# Constraining Curvature Density Parameter with Time-Delay Strong Lenses and Complimentary Probes: a Forecast for Next-Generation Surveys

Project Report

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#### Abstract

It is of great interest among cosmologists to obtain a high-fidelity constraint on curvature density parameter  $\Omega_k$  of late universe that is independent of the early universe Cosmic Microwave Background measurement, and this report presents a framework to do so with the combined data of time-delay distances from strong gravitational lenses and complimentary probes. Simulated data of next-generation surveys are used in model-dependent and non-parametric analysis where Markov chain Monte Carlo sampling is used to constrain  $\Omega_k$ . By assuming a non-flat wCDM model, the addition of strong lensing data into the analysis with data of Type Ia supernovae and Baryon Acoustic Oscillation breaks the w- $\Omega_k$  degeneracy, placing a strong constraint on  $\Omega_k$ . Similar order of magnitude of constraint on  $\Omega_k$  was obtained in non-parametric analysis. This serves as a promising forecast for next-generation surveys on constraining  $\Omega_k$ .

### 1 Introduction

The curvature density parameter,  $\Omega_k$ , is an important cosmological parameter that determines the geometry of our Universe. Probes such as Type Ia supernovae (SNe Ia), strong gravitational lenses and baryon acoustic oscillations (BAO) give estimates on  $\Omega_k$  of the late universe, leading to possible ways of obtaining a high-fidelity constraint on  $\Omega_k$  independent of the early-universe Cosmic Microwave Background (CMB) measurement. The constraint placed on  $\Omega_k$  by SNe Ia, strong gravitational lenses and BAO individually have been widely analysed and discussed in past literature. A promising study on constraining  $\Omega_k$  is to look at a large number of time-delay distance measurements from strong gravitational lenses and combine it with complimentary probes, hoping that the combined data can overcome the deficiency of limited lenses events in previous analysis such as that by H0LiCOW on 6 lenses [1] and degeneracies among cosmological parameters in single-probe analysis.

In light of the next-generation survey from the Rubin Observatory's legacy survey of space and time (LSST), it is expected that hundreds of strong lensing events will be observed during its 10-year survey baseline [2]. This will provide a large dataset for studying the ability of strong lenses to constrain  $\Omega_k$ . In this project, we use simulated LSST-like data [3] to explore 3 tasks. Firstly, we combine it with a past survey, Pantheon [4], of apparent magnitude measurement of SNe Ia to see the improvement made on constraining  $\Omega_k$  by adding lenses into the analysis. Different models of dark energy in a non-flat universe, i.e.  $o\Lambda$ CDM and owCDM, are used in the process of generating mock values of "measurements" for each next-generation survey. Then, we use simulated next-generation Roman-like SNe Ia data [5] and DESI-like BAO data [6] and combine them with LSST-like data to see their ability on constraining  $\Omega_k$ . Finally, we use Gaussian Process to fit Hubble parameter  $H_z$  and look at a non-parametric method to analyse the constraint that lenses and complimentary probes place on  $\Omega_k$ . We hope that this project provides an elementary but instructive forecast on using next-generation surveys to constrain  $\Omega_k$ .

# 2 Methodology

#### 2.1 Formulae in cosmology

The three late-universe probes used in this project are strong gravitational lenses, SNe Ia and BAO. Each of them has a measurable quantity that is linked to the curvature density parameter  $\Omega_k$  via some formulae. For the purpose of having a reference section for this brief report, relevant formulae will be listed below without derivations.

For lenses, we are interested in measuring time-delay distances  $D_{\Delta t}$  of a lens at redshift  $z_l$  and a source at  $z_s$ :

$$D_{\Delta t} = \frac{(1 + z_s)D_{A,l}D_{A,s}}{D_{A,ls}},\tag{1}$$

where  $D_{A,l}$  and  $D_{A,s}$  are angular diameter distances at lens and source respectively, and  $D_{A,ls}$  is the angular diameter distance between lens and source.  $D_A$  at a given redshift can be calculated by  $D_L/(1+z)$  where  $D_L$  is the luminosity distance at the same redshift with which the formula will be given below.  $\Omega_k$  appears in  $D_L$ .

