

# Optimal Parameter Constraints on Dark Energy Models

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# 1 Introduction

- The cosmic narrative, as described by the standard Big Bang Theory, attributes the universe's accelerated expansion to a cosmological constant known as dark energy, constituting a substantial 70% of the total energy budget of the universe.
- However, the precision of modern observational tools has ushered in an era of detailed scrutiny, revealing intriguing inconsistencies that challenge the robustness of this established model.
- These observational discrepancies have prompted a critical reexamination of our cosmological understanding, leading to the exploration of alternative models that incorporate non-zero spatial curvature and dynamic dark energy.
- In this context, this project endeavours to investigate and constrain four distinct cosmological models that diverge from the conventional paradigm.
- By using standardized, lower-redshift observations, this project aims to shed light on the details of cosmic evolution, offering insights into the nature of spatial curvature and the dynamical behaviour of dark energy.
- The foundation of the analysis rests on the combined use of Hubble parameter ( $H(z)$ ) and Baryon Acoustic Oscillation (BAO) data, enabling the identification of the most likely cosmological scenarios.

## 2 Cosmological Models

### 2.1 $\Lambda$ CDM model

- The Hubble parameter in the  $\Lambda$ CDM model (CDM - Cold Dark Matter) is

$$H(z) = H_0 \sqrt{\Omega_{m0}(1+z)^3 + \Omega_{k0}(1+z)^2 + \Omega_\Lambda} \quad (1)$$

- In the above equation  $H_0$  is the Hubble constant and  $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\Omega_\Lambda$  are the present values of the non-relativistic matter density parameter, the spatial curvature energy density parameter, and the cosmological constant energy density parameter, respectively.
- $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\Omega_\Lambda$  are related through the equation  $\Omega_{m0} + \Omega_{k0} + \Omega_\Lambda = 1$
- In the **spatially non-flat  $\Lambda$ CDM** model, the conventional choice of free parameters is  $\Omega_{m0}$ ,  $\Omega_{k0}$  and  $H_0$ , while in the **spatially-flat  $\Lambda$ CDM** model we use the same set of free parameters but now with  $\Omega_{k0} = 0$ .

### 2.2 XCDM model

- The Hubble parameter in the XCDM model is

$$H(z) = H_0 \sqrt{\Omega_{m0}(1+z)^3 + \Omega_{k0}(1+z)^2 + \Omega_{X0}(1+z)^{3(1+\omega_X)}} \quad (2)$$

- In the above equation  $H_0$  is the Hubble constant and  $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\Omega_{X0}$  are the present values of the non-relativistic matter density parameter, the spatial curvature energy density parameter, and the X-fluid dark energy density parameter, respectively.
- $\omega_X$  is the equation of state parameter of the X-fluid (the ratio of the pressure to the energy density).
- $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\Omega_{X0}$  are related through the equation  $\Omega_{m0} + \Omega_{k0} + \Omega_{X0} = 1$
- In the **spatially non-flat XCDM** model, the conventional choice of free parameters is  $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\omega_X$  and  $H_0$ , while in the **spatially-flat XCDM** model we use the same set of free parameters but now with  $\Omega_{k0} = 0$ .
- In the XCDM parameterization when  $\omega_X = -1$  the  $\Lambda$ CDM model is recovered.

## 2.3 $\phi$ CDM model

- The Hubble parameter in the XCDM model is

$$H(z) = H_0 \sqrt{\Omega_{m0}(1+z)^3 + \Omega_{k0}(1+z)^2 + \Omega_\phi(z, \alpha)} \quad (3)$$

- In the above equation **H<sub>0</sub>** is the Hubble constant and  $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\Omega_\phi$  are the present values of the non-relativistic matter density parameter, the spatial curvature energy density parameter, and the scalar field dynamical dark energy density parameter, respectively.
- $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\Omega_\phi(z, \alpha)$  are related through the equation  $\Omega_{m0} + \Omega_{k0} + \Omega_\phi(z, \alpha) = 1$
- In the **spatially non-flat  $\phi$ CDM** model, the conventional choice of free parameters is  $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\alpha$  and  $H_0$ , while in the **spatially-flat  $\phi$ CDM** model we use the same set of free parameters but now with  $\Omega_{k0}=0$ .
- In the  $\phi$ CDM parameterization when  $\alpha = 0$  the  $\Lambda$ CDM model is recovered.

## 3 Data

- In this project a combination of  $H(z)$ , BAO and QSO data is used.

