Any title, doesn't really matter

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This may be an abstract one day.

Introduction.—

- Bayesian Parameter Estimation.
- Generative Adversarial Networks.
- GWs are the shit.

Methods—

- How do you incorporate priors.
- Convergence.
- Volume of training data.
- Modifications from standard GAN (most of section).

Results—

GW150914. Show PE estimates on mass, spins, etc. Small section on waveform reconstruction. Plot of this as well.

 \bullet Which priors were used. Same as GW150914 analysis.

Conclusions—

Percision that we get, speed (sell this hard),

- Sell speed, with the right caveats.
- Waveform reconstruction.
- We are not model indepenent.
- Non-Gaussian noise.

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