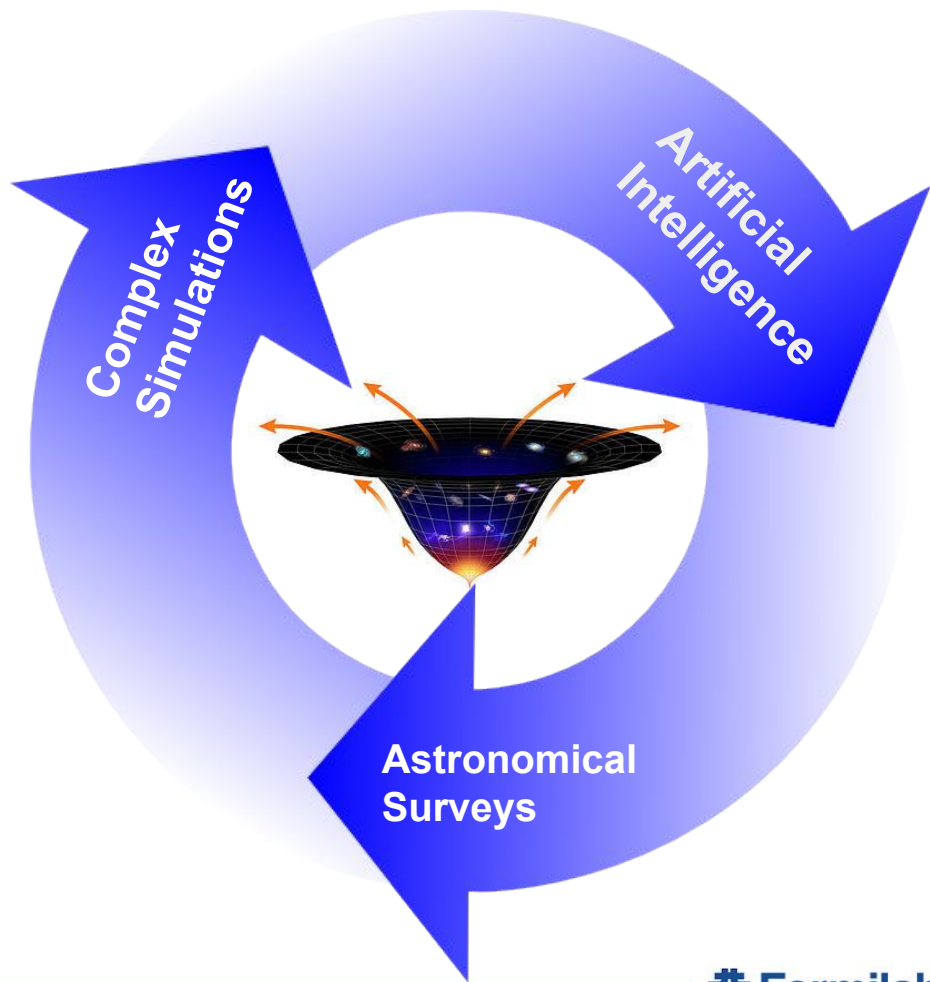
Astronomy



Quantum Science

My science interests

- Formation and evolution of structure in the Universe
- Build robust, trustworthy and understandable AI
- Leverage and use all available **datasets** - learn from their similarities and differences



My Projects

I

Convolutional Neural Networks

Classification of astrophysical objects and inference of the physical parameters that describe them.

II

Combining datasets and building algorithm robustness

Domain adaptation for learning domain invariant features.
Building models that work in multiple data domains.
Understanding adversarial robustness.

III

Object detection with R-CNNs

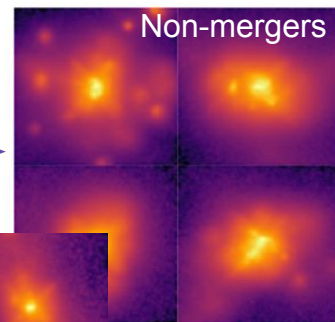
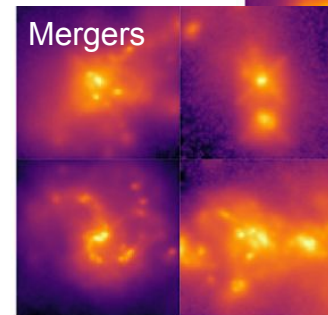
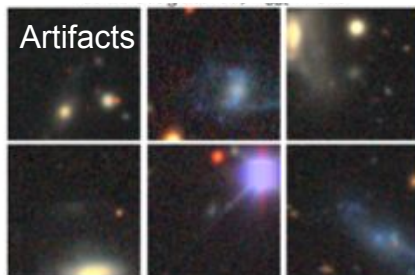
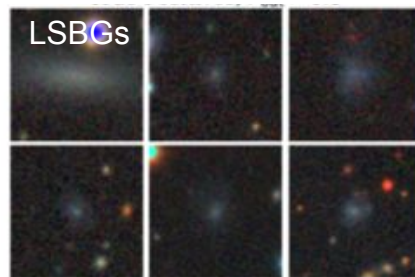
Using detection algorithms to help find low surface brightness objects in survey data or detect artifacts and other problems.

