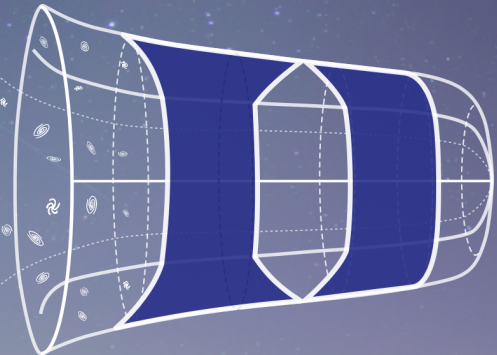


SNe Ia growth-rate measurements with Rubin-LSST simulations: intrinsic scatter systematics

B. Carreres, R. Chen, E. Peterson, D. Scolnic, C. Ravoux, D. Rosselli, M. Acevedo, J. E. Bautista, D. Fouchez, L. Galbany, B. Racine and The LSST Dark Energy Science Collaboration

 arXiv 2505.13290



Duke Cosmology



$f\sigma_8$ as a probe for general relativity

Cosmological principle:
Universe is homogeneous

Image credits: Illustris TNG

$f\sigma_8$ as a probe for general relativity

Observations:
Universe is *not exactly* homogeneous

Image credits: Illustris TNG

$f\sigma_8$ as a probe for general relativity

Observations:

Universe is *not exactly* homogeneous

Matter density fluctuations:

$$\delta(\mathbf{x}) = \frac{\rho(\mathbf{x})}{\bar{\rho}} - 1$$

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σ_8 : fluctuation over sphere of
 $8 \text{ Mpc} \cdot h^{-1}$ radius

$$\delta(\mathbf{x}) = \sigma_8 \tilde{\delta}(\mathbf{x})$$

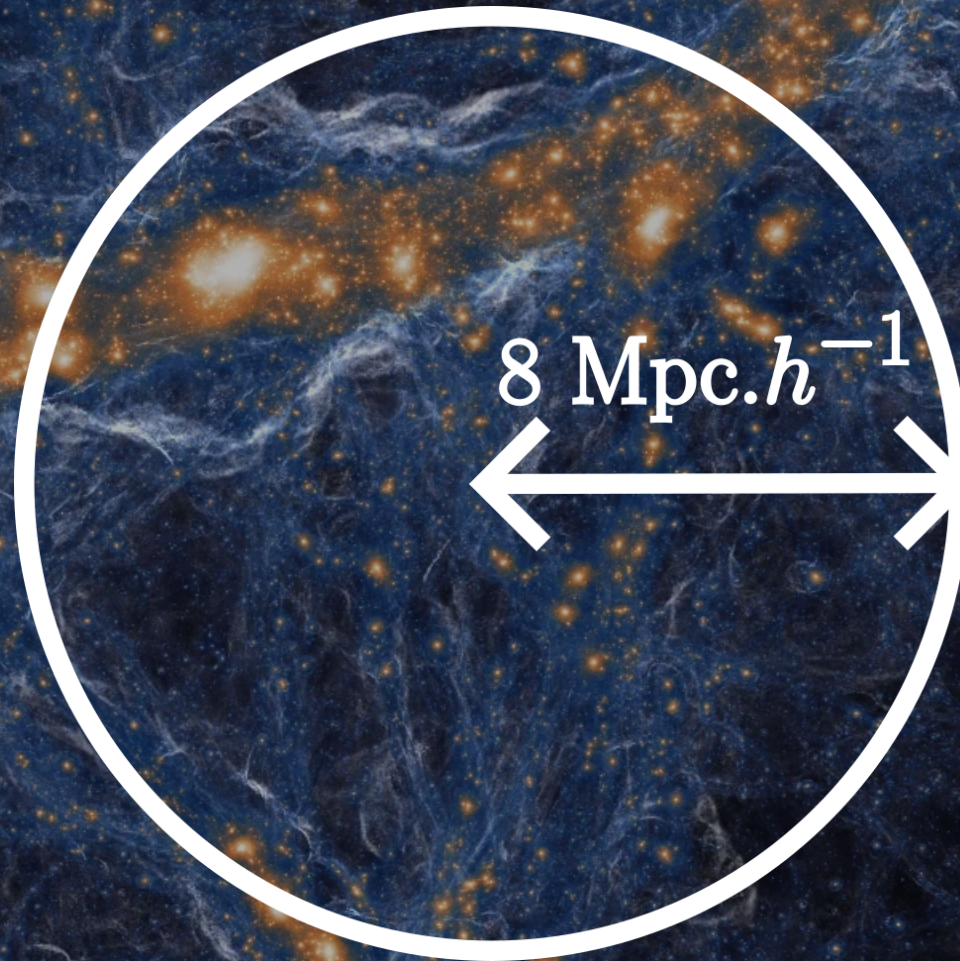


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8 Mpc. h^{-1} radius

$$\delta(\mathbf{x}) = \sigma_8 \tilde{\delta}(\mathbf{x})$$

Fluctuations evolve over time

$$\delta(\mathbf{x}, t) = \delta(\mathbf{x}) D(t)$$

$z = 20.0$

50 Mpc/h

Image credits: Illustris TNG

$f\sigma_8$ as a probe for general relativity

Evolution of structures:
Dark energy vs Gravity

$$f = \frac{d \ln D}{d \ln a}$$

Dark Energy
Gravity

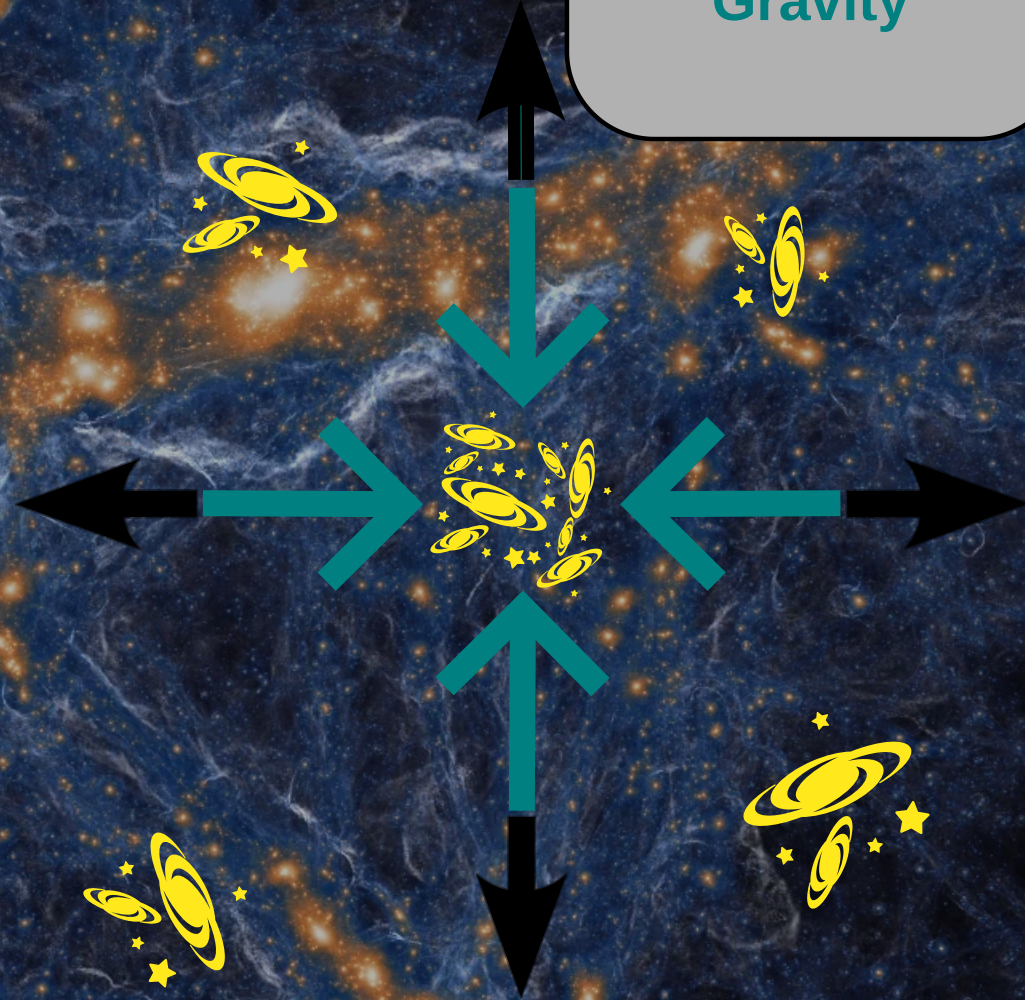


Image credits: Illustris TNG

$f\sigma_8$ as a probe for general relativity

Evolution of structures:
Dark energy vs Gravity

$$f = \frac{d \ln D}{d \ln a}$$

$$f \simeq \Omega_m^\gamma$$

General Relativity + Λ CDM: $\gamma \simeq 0.55$

$\gamma \equiv$ Growth index

f measurement is generally
degenerate with σ_8 !
 $\Rightarrow f\sigma_8$

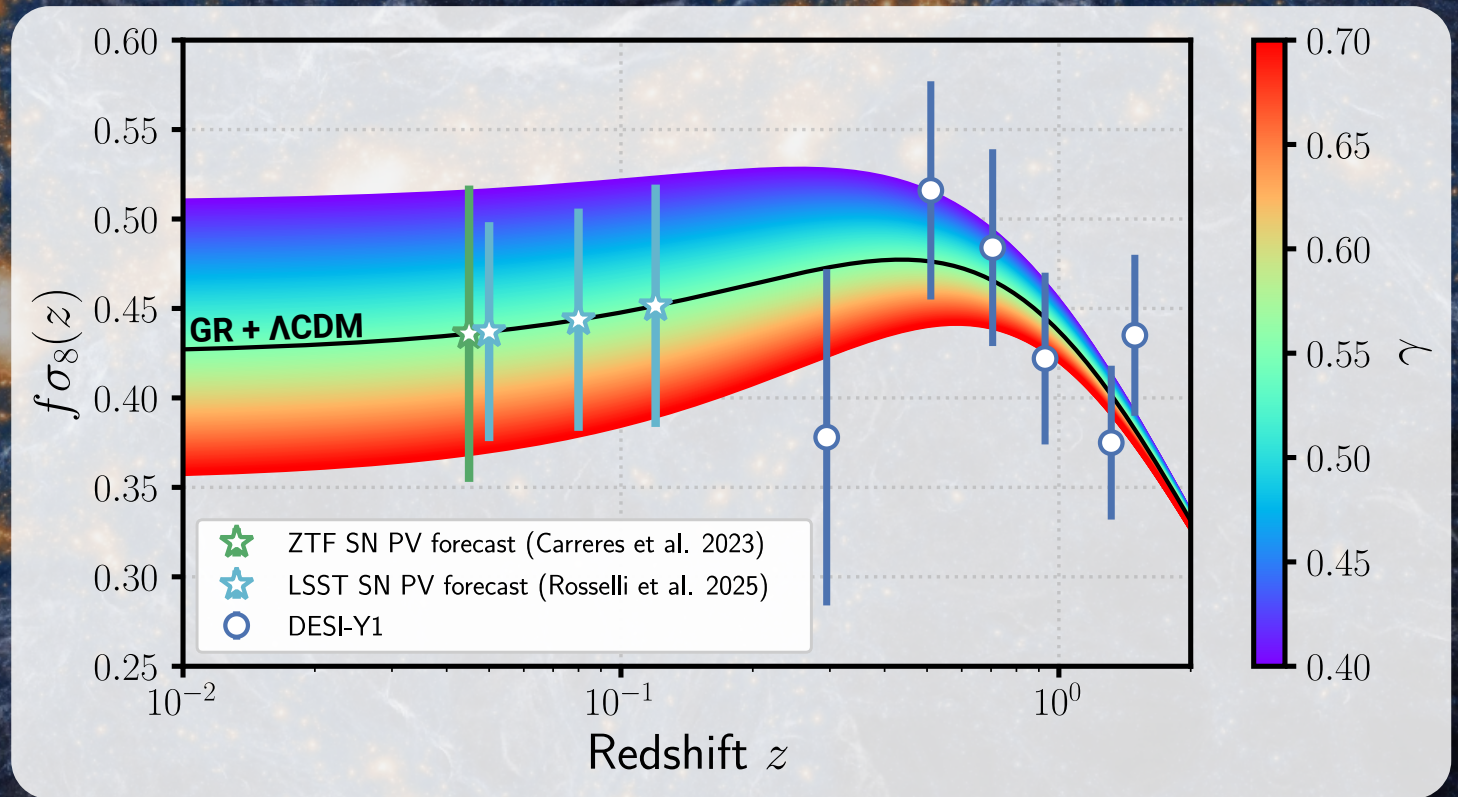
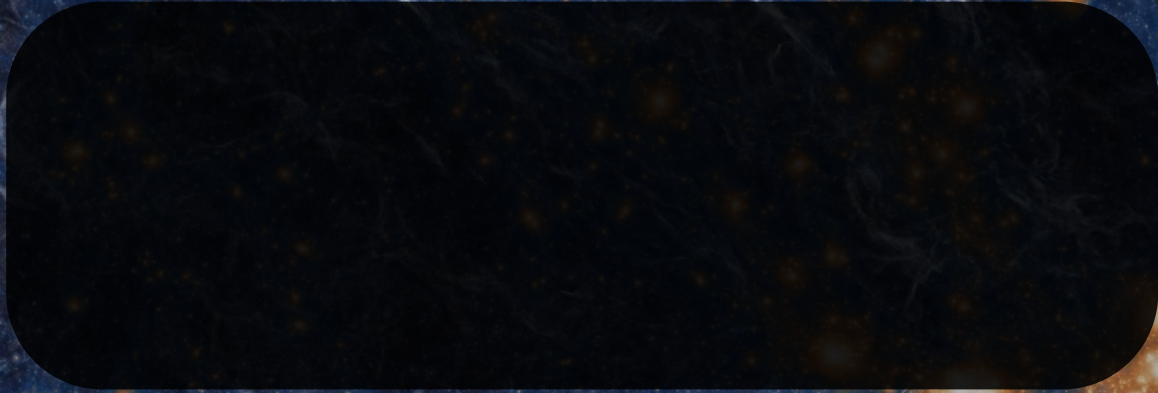


Image credits: Illustris TNG

How to measure $f\sigma_8$?



Dark Energy
Gravity

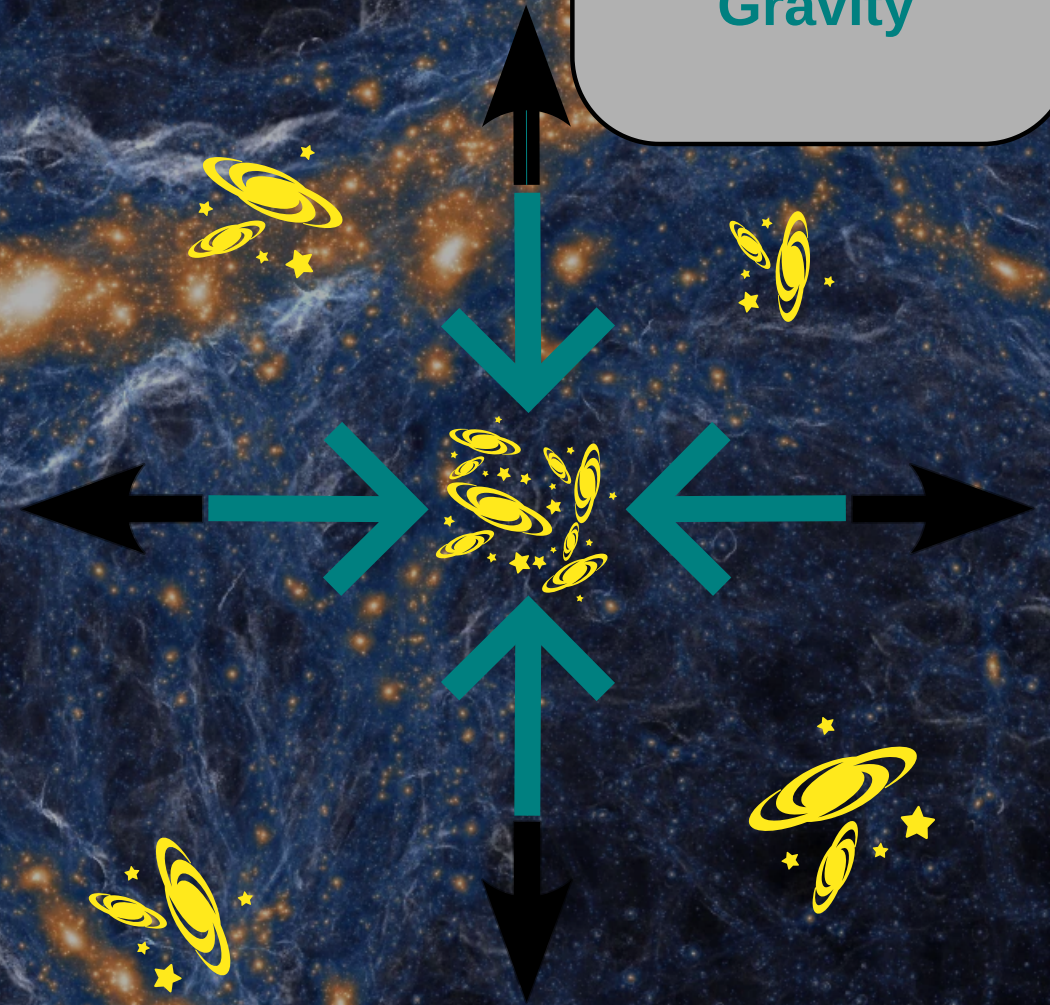


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Velocities are probes of $f\sigma_8$!

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Gravity
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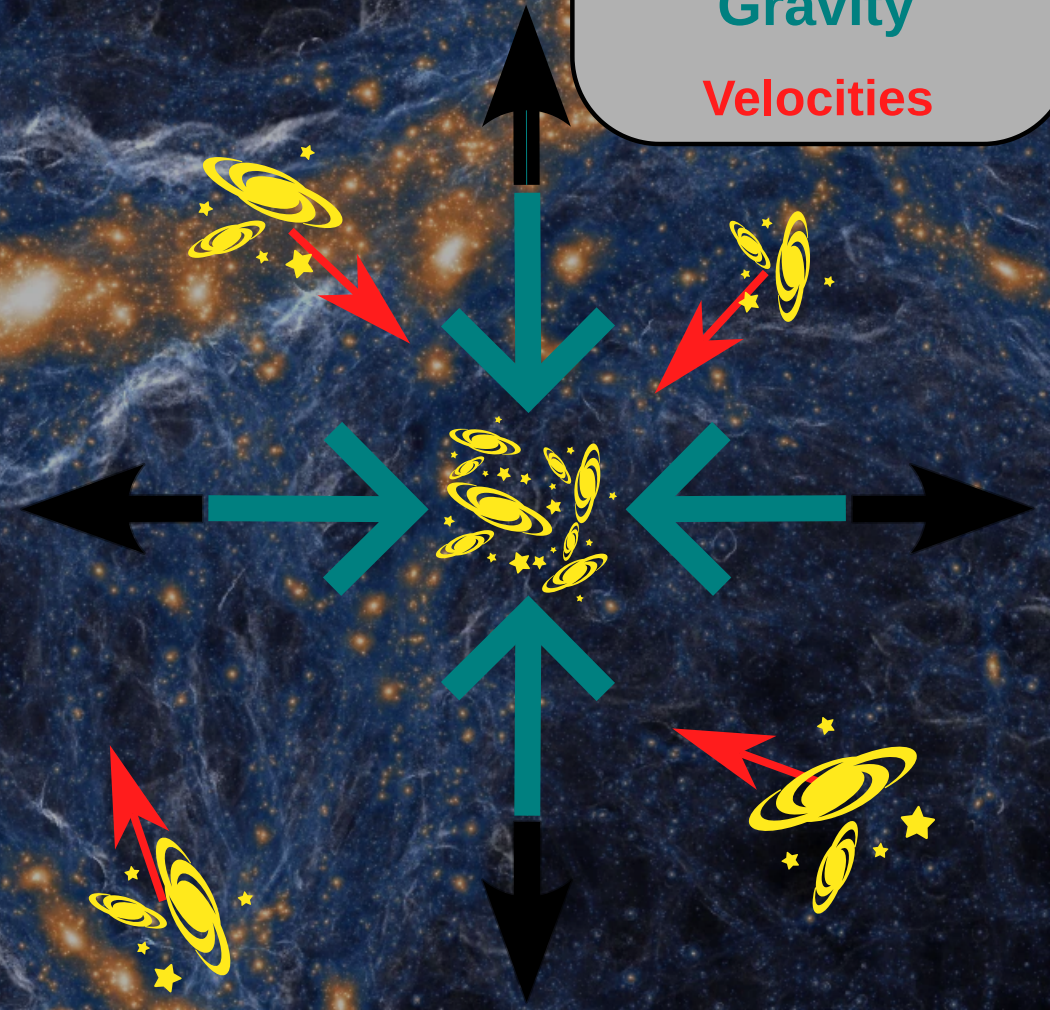


Image credits: Illustris TNG

How to measure $f\sigma_8$?

Velocities are probes of $f\sigma_8$!

