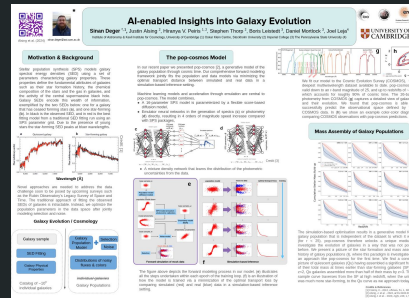




Alsing et al. (2024)  
arXiv:2402.00935



Poster Session B - May 1st



## AI-enabled Insights into Galaxy Evolution with pop-cosmos

**Sinan Deger**

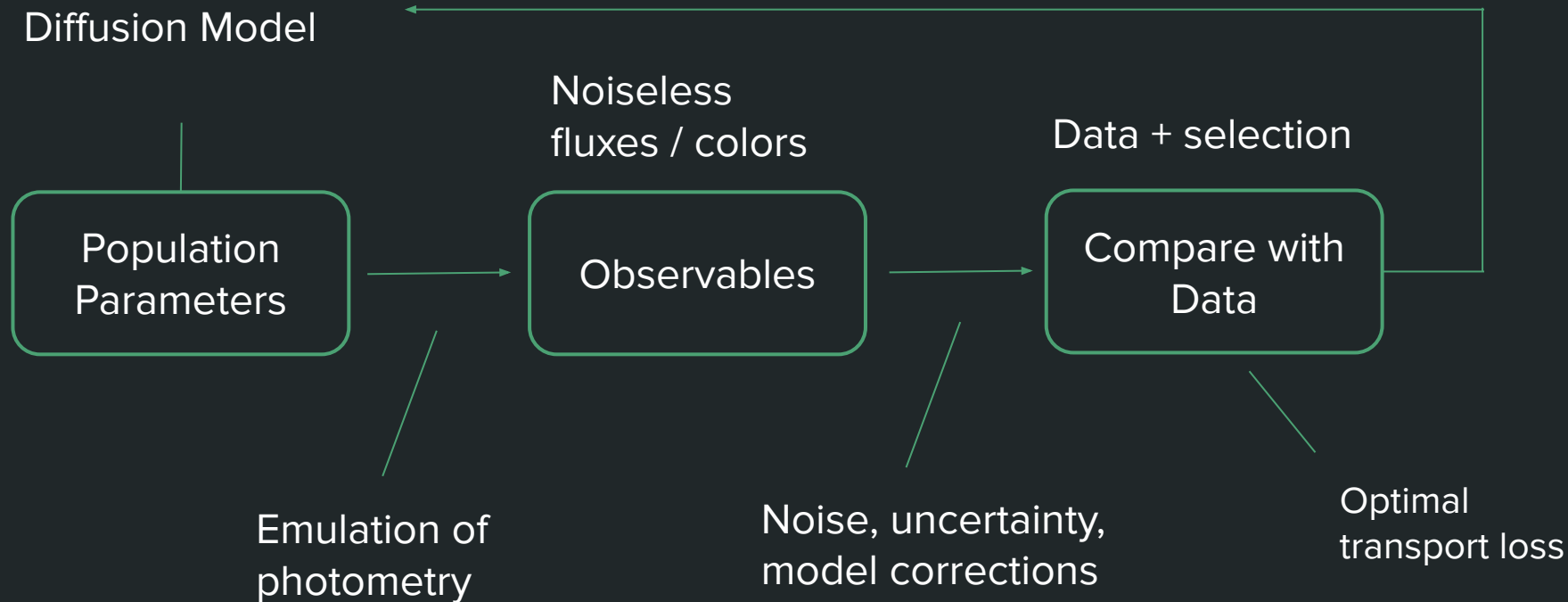
*Institute of Astronomy & Kavli Institute for Cosmology  
University of Cambridge*

With Justin Alsing, Hironya Peiris, Stephen Thorp<sup>\*</sup>, Boris Leistedt,  
Daniel Mortlock, and Joel Leja

