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# Late-Time Constraints on Future Singularity Dark Energy Models from Geometry and Growth

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Article

# Late-Time Constraints on Future Singularity Dark Energy Models from Geometry and Growth

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## Abstract

We confront two future-singularity dark-energy templates—sudden future singularities (SFSs) and finite scale factor singularities (FSFSs)—with late-time geometric probes and redshift-space distortion growth data. We compute the observable growth  $f\sigma_8(z)$  by solving the full linear perturbation system (including the standard fiducial cosmology rescaling of RSD measurements) and build a joint  $\chi^2$  from Pantheon+SH0ES SNe Ia,  $H(z)$ , DESI AP-only BAO, and  $f\sigma_8$ . Parameter constraints are obtained via grid-based profiling over nuisance parameters and the singularity time location parameter. We compare the viability and goodness of fit of the singularity scenarios to the  $\Lambda$ CDM reference.

**Keywords:** large-scale structure; growth of perturbations; dark energy; dynamical dark energy; cosmological singularities; DESI; Euclid; WFIRST

## 1. Introduction

Late-time cosmology is currently constrained primarily by *geometrical* probes, which map the background expansion through integrated distance–redshift relations. These include Type Ia supernovae, baryon acoustic oscillations, and direct measurements of  $H(z)$  from cosmic chronometers and related techniques [1–12]. While these data tightly constrain the expansion history, they are intrinsically limited in their ability to identify the physical origin of cosmic acceleration: within general relativity, many distinct dark-energy histories can reproduce nearly identical background distances.

A complementary and more discriminating handle is provided by *dynamical* information, i.e., the growth of cosmic structure, commonly encoded in the observable  $f\sigma_8(z)$  inferred from redshift-space distortions (RSD) in galaxy surveys [13–17]. Even though current growth data are typically less constraining than geometric probes when taken alone, their combination can substantially tighten constraints on non-standard scenarios and can test whether a given background model yields an acceptable perturbation history. This is particularly relevant for models that mimic  $\Lambda$ CDM at the background level but predict a modified growth rate, including both dynamical dark energy and modifications of gravity [18–21].

Among phenomenological late-time departures from  $\Lambda$ CDM within general relativity, an interesting class is provided by effective dark-energy models whose evolution can be associated with finite-time future singularities. These include the Sudden Future Singularity (SFS, Type II) and the Finite Scale Factor Singularity (FSFS, FSF, Type III) scenarios [22–24]. In this work we treat such models as late-time effective descriptions and assess them empirically: we do not assume that a singularity must be realized in nature, but instead ask whether the corresponding background-plus-perturbation evolution can simultaneously fit current late-time data.



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Motivated by this goal, we perform a joint, late-Universe comparison of SFS and FSFS models against a reference  $\Lambda$ CDM cosmology using a unified likelihood that combines geometric and growth information. On the geometric side we employ Type Ia supernovae,  $H(z)$  measurements, and an Alcock–Paczyński-type BAO observable  $F_{\text{AP}}(z) = D_M(z)H(z)/c$  (here implemented using recent DESI measurements). DESI BAO-only constraints are compatible with a flat  $\Lambda$ CDM expansion history. At the same time, DESI analyses in the  $w_0w_a$ CDM extension and in combination with CMB and/or supernova data report a statistically significant preference for evolving dark energy, whose exact significance depends on the adopted data combination [25]. On the growth side we use a compilation of  $f\sigma_8(z)$  data from RSD and account for the standard geometrical rescaling associated with the fiducial cosmology adopted in each measurement [13,14,26]. To ensure numerical robustness in the presence of non-physical regions of parameter space (e.g., backgrounds leading to integration failure), we adopt a grid-based exploration with profile-likelihood minimization over nuisance parameters. In particular, we profile analytically over the supernova magnitude offset and over the clustering amplitude  $\sigma_8$ , optionally including a Gaussian prior on  $\sigma_8$ .

The paper is organized as follows: In Section 3 we define the SFS and FSFS background parameterization and summarize the mapping to the late-time observables  $H(z)$  and  $D_A(z)$ . In Section 2 we present the linear perturbation system used to compute the scale-dependent growth and the prediction for  $f\sigma_8(z)$  at a fixed wavenumber. In Section 4 we describe the data sets and the construction of the total (profiled)  $\chi^2$ , including the fiducial cosmology rescaling of RSD points and the treatment of covariances. Our results—best-fit regions, goodness of fit, and a comparison to  $\Lambda$ CDM—are presented in Section 5. We conclude in Section 6 with a discussion of the implications and limitations of this late-time analysis.

## 2. Linear Perturbations and Growth of Structure

The growth of matter perturbations provides a dynamical probe complementary to purely geometrical observables. In general relativity, the exact linear perturbation equations predict a scale dependence of the growth rate on sufficiently large scales, even for  $\Lambda$ CDM, once the sub-Hubble approximation is relaxed [27–29].

This effect becomes particularly relevant for dynamical dark energy models, including future singularity scenarios, where dark energy perturbations may not be negligible and can couple nontrivially to matter perturbations at late times [30]. In this work we therefore solve the full set of linear perturbation equations without invoking the sub-Hubble or quasi-static approximations.

We adopt the Newtonian gauge for scalar perturbations and assume negligible anisotropic stress, so that the perturbed line element can be written as

$$ds^2 = -(1 + 2\Phi) dt^2 + (1 - 2\Phi) a^2 \gamma_{ij} dx^i dx^j, \quad (1)$$

where  $\gamma_{ij}$  denotes the spatial metric and  $\Phi$  is the Newtonian potential. For a spatially flat universe filled with pressureless, nonrelativistic dark matter with density  $\rho_m$  and an exotic dark energy component with density  $\rho_{de}$ , the background dynamics are described by the Friedmann and continuity equations,

$$H^2 = \frac{8\pi G}{3} (\rho_m + \rho_{de}), \quad (2)$$

$$\dot{\rho} = -3H(\rho + p), \quad (3)$$

where  $\rho = \rho_m + \rho_{de}$  and  $p$  is the total pressure.

Linearizing Einstein’s equations around this background in the Newtonian gauge yields, after some algebra, the following evolution equations for the potential  $\Phi$ , the matter density contrast  $\delta \equiv \delta\rho_m/\rho_m$ , and the velocity potential of dark matter (encoded in  $v_f \equiv -va$ ):

$$\ddot{\Phi} = -4H\dot{\Phi} + 8\pi G\rho_{de} w_{de} \Phi, \tag{4}$$

$$\dot{\delta} = 3\dot{\Phi} + \frac{k^2}{a^2} v_f, \tag{5}$$

$$\dot{v}_f = -\Phi, \tag{6}$$

subject to the constraint equations

$$3H(H\Phi + \dot{\Phi}) + \frac{k^2}{a^2} \Phi = -4\pi G \delta\rho_m, \tag{7}$$

$$(H\Phi + \dot{\Phi}) = -4\pi G \rho_m v_f. \tag{8}$$

Here,  $k$  denotes the comoving wavenumber of the perturbation mode.

If one assumes that the relevant modes are well inside the Hubble radius,  $k^2/a^2 \gg H^2$ , and that the gravitational potential varies slowly in time, the system above reduces to the familiar scale-independent evolution equation

$$\ddot{\delta}_m + 2H\dot{\delta}_m - 4\pi G\rho_m\delta_m = 0, \tag{9}$$

for  $\delta_m$ . Relaxing only the sub-Hubble assumption, while still keeping the potential quasi-static, leads instead to a scale-dependent growth equation [28,29],

$$\ddot{\delta}_m + 2H\dot{\delta}_m - \frac{4\pi G\rho_m\delta_m}{1 + \zeta(a, k)} = 0, \tag{10}$$

where the function

$$\zeta(a, k) = \frac{3a^2 H(a)^2}{k^2} \tag{11}$$

encodes the departure from the sub-Hubble limit. In the present analysis we do not employ any of the above approximations. All growth predictions are obtained by numerically solving the complete system of linear perturbation Equations (4)–(6) together with the background evolution specified in Section 3. The observable  $f\sigma_8(z)$  used in the likelihood analysis is constructed directly from these numerical solutions.