Velocity statistics directly depend on $f\sigma_8$:
 $\langle v(\mathbf{x}_i)v(\mathbf{x}_j) \rangle \propto (f\sigma_8)^2$

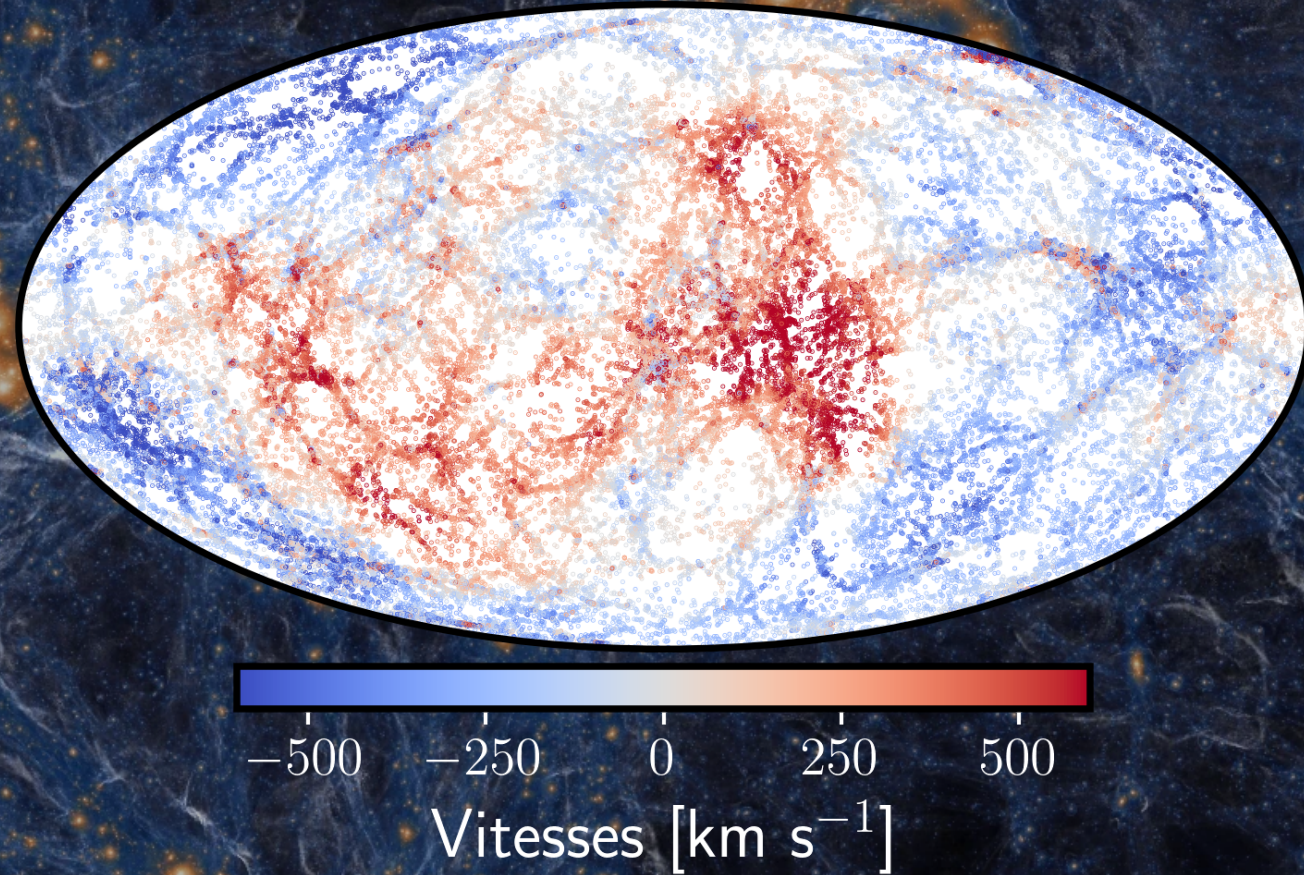
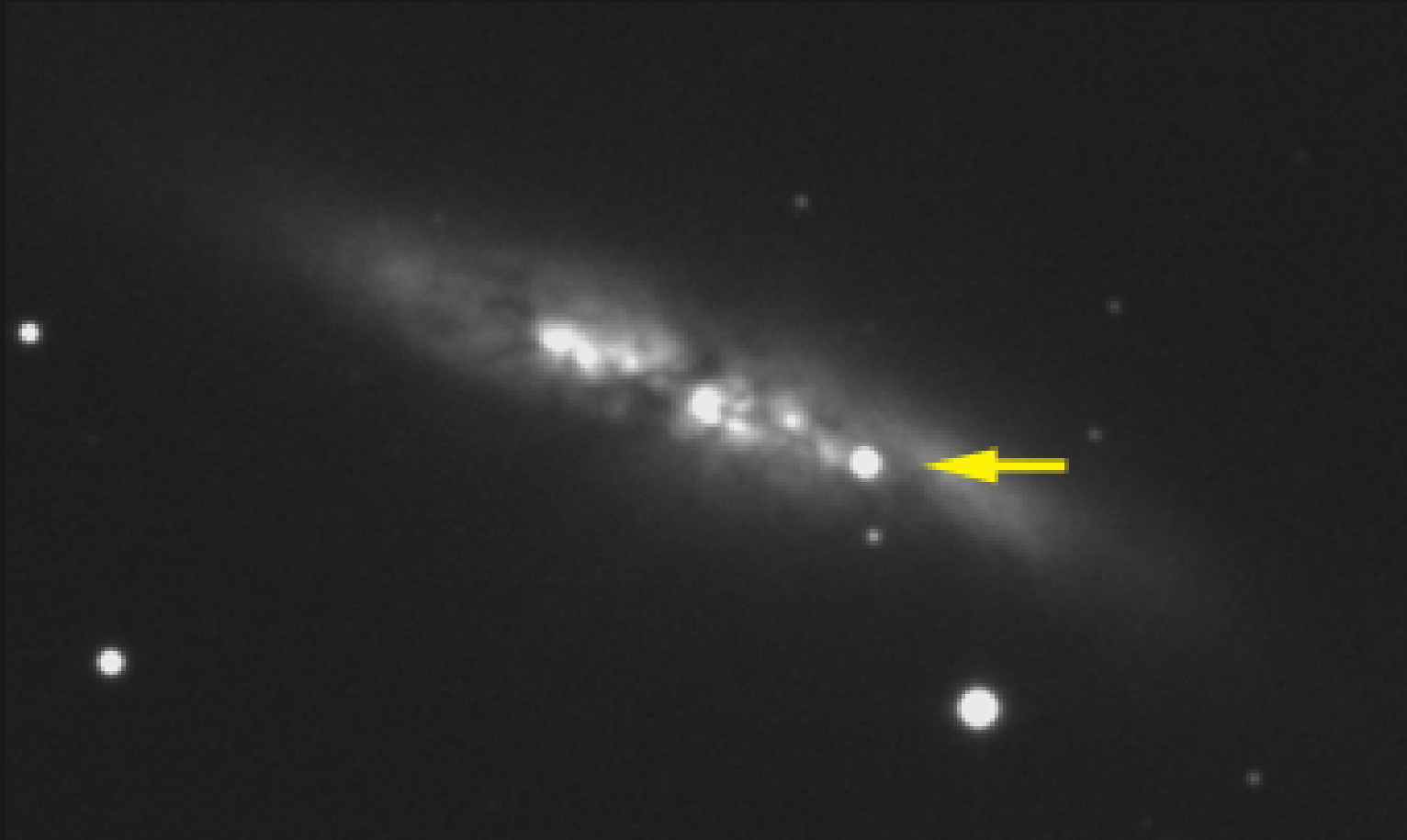


Image credits: Illustris TNG

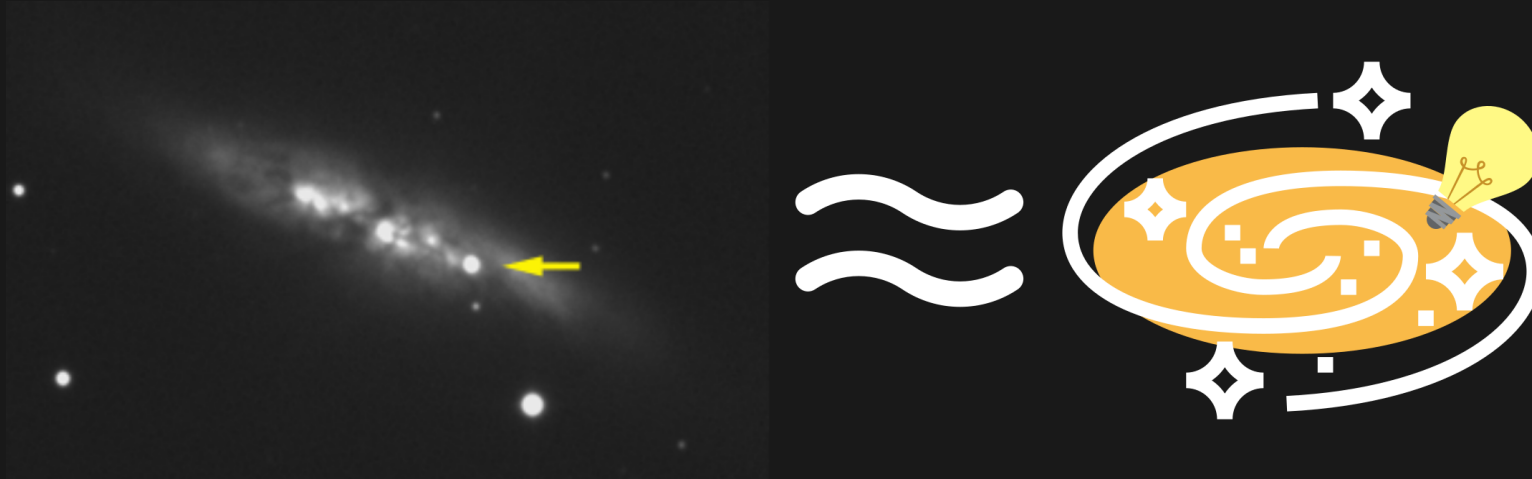
Estimating velocities from SNe Ia

Type Ia supernovae are exploding white dwarfs!

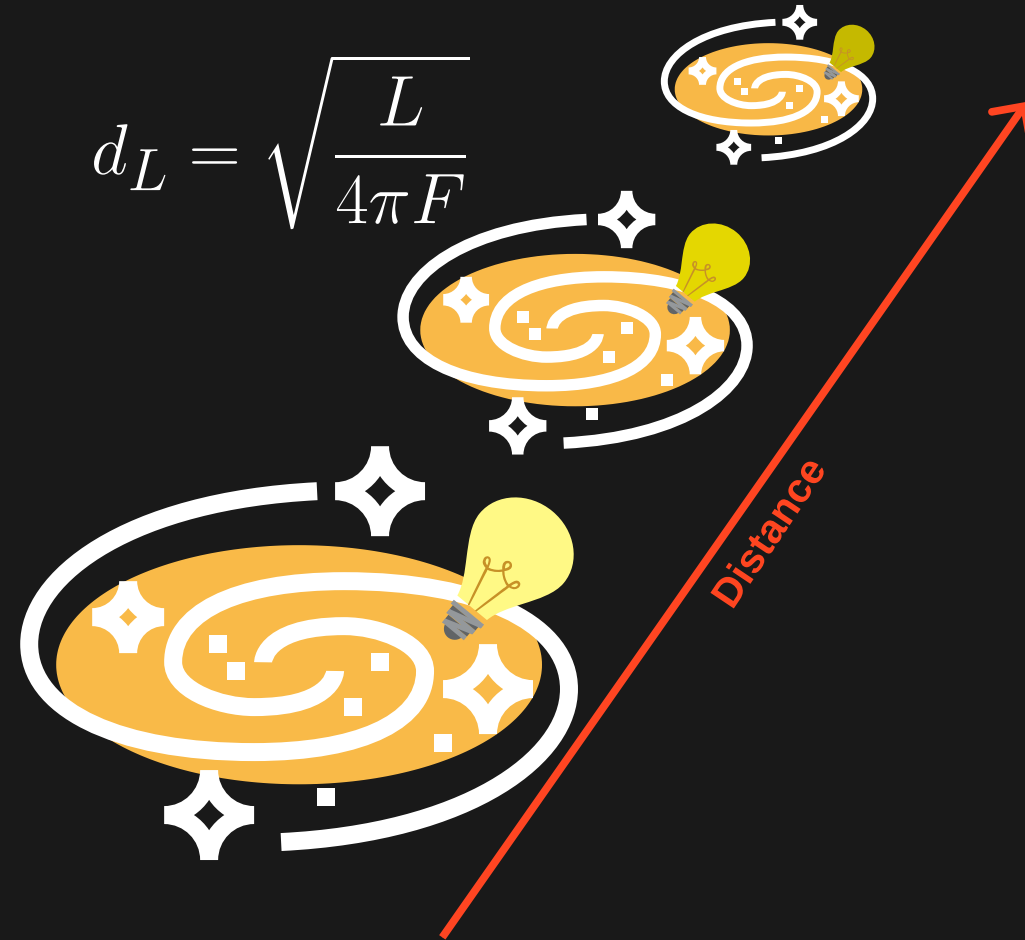


Estimating velocities from SNe Ia

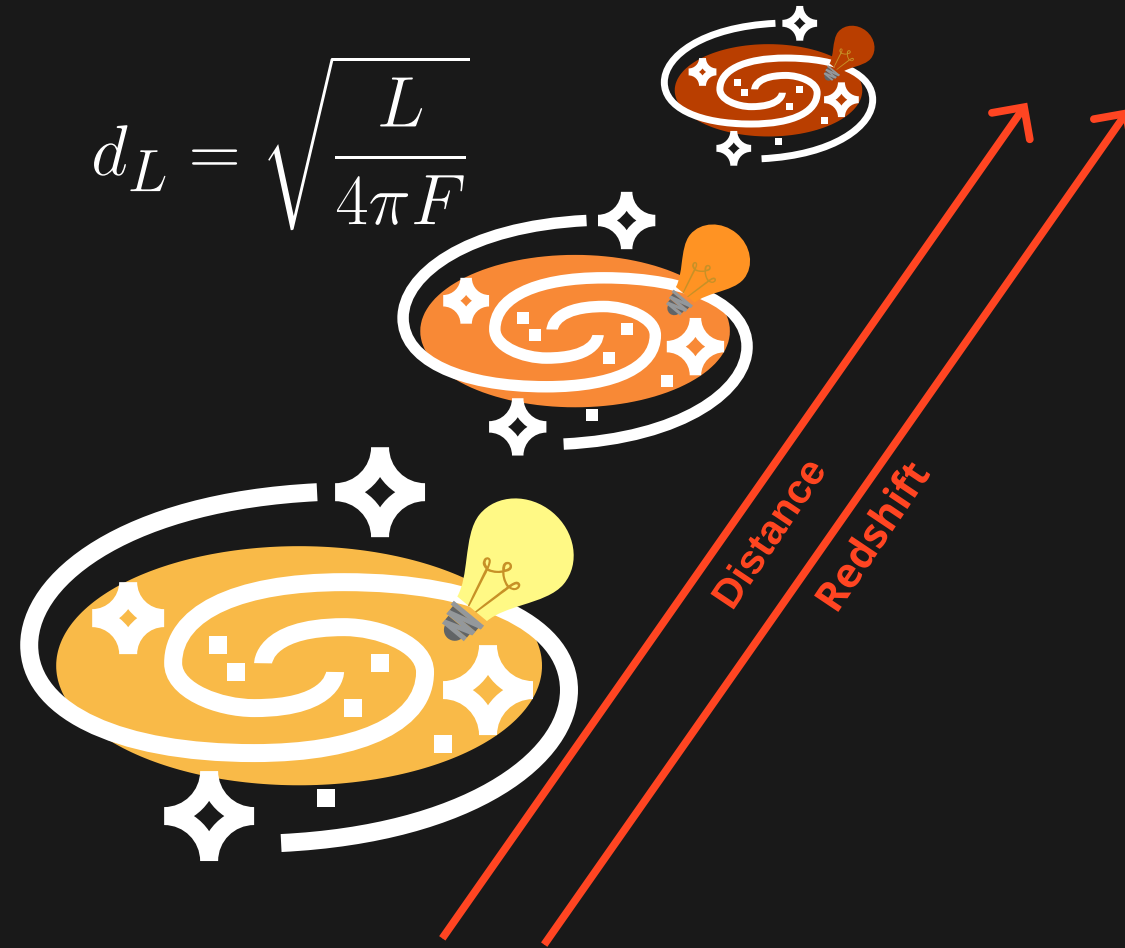
Type Ia supernovae are standard candles!



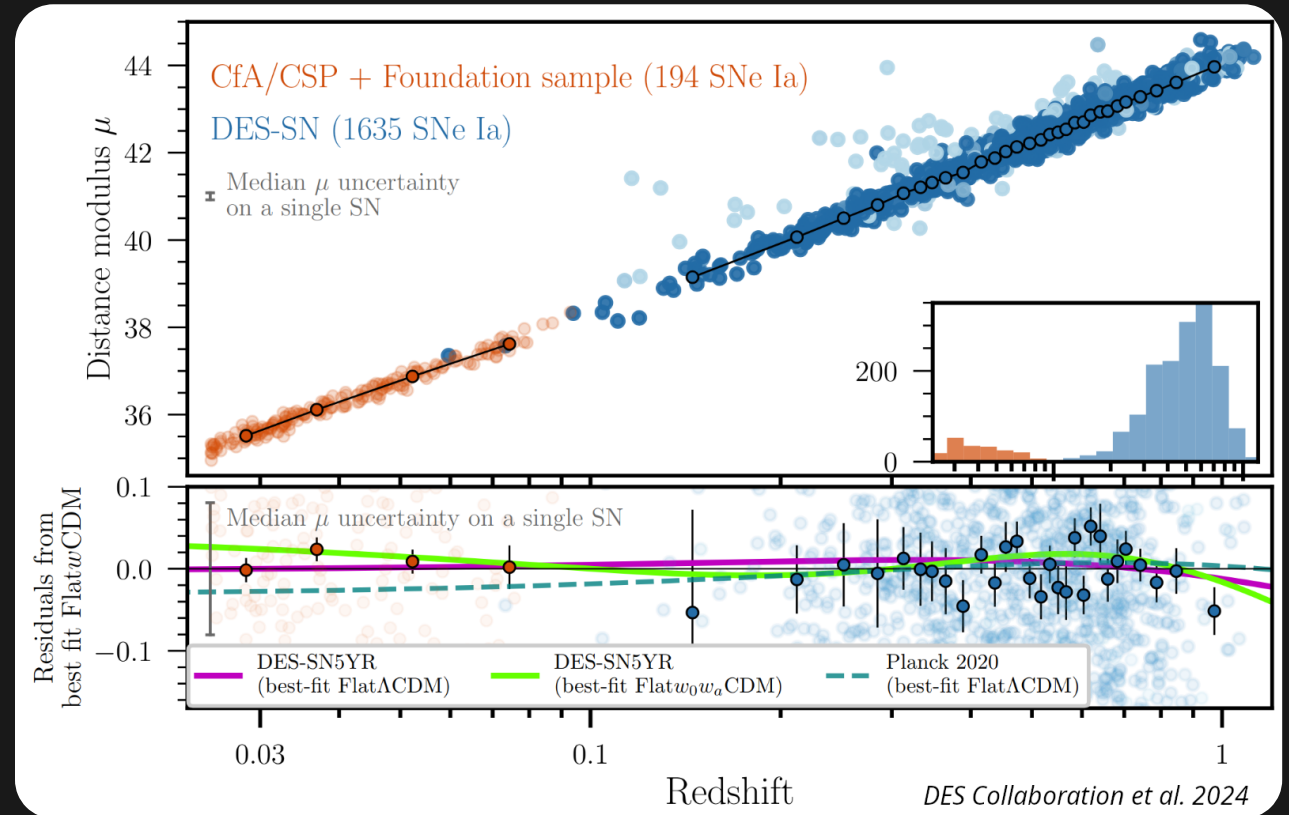
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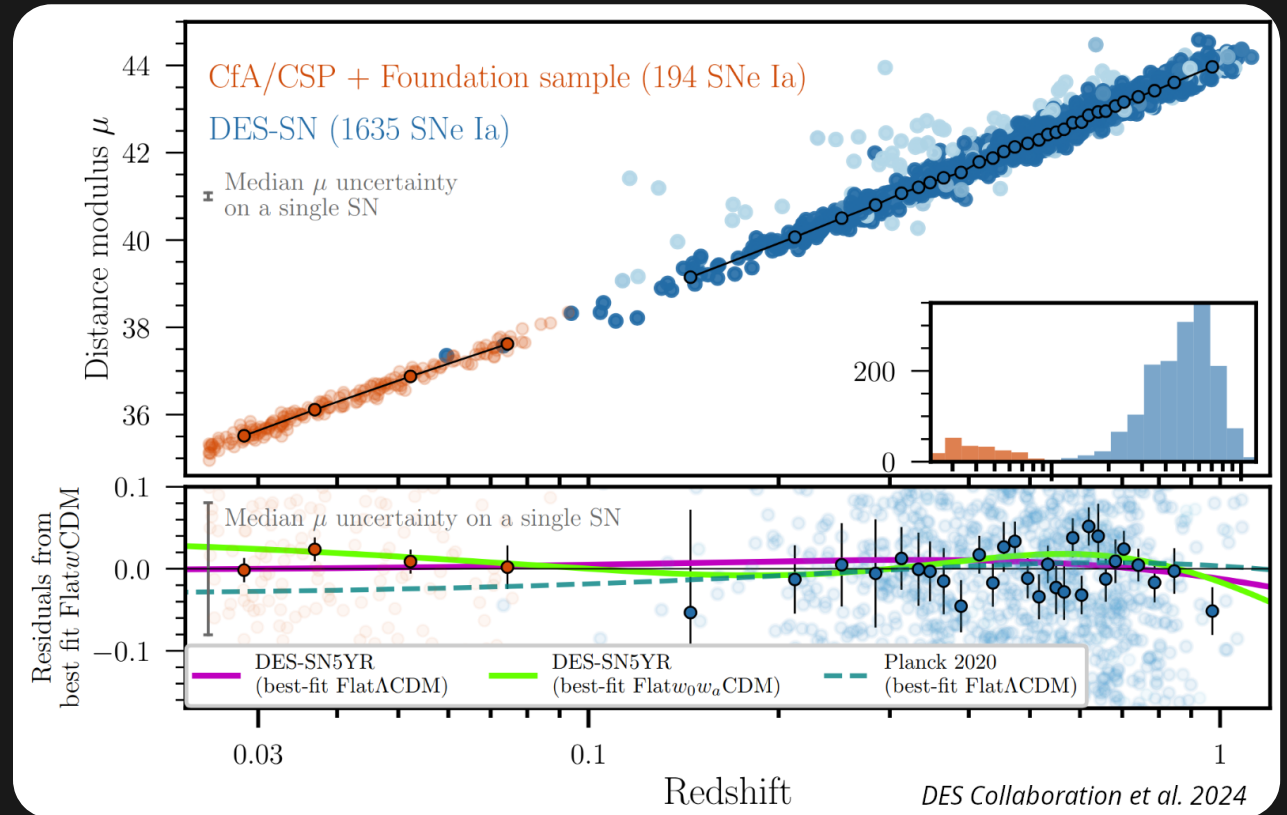


Estimating velocities from SNe Ia



Estimating velocities from SNe Ia

Distance modulus:
$$\mu = 5\log(d_L/10 \text{ pc}) = m - M$$



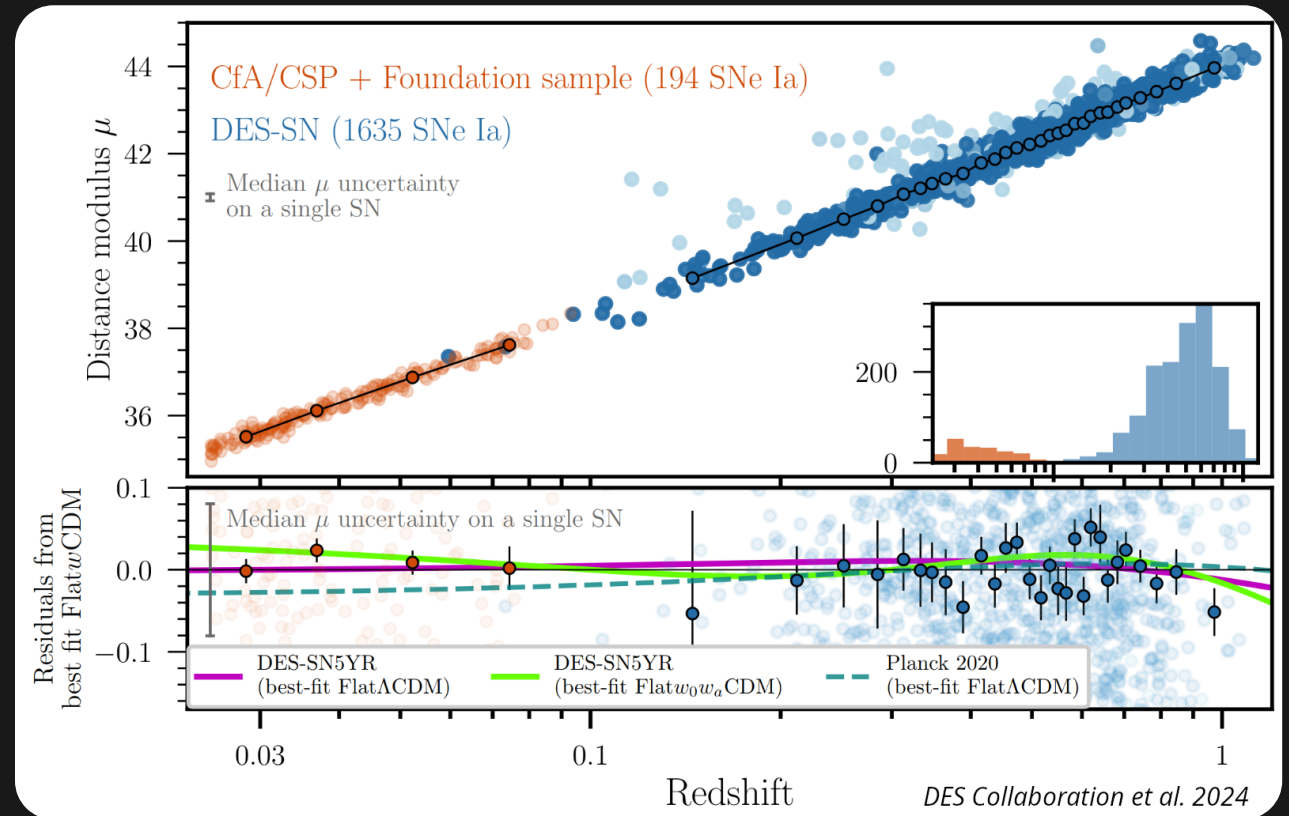
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SNe Ia luminosity is corrected for known correlations!

