

IFT Summer school: Statistics exercises

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Background Theory

The normalised Friedmann equation of the universe can be written as:

$$\left(\frac{H}{H_0}\right)^2 = \Omega_{r,0} a^{-4} + \Omega_{m,0} a^{-3} + \Omega_{k,0} a^{-2} + \Omega_{\Lambda,0}, \quad a = \frac{1}{1+z} \quad (1)$$

where:

- a is the time-dependent scale-factor of the universe, normalised to 1 today.
- z is the observed redshift of an object located at a previous epoch.
- $H = \frac{d}{dt} \log a$ is the Hubble parameter, with present value H_0 .
- $\Omega_i = 8\pi G\rho_i/3H^2$ is the density parameter for component i , with present value $\Omega_{i,0}$.

From these definitions, you should note that $\Omega_{i,0} \sim 1$, and that by setting $a = 1$ (today), $1 = \Omega_{r,0} + \Omega_{m,0} + \Omega_{k,0} + \Omega_{\Lambda,0}$.

With this, along with a little geometry, we may compute several cosmological distance measures:

$$d_L = \frac{d_M}{1+z} \quad (\text{Luminosity dist.}) \quad (2)$$

$$d_M = \begin{cases} \frac{d_H}{\sqrt{\Omega_{k,0}}} \sinh \frac{d_C \sqrt{\Omega_{k,0}}}{d_H} & \Omega_{k,0} > 0 \\ d_C & \Omega_{k,0} = 0 \\ \frac{d_H}{\sqrt{|\Omega_{k,0}|}} \sin \frac{d_C \sqrt{|\Omega_{k,0}|}}{d_H} & \Omega_{k,0} < 0 \end{cases} \quad (\text{Transverse comoving dist.}) \quad (3)$$

$$d_C = d_H \int_0^z \frac{dz}{\sqrt{\Omega_{r,0} a^{-4} + \Omega_{m,0} a^{-3} + \Omega_{\Lambda,0}}} \quad (\text{Comoving dist.}) \quad (4)$$

$$d_H = \frac{c}{H_0} \quad (\text{Hubble dist.}) \quad (5)$$

For specific kinds of supernovae/astronomical objects, we can directly measure the magnitude $\mu = m - M$, which is related to their luminosity distance d_L :

$$\mu = 5(\log_{10} d_L - 1) \quad (\text{obviously...}) \quad (6)$$

We measure the magnitudes μ and redshifts z of N supernovae, and combine them into two data vectors $y = (\mu_1, \dots, \mu_N)$, $x = (z_1, \dots, z_N)$. These magnitudes also come with an associated error, or more precisely a covariance matrix Σ .¹ Given a specific cosmology defined by the parameters $\theta = (H_0, \Omega_{r,0}, \Omega_{m,0}, \Omega_{\Lambda,0}, \Omega_{k,0})$, we can compute the luminosity distance (2) as a function of redshift, and thus the theoretical magnitude of the object $\hat{\mu}(z; \theta)$ for any given z , and thus the theoretical data vector $\hat{y}(\theta) = (\hat{\mu}(z_1; \theta), \dots, \hat{\mu}(z_N; \theta))$. The likelihood of observing this data is then:

$$\mathcal{L}(\theta) = P(y, x|\theta) = \frac{1}{\sqrt{\det 2\pi\Sigma}} \exp\left(-\frac{1}{2}[y - \hat{y}(\theta)]^T \Sigma^{-1} [y - \hat{y}(\theta)]\right) \quad (7)$$

This object forms the centre of our inference. The data are taken from: <http://supernova.lbl.gov/Union/> which you should visit to see some example figures.

1 Set up

Copy the work directory into your local user area, and source the relevant files:

```
ssh -X <username>@hydra.ift.uam-csic.es
cp -r /home/prof4/PolyChord ~/PolyChord
cd ~/PolyChord
source modules
```

Alternative: Local install for unix-like systems (MAC, linux)

```
wget https://www.mrao.cam.ac.uk/~wh260/PolyChord.tar.gz
tar -xvf PolyChord.tar.gz
cd ~/PolyChord
make veryclean
make PyPolyChord
export LD_LIBRARY_PATH=$LD_LIBRARY_PATH:$PWD/lib
export LD_PRELOAD=/usr/lib/openmpi/libmpi.so
```

(NB: You will need to substitute the location of your openmpi library into LD_PRELOAD)

You will also need the python modules (both pip-installable):

¹These statements are grossly misleading. Much of the research/controversy in this field goes into correctly modelling the errors in the Hubble distance ladder

- getdist <https://pypi.python.org/pypi/GetDist/>
- MPI4py <https://pypi.python.org/pypi/GetDist/>

as well as a gfortran compiler + mpi. Alternatively, you can run `make veryclean`; make PyPolyChord MPI=, if you don't want the faff of setting up MPI.

