



# Density-velocity method for $\sigma_8$

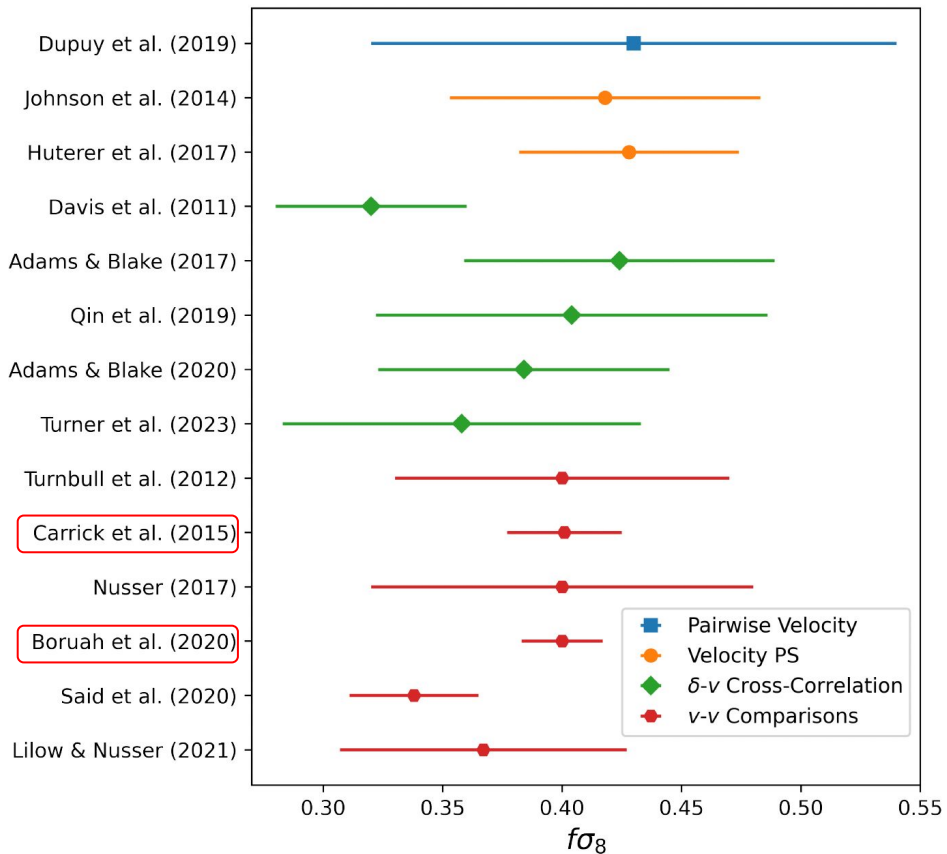
Pauline Zarrouk (CNRS/LPNHE & Sorbonne Université)

1. Methodology
2. Density-velocity projects using ZTF DR2.5 SN Ia @LPNHE

# Methodology: literature and nomenclature

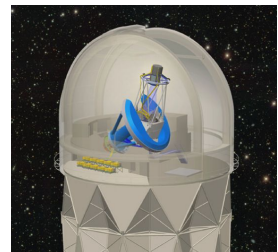
The following names correspond to the same method:

- density-velocity method/comparison
- velocity-velocity comparison
- reconstruction-and-scaling method

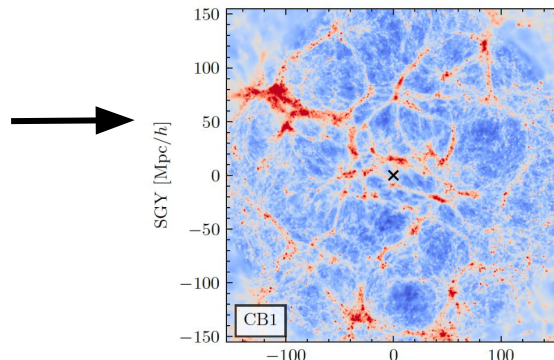
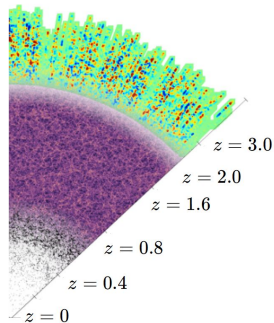