<sup>\*</sup>At EuCAIFCon24

# The pop-cosmos Framework - An Overview

Diffusion Model

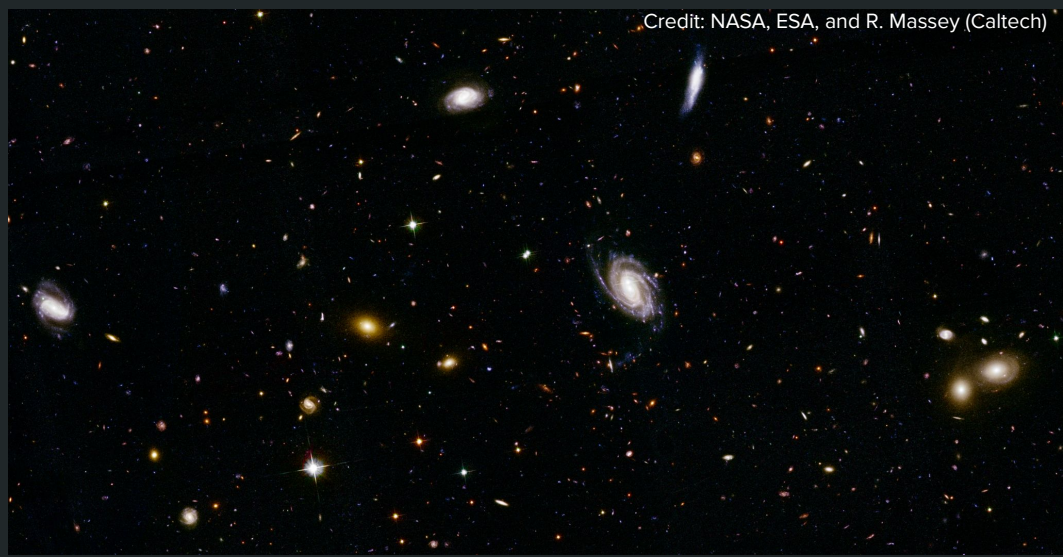


**We directly optimize population parameters on the data space by minimizing the optimal transport loss**

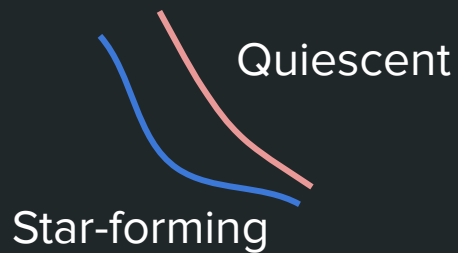
**Training loop for pop-cosmos**

# Fitting pop-cosmos to Observations

We fit pop-cosmos to COSMOS20 (Weaver+22), a deep galaxy survey with observations from the ultraviolet to the infrared.