$$\mu_{\text{obs}} = m_B - (M_B)$$



Estimating velocities from SNe Ia

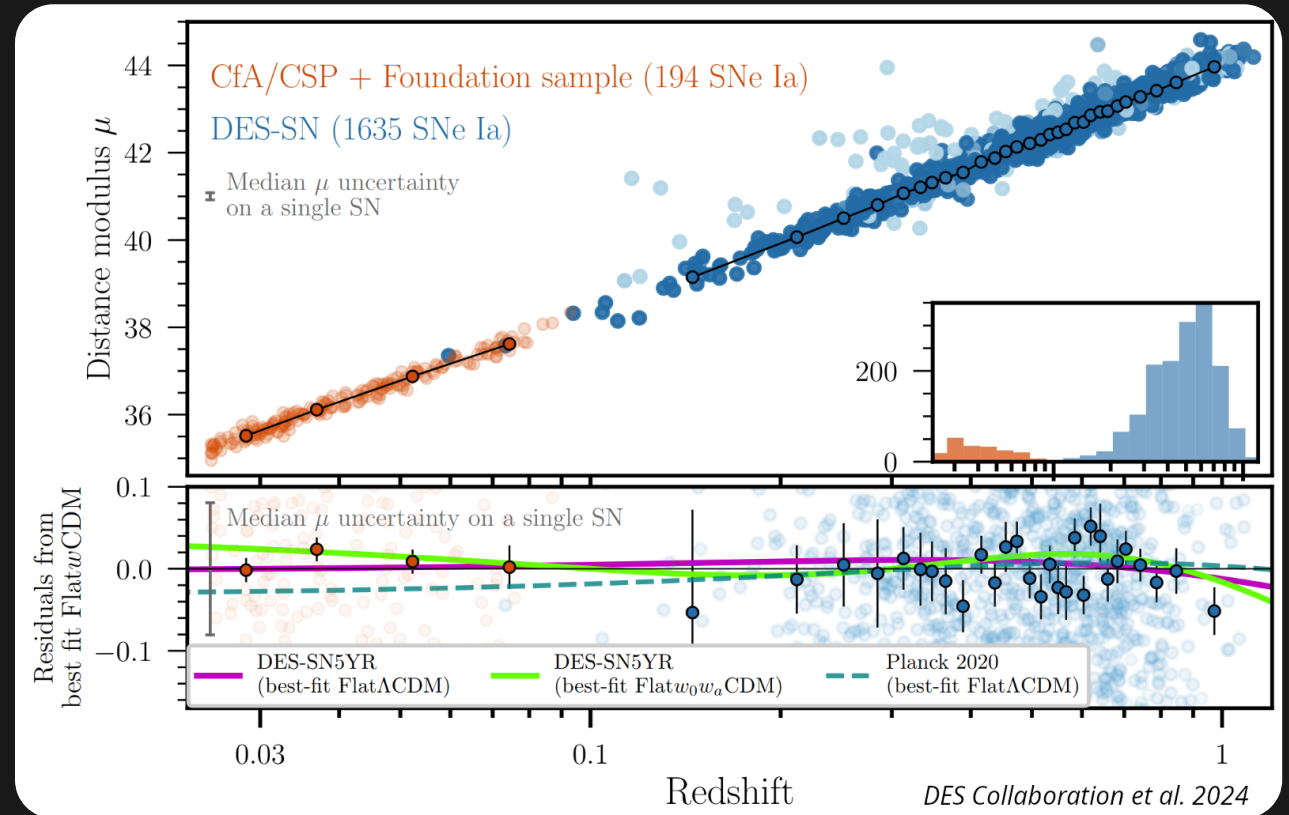
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- Brighter-slower: higher stretch $x_1 \Rightarrow$ brighter SNIa

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Estimating velocities from SNe Ia

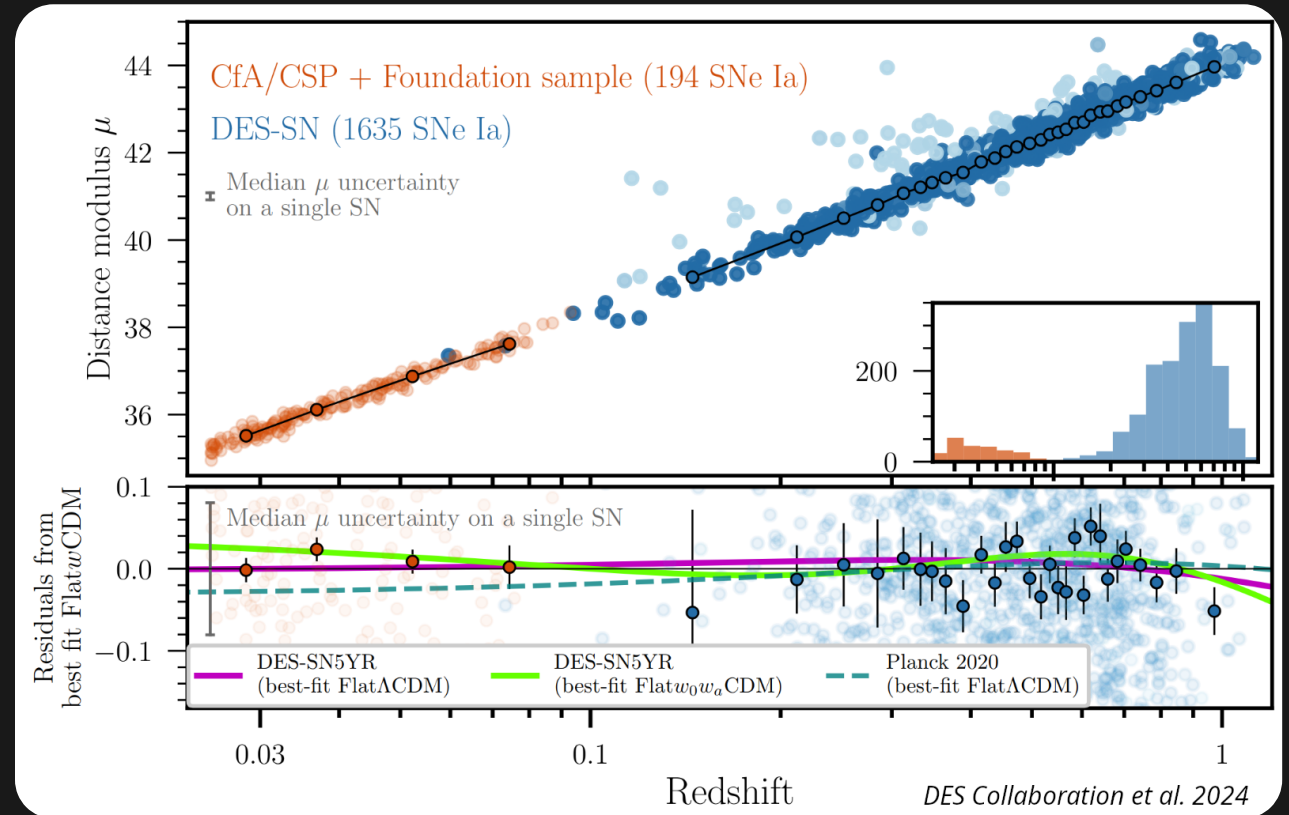
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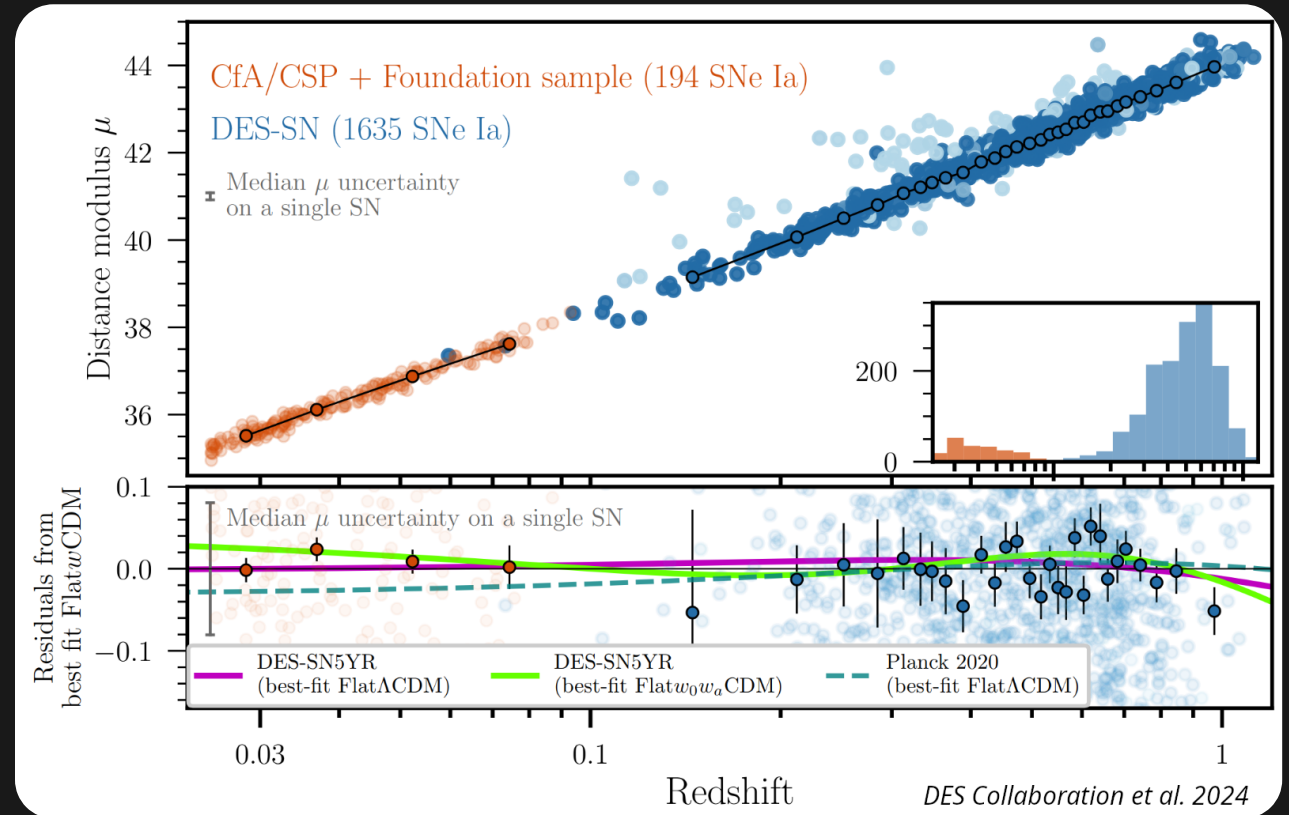
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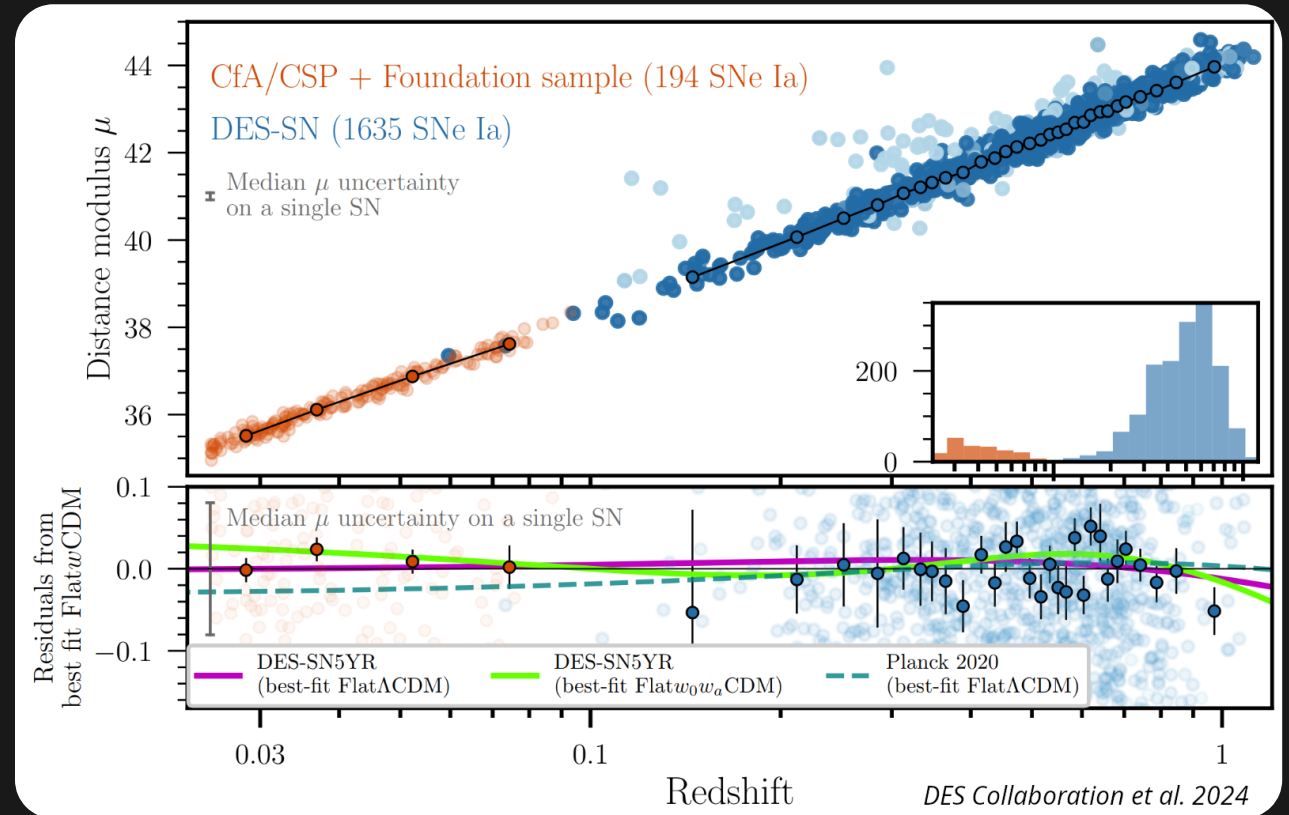
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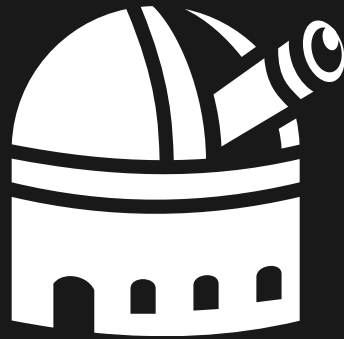
Remaining intrinsic scatter $\sigma_\mu \sim 0.12 \text{ mag}$



Estimating velocities from SNe Ia

Observed redshifts:

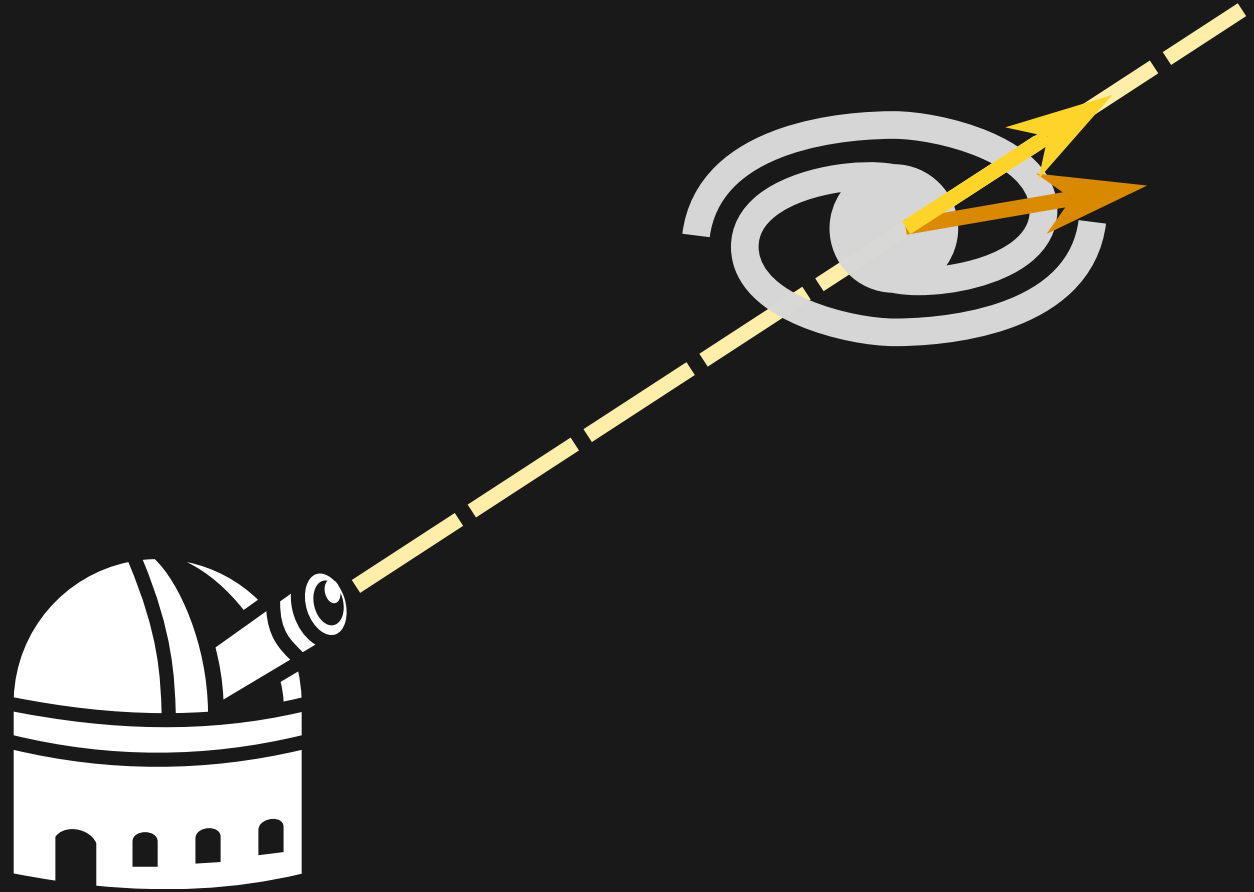
$$1 + z_{\text{obs}} = (1 + z_{\text{cos}})(1 + z_p) \text{ with } z_p \simeq v_p/c$$



Estimating velocities from SNe Ia

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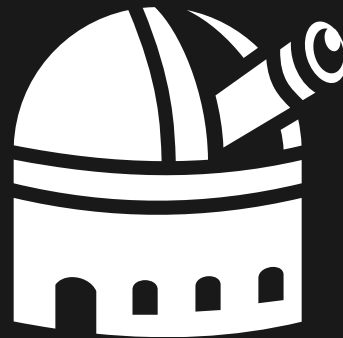
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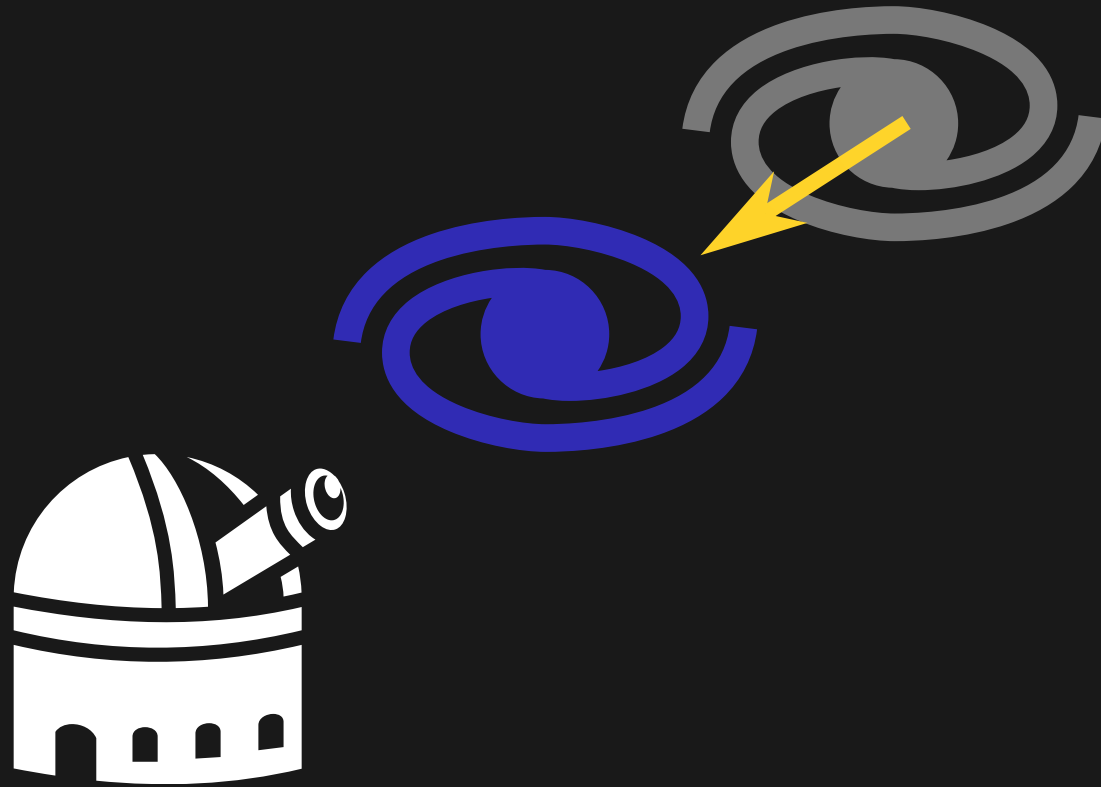
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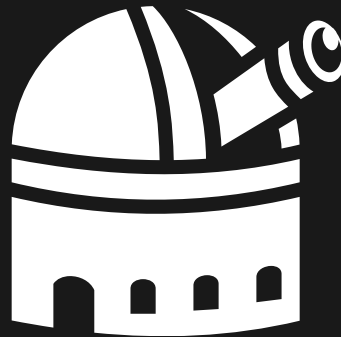
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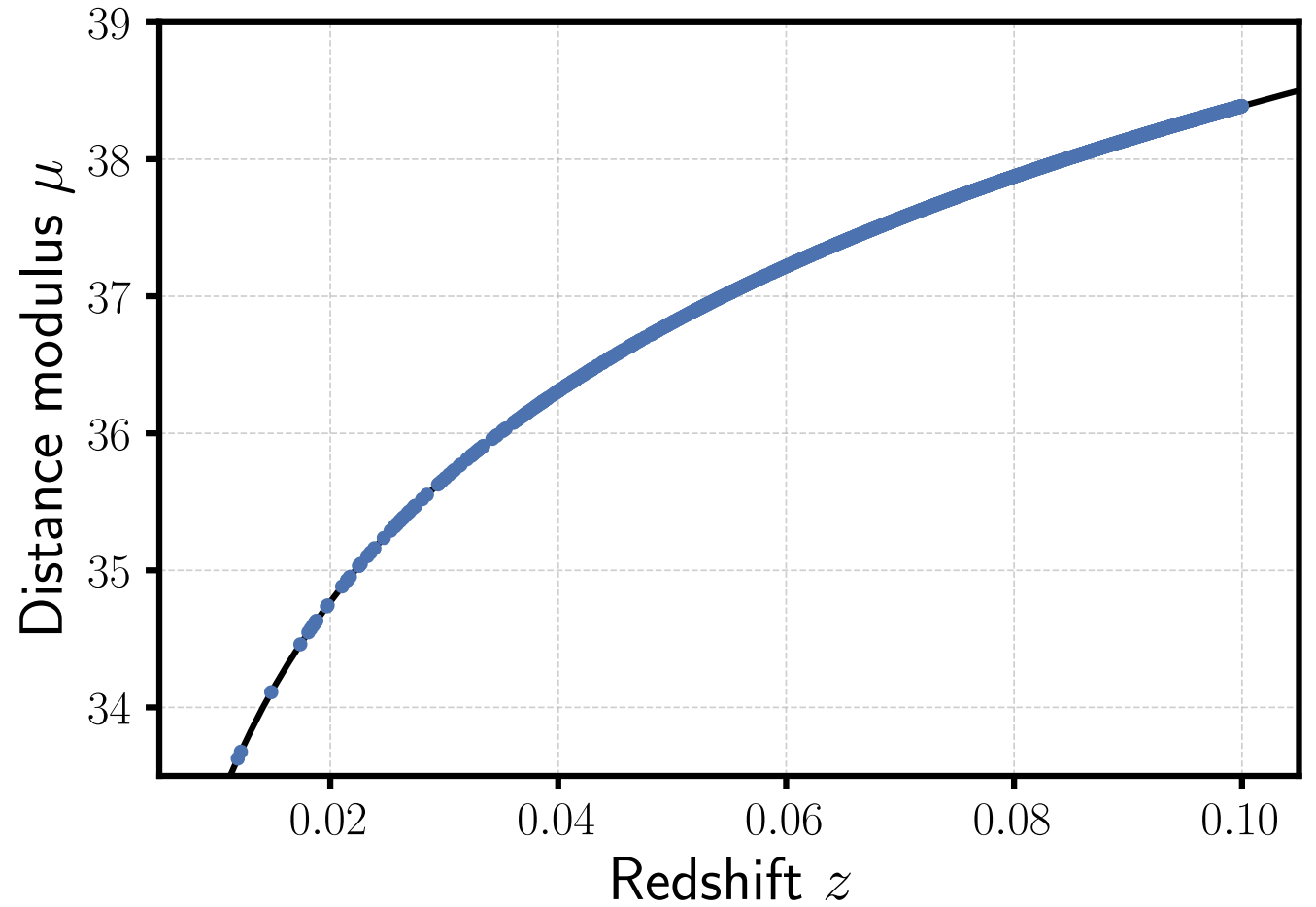
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$$\Delta\mu = \mu_{\text{obs}} - \mu_{\text{th}}(z_{\text{obs}}) = 0$$