# Methodology: example of DESI x ZTF

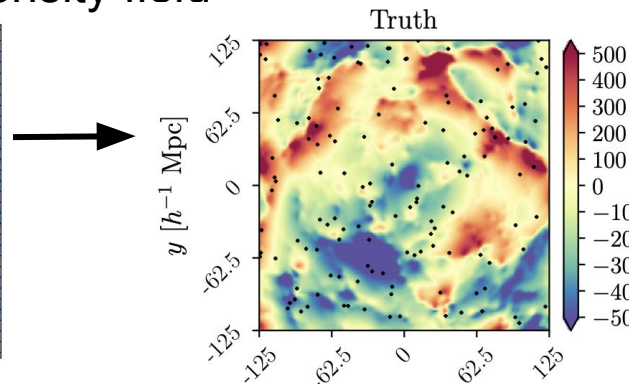
Step 1: Build a model of the velocity field from the galaxy density field



catalogue of galaxy redshifts



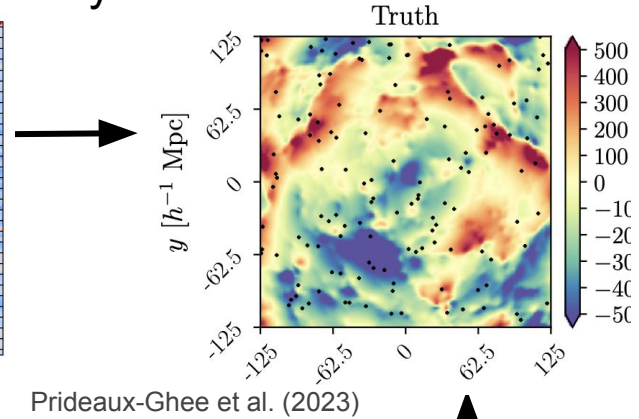
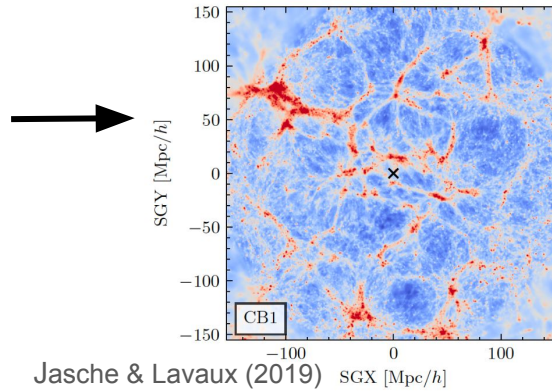
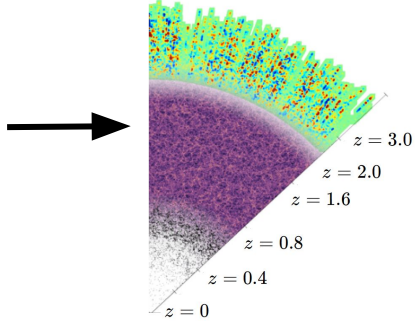
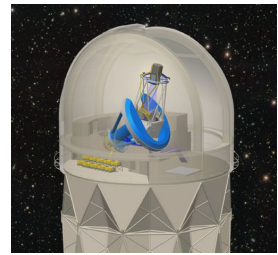
Jasche & Lavaux (2019)



Prideaux-Ghee et al. (2023)

