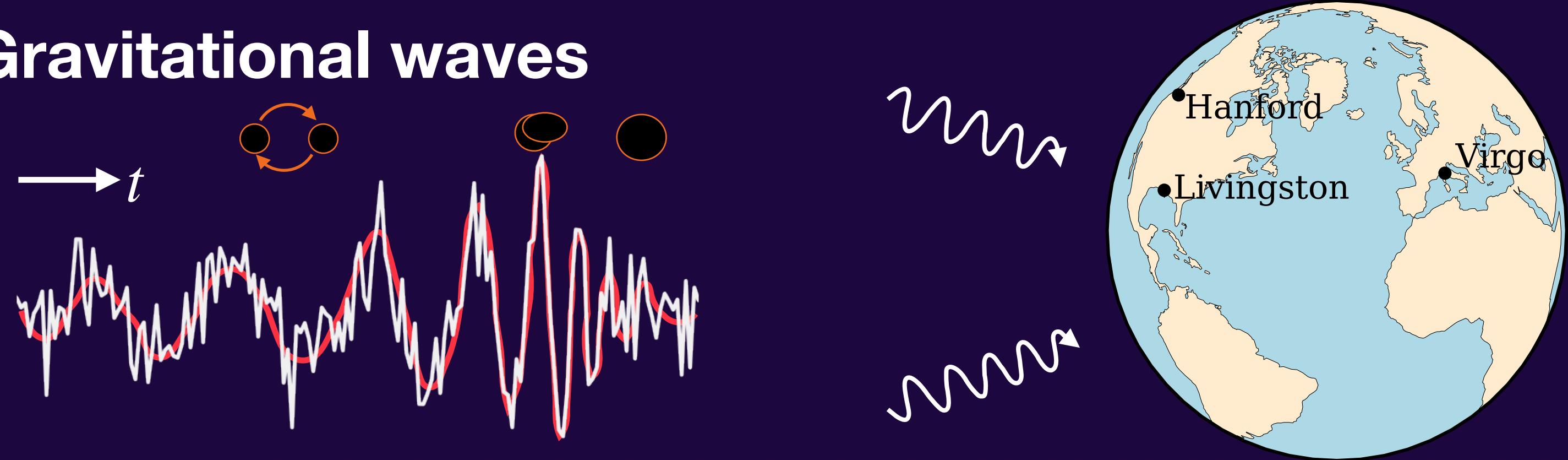


Flexible Gravitational-Wave Parameter Estimation with Transformers

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Gravitational waves



Goal: Analyze signals \rightarrow posterior distribution of black hole mergers