Estimating velocities from SNe Ia

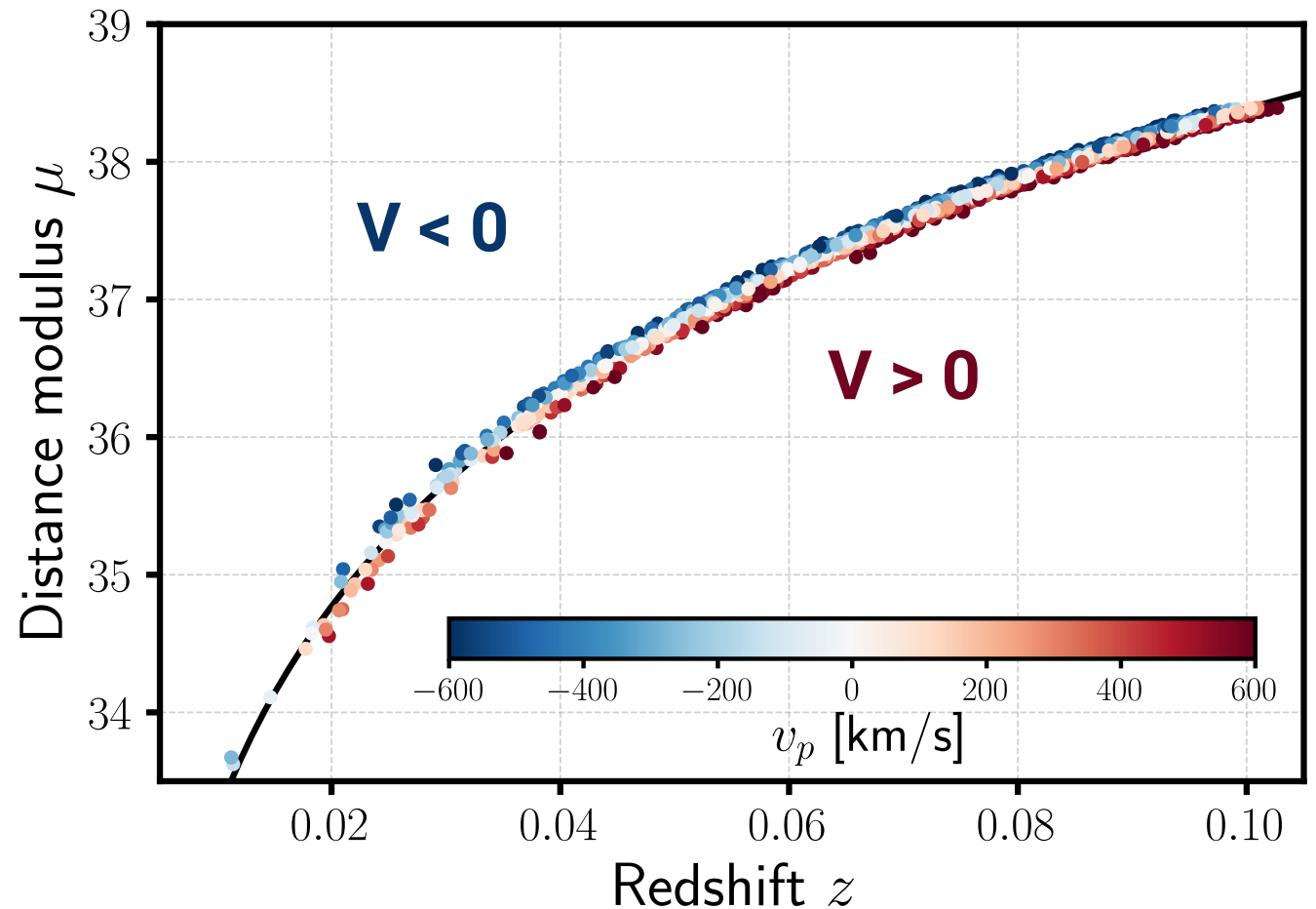
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$$\Delta\mu = \mu_{\text{obs}} - \mu_{\text{th}}(z_{\text{obs}}) \simeq -\frac{5}{c \ln 10} \left(\frac{(1+z)c}{H(z)r(z)} - 1 \right) v_p$$

$$\sigma_{v_p} \simeq -\frac{c \ln 10}{5} \left(\frac{(1+z)c}{H(z)r(z)} - 1 \right)^{-1} \sigma_{\Delta\mu}$$



Estimating velocities from SNe Ia

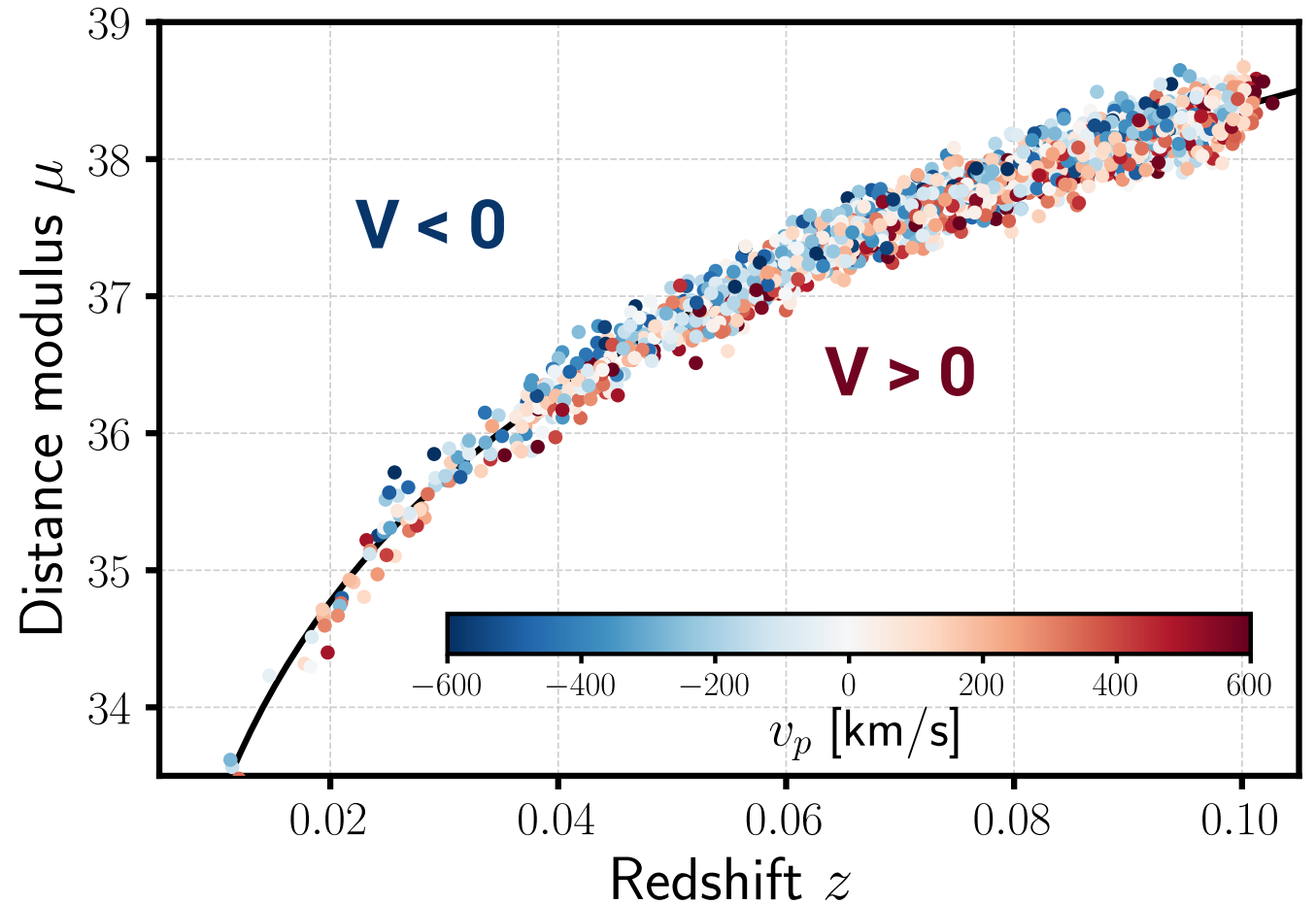
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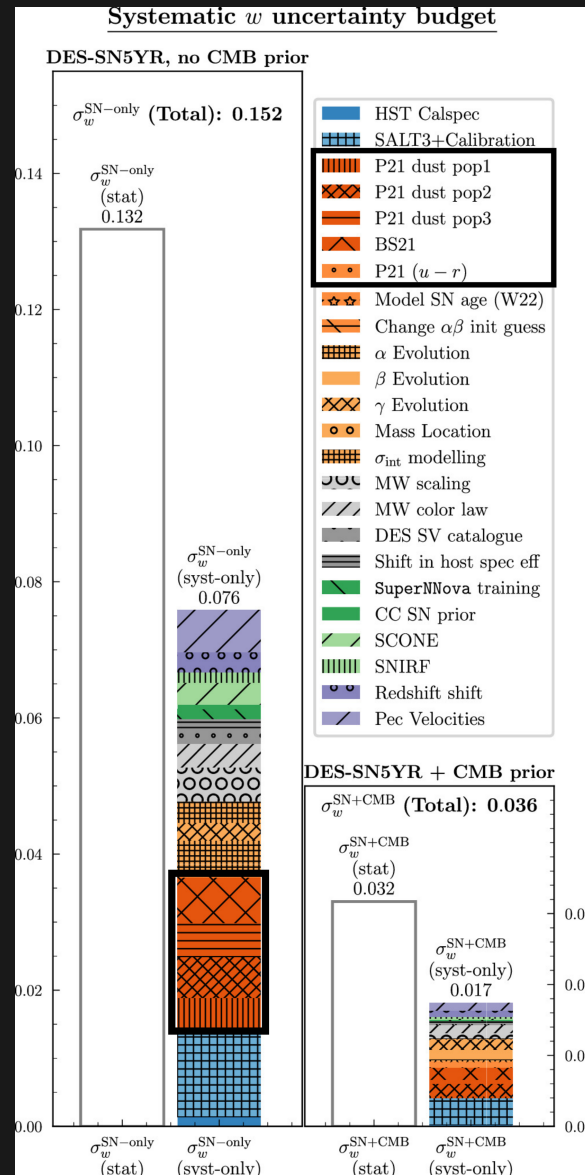
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The intrinsic scatter of SNe Ia

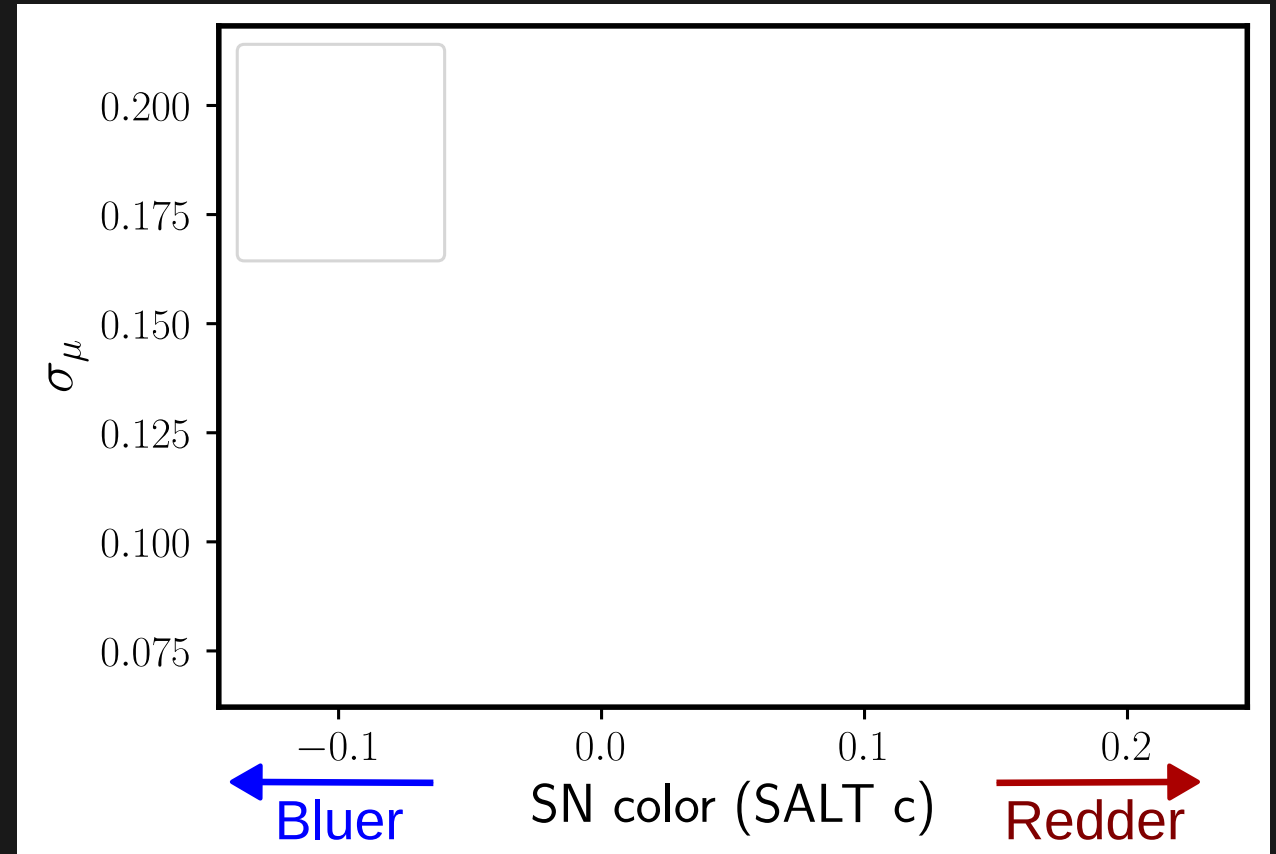
Main systematic of DES 5-year dark energy analysis (*Vincenzi et al. 2023*)!!!

What about $f\sigma_8$???



The intrinsic scatter of SNe Ia

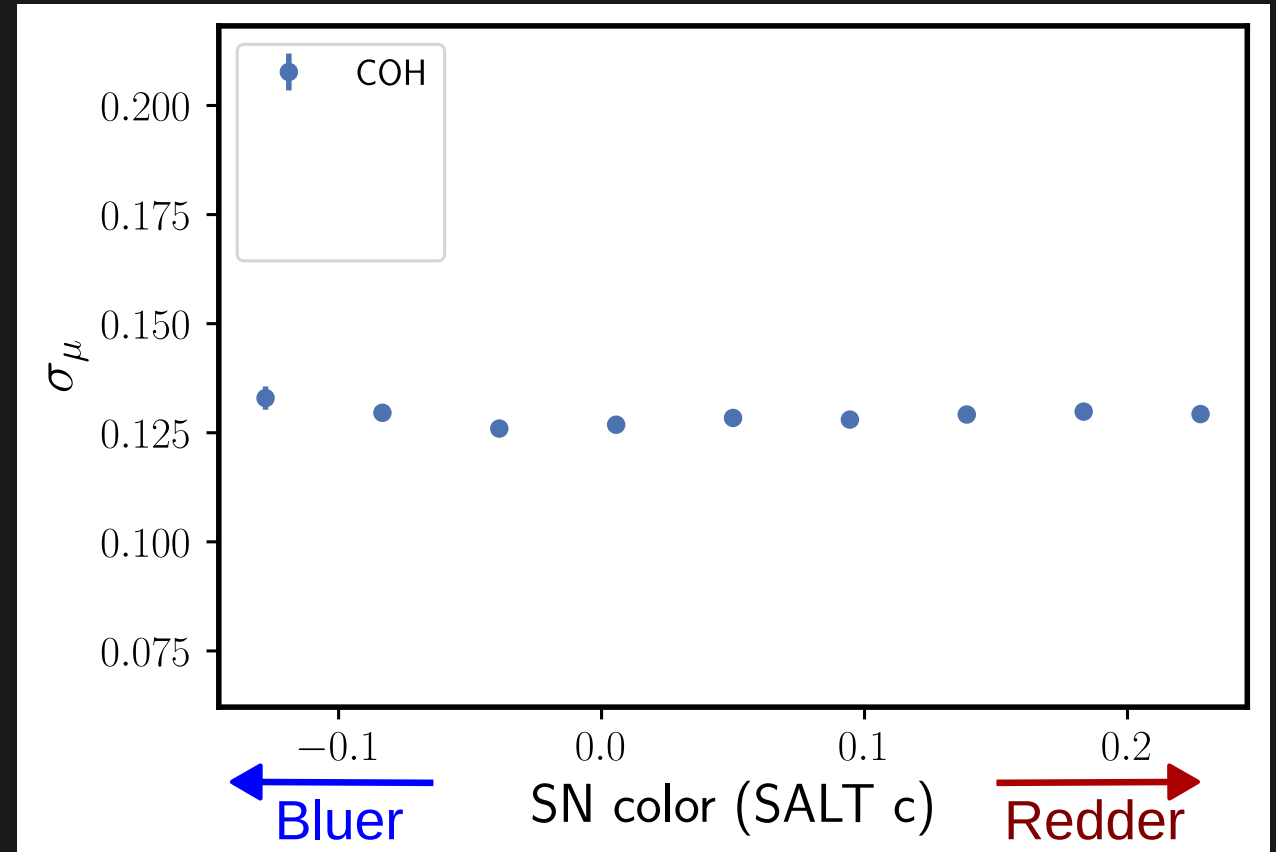
In this work we considered 4 models of intrinsic scatter



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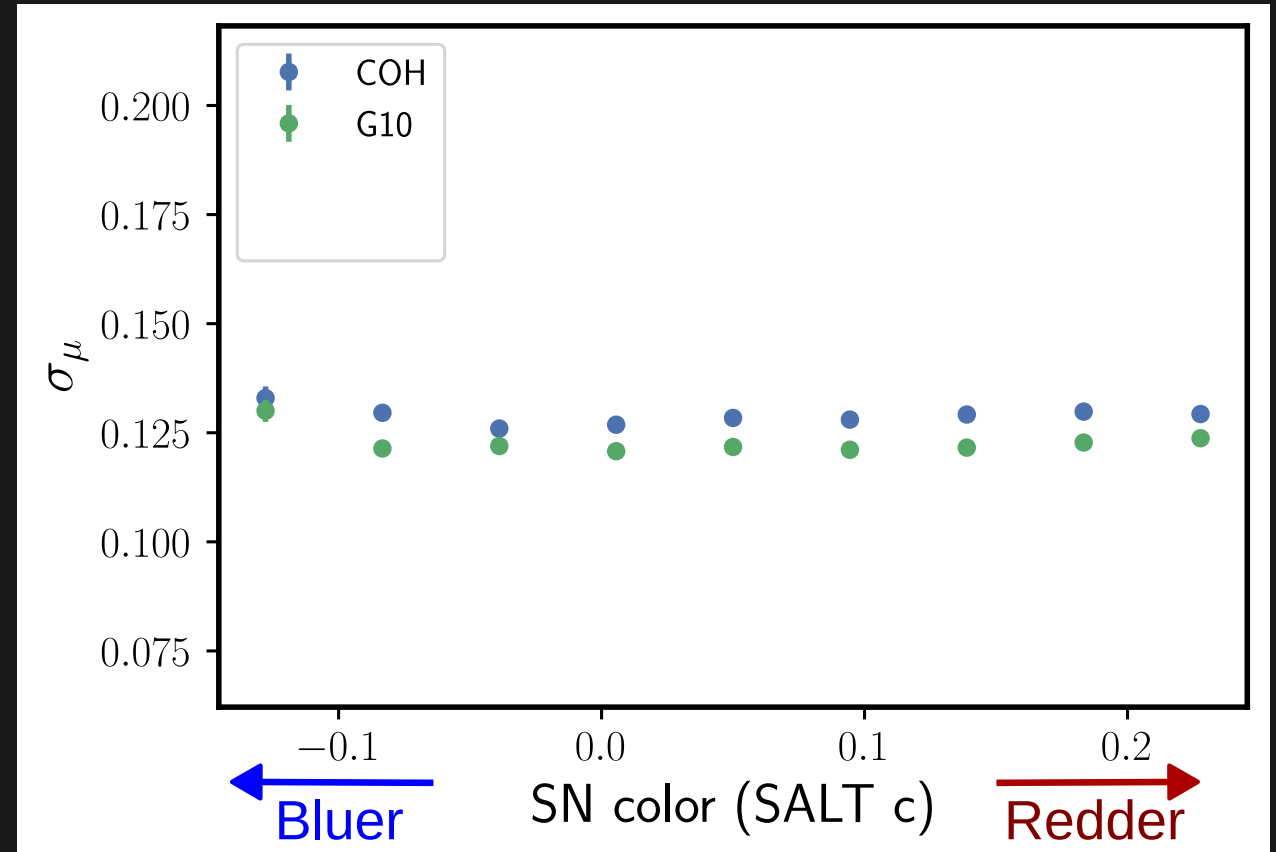
- Random coherent scatter (COH)
Achromatic
Unrealistic



The intrinsic scatter of SNe Ia

In this work we considered 4 models of intrinsic scatter

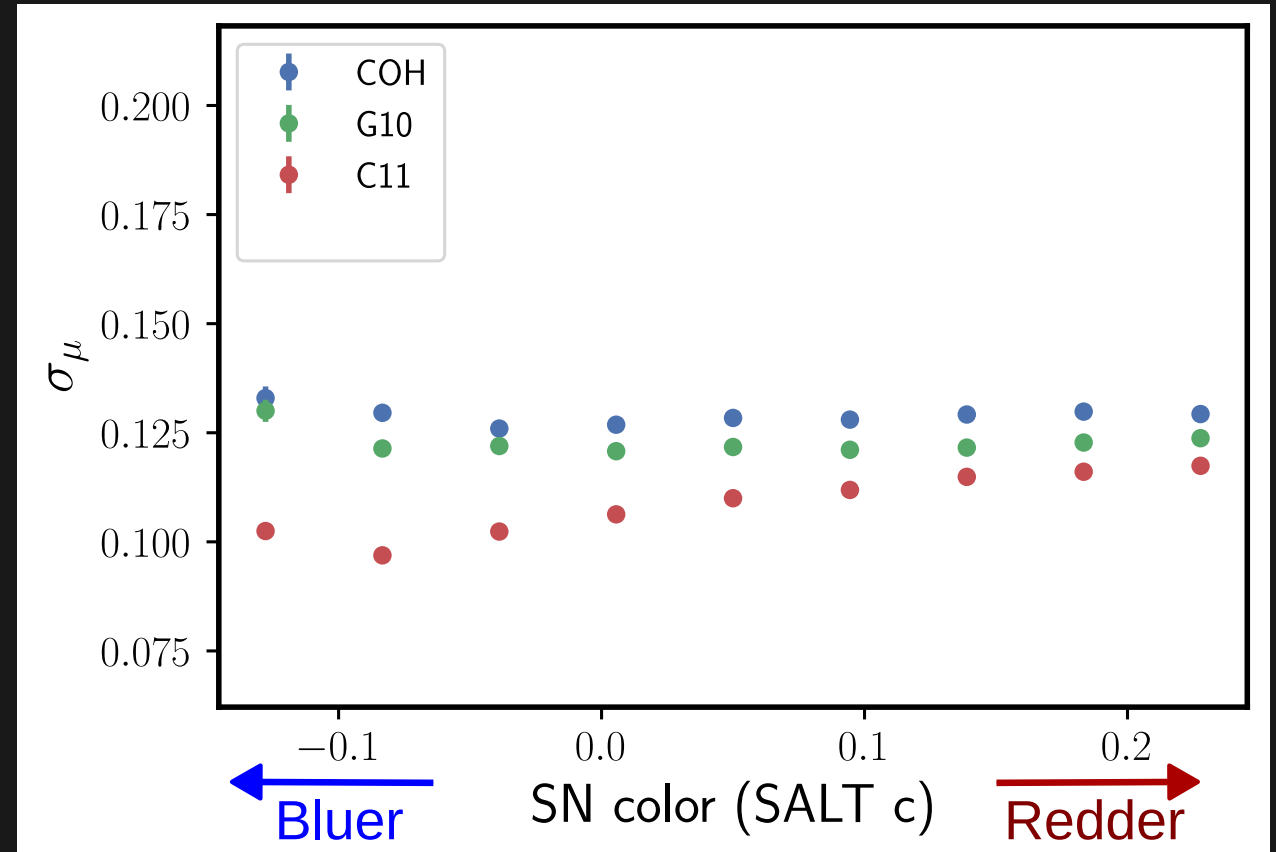
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Historically used (Pantheon, Pantheon+)