### 3.1 $H(z)$ data

- This data is crucial for understanding the evolution of the universe and can provide insights into the nature of dark energy, dark matter, and the overall dynamics of the cosmos.
- $H(z)$  is Hubble parameter as a function of redshift.
- In this project 31  $H(z)$  data measurements from [1] are used. These were determined using the cosmic chronometric technique.
- [Table15](#) contains the  $H(z)$  data.  $H(z)$  and  $\sigma_H$  have units of  $Kms^{-1}Mpc^{-1}$

### 3.2 BAO Data

- BAO typically stands for Baryon Acoustic Oscillations, a feature in the universe's large-scale structure. Baryon Acoustic Oscillations are fluctuations in the density of visible matter (baryons) that occurred shortly after the Big Bang. These fluctuations have left a signature in the distribution of galaxies and other cosmic structures we observe today.
- BAO data is often used in cosmology to study the universe's large-scale structure and measure the expansion rate. By analyzing the distribution of galaxies, astronomers can study the imprints of Baryon Acoustic Oscillations and use them as a cosmological ruler to probe the geometry and expansion history of the universe.
- In this project 11 BAO data measurements from [2] are used.
- [Table16](#) contains the BAO data.
- In [Table16](#)  $D_M(r_{s, fid}/r_s)$ ,  $D_V(r_{s, fid}/r_s)$ ,  $r_s$ , and  $r_{s, fid}$  have units of Mpc, while  $H(z)(r_s/r_{s, fid})$  has units of  $Kms^{-1}Mpc^{-1}$  and  $D_A/r_s$  is dimensionless.

$$r_s = \frac{55.154 \exp[-72.3(\Omega_{vo}h^2 + 0.0006)^2]}{(\Omega_{bo}h^2)^{0.1280}(\Omega_{cbo}h^2)^{0.25351}} \quad (4)$$

- Here  $\Omega_{cbo} = \Omega_{co} + \Omega_{bo} = \Omega_{mo} - \Omega_{vo}$  with  $\Omega_{cbo}$ ,  $\Omega_{co}$ ,  $\Omega_{bo}$  and  $\Omega_{vo} = 0.0014$  being the current values of the CDM + baryonic matter, CDM, baryonic matter, and neutrino energy density parameters, respectively, and the Hubble constant **H<sub>0</sub> = 100h Kms<sup>-1</sup>Mpc<sup>-1</sup>**
- The subscript of 'o' on a given quantity denotes the current value of that quantity.

- Additionally,  $\Omega_{b0}h^2$  is slightly model dependent; the values of this parameter that are used in this project are from [3] and are given Table17
- BAO provides observers with a “standard ruler” which can be used to measure cosmological distances. These distances can be computed in a given cosmological model, so measurements of them can be used to constrain the parameters of the model in question.
- The transverse co-moving distance is

$$D_M(z) = \begin{cases} D_C(z) & \text{if } \Omega_{k0} = 0, \\ \frac{c}{H_0\sqrt{\Omega_{k0}}} \sinh\left(\frac{D_C H_0}{c} \sqrt{\Omega_{k0}}\right) & \text{if } \Omega_{k0} > 0, \\ \frac{c}{H_0\sqrt{|\Omega_{k0}|}} \sin\left(\frac{D_C H_0}{c} \sqrt{|\Omega_{k0}|}\right) & \text{if } \Omega_{k0} < 0. \end{cases}$$

- $D_H = \frac{c}{H(z)}$
- $D_C = \frac{c}{H_0} \int_0^z \frac{dz'}{E(z')}$ ,
- The volume-averaged angular diameter distance is  $D_V(z) = [\frac{cz}{H_0} \frac{D_M^2(z)}{E(z)}]^{1/3}$
- The angular diameter distance at redshift (z) is  $D_A(z) = \frac{D_M(z)}{1+z}$

### 3.3 QSO Data

- It is found that compact structures in intermediate-luminosity radio quasars could serve as standard cosmological rulers. The QSO data used in this project come from a newly compiled sample of these standard rulers from observations of 120 intermediate-luminosity quasars taken over a redshift range of  $0.46 < z < 2.76$ , with angular sizes  $\theta_{\text{obs}}(z)$  and redshifts(z) listed in Table18 and Table19

## 4 Method

### 4.1 Chi-square Minimization

- In this project, the  $\chi^2$  statistic is used to find the best-fitting parameter values and limits for a given model. Most of the data points we use are uncorrelated. Most of the data we analyze are uncorrelated. For uncorrelated data points,  $\chi^2$  is given by

$$\chi^2(p) = \sum_{i=1}^N \frac{[A_{\text{th}}(p; z_i) - A_{\text{obs}}(z_i)]^2}{\sigma_i^2} \quad (5)$$

