

Accelerated Bayesian inference of multimessenger astronomy with jax and machine learning

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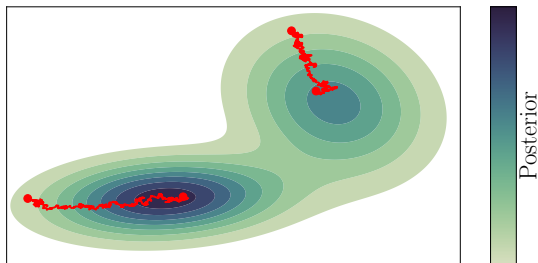


Bayesian inference

Bayesian inference: get **posterior** of parameters θ from data d

$$p(\theta|d) = \frac{p(d|\theta)p(\theta)}{p(d)}$$

Problem: Computationally expensive for multimessenger data + EOS



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 - Speed-of-sound extension scheme
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Next step: Infer EOS parameters directly from multimessenger data, upgrade software for future detectors

JAX = NUMPY + composable transformations:

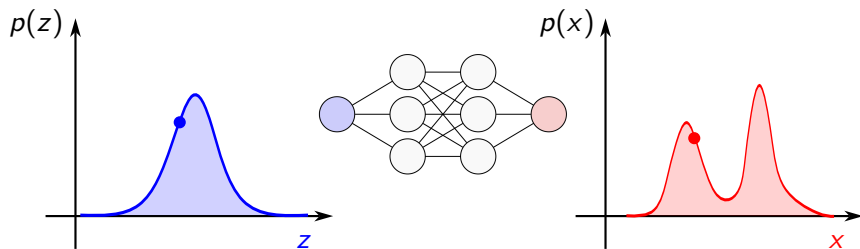
- ① Automatic differentiation
- ② Just-in-time (JIT) compilation
- ③ GPU acceleration
- ④ Parallelization



Potential for accelerating science, without resorting to machine learning!

Normalizing flows

- Generative machine learning model
- Learn mapping between **latent** and **parameter** space
- Enable approximate sampling from complicated distributions




Accelerating Bayesian inference for MMA: progress

- Gravitational waves (arXiv:2404.11397)
 - Waveforms: $\mathcal{O}(10^3)$ faster
 - Analyze binary neutron stars in 15 – 30 minutes rather than $\mathcal{O}(\text{hours})$


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- Future work:
 - Next-generation gravitational wave detectors
 - Combine into one multimessenger analysis pipeline

Thanks for listening!

Our JAX ecosystem:

- Normalizing flow-enhanced MCMC sampler:  kazewong/flowMC
- Gravitational waveforms:  tedwards2412/ripple
- Gravitational waves inference toolkit:  kazewong/jim
- EOS code and TOV solver:  tsunhopang/jose (*work in progress*)
- Kilonovae and GRB:  ThibeauWouters/fiesta (*work in progress*)