

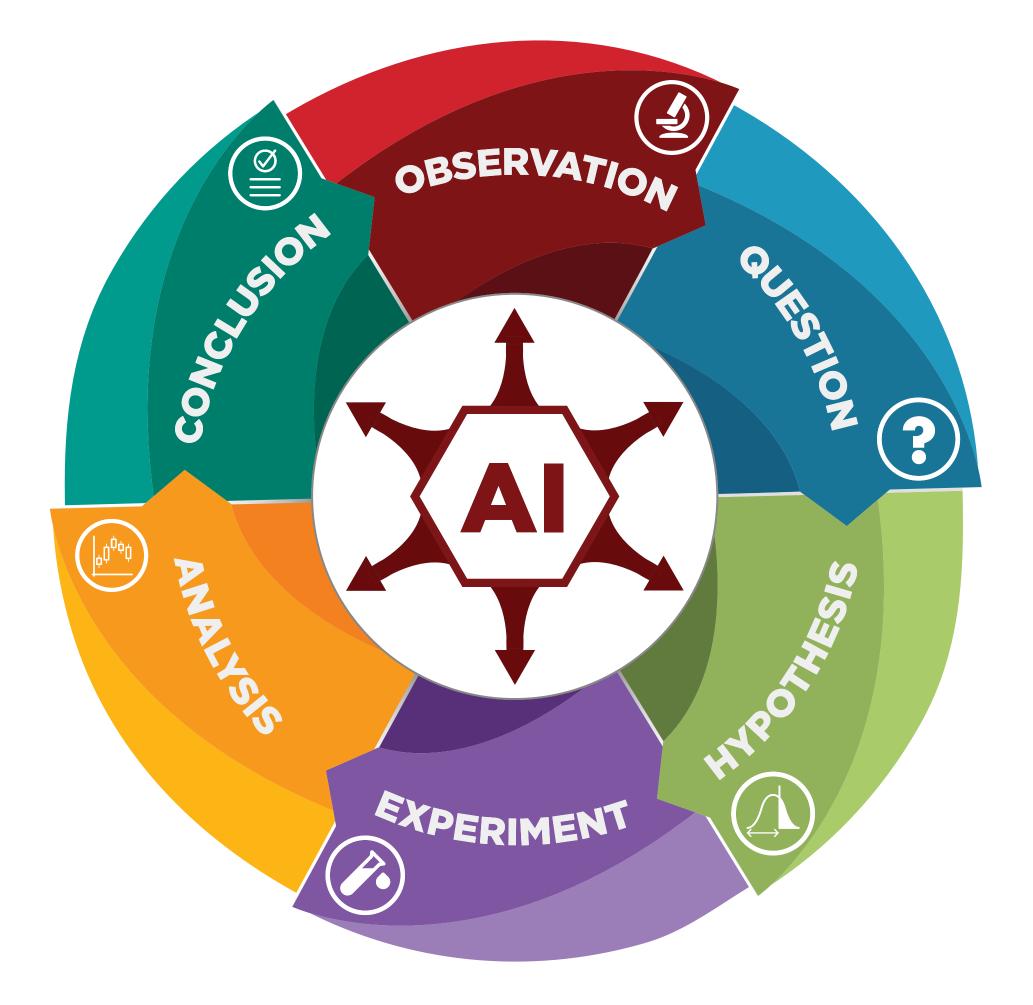
## **Groundbreaking Discoveries and Translation**

- Develop a new understanding of the laws of nature and rules of life
- Accelerate affordable drug development
- Engineer green materials
- Build quantum computers
- Develop sustainable climate policies



image credit: https://www.greenbiz.com/article/whats-your-sustainability-moonshot







#### Al will fundamentally change the nature and pace of scientific discovery, influencing data analysis, hypothesis generation, simulation, and experimental design





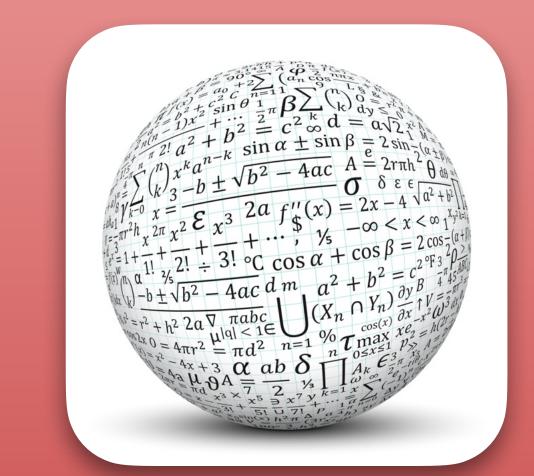
WILL KNIGHT BUSINESS AUG 10, 2022 7:00 AM

## Sloppy Use of Machine Learning Is Causing a 'Reproducibility Crisis' in Science

Al hype has researchers in fields from medicine to sociology rushing to use techniques that they don't always understand-causing a wave of spurious results.



Developing Al4Science without understanding Al foundations is like developing biotech without understanding biology.



