SN Cosmology with BEAMS

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LSST SN Workshop - Pitt November 2016

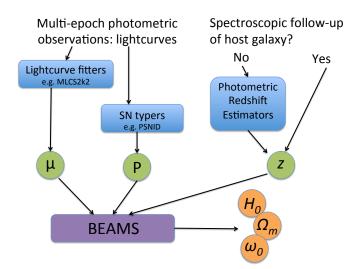


Motivation

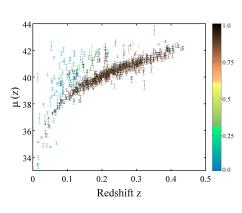
- doing cosmology with SNe, without spectroscopy
- incorporate type probably
- use the full (contaminated) sample, without making cuts
- error ellipses on cosmological parameters



What does BEAMS do?



Photometric Sample from SDSS-II SN Survey



Hlozek et al. 2012

- z from host galaxy spectroscopic redshift
- μ from MLCS2k2 (Jha et al. 2007)
- color coded using P_{Ia} from the PSNID typer (Sako et al. 2008, 2011)
- we want to fit for two populations: SNIa and non-SNIa



Maths:

$$P(\boldsymbol{\theta}|\boldsymbol{d}) \propto P(\boldsymbol{\theta}) \times \prod_{i=1}^{N} \left\{ P(d_i|\boldsymbol{\theta}, \text{SN Ia}) P_{\text{Ia},i} + P(d_i|\boldsymbol{\theta}, \text{not SN Ia}) (1 - P_{\text{Ia},i}) \right\}$$

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Words:

Posterior probability for total population \propto prior on cosmological parameters \times $\prod_{i=1}^{N}$ (likelihood for Ia \times probability that it is a SN Ia + likelihood for not Ia \times probability that it is not a SN Ia)

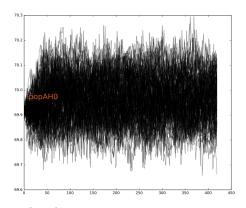


$$P(d_i|\boldsymbol{\theta}, \text{SN Ia}): \boldsymbol{\theta} = H_0, \Omega_m, \Omega_{\Lambda}$$

 $P(d_i|\boldsymbol{\theta}, \text{SN Ia})$: $\boldsymbol{\theta} = H_0, \Omega_m, \Omega_\Lambda$ $P(d_i|\boldsymbol{\theta}, \text{not SN Ia})$: fitting several sub-populations... do you need a single mean, binned means?



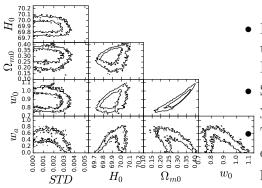
Maximizing the Posterior Probability



MCMC walkers.

- BEAMS is implemented using MCMC. (*emcee* Python package.)
- See Renee's talk yesterday.

Maximizing the Posterior Probability



Covariance of cosmological parameters.

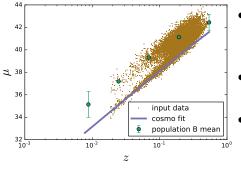
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This gives full error ellipses on all the fit parameters.



Maximizing the Posterior Probability



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Fits to mock SNe data.



• Incorporating full photo-z PDFs.



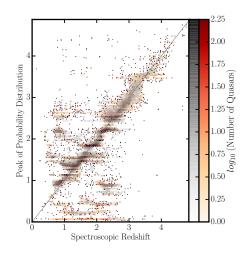
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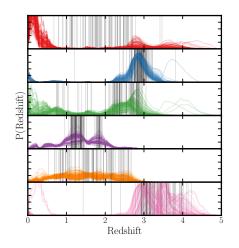
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- Comparison of lightcurve fitting algorithms, how does that effect the output cosmology?

More about Photo-z PDFs (of quasars...)

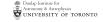


• Photometric redshifts can be drastically different from spectroscopic redshifts.

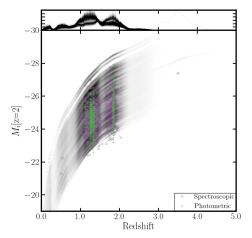
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- Implementing this is practice is difficult, but has been done for quasar luminosity functions.

 (Kelly et al. 2008)
- I have plans to do this for BEAMS.