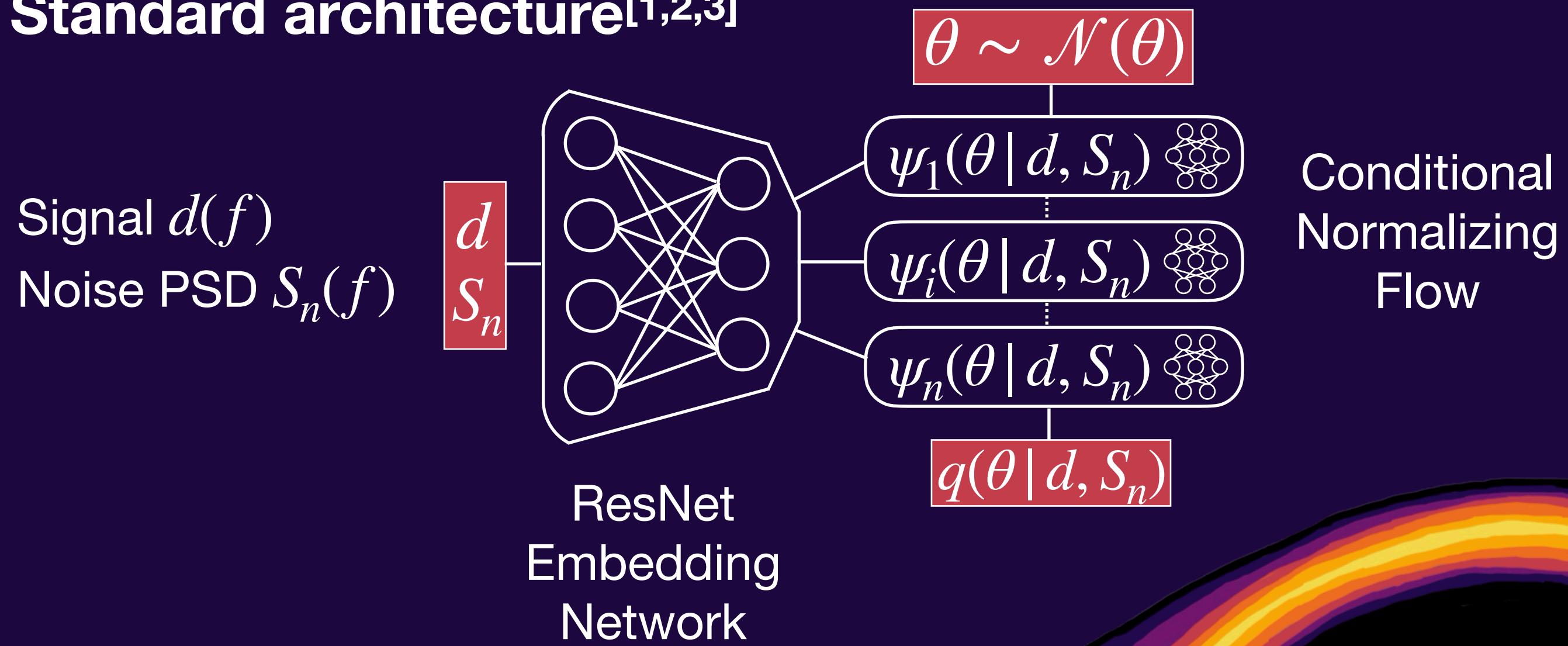
Problem: Real data is messy

\rightarrow Re-train model to adapt to different data analysis settings

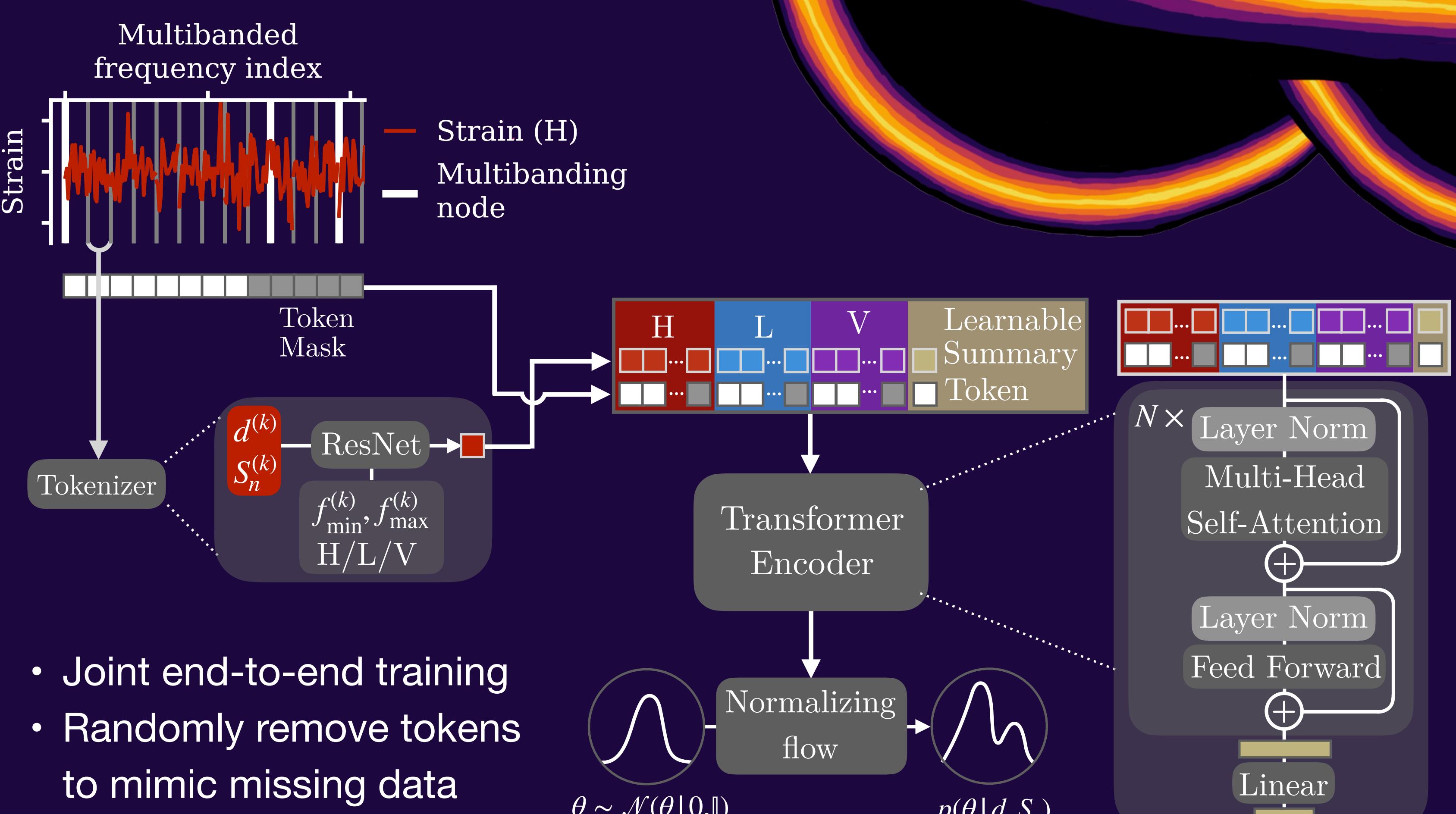
Solution: Flexible transformer architecture and masking procedure during training

