



# HYPERION:

a Normalizing Flow based pipeline for the rapid parameter estimation of eccentric Close Encounters

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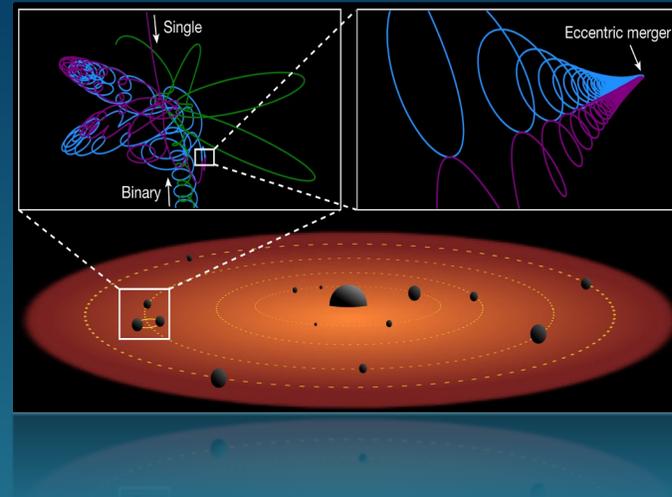
V Gravi Gamma Nu Workshop

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# Close Encounters in brief

Samsing J. (2022)



## ➤ High eccentricity ( $e \sim 1$ ):

⇒ GW emission: Repeated Bursts at each periastron

⇒  $v \sim 0.7c \rightarrow$  strong field regime / GR Tests

⇒ f-modes excitation in Neutron Stars  $\rightarrow$  Equation Of State constraints

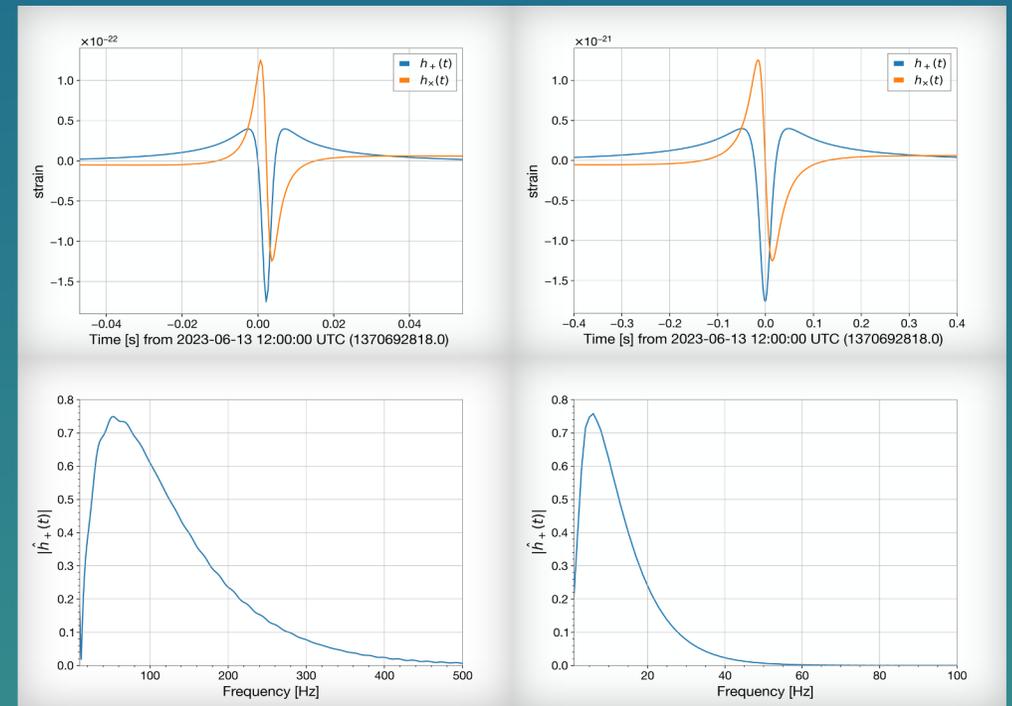
⇒ Dynamical formation channel  $\rightarrow$  Astrophysics