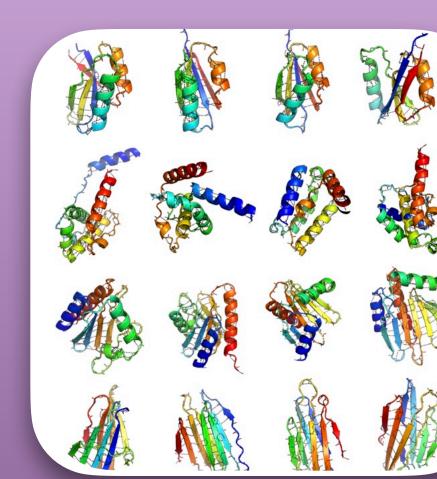
### Uncovering new laws of nature



### Al-guided scientific measurement

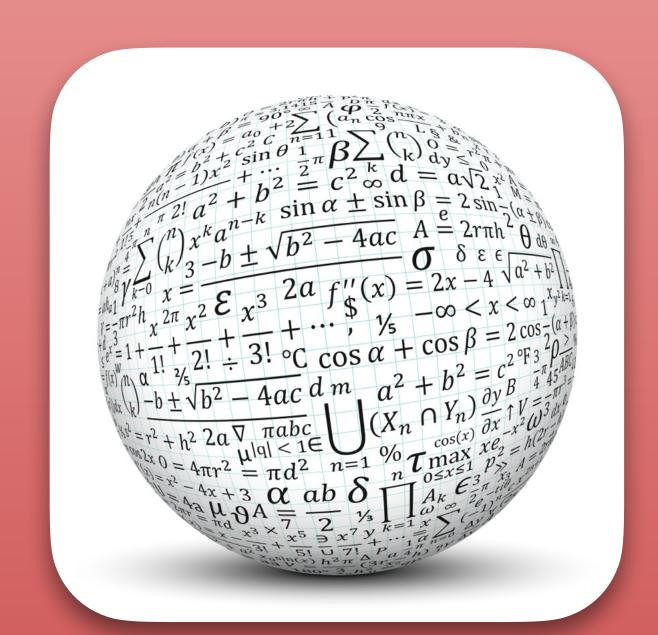


## Physicsinformed AI

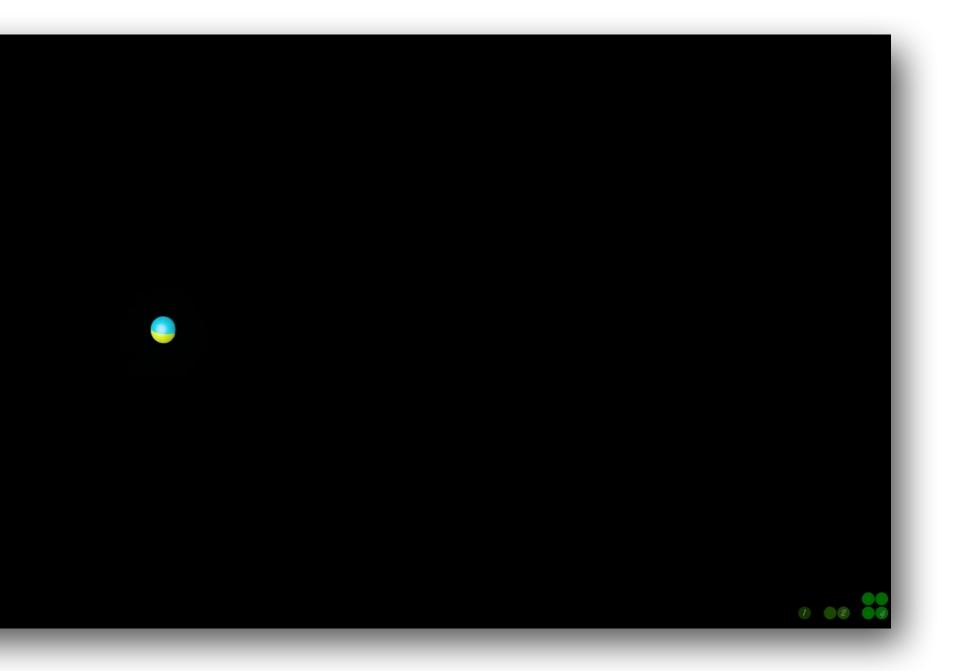


Generative Al for Science





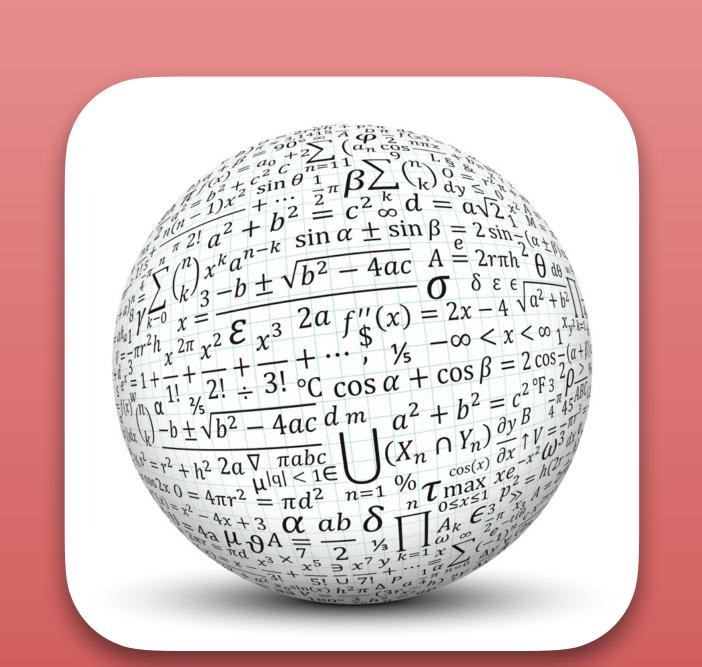
Given observations of a system, use Al to uncover the governing physical laws



 $\frac{dx}{dt} = \sigma(y - x)$ dyz) - ydt  $\frac{dz}{dt} = xy - \beta z$ 

https://www.youtube.com/watch?v=q3kNHomvU0k





Given observations of a system, use Al to uncover the governing physical laws