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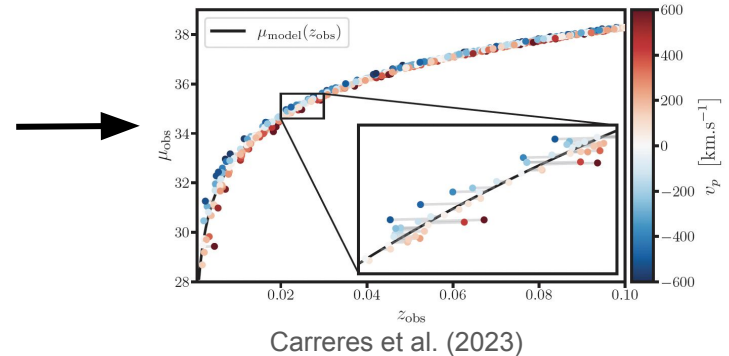
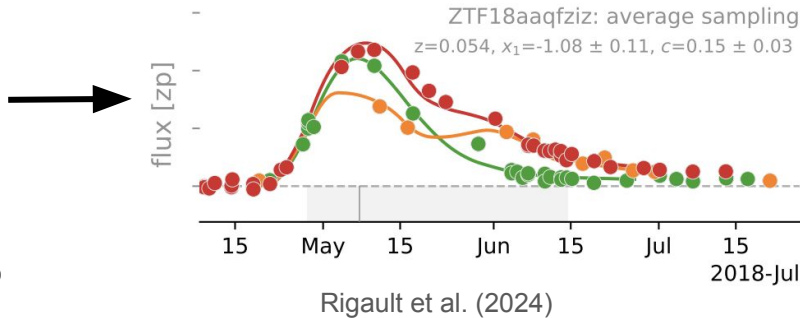
Step 1: Build a model of the velocity field from the galaxy density field



Step 2: Compare with velocity measurements from direct distance tracers



Here SN Ia  
but also TF, FP



$f\sigma_8$

# Methodology: Step 1

Build a **linear** model of the velocity field from the galaxy density field

linear growth rate of structures

$$v(\mathbf{r}) = \frac{H_0 f(\Omega_m)}{4\pi b_g} \int \delta_g(\mathbf{r}') \frac{(\mathbf{r}' - \mathbf{r})}{|\mathbf{r}' - \mathbf{r}|^3} d^3\mathbf{r}'.$$

galaxy bias

Amplitude of mass fluctuations within spheres of  $R = 8 \text{ Mpc/h} \rightarrow \sigma_8$

$$\sigma^2(R, z) = \frac{1}{2\pi^2} \int_0^\infty k^2 P_{\text{lin}}(K, z) |W(k; R)|^2 dk,$$

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galaxy bias

→ Smooth the density field using a Gaussian kernel of width  $R_G$ :  $W(k) = \exp\left(\frac{-k^2 R_G^2}{2}\right)$

Optimal smoothing scale that minimises the variance of the velocity field while ensuring an unbiased reconstruction is  $R_G = 3-5 \text{ Mpc/h}$  (Carrick et al. 2015, Hollinger & Hudson 2021)

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→ account for galaxy selection effect: luminosity weighting in Carrick et a. 2015, tests of various definitions using simulations in Hollinger & Hudson 2021



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$f(\Omega_m)$   $b_g$  galaxy bias

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→ **dominant source of systematics for 2M++: cosmic variance** (Hollinger & Hudson 2021, Hollinger & Hudson 2023, Blake & Turner 2024) → **need mock-based covariance**

# Reconstruction methods for density and velocity fields

Comparison of reconstruction methods against catalogues of direct distance tracers  
(Stiskalek et al. *in prep*)

Reconstruction methods considered:

- **linear theory 2M++**: luminosity-weighted density field derived from the positions of galaxies in 2M++ assuming linear theory (Carrick et al. 2015)
- **neural-network 2M++** from Lilow et al. (2024)
- **CSiBORG (constrained simulations in BORG) 2M++**: obtained by imposing the initial conditions on the density field derived from BORG applied to 2M++. See Jasche & Lavaux (2019), Bartlett et al. (2021), Stopyra et al. (2024)
- **constrained realisations Cosmicflows-2**: Sorce (2018, 2020)
- **HMC forward-modelling Cosmicflows-4**: Courtois et al. (2023)