For SNe Ia, what we measure is the apparent magnitude  $m_B$  given by:

$$m_B = 5\lg(D_L) + 25 + M_B,$$
 (2)

where  $M_B$  is the absolute magnitude and  $D_L$  is the luminosity distances at z.  $D_L$  is given by:

$$D_L = \frac{c(1+z)}{H_0\sqrt{|\Omega_k|}}F(\sqrt{|\Omega_k|}\int_0^z \frac{\mathrm{d}z'}{E(z')}),\tag{3}$$

where function F is sin for  $\Omega_k \leq 0$  and sinh for  $\Omega_k > 0$ .

For BAO, we measure  $H_z$ , the Hubble parameter. Its relation to  $H_0$ ,  $\Omega_k$  and other density parameters is trivially derived from the Friedman equations.

#### 2.2 Statistical setup

The method to get constraints on parameters in this project is Markov chain Monte Carlo (MCMC), a sampling technique in Bayesian inference. Using MCMC, we drawn samples for cosmological parameters to get distributions of them based on measured data provided. The mean and the standard deviation of the distributions become the constraints placed on cosmological parameters.

In Bayesian inference, a prior distribution and a likelihood function are required. Throughout the project, we use conservative uniform prior distributions given by the table below for each cosmological parameter marginalised:

Parameter	Prior
$H_0$	U(0, 150)
$\Omega_m$	U(0.05, 0.5)
$\Omega_k$	U(-0.5, 0.5)
w	U(-2.5, 0.5)
$M_B$	U(-38.4, 0)

The likelihood function is set to return a  $\chi^2$  term for lenses and BAO in this form:

$$\chi^2 = \sum_{\text{all probes used}} \sum_i \frac{(x_i - x_{m,i})^2}{\sigma_{x_i}^2}.$$
 (4)

 $x_i$ ,  $x_{m,i}$  and  $\sigma_{x_i}$  indicate measured value, modelled value and percentage uncertainty in measurement  $x_i$  respectively. For example, for lenses,  $x_i$  is supposed to be a measured value of  $D_{\Delta t}$ ; for BAO,  $x_i$  is a measured value of  $H_z$ . For SNe Ia, the  $\chi^2$  term takes this form instead:

$$\chi^2 = \Delta m_B^T \cdot C^{-1} \cdot \Delta m_B, \tag{5}$$

where  $\Delta m_B = m_B - m_{B \text{model}}$  is a vector of difference between measured and modelled apparent magnitudes and  $C = D_{\text{stat}} + C_{\text{sys}}$  is the covariance matrix that takes into account both statistical and systematic uncertainty.

Data analysis in this project is done with Python. The key package used to calculate cosmological quantities in many cases is astropy.cosmology. The package used to run MCMC is emcee, and the package to do Gaussian Process is george. Note that in most cases in this project, we are doing a forecast for next-generation surveys, and the "measured" dataset is a simulated dataset generated from a mock universe of either FlatwCDM,  $o\Lambda$ CDM or owCDM model using methods in astropy.cosmology.

In most cases, we work in a mock owCDM universe with cosmological parameters set to be  $(H_0, \Omega_M, \Omega_k, w, M_B) = (72, 0.3, 0, -1, -19.2)$  and run MCMC analysis with  $(n_{\text{walkers}}, n_{\text{samples}}) = (32, 20000)$ . We will refer to such a statistical setting and default values of cosmological parameters as the "standard setting".