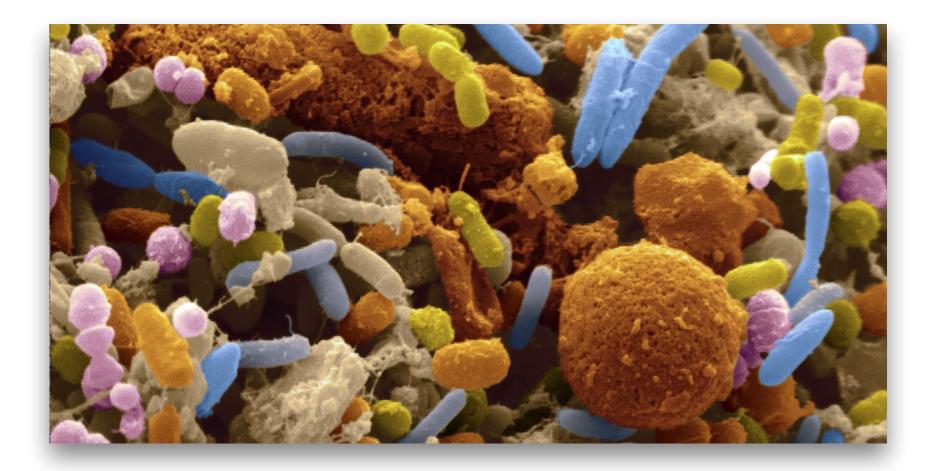




Vincenzo Vitelli

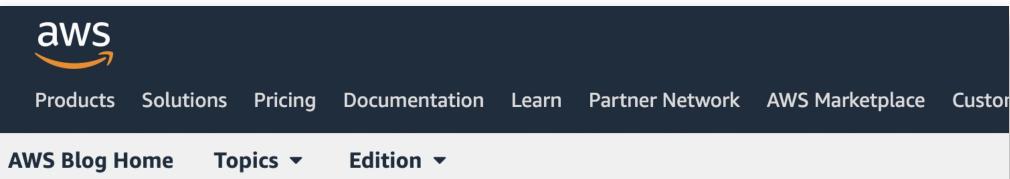


Use Al to design better experiments, simulations, and Sensors



Rina Barber uncertainty quantification foundations

#### Seppe Kuehn adaptive design of microbial communities



#### **AWS Machine Learning Blog**

#### **Introducing Fortuna: A library for uncertainty** quantification

by Gianluca Detommaso, Alberto Gasparin, Cedric Archambeau, Michele Donini, Matt and Andrew Gordon Wilson | on 16 DEC 2022 | in Amazon Machine Learning, Artificia Intelligence, Foundational (100) | Permalink | 🗩 Comments | 🏞 Share

