



Observatoire astronomique  
de Strasbourg



Leibniz-Institut für  
Astrophysik Potsdam

# Near Field Cosmology

From an Observational  
to a Numerical local Universe

Jenny Sorce

*Fuerteventura, Spain, September 2017*

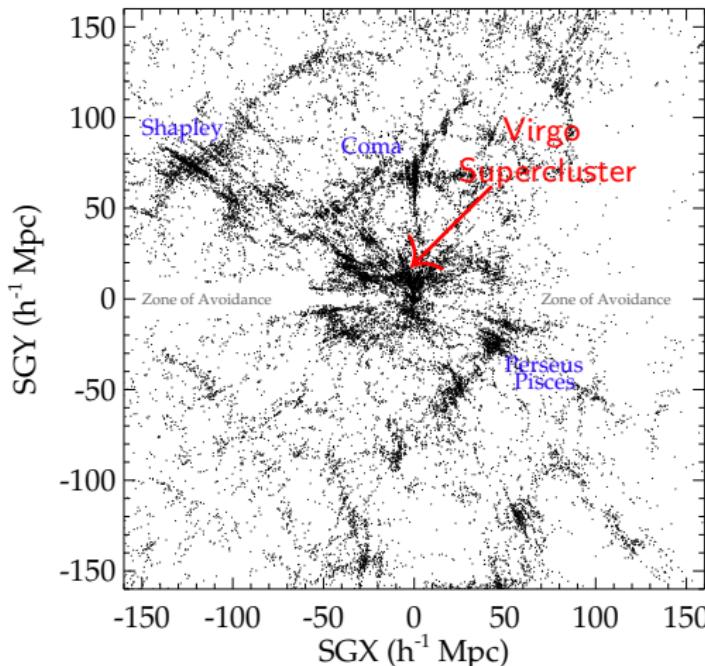
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## Definition (IAU Division H):

- Near Field Cosmology: “[...] increasing interest in studying the **local Universe (near field)** as distinct from the high redshift universe (far field).”
- local Universe: “defined by the distance ( $\sim 10$  Mpc) over which stellar populations in galaxies can be resolved by the HST. [...] **extend to include Virgo** ( $\sim 15$  Mpc) [...] to cover the full range of galaxy environments, from voids to massive groups and clusters. **In an era of ELTs**, [...] **possible to extend** [...] to even **greater distances**.”

# The local Universe in this lecture

Size of the LSS, walls, etc: 100's of Mpc



Virgo Supercluster

Virgo Cluster  
distance:  $\sim 15$  Mpc



Local Group  
size:  $\sim 2$  Mpc

Milky Way  
size:  $\sim 30$  kpc



Andromeda  
distance:  
 $\sim 750$  kpc

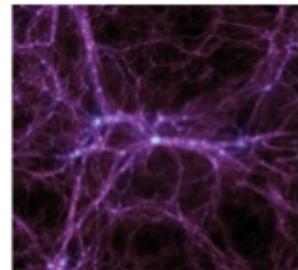
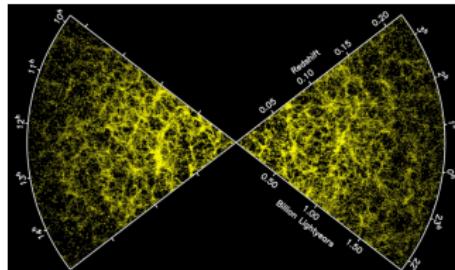
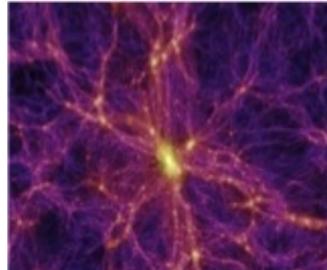


$$1 \text{ pc} = 3.0857 \times 10^{13} \text{ km} = 3.26 \text{ light-years}$$

# Motivation

See also Rien van der Weijgaert's lecture for simulated/observed LSS

$\Lambda$ CDM works well on large scales (simulations vs. observations):



2dF redshift survey, Colless 1999 & Millennium runs, Springel et al. 2005 and 2008