### 3. FSFS and SFS as Dynamical Dark Energy Candidates

The SFS and FSFS show up within the framework of the Einstein–Friedmann cosmology governed by the standard field equations, Equation (2), and the energy-momentum conservation law, Equation (3). We obtain the SFS and FSFS scenarios with the scale factor in the following form:

$$a(t) = a_s \left[ \delta + (1 - \delta) \left( \frac{t}{t_s} \right)^m - \delta \left( 1 - \frac{t}{t_s} \right)^n \right]. \tag{12}$$

This scale factor ansatz fully specifies the background evolution and will be used consistently throughout the data analysis described in Section 4. In this work we fix the exponent  $m$  to the matter-dominated value

$$m = \frac{2}{3}. \tag{13}$$

This choice ensures that at early times ( $t \ll t_s$ ) the scale factor reduces to the standard Einstein–de Sitter behavior  $a(t) \propto t^{2/3}$  associated with pressureless matter, i.e., the ansatz reproduces a conventional pre-acceleration expansion history. Since our goal is a late-time, phenomenological test of singularity-inspired departures from  $\Lambda$ CDM, and the data employed here constrain primarily the late Universe, we keep  $m$  fixed and constrain the remaining parameters that control the late-time deviation and the approach to the exotic singularity (notably  $n, \delta$ , and the time location parameter  $y_0$  introduced below /in Section 4). In general, an appropriate choice of the constants ( $b, t_s, a_s, m, n$ ) is necessary [22,31–33]. For both cases, the SFS as well as the FSFS model, the evolution starts with the standard big-bang singularity at  $t = 0$  ( $a = 0$ ), and evolves to an exotic singularity for  $t = t_s$ , where  $a = a_s \equiv a(t_s)$  is a constant [32–34]. Accelerated expansion in an SFS universe is assured with a negative  $b$ , while for an FSFS universe  $b$  has to be positive. In order to have the SFS,  $n$  has to be within the range  $1 < n < 2$ ; while for an FSFS,  $n$  has to obey the condition  $0 < n < 1$ . For the SFS at  $t = t_s, a \rightarrow a_s, q \rightarrow q_s = \text{const.}, p \rightarrow \infty$ ; while for an FSFS the energy density  $\rho$  also diverges and one has: for  $t \rightarrow t_s, a \rightarrow a_s, \rho \rightarrow \infty$ , and  $p \rightarrow \infty$ , where  $a_s, t_s$  are constants and  $a_s \neq 0$ . In both scenarios the non-relativistic matter scales as  $a^{-3}$ , i.e.,

$$\rho_m = \Omega_{m,0} \rho_0 \left(\frac{a_0}{a}\right)^3, \tag{14}$$

and the evolution of the exotic (dark energy) fluid  $\rho_{de}$  can be determined by taking the difference between the total energy density  $\rho$  and the energy density of the non-relativistic matter, i.e.,

$$\rho_{de} = \rho - \rho_m. \tag{15}$$

In those scenarios the dark energy component is also responsible for the exotic singularity at  $t \rightarrow t_s$ . The dimensionless energy densities are defined in a standard way as

$$\Omega_m = \frac{\rho_m}{\rho}, \quad \Omega_{de} = \frac{\rho_{de}}{\rho}. \tag{16}$$

For the dimensionless exotic dark energy density we have the following expression:

$$\Omega_{de} = 1 - \Omega_{m,0} \frac{H_0^2}{H^2(t)} \left(\frac{a_0}{a(t)}\right)^3 = 1 - \Omega_m. \tag{17}$$

The barotropic index of the equation of state for the dark energy is defined as

$$w_{de} = p_{de} / \rho_{de}. \tag{18}$$

The singularity scenarios considered in this work were also tested as candidates for dynamical fine structure cosmology [35]. In that approach, the dark energy is sourced by a scalar field that couples to the electromagnetic sector of the theory; the knowledge about the effective evolution of the dark energy density and the dark energy equation of state evolution is sufficient to estimate the resulting fine structure evolution.

#### 4. Data Sets and Likelihood Construction

We confront the SFS and FSF singularity scenarios with four classes of late-time probes: (i) the Pantheon + SH0ES Type Ia supernova compilation (distance moduli with the full STAT+SYS covariance) [36–38], (ii) observational Hubble data (OHD)  $H(z)$  measurements [39–45], and (iii) DESI DR2 BAO constraints in the Alcock–Paczynski (AP) combination only [25,46,47], and (iv) redshift-space distortion (RSD) growth measurements in the form of a compilation of  $f\sigma_8(z)$  points with a non-diagonal covariance matrix [13,14,16,48,49]<sup>1</sup>. For reference we also evaluate the corresponding likelihood in  $\Lambda$ CDM.

Throughout, we fix  $H_0 = 67.36 \text{ km s}^{-1} \text{ Mpc}^{-1}$  and  $\Omega_{m,0} = 0.3$  and treat the parameters of the singularity ansatz (including the time location parameter  $y_0$ ) as the quantities to be constrained. We also keep a fixed effective comoving wavenumber  $k_{\text{eff}} = 0.1$  (in the same convention as used in the perturbation equations below).

#### 4.1. Background Evolution and Geometric Functions

Using the background evolution defined in Section 3, we construct all geometrical and dynamical observables entering the likelihood.

For numerical stability and consistency across all data sets, the functions  $t(z)$  and the dimensionless comoving distance  $r(z)$  are obtained by solving the coupled ODE system

$$\frac{dt}{dz} = -\frac{1}{(1+z)H(t(z))}, \quad \frac{dr}{dz} = \frac{H_0}{H(t(z))}, \quad t(0) = t_0, \quad r(0) = 0, \quad (19)$$

from which we construct

$$E(z) \equiv \frac{H(z)}{H_0}, \quad D_A(z) = \frac{r(z)}{1+z}, \quad D_M(z) = r(z). \quad (20)$$

#### 4.2. Pantheon + SH0ES SNe Ia with Analytic Profiling over the Absolute Magnitude

For Pantheon + SH0ES [36–38], we compute the theoretical distance modulus vector

$$\mu_{\text{th}}(z) = 5 \log_{10} \left[ \frac{D_L(z)}{\text{Mpc}} \right] + 25, \quad D_L(z) = \frac{c}{H_0} r(z) (1 + z_{\text{hel}}), \quad (21)$$

evaluated at the CMB-frame redshifts  $z_{\text{cmb}}$  (with the standard  $(1 + z_{\text{hel}})$  factor for the luminosity distance). We use the published full covariance matrix  $\mathbf{C}_{\text{SN}}$  (STAT + SYS).

Rather than fitting the absolute magnitude nuisance parameter explicitly, we profile over it analytically. Writing  $\mathbf{1}$  for the all-ones vector and  $\Delta = \mu_{\text{obs}} - \mu_{\text{th}}$ , the profiled supernova chi-square is implemented as

$$\chi_{\text{SN}}^2 = \Delta^T \mathbf{C}_{\text{SN}}^{-1} \Delta - \frac{(\Delta^T \mathbf{C}_{\text{SN}}^{-1} \mathbf{1})^2}{\mathbf{1}^T \mathbf{C}_{\text{SN}}^{-1} \mathbf{1}} + \ln \left[ \frac{\mathbf{1}^T \mathbf{C}_{\text{SN}}^{-1} \mathbf{1}}{2\pi} \right], \quad (22)$$

which is the quantity minimized in our scan.

#### 4.3. Observational Hubble Data

For the OHD compilation we use a standard Gaussian likelihood with diagonal errors,

$$\chi_H^2 = \sum_i \left[ \frac{H_{\text{obs}}(z_i) - H_{\text{th}}(z_i)}{\sigma_{H,i}} \right]^2, \quad (23)$$

where  $H_{\text{th}}(z_i) = H(t(z_i))$  is computed from the ODE solution for  $t(z)$ .