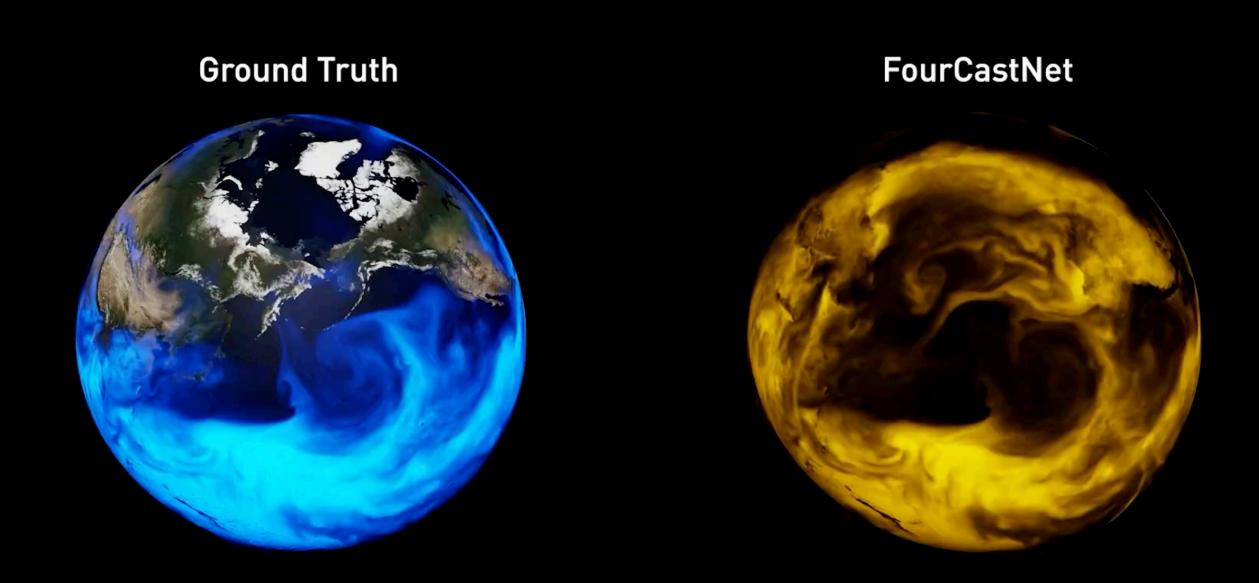
Proper estimation of predictive uncertainty is fundamental in applications that involv decisions. Uncertainty can be used to assess the reliability of model predictions, trigo intervention, or decide whether a model can be safely deployed in the wild.







Optimally leverage physical models and experimental or observational data

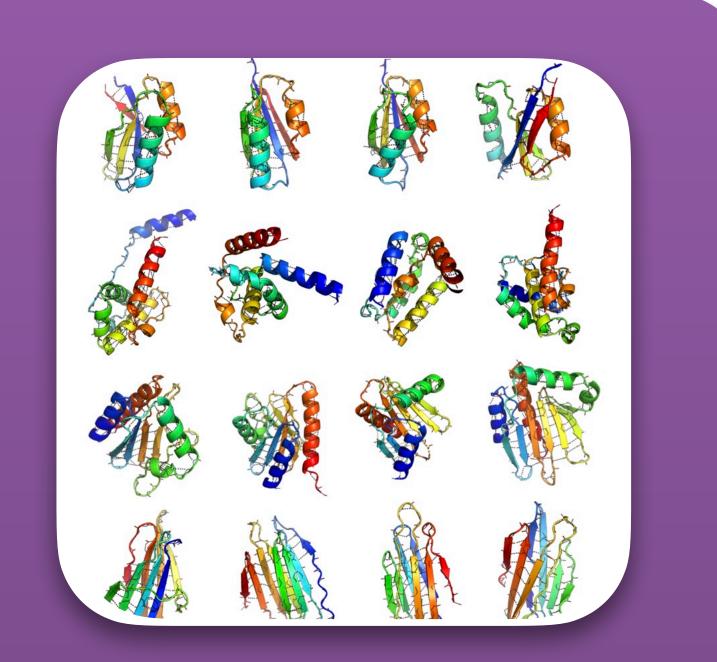


#### Pedram Hassanzadeh learned emulators of climate simulations

https://www.youtube.com/watch?v=nuT\_U1AQz3g







Generate new examples of physical objects or processes to aid discovery

- conditions
- horizon

Emulators are important for understanding statistical trends (e.g., hurricane frequency and severity over long time horizons)

