

## Fall 2016 Astro 250: Stellar Populations

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### Problem Set 2 – Intro to Probability and Stats

Assigned: 9/12/16

Due: 9/26/16

All problem sets should be completed in your public github repository. Problems that require written solutions should be completed using LaTeX. Coding exercises should be done in a jupyter notebook. For clarity, you may want to put each coding problem in a separate notebook.

#### Problem 1.

Consider a single power-law IMF of the form:

$$P(M|\theta) = c M^{-\alpha} \quad (1)$$

where

$$c = \frac{1}{\int_{M_{min}}^{M_{max}} M^{-\alpha} dM} \quad (2)$$

and  $\theta = (M_{min}, M_{max}, \alpha)$ .

For simplicity, assume perfect knowledge of the masses and that observational effects are negligible.

(a) Write code that generates a list of  $N$  stellar masses between a given  $M_{min}$  and  $M_{max}$  from a power-law distribution with an index of  $\alpha$ .

(b) Write code that will perform inference on the set of fake data you generated in part (a) using `emcee`.

(c) Generate a fake dataset assuming  $M_{min} = 3 M_{\odot}$ ,  $M_{max} = 15 M_{\odot}$ ,  $N = 1000$  and  $\alpha = 1.35$ . In your inference code, let  $\alpha$  and  $M_{max}$  be free parameters (but fix  $M_{min} = 3 M_{\odot}$ ). Given this fake dataset, to what precision can you constrain  $\alpha$  and  $M_{max}$ ?

(d) Show how the precision to which  $\alpha$  and  $M_{max}$  can be recovered depends on  $N$ , from  $N \sim 10$  to  $N \sim 10,000$ . Summarize your results in plots. It is OK to discretely and uniformly select values of  $N$  in  $\log_{10}$  space (e.g.,  $\log_{10}(N) = 1, 2, 3, 4$ ). *Hint: In the limit that  $N$  is a small number, you may want to generate multiple datasets to verify the fidelity of your confidence intervals, as stochastic effects can be important.*

### Problem 2.

In this problem, we will attempt to re-create Salpeter’s original IMF measurement.

(a) Using the data for mass and number density in Table 2 and/or Figure 2 in Salpeter (1955), fit a power-law using an optimizer or least squares fitter (e.g., `scipy.optimize`).

(b) Same as part (a) only using your own inference code and `emcee`. Compare the two results: How close are they to one another? How close are they to the value reported in Salpeter (1955)?

### Problem 3.

There are claims in the literature that the low-mass IMF slope may systematically deviate from the Galactic value in low-mass dwarf galaxies (e.g., Wyse et al. 2002; Geha et al. 2013). However, these measurements are made over a fairly limited mass range (usually  $\sim 0.5 - 0.8 M_\odot$ ) and done so assuming that a power-law is a reasonable approximation for the low-mass IMF.

Suppose the true IMF for stars with  $M < 1 M_\odot$  in all galaxies is actually a Chabrier IMF, i.e., a log-normal at low-masses. Ignoring corrections for stellar multiplicity, this IMF has the functional form:

$$\xi(m)\Delta m = \frac{0.15}{m} \exp \frac{-(\log(m) - \log(0.08))^2}{(2 \times 0.69)^2} \quad (3)$$

(a) Using the Chabrier IMF from above, generate a list of  $N=10,000$  (perfectly known) stellar masses between  $0.5$  and  $0.8 M_\odot$ .

(b) Now, assuming a single-slope power-law IMF model (as done in the literature), infer the value of the spectral index  $\alpha$ . How does this compare with the canonical Kroupa IMF found in the Milky Way?

### Problem 4.

Using python-FSPS:

(1) Generate the spectrum for a 10 Myr simple stellar population (assume no dust, fixed metallicity, etc – the only variable of interest is age). Plot how the spectrum from  $1500 - 10000 \text{\AA}$  changes for three different high-mass IMF ( $> 1 M_\odot$ ) values:  $\alpha = 0.8, 1.3, 1.8$ , holding the lower portions of the IMF fixed.

(2) Generate the spectrum of a 10 Gyr simple stellar population. Plot how the spectrum from  $5000 - 20000 \text{\AA}$  changes for three different IMF forms: Salpeter IMF, a Kroupa IMF, and the van Dokkum IMF.