

Tasks to be completed for project:

April 25 (Sat): Uploaded bare-bones code to repo; generalised from Bayesian Inference assignment to have multiple parameters and arbitrary posteriors/priors - Kyle

May 4 (Mon): presentation review (all) - to be on the earlier (safer) side

May 5 and 6 (Tue-Wed): buffer days

May 7 (Thu): in-class presentations (all)

May 8 (Fri): delivery of code and presentation slides

Use basic algorithm to run different tests over and over; all “physics functions” and priors, etc as separate scripts with functions that can be imported

Steps for Task 1:

1. Have basic posterior sampling, chain, and convergence test running to use to test everything else - **Kyle; aim for end of Sun April 26**
2. Globally import data table from file - **Pratik**
3. Def mu_model (for 3 different cases of OmegaK) - **Pratik/Sudheer; try to have it done after OH on Monday, April 27**
 - a. OmegaK is based on OmegaM and OmegaLambda
 - b. Analytically check where degeneracy appears (discuss during office hours)?
4. Define a function for the priors on all 4 parameters (OmegaM, OmegaLambda, H0, and M) such that M can have a uniform or a gaussian prior - **Avinash; try to have it done by Sun April 26**
 - a. Uniform for all
5. Define likelihood functions - **Avinash; try to have it done by Sun April 26**
 - a. 1likelihood function - with and without systematic errors (systematic errors: using covariance matrix from additional data file; no systematic error: just diagonal terms using error from the main dataset)
6. Define posterior function for the chain to sample from - **Kyle**
 - a. Based on likelihood and priors
7. Create and propagate chain using MH and sampling from posteriors - **Kyle**
 - a. Gaussian generating function - independent for each parameter
8. Define a convergence test for the MCMC - **Kyle**
9. Marginalise (integrate numerically) over H0 and M to get posterior only in terms of OmegaM and OmegaLambda - **Sudheer**
10. Plot posterior for OmegaM and OmegaLambda - **Sudheer; aim for Friday May 1**
 - a. Include both analyses (with and without errors) in the same plot

Steps for Task 2:

11. Repeat same analysis as Task 1 except now marginalise over Ω_M , Ω_Λ , and M - **Pratik**

```
def mu_model(param, data)
    params = array of  $\Omega_M$ ,  $\Omega_\Lambda$ ,  $H_0$ ,  $M$ 
    data = array of  $m$ ,  $z$ ,  $merr$  for each SN
    returns mu_model for each SN (array of length 40)
```