Chaotic systems exhibit extreme sensitivity to initial

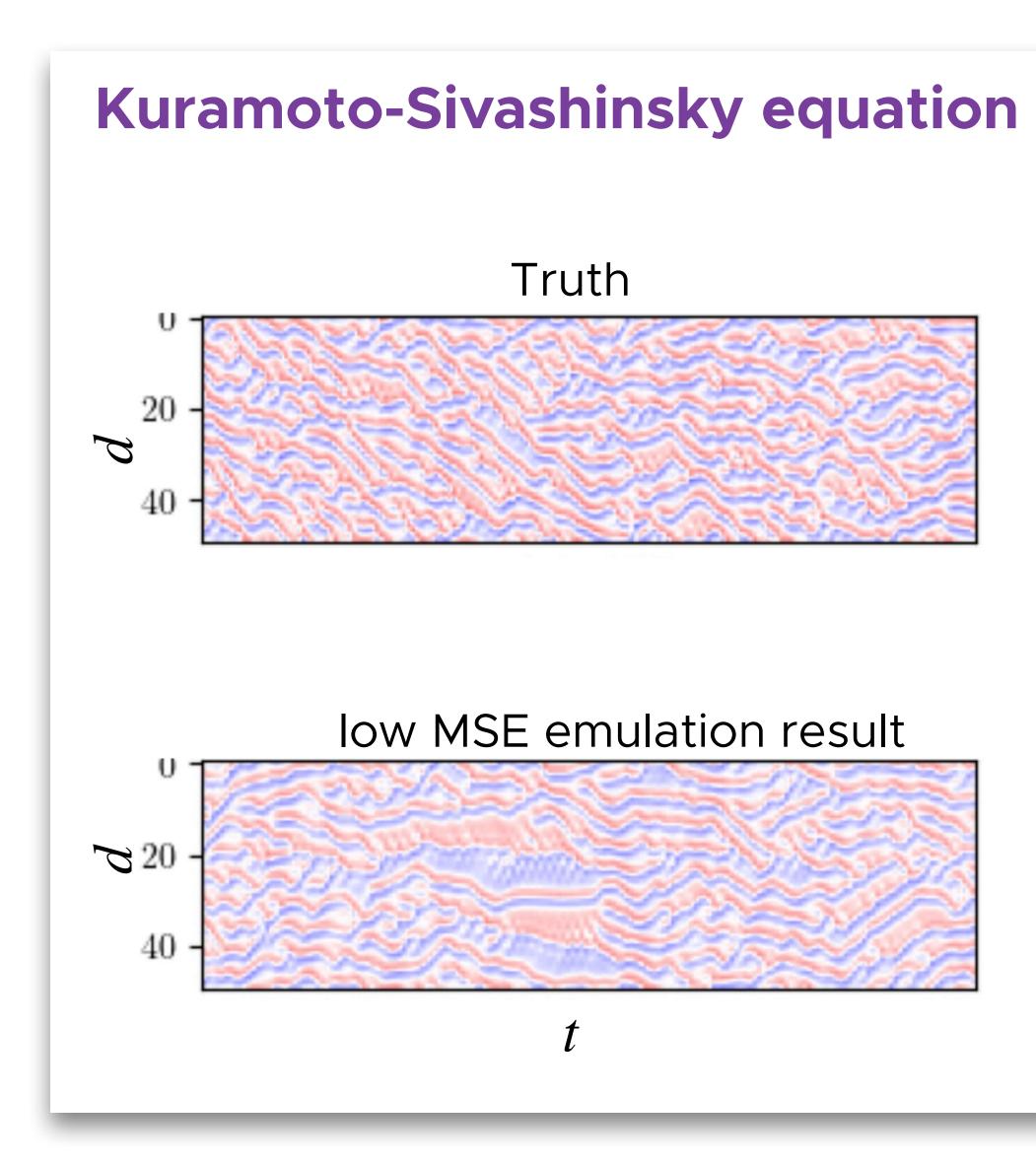
Emulators trained to give accurate "point" forecasts (e.g. small MSE) are doomed to failure past a short-term