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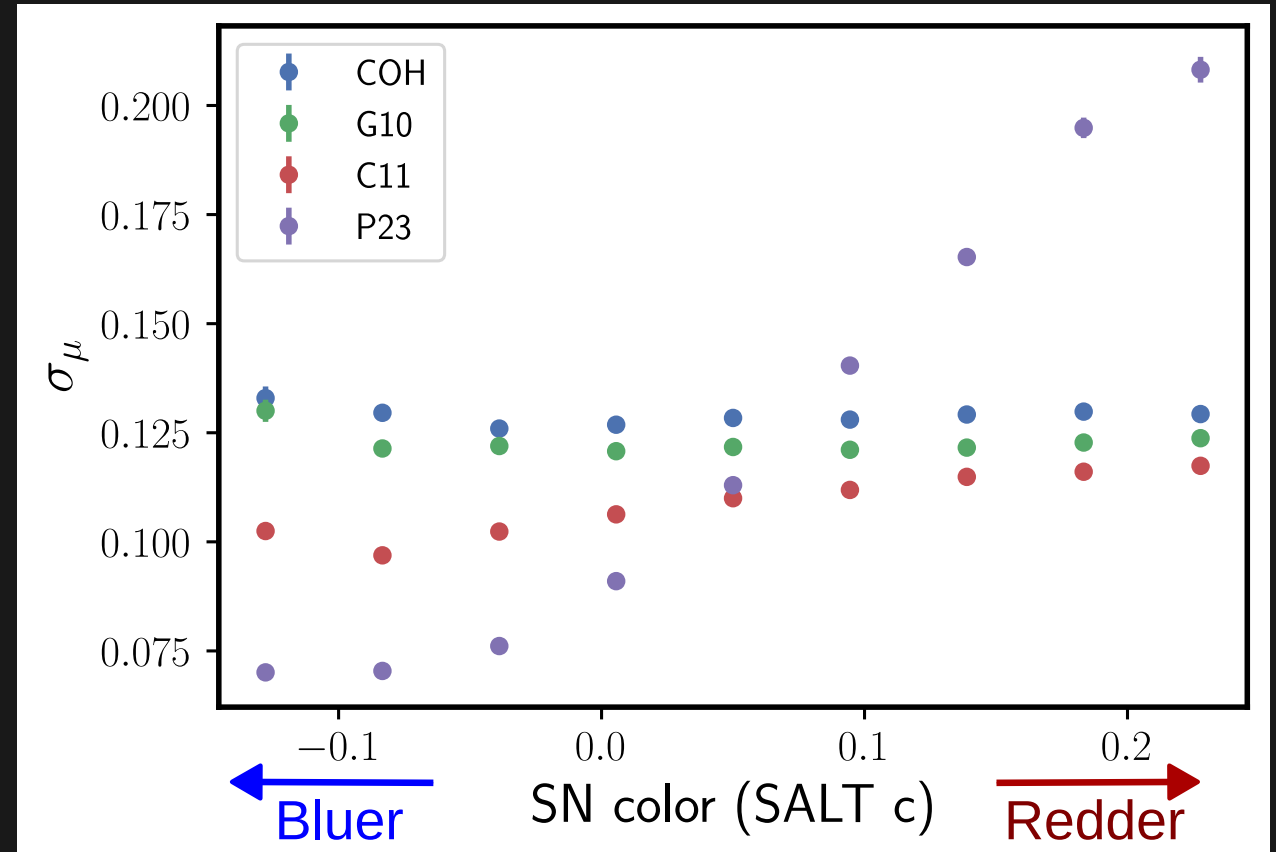
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The intrinsic scatter of SNe Ia

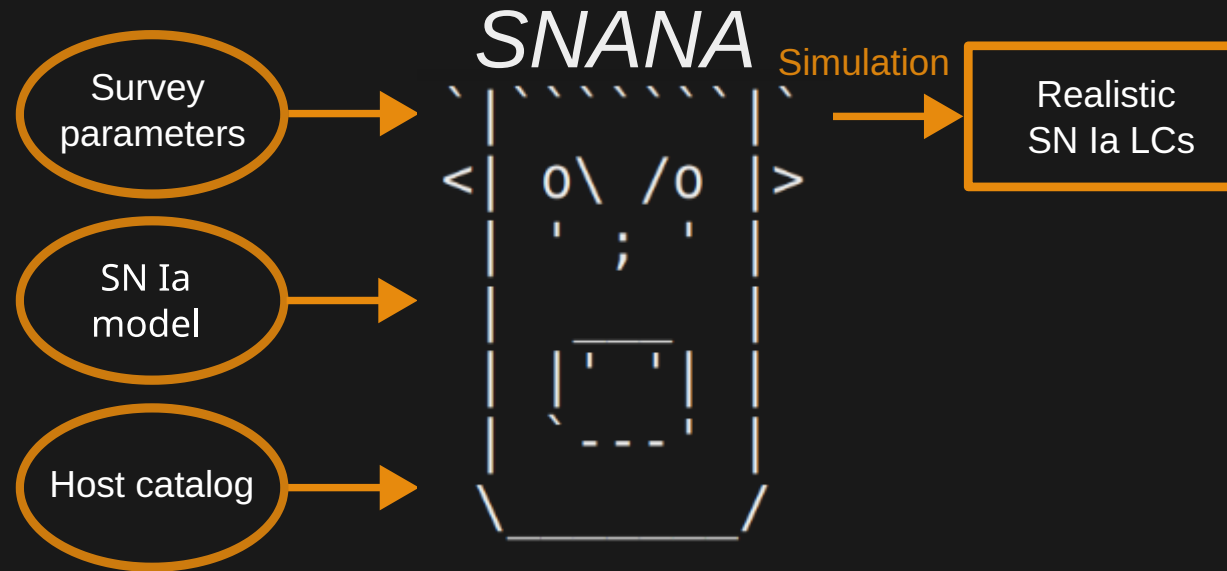
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Historically used (Pantheon, Pantheon+)
- The BS21 model (Brout & Scolnic 2021),
parameters from Popovic et al. 2023 (P23):
Dust-based model
Currently favored by data (DES 5-year)



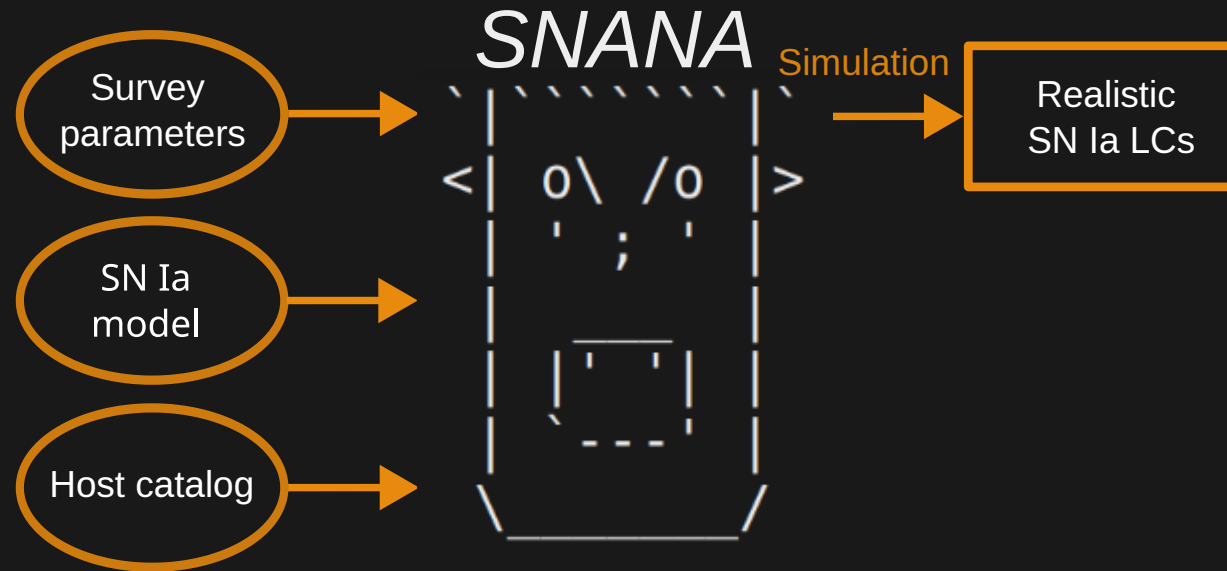
Rubin-LSST Simulations

We used the SNANA software (*Kessler et al. 2009*) to simulate the 10 years of the Rubin-LSST survey!



Rubin-LSST Simulations

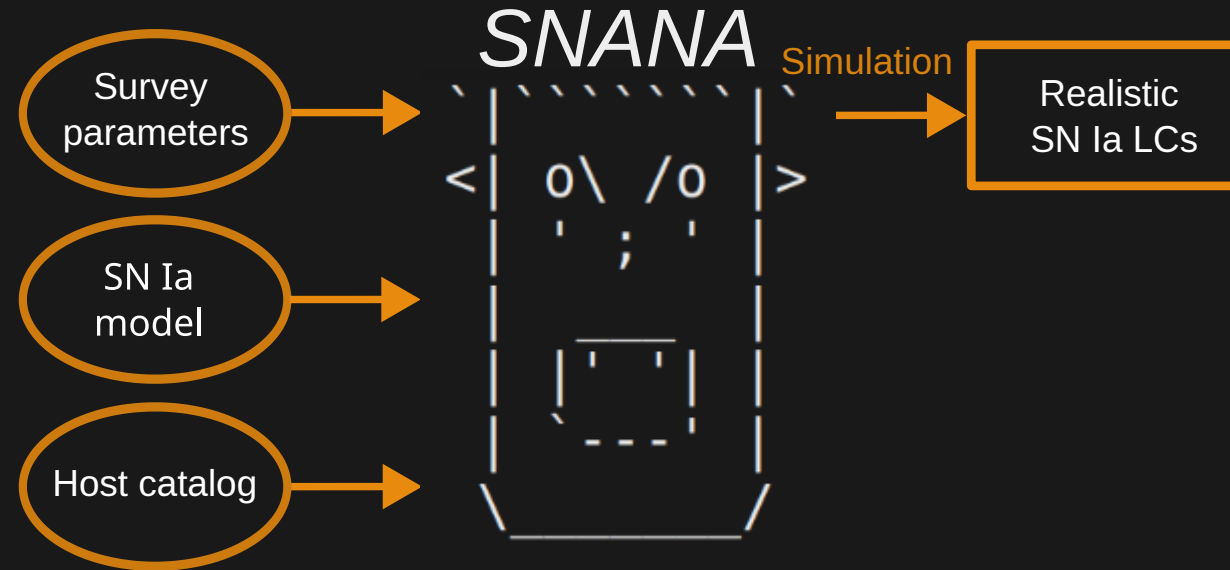
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- Survey parameters from LSST survey simulation (OpSim)

Rubin-LSST Simulations

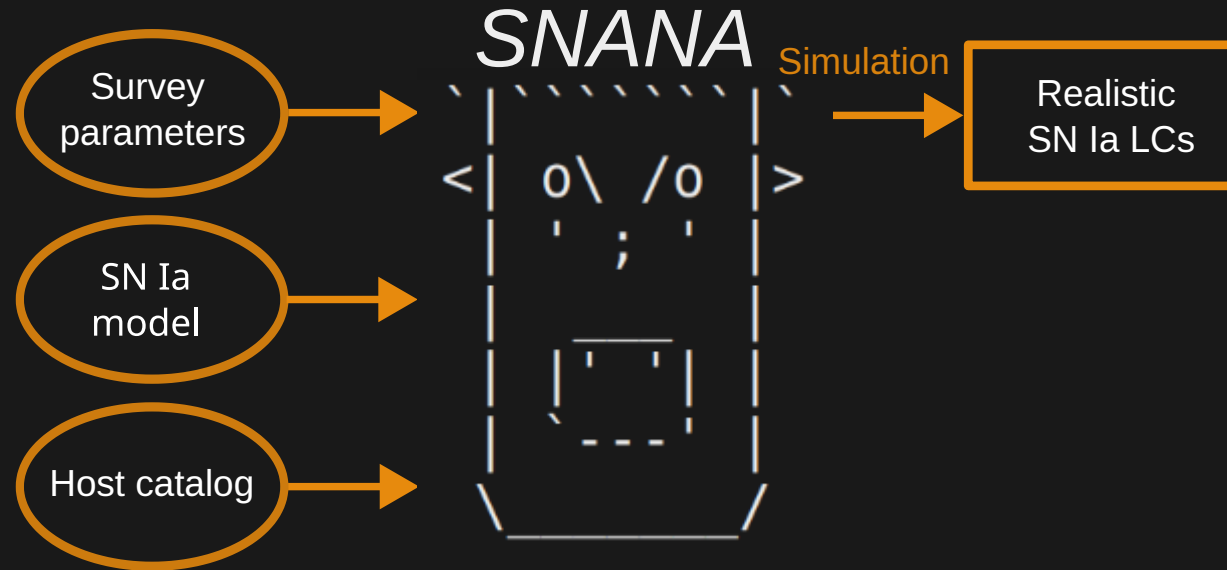
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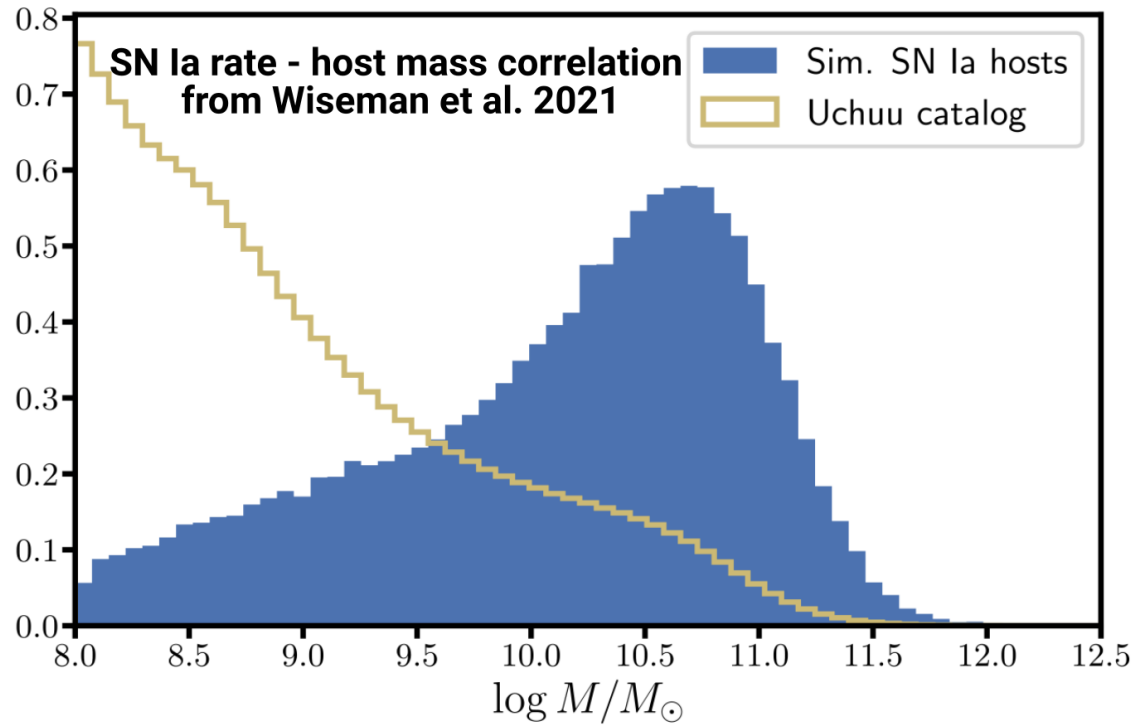
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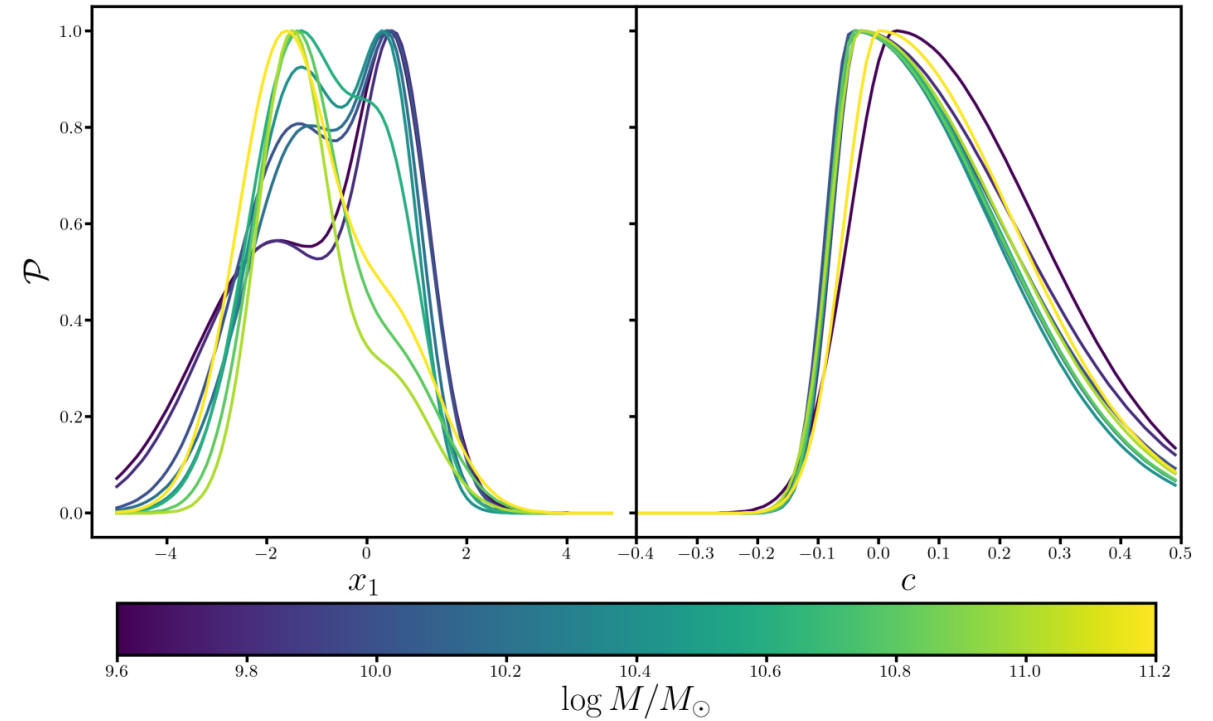


- Survey parameters from LSST survey simulation (OpSim)
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- Host catalog: Uchuu UniverseMachine N-body simulation (*Ishiyama et al. 2021, Aung et al. 2022*)

