



Non-parametric Bayesian methods for
cosmological parameter estimation

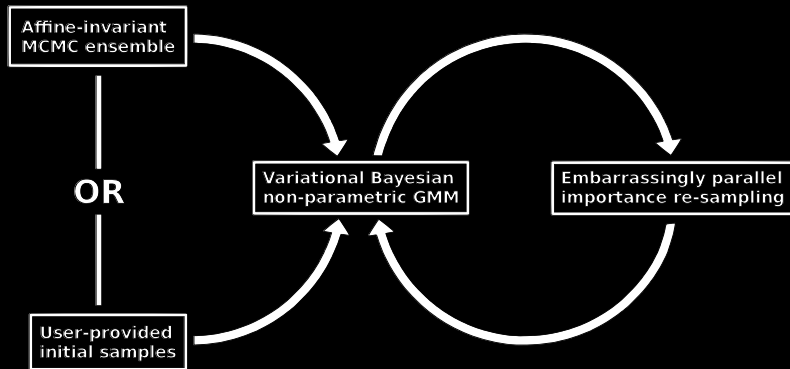
Ben Moews & Joe Zuntz
Institute for Astronomy (IfA)
School of Physics & Astronomy
The University of Edinburgh

Status quo of cosmological parameter estimation

- ▶ Modern problems deal with high-dimensional estimation
- ▶ Computing posterior values is computationally expensive
- ▶ MCMC algorithms, even using multiple walkers, are slow
- ▶ Distributions often can't be ballparked in a reasonable way
- ▶ Making this an embarrassingly parallel problem would help

Non-parametric Bayesian solutions to this problem

- ▶ Cut down the initial parallelization-limited sampling step
- ▶ Parallelize a repeated Bayesian fitting and re-sampling step



Simple, fast and robust high-dimensional CPE

Callable with one function, only 3 *required* inputs:

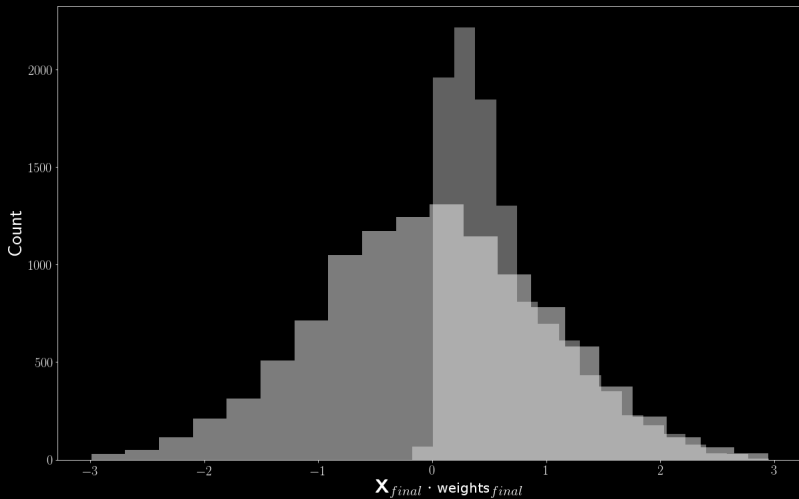
- ▶ The set of ranges for the parameters
- ▶ The function that returns a posterior
- ▶ The number of samples in the output

Simple, fast and robust high-dimensional CPE

- ▶ Building on related PMC approaches, e.g. CosmoPMC
- ▶ Dynamically shrinking convergence threshold for fitting
- ▶ Robust against potentially dominating high-posterior points
- ▶ Use of kernel density estimation for low-dimensional problems
- ▶ Optional return of the model and its importance weights
- ▶ Built-in option for multi-process parallelization and MPI

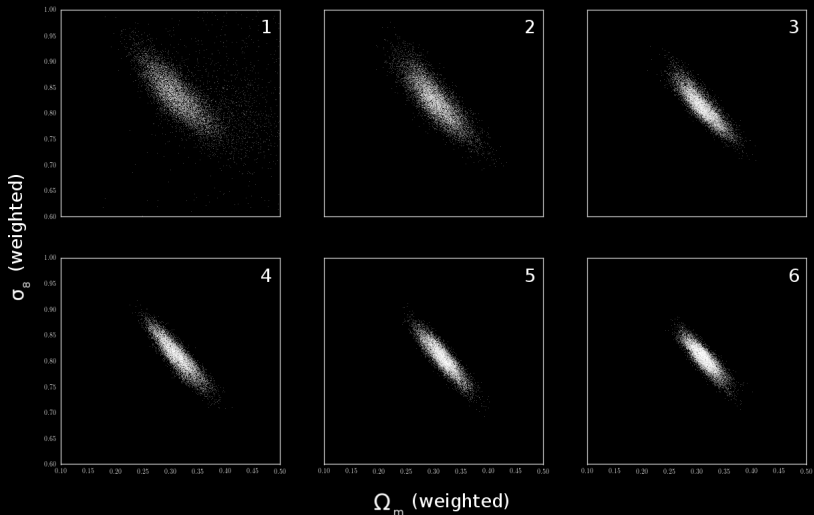
Testing the functionality on a toy example

Distribution recovery



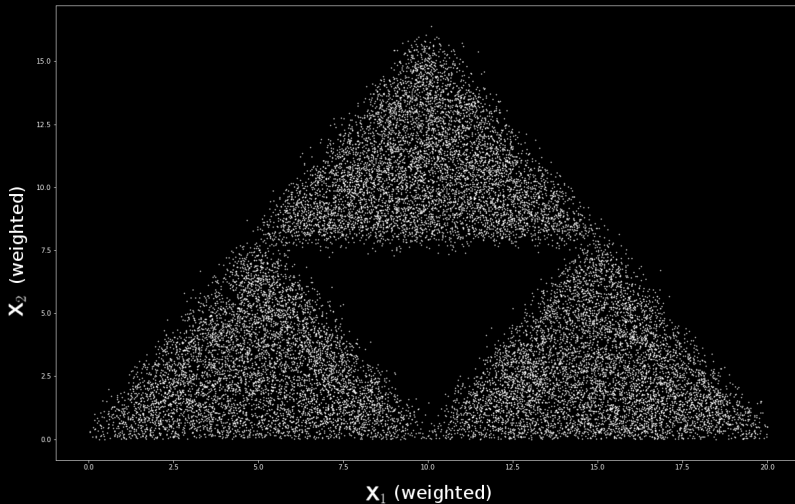
Testing gaussbock on an actual DES SV example

Model-drawn samples over time



Testing approximations of uncommon posteriors

Unusual posterior distributions



Spread the word by testing and using it yourself

It's on PyPI, and easy to install with a single console command!

```
pip install gaussbock
```

(The paper will appear on arXiv shortly.)

Questions?



gaussbock

Contact: ben.moews@ed.ac.uk