#### 4.4. DESI DR2 BAO: AP-Only Constraints

From DESI DR2, we employ only the AP combination [46]

$$F_{\text{AP}}(z) = \frac{D_M(z) H(z)}{c}, \quad (24)$$

for the redshift bins where AP information is provided. We build the covariance matrix using the quoted uncertainties and correlation coefficients and evaluate

$$\chi_{\text{BAO,AP}}^2 = (\mathbf{F}_{\text{AP,obs}} - \mathbf{F}_{\text{AP,th}})^T \mathbf{C}_{\text{AP}}^{-1} (\mathbf{F}_{\text{AP,obs}} - \mathbf{F}_{\text{AP,th}}). \quad (25)$$

#### 4.5. Growth Data $f\sigma_8(z)$ with Geometric Rescaling and Profiled $\sigma_8$

We use a compilation of RSD measurements [48,49] in the form of  $f\sigma_8(z)$  with a covariance matrix that includes non-diagonal blocks for selected sub-samples (WiggleZ and SDSS-IV), while the remaining points are treated as uncorrelated.

Following standard practice, each measurement is corrected for the fiducial cosmology assumed in the original analysis [47]. Denoting by  $E_{\text{fid}}(z)$  and  $D_{A,\text{fid}}(z)$  the dimensionless Hubble parameter and angular diameter distance in the fiducial cosmology (parameterized by  $\Omega_{m,0}^{\text{fid}}$  as provided with each point), the prediction entering the likelihood is rescaled as

$$[f\sigma_8(z)]_{\text{pred}} = \frac{E_{\text{fid}}(z) D_{A,\text{fid}}(z)}{E(z) D_A(z)} [f\sigma_8(z)]_{\text{th}}. \tag{26}$$

For the fiducial functions we use

$$E_{\text{fid}}(z) = \sqrt{\Omega_{m,0}^{\text{fid}}(1+z)^3 + (1 - \Omega_{m,0}^{\text{fid}})}, \quad D_{A,\text{fid}}(z) = \frac{1}{1+z} \int_0^z \frac{dz'}{E_{\text{fid}}(z')}. \tag{27}$$

The RSD measurements constrain the standard growth combination

$$A(z) \equiv f(z)\sigma_8(z), \tag{28}$$

where  $f(z) = d \ln \delta / d \ln a$  is the logarithmic growth rate and  $\sigma_8(z)$  denotes the rms amplitude of linear matter fluctuations smoothed with a top-hat window of radius  $8 h^{-1} \text{Mpc}$  at redshift  $z$ . In linear theory  $\sigma_8(z)$  scales with the growth factor, which motivates Equations (28)–(30) and the linear scaling of the theory vector with the present-day amplitude  $\sigma_8 \equiv \sigma_{8,0}$ . In the SFS/FSF templates we therefore treat  $\sigma_{8,0}$  as a nuisance amplitude and determine it by analytic profiling in the growth sector (optionally with the adopted Planck Gaussian prior), returning the corresponding best-fit  $\hat{\sigma}_8$  reported in Table 1.

**Table 1.** Best-fit  $\chi^2$  values for the combined (ALL), geometry-only (GEOM), and growth-only (GROW) analyses. For each data set we report the minimum  $\chi^2$ , the offset  $\Delta\chi^2 \equiv \chi_{\text{min}}^2 - \chi_{\text{min},\Lambda\text{CDM}}^2$ , and the corresponding best-fit parameters in the scanned  $(n, \delta)$  grid after profiling over the remaining parameters (including the inner minimization over  $y_0$ ; for growth,  $\sigma_8$  is profiled with the adopted prior).

Data Set	Model	$\chi_{\text{min}}^2$	$\Delta\chi^2$	$n$	$\delta$	$y_0$	$\sigma_8$
ALL	$\Lambda\text{CDM}$	2925.53	0.00	–	–	–	0.814213
ALL	SFS	3092.43	+166.90	1.9999	−0.330067	0.9999	0.796342
ALL	FSFS	2981.37	+55.84	0.0100	0.941000	0.677318	0.812095
GEOM	$\Lambda\text{CDM}$	2909.56	0.00	–	–	–	–
GEOM	SFS	2926.37	+16.80	1.9999	−0.313568	0.9999	–
GEOM	FSFS	2947.81	+38.25	0.0264983	0.850015	0.699058	–
GROW	$\Lambda\text{CDM}$	15.9714	0.0000	–	–	–	0.814213
GROW	SFS	134.263	+118.29	1.9999	−0.792020	0.9999	0.800279
GROW	FSFS	12.5879	−3.38	0.0429967	0.866680	0.733248	0.816469

##### 4.5.1. Perturbations and $f\sigma_8$

The growth prediction is computed by solving a first-order system for the matter density contrast  $\delta(t)$  and auxiliary variables (velocity potential and metric potential variables) on the given background. Initial conditions are imposed at  $z_i = 1090$ , with  $\delta(z_i) = 10^{-3}$  and metric/velocity initial conditions constructed from the same Poisson-type relation used consistently across models.

From the numerical solution we compute

$$f(z) = \frac{d \ln \delta}{d \ln a}, \quad \sigma_8(z) = \sigma_{8,\text{ref}} \frac{\delta(z)}{\delta(0)}. \tag{29}$$

In practice, the code uses  $\sigma_{8,\text{ref}} = 0.8163$  as the reference normalization built into the definition of  $\sigma_8(z)$ , and constructs the theory vector at this reference amplitude,

$$[f\sigma_8(z)]_{\text{th,ref}} = f(z) \sigma_8(z) \Big|_{\sigma_{8,\text{ref}}}. \tag{30}$$

For an arbitrary present-day amplitude  $\sigma_8$  the prediction scales linearly,

$$[f\sigma_8(z)]_{\text{th}} = \left( \frac{\sigma_8}{\sigma_{8,\text{ref}}} \right) [f\sigma_8(z)]_{\text{th,ref}}. \tag{31}$$

#### 4.5.2. Analytic Profiling of $\sigma_8$ with a Planck Prior

We treat  $\sigma_8$  as a nuisance parameter with a Gaussian prior centered on the Planck value [49]:

$$\pi(\sigma_8) \propto \exp \left[ -\frac{(\sigma_8 - \sigma_{8,P})^2}{2 \Delta\sigma_{8,P}^2} \right], \quad \sigma_{8,P} = 0.8163, \quad \Delta\sigma_{8,P} = 0.006. \tag{32}$$

Since the model prediction is linear in  $\sigma_8$ , the combined growth chi-square

$$\chi_G^2(\sigma_8) = (\mathbf{d} - \alpha \mathbf{p})^T \mathbf{C}^{-1} (\mathbf{d} - \alpha \mathbf{p}) + \left( \frac{\sigma_8 - \sigma_{8,P}}{\Delta\sigma_{8,P}} \right)^2, \quad \alpha \equiv \frac{\sigma_8}{\sigma_{8,\text{ref}}}, \tag{33}$$

(where  $\mathbf{p}$  is the rescaled theory vector at  $\sigma_{8,\text{ref}}$  and  $\mathbf{d}$  is the data vector) is minimized analytically with respect to  $\sigma_8$ , yielding both the profiled minimum  $\chi_{G,\text{prof}}^2$  and the best-fit  $\hat{\sigma}_8$ .

#### 4.6. Total Goodness of Fit and Profiling in $y_0$

The total goodness of fit for a given model point is defined as

$$\chi_{\text{tot}}^2 = \chi_{\text{SN}}^2 + \chi_H^2 + \chi_{\text{BAO,AP}}^2 + \chi_{G,\text{prof}}^2, \tag{34}$$

where the growth term is already profiled over  $\sigma_8$  with the Planck prior.

For each pair  $(n, \delta)$  we further profile over  $y_0$  by minimizing  $\chi_{\text{tot}}^2(m, n, \delta, y_0)$  in one dimension using a multi-start strategy with box constraints  $y_0 \in [0.03, 0.9999]$  and additional background viability checks (positivity of  $a(y)$  and  $H(y)$  on the evolution interval and non-negative effective dark energy density  $\rho_{\text{DE}}$ ). In the scans reported below we keep  $m$  fixed to (e.g.,  $m = 2/3$ ) as motivated in Section 3, and perform a dense grid search in  $(n, \delta)$ , storing the profiled best-fit values of  $y_0$  and  $\hat{\sigma}_8$  at each grid point.

#### 4.7. Geometry-Only and Growth-Only Fits, and Profiled Contour Regions

In addition to the combined fit, we explicitly assess how the two classes of observables—(i) purely geometrical probes and (ii) growth-of-structure measurements—constrain the singularity parameters and whether they prefer compatible regions of parameter space.