DINGO (Deep INference for Gravitational wave Observations)

Standard architecture^[1,2,3]



DINGO-T1: Architecture

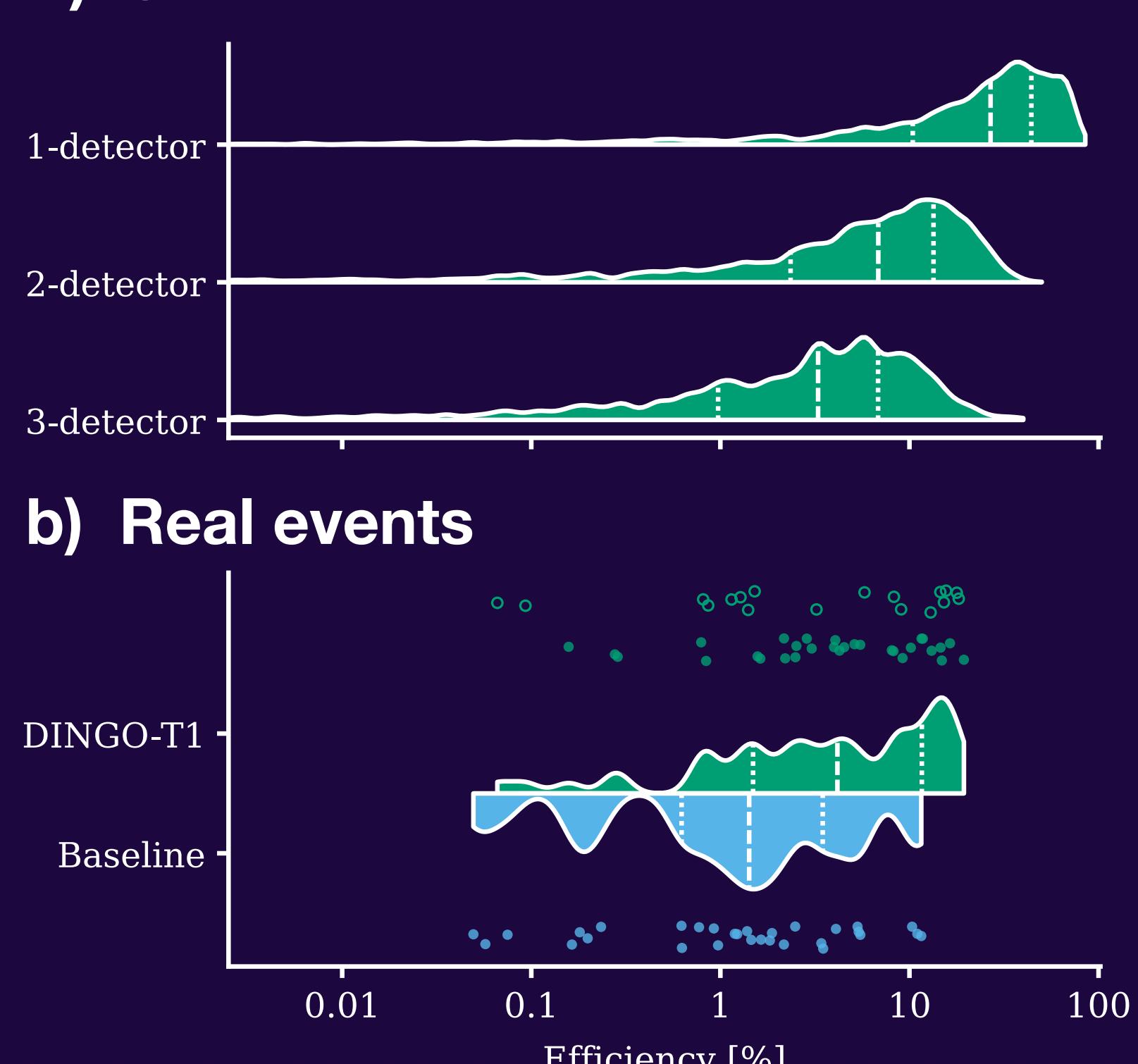


Masking strategies



Performance

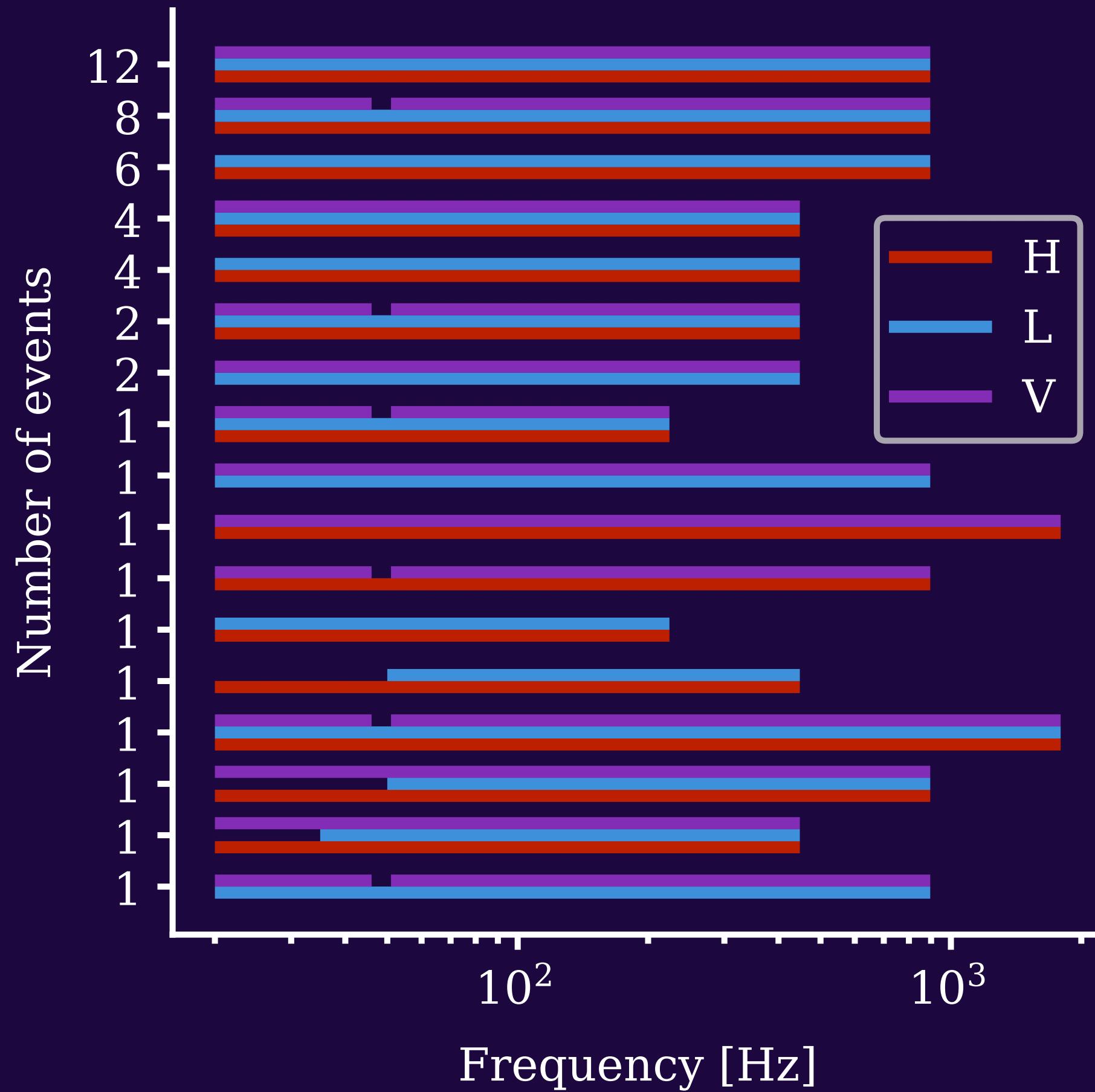