- Here p is the set of model parameters, for example,  $p = (H_0, \Omega_{m0})$  in the flat  $\Lambda$ CDM model,  $z_i$  is the redshift at which the measured value is  $A_{\text{obs}}(z_i)$  with one standard deviation uncertainty  $\sigma_i$ , and  $A_{\text{th}}(p; z_i)$  is the predicted value computed in the model under consideration. The  $\chi^2$  expression in eq.(5) holds for the  $H(z)$  measurements listed in Table15 and the BAO measurements listed in the last five lines of Table16.
- The measurements in the first six lines of Table16 are correlated, in which case  $\chi^2$  is given by

$$\chi^2(p) = [A_{\text{th}}(p) - A_{\text{obs}}]^T C^{-1} [A_{\text{th}}(p) - A_{\text{obs}}] \quad (6)$$

where  $C^{-1}$  is the inverse of the covariance matrix C.

$$C = \begin{bmatrix} 624.707 & 23.729 & 325.332 & 8.34963 & 157.386 & 3.57778 \\ 23.729 & 5.60873 & 11.6429 & 2.33996 & 6.39263 & 0.968056 \\ 325.332 & 11.6429 & 905.777 & 29.3392 & 515.271 & 14.1013 \\ 8.34963 & 2.33996 & 29.3392 & 5.42327 & 16.1422 & 2.85334 \\ 157.386 & 6.39263 & 515.271 & 16.1422 & 1375.12 & 40.4327 \\ 3.57778 & 0.968056 & 14.1013 & 2.85334 & 40.4327 & 6.25936 \end{bmatrix}$$

- For the QSO data the below formula is used:

$$\chi^2(p) = \sum_{i=1}^N \left[ \frac{\theta_{\text{th}}(p; z_i) - \theta_{\text{obs}}(z_i)}{\sigma_i + 0.1\theta_{\text{obs}}(z_i)} \right]^2 \quad (7)$$

Where  $\theta_{\text{th}}(p; z_i)$  is the model-predicted value of the angular size,  $\theta_{\text{obs}}(z_i)$  is the measured angular size at redshift  $z_i$ , and  $\sigma_i$  is the uncertainty on the measurement made at redshift  $z_i$ . The term is proportional to  $\theta_{\text{obs}}(z_i)$  in the denominator is added to  $\sigma_i$  to account for systematic uncertainties in the angular size measurements.

## 4.2 MCMC Analysis

- In this project **Markov Chain Monte Carlo (MCMC)** statistical technique is used to estimate the distribution of model parameters in a Bayesian framework. MCMC is particularly valuable when dealing with complex models or models for which the likelihood function is difficult to compute.
- **Bayesian Inference** : MCMC is often used in Bayesian statistical analysis. Bayesian inference involves updating our beliefs about a parameter based on new evidence or data. The updated belief is represented by a probability distribution, and MCMC helps in sampling from this distribution.
- **Markov Chain** : MCMC involves creating a Markov Chain, which is a sequence of random states where the probability of transitioning from one state to another depends only on the current state (Markov property). The chain is constructed in such a way that it eventually reaches a state where the distribution of interest is approximated.
- **Metropolis-Hastings Algorithm** : One of the common algorithms used in MCMC is the Metropolis-Hastings algorithm. It involves proposing new states based on the current state, accepting or rejecting the proposed state based on an acceptance criterion, and then updating the current state accordingly.
- **EMCEE** library in **Python** is used to implement Monte Carlo Markov Chain analysis for constraining cosmological parameters.

## 4.3 Chi-square minimization v/s MCMC

### 4.3.1 Chi-square Minimization:

- **Simplicity and Speed**: Chi-square minimization is relatively straightforward and computationally efficient. It is suitable for problems with simple models and well-defined parameter spaces.
- **Deterministic Optimization**: Chi-square minimization seeks to find the parameters that minimize the difference between observed and expected values. It provides a deterministic solution, and the final parameter values are chosen to best fit the data.
- **Assumptions**: Chi-square minimization assumes that the errors are normally distributed and that the model is correctly specified. It may not be robust in the presence of outliers or in cases where the model assumptions are violated.