# 3 Results and analysis

## 3.1 Combining LSST lenses with a past survey

It is well-known that for parametric analysis on SNe Ia, there exists a degeneracy between Hubble's constant  $H_0$  and  $M_B$ . This is because both of them enter equation (2) as additive terms. Similar degeneracies exist between w in equation of state of dark energy and density parameters  $\Omega_k$  and  $\Omega_M$ . These set a limit on the ability of single-probe data to constrain  $\Omega_k$ . To see these, we use the redshifts in binned Pantheon dataset and generate  $m_B$  measurements in the standard setting. Figure 1 (a) clear shows the degeneracies and limitations on using SNe Ia alone to constrain  $\Omega_k$ :  $\Omega_k = 0.090^{+0.248}_{-0.320}$ , a weak constraint.

On the other hand, during the 10-year LSST survey, approximately 310 lensing events are expected to be observed. We call the 310 pairs of  $(z_l, z_s)$  in the source file given by LSST DESC data products[3] data that satisfies a LSST-like distribution, or LSST-like data. For model-dependent analysis, we assumed the standard setting. LSST-like data was used to generate  $D_{\Delta t}$  measurements using equation (1). The number of effective lensing events can increase if quasars and other similar probes are taken into account, and at some instances in this project we increased the number to 3000 by drawing samples from the original source file and appended to it while maintaining the LSST-like distribution. The uncertainty of each simulated measurement was generated as a random number between 6%-10%. It is also widely understood that the combination of angular diameter distances in equation (1) makes  $D_{\Delta t}$  approximately inversely proportional to  $H_0$ , placing a strong constraint on  $H_0$ . However, even with 3000 lenses, the single-probe situation on constraining  $\Omega_k$  is not optimistic, either. MCMC analysis with standard setting shows that curvature density parameter is weakly constrained by lenses:  $\Omega_k = 0.337^{+0.125}_{-0.317}$  (fig. 1 (b)).

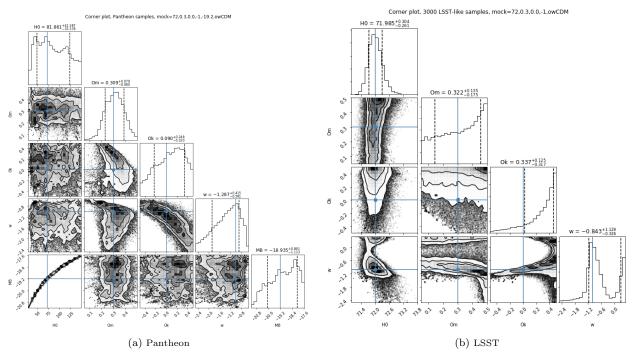


Figure 1: Constraints placed on cosmological parameters by 2 single-probe datasets

While lenses are decent at constraining  $H_0$  and SNe Ia are not, it is promising to study how the combined

data can compliment each other in constraining cosmological parameters, especially that of the  $\Omega_k$  which both individual probe analysis fail to place a strong constraint alone. Looking at the shapes of the w- $\Omega_k$  contours in fig. 1, we see that Pantheon produces a "banana" shaped graph indicating degeneracy, while LSST has a more horizontal graph, indicating strong constraint on w and weak constraint on  $\Omega_k$ . By adding equation (4) and (5) together as the output of log-likelihood function and run MCMC with standard setting, we obtain the constraint by Pantheon+LSST:  $\Omega_k = -0.058^{+0.112}_{-0.113}$ . A more detailed table of results of constraints placed on various parameters in non-standard settings can be found in Appendix A.

We can see that the addition of lenses is helping to break the degeneracy between w and  $\Omega_k$  in SNe Ia analysis, as seen from fig. 2, hence complimenting the ability of the past survey Pantheon on constraining  $\Omega_k$ . However, due to the limitation in precision of measurement in past surveys, the systematic uncertainty of  $m_B$  measurements in Pantheon is relatively large. The constraint achieved on  $\Omega_k$  is still not desirable.

