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Fakultät für Physik  
Statistical Methods - An Introduction  
Problem Sheet 4 (SS12)

To be submitted at May, 14th, 2012 to jochen.weller@usm.lmu.de.

## Type Ia Supernovae Magnitude Statistics

Type Ia Supernovae are among the brightest objects in the Universe. They are all almost the same brightness, with differences that can be standardized to less than 10 percent. The accelerating cosmic expansion first inferred from observations of these objects (Riess et al. 1998; Perlmutter et al. 1999) indicates unexpected gravitational physics, frequently attributed to the dominating presence of a *dark energy* with negative pressure.

We will now use a SN Ia data sample compiled by Riess et al. 2004 to estimate the matter content of the Universe.

After downloading the supernovae dataset `sn_data_riess.dat` on lecture webpage, we need to:

- a) Read data points using the `scan` command: we need columns 2,3,4 which correspond to redshift  $z$ , magnitude  $m$  and error on magnitude  $\sigma$  (be careful to skip the header). Define a table called `data` with these three columns: it will be useful later.
- b) Calculate the theoretical magnitude, assuming a flat Universe with  $H_0 = 72.0 [kms^{-1}Mpc^{-1}]$ ,  $c = 300000 [kms^{-1}]$ . The luminosity distance  $d_L$  which is given by

$$d_L = c(1+z) \int_0^z \frac{dz}{H(z)},$$

is implemented in the `dL.R` code, which can be downloaded from lecture webpage.

Define then the magnitude  $m$  as a function of  $z, \Omega_m, M$ , remembering its definition

$$m = M + 5 \log_{10}(H_0 \cdot d_L) .$$

[Note that this is obtained from

$$m = M_{int} + 5 \log_{10}(d_L) + 25 ,$$

where

$$M = M_{int} - 5\log_{10}(H_0) + 25$$

and  $M_{int}$  is the intrinsic magnitude.]

- c) Define  $\chi^2$  as a function of  $\Omega_m$  and  $M$ , according to its definition

$$\chi^2 = \sum_{i=1}^N \frac{(m_i - m_{th}(z_i))^2}{\sigma_i^2},$$

where  $N$  is the number of data points,  $m_i$  is magnitude data,  $m_{th}$  is theoretical magnitude,  $\sigma_i$  is the error on data.

- d) Find the minimum of  $\chi^2$  (hint: avoid calculating the derivative and use `for` loops instead) and print out also the minimum values for the parameters  $\Omega_m$  and  $M$ . Let vary  $\Omega_m$  in  $[0.0, 1.0]$  with step 0.05 and  $M$  in  $[15.5, 16.5]$  with step 0.01.
- e) Calculate the posterior probability as a function of  $\Omega_m$  and  $M$ , according to its definition

$$p = e^{-\frac{1}{2}(\chi^2 - \chi_{min}^2)},$$

where  $\chi_{min}^2$  is the value calculated in the previous step.

- f) Plot  $\Omega_m$ - $M$  contour plot (hint: see Lecture3b).
- g) Marginalize over  $M$  in the range  $[13.0, 18.0]$  (hint: see Lecture3c).
- h) Plot the 1-d marginalized likelihood as a function of  $\Omega_m$ .

10 POINTS