Simulation: Correlations SN Ia - host

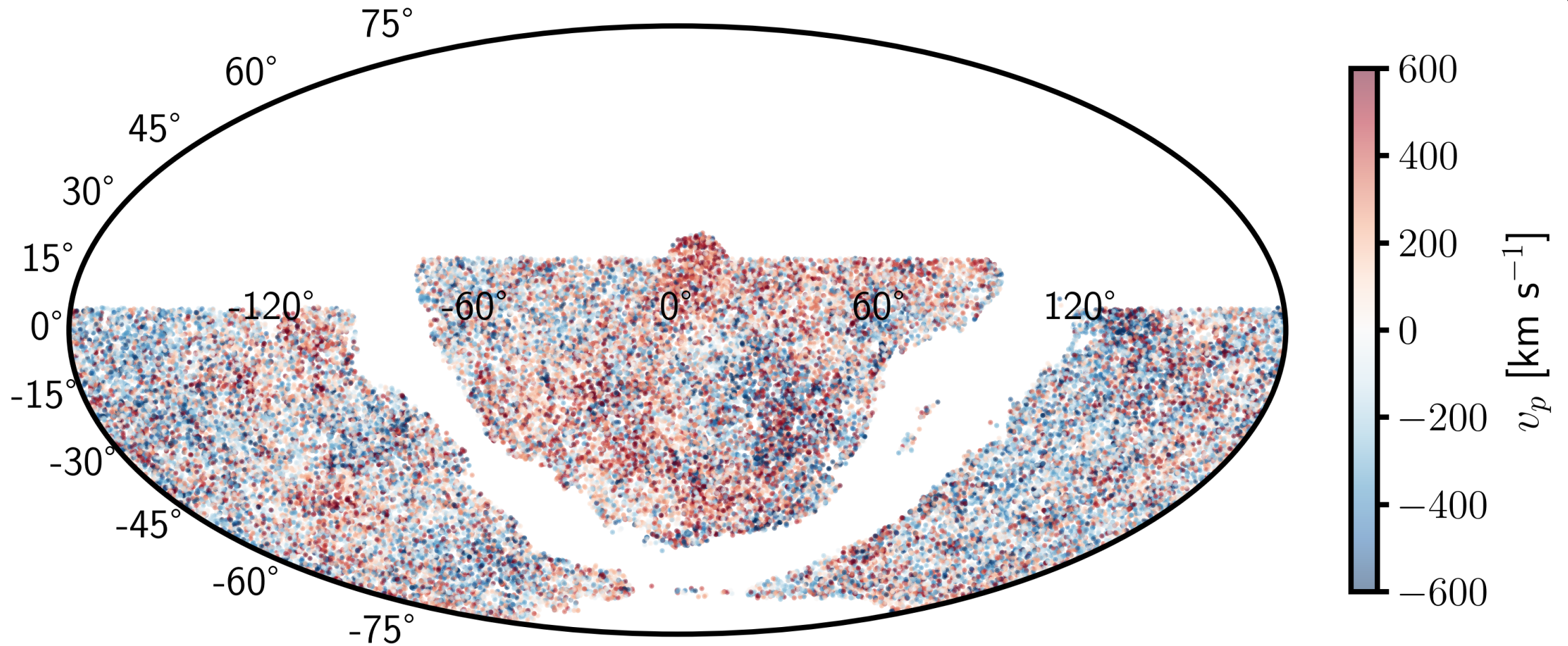


SALT parameters - host mass correlation from Popovic et al. 2021



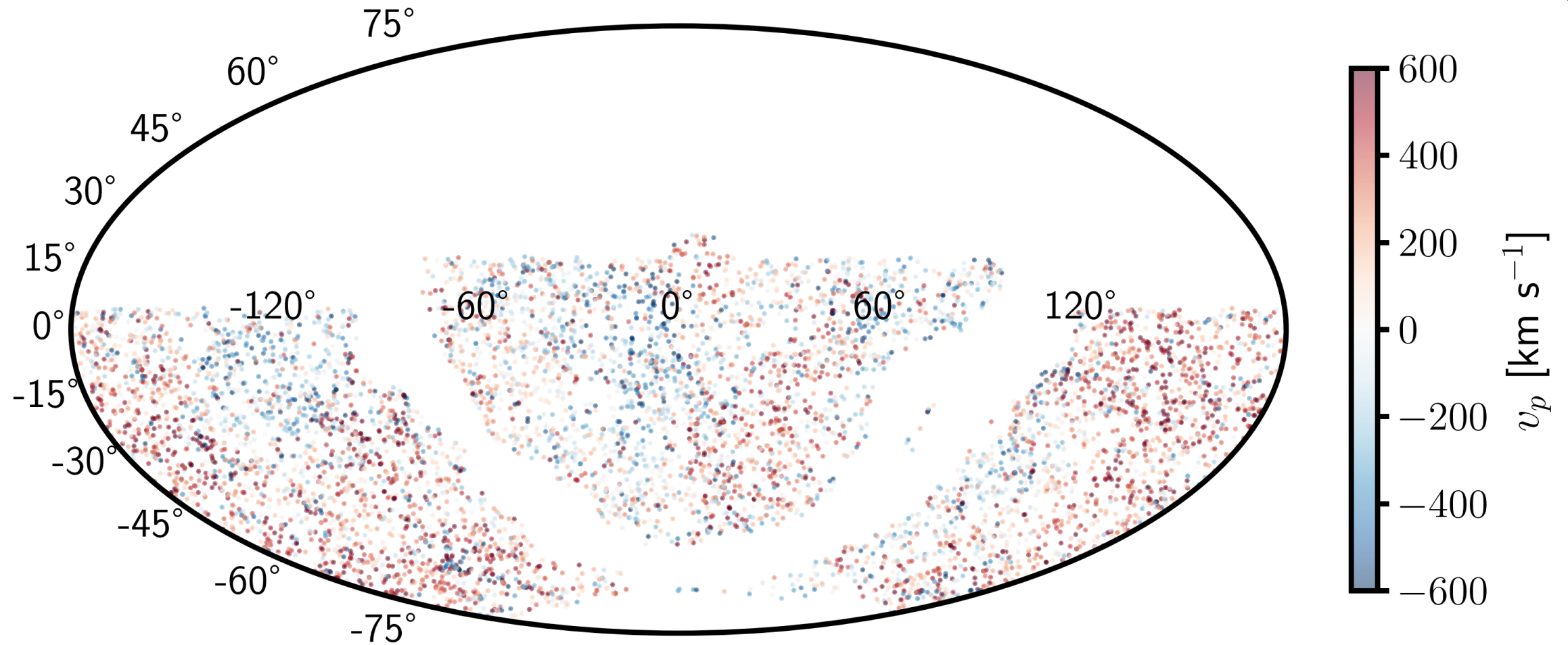
Simulation: the SNe Ia simulated sample

Simulation up to $z \sim 0.16 \Rightarrow N_{\text{SN}} \sim O(50\,000)$

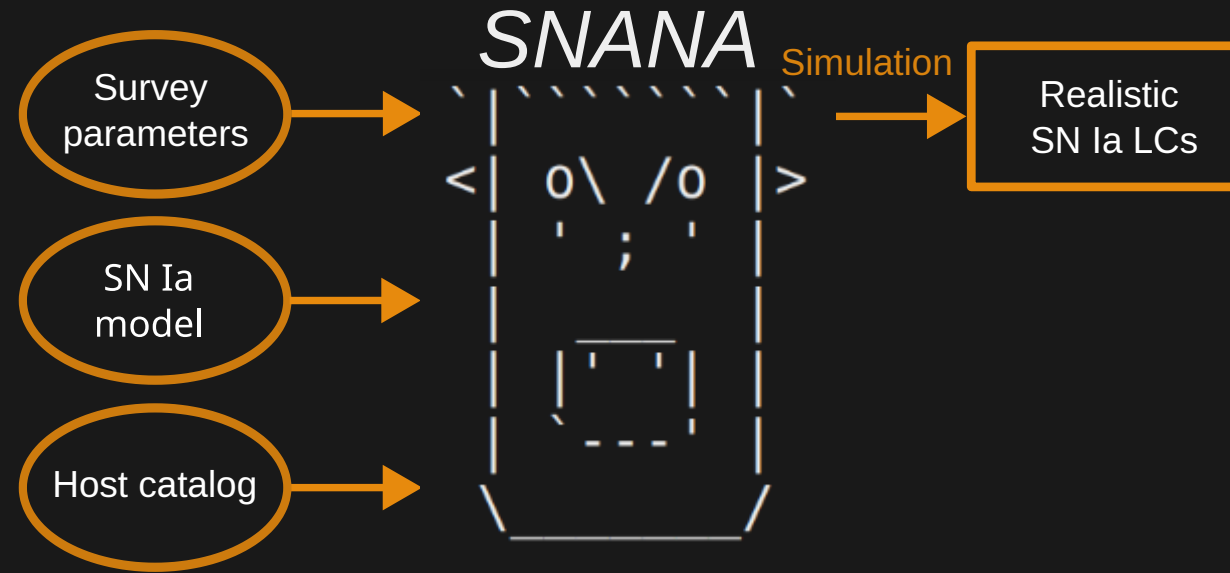


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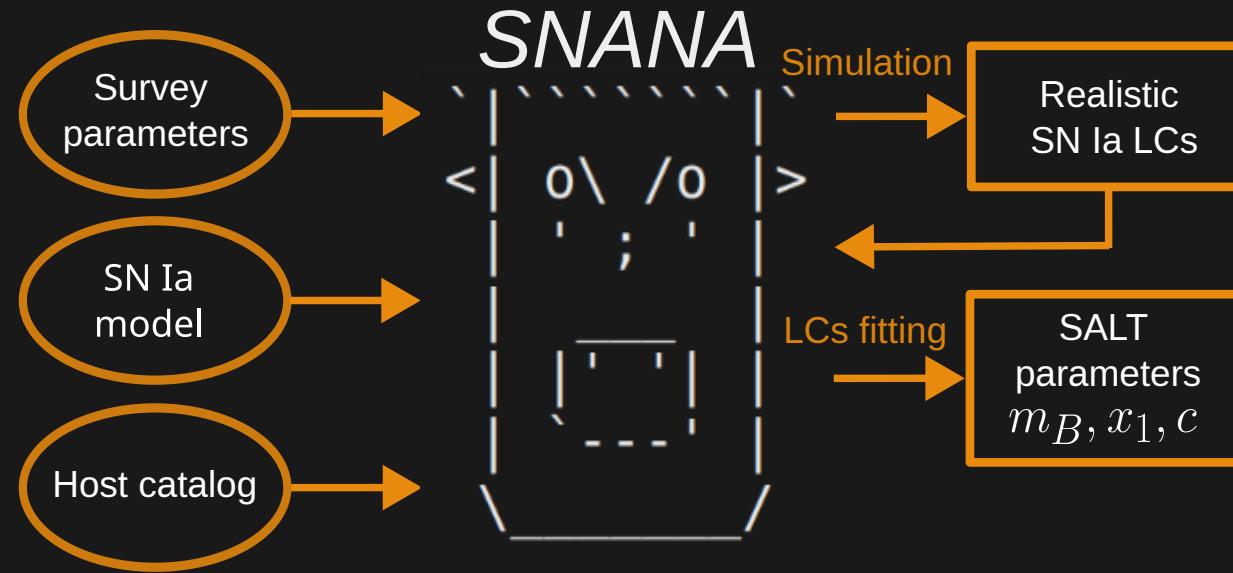
Using SNe Ia up to $z \sim 0.1 \Rightarrow N_{\text{SN}} \sim O(6\,600)$



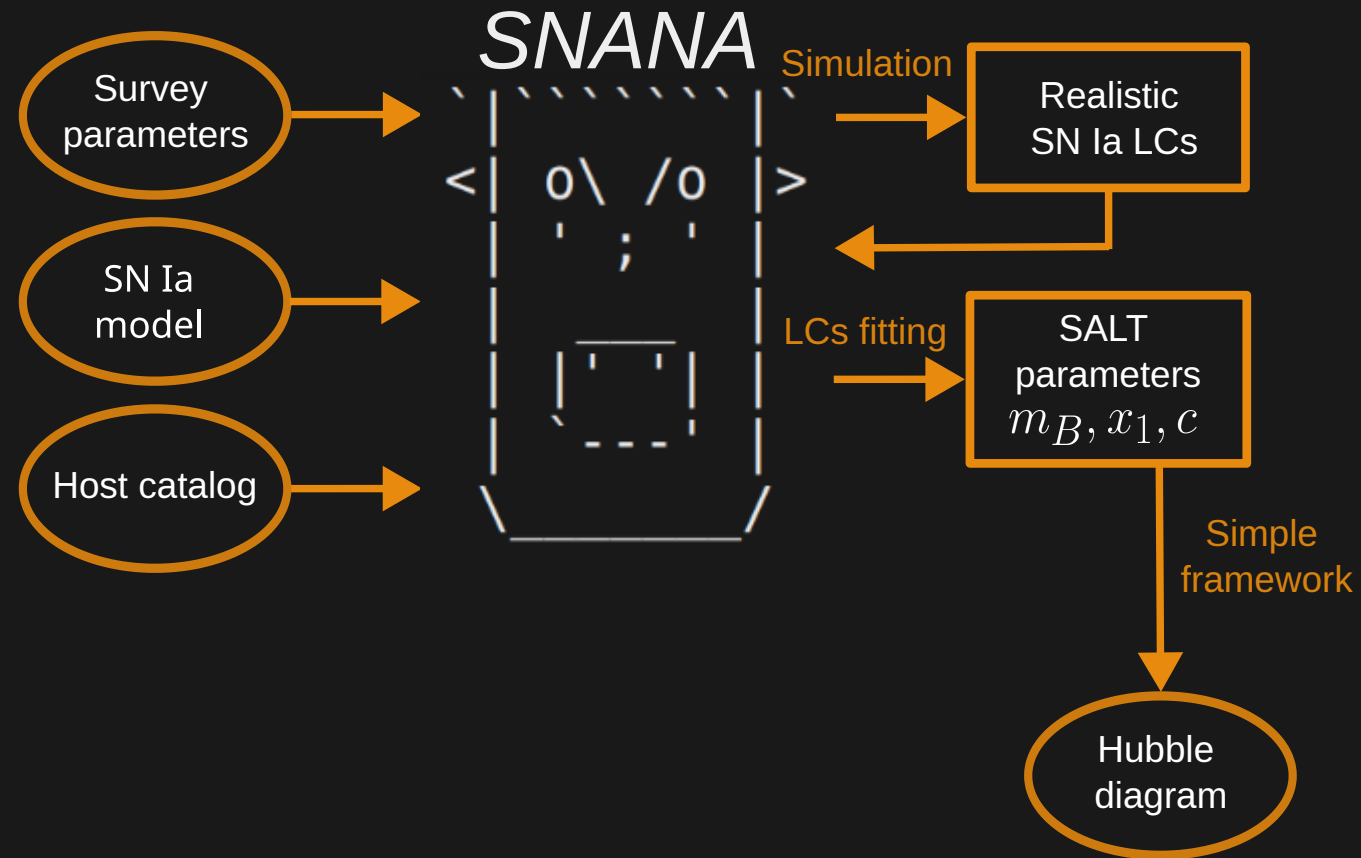
Building the Hubble diagram



Building the Hubble diagram



Building the Hubble diagram



Building the Hubble diagram: simple framework

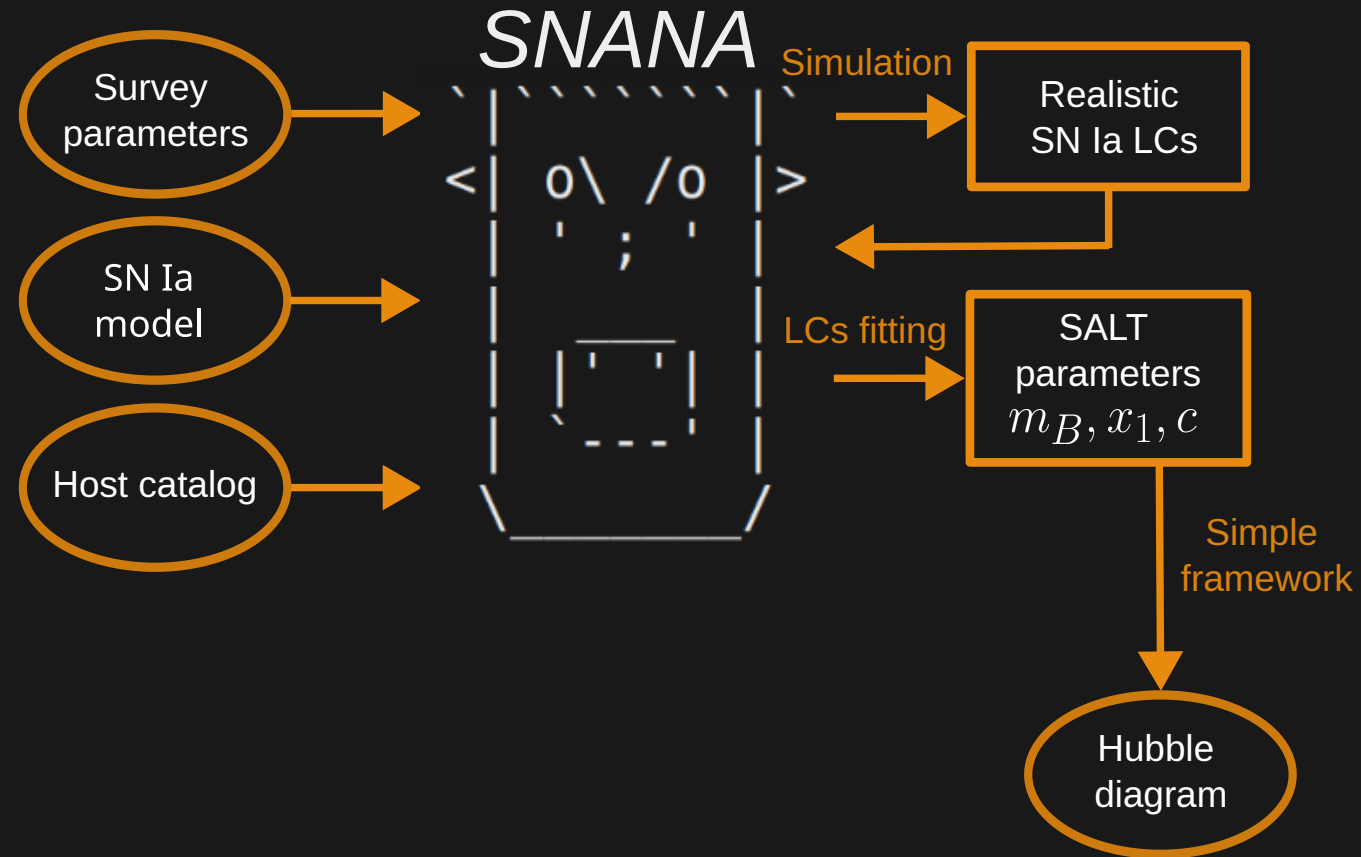
Fit for the Tripp relation along $f\sigma_8$:

$$\mu_{\text{obs}} = m_B - (M_0 - \alpha x_1 + \beta c + \Delta_M(\mathbf{M}_{\text{host}}; \gamma))$$

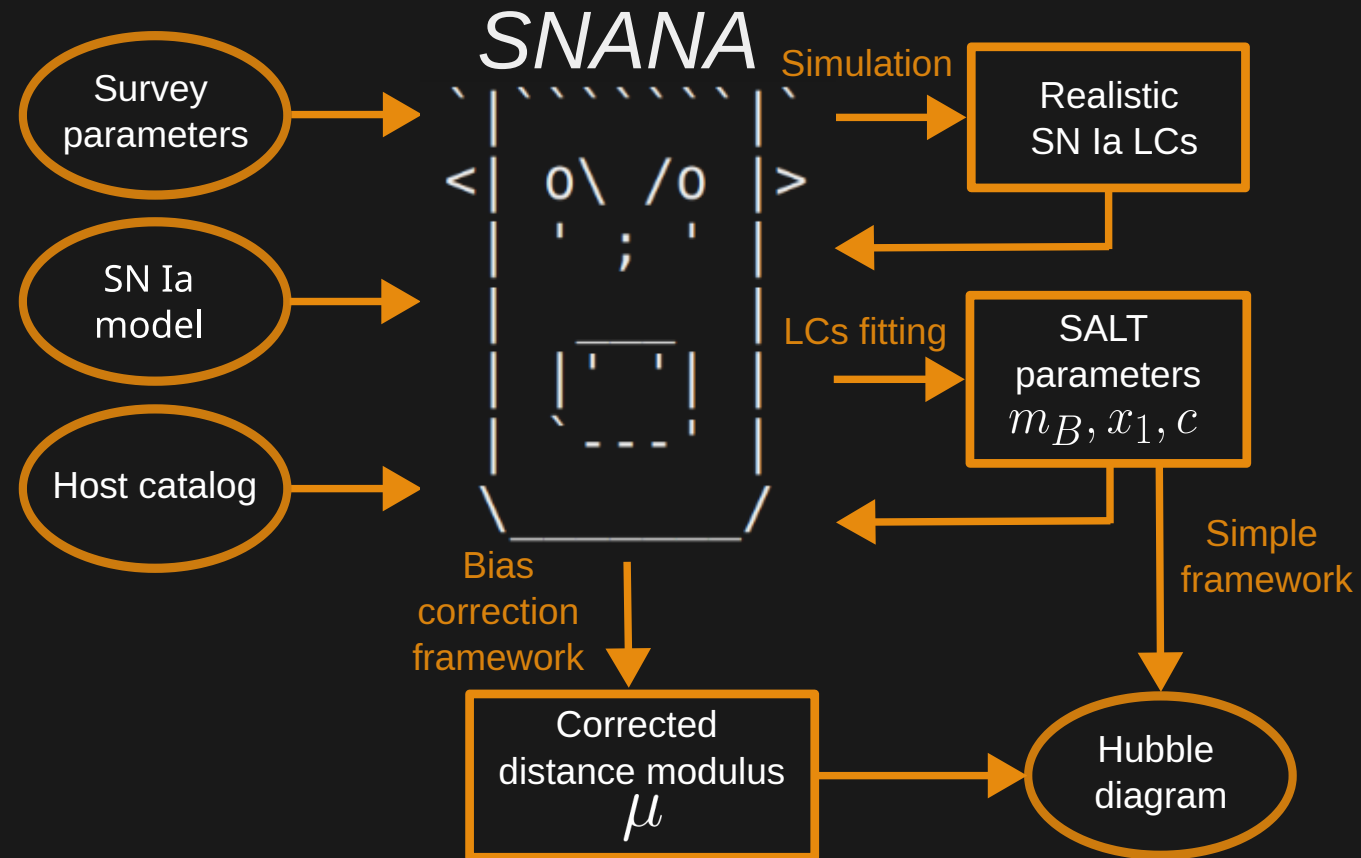
$$\sigma_{\mu}^2 = \sigma_{\text{obs}}^2 + \sigma_{\text{int}}^2$$

$M_0, \alpha, \beta, \gamma$ and σ_{int} are free parameters.

Building the Hubble diagram



Building the Hubble diagram



Building the Hubble diagram: BBC framework

$$\mu_{\text{obs,BBC}} = m_B - (\textcolor{red}{M}_0 - \textcolor{red}{\alpha}x_1 + \textcolor{red}{\beta}c + \Delta_M(\mathbf{M}_{\text{host}}; \textcolor{red}{\gamma})) + \delta_{\text{corr.}}$$

$$\sigma_{\mu}^2 = \sigma_{\text{obs}}^2 + \textcolor{red}{\sigma}_{\text{int}}^2$$

Building the Hubble diagram: BBC framework

$$\mu_{\text{obs,BBC}} = m_B - (\textcolor{red}{M}_0 - \textcolor{red}{\alpha}x_1 + \textcolor{red}{\beta}c + \Delta_M(\mathbf{M}_{\text{host}}; \textcolor{red}{\gamma})) + \delta_{\text{corr.}}$$

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$\delta_{\text{corr.}}$ is obtained by:

Building the Hubble diagram: BBC framework

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- Running an extra-large simulation ($\sim 40 \times$ LSST) and fitting the Hubble diagram

Building the Hubble diagram: BBC framework

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Building the Hubble diagram: BBC framework

$$\mu_{\text{obs,BBC}} = m_B - (\textcolor{red}{M}_0 - \textcolor{red}{\alpha}x_1 + \textcolor{red}{\beta}c + \Delta_M(\mathbf{M}_{\text{host}}; \textcolor{red}{\gamma})) + \delta_{\text{corr.}}$$

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- Computing the correction in each cell $\delta_{\text{corr.}} = \langle \mu_{\text{obs}} - \mu_{\text{fid}} \rangle_{\text{cell}}$

Building the Hubble diagram: BBC framework

$$\mu_{\text{obs,BBC}} = m_B - (\textcolor{red}{M}_0 - \textcolor{red}{\alpha}x_1 + \textcolor{red}{\beta}c + \Delta_M(\mathbf{M}_{\text{host}}; \textcolor{red}{\gamma})) + \delta_{\text{corr.}}$$

$$\sigma_{\mu}^2 = \sigma_{\text{obs}}^2 + \textcolor{red}{\sigma}_{\text{int}}^2$$

$\delta_{\text{corr.}}$ is obtained by:

- Running an extra-large simulation ($\sim 40 \times$ LSST) and fitting the Hubble diagram
- Binning over the parameters $p = \{z_{\text{obs}}, x_1, c, M_{\text{host}}\}$
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- Interpolate over the cells to obtain $\delta_{\text{corr.}}(p)$

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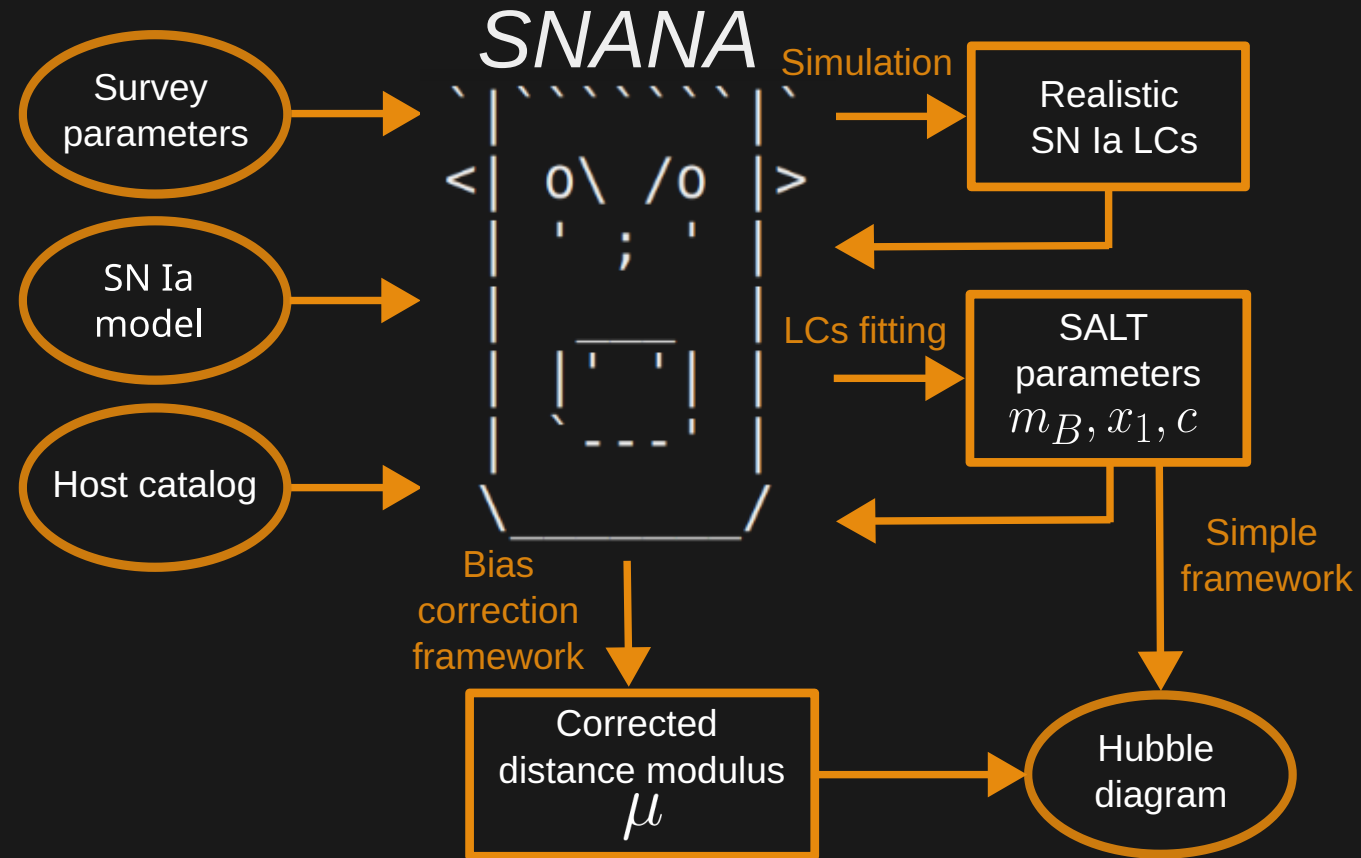
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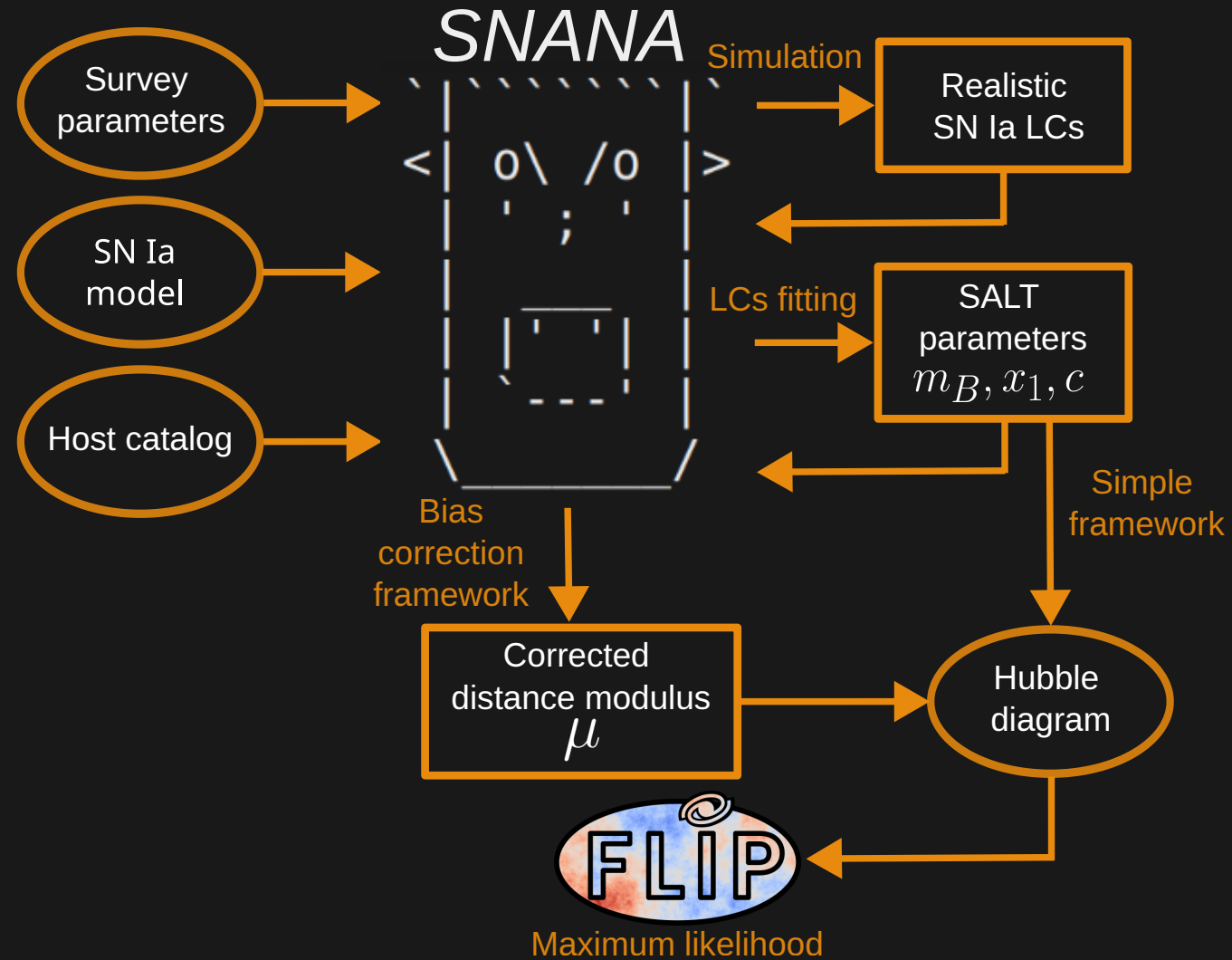
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- Interpolate over the cells to obtain $\delta_{\text{corr.}}(p)$