But problems on the small scales, e.g.:

- missing satellite galaxies and dwarfs (e.g. Klypin et al. 1999 ; Moore et al. 1999 ; Zavala et al. 2009) , etc
- size of voids (e.g. Tikhonov & Klypin 2009)
- preferential distribution of the Milky Way's satellites in a pancake shape-like rather than an isotropic distribution (e.g. Kroupa et al. 2005)



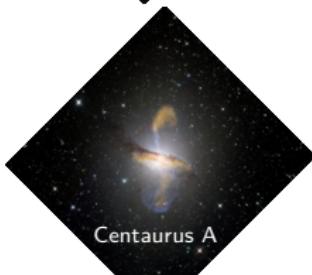
## Is this due to the fact that we reside in a given environment?

Our measurements, conclusions, local and far observations might be biased by its characteristics, e.g.:

- variation of the 'local' Hubble Constant with density (Wojtak et al. 2014)
- impact of the gravitational redshift due to the local gravitational potential (Wojtak et al. 2015)



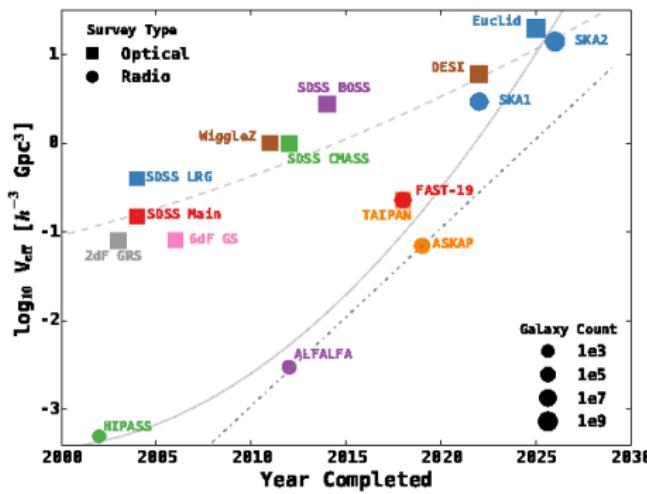
but it is the **best and most observed**  
Volume → Focus ! → detailed  
observations, map, expansion ( $H_0$ )



# Observational

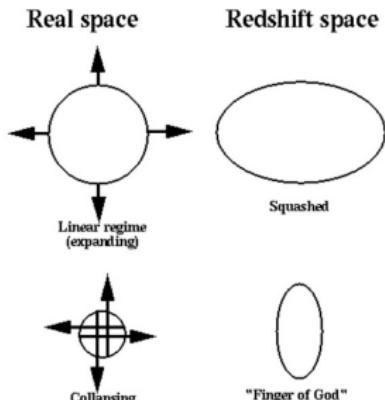
# Mapping the local Universe

From redshift survey to distance measurements:



⇒ redshift survey approximate distance measurements:  $d \sim v/H_0$

See Will Percival & Yi Zheng's lectures for detailed explanations



Kaiser 1987; Hamilton 1997  
also non-linear motions

$$z_{obs} = (\lambda_r - \lambda_e)/\lambda_e$$

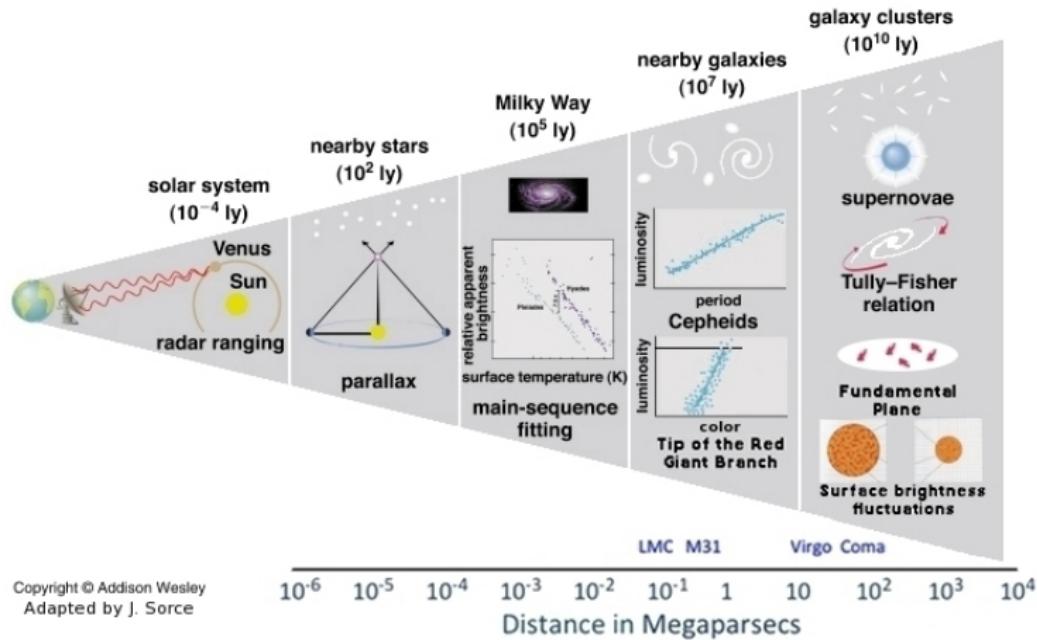
but  $\lambda_r \neq a \times \lambda_e$  because of  $v_{pec}$

$$v_{tot} = H_0 d + v_{pec}$$

$$( \neq cz_{obs} = (1 + cz_{exp})(1 + cz_{pec}) - 1 )$$

direct distance measurements ⇒ distance indicators (and eventually access  $v_{pec}$ )

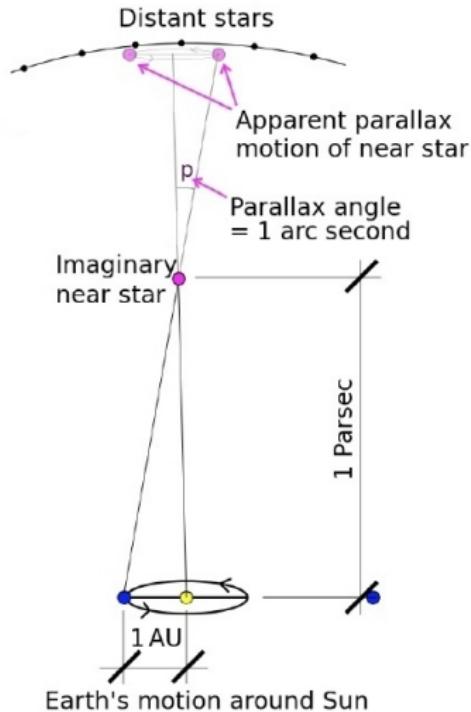
# Cosmic ladder & a few examples



standard candle: celestial object with **absolute magnitude** assumed to **not vary with age/distance**.

distance modulus:  $\mu = m - M = 5 \log_{10}(d(\text{Mpc})) + 25$  (m → measure, M → distance indicator)

# Parallax: apparent motion of stars caused by Earth's motion



First measured in 1838  
by Bessel with an  
heliometer (two  
semi-lenses, adjustable  
until superimposition)

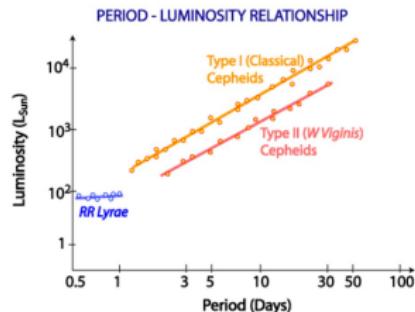
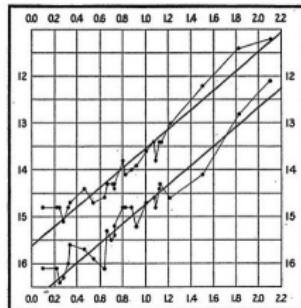
$$d = 1/p \text{ pc}$$

(pc: distance at which a  
star has  $p=1''$ )

