A more accurate Parameterization based on cosmic Age (MAPAge)

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Abstract Recently, several statistically significant tensions between different cosmological datasets have raised doubts about the standard Lambda cold dark matter (ΛCDM) model. A recent letter (Huang 2020) suggests to use “Parameterization based on cosmic Age” (PAge) to approximate a broad class of beyond-ΛCDM models, with a typical accuracy ∼ 1% in angular diameter distances at z ≲ 10. In this work, we extend PAge to a More Accurate Parameterization based on cosmic Age (MAPAge) by adding a new degree of freedom η2. The parameter η2 describes the difference between physically motivated models and their phenomenological PAge approximations. The accuracy of MAPAge, typically of order 10^{-3} in angular diameter distances at z ≲ 10, is significantly better than PAge. We compare PAge and MAPAge with current observational data and forecast data. The conjecture in Huang (2020), that PAge approximation is sufficiently good for current observations, is quantitatively confirmed in this work. We also show that the extension from PAge to MAPAge is important for future observations, which typically require sub-percent accuracy in theoretical predictions.

Key words: cosmology — cosmological parameters — observations

1 INTRODUCTION

At the end of the last century, the observed extra dimming of Type Ia supernovae (SNe) led to the introduction of dark energy into standard cosmology (Perlmutter et al. 1999; Riess et al. 1998; Scolnic et al. 2018). In the concordance Λ cold dark matter (ΛCDM) model, dark energy is interpreted as the cosmological constant invented by Albert Einstein. Despite the great observational success of ΛCDM model in the last two decades (Planck Collaboration et al. 2020; Alam et al. 2021; DES Collaboration et al. 2021), the fine tuning problem and the coincidence problem of the cosmological constant remain to be a pain in the neck for at least some, if not all, of the theoreticians (Weinberg 1989; Zlatev et al. 1999). The concordance model was further challenged by the recently emerged observational discrepancy between directly and indirectly measured Hubble constants (Riess et al. 2021; Wong et al. 2020) and a ∼ 3σ tension between the measurements of the matter clustering (Asgari et al. 2021). The theoretical unnaturalness and observational tensions have motivated a plethora of beyond-ΛCDM models (Caldwell et al. 1998; Zlatev et al. 1999; Chiba et al. 2000; Armendariz-Picon et al. 2000; Dvali et al. 2000; Kamenshchik et al. 2001; Bento et al. 2002; Capozziello et al. 2003; Amendola 2000; Huang 2016), although in the Bayesian view none of them has been proven to be significantly more competitive than ΛCDM. The zoology of the models, namely the exercise of computing Bayesian evidence for all the models, if ever possible, may not be a pleasant job. Preferred choices are often more phenomenological models such as a perfect-fluid dark energy with its equation of state being a constant (wCDM model) or a linear function of the scale factor (w0−wa model) (Chevallier & Polarski 2001; Linder 2003). More blind methods such as Taylor expansion (Visser 2004) and Gaussian process (Shafieloo et al. 2012) are sometimes used to explore more complex scenarios.

The Parameterization based on the cosmic Age (PAge), recently proposed by Huang (2020), is somewhat in between. It is a semi-blind model capturing common physical features of many models, such as matter dominance at high redshift and that the energy density of the universe decreases with time. In the Bayesian view, PAge is more economic than many other bottom-up methods in the sense that it only contains one more parameter than ΛCDM. Physical models can be mapped to PAge space either by simply matching the deceleration parameter at some pivot redshift or by doing a least square fitting of cosmological observables. It has been shown that
Relative errors of distance moduli are typically \( \lesssim 1\% \) at \( z \lesssim 10 \) (Luo et al. 2020; Huang et al. 2021). Empirically such an accuracy is good enough for utilizing current cosmological data at \( z \lesssim 10 \). Indeed, PAge has been applied to many currently available data sets and yielded fruitful results (Huang 2020; Luo et al. 2020; Huang et al. 2021; Cai et al. 2021).

Tiny deviation from PAge, however, may become measurable with future cosmological surveys. For instance, the ongoing ground-based project of Dark Energy Spectroscopic Instrument (DESI) will measure angular diameter distance and Hubble parameter to percent-level accuracy up to redshift \( z \sim 1.5 \) (DESI Collaboration et al. 2016). The precision of measurements is expected to be further improved by the space missions of Euclid satellite and the Chinese Space Station Telescope Optical Survey in the near future (Amendola et al. 2018; Gong et al. 2019).

Thus, aiming at future cosmological surveys, we extend PAge to a “More Accurate Parameterization based on the cosmic Age” (MAPAge) by adding a new degree of freedom into PAge, of which the ansatz will be given immediately below in Section 2. We show in Section 3 that the new degree of freedom in MAPAge is not well constrained by current observations, thus explicitly confirming that PAge is accurate enough for current data. In Section 4, we forecast the constraint on MAPAge parameters with simulated data of baryon acoustic oscillations (BAO) from DESI. We conclude and discuss in Section 5.

### 2 MODEL

In PAge approximation, the expansion rate of the Universe is parameterized as (Huang 2020)

\[
\frac{H}{H_0} = 1 + \frac{2}{3} \left( 1 - \eta \right) \left( \frac{1}{H_0 t} - 1 \right),
\]

where \( t \) is the cosmological time and \( H \) is the Hubble parameter. Here \( H_0 \approx 100h \text{ km s}^{-1} \text{Mpc}^{-1} \) is the Hubble constant. The age parameter \( p_{age} \equiv H_0 t_0 \) is the present cosmological time \( t_0 \) expressed in unit of \( H_0^{-1} \). The phenomenological parameter \( \eta (\eta < 1) \) characterizes the deviation from Einstein de-Sitter universe (flat CDM model).

In numeric calculation, the correspondence between the cosmological time \( t \) and the redshift \( z \) is obtained by inverting the monotonic function

\[
z(t) = e^{-\int_0^t H(t') dt'} - 1.
\]

It can be trivially shown that \( H(z) \) is automatically guaranteed to be a monotonically increasing function. At high redshift \( z \gg 1 \), PAge asymptotically approaches the matter-dominated behavior \( a \propto t^{2/3} \), where \( a = \frac{1}{1+z} \) is the scale factor. (As a late-universe phenomenological approximation, PAge ignores the very short period of radiation-dominated era.) These key features makes PAge a very compact approximate description of many physical models.

The More Accurate version of PAge, MAPAge, is formulated to preserve the aforementioned advantages,

\[
\frac{H}{H_0} = 1 + \frac{2}{3} \left( 1 - (\eta + \eta_2) \frac{H_0 t}{p_{age}} + \eta_2 \left( \frac{H_0 t}{p_{age}} \right)^2 \right) \times \left( \frac{1}{H_0 t} - 1 \right),
\]

### Table 1 MAPAge Approximation: Maximum Relative Errors in \( D_A(z) \) (0 ≤ \( z \) ≤ 10)

| Models    | Parameters          | \( p_{age} \) | \( \eta \) | \( \eta_2 \) | max \( \frac{\Delta D_A}{D_A} \) |
|-----------|---------------------|---------------|-----------|-------------|----------------------------------|
| CDM       | \( \Omega_m = 1 \)  | \( \frac{1}{5} \) | 0                     | 0                       | 0.000                            |
| nonflat CDM | \( \Omega_m = 0.3, \Omega_k = 0.7 \) | 0.809 | -0.0389 | -0.373 | 3.71 \times 10^{-3} |
| \( \Lambda \)CDM | \( \Omega_m = 0.3 \) | 0.964 | 0.377 | 0.0752 | 1.53 \times 10^{-4} |
| nonflat \( \Lambda \)CDM | \( \Omega_m = 0.5, \Omega_k = 0.2 \) | 0.797 | 0.113 | -0.0596 | 7.52 \times 10^{-4} |
| wCDM      | \( \Omega_m = 0.3, w = -1.2 \) | 0.991 | 0.664 | -0.148 | 5.77 \times 10^{-4} |
| nonflat \( w \)CDM       | \( \Omega_m = 0.33, \Omega_k = -0.25, w = -0.8 \) | 0.967 | 0.215 | 0.411 | 1.85 \times 10^{-3} |
| \( w_0 - w_0 \)CDM      | \( \Omega_m = 0.3, w_0 = -1, w_a = 0.3 \) | 0.953 | 0.372 | -0.0131 | 4.69 \times 10^{-4} |
| nonflat \( w_0 - w_0 \)CDM | \( \Omega_m = 0.25, \Omega_k = 0.1, w_0 = -1.2, w_a = -0.2 \) | 1.009 | 0.629 | -0.197 | 1.78 \times 10^{-3} |
| GCG       | \( \Omega_b = 0.05, A = 0.75, \alpha = 0.1 \) | 0.956 | 0.426 | 0.0386 | 5.66 \times 10^{-4} |
| DGP       | \( \Omega_m = 0.3 \) | 0.907 | 0.148 | 0.0252 | 4.20 \times 10^{-5} |

**Fig. 1** PAge approximation and MAPAge approximation: relative errors of angular diameter distance. Model parameters are given in Table 1.
where the new degree of freedom $\eta_2$ ($-1 < \eta_2 < 1$) can be regarded as a cubic-order correction to PAge approximation. When $\eta_2 = 0$, MAPAge degrades to PAge.

The cosmological models listed in Luo et al. (2020) and Huang et al. (2021) can be approximately mapped into MAPAge space with better precision. A least-square fitting of the dimensionless quantity $H_t$ (as a function of redshift) is applied in the redshift range $0 \leq z \leq 10$ to determine the phenomenological parameters $\eta$ and $\eta_2$. Table 1 shows the fitting accuracy of angular diameter distances $D_A$. Comparing these results with the fitting accuracy of PAge, given in Luo et al. (2020) and Huang et al. (2021), we find that the precision of MAPAge approximation is typically ~ an order of magnitude better than PAge. Such a comparison is not entirely fair, however, because in Luo et al. (2020) and Huang et al. (2021) the $\eta$ parameter of PAge is differently determined, by matching the deceleration parameter $q_0$ at redshift zero. To obtain a more fair comparison, we apply the same least square fitting method to PAge. The $D_A$ fitting errors of PAge and MAPAge for a few models are shown in Figure 1. The result again confirms MAPAge’s superiority in fitting accuracy.

### 3 CURRENT CONSTRAINTS

To study the observational constraint on MAPAge parameters, we compile the BAO data sets from 6dF Galaxy Survey and Sloan Digital Sky Survey (SDSS) (Beutler et al. 2011; Ross et al. 2015; Alam et al. 2017; du Mas des Bourboux et al. 2017; Ata et al. 2018) and the Pantheon SNe data (Scolnic et al. 2018). A $\chi^2$-square likelihood is applied to both data sets. The collected data points and covariance matrices can be found in Luo et al. (2020) and Scolnic et al. (2018), respectively. We perform Monte Carlo Markov Chain (MCMC) calculation for both PAge and MAPAge. Uniform priors are applied on $h r_d \in [0, 200 \text{Mpc}]$, $p_{\text{age}} \in [0.8, 1.2]$, $\eta \in [-2, 1]$, and $\eta_2 \in [-1, 1]$.

We present the marginalized constraints (mean and 68% confidence-level bounds) on PAge and MAPAge parameters in Table 2. The marginalized $1\sigma$, $2\sigma$ contours

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**Table 2** Marginalized Constraints on Page and MAPAge Parameters

| Models        | Data                  | $h r_d$/Mpc | $p_{\text{age}}$ | $\eta$  | $\eta_2$ | $\chi^2_{\text{min}}$/d.o.f. |
|---------------|-----------------------|-------------|------------------|---------|----------|-----------------------------|
| PAge          | current BAO+SNe       | 101.2 ± 1.2 | 0.975 ± 0.013    | 0.372 ± 0.065 | -        | 0.987                       |
| MAPAge        | current BAO+SNe       | 102.0 ± 1.3 | 0.968 ± 0.014    | 0.50 ± 0.12   | -0.32 ± 0.24 | 0.985                       |
| MAPAge        | BAO forecast          | 100.4 ± 0.74| 0.960 ± 0.0079   | 0.394 ± 0.044 | 0.04 ± 0.12 | 0.00284                     |

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**Fig. 2** The marginalized constraints on PAge and MAPAge parameters. The inner and outer contours enclose 68% and 95% confidence regions, respectively.
for both PAge and MAPAge parameters are shown in Figure 2 for a comparison. The apparent worsen
measurement of $\eta$ in MAPAge is due to the strong
degeneracy between $\eta$ and $\eta_2$, as shown by the red $\eta$-$\eta_2$ contour in Figure 2. The constraints on $h r_d$ and
$\rho_{\text{age}}$ parameters, different from $\eta$, are not significantly
influenced by the inclusion of $\eta_2$. Here we see that the
coincidental proximity $\rho_{\text{age}} \approx 1$, which has inspired some
recent discussion about whether we are living in a special
cosmic era (Avelino & Kirshner 2016), is a rather robust
result that does not rely on the $\Lambda$CDM framework.

The strong degeneracy between $\eta$ and $\eta_2$ suggests that
the data cannot distinguish MAPAge and PAge very well.
It is therefore unnecessary to use MAPAge for current
data, in the spirit of Occam’s razor principle. As an easy
and efficient parameterization, PAge approximation still
serves as a sufficiently powerful tool to investigate late-
time cosmological expansion history beyond the $\Lambda$CDM
physics, such as the late-time cosmic acceleration (Huang
2020; Luo et al. 2020), the statistically significant $H_0$
crisis (Riess et al. 2021; Wong et al. 2020) and $S_8$
tension (Asgari et al. 2021). However, with the rapid
accumulation of cosmological data, the situation might
change within a few years. In the next section we proceed
to explore the future prospects of MAPAge.

4 FORECAST

DESI is an ongoing Stage IV ground-based dark energy
experiment. It is designed to study BAO and the growth
of structure with a wide-area galaxy and quasar redshift
survey. DESI experiment aims to provide more precise
and higher quality observation data, at least an order of
magnitude improvement over SDSS both in the comoving
volume and the number of galaxies (DESI Collaboration
et al. 2016). These more precise data can greatly advance
our understanding of the cosmic evolution history and the
nature of cosmic dark components.

We adopt the DESI BAO forecast data from different
tracers across the redshift range $[0, 3.5]$ and covering
14,000 square degrees of the sky. The radial and
transverse BAO forecast are listed in table 2.3, table 2.5
and table 2.7 of DESI Collaboration et al. (2016),
respectively. The baseline cosmological model employed
in DESI forecast has been mentioned in section 2.4.1
of DESI Collaboration et al. (2016), whose fiducial
values are detailedly summarized in table 5 of Planck
Collaboration et al. (2014). This fiducial flat-$\Lambda$CDM

Fig. 3 Marginalized constraints on MAPAge parameters with current BAO+SNe and forecast DESI BAO data. The grey
dashed horizontal and vertical lines show the mapping values of MAPAge which correspond to the fiducial $\Lambda$CDM
cosmology applied in forecast.
cosmology approximately corresponds to \( \mathcal{H}_r = 100.13 \), \( \rho_{\text{Page}} = 0.957 \), \( \eta = 0.370 \) and \( \eta_2 = 0.072 \) in MAP\text{Age} language. We perform MCMC analysis with a \( \chi^2 \)-square likelihood. The last row of Table 2 lists the marginalized posteriors of MAP\text{Age} parameters, which are consistent with the mapping results of MAP\text{Age} derived from the fiducial cosmology.

In Figure 3 we compare the constraints on MAP\text{Age} parameters with current BAO+SNe data and with forecast DESI BAO data. The apparent shrinking of the parameter by parameter contours from now (BAO + SNe) to future (DESI) suggests that MAP\text{Age} will become more and more favorable as the clock ticks.

5 CONCLUSIONS

Among many phenomenological late-universe expansion reconstruction methods, PA\text{ge} has been proven to be a very compact and robust one. In this work, we extended the PA\text{ge} approximation to a more precise framework (MAP\text{Age}) to embrace the forthcoming era of hyper-precision cosmology. MAP\text{Age} approximation inherits all the advantages of PA\text{ge} and can approximate the expansion history of many models to \( \sim 10^{-3} \) level. We took the ongoing Stage IV experiment DESI as an example and showed the superiority of MAP\text{Age} for cosmological surveys in the near future.

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