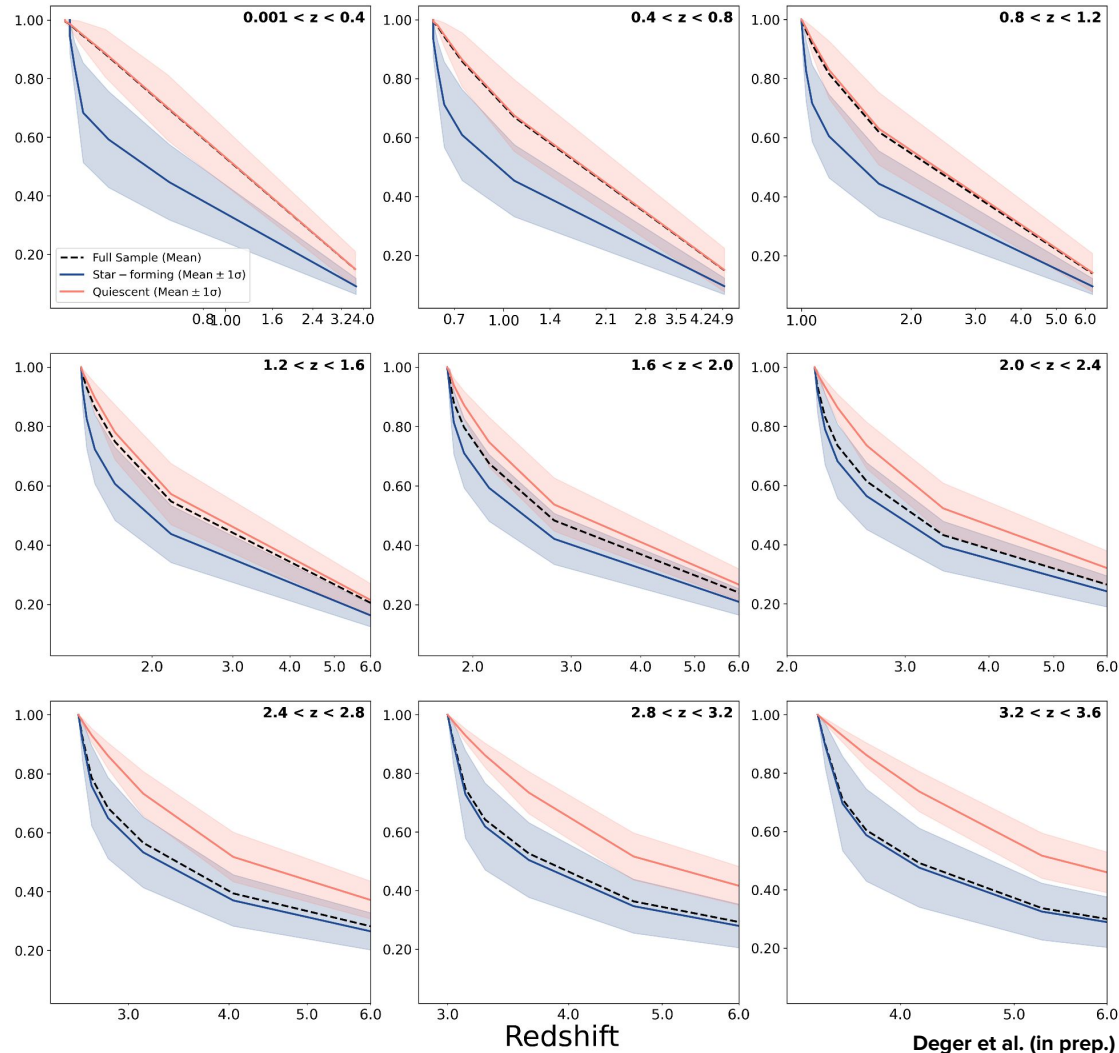


**The simulation-based optimization approach in pop-cosmos results in a generative model representative of general galaxy populations**



Our AI-enabled model  
unlocks a unique way to  
investigate the evolution of  
galaxy populations across  
90% of cosmic time ( $z < 4$ ).

Cumulative Stellar Mass Build-up



# Convolutional neural network search for long-duration transient gravitational waves from glitching pulsars

Rodrigo Tenorio - University of the Balearic Islands & IAC3

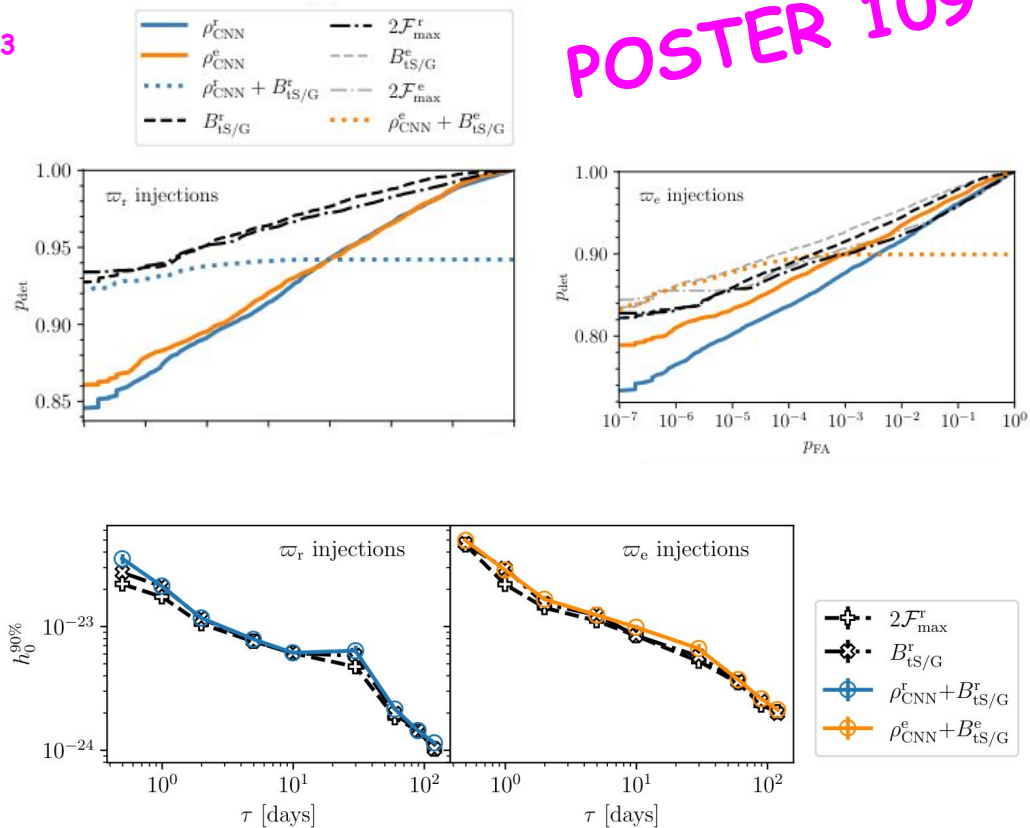
We present a machine-learning search for transient continuous gravitational waves (CWs) sourced by a glitch in the Vela pulsar during the Advanced LIGO O2 observing run.