a) Simulated data



b) Real events

Real data is messy:

48 events with 17 different data analysis settings



- Missing detectors
- Changes in f_{min} & f_{max}
- Remove small range $[f_{low}, f_{high}]$

\rightarrow We would need to train 17 different DINGO models!

Validation with importance sampling^[3]

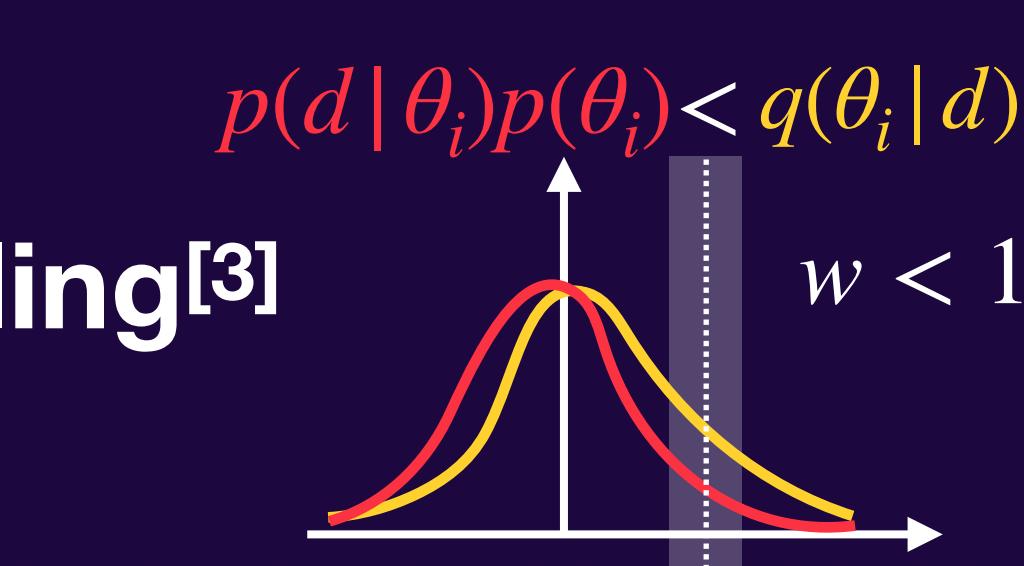
Compare learned NPE density and likelihood

$$\frac{p(\theta|d)}{q(\theta|d)} \propto w_i = \frac{p(d|\theta_i)p(\theta_i)}{q(\theta_i|d)}$$