⇒ Possible e.m. counterparts  $\rightarrow$  Multimessenger Astronomy

Waveform Example with EFB-T model (Loutrel N. (2020))

## ➤ Challenges for PE and detection:

- Low SNR with current sensitivities
- Burst-like signal with small frequency evolution



$10 M_{\odot} + 10 M_{\odot}$

$100 M_{\odot} + 100 M_{\odot}$

# Normalizing Flows

Papamakarios G. (2019)

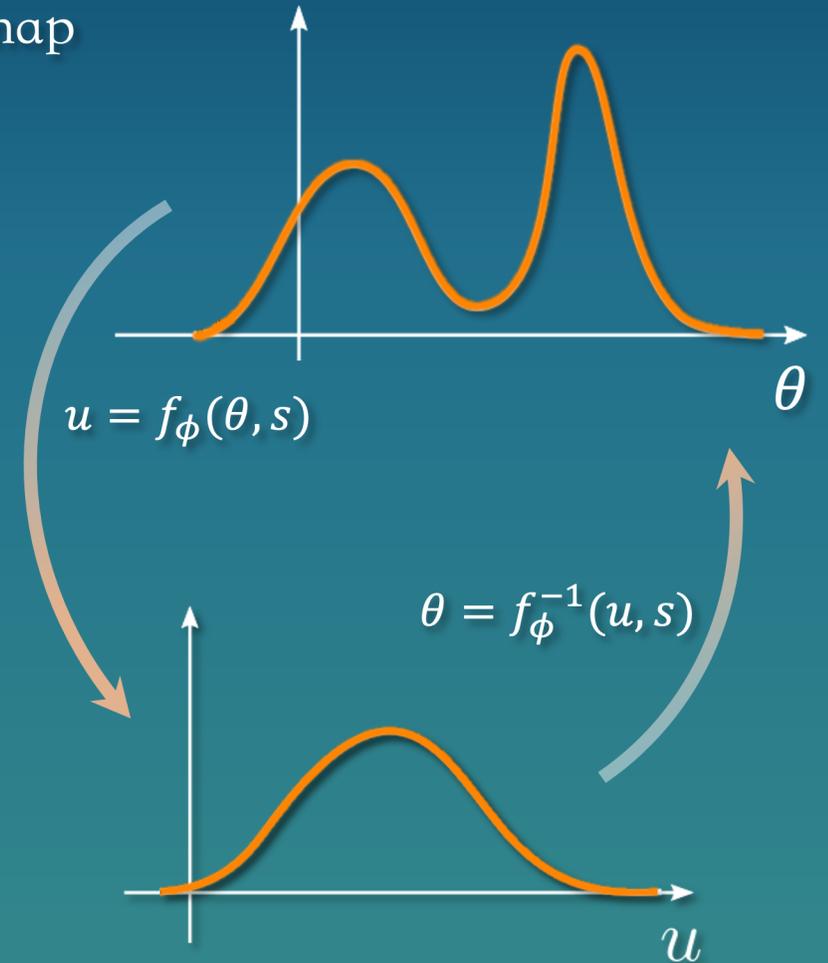
- **Key idea:** parametrize an analytical function  $f: \mathbb{R}^N \rightarrow \mathbb{R}^N$  to map GW parameters into a simpler distribution

$$\theta \xrightarrow{f(s)} u \sim \pi(u) = \mathcal{N}(0, \mathbb{I})$$

- $f$  must:

- be **invertible** and **differentiable**
- depend on parameters  $\phi \leftarrow \text{NN output}$