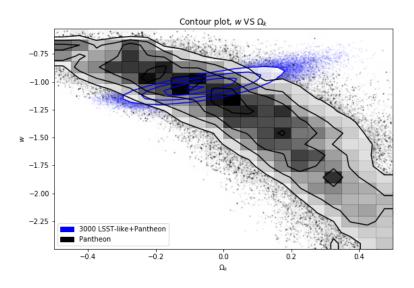


Figure 2: Comparison of w- $\Omega_k$  contour with and without lenses

#### 3.2 Combining LSST lenses with next-generation surveys

To make fairer comparison and forecast for future analysis, we combine LSST lenses with next-generation surveys. The next-generation SNe Ia survey we used is Roman which has a systematic uncertainty matrix of approximately  $10^1$  magnitude smaller at every entry compared to that of *Pantheon*. We take simulated redshifts from *Hounsell et al 2018* [5] and generate  $m_B$  measurement from those redshifts under the same mock universe using equation (2). An intrinsic magnitude scattering of a Gaussian distribution of  $\mu = -19.2$ mag and  $\sigma = 0.02$ mag is added to  $M_B$  for the binned Roman data to account for statistical distribution of absolute magnitude of SNe Ia. Note that by adding the intrinsic scattering, the mock universe is close to but not exactly under standard setting. This potentially causes a bias in the cosmological parameters obtained by MCMC from that of the mock values, and the magnitude and sign of bias depend on which random seed is used to generate the intrinsic scattering when applying numpy.random.seed() method.

By running MCMC under standard setting with intrinsic scattering on single-probe and multi-probe datasets, we obtain the w- $\Omega_k$  contour plot shown in fig. 3. Notice how the inclusion of lenses and BAO dataset help to break the w- $\Omega_k$  degeneracy in SNe Ia dataset. The introduction of intrinsic scattering generated by

random seeds causes a noticeable bias in the w- $\Omega_k$  contour plot from the mock values w = -1,  $\Omega_k = 0$ , and fig. 4 shows such biases for seed 20, 21 and 22. Take note that if no seed number is mentioned in a plot, it is using seed 20 by default.

The constraint obtained by combining 3 probes and averaging for seed 20, 21 and 22 on curvature is:  $\Omega_k = -0.037^{+0.014}_{-0.014}$ . The  $1\sigma$  interval has almost reduced by an order of magnitude as compared to the analysis with Pantheon, as the simulated Roman dataset has a much smaller systematic uncertainty in its covariance matrix, reducing the width of the banana-shaped contour in fig. 2, and the inclusion of the horizontal lenses contour helps to break the degeneracy of SNe Ia and BAO datasets, placing a strong contraint on  $\Omega_k$ .

A more detailed record of results under other non-standard setting for the constraint by next-generation surveys on all relevant cosmological parameters can be found in Appendix B.