α, β, γ and σ_{int} are fitted prior to $f\sigma_8$

The maximum likelihood method



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The Maximum likelihood method is implemented within the  package (*Ravoux, Carreres et al. 2025*)

We want to maximize the likelihood function:

$$\mathcal{L}(f\sigma_8; v_p) \propto (2\pi)^{-\frac{N}{2}} |\mathbf{C}(f\sigma_8)|^{-\frac{1}{2}} \exp \left(-\frac{1}{2} \mathbf{v}_p^T \mathbf{C}(f\sigma_8)^{-1} \mathbf{v}_p \right)$$

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$$\mathbf{C}(f\sigma_8) = \mathbf{C}^{\text{obs}} + \mathbf{C}^{vv}(f\sigma_8)$$

The maximum likelihood method: velocity covariance

The covariance of the velocity field is:

$$\langle v_i(\mathbf{r}_i) v_j(\mathbf{r}_j) \rangle = C_{ij}^{vv} \propto (f\sigma_8)^2 \int_{k_{\min}}^{k_{\max}} P(k) W_{ij}(k; \mathbf{r}_i, \mathbf{r}_j) dk$$

$\mathcal{L}(p) \propto \exp \left[-\frac{1}{2} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}^T \begin{bmatrix} C_{12} & C_{21} & & \\ & C_{22} & & \\ & & C_{33} & \\ & & & C_{44} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} \right]$

$^{-1} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$

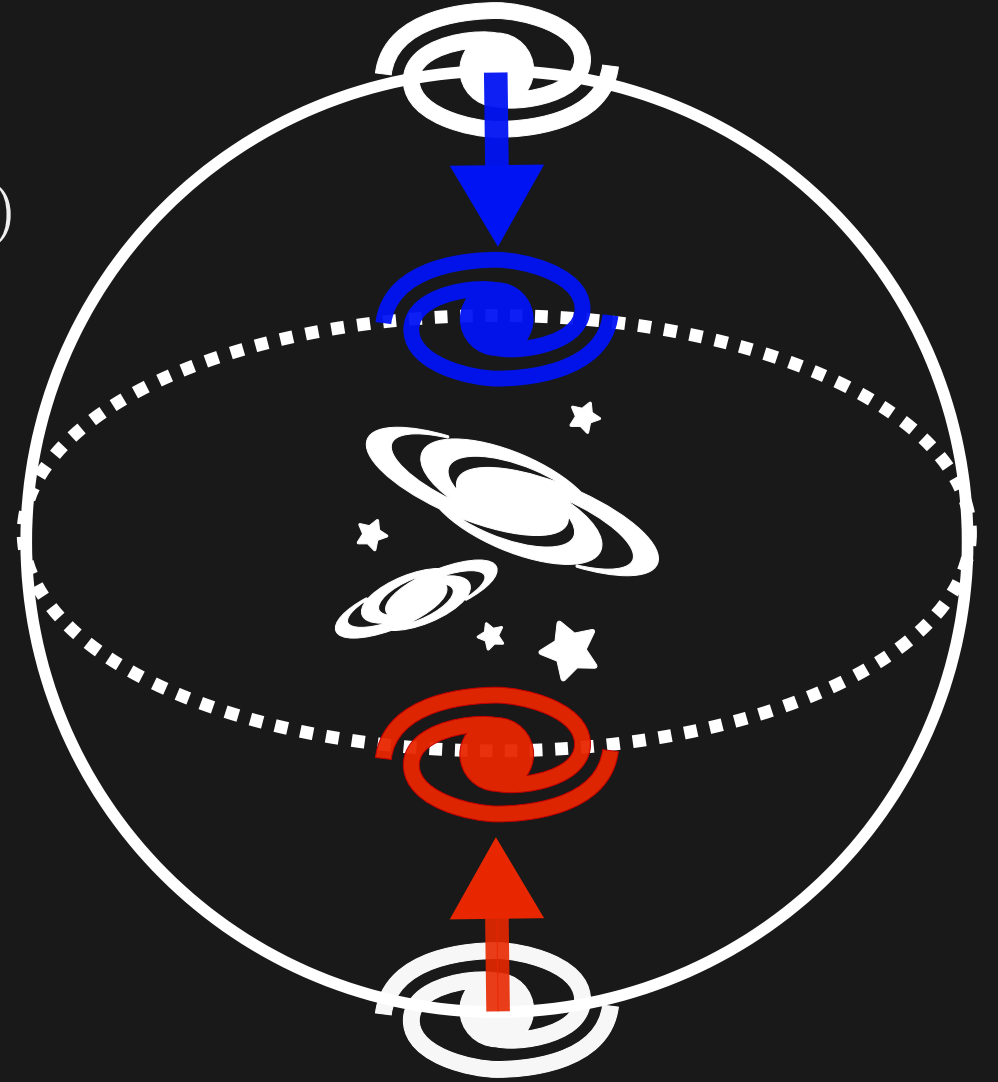
Credits: Corentin Ravoux

The σ_u redshift space parameter

Positions are evaluated using $z_{\text{obs}} \Rightarrow$ Redshift Space Distorsion

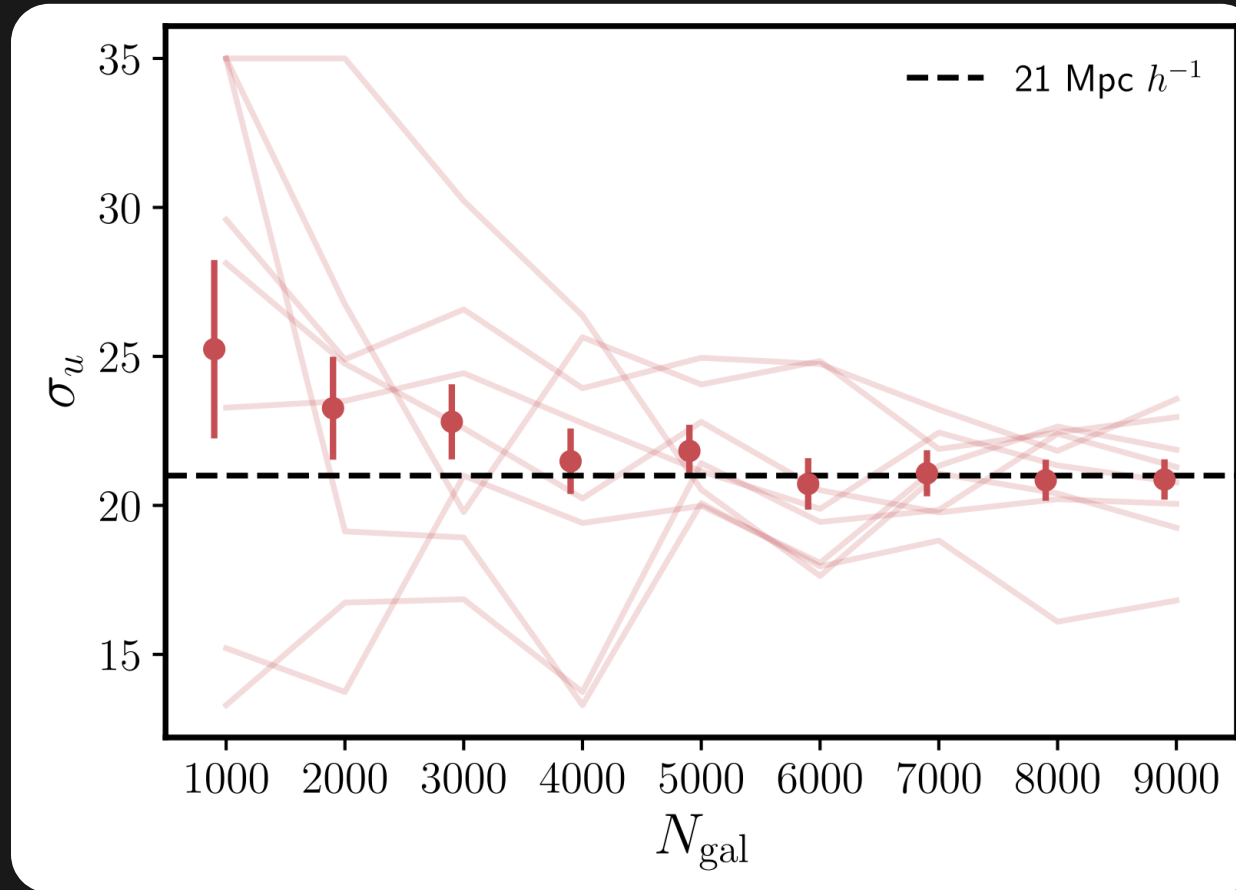
Empirical damping introduced in *Koda et al. 2014*: $D_u = \text{sinc}(k\sigma_u)$

$$C_{ij}^{vv} \propto (f\sigma_8)^2 \int_{k_{\min}}^{k_{\max}} P(k) D_u(k, \sigma_u)^2 W_{ij}(k; \mathbf{r}_i, \mathbf{r}_j) dk$$

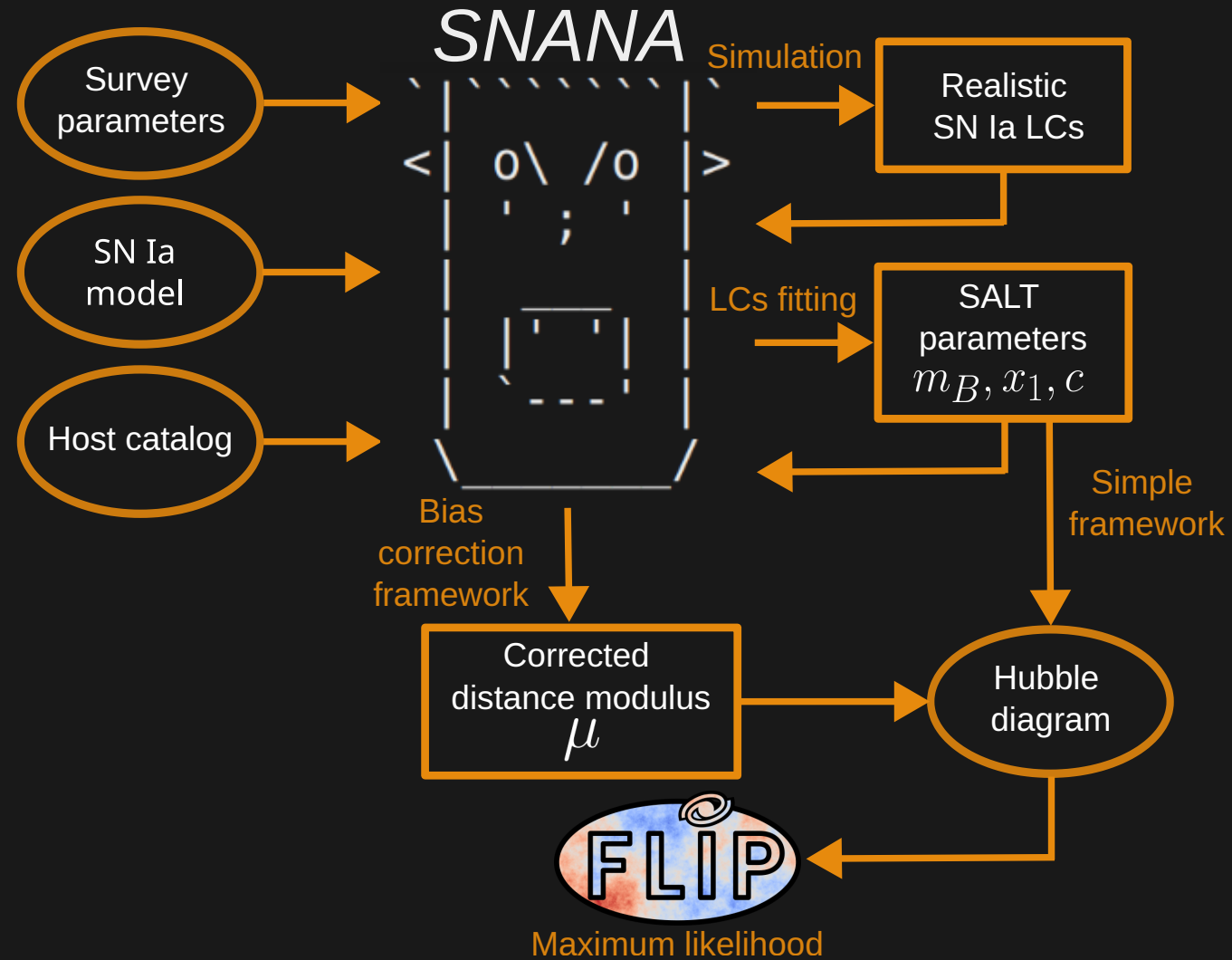


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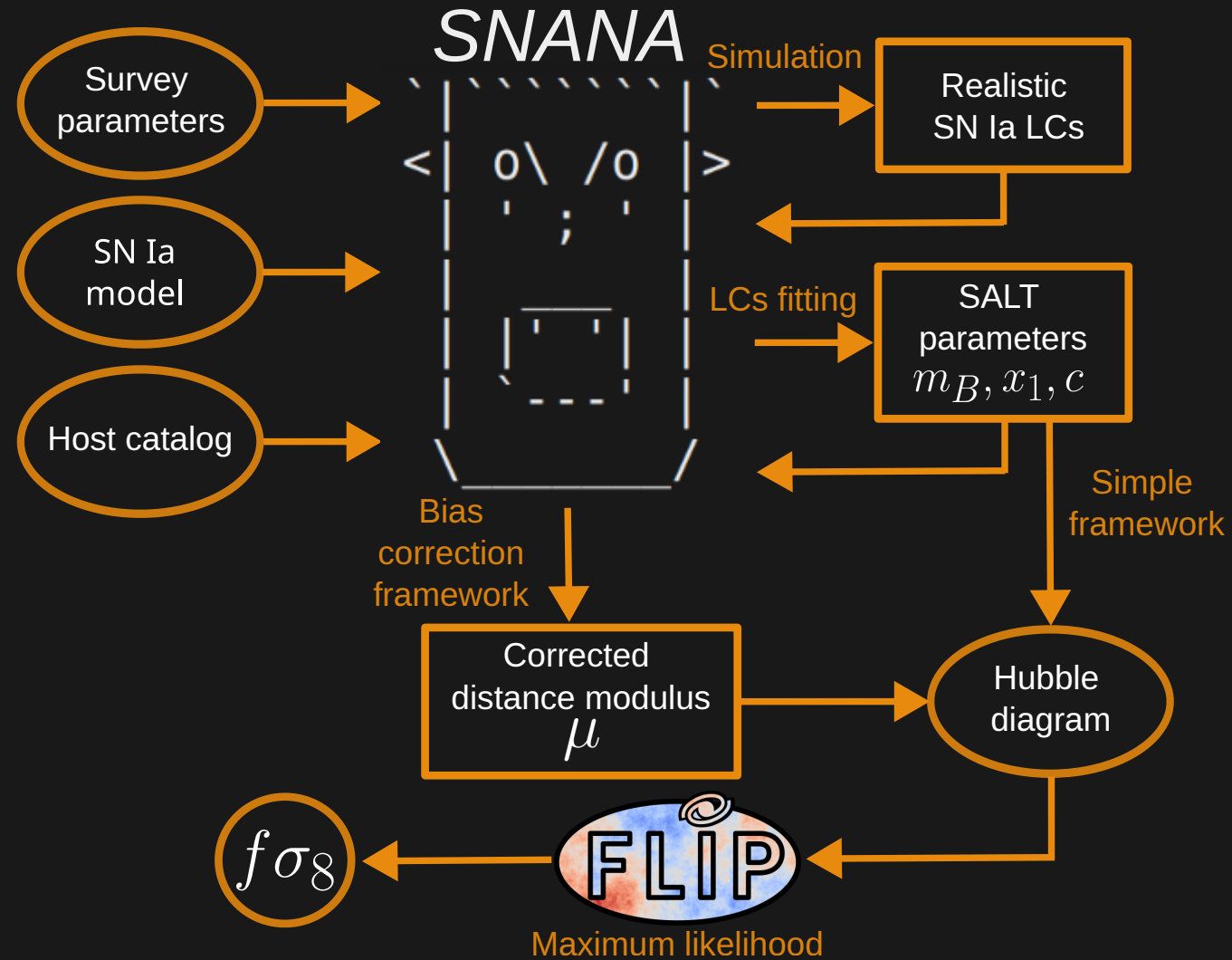
From a fit of true vel. from randomly sampled galaxies of the Uchuu simulation we found $\sigma_u \simeq 21 \text{ Mpc } h^{-1}$