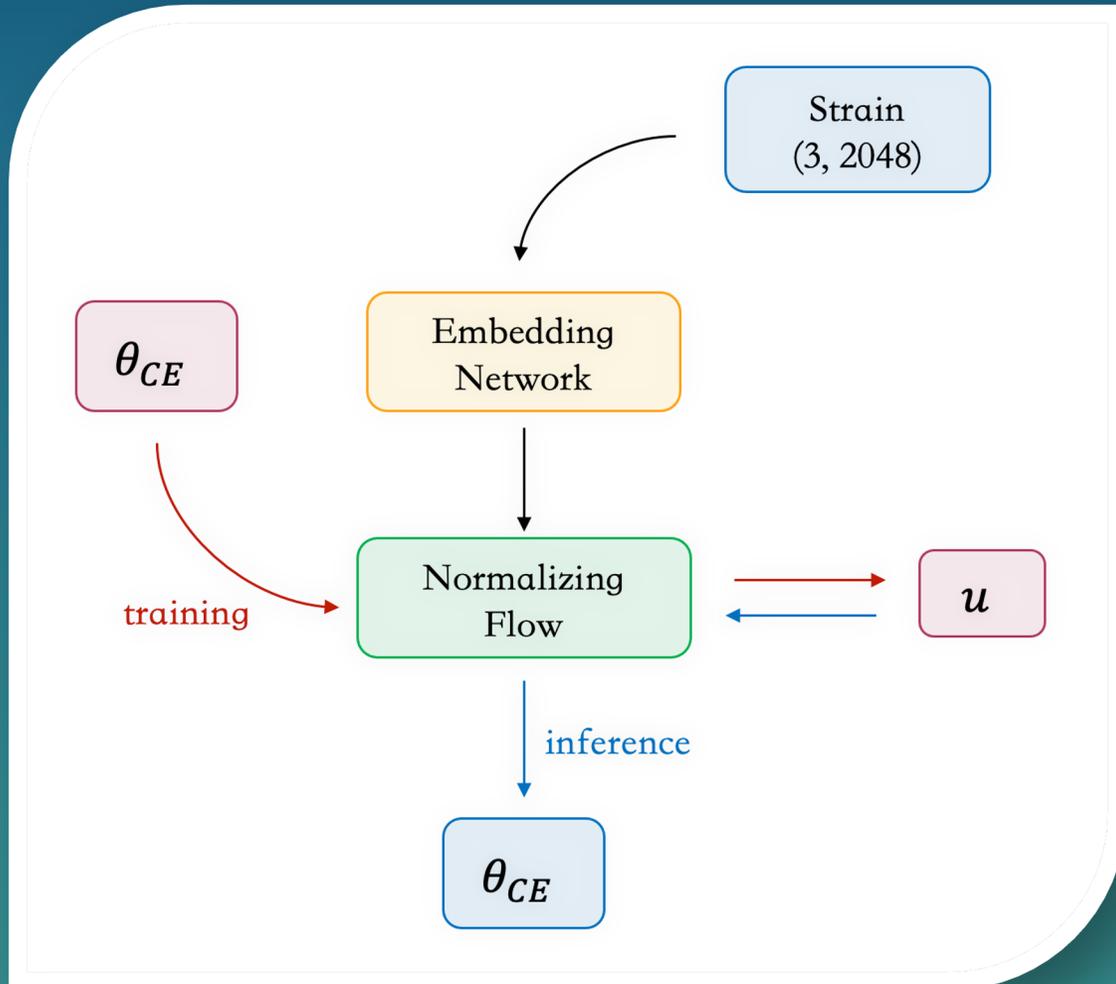
$$p(\theta|s) \approx q(\theta|s) = \pi(f_\phi(\theta, s)) \left| \det J(f_\phi(\theta, s)) \right|$$



# HYPERION: Model Overview

*De Santi et al., Phys. Rev. D 109, 102004 (2024)*

➤ **HYP**er fast close **EncounteR** Inference from **O**bservations with **N**ormalizing flows



➤ Developed in **Python + PyTorch**

➤ **Input:** Strain Timeseries

➤ **Output:**  $\theta_{CE} = \{M, q, e_0, \bar{p}_0, d_L, \delta t_p, \alpha, \delta\}$

➤ Core modules:

- **Embedding Network:** CNN, extracts features
- **Normalizing Flow:** coupling layers, reconstructs posterior distribution
- **~ 180 M** parameters

# HYPERION Performances

➤ Comparison with standard PE tools e.g. [Bilby](#): ( *Ashton G. et al. (2018)* ) :

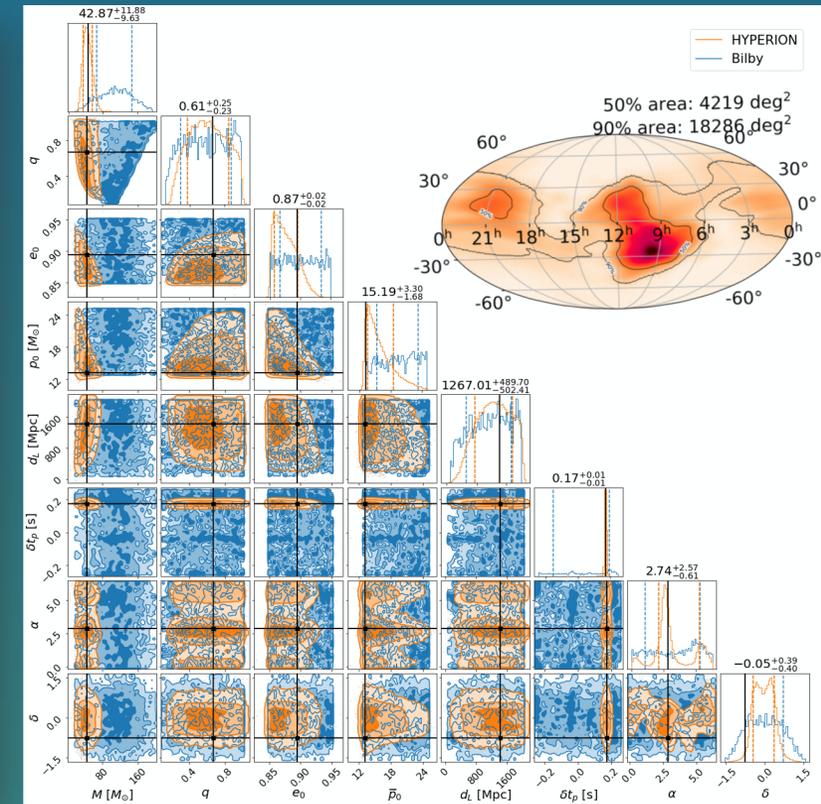
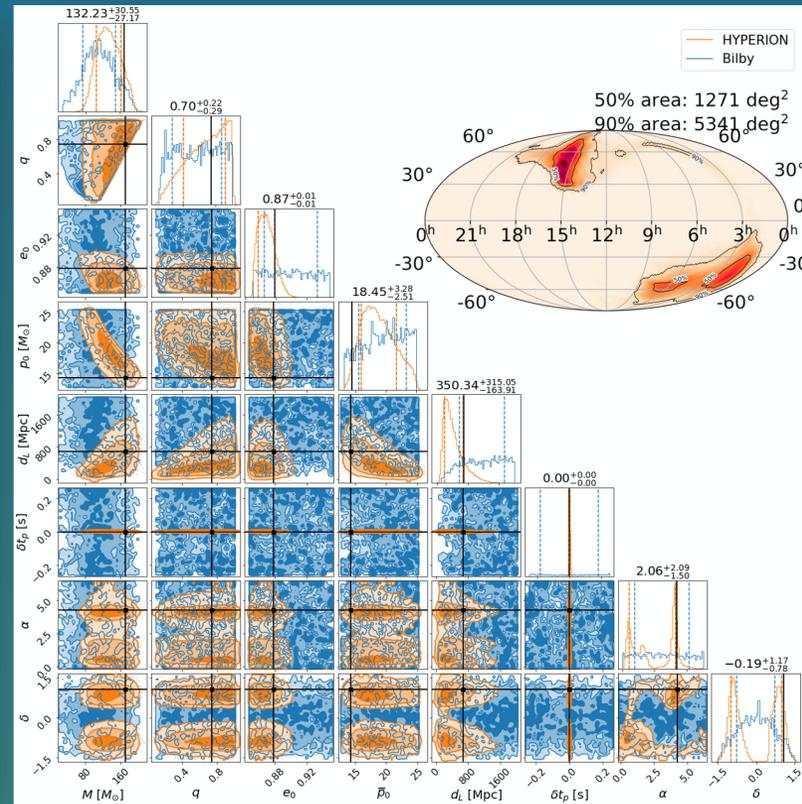
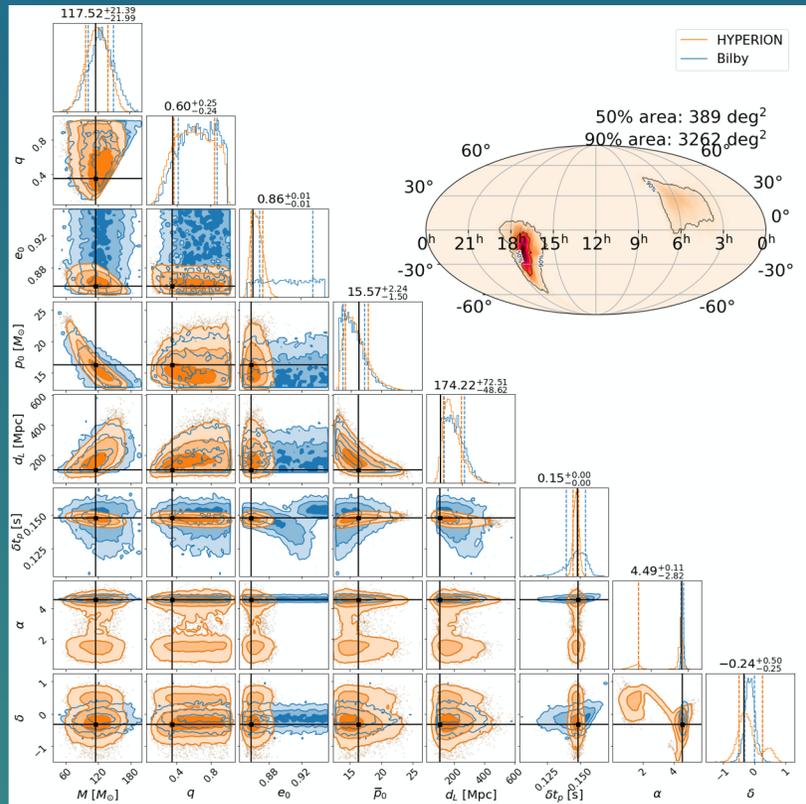
	Posterior samples	Inference time
Bilby	$\sim 5 \times 10^3$	$\sim 10$ h
HYPERION (CPU)	$5 \times 10^4$	$\sim 16$ s
HYPERION (GPU)	$5 \times 10^4$	$\sim 0.5$ s