### 4.3.2 MCMC Analysis:

- **Bayesian Inference**: MCMC is often employed in Bayesian statistical analysis, allowing for the estimation of the posterior distribution of parameters. It provides a full probabilistic description of the parameter uncertainties.
- **Dealing with Uncertainties**: MCMC naturally handles uncertainties in parameter estimation. It provides a range of plausible parameter values rather than a single-point estimate. This can be crucial when dealing with uncertain or incomplete information.
- **Incorporating Priors**: Bayesian MCMC allows the incorporation of prior information into the analysis. This is beneficial when we have prior knowledge about the parameters that we want to incorporate into the analysis.

### 4.3.3 WHY BOTH?

- **Validation and Robustness:** Using both methods allows us to cross-validate our results. If the results obtained from Chi-square minimization and MCMC analysis agree, it adds confidence to our findings.
- **Model Checking:** Different methods can help in checking the adequacy of our chosen model. Chi-square minimization might provide a best-fit solution, while MCMC allows us to explore the full parameter space and assess the credibility of different model scenarios.
- **Complementary Information:** Each method provides different types of information. Chi-square minimization gives a single best-fit solution, while MCMC provides a distribution of possible parameter values. Combining them can give a more comprehensive understanding of the problem.

## 5 Results

### 5.1 $H(z)$ constraints

#### 5.1.1 $\Lambda$ CDM model

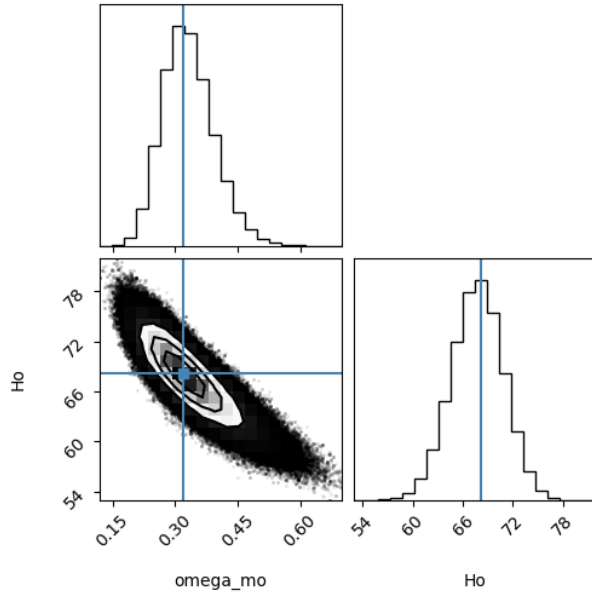
- **Flat  $\Lambda$ CDM model :** In the **spatially flat  $\Lambda$ CDM** model, the free parameters are  $\Omega_{m0}$  and  $H_0$ .

Table 1: Chi-square minimization

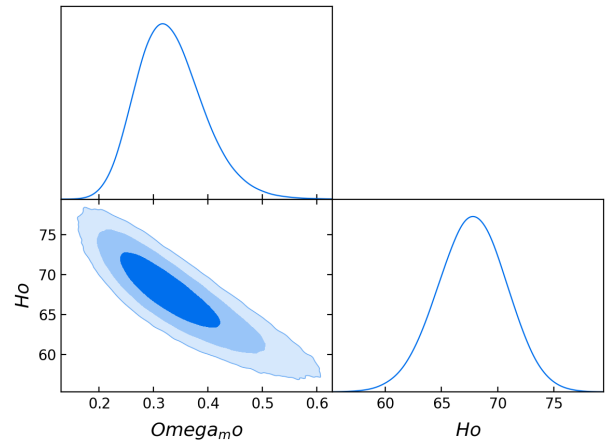
parameter	optimal value
$\Omega_{m0}$	0.3195
$H_0$	68.1499