The resulting pipeline is about 80 times faster than state-of-the-art pipelines at less than a 10% loss in sensitivity.

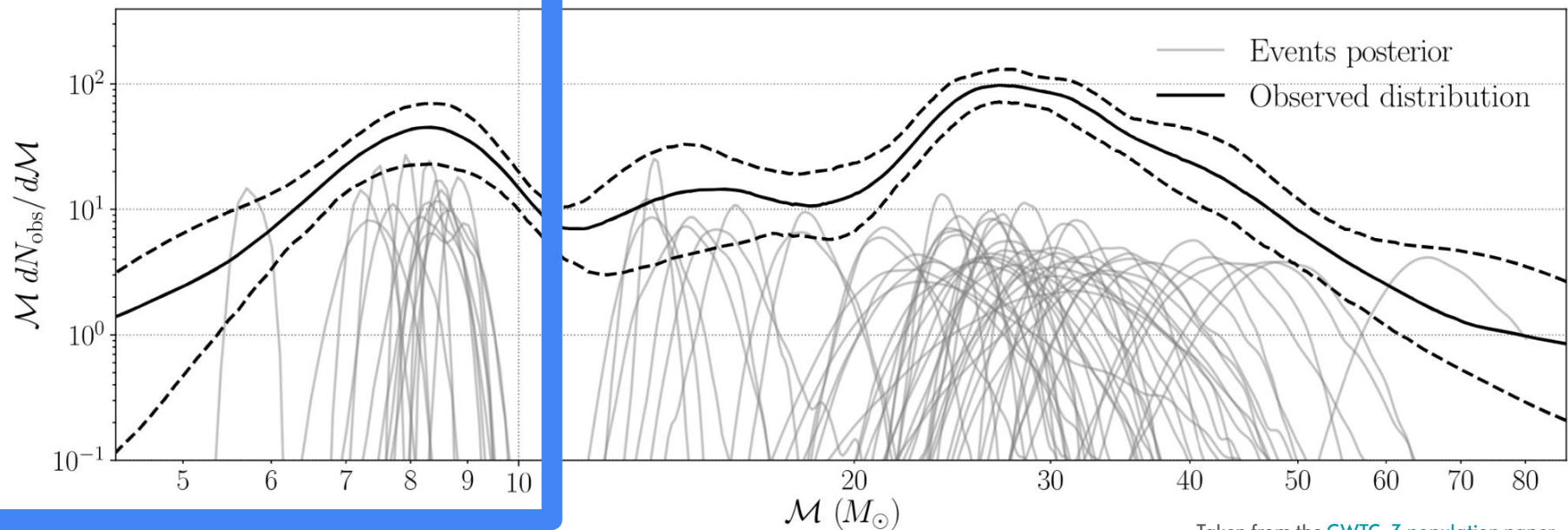
Transient CWs are decaying CWs produced by transient deformations in neutron stars such as decaying mountains or r-modes.

Physical amplitude-evolution models are computationally prohibitive due to the unknown duration and start time of the signal.

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**Do we train our SBI  
models optimally?**



Taken from the [GWTC-3 population paper](#)



# Tuning Neural Posterior Estimation for Gravitational Wave Inference

- Efficient training
- Low hardware requirements
- Yet, **quick inference!**

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