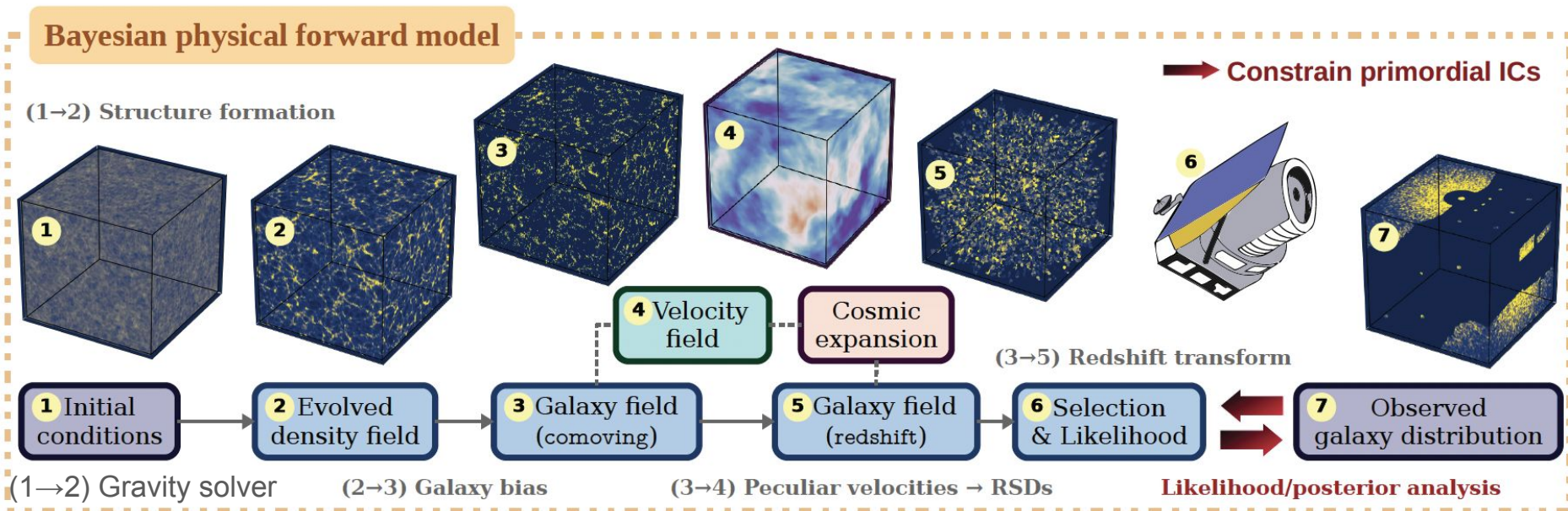
Data considered:

- Tully-Fisher
- **SN Ia**: LOSS, Foundation  $\rightarrow M_{\text{SN}}$ ,  $x_0$  (stretch),  $x_1$  (colour)