My Projects

Convolutional Neural Networks

“DeepMerge”: Ćiprijanović et al. 2020.
“DeepShadows”: Tanoglidis et al. 2020.

- Using deep learning (CNNs) to **distinguish between merging and non-merging galaxies**, which are crucial for understanding of galaxy evolution.
- We show it is **possible even for very distant galaxies!**



- Using deep learning to **distinguish low surface brightness galaxies (LSBGs) from artifacts** in DES data.
- Crucial for building their catalogues, LSBGs might be the most numerous type of galaxies!

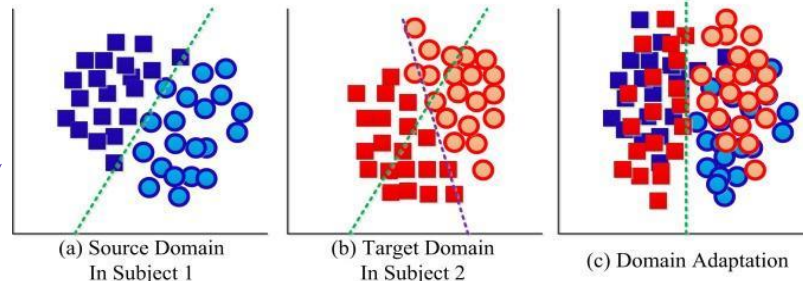
My Projects

II

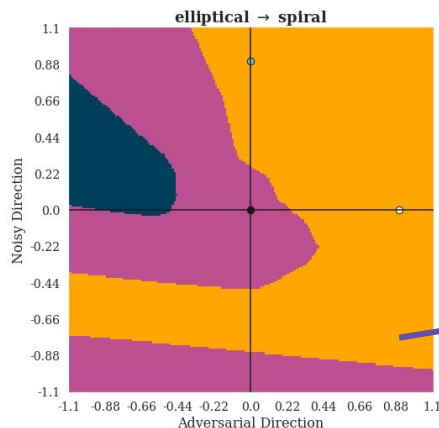
**Combining datasets
and improving
algorithm robustness**

You heard more in from
Kathryn Downey the
previous talk !

“DeepMerge II”:
Ćiprijanović et al. 2021.



- **Domain adaptation methods** allow us align data distributions and find common decision boundary.
- We can use **combine knowledge from simulation with new and unlabeled observations from astronomical surveys!**



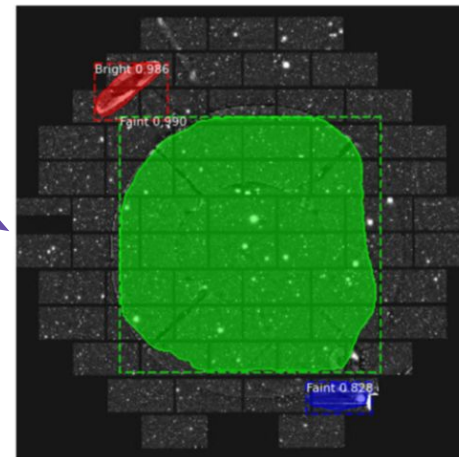
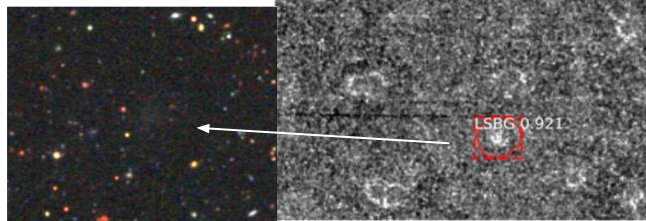
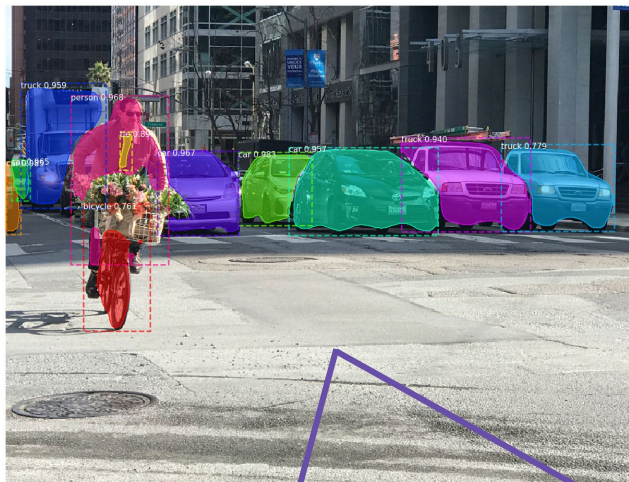
- **Exploring latent space of the model**, and how observational effects like **noise** can **cause errors in predictions**.