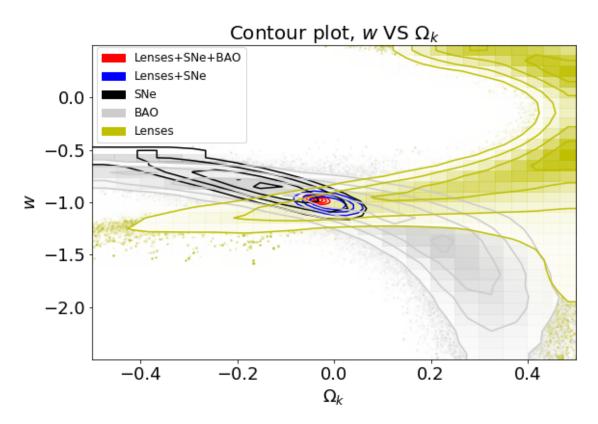


Figure 3: Comparison of w- $\Omega_k$  contour with different probes

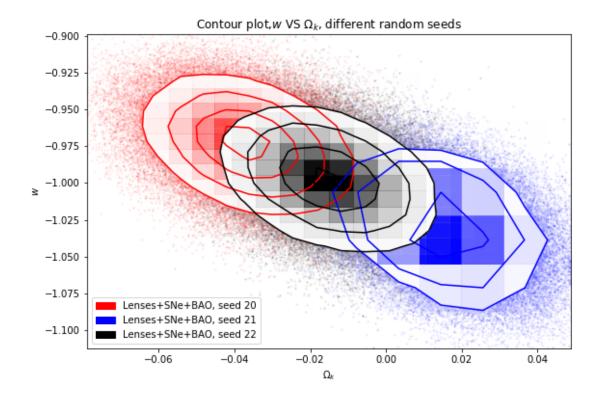


Figure 4: Comparison of w- $\Omega_k$  contour with different seeds

### 3.3 Non-parametric inference

It is also worthwhile to consider the effect of adding lenses into non-parametric analysis on constraining  $\Omega_k$ . To carry out non-parametric analysis, we no longer relying on assuming a mock universe with a set of truth values of w = -1,  $\Omega_k = 0$ , etc. What we do is interpolating values of  $E(z) = H_z/H_0$ , which appears in equation (3), using Gaussian Process (GP). E(z), the normalised Hubble parameter, appears in all cosmological distances involved in this project. It is a function of density parameters and w, and the form of this function depends on the model chosen. GP does not require an assumed model of the universe, but rather it is interpolating values statistically using a few given data points. We mentioned that  $H_z$  will be measured in BAO survey, so, we use 19 simulated BAO  $H_z$  "measurement" to run GP using package george with exponential-squared kernel and optimal hyper-parameters. The optimal hyper-parameter for this kernel, A (amplitude) and l (length scale), are determined by MCMC.

The outcome of GP is an interpolated curve for  $H_z$  between redshift 0 and 1.85. This curve of  $H_z$  is then used to calculate E(z) which appears in equation (1)&(3) for  $D_{\Delta t}$  and  $D_L$ . This enables the previously modelled values in equation (4)&(5) to be generated in a non-parametric way. The rest of the steps is the same as those in section 4. We run MCMC with 310 lenses cases and  $(n_{\text{walkers}}, n_{\text{samples}}) = (32, 20000)$ . Figure 5 shows the constraints obtained with Roman and 310 LSST-like lenses. This again corroborates with our previous discussion that SNe Ia gives poor constraint on  $H_0$  and lenses give poor constraint on  $\Omega_k$ , and they are complimenting each other to give strong constraints on both  $H_0$  and  $\omega_k$ .

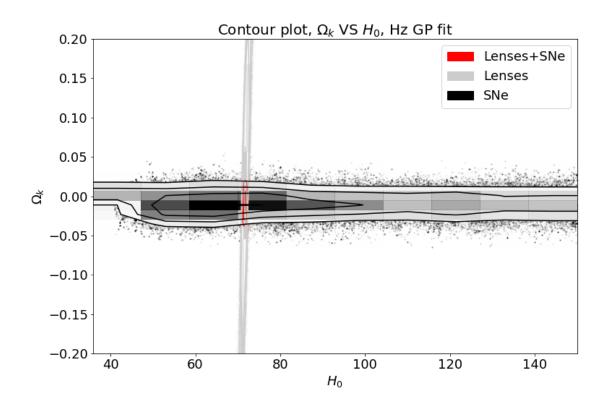


Figure 5: Comparison of  $\Omega_k$ - $H_0$  contour in non-parametric analysis

Figure 6 shows a summary of non-parametric analysis with 310 lenses on constraining parameters. The curvature density parameter is constrained to be:  $\Omega_k = -0.009^{+0.015}_{-0.015}$ . This is comparable to the constraint obtained in model-dependent analysis with 3000 lenses! The histograms shows the improvements made by adding lenses into the analysis. A detailed record of result can be found in Appendix C.