# What is BORG?

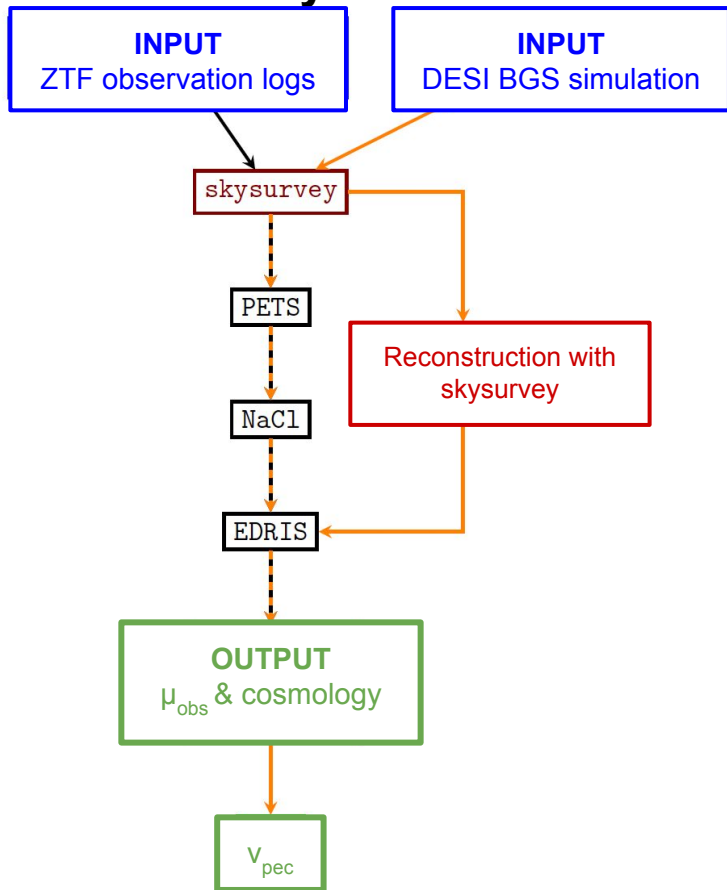
BORG: Bayesian Origin of Reconstruction Galaxies

Jasche & Wandelt (2013), Jasche & Lavaux (2019)



# Methodology: Step 2

## Peculiar velocity measurements from ZTF DR2.5 SN Ia @LPNHE



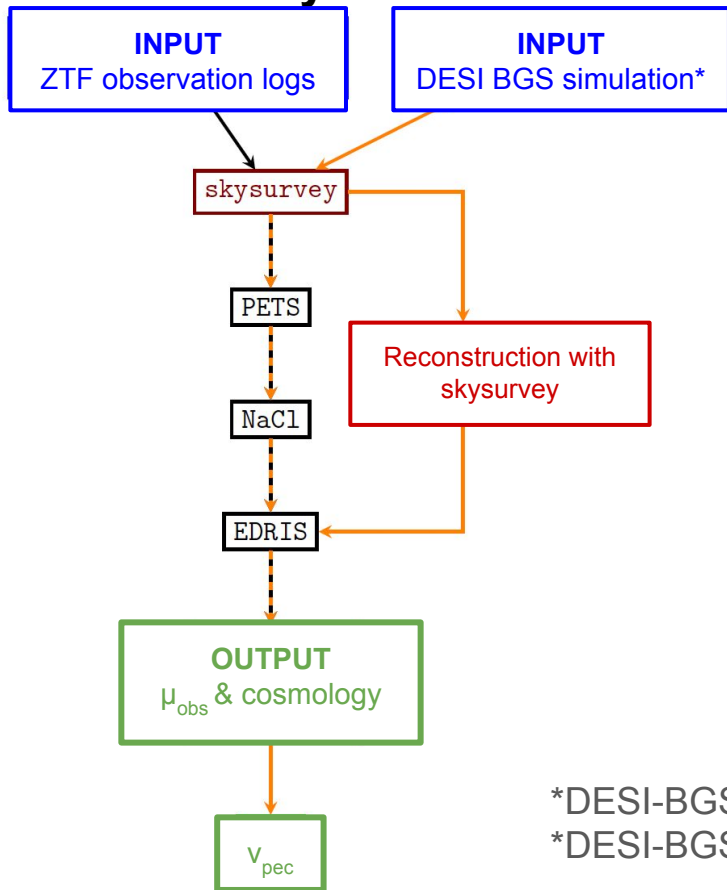
The goal is twofold:

- Use one single pipeline for all cosmological measurements from ZTF DR2.5 SN Ia, including peculiar velocities  
→ Lemaitre pipeline
- Provide unbiased peculiar velocity measurements by first testing the pipeline on simulations

