Constraining Dark Energy: First Results from the SDSS-II Supernova Survey J. Craig Wheeler Department of Astronomy University of Texas at Austin (adapted from presentation by Josh Frieman) Texas Cosmology Network Meeting October 29, 2009

1998-2009 SN Ia Synopsis

- Substantial increases in both quantity and quality of SN Ia data: from several tens of relatively poorly sampled light curves to many hundreds of well-sampled, multiband light curves from rolling surveys
- Extension to previously unexplored redshift ranges: z>1 and 0.1<z<0.3
- Extension to previously underexplored rest-frame wavelengths (NIR)
- Vast increase in spectroscopic data
- Identification of SN Ia subpopulations (host galaxies)
- Entered the systematic error-dominated regime, but with pathways to reduce sys. errors

ESSENCE

Wood-Vasey, etal Miknaitis, etal SNLS Astier, etal

But redshift desert remains



SDSS II Supernova Survey Goals

- Obtain few hundred *high-quality** SNe Ia light curves in the `redshift desert' *z*~0.05-0.4 for continuous Hubble diagram
- Probe Dark Energy in *z* regime complementary to other surveys
- Well-observed sample to anchor Hubble diagram, train light-curve fitters, and explore systematics of SN Ia distances
- Rolling search: determine SN/SF rates/properties vs. *z*, environment
- Rest-frame *u*-band templates for z > 1 surveys
- Large survey volume: rare & peculiar SNe, probe outliers of population

SDSS-II SN Survey Team

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Use the SDSS 2.5m telescope

- September 1 November 30 of 2005-2007
- Scan 300 square degrees every 2 days
- Obtain densely sampled multi-color light curves
- Results today from 2005 season

Kessler, et al 09; Lampeitl et al 09; Sollerman et al 09



Searching For Supernovae

Difference

Search







2005 •



- 3,694 candidates
- 193 confirmed Ia
- 2007 •

•

175 confirmed Ia

•Positional match to remove movers •Insert fake SNe to monitor efficiency



SDSS SN Photometry

Holtzman etal (2008)







Spectroscopic Target Selection





Spectroscopic Target Selection



SN and Host Spectroscopy



MDM 2.4m NOT 2.6m APO 3.5m NTT 3.6m KPNO 4m WHT 4.2m Subaru 8.2m **HET 9.2m** Keck 10m Magellan 6m TNG 3.5m SALT 10m



Spectroscopic Deconstruction

2005kg SNIa z=0.3904 SN Phase=+9d

2005fr SNIa z=0.286 SN Phase=+7d



Zheng, et al (2008)

Extract R_V by matching colors of SDSS SNe to MLCS simulations

$$\left\langle R_{V}\right\rangle = \frac{A_{V}}{E(B-V)} \approx 2$$

- MLCS previously used Milky Way avg R_V=3.1
- Lower R_V more consistent with SALT color law



D. Cinabro



Determine priors and efficiencies from data and Monte Carlo simulations

Monte Carlo Simulations match data distributions



Use recorded observing conditions (local sky, zero-points, PSF, etc)



Hubble Diagram



Hubble Diagram with SDSS SNe

103 SNe Ia from first season



Kessler etal (2009) Lampeitl etal (2009) Sollerman etal (2009)

A Tale of Two Fitters

MLCS (Multicolor Light Curve Shape; MLCS2k2, Jha, 2002):

- U-band model trained only on low-redshift (observer-frame) U-band data (calibration, atmospheric variations)
- Assumes *all* excess color due to dust extinction (some of it must be); dust prior dominates at high-redshift

SALT (Spectral Adaptive Lightcurve Template, Guy et al 2005):

- Global fit for color/dust parameter β : minimizing Hubble-scatter can lead to bias
- Trend toward bluer colors at high-*z*: if allow $\beta(z)$, see strong trend with redshift

•Retrain and refine the models with newer data

Correct Distance Estimates for Brightness-decline relation and dust extinction

> MLCS2k2 Light-curve templates Jha, etal (2007)

time-dependent model "vectors" trained on Low-z SNe



$$\vec{m_X}(t - t_0) = \vec{M_X^0} + \mu_0 + \vec{\zeta_X} (\alpha_X + \beta_X / R_V) A_V^0 + \vec{P_X} \Delta + \vec{Q_X} \Delta^2,$$

(plus K-corrections)
ime of maximum distance modulus dust law extinction stretch/decline rate

SALT-II Light-curve Fits

• Fit each light curve using rest-frame *spectral* surfaces:

$$\frac{dF_{\text{rest}}}{d\lambda}(t,\lambda) = x_0 \times [M_0(t,\lambda) + x_1 \times M_1(t,\lambda)] \times \exp[c \times CL(\lambda)]$$

light-curve shape color term

• Light curves fit individually, but distances only estimated globally:

$$\mu_i = m^*_{Bi} - M + lpha \cdot x_{1,i} - eta \cdot c_i$$

Global fit parameters, determined along with cosmological parameters

•Differences from MLCS: not trained just on low-redshift data; flat priors on model parameters, espec. color; color variations not assumed to come only from dust

SDSS+Nearby SNe Only

SALT-II



SDSS+Nearby+SNLS+ESSENCE+HST



SALT2 distance moduli for SNLS SNe systematically higher than MLCS ₂₄

Systematic Errors are Dominant

SN+BAO+CMB constraints

MLCS vs. SALT discrepancy is NOT associated with SDSS SNe



SALT vs MLCS template light curves

Diagnosis:

Large difference in Light-curve model in U-band

Use of prior on extinction in MLCS



Systematic Errors (and Controls)

- Dust and SN color variation (multi- λ , NIR, high S/N)
- Selection effects (artificial SNe, Monte Carlo simulations)
- Population evolution (SN properties vs host environment)
- Photometric calibration (system calibration (lasers, etc) & cross-calibration of systems)
- Sample purity (spectroscopy)
- All (subdivide large samples to cross-check)
- Clear pathway to progress

The Future is (Mostly) Photometric

- Pan-STARRS, Dark Energy Survey, LSST: thousands to hundreds of thousands of SN light curves to $z\sim 1$
- Limited spectroscopic resources for follow-up, replace spectroscopic with photometric SN classification
- Very large samples will enable subdivision to study correlations and control systematics
- Photometric-redshift precision at high-*z* does not critically impact DE constraints but degrades (SN-type) sample purity
- SN spectroscopic subsamples (both Ia and non-Ia) required to quantify purity and define SN Ia color selection. SDSS SN test of photometric classifier: ~92% complete, 5% impurity (M. Sako)

Conclusions

• Robust evidence for cosmic acceleration from Supernovae and other probes

•Systematic errors pose challenges to reaching greater precision in dark energy properties

•We have data in hand to help resolve these issues for SNe: retraining light-curve models using SDSS, CSP, CfA, SNF,...

• Future experiments will/must be designed to control systematic errors and exploit complementarity of multiple probes: HETDEX, DES, PanSTARRS, PAU, LSST, JDEM,...

Dust Extinction Law: R_V



Jha



Determine priors and efficiencies from data and Monte Carlo simulations

Determine Survey Efficiencies



SALT vs MLCS

Diagnosis:

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