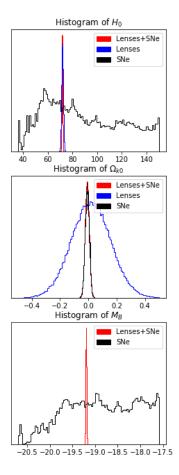


Figure 6: Summary of constraints with LSST and Roman datasets in non-parametric analysis

### 4 Conclusion

In this project, we explored the ability of lenses to constrain  $\Omega_k$  and combine it to a past survey and future surveys to see the improvements made on constraining  $\Omega_k$  in both model-dependent and independent analysis. Result shows optimistic forecast on adding lenses into analysis to get strong constraint on  $\Omega_k$ . When adding lenses to the past Pantheon survey, the constraint obtained is  $\Omega_k = -0.058^{+0.122}_{-0.113}$ . When combining lenses with simulated next-generation surveys, the constraint is  $\Omega_k = -0.037^{+0.014}_{-0.014}$ . In non-parametric analysis, the constraint achieved is  $\Omega_k = -0.009^{+0.015}_{-0.015}$  for 310 lenses.

Although failed to place a strong constraint on  $\Omega_k$  alone, lenses contributes positively at constraining  $\Omega_k$  when combined with other probes explored in this project, as lenses compliments the weakly constrained parameters of other probes (e.g.  $H_0$ ), hence improving the constraint achieved on all parameters overall. It will be promising to study what constraint on  $\Omega_k$  can be achieved when combining lenses with other consistent and complimentary probes, as well as analysis with more complicated model of the universe such as  $ow_0w_a{\rm CDM}$  and slow-roll model. Hopefully, this project can serve as an elementary but instructive forecast on using next-generation surveys to constrain  $\Omega_k$ .

### References

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- [3] LSST Dark Energy Science Collaboration (LSST DESC) et al. "The LSST DESC DC2 Simulated Sky Survey". In: 253.1, 31 (Mar. 2021), p. 31. DOI: 10.3847/1538-4365/abd62c. arXiv: 2010.05926 [astro-ph.IM].
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- [5] R Hounsell et al. "Simulations of the WFIRST supernova survey and forecasts of cosmological constraints". In: *The Astrophysical Journal* 867.1 (2018), p. 23.
- [6] Kyle S Dawson et al. "The SDSS-IV extended Baryon Oscillation Spectroscopic Survey: overview and early data". In: *The Astronomical Journal* 151.2 (2016), p. 44.