We define the two partial objective functions

$$\chi_{\text{geom}}^2 \equiv \chi_{\text{SN}}^2 + \chi_H^2 + \chi_{\text{BAO,AP}}^2, \quad \chi_{\text{grow}}^2 \equiv \chi_{G,\text{prof}}^2, \tag{35}$$

where  $\chi_{G,\text{prof}}^2$  denotes the growth contribution after analytic profiling over  $\sigma_8$  (with the adopted Planck prior), while  $\chi_{\text{SN}}^2$  is already profiled over the absolute SN magnitude

nuisance parameter as described above. For each model we then perform two additional scans: a geometry-only scan minimizing  $\chi_{\text{geom}}^2$  and a growth-only scan minimizing  $\chi_{\text{grow}}^2$ .

As for the total likelihood, we profile over the time location parameter  $y_0$  separately for each data subset, i.e.,

$$\chi_{\text{geom,prof}}^2(n, \delta) \equiv \min_{y_0} \chi_{\text{geom}}^2(m, n, \delta, y_0), \quad \chi_{\text{grow,prof}}^2(n, \delta) \equiv \min_{y_0} \chi_{\text{grow}}^2(m, n, \delta, y_0), \quad (36)$$

subject to the same box constraints and background-viability filters used in the combined analysis. At each  $(n, \delta)$  grid point we store the corresponding profiled best-fit values  $\hat{y}_0$  (and, for growth, the best-fit  $\hat{\sigma}_8$  returned by the analytic minimization).

To visualize the constraints, we construct profile-likelihood contour plots in the  $(n, \delta)$  plane for both SFS and FSF and for each subset (GEOM and GROW). Specifically, for  $X \in (\text{geom}, \text{grow})$  we define

$$\Delta\chi_X^2(n, \delta) \equiv \chi_{X,\text{prof}}^2(n, \delta) - \chi_{X,\text{prof,min}}^2, \quad (37)$$

where  $\chi_{X,\text{prof,min}}^2$  is the minimum of the corresponding profiled surface within the scanned domain. Since  $(n, \delta)$  are the two displayed parameters, we report the nominal joint confidence regions using the standard two-parameter thresholds  $\Delta\chi^2 = 2.30, 6.18, 11.83$  for  $1\sigma, 2\sigma,$  and  $3\sigma$ , respectively, and mark the best-fit grid point for each subset on the contour plots. This split presentation provides a direct, model-by-model diagnostic of whether tensions originate predominantly from the background (geometrical) sector or from the perturbation (growth) sector.

Because the profiling in  $y_0$  (and, for growth, in  $\sigma_8$ ) is performed independently for each data subset, the locations of the geometry-only and growth-only minima need not coincide. Consequently, the minimum of the combined objective function is not generally equal to the sum of the separately profiled minima,

$$\chi_{\text{tot,min}}^2 \neq \chi_{\text{geom,min}}^2 + \chi_{\text{grow,min}}^2, \quad (38)$$

with equality only if both subsets are minimized by the same parameter point.

## 5. Results

### 5.1. Overview of the Scan and Profiled Likelihood

We confront the SFS and FSF future-singularity scenarios with a joint late-time likelihood built from Pantheon + SH0ES SNe Ia [36–38], observational Hubble data (OHD) [48,49], DESI DR2 BAO constraints in the AP combination only [46,47], and a compilation of  $f\sigma_8(z)$  growth measurements [13,14,16,48,49]. The background evolution is specified by the analytic scale factor ansatz introduced in Section 3 and implemented in Section 4. Throughout this work we fix  $(H_0, \Omega_{m,0})$  and  $m$ , and scan the singularity parameters  $(n, \delta)$  on a dense grid. For each grid point we profile over the time location parameter  $y_0$  by minimizing the total chi-square Equation (34) subject to basic viability requirements (positivity of  $a$  and  $H$  on the evolution interval and  $\rho_{\text{DE}} \geq 0$  along the sampled background history—see Table 2 for an explicit list of filters). Because these viability requirements act as *hard cuts* and each likelihood evaluation involves stiff ODE solves, the resulting profiled objective  $\chi^2(n, \delta)$  is generally *non-smooth* and may contain holes or disconnected viable patches (especially for the FSF/FSFS template at large  $\delta$ ). In such a setting, a global MCMC exploration would be inefficient (low acceptance, poor mixing, and extensive tuning would be required to robustly handle frequent solver failures and hard rejections). We therefore adopt a deterministic *profile-likelihood mapping* on a regular  $(n, \delta)$  grid, with explicit profiling over  $y_0$  (and analytic profiling over the  $\sigma_8$  nuisance with a Planck prior), and we infer

confidence regions from  $\Delta\chi^2$  contours. The stability of the inferred regions is verified by grid-resolution tests and local refinement scans (Section 5.6 and Table 3).

**Table 2.** Viability filters applied in the construction of the profiled  $\chi^2$  surfaces. Grid points failing any filter are rejected (equivalently assigned  $\chi^2 = \infty$ ) and do not contribute to the profiled maps.

Filter	Criterion	Motivation/Failure Mode
Time location prior	$0.03 \leq y_0 < 1$	Excludes singularities in the past; enforces the adopted model domain.
Background positivity	$a(t) > 0$ and $H(t) > 0$ on the integration interval	Rejects non-expanding or ill-defined backgrounds.
Dark-energy viability	$\rho_{DE}(t) \geq 0$ along the sampled background	Excludes unphysical effective-energy histories in the present setup.
ODE integrability (background)	Successful background solve (no NaN/Indeterminate)	Discards points where the background ODE system fails or becomes numerically unstable.
ODE integrability (perturbations)	Successful perturbation solve (no NaN/Indeterminate)	Discards points where the linear-growth system cannot be evolved reliably.

**Table 3.** Validation of grid-resolution stability for the all-observables profiled best fit. We compare the baseline full-range scan ( $61 \times 61$ ), the higher-resolution full-range scan ( $81 \times 81$ ), and a local refinement (zoom) around the best-fit neighborhood.

Model	Scan	$n_{\min}$	$\delta_{\min}$	$y_{0,\min}$	$\chi^2_{\min}$
SFS	$61 \times 61$	1.9999	-0.330067	0.9999	3092.43
SFS	$81 \times 81$	1.9999	-0.321818	0.9999	3092.65
SFS	zoom ( $31 \times 31$ )	1.9999	-0.326667	0.9999	3092.41
FSFS	$61 \times 61$	0.01	0.941	0.677318	2981.37
FSFS	$81 \times 81$	0.01	0.941	0.677076	2981.58
FSFS	zoom ( $21 \times 21$ )	0.0135	0.9185	0.693061	2980.35

The growth contribution is evaluated from the full (scale-dependent) perturbation system (Section 2), including the fiducial cosmology rescaling of each  $f\sigma_8$  measurement, and with analytic profiling over  $\sigma_8$  using a Gaussian Planck prior (Section 4).

For reference, we also evaluate the same data sets in a fixed  $\Lambda$ CDM model (with the same  $(H_0, \Omega_{m,0})$  used throughout) and use its best-fit predictions as baseline curves in the figures below.

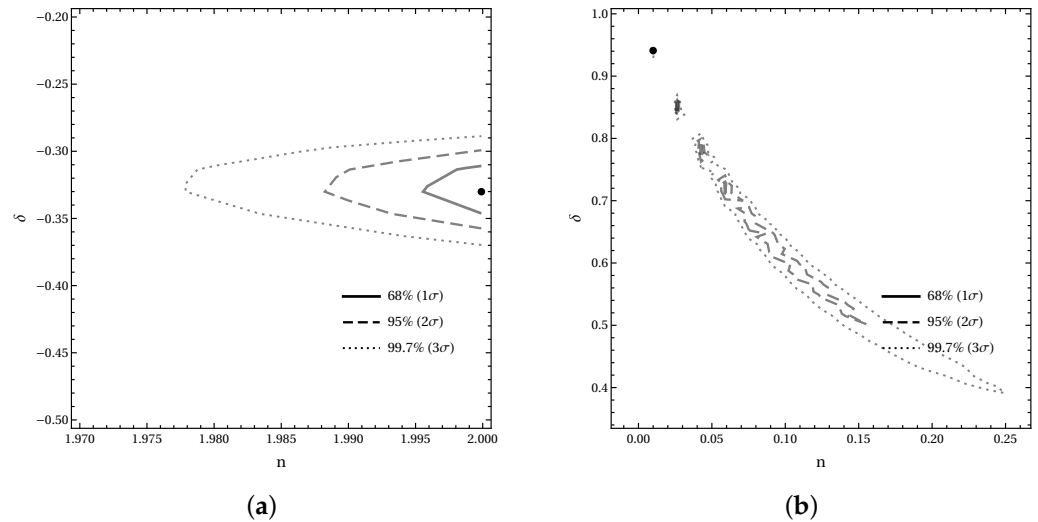
### 5.2. Profile–Likelihood Constraints in the $(n, \delta)$ Plane