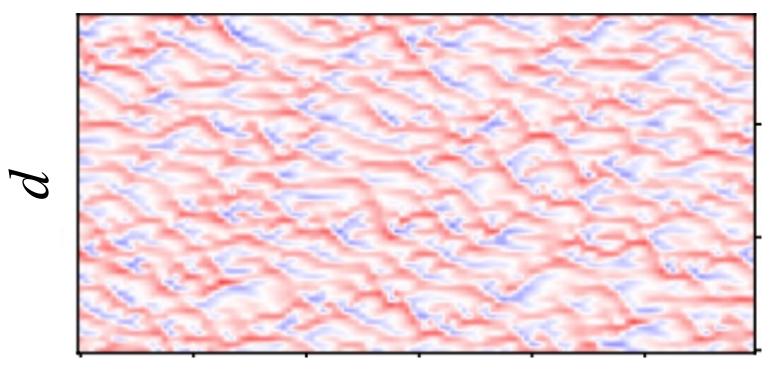


## Failure of MSE-minimizing emulators

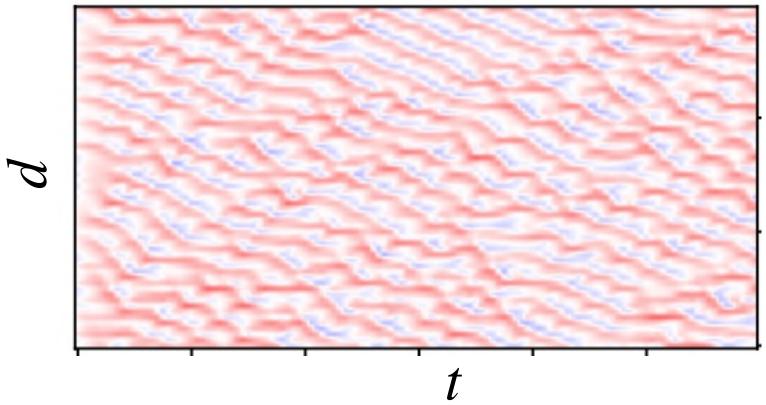


#### Lorenz 96 ODE

#### Truth



#### low MSE emulation result



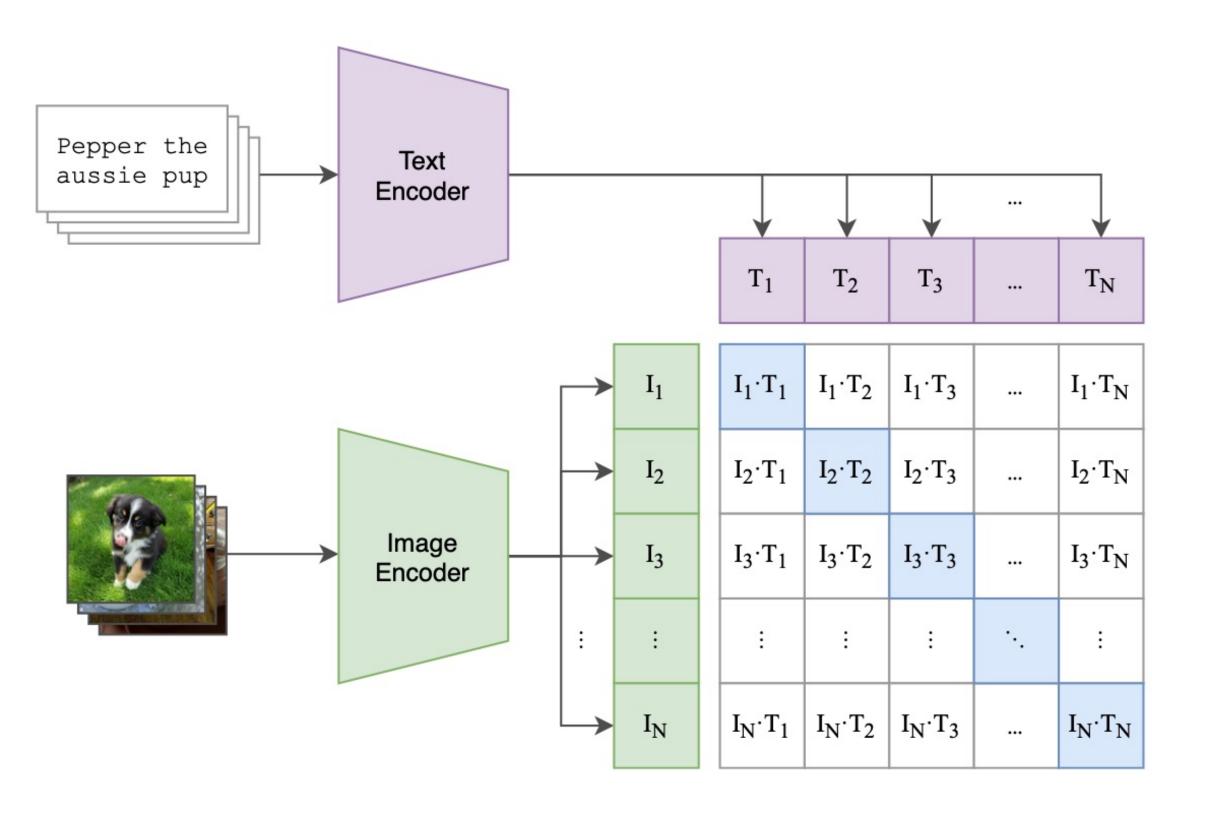


- statistics
- Contrastive learning preserves chaotic attractors

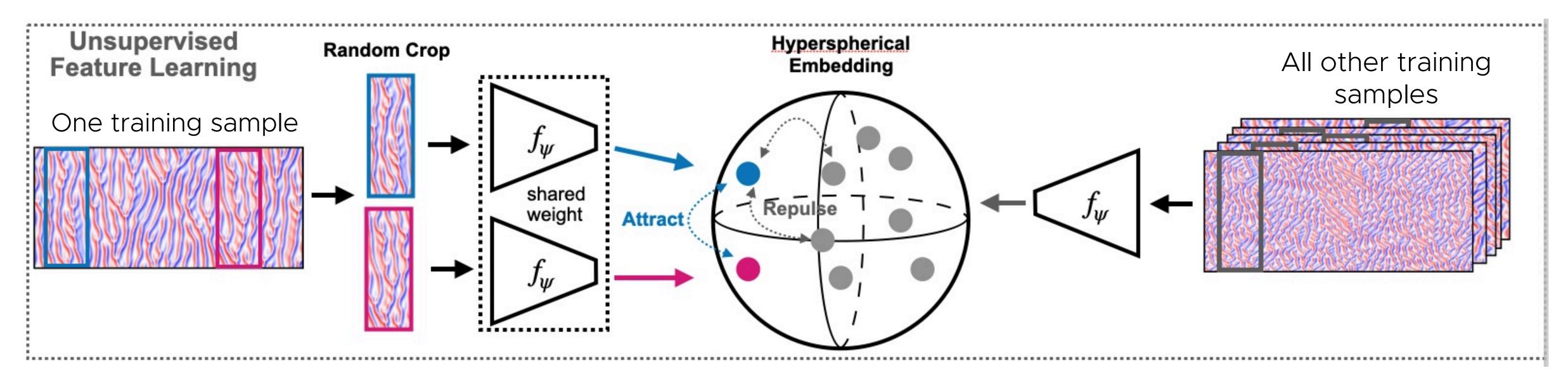
Example: CLIP (Contrastive Language-Image Pre-training)