Table 2: MCMC Analysis

parameter	optimal value
$\Omega_{m0}$	0.3324
$H_0$	67.7187



(a) Corner Plot



(b) Contour Plot

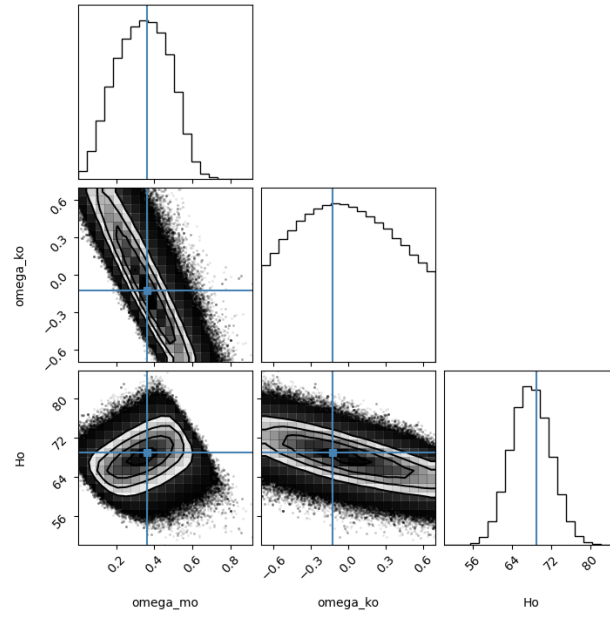
- **Non-Flat  $\Lambda$ CDM model :** In the **spatially non-flat  $\Lambda$ CDM** model, the free parameters are  $\Omega_{m0}$ ,  $\Omega_{k0}$  and  $H_0$ .

Table 3: Chi-square minimization

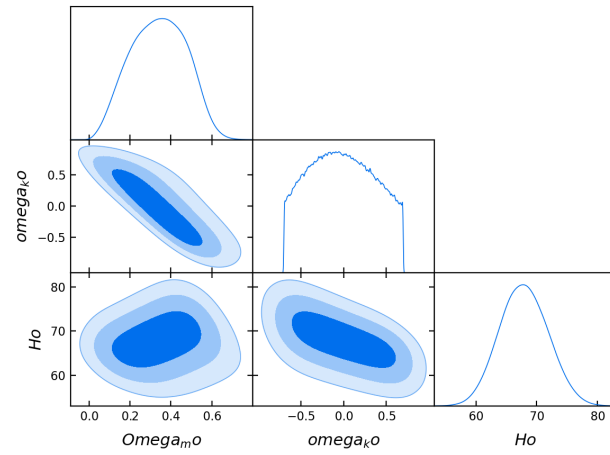
parameter	optimal value
$\Omega_{m0}$	0.3597
$\Omega_{k0}$	-0.1259
$H_0$	68.9987

Table 4: MCMC Analysis

parameter	optimal value
$\Omega_{m0}$	0.3393
$\Omega_{k0}$	-0.0162
$H_0$	67.8891



(a) Corner Plot



(b) Contour Plot



### 5.1.2 XCDM model

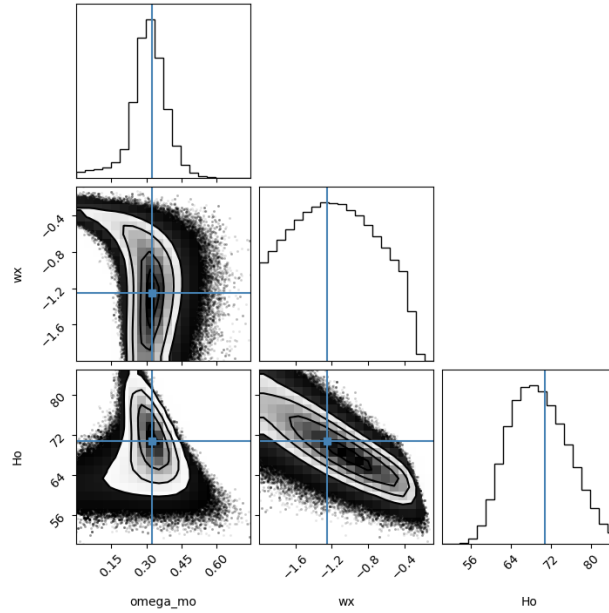
- **Flat XCDM model** : In the **spatially flat XCDM** model, the free parameters are  $\Omega_{m0}, \Omega_X$  and  $H_0$ .

Table 5: Chi-square minimization

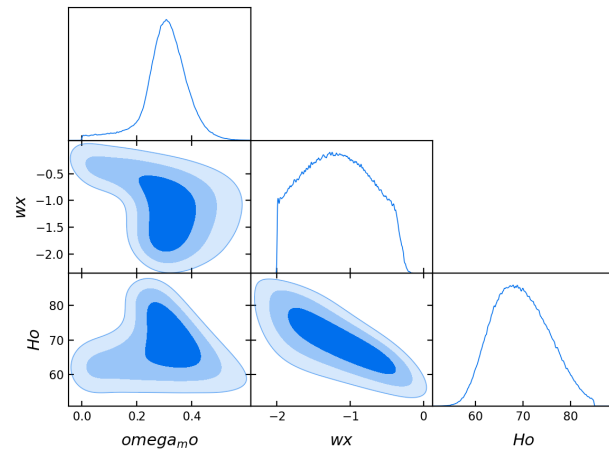
parameter	optimal value
$\Omega_{m0}$	0.3223
$\omega_X$	-1.2497
$H_0$	70.8207