For each model (SFS and FSF) we construct the profiled surface

$$\chi^2_{\text{prof}}(n, \delta) \equiv \min_{y_0} \chi^2_{\text{tot}}(m, n, \delta, y_0), \tag{39}$$

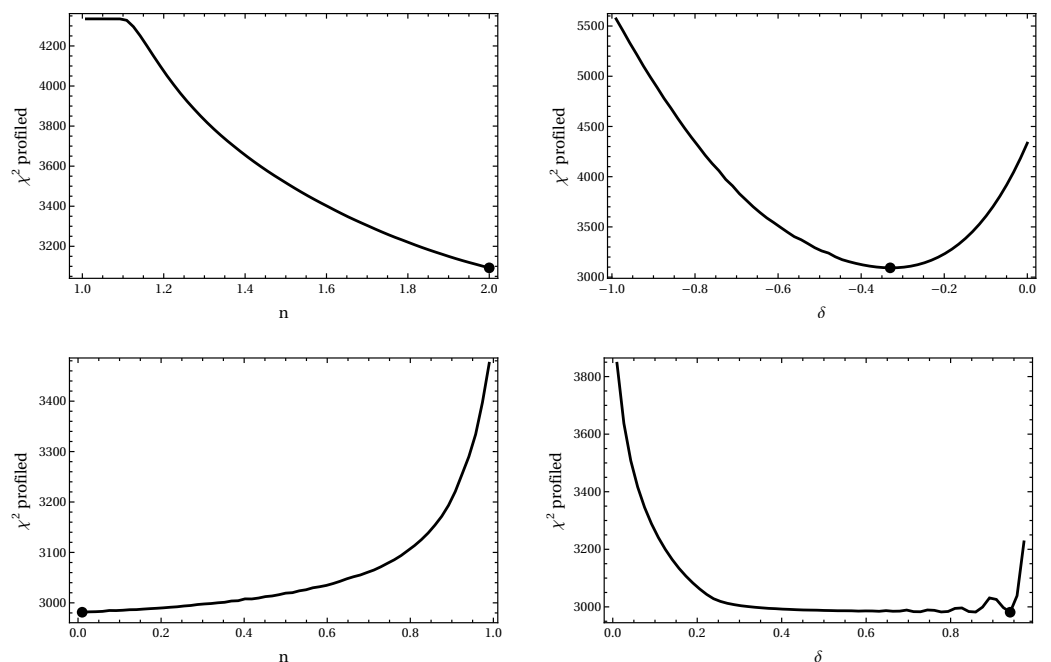
and define  $\Delta\chi^2(n, \delta) = \chi^2_{\text{prof}}(n, \delta) - \chi^2_{\text{prof},\min}$ , where  $\chi^2_{\text{prof},\min}$  is the global minimum of the profiled surface within the scanned domain. Since  $(n, \delta)$  are the two displayed parameters, we use the standard two-parameter thresholds  $\Delta\chi^2 = \{2.30, 6.18, 11.83\}$  to indicate the nominal  $1\sigma, 2\sigma,$  and  $3\sigma$  confidence regions.

Figure 1 shows that the two singularity templates exhibit qualitatively different constraint patterns. In the SFS case the fit improves toward the upper end of the scanned  $n$ -range. Motivating, dedicated, high-resolution scans in the vicinity of the boundary (and/or a broadened prior range) to assess the stability of the minimum and the local degeneracy direction. For FSF, the preferred region is pushed toward small  $n$ .



**Figure 1.** Two-dimensional confidence contours in the  $(n, \delta)$  plane for the SFS (a) and FSFS (b) models, derived from the  $(n, \delta)$  grid after profiling over the remaining parameters at each grid point. Filled markers indicate the best-fit points for the corresponding data subset.

To complement the two-dimensional contours, Figure 2 presents one-dimensional profiles obtained from the same  $(n, \delta)$  grid by minimizing over the remaining parameters at each grid point, including profiling in  $y_0$ . These profiles provide a compact diagnostic of whether the best fit is driven toward the edges of the scanned parameter domain.



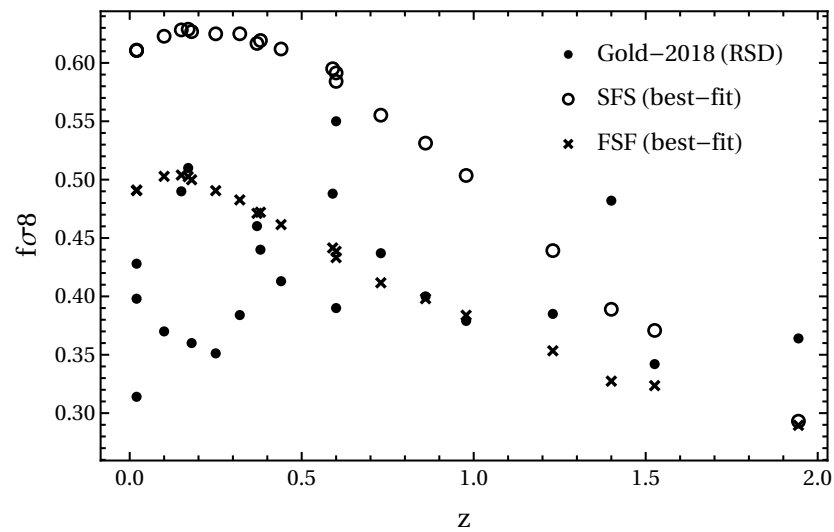
**Figure 2.** One-dimensional profiled  $\chi^2$  constraints derived from the  $(n, \delta)$  grid by minimizing over the remaining parameters (including the inner minimization over  $y_0$ ) at each grid point. (Top row): SFS. (Bottom row): FSF. Filled markers indicate the best-fit points for the corresponding data subset.

### 5.3. Best-Fit Predictions for Background and Growth Observables

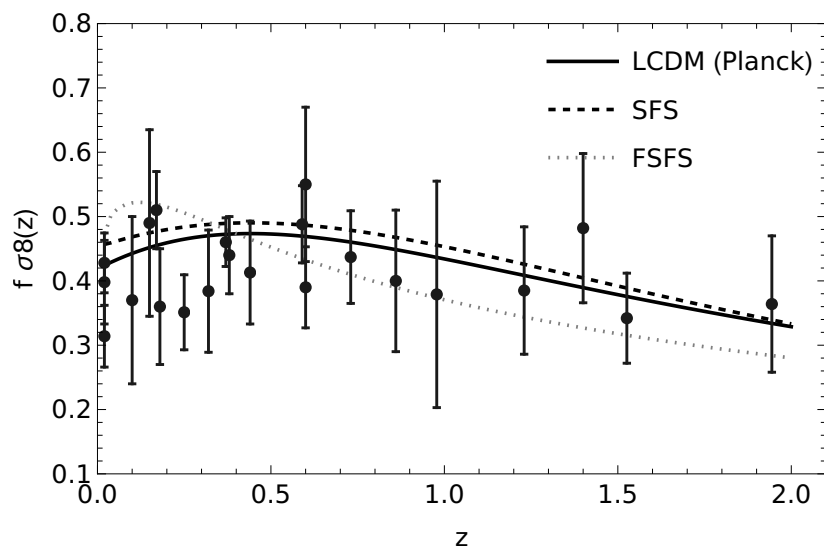
We next compare the best-fit predictions to each data set through residual-style diagnostic plots. These figures provide an intuitive, dataset-by-dataset view of where the models succeed or fail, complementing the numerical  $\chi^2$  summary given below.