## A new hope

#### Instead of focusing on small MSE, want emulator to preserve invariant measures of chaotic attractors and the corresponding time-invariant



## **Unsupervised contrastive feature loss (CL)**



Learn embedding of short time windows so that:

- nearby in embedding space and

two time windows from same simulation (i.e. with same chaotic attractors) are

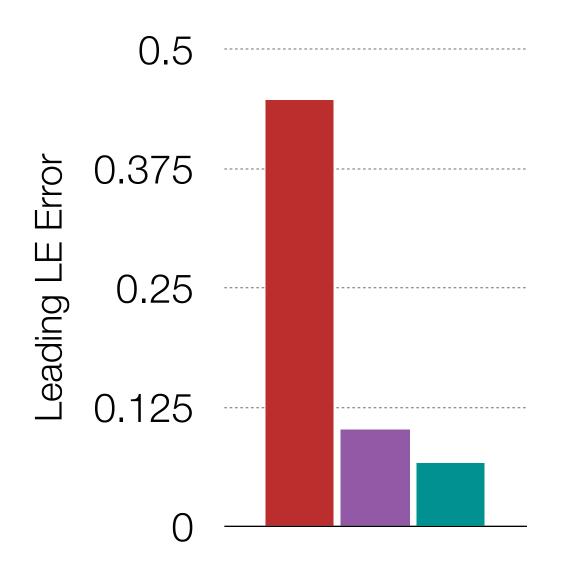
two time windows from different simulations are far away in embedding space