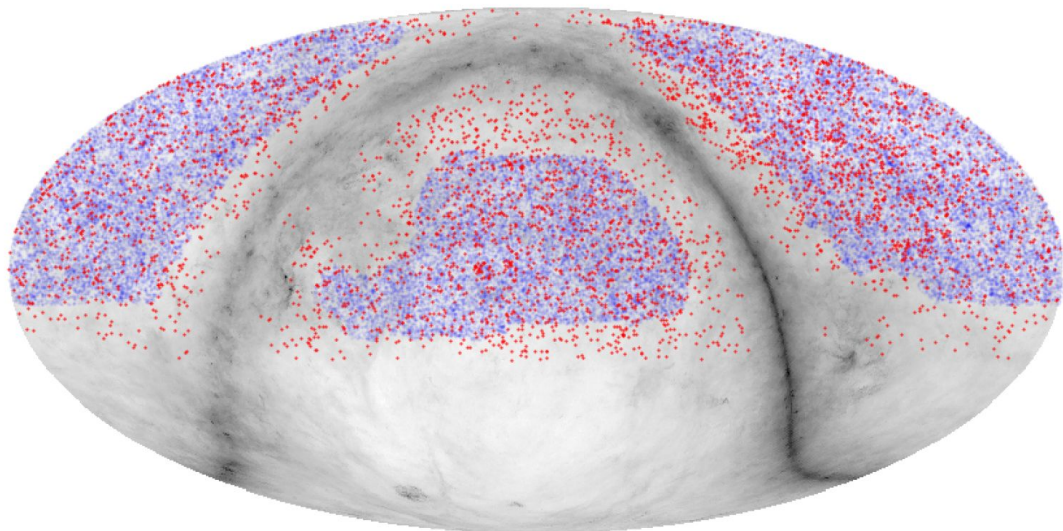
Work led by Antoine Gilles--Lordet, Mahmoud Osman, Pauline Zarrouk and Nicolas Regnault