My Projects



Object detection with R-CNNs

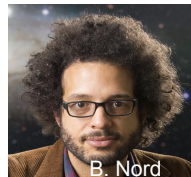
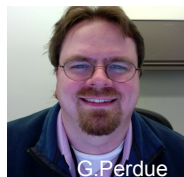
- Using **Mask R-CNNs** used for **object detection** and masking to search for hard-to-find **LSBGs**, as well as **ghosts** and other problematic artefacts in **DES data**.



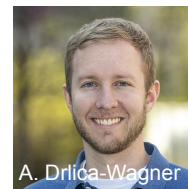
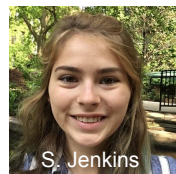
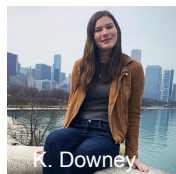
“DeepGhostbusters”:
Tanoglidis et al. *in preparation*

Big thanks to all my amazing collaborators!

Fermilab



University of Chicago

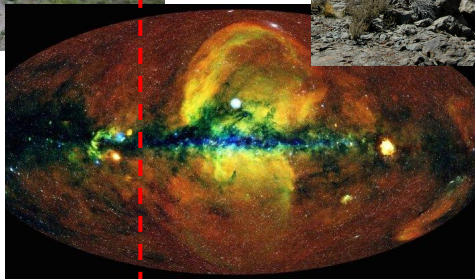
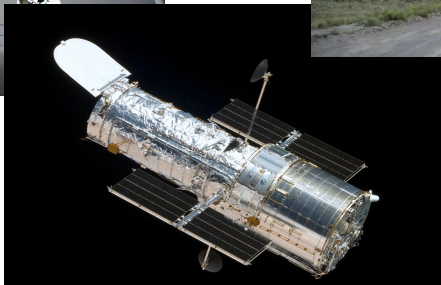
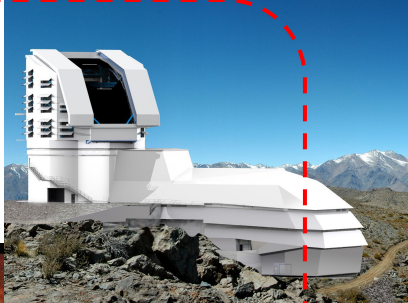


Argonne, Oakridge



Space Telescope Science Institute





Are we alone?

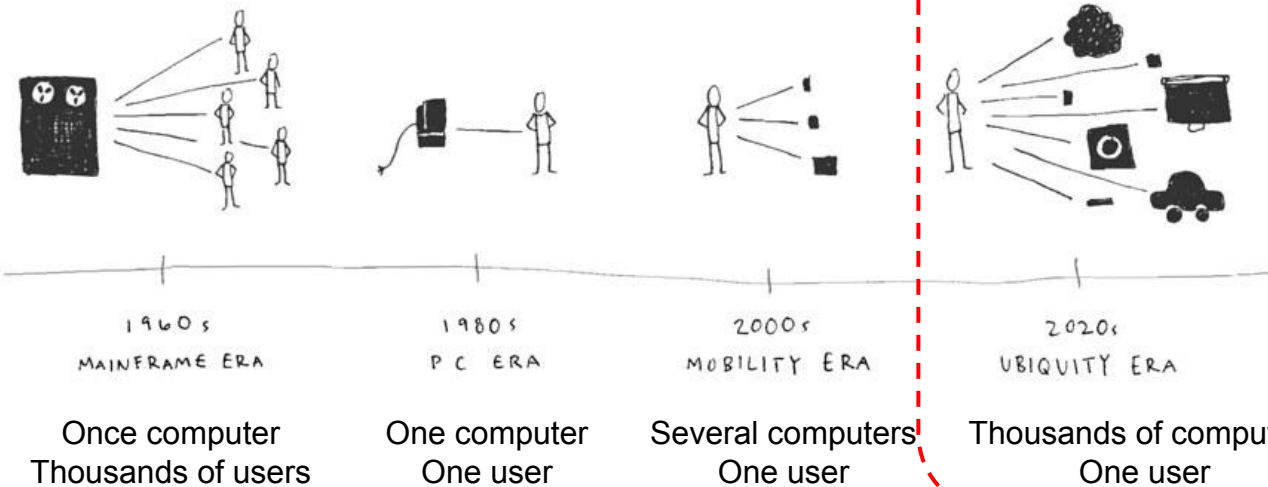
Exoplanets, search for life

How did we get here?

Stars and elements, galaxies and their history, structure formation

How does our Universe work?

The extremes of nature, dark matter and dark energy



Thank you!

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