# Appendix A

Constraints on parameters with simulated lenses and Pantheon

Cosmological model		$H_0$	$\Omega_m$	$\frac{\Omega_k}{\Omega_k}$	$\overline{w}$	$M_B$
1000 Strong lenses + SNe Ia						
$o\Lambda { m CDM}$	Mock MCMC	$72 \\ 71.906^{+0.324}_{-0.327}$	$0.3 \\ 0.312^{+0.047}_{-0.048}$	$0.00 \\ -0.039^{+0.118}_{-0.110}$	-1 ≡ -1	$-19.2 \\ -19.207^{+0.018}_{-0.018}$
$2000 \ Strong \ lenses + SNe \ Ia$						
$o\Lambda{ m CDM}$	Mock MCMC	$72 \\ 71.939^{+0.254}_{-0.255}$	$0.3 \\ 0.307^{+0.039}_{-0.040}$	$0.00 \\ -0.024^{+0.093}_{-0.090}$	-1 ≡ -1	$-19.2 \\ -19.205^{+0.016}_{-0.016}$
$3000 \ Strong \ lenses + SNe \ Ia$						
$o\Lambda \mathrm{CDM}$	Mock MCMC	$72 \\ 71.959^{+0.212}_{-0.215}$	$0.3 \\ 0.304^{+0.035}_{-0.035}$	$\begin{array}{c} 0.00 \\ -0.017^{+0.080}_{-0.076} \end{array}$	-1 ≡ −1	$-19.2 \\ -19.205^{+0.015}_{-0.015}$
$o\Lambda { m CDM}$	Mock MCMC	$72 \\ 71.952^{+0.210}_{-0.211}$	$0.3 \\ 0.304^{+0.035}_{-0.035}$	$0.03 \\ 0.010^{+0.081}_{-0.078}$	-1 ≡ −1	$-19.2 \\ -19.205^{+0.015}_{-0.015}$
$o\Lambda { m CDM}$	Mock MCMC	$72 \\ 71.953^{+0.211}_{-0.211}$	$0.3 \\ 0.304^{+0.035}_{-0.036}$	$0.05 \\ 0.030^{+0.084}_{-0.080}$	-1 ≡ -1	$-19.2 \\ -19.205^{+0.015}_{-0.015}$
$o\Lambda { m CDM}$	$_{\rm MCMC}^{\rm Mock}$	$72 \\ 71.954^{+0.212}_{-0.214}$	$0.3 \\ 0.304^{+0.034}_{-0.035}$	$-0.03 \\ -0.048^{+0.078}_{-0.075}$	-1 ≡ -1	$-19.2 \\ -19.205^{+0.015}_{-0.015}$
$o\Lambda { m CDM}$	Mock MCMC	$72 \\ 71.954^{+0.213}_{-0.213}$	$0.3 \\ 0.304^{+0.034}_{-0.034}$	$-0.05 \\ -0.066^{+0.076}_{-0.072}$	-1 ≡ −1	$-19.2 \\ -19.205^{+0.015}_{-0.015}$
$ow\mathrm{CDM}$	Mock MCMC	$72 \\ 72.059^{+0.285}_{-0.275}$	$0.3 \\ 0.337^{+0.071}_{-0.076}$	$0\\-0.058^{+0.122}_{-0.113}$	$ \begin{array}{c} -1 \\ -1.043^{+0.087}_{-0.078} \end{array} $	$-19.2 \\ -19.205^{+0.015}_{-0.015}$
3000 Strong lenses						
$\mathrm{F}w\mathrm{CDM}$	Mock MCMC	$72 \\ 71.997^{+0.319}_{-0.287}$	$0.3 \\ 0.313^{+0.131}_{-0.162}$	$0 \equiv 0$	$ \begin{array}{r} -1 \\ -1.029^{+0.063}_{-0.081} \end{array} $	-