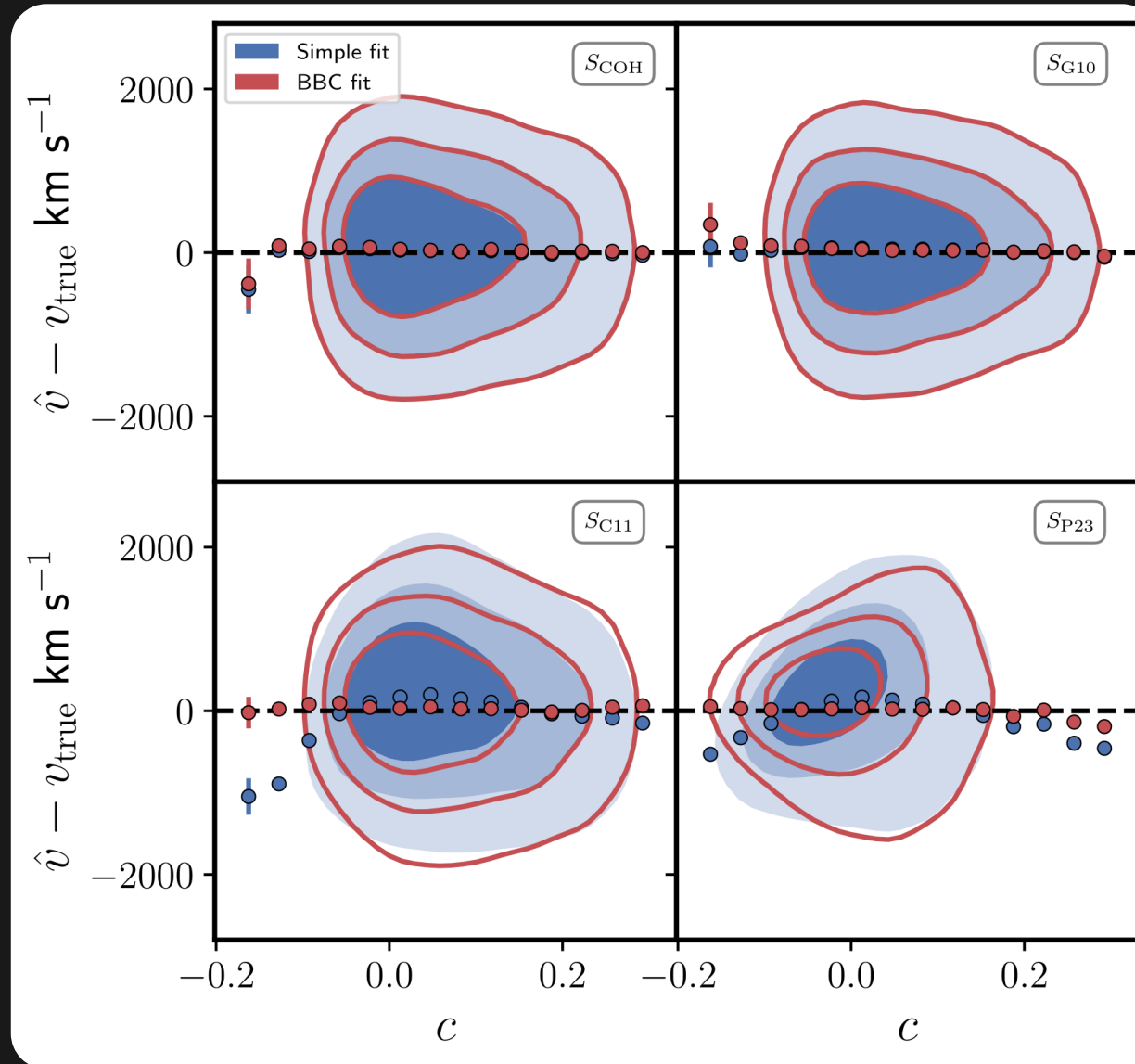
Fitting for $f\sigma_8$



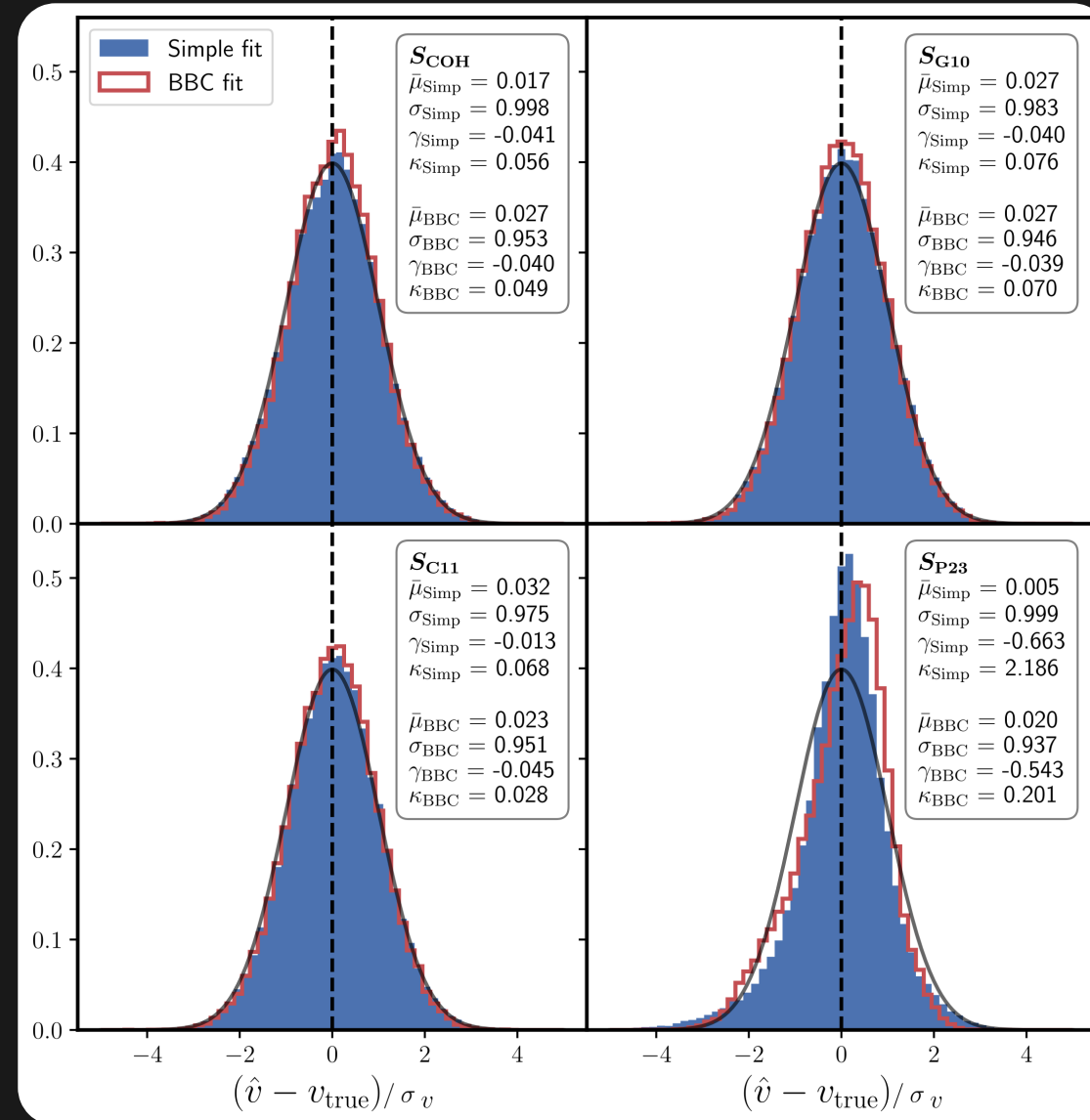
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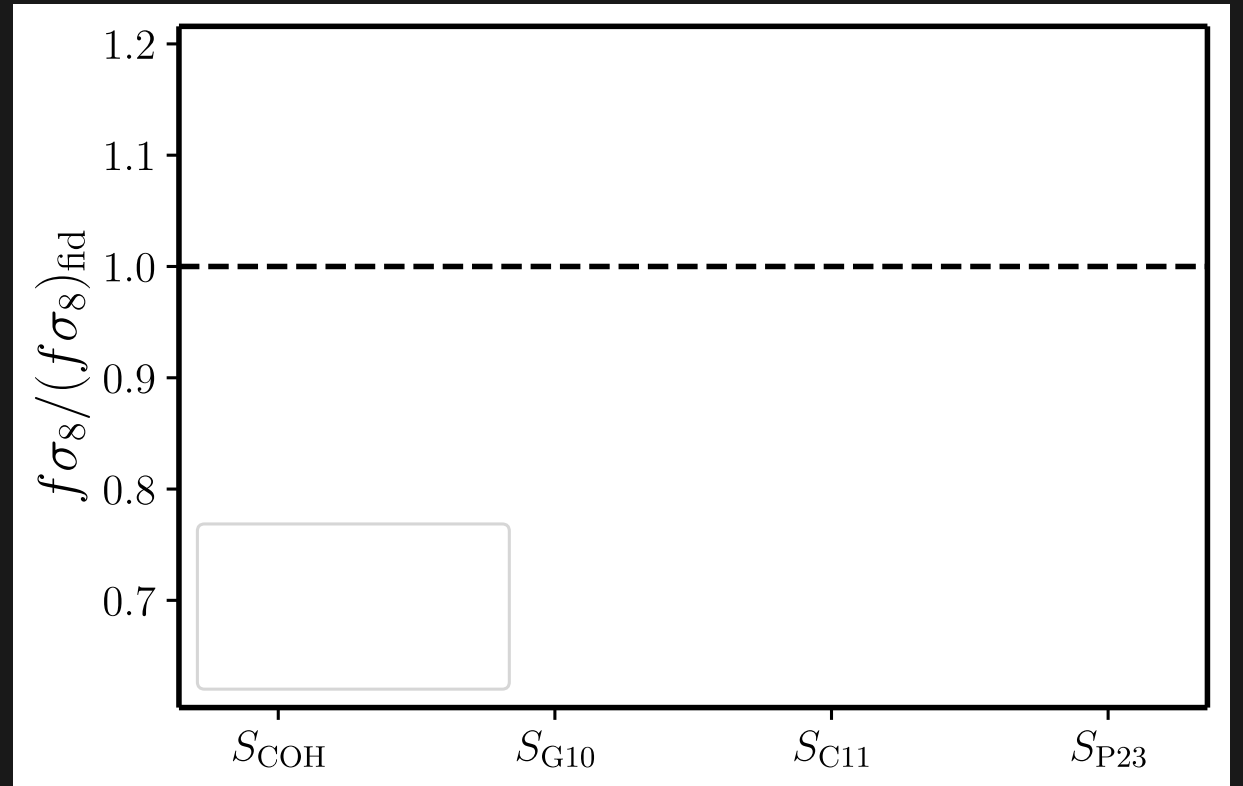
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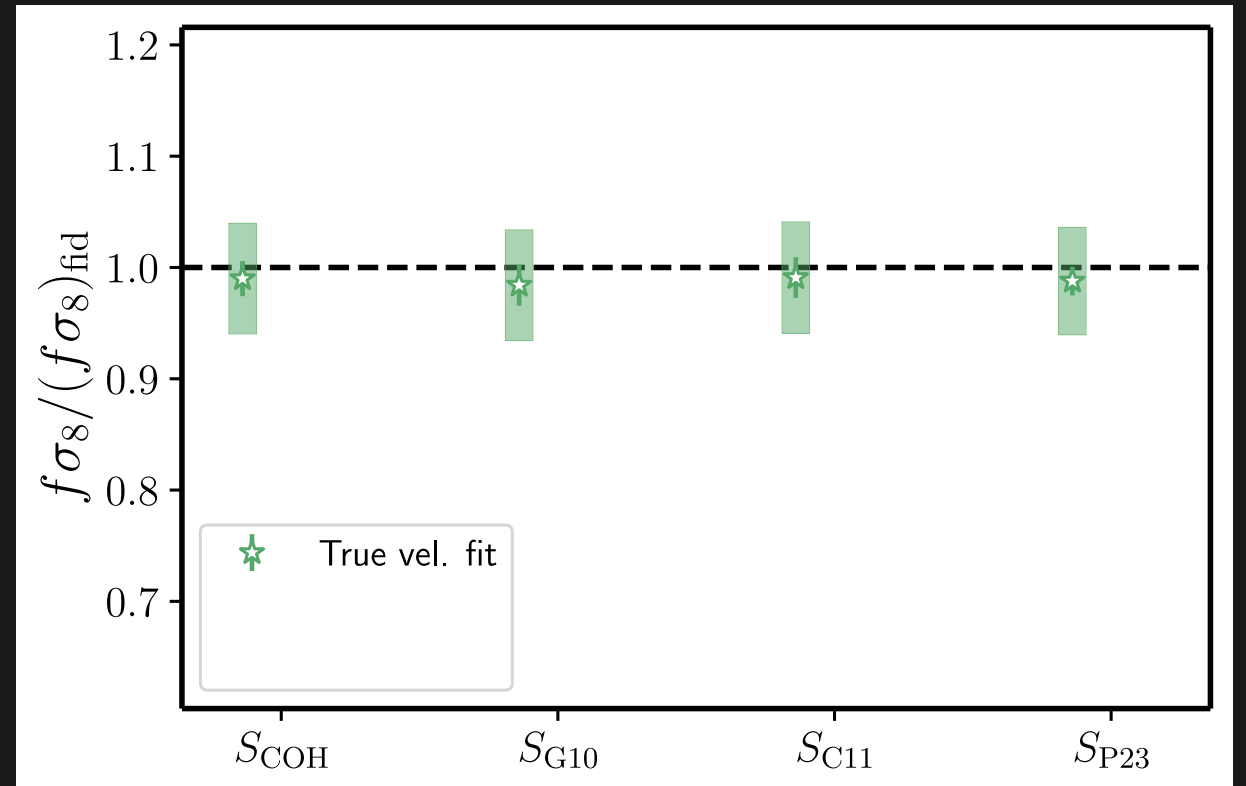


Results: $f\sigma_8$ fit for different scatter models



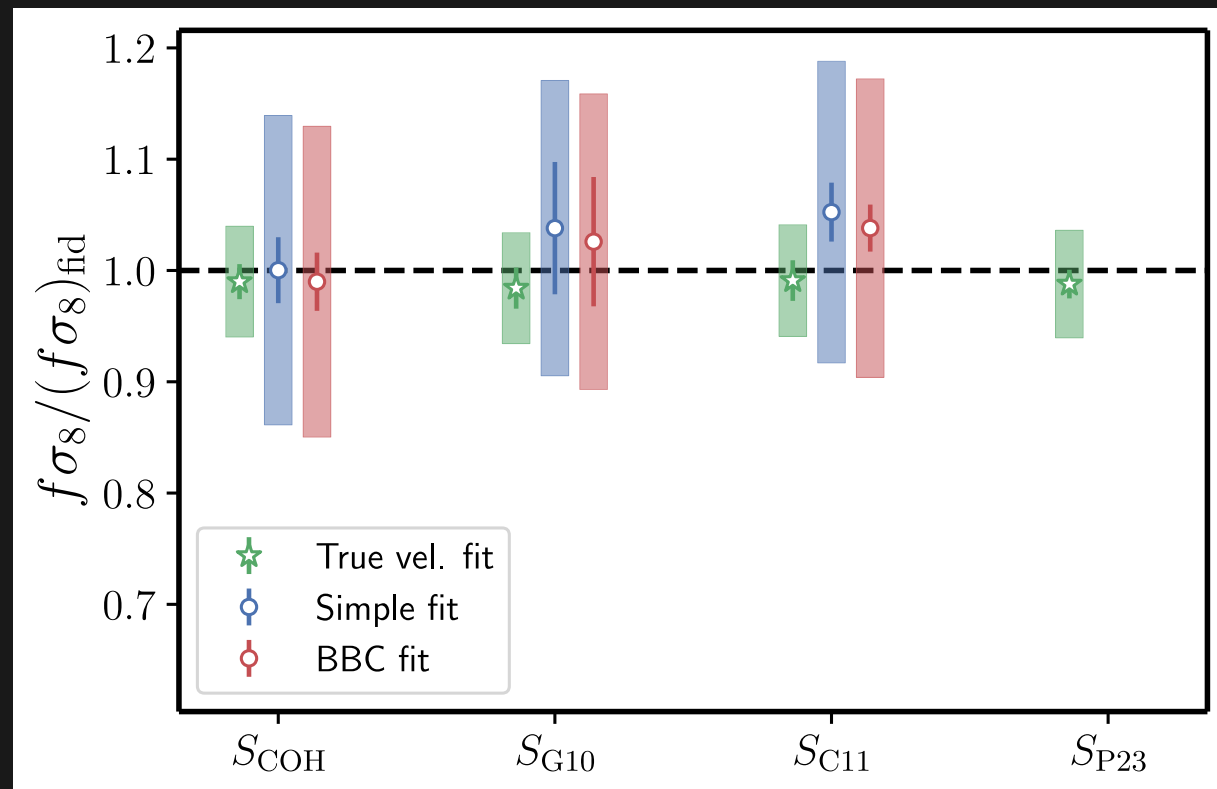
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- True vel. fit:
Unbiased $f\sigma_8$
 $\sigma_{f\sigma_8} \sim 5\%$



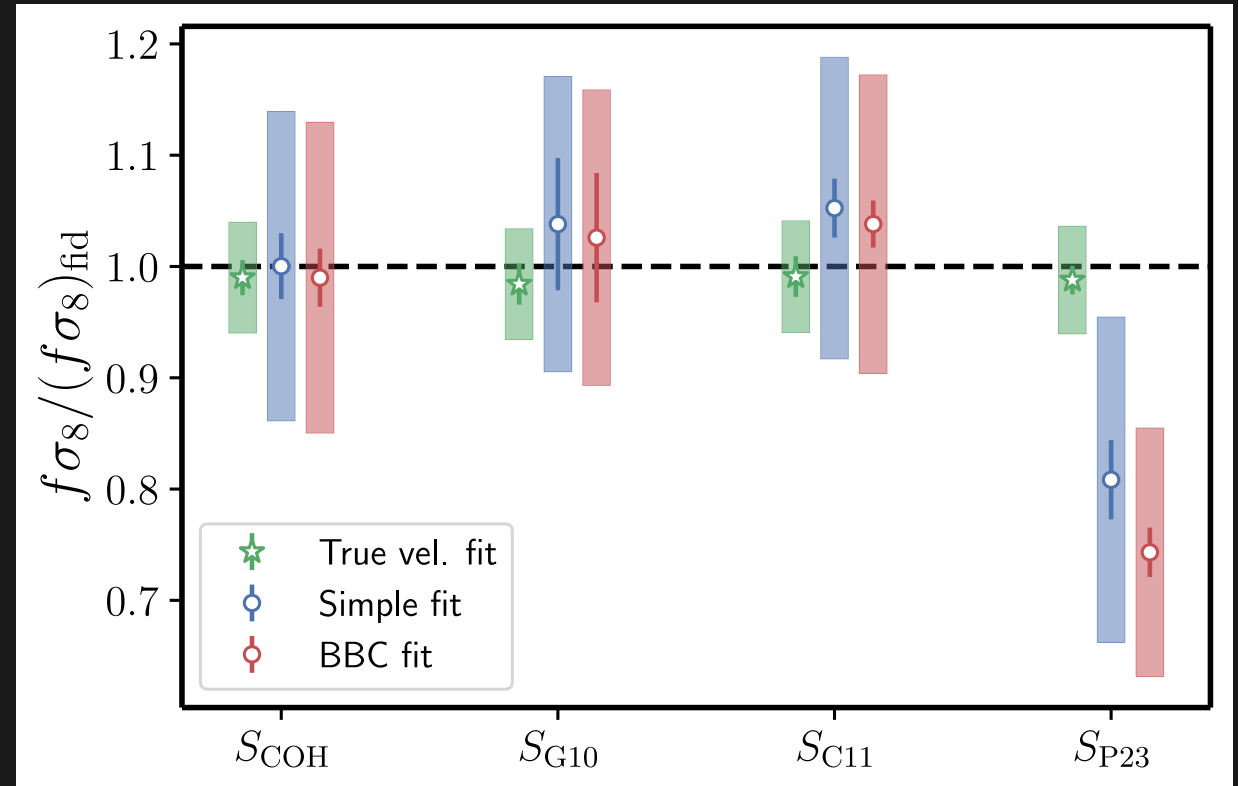
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- P23 - Simple fit: $\sigma_{f\sigma_8} \sim 15\%$
P23 - BBC fit: $\sigma_{f\sigma_8} \sim 11\%$
Results for P23 are biased by $> 20\%$!!!



Results: Systematics - BS21 parameter variations

Errors on BS21 parameters propagated as in DES 5-year analysis (*Vincenzi et al. 2024*)

No change in $f\sigma_8$ fit!!!

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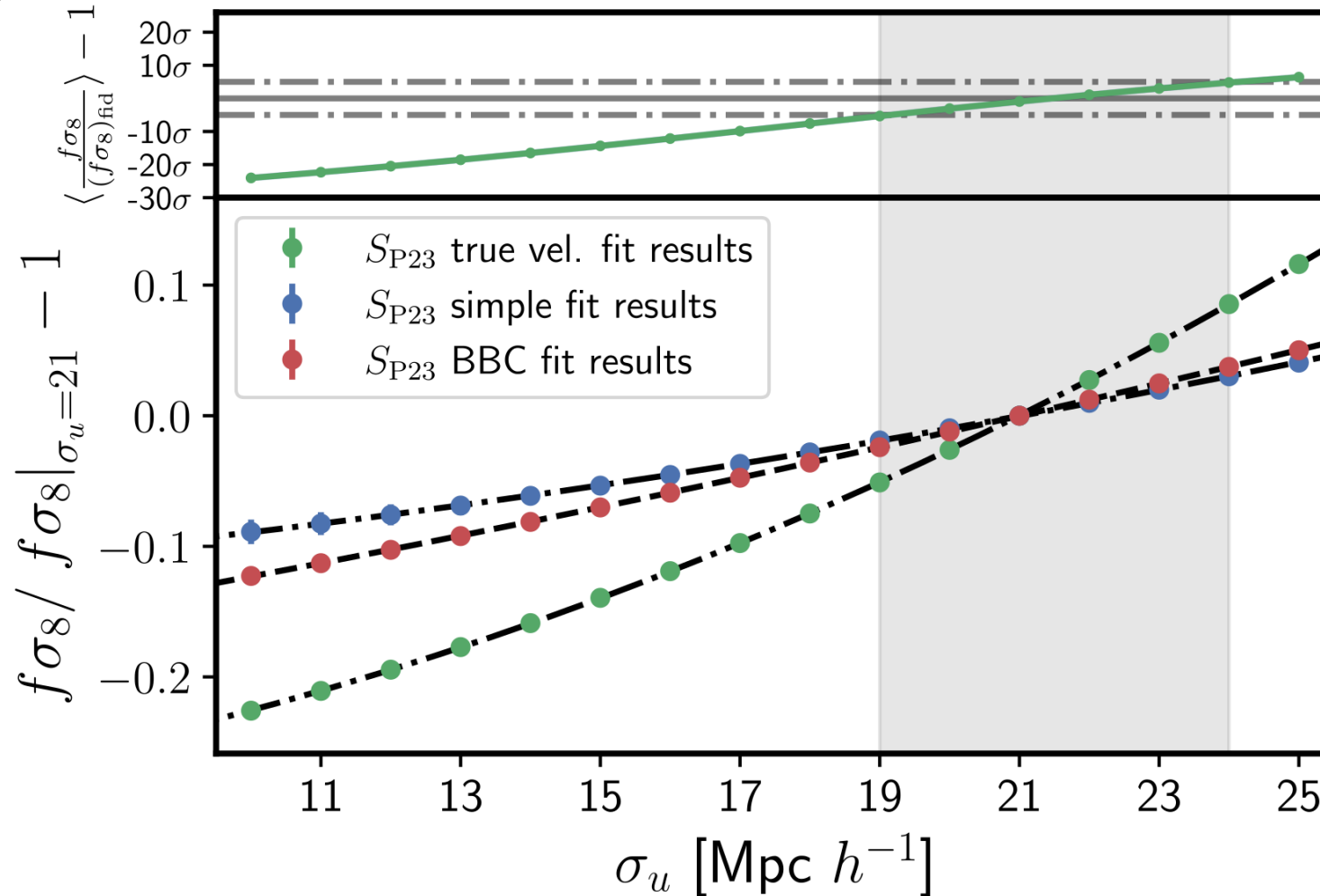
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	$\langle f\sigma_8 \rangle / (f\sigma_8)_{\text{fid}}$	$\sqrt{\langle \sigma_{f\sigma_8}^2 \rangle}$
BBC fit	0.649 ± 0.028	12.2%
BBC + int. scat. cov fit	0.647 ± 0.028	12.2%

Results: Systematics - σ_u RSD parameter

$$\Delta\sigma_u \sim 18.5 - 23.5 \text{ Mpc } h^{-1} \Rightarrow \sigma_{f\sigma_8}^{\sigma_u} \sim 6\%$$



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What's next?

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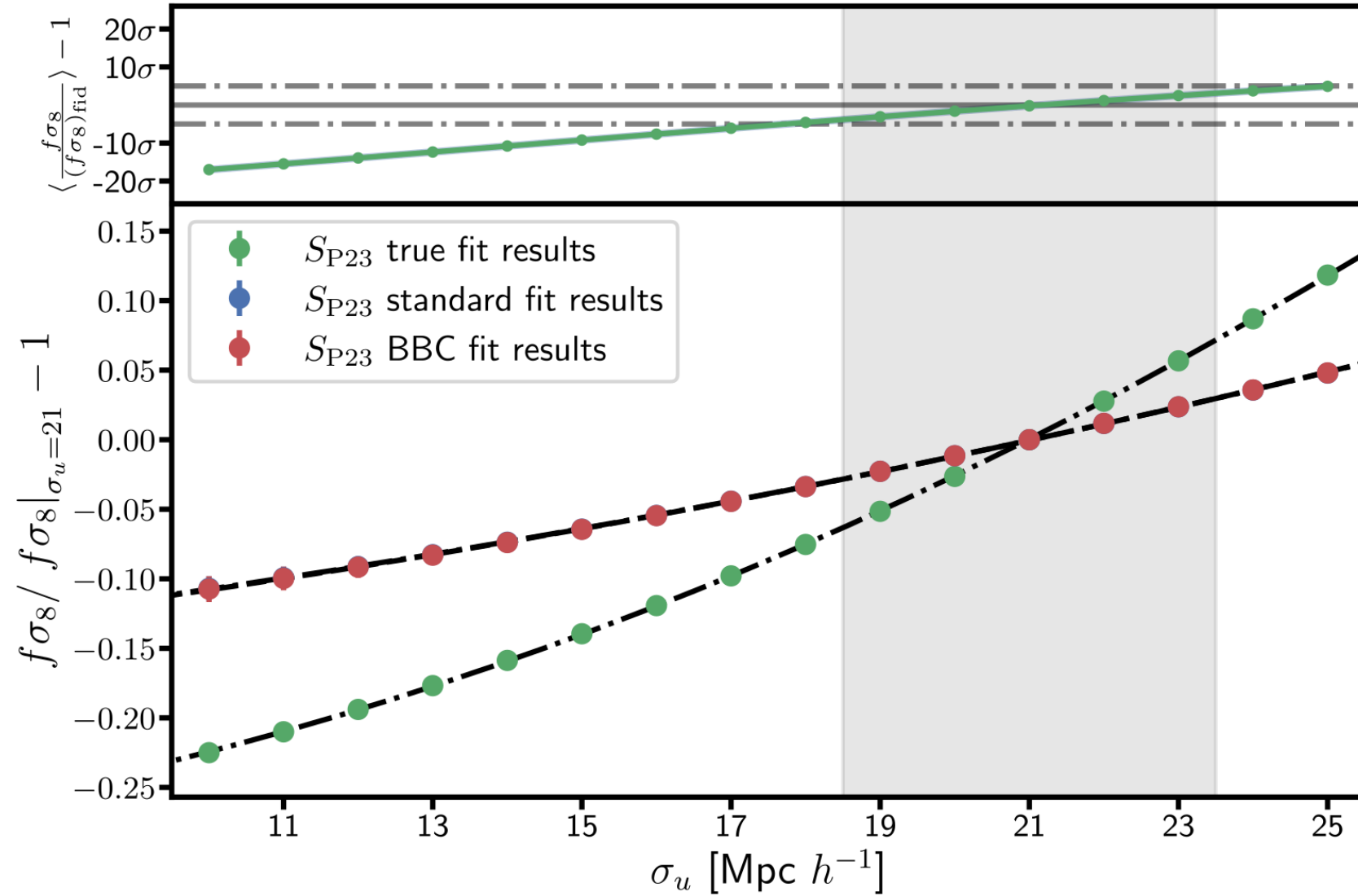
- Is the BS21 model prediction correct? Will we see non-Gaussianities in data?
- Is it possible to find a better RSD parametrisation than σ_u ?

Thank you for your attention !

Full $f\sigma_8$ results

Models	True fit			Standard fit			BBC fit		
	$\langle f\sigma_8 \rangle / (f\sigma_8)_{\text{fid}}$	$\sqrt{\langle \sigma_{f\sigma_8}^2 \rangle}$ (%)	STD($f\sigma_8$) (%)	$\langle f\sigma_8 \rangle / (f\sigma_8)_{\text{fid}}$	$\sqrt{\langle \sigma_{f\sigma_8}^2 \rangle}$ (%)	STD($f\sigma_8$) (%)	$\langle f\sigma_8 \rangle / (f\sigma_8)_{\text{fid}}$	$\sqrt{\langle \sigma_{f\sigma_8}^2 \rangle}$ (%)	STD($f\sigma_8$) (%)
S _{RND}	1.008 ± 0.015	4.7	4.3	0.979 ± 0.023	12.7	6.7	0.983 ± 0.024	12.5	6.8
S _{G10}	1.000 ± 0.018	4.7	5.2	1.013 ± 0.047	12.0	13.2	1.014 ± 0.049	11.9	13.7
S _{C11}	1.001 ± 0.017	4.7	4.9	0.999 ± 0.029	10.5	8.3	1.000 ± 0.026	10.3	7.3
S _{P23}	0.999 ± 0.013	4.6	3.6	0.834 ± 0.035	13.6	12.0	0.805 ± 0.024	10.3	8.5

σ_u syst. for G10



HD residuals