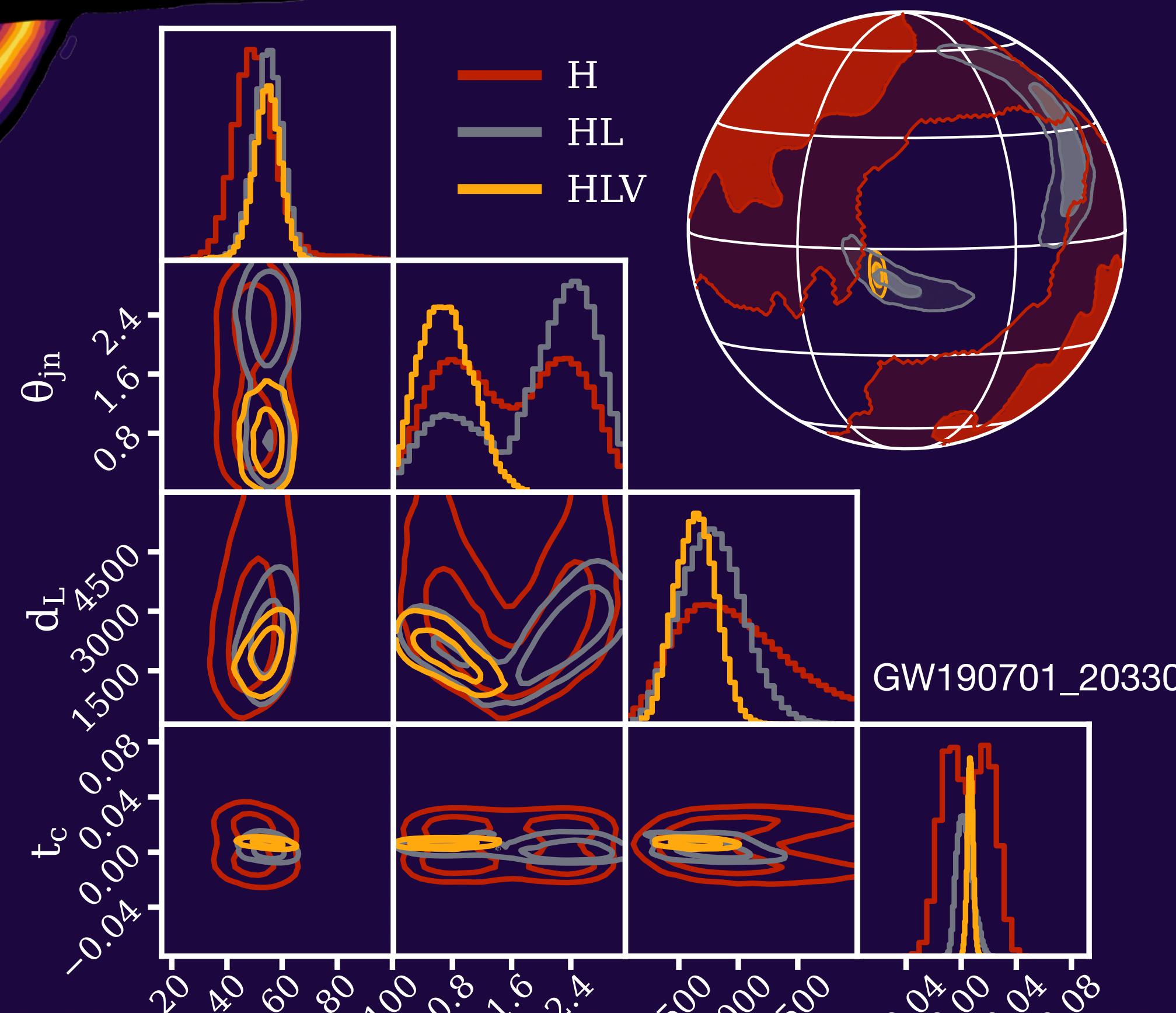
Likelihood · Prior
Proposal (NPE)