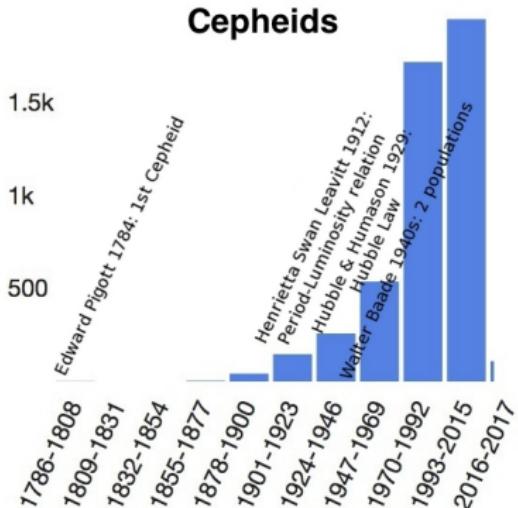
©Sparke & Gallagher

# Cepheids - standard candle: $\mu \propto \log_{10}(P)$

**Cepheid**: star varying in brightness with a well-defined stable period and amplitude.



Luminosity vs. Period (logarithmic scales):  
Left: Cepheids in the SMC (Leavitt 1912) - Right: credit: ATNF

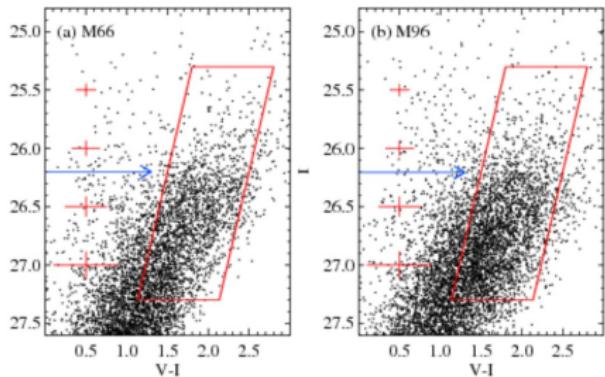


- More: metallicity (Webb et al. 1998), etc
- Big projects: e.g. HST key project (Freedman et al. 2001)

$\kappa$ -mechanism (opacity increases with temperature): atmosphere moves inward  $\Rightarrow$  denser & more opaque  $\Rightarrow$  heats up  $\Rightarrow$  pressure pushes the layer back out (repeat)

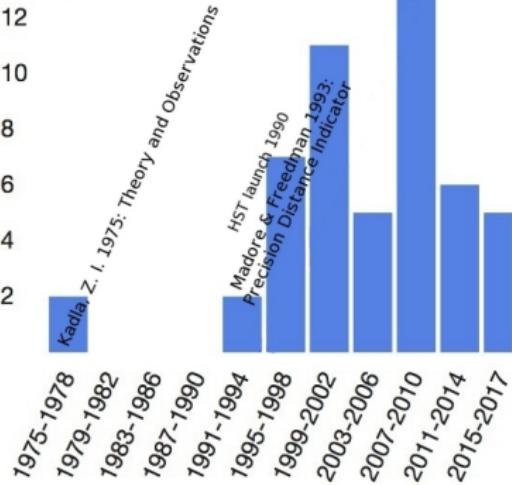
# TRGB - standard candle: $M_I \sim -4.0$

**TRGB:** Tip of the Red Giant Branch.



Lee et al. 2013 - edge-detection = sharp discontinuity

## Tip of the Red Giant Branch



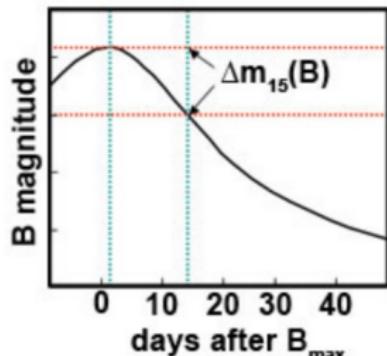
- More: metallicity (Mager et al. 2008), etc
- Big projects: e.g. Carnegie-Chicago Hubble project (Hatt et al. 2017)

Helium core at pressure and temperature to undergo nuclear fusion  $\Rightarrow$  temperature increases  $\Rightarrow$  sharp discontinuity in the evolutionary track of the star on the HR diagram.