2 Plotting the data (5-10 mins)

1. Run the script which produces a plot of the supernovae:

```
python plot_sn.py
gedit_bg plot_sn.py &
```

(`gedit_bg` is an alias that was loaded when you sourced “modules”, which just pipes the `stderr` to `/dev/null`. `gedit` is a simple gui text editor. Feel free to substitute your own preferred text editor.²)

2. Modify the script so that instead of plotting magnitudes, you plot luminosity distances d_L against redshift z (Hint: consider equation (6), and make sure you calculate the errors correctly).
3. Question: Why can't we turn this plot into something more theoretically intuitive, such as scale-factor against cosmic time?

3 Metropolis Hastings (20-30 mins)

I have written up the Gaussian likelihood for you (if you are interested, it's all in `SNE/supernova_data.py`). For now, we will assume that the universe is flat ($\Omega_{k,0} = 0$), with only matter and dark energy

4. Write a two-dimensional Metropolis-Hastings algorithm to explore the likelihood for flat universe with only dark energy and matter in it. You should choose your parameters to be H_0 and $\Omega_{m,0}$, with the value of dark energy therefore being $\Omega_{\Lambda,0} = 1 - \Omega_{m,0}$. You should start by modifying the file `MH.py` (If you haven't managed to debug your code after 20 minutes or so, you can copy the solution in `answers/MH_1.py`)
5. Question: Comment on the properties of the chain produced
6. How do the properties of the chain change if you choose a more sensible starting point? or a different step size? Even if you tune these correctly, do you notice anything funny about the properties of your chains?
7. Plot your chains using `getdist`:

²In my opinion, any text editor is fine, so long as its vim or emacs.

```

O_L_array = 1. - np.array(O_m_array)
samples = np.array([H0_array, O_m_array, O_L_array]).T
weights = np.array(count_array)
names = ['H0', 'O_m', r'O_L*']
labels = ['H_0', r'\Omega_m', r'\Omega_\Lambda']
samples = getdist.MCSamples(samples=samples,
                             weights=weights,
                             names=names,
                             labels=labels)
g = getdist.plots.getSubplotPlotter()
g.triangle_plot(samples, filled=True)
plt.show()

```

8. Is there anything different if you use the likelihood that doesn't take into account systematic errors? (change `loglikelihood_sys` for `loglikelihood_nosys`)

4 PolyChord (20 mins)

9. Start by using `polychord` to plot what you just worked on:

```
mpirun -np 1 python run_PyPolyChord.py
```

You can change `-np 1` to a higher number to increase the number of MPI cores which `PolyChord` runs on. What is the fundamental difference between the likelihood which includes systematic errors, and the one that does not? Have a look at the script to see how `PolyChord` is set up. Most of the code is purely interfacing our code with `PolyChord`'s input, but some of the settings can be important.

10. Now for the punch-line. Run:

```
mpirun -np 1 python run_all.py
```

This will run three models:

- (a) matter + dark energy (flat)
- (b) matter + dark energy + curvature
- (c) matter + curvature (no dark energy).

After it's done, what can you say about both the evidences of each model, and the comparison of the posteriors with & without curvature?

11. Modify the above script by adding in/removing components of the universe (radiation etc).
12. Plot the predictive posterior distribution using:

```
python compute_contours.py
python plot.py
```

This plots the predictive posterior distribution for the flat matter dark energy universe. Modify `compute_contours.py` to do it for the curved matter dark energy universe.

5 Bonus Questions/extended investigations

To be attempted in any order

13. Modify code to allow for a dark energy with a variable equation of state parameter w (i.e. change $\Omega_{\Lambda,0} \rightarrow \Omega_{\Lambda,0} a^{-3(1+w)}$).
14. Modify your MH algorithm to include convergence diagnostics, such as split-Rhat. A good reference can be found in the stan manual: <https://github.com/stan-dev/stan/releases/download/v2.14.0/stan-reference-2.14.0.pdf> (28.3. Initialization and Convergence Monitoring)
15. How is run-time and evidence accuracy affected by the PolyChord setting `nlive`?
16. How sensitive are your conclusions to the prior widths (qualitatively and quantitatively)?