Performance criterion:
Sample efficiency

$$\epsilon = \frac{1}{N} \left(\sum_i w_i \right)^2$$

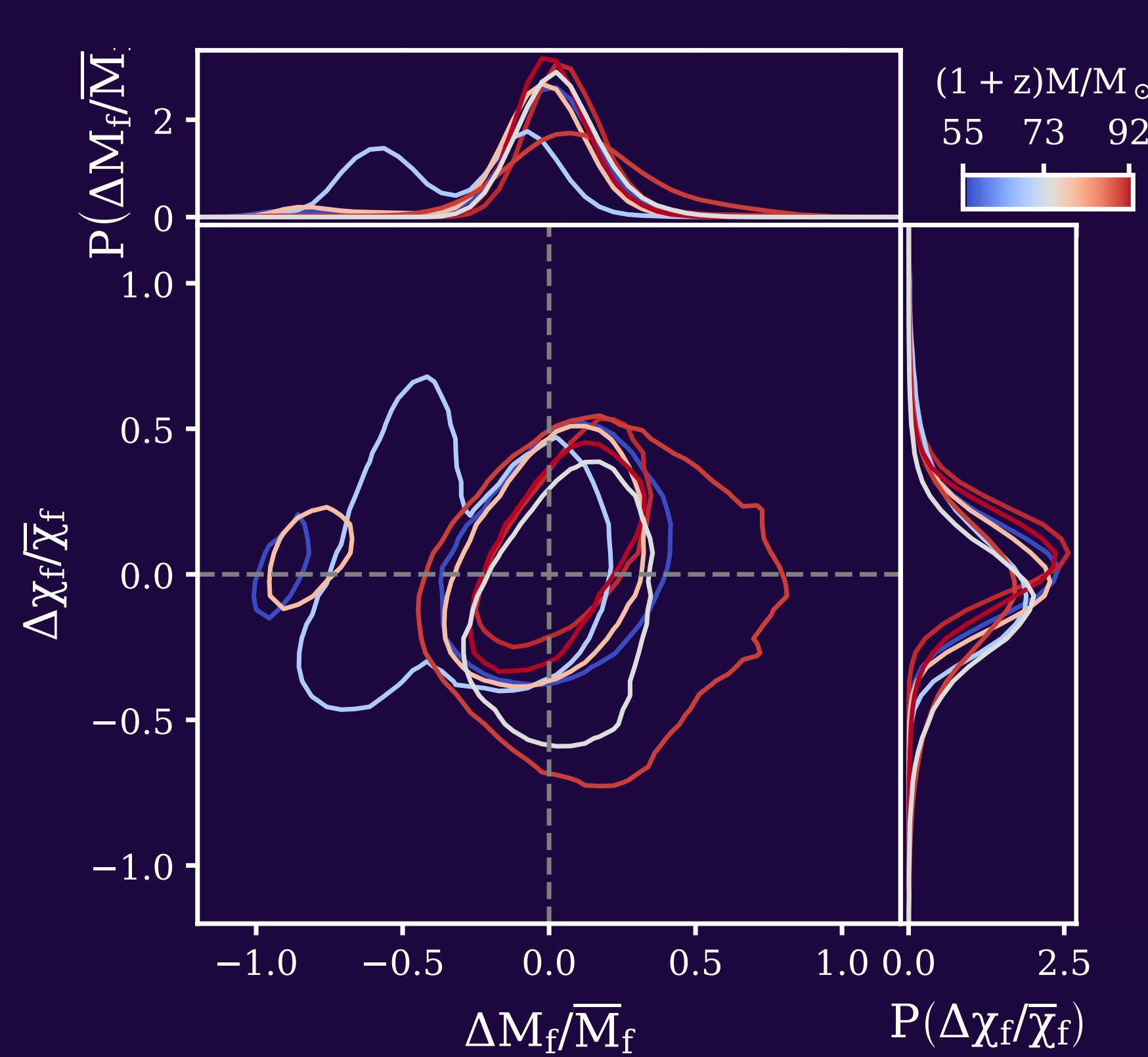


Flexible analysis



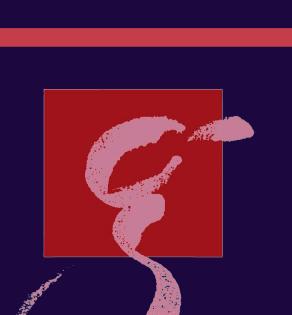
Tests of general relativity

- Analyze inspiral and postinspiral part of signal separately
- Check whether parameters agree



References

- [1] Dax+, Real-Time GW Science with NPE, PRL 2021
- [2] Dax+, Group Equivariant NPE, ICLR 2021
- [3] Dax+, Neural IS for Rapid and Reliable GW Inference, PRL 2023



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