#### 5.3.1. Growth: $f\sigma_8(z)$

Figures 3 and 4 compare the growth compilation with the best-fit SFS and FSF predictions. In each case, the theory vector is constructed using the full perturbation evolution on the best-fit background, including the fiducial cosmology rescaling and analytic profiling over  $\sigma_8$  with the adopted prior, so the plotted points correspond to the quantities entering the growth likelihood.



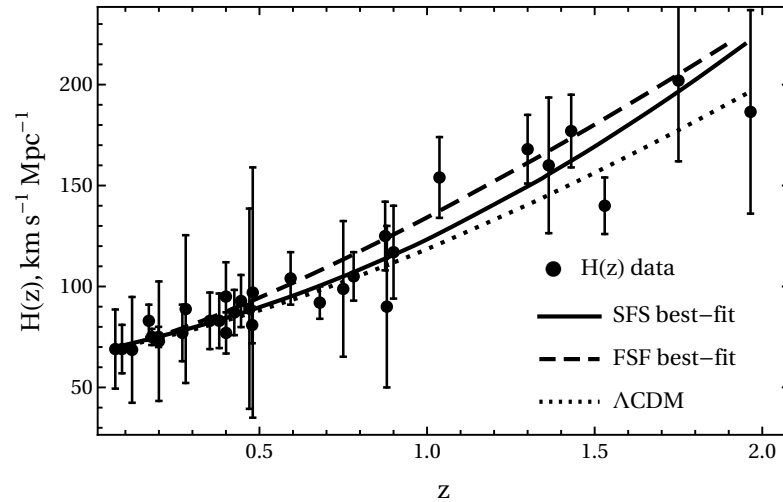
**Figure 3.** Comparison of the RSD growth data  $f\sigma_8(z)$  with the best-fit predictions of the SFS and FSFS models. Filled circles show the observational compilation, while open circles and crosses denote the SFS and FSFS best-fit theoretical values evaluated at the same redshifts, respectively. Values are normalized by the profiled  $\sigma_8$  (i.e., the amplitude is obtained by minimizing the growth-sector contribution to  $\chi^2$  with respect to  $\sigma_8$  for each model).



**Figure 4.** Redshift-space distortion measurements of  $f\sigma_8(z)$  compared to the best-fit predictions of the SFS and FSFS models, as well as to the reference  $\Lambda$ CDM curve evaluated at the same redshifts. The SFS/FSFS theory vectors correspond to the predictions entering the growth  $\chi^2$ , including the Alcock–Paczynski rescaling and profiling over  $\sigma_8$ .

### 5.3.2. Expansion Rate: $H(z)$

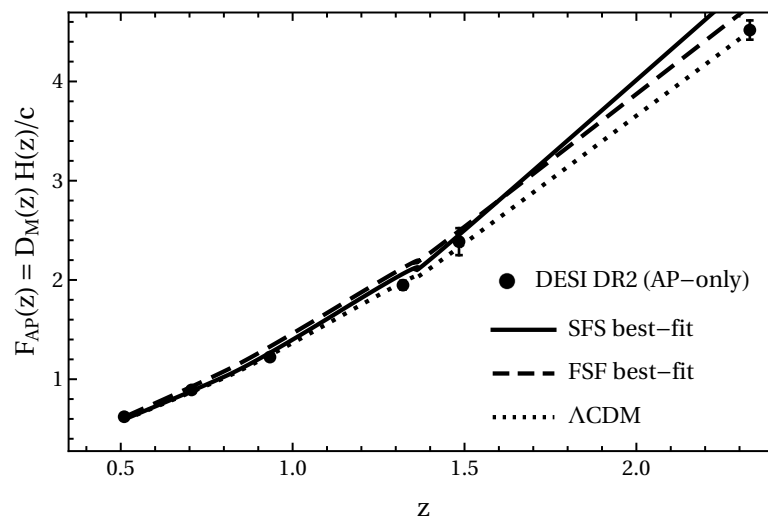
Figure 5 shows the observational Hubble data together with the best-fit background expansion histories of SFS and FSF, compared to the reference  $\Lambda$ CDM background. This plot provides a direct visual summary of the OHD contribution to the joint likelihood.



**Figure 5.** Observational Hubble data  $H(z)$  (points with  $1\sigma$  uncertainties) compared with the best-fit predictions of the SFS and FSFS models and with the reference  $\Lambda$ CDM cosmology.

### 5.3.3. BAO AP-Only: $F_{AP}(z)$

Figure 6 compares the AP-only BAO constraints expressed through  $F_{AP}(z) = D_M(z) H(z)/c$  with the best-fit predictions. In this setup,  $F_{AP}$  acts as a geometric consistency test that complements SNe Ia and OHD.

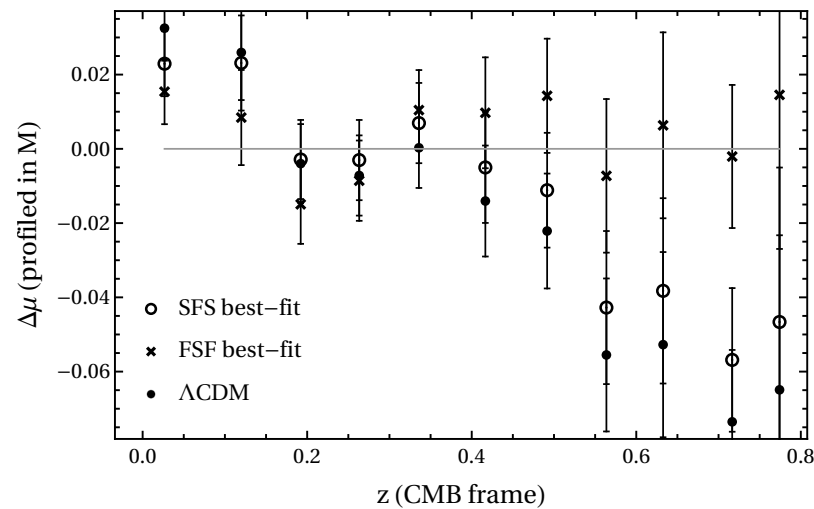


**Figure 6.** Alcock–Paczynski (AP) BAO constraints shown in terms of  $F_{AP}(z) \equiv D_M(z) H(z)/c$ . Black points with error bars correspond to the DESI DR2 AP-only measurements, while the curves show the best-fit predictions of the SFS (solid) and FSFS (dashed) models, compared with the  $\Lambda$ CDM reference (dotted).

### 5.3.4. Supernovae: Binned Residuals with Profiled Magnitude

Figure 7 shows binned SNe Ia residuals constructed after projecting out the same absolute-magnitude mode that is profiled analytically in the SN likelihood. This represen-

tation highlights possible redshift-dependent mismatches without reintroducing sensitivity to the nuisance normalization.



**Figure 7.** Binned Type Ia supernova distance-modulus residuals  $\Delta\mu$  as a function of CMB-frame redshift. At each redshift, the residuals are computed after removing the same profiled mode in the absolute magnitude as used in the SN  $\chi^2$ . Points represent averages in redshift bins (30 bins, requiring at least 10 SNe per bin), and error bars indicate the standard error of the mean within each bin. The best-fit SFS (open circles) and FSF (crosses) predictions are compared to the  $\Lambda$ CDM reference (filled dots). The horizontal line marks  $\Delta\mu = 0$ .

5.4. Goodness of Fit,  $\Lambda$ CDM Comparison, and Geometry vs. Growth Split

At the global best-fit point of each model, Equation (34), we compute the individual contributions and record, in addition, the profiled best-fit value  $\hat{\sigma}_8$  returned by the analytic minimization in the growth sector. A compact summary of the best-fit parameter values and  $\chi^2$  contributions is given in Table 1.

To diagnose whether tensions arise primarily from background probes or from the perturbation sector, we also carry out separate scans for the geometry-only objective function  $\chi^2_{\text{geom}} = \chi^2_{\text{SN}} + \chi^2_H + \chi^2_{\text{BAO,AP}}$  and for the growth-only objective function  $\chi^2_{\text{grow}} = \chi^2_{G,\text{prof}}$ , profiling over  $y_0$  independently in each case. The resulting geometry-only and growth-only constraints in the  $(n, \delta)$  plane are shown in Figure 8 (SFS) and Figure 9 (FSF).