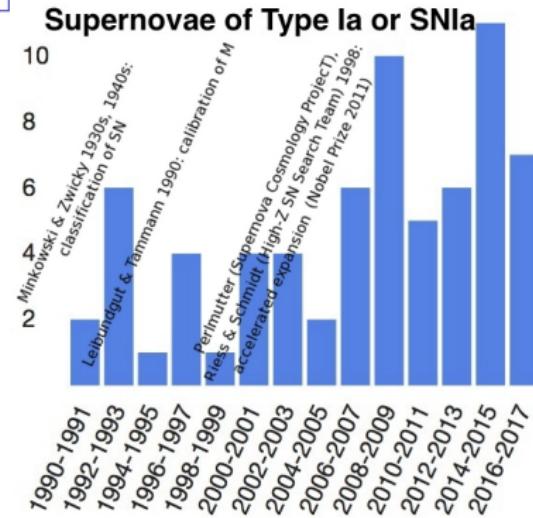
# SNIa - standard candle: $M_v = -19.3$

See Yun Wang's lecture for detailed explanations

**SNIa:** occurs in **binary** systems in which one of the stars is a low rotation rate carbon-oxygen **white dwarf** with  $M_{\text{limit}} = 1.44M_{\odot}$



Credit: Swinburne University of Technology

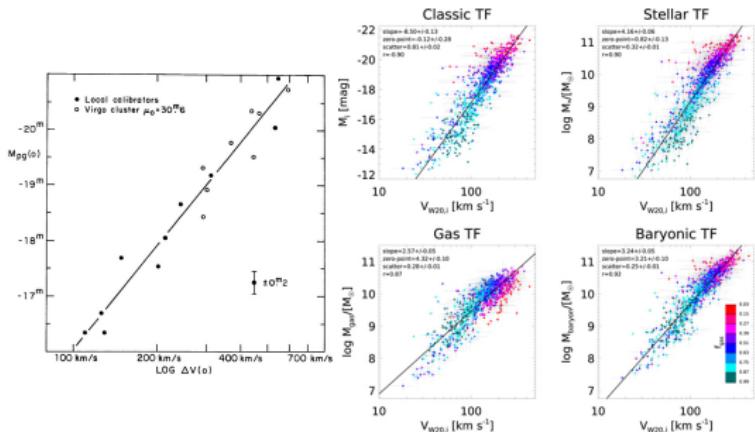


- More: environment (Rigault et al. 2013), etc
- Big projects: e.g. Nearby Supernovae Factory (Aldering et al. 2002)

Mass accretion from companion  $\Rightarrow$  core reaches carbon-T<sub>fusion</sub> close to M<sub>limit</sub>  $\Rightarrow$  SN explosion: C&O converted into heavier elements within a few s., T=billions of deg., released energy unbinds the star  $\Rightarrow$  shock wave, particles at  $\sim 6\%$  of light-speed (most accepted scenario).

# Tully-Fisher: $M \propto \log_{10}(v)$

**classic TF:** luminosity - rotation rate.



Tully & Fisher 1977 - Bradford et al. 2017

$$M_{\text{baryonic}} = M_* + M_{\text{gas}}$$

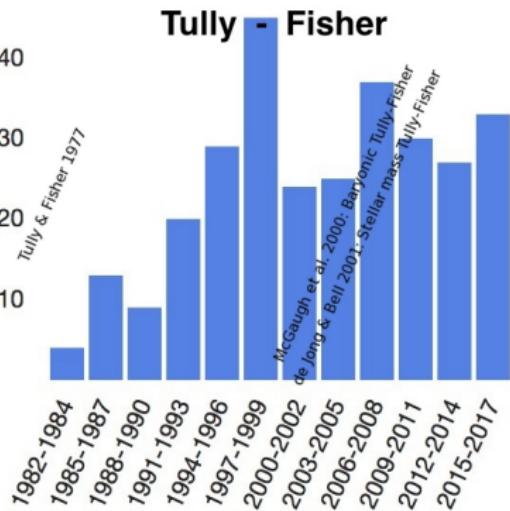
Virial theