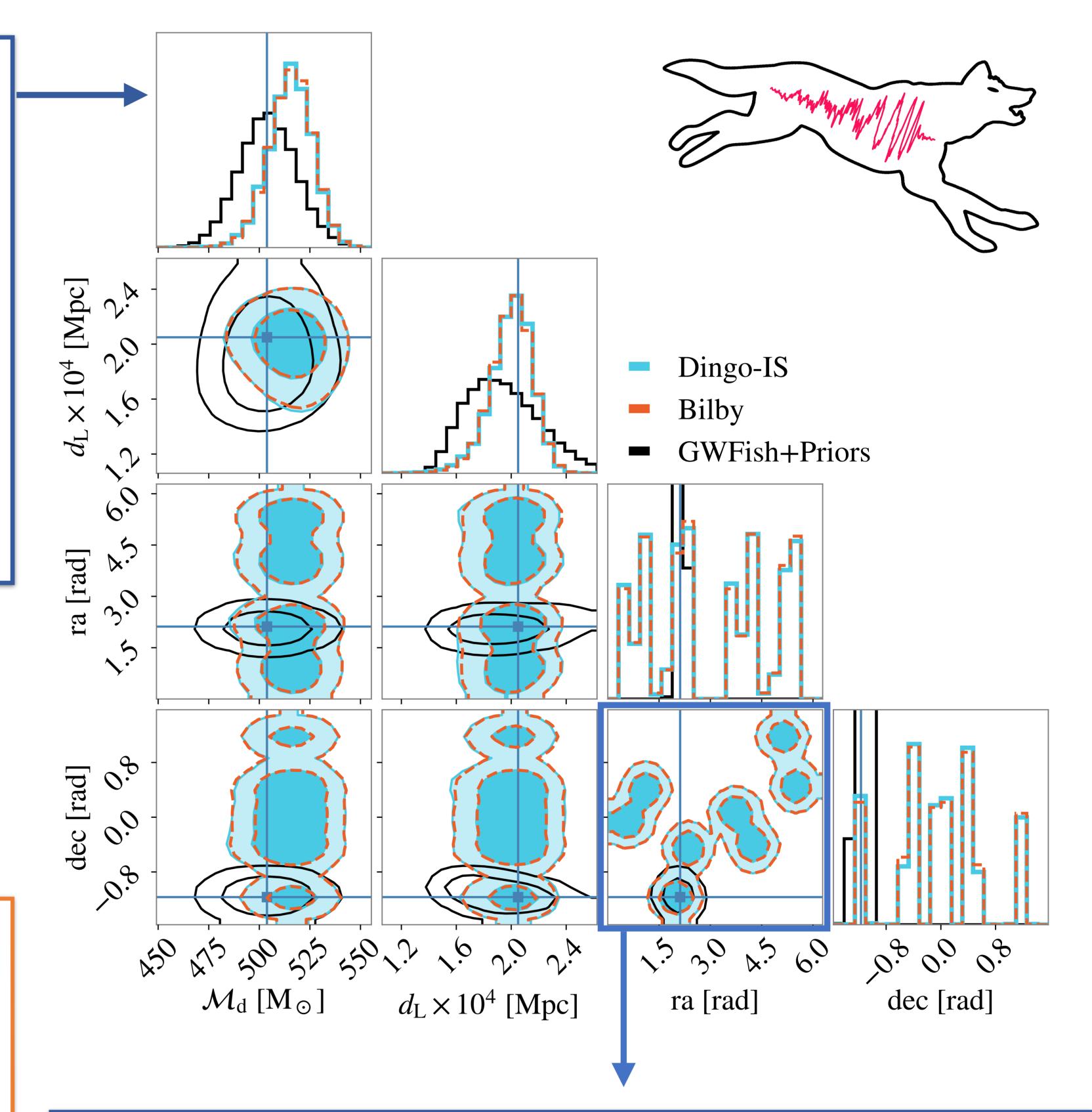
## Fast and accurate parameter estimation of high-redshift sources with the Einstein Telescope

In its triangular configuration, the Einstein Telescope will provide more precise distance measurements than sky localization.

**Background:** The Einstein Telescope will be a key instrument for detecting GWs in the coming decades [Punturo et al. 2010, Maggiore et al. 2020, Branchesi et al. 2023, Abac et al. 2025]. However, analyzing the data and estimating source parameters will be challenging, especially given the large number of expected detections – of order  $10^5$  per year – which makes current methods based on stochastic sampling impractical [Couvares et al. 2021]

## Main result

Posterior samples from Dingo-IS [Dax et al. 2021] and Bilby [Romero-Shaw et al. 2020] indicate that the two distributions are effectively identical. The corner plot shows also multimodalities that arise due to the geometry of a triangular detector [Singh and Bulik 2021]. These modes are absent by construction when using the Fisher information matrix approximation [GWFish+Priors, Dupletsa et al. 2025].



## Methods

In <u>Santoliquido et al. 2025</u>, we use Dingo-IS to perform Neural Posterior Estimation (NPE) of high-redshift events observable with the Einstein Telescope. NPE is a likelihood-free inference technique that leverages normalizing flows to approximate posterior distributions [<u>Dax et al. 2023</u>]. Once trained, this approach enables fast and accurate inference—typically requiring only a few minutes per source. To correct for any residual bias, we employ importance sampling,

$$w_i = \frac{\mathcal{L}(d \mid \theta)\pi(\theta)}{q(\theta \mid d)}$$

where  $\mathcal{L}(d\,|\,\theta)$  is the likelihood,  $\pi(\theta)$  the prior and  $q(\theta\,|\,d)$  is the neural approximation of the posterior.