Because  $y_0$  (and, for growth,  $\sigma_8$ ) is profiled independently in the geometry-only and growth-only analyses, the reported minima correspond to different profiled objective functions. Consequently,

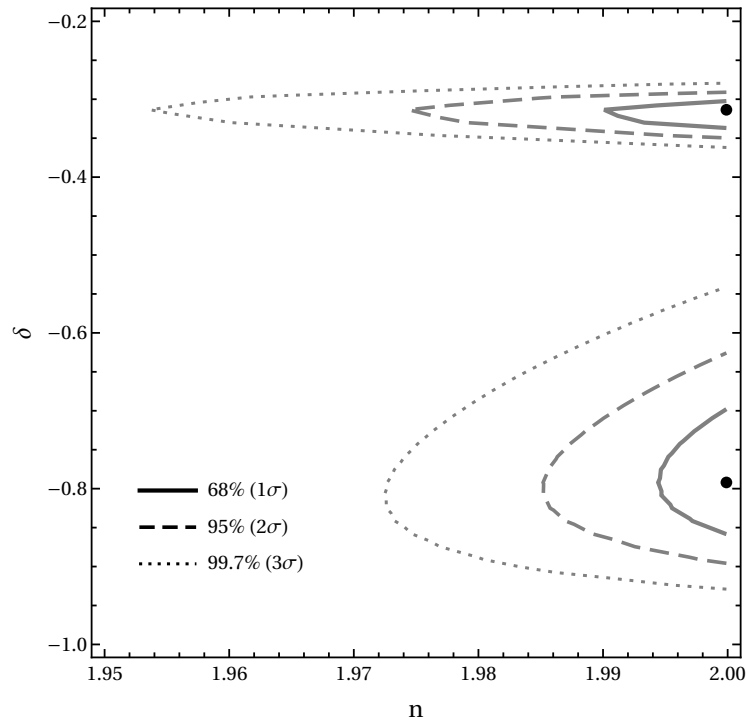
$$\chi^2_{\text{tot,min}} \neq \chi^2_{\text{geom,min}} + \chi^2_{\text{grow,min}}, \tag{40}$$

and, more generally,

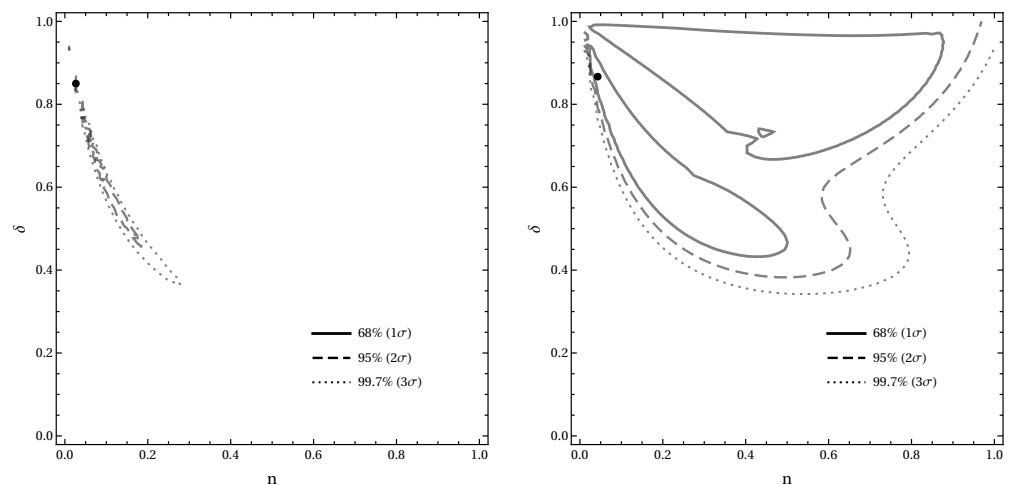
$$\min(\chi^2_{\text{geom}} + \chi^2_{\text{grow}}) \geq \min(\chi^2_{\text{geom}}) + \min(\chi^2_{\text{grow}}), \tag{41}$$

with equality only if both contributions are minimized by the same parameter point.

The split analysis highlights that the two singularity templates are constrained in markedly different ways. Within the scanned domain and under the adopted priors, SFS is primarily penalized by the growth sector, while its geometry-only fit is only moderately degraded relative to  $\Lambda$ CDM. Conversely, FSF can match the growth data competitively (and, in the present scan, slightly improves the growth-only minimum relative to the  $\Lambda$ CDM reference), but it is disfavored by geometric probes, which dominate the degradation in the combined fit.



**Figure 8.** Two-dimensional confidence regions in the  $(n, \delta)$  plane for the SFS model obtained from geometric probes (SN Ia + BAO AP-only +  $H(z)$ ) and from growth data ( $f\sigma_8$ ). In each case the contours correspond to  $\Delta\chi^2 = \{2.30, 6.18, 11.83\}$ , i.e., 68%, 95%, and 99.7% joint confidence levels for two parameters, after profiling over the remaining parameters (including  $y_0$  and, for growth,  $\sigma_8$ ). Filled markers indicate the best-fit points for the corresponding data subset.



**Figure 9.** Two-dimensional confidence regions in the  $(n, \delta)$  plane for the FSFS model obtained from the grid scan after profiling over the remaining parameters (including the inner minimization over  $y_0$ ). **(Left):** geometric fit ( $\chi^2_{\text{geom}}$ ). **(Right):** growth fit ( $\chi^2_{\text{grow}}$ ). Contours correspond to the standard two-parameter  $\Delta\chi^2$  thresholds. Filled markers indicate the best-fit points for the corresponding data subset.

5.5. Model-Selection Diagnostics (AIC/BIC)

In addition to the explicit  $\chi^2$  breakdown and the residual diagnostics, one may quote information criterion statistics as a compact (albeit heuristic) summary of the balance between goodness of fit and model complexity. For Gaussian errors, the Akaike and Bayesian information criteria are defined as [50,51]

$$\text{AIC} \equiv \chi^2_{\text{min}} + 2k, \quad \text{BIC} \equiv \chi^2_{\text{min}} + k \ln N, \quad (42)$$

where  $k$  is the number of fitted (free) parameters and  $N$  is the number of data points used in the given fit.

In our setup, the number of data points entering each likelihood block is

$$N_{\text{SN}} = 1701, \quad N_H = 31, \quad N_{\text{BAO,AP}} = 6, \quad N_{f\sigma_8} = 22, \quad (43)$$

hence  $N_{\text{geom}} = N_{\text{SN}} + N_H + N_{\text{BAO,AP}} = 1738$ ,  $N_{\text{grow}} = N_{f\sigma_8} = 22$ ,  $N_{\text{all}} = N_{\text{geom}} + N_{\text{grow}} = 1760$ . For parameter counting, we treat the singularity parameters  $(n, \delta)$  and the time location parameter  $y_0$  as free in SFS/FSFS, and we include the growth normalization  $\sigma_8$  as a nuisance parameter in the growth sector (profiled analytically with the adopted Planck prior). With  $(H_0, \Omega_{m,0}, m)$  fixed throughout, this implies

$$k_{\Lambda\text{CDM}} = \begin{cases} 0, & \text{GEOM,} \\ 1, & \text{GROW/ALL,} \end{cases} \quad k_{\text{SFS}} = k_{\text{FSFS}} = \begin{cases} 3, & \text{GEOM,} \\ 4, & \text{GROW/ALL,} \end{cases} \quad (44)$$

so that the relative penalty with respect to  $\Lambda\text{CDM}$  is  $\Delta k = 3$  in all three comparisons. Accordingly,

$$\Delta\text{AIC} = \Delta\chi^2 + 2\Delta k = \Delta\chi^2 + 6, \quad \Delta\text{BIC} = \Delta\chi^2 + \Delta k \ln N = \Delta\chi^2 + 3 \ln N, \quad (45)$$

with  $N = N_{\text{geom}}, N_{\text{grow}}, N_{\text{all}}$  depending on the data set.

The resulting information criterion values for all three data combinations (GEOM, GROW, and ALL) are summarized in Table 4. For each data set we list the minimum  $\chi^2$ , together with  $\Delta\chi^2$ ,  $\Delta\text{AIC}$ , and  $\Delta\text{BIC}$  relative to the corresponding  $\Lambda\text{CDM}$  baseline (computed using the same  $N$  and  $\Delta k = 3$  as detailed above).