Jiang, Lu, Orlova, and Willett 2023

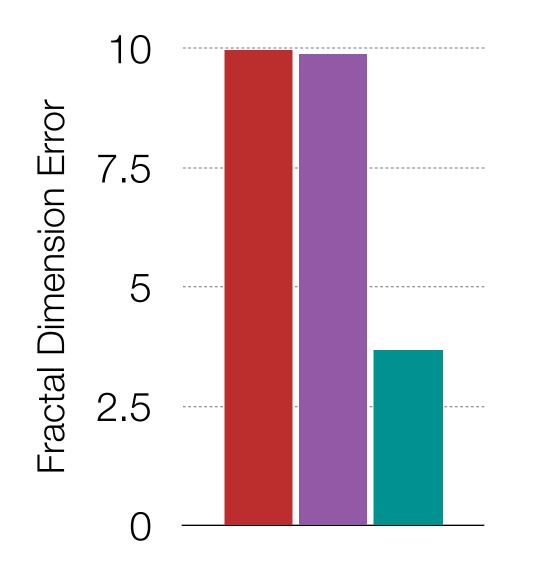
## Lorenz 96 Illustration

#### (Unknown) governing equatio

MSE loss
Optimal transport loss (assume extra knowledge)
Contrastive loss

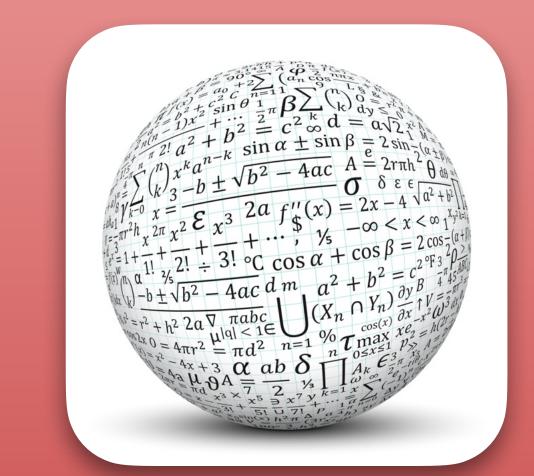


on: 
$$\frac{du_i}{dt} = (u_{i+1} - u_{i-2})u_{i-1} - u_i + F$$



Jiang, Lu, Orlova, and Willett 2023





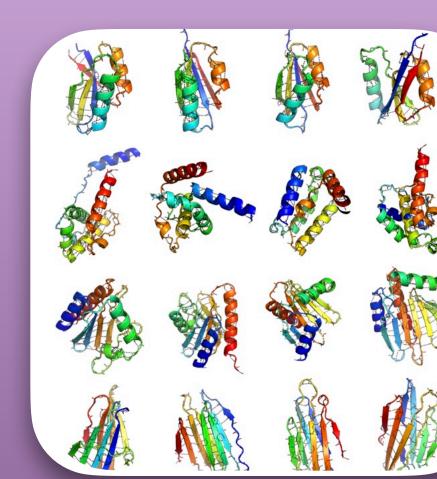
### Uncovering new laws of nature



### Al-guided scientific measurement

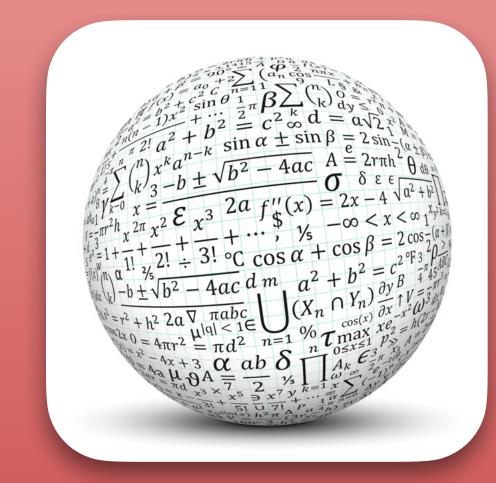


## Physicsinformed AI



Generative Al for Science





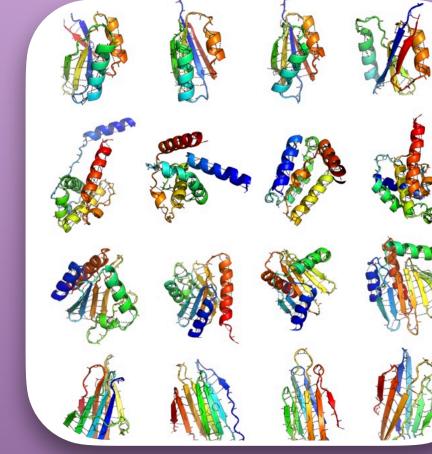
## Uncovering new laws of nature



**Al-guided** scientific measurement







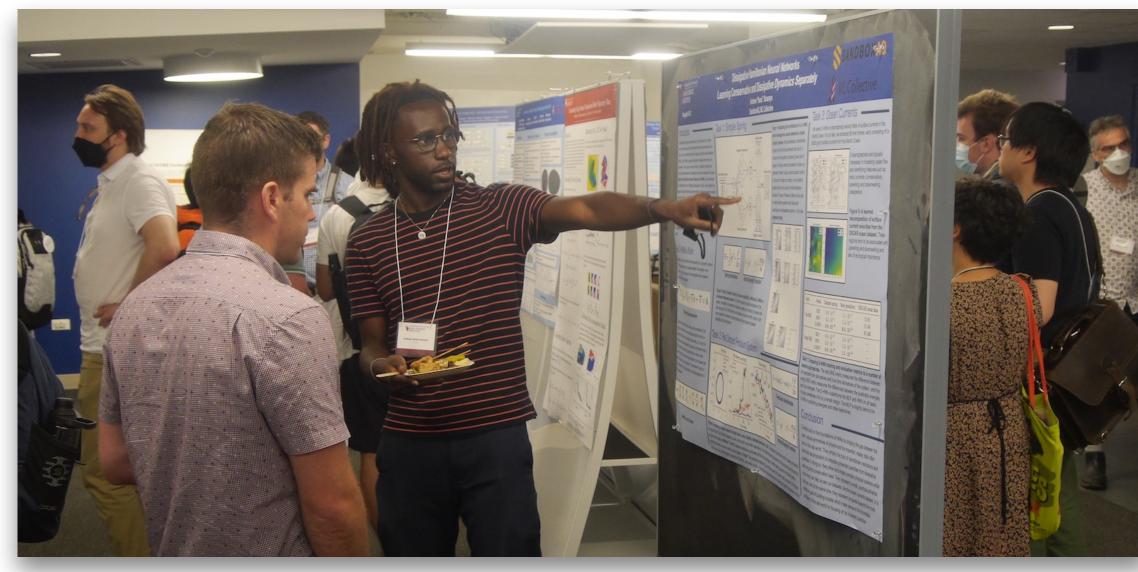
## **Physics**informed Α

### Generative **Al for** Science

#### **Al Foundations**





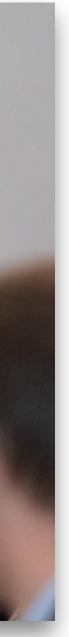




Caltech Concaco

The University of Chicago and Caltech Conference on AI + Science

Sponsored by the Margot and Tom Pritzker Foundation



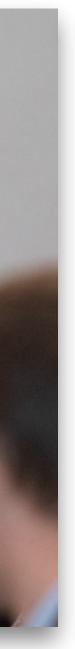






Training activities (summer schools, workshops, crossdisciplinary collaborations) **catalyze** ground-breaking research and **accelerate** workforce development

Sponsored by the Margot and Tom Pritzker Foundation







- Cross-disciplinary co-mentoring
- "Unseminars", including free-form Q&A sessions with experts
- Summer schools: 208 trainees from 66 institutions
- Conferences: 4600 views of conference videos
- Annual retreats
  - Communications workshops
  - Carefully designed science brainstorming sessions with cross-disciplinary teams

Key Catalysts

#### The Eric and Wendy Schmidt Al in Science **Postdoctoral Fellowship**



#### **E**SCHMIDT **FUTURES**



# Al foundations fuel Al4Science like biology fuels biotech.

## Thank you! willett@uchicago.edu



National Institute for Theory and Mathematics in Biology

#### THE UNIVERSITY OF CHICAGO

DATA SCIENCE INSTITUTE