**10<sup>4</sup> TIMES FASTER!**

➤ SNR  $\approx 30$

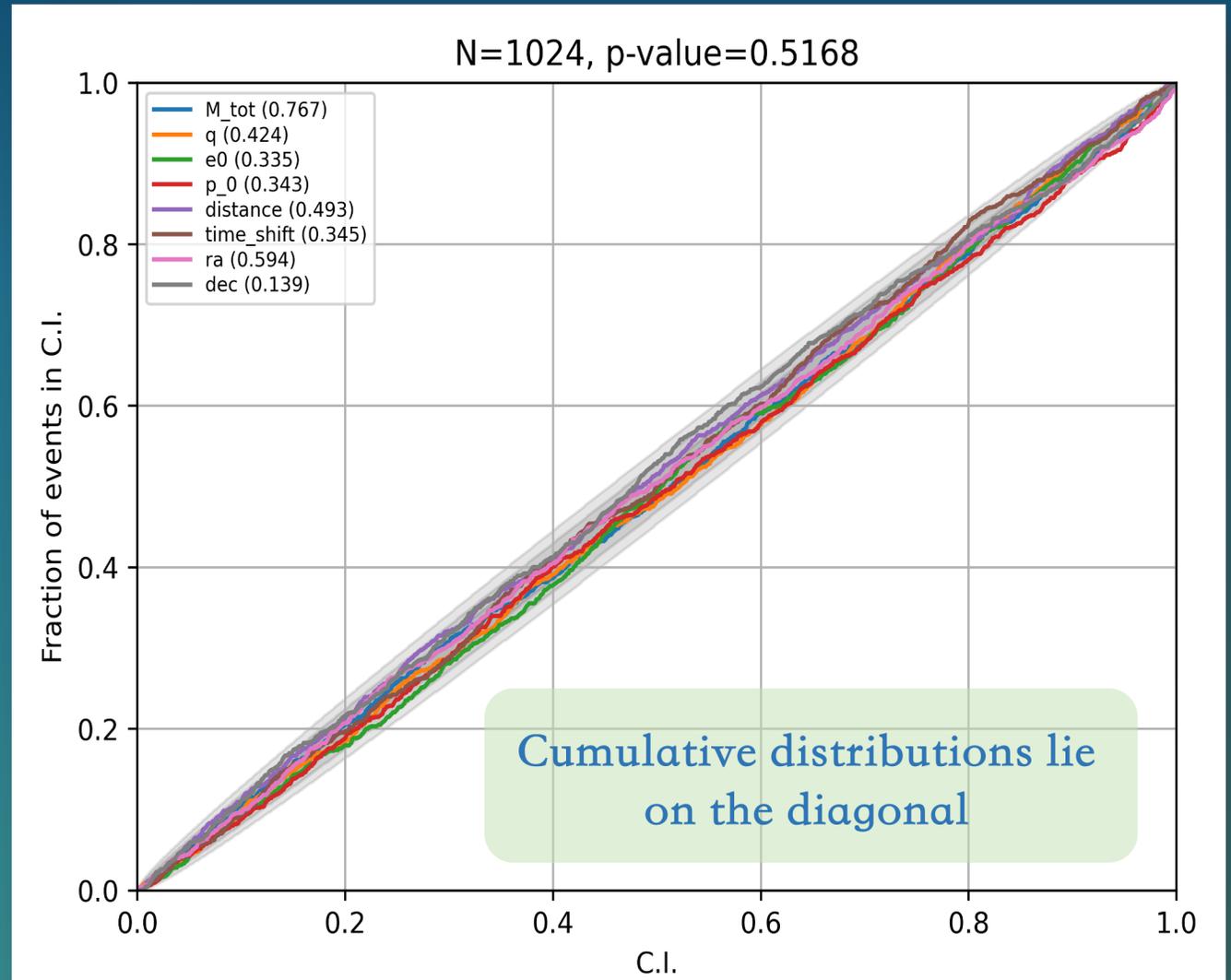
➤ SNR  $\approx 12$

➤ SNR  $\approx 6$



# HYPERION Performances

1. Obtain posterior for 1024 events from the test dataset
2. For each parameter compute the percentile score of the true value in the marginalized posterior
3. Take the cumulative distributions
4. Kolmogorov-Smirnov Test (95%)
  - Promising approach on simulations



# Conclusion & Future prospects

## ➤ Normalizing Flows for Close Encounters :

- CEs are interesting but challenging sources
- Likelihood – free approach
- $\sim 10^4$  times **faster** than traditional methods → low latency → e.m. follow-up
- To our knowledge, first application to CE
- Method paper: *De Santi et al., Phys. Rev. D 109, 102004 (2024)*

## ➤ Future work :

- Test detection capabilities on real data
- Perform systematic searches over O3

*Thank you for the attention!  
If you're curious about this work, please come and chat!*