Table 6: MCMC Analysis

parameter	optimal value
$\Omega_{m0}$	0.3049
$\omega_X$	-1.1726
$H_0$	69.5819



(a) Corner Plot



(b) Contour Plot

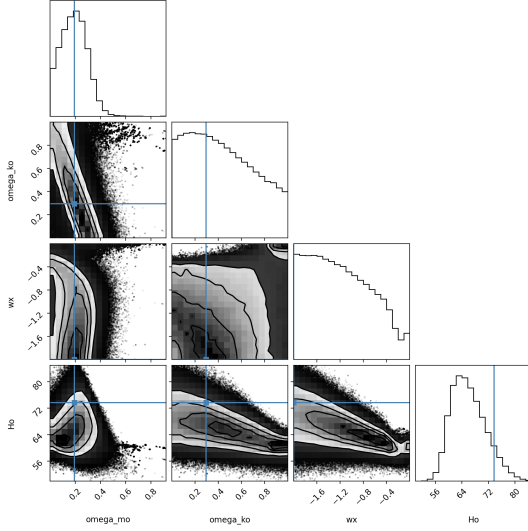
- **Non-Flat XCDM model** : In the **spatially non-flat XCDM** model, the free parameters are  $\Omega_{m0}$ ,  $\Omega_{k0}$ ,  $\omega_X$  and  $H_0$ .

Table 7: Chi-square minimization

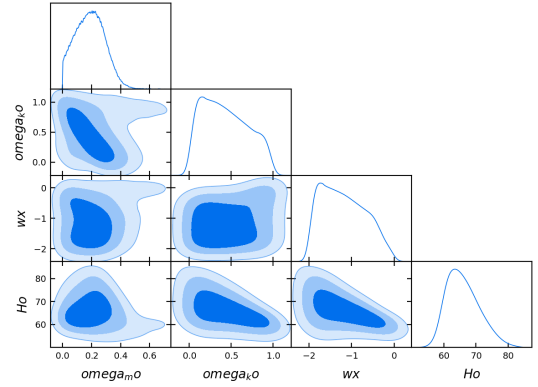
parameter	optimal value
$\Omega_{m0}$	0.1943
$\Omega_{k0}$	0.2940
$\omega_X$	-2
$H_0$	73.7042

Table 8: MCMC Analysis

parameter	optimal value
$\Omega_{m0}$	0.1920
$\Omega_{k0}$	0.4325
$\omega_X$	-1.1886
$H_0$	66.0597



(a) Corner Plot



(b) Contour Plot

## 5.2 H(z) and BAO constraints

### 5.2.1 $\Lambda$ CDM model

- **Flat  $\Lambda$ CDM model** : In the **spatially flat  $\Lambda$ CDM** model, the free parameters are  $\Omega_{m0}$  and  $H_0$ .

Table 9: Chi-square minimization

parameter	optimal value
$\Omega_{m0}$	0.0014
$H_0$	96.3655

Table 10: MCMC Analysis

parameter	optimal value
$\Omega_{m0}$	0.0014
$H_0$	75.9917

### 5.2.2 XCDM model

- **Flat XCDM model** : In the **spatially flat XCDM** model, the free parameters are  $\Omega_{m0}$ ,  $\omega_X$  and  $H_0$ .

Table 11: Chi-square minimization

parameter	optimal value
$\Omega_{m0}$	0.0014
$\omega_X$	-0.8070
$H_0$	94.4642

Table 12: MCMC Analysis

parameter	optimal value
$\Omega_{m0}$	0.0014
$\omega_X$	-0.9641
$H_0$	74.0892

Table 13: H(z) data

z	H(z)	$\sigma_H$
0.07	69	19.6
0.09	69	12
0.12	68.6	26.2
0.17	83	8
0.179	75	4
0.199	75	5
0.20	72.9	29.6
0.27	77	14
0.28	88.8	36.6
0.352	83	14
0.3802	83	13.5
0.4	95	17
0.4004	77	10.2
0.4247	87.1	11.2
0.4497	92.8	12.9
0.47	89	50
0.4783	80.9	9
0.48	97	62
0.593	104	13
0.68	92	8
0.781	105	12
0.875	125	17
0.88	90	14
0.90	117	23
1.037	154	20
1.3	168	17
1.363	160	33.6
1.43	177	18
1.53	140	14
1.75	202	40
1.965	186.5	50.4