No. of strong gravitational lensing events = 3000

No. of binned SNe Ia events = 40

No. of walkers in MCMC = 32

No. of samples in MCMC = 20000

numpy.random.seed() used in all random processes: 20

# Appendix B

Constraints on parameters with simulated lenses and other probes

Constraints on parameters with simulated lenses and other probes						
Cosmological mod	del	$H_0$	$\Omega_m$	$\Omega_k$	w	$M_B$
Lenses						
CDM	Mock	72	0.3	0	-1	-
owCDM	$MCMC^*$	$71.985^{+0.304}_{-0.261}$	$0.322^{+0.135}_{-0.175}$	$0.337^{+0.125}_{-0.317}$	$-0.843^{+1.129}_{-0.226}$	-
		-0.261	-0.175	-0.317	-0.320	
SNe Ia						
Sive iu	Mock	72	0.3	0	_1	-10.2
owCDM	MCMC	74 121+28.898	0.0	0.156+0.121	0.020+0.155	$-19.2 \\ -19.111^{+0.713}_{-0.718}$
	MOMO	$^{14.121}_{-20.841}$	$0.302_{-0.074}$	$-0.150_{-0.150}$	$-0.829_{-0.167}$	$-19.111_{-0.718}$
BAO						
BAO	M 1	70	0.9	0	1	
owCDM	Mock	72	0.3	0	-1	-
	MCMC	$72.738^{+5.279}_{-3.941}$	$0.246^{+0.036}_{-0.078}$	$0.180^{+0.166}_{-0.351}$	$-1.215_{-0.658}^{+0.662}$	-
Lenses + SNe Ia						
$o\Lambda { m CDM}$	Mock	72	0.3	0	-1	-19.2
ONODIVI	MCMC	$71.957^{+0.112}_{-0.112}$	$0.317^{+0.005}_{-0.005}$	$-0.017^{+0.018}_{-0.017}$	$\equiv -1$	$-19.2 \\ -19.192^{+0.007}_{-0.007}$
owCDM	Mock	72	0.3	0	-1	-19.2
owCDM	MCMC	$72.004^{+0.259}_{-0.261}$	$0.318^{+0.007}_{-0.008}$	$-0.012^{+0.028}_{-0.028}$	$-1.008^{+0.042}_{-0.044}$	$ \begin{array}{r} -19.2 \\ -19.191^{+0.009}_{-0.009} \end{array} $
		0.201	0.000	0.020	0.011	0.003
Lenses + SNe Ia + BAO						
·	Mock	72	0.3	0	-1	-19.2
$o\Lambda \mathrm{CDM}$	MCMC	$71.905^{+0.100}$	$0.315^{+0.005}$	$-0.031^{+0.013}$	= -1	$-19.2 \\ -19.201^{+0.004}_{-0.004}$
	WOW	11.505_0.101	0.010_0.005	0.001_0.013	_ 1	13.201-0.004
	Mock	72	0.3	0	-1	-19.2
owCDM	MCMC1*					$-19.199^{+0.004}_{-0.004}$
	MCMCI	11.181_0.148	$0.515_{-0.006}$	$-0.050_{-0.014}$	$-0.974_{-0.024}$	$-19.199_{-0.004}$
	M 1	70	0.9	0	1	10.0
owCDM	Mock	72	0.3	0	-1	$ \begin{array}{r} -19.2 \\ -19.201^{+0.004}_{-0.004} \end{array} $
	$MCMC2^*$	$72.170_{-0.164}^{+0.166}$	$0.296^{+0.006}_{-0.006}$	$0.014_{-0.015}^{+0.014}$	$-1.028^{+0.027}_{-0.026}$	$-19.201_{-0.004}^{+0.004}$
	3.5				_	40.0
owCDM	Mock	72	0.3	0	-1	-19.2
	$MCMC3^*$	$71.954^{+0.145}_{-0.150}$	$0.306^{+0.006}_{-0.006}$	$-0.015^{+0.014}_{-0.014}$	$-0.997^{+0.024}_{-0.024}$	$-19.202^{+0.004}_{-0.004}$

No. of strong gravitational lensing events = 3000

No. of binned SNe Ia events = 40

No. of BAO measurements = 18

No. of walkers in MCMC = 32

No. of samples in MCMC = 20000

<sup>\*</sup> MCMC & MCMC1 indicate seed=20, MCMC2 indicates seed=21, MCMC3 indicates seed=22

# Appendix C

Constraint on parameters in non-parametric analysis

Constraint on parameters in non parameter analysis					
Probe		$H_0$	$\Omega_k$	$M_B$	
SNe Ia	Mock MCMC	$72 \\ 68.619^{+14.652}_{-11.738}$	$0\\-0.011^{+0.016}_{-0.016}$	$-19.2 \\ -19.295^{+0.419}_{-0.406}$	
Strong lenses	Mock MCMC	$72 \\ 71.183^{+1.335}_{-1.328}$	$  0 \\  0.070^{+0.116}_{-0.115} $	-	
Strong lenses+SNe Ia	Mock MCMC	$72 \\ 70.440^{+0.723}_{-0.702}$	$0\\-0.008^{+0.015}_{-0.014}$	$-19.2 \\ -19.238^{+0.022}_{-0.021}$	

No. of strong gravitational lensing events = 310

No. of binned SNe Ia events =40

No. of walkers in MCMC = 32

No. of samples in MCMC = 20000