# Methodology: Step 2

## Peculiar velocity measurements from ZTF DR2.5 SN Ia @LPNHE



Joint DESI BGS x ZTF SN Ia simulations

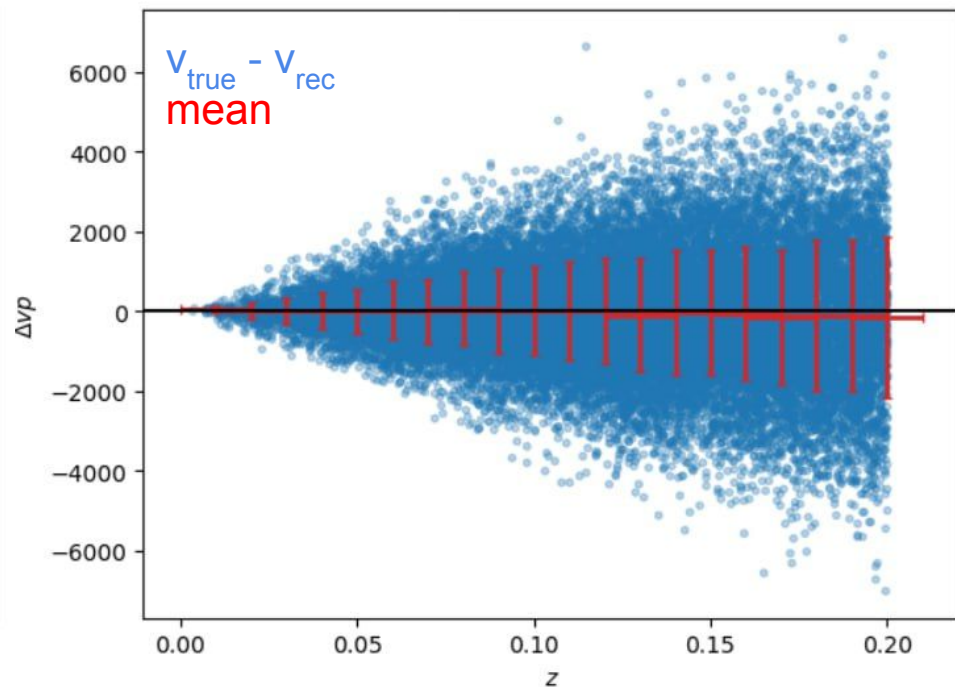
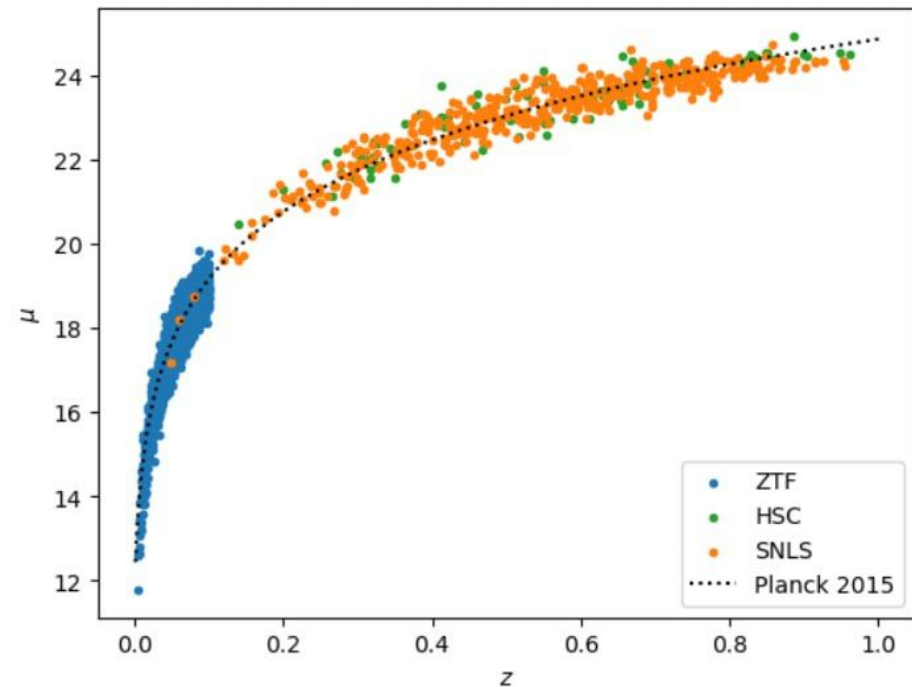


\*DESI-BGS Uchuu (Prada et al. 2024)

\*DESI-BGS AbacusSummit (Smith et al. 2024)

# Methodology: Step 2

## Peculiar velocity measurements from ZTF DR2.5 SN Ia @LPNHE



Next steps: perform systematic tests of  $v_{pec}$  measurements with Lemaitre pipeline following the Data Challenge plan (Mahmoud)

# Density-velocity projects @ LPNHE using ZTF DR2.5 SN Ia

## Using 2M++ for Step 1

- Use CSiBORG-2M++ for the reconstruction of the density and velocity fields
- compare with peculiar velocities measurements from ZTF DR2.5 SN Ia
- Goal: measure  $\sigma_8$ 
  - generate ZTF SN Ia within these constrained simulations using skysurvey
  - use full covariance from CSiBORG 2M++ x ZTF SN Ia
  - timeline: July-Sept 2025

## Using DESI BGS for Step 1

- Field-level inference using BORG applied to the DESI BGS at  $z < 0.2$  to reconstruct the initial conditions and the density and velocity fields → create CSiBORG-DESI BGS
- compare with peculiar velocities measurements from ZTF DR2.5 SN Ia
- Goal: measure  $\sigma_8$ 
  - timeline: post DR2.5

**Next:** Towards joint density and velocity field-level inference using BORG

Thanks for your attention!