Table 14: BAO data

z	Measurement	Value
0.38	$D_M(r_{s,fid}/r_s)$	1512.39
0.38	$H(z) (r_s/r_{s,fid})$	81.2087
0.51	$D_M(r_{s,fid}/r_s)$	1975.22
0.51	$H(z) (r_s/r_{s,fid})$	90.9029
0.61	$D_M(r_{s,fid}/r_s)$	2306.68
0.61	$H(z) (r_s/r_{s,fid})$	98.9647
0.122	$D_V(r_{s,fid}/r_s)$	$539 \pm 17$
0.81	$D_A/r_s$	$10.75 \pm 0.43$
1.52	$D_V(r_{s,fid}/r_s)$	$3843 \pm 147$
2.34	$D_H/r_s$	8.86
2.34	$D_M/r_s$	37.41

Table 15: Baryon densities for the models

Model	$\Omega_{\text{bo}} h^2$
Flat $\Lambda$ CDM	0.02225
Nonflat $\Lambda$ CDM	0.02305
Flat XCDM	0.02229
Nonflat XCDM	0.02305
Flat $\phi$ CDM	0.02221
Nonflat $\phi$ CDM	0.02303

[1, 4, 3, 2]

## References

- [1] J. Ryan(1), S. Doshi(2, 1), and B. Ratra(1), “Constraints on dark energy dynamics and spatial curvature from hubble parameter and baryon acoustic oscillation data,” 2018.
- [2] S. Cao, *Cosmological constraints from standardized non-CMB observations*. PhD thesis, Kansas State University, KS, 2023.
- [3] J. Ryan(1), Y. Chen(2), and B. Ratra(1), *Baryon acoustic oscillation, Hubble parameter, and angular size measurement constraints on the Hubble constant, dark energy dynamics, and spatial curvature*. PhD thesis, (1) Department of Physics, Kansas State University, 116 Cardwell Hall, Manhattan, KS 66502, USA (2) Key Laboratory for Computational Astrophysics, National Astronomical Observatories, Chinese Academy of Sciences, Beijing, 100012, China, 2019.
- [4] N. Khadka(1), M. L. Martínez-Aldama(2, 3), M. Zajaček(4), B. Czerny(2), and B. Ratra(1), *Do reverberation-measured  $H\beta$  quasars provide a useful test of cosmology?* PhD thesis, (1)Department of Physics, Kansas State University, 116 Cardwell Hall, Manhattan, KS 66506, USA (2)Center for Theoretical Physics, Polish Academy of Sciences, Al. Lotników 32/46, 02-668 Warsaw, Poland (3)Departamento de Astronomia, Universidad de Chile, Camino del Observatorio 1515, Santiago, Chile (4)Department of Theoretical Physics and Astrophysics, Faculty of Science, Masaryk University, Kotlářská 2, 602 00 Brno, Czech Republic, 2022.

Table 16: QSO data

Source	z	$\theta$	$\sigma_g$
P 2351-006	0.462	2.743	0.027
3C 279	0.5362	1.454	0.052
P 0252-549	0.539	1.049	0.077
P 1136-13	0.558	1.974	0.048
P 0403-13	0.574	2.399	0.077
P 0920-39	0.591	1.933	0.097
3C 345	0.5928	1.525	0.041
GC 1104 + 16	0.632	1.762	0.15
P 0113-118	0.67	2.218	0.108
P 2329-415	0.6715	1.292	0.208
GC 2251 + 13	0.673	1.972	0.11
P 2344 + 09	0.677	1.904	0.075
1803 + 78	0.68	1.244	0.129
0651 + 82	0.71	1.454	0.245
P1354 + 19	0.72	1.415	0.074
GC 1636 + 47	0.74	1.139	0.076
P 1237-10	0.752	1.518	0.058
P 2135-248	0.819	0.897	0.304
3C 179	0.846	1.953	0.144
P 0915-213	0.847	1.758	0.094
GC 1213 + 35	0.857	1.73	0.183
P 0454-46	0.858	2.094	0.049
P 1252 + 11	0.871	1.635	0.249
P 1055 + 01	0.888	1.144	0.059
P 0537-441	0.894	1.149	0.041
P 0537-158	0.947	1.747	0.087
P 2354-11	0.96	1.661	0.121
P 1933-400	0.965	1.81	0.063
GC 0237 + 04	0.978	0.568	0.426
P 2326-477	1.299	2.07	0.047
P 0448-392	1.302	1.387	0.064
P 1449-012	1.319	1.682	0.064
P 2312-319	1.323	0.528	0.135
P 1327-311	1.326	0.438	0.863
P 1615 + 29	1.339	1.608	0.117
P 1831-711	1.356	1.46	0.058
P 0522-611	1.4	1.168	0.096
P 0005-239	1.407	0.758	0.14
P 2320-035	1.41	0.596	0.171
GC 0820 + 56	1.417	0.424	0.114
GC 0805 + 41	1.42	1.481	0.306
P 1532 + 01	1.435	1.172	0.074
P 2335-18	1.446	1.81	0.168
P 1030-357	1.455	1.718	0.073
AO 0952 + 17	1.472	1.484	0.223
GC 2253 + 41	1.476	1.327	0.19
P 0524-460	1.479	1.784	0.059
P 2052-47	1.489	1.292	0.082
P 0220-349	1.49	1.243	0.059
P 2058-297	1.492	0.87	0.429
4C 46.29	1.5586	2.446	0.216
P 2227-08	1.5595	1.038	0.109
P 0406-127	1.563	0.796	0.168
P 0837 + 035	1.57	1.497	0.1
P 1104-445	1.598	1.743	0.081
P 1351 + 021	1.6077	1.788	0.034
P 0127 + 145	1.6301	1.788	0.051
P 1130 + 009	1.633	0.659	0.203
P 0208-512	0.999	1.031	0.051
P 0355-483	1.016	1.517	0.033