We stress that AIC/BIC are only heuristic summaries; the more informative comparison remains the explicit chi-square breakdown and the residual plots for each data set.

**Table 4.** Information criterion summary for the geometry-only (GEOM), growth-only (GROW), and combined (ALL) fits. We report  $\chi^2_{\text{min}}$  for each model, and the offsets  $\Delta\chi^2$ ,  $\Delta\text{AIC}$ , and  $\Delta\text{BIC}$  computed with respect to the corresponding  $\Lambda\text{CDM}$  baseline for the same data set. Here  $N_{\text{geom}} = 1738$ ,  $N_{\text{grow}} = 22$ , and  $N_{\text{all}} = 1760$ . The parameter counts are  $k_{\Lambda\text{CDM}} = 0$  (GEOM) and  $k_{\Lambda\text{CDM}} = 1$  (GROW/ALL), while  $k_{\text{SFS}} = k_{\text{FSFS}} = 3$  (GEOM) and  $k_{\text{SFS}} = k_{\text{FSFS}} = 4$  (GROW/ALL), so  $\Delta k = 3$  for all comparisons.

Data Set	Model	$N$	$\chi^2_{\text{min}}$	$\Delta\chi^2$	$\Delta\text{AIC}$	$\Delta\text{BIC}$
ALL	$\Lambda\text{CDM}$	1760	2925.53	0.00	0.00	0.00
ALL	SFS	1760	3092.43	166.90	172.90	189.32
ALL	FSFS	1760	2981.37	55.84	61.84	78.26
GEOM	$\Lambda\text{CDM}$	1738	2909.56	0.00	0.00	0.00
GEOM	SFS	1738	2926.37	16.81	22.81	39.19
GEOM	FSFS	1738	2947.81	38.25	44.25	60.63
GROW	$\Lambda\text{CDM}$	22	15.97	0.00	0.00	0.00
GROW	SFS	22	134.26	118.29	124.29	127.56
GROW	FSFS	22	12.59	−3.38	2.62	5.89

### 5.6. Validation: Grid-Resolution Stability and Local Refinement

To assess the numerical stability of the profiled grid results, we performed two cross-checks for the *all-observables* likelihood: (i) a full-range scan with increased grid resolution (from  $61 \times 61$  to  $81 \times 81$  points in the same  $(n, \delta)$  domain) and (ii) a local “zoom” refinement around the profiled best-fit neighborhood. In all cases we used identical priors and the same one-dimensional profiling over the time location parameter  $y_0$  at each grid node. The resulting best-fit locations and minimum values are summarized in Table 3.

### 5.6.1. Sudden Future Singularity

The  $81 \times 81$  full-range scan reproduces the same boundary-dominated best-fit behavior found in the  $61 \times 61$  baseline scan, with only a minor shift along the  $\delta$  direction (Table 3). A dedicated local refinement in the neighborhood of the minimum (zoom scan) yields a consistent best-fit region and a practically unchanged minimum  $\chi^2$ , indicating that the inferred constraints are not sensitive to the baseline grid spacing in  $(n, \delta)$ .

### 5.6.2. Finite Scale Factor Singularity

As summarized in Table 3, increasing the global grid resolution from  $61 \times 61$  to  $81 \times 81$  preserves the location of the best-fit region in the profiled  $(n, \delta)$  surface. A dedicated local refinement yields a slightly lower  $\chi_{\min}^2$  and shifts the profiled minimum along the same shallow valley. This behavior is expected because, in the high- $\delta$  regime, the one-dimensional profile in  $y_0$  becomes numerically rugged due to stiff ODE solves combined with hard viability cuts. Concretely, for the  $21 \times 21$  local refinement, only 252 out of 441 grid nodes remain viable after the filters in Table 2 (the rest are rejected by non-physical background histories and/or numerical failures). Importantly, the resulting  $\mathcal{O}(1)$  changes in the minimum do not modify the inferred confidence contours or any qualitative conclusions drawn from the baseline maps.

Given this level of stability under both global refinement and local zooming, we adopt the  $61 \times 61$  grid as the reference for figures and summary tables in the main text, the  $81 \times 81$  full-range runs and the local zoom refinements yield indistinguishable best-fit regions and only sub-percent shifts in the profiled minima; hence, the higher-resolution and zoom scans are used solely as validation checks (Table 3).

## 6. Conclusions

We have tested two singularity-inspired dynamical dark-energy templates, SFS and FSFS, against a combined late-time likelihood built from Pantheon + SH0ES SNe Ia [36–38], OHM  $H(z)$  measurements [39–45], DESI DR2 BAO constraints in the AP combination [25,46,47], and a compilation of RSD growth measurements in the form of  $f\sigma_8(z)$  [13,14,16,48,49]. A central feature of the analysis is that the growth predictions are obtained from the full (scale-dependent) linear perturbation system evolved on the model background, with a consistent fiducial cosmology rescaling of each RSD point and analytic profiling over  $\sigma_8$  with a Planck-motivated Gaussian prior. The parameter inference is performed via dense scans in the  $(n, \delta)$  plane with profiling over the singularity time location parameter  $y_0$  at each grid point, subject to background viability filters. The results show a clear qualitative separation between the two scenarios when assessed jointly and also when split into geometry-only and growth-only constraints. For SFS, the best-fit region is driven toward the upper boundary of the explored  $n$  range, and while the geometry-only sector can be accommodated at a moderately degraded level relative to  $\Lambda$ CDM, the growth likelihood strongly penalizes the model within the scanned domain. In the combined analysis, this manifests as a substantial deterioration of the total goodness of fit relative to the  $\Lambda$ CDM reference.

For FSFS, the pattern is different: the growth sector can be fitted competitively (and, within the present scan, yields a slightly improved growth-only minimum relative to  $\Lambda$ CDM), but geometric probes disfavor the model and dominate the degradation of the combined fit. The geometry vs. growth contour comparison therefore indicates that, in the adopted parametrization and prior ranges, FSFS is constrained primarily through background observables, whereas SFS is constrained primarily through structure-growth data.

These conclusions should be interpreted within the deliberately restricted scope of the present analysis:  $(H_0, \Omega_{m,0})$  and  $m$  are held fixed; BAO information is used only through the AP combination (without an absolute ruler calibration in this setup); and  $\sigma_8$  is treated as a nuisance parameter with an external prior.

The SFS/FSFS scenarios considered here are defined via a phenomenological parametrization of the scale factor  $a(t)$ . Consequently, a substantial fraction of parameter combinations leads to either no viable background evolution or to configurations that fall outside the physical domain of the present formulation. In particular, the parameter  $y_0$  controls the location of the future singularity in the adopted parametrization, and values  $y_0 > 1$  would correspond to a singularity in the past and therefore to non-physical solutions within the present setup. We thus impose  $y_0 < 1$  as a hard prior and discard all parameter points that fail to produce a physical background (and, when relevant, a well-defined growth solution). Within the physical domain ( $y_0 < 1$ ) and for fixed  $m = 2/3$ , the SFS likelihood improves monotonically towards the boundary  $y_0 \rightarrow 1^-$  and  $n \rightarrow 2^-$ . Accordingly, the best-fit points saturate the imposed upper bounds in both the joint and split analyses. We interpret this behavior as a limitation of the underlying scale factor parametrization: the fit is driven towards the boundary where the model most closely mimics a standard background evolution, rather than as a robust prediction of an imminent singularity. Importantly, even in this boundary regime the SFS model provides a very poor description of the  $f\sigma_8$  data, which dominates the degradation of the combined fit relative to  $\Lambda$ CDM.

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## Abbreviations

The following abbreviations are used in this manuscript:

$\Lambda$ CDM	Lambda Cold Dark Matter
SFS	Sudden Future Singularity
FSFS	Finite Scale Factor Singularity
DESI	Dark Energy Spectroscopic Instrument
CMB	Cosmic Microwave Background
BAO	Baryon Acoustic Oscillations
SN Ia	Type Ia Supernovae
LSS	Large-Scale Structure

## Note

- <sup>1</sup> All calculations and figures were produced using *Mathematica* (Wolfram Research, Inc., *Mathematica*, Version 13.0, Champaign, IL (2021)).

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