Table 17: QSO data

Source	z	$\theta$	$\sigma_g$
P 0906 + 01	1.018	1.794	0.081
P 0130-17	1.02	1.544	0.197
OJ 320	1.025	1.552	0.122
P 1348-289	1.034	1.797	0.163
P 2356 + 196	1.066	1.155	0.453
GC 1514 + 19	1.07	0.76	0.07
P 0122-00	1.08	1.529	0.16
GC 1144 + 40	1.088	1.399	0.258
GC 1335 + 55	1.0987	0.88	0.581
P 2303-052	1.139	1.638	0.086
P 1210 + 134	1.141	1.727	0.053
P 2329-16	1.153	1.758	0.054
P 1438-347	1.159	1.277	0.069
P 2332-017	1.184	1.485	0.048
P 1127-14	1.187	1.611	0.089
P 2329-384	1.202	1.099	0.081
P 1004-018	1.212	1.254	0.088
QC 08211 + 39	1.216	2.622	0.085
OV 591	1.22	2.032	0.147
P 1823-455	1.244	1.38	0.064
1150 + 81	1.25	2.44	0.151
VRO 40.09.02	1.252	2.109	0.111
GC 1020 + 40	1.254	1.544	0.19
GC 0537 + 53	1.275	1.046	0.381
P 0514-16	1.278	1.883	0.146
P 0405-385	1.285	0.979	0.111
OR 186	1.29	1.583	0.107
GC 0707 + 47	1.292	0.517	0.77
P 0511-220	1.296	1.2	0.187
P 0229-398	1.646	1.431	0.093
NRAO 512	1.66	1.245	0.214
P 0922 + 005	1.72	1.005	0.108
P 0202-17	1.74	1.464	0.124
P 1148-171	1.751	1.602	0.205
DW 1403-08	1.763	1.629	0.139
P 2320-021	1.774	1.25	0.191
P 0108-079	1.776	0.66	0.15
P 1451-400	1.81	1.136	0.047
P 1034-374	1.821	0.518	0.082
P 0805-07	1.837	0.85	0.495
0633 + 73	1.85	1.635	0.307
OK 492	1.873	1.13	0.286
OP - 192	1.89	1.537	0.059
OH - 230	1.9	1.396	0.194
GC 1656 + 34	1.939	1.526	0.163
P 0048-071	1.975	1.207	0.094
GC 0119 + 24	2.025	1.7	0.279
OF 036	2.048	1.901	0.148
P0226-038	2.055	1.595	0.057
GC 1325 + 43	2.073	1.422	0.368
P 2319 + 07	2.09	1.916	0.132
P 1116 + 12	2.118	0.564	0.664
P 2145-17	2.13	2.264	0.048
P 1020 + 191	2.136	1.449	0.077
P 0642-349	2.165	1.097	0.223
P 1032-199	2.198	1.434	0.036
P 2314-409	2.448	1.676	0.045
GC 1337 + 63	2.5584	1.598	0.323
P 0329-255	2.685	1.506	0.101
P 0136 + 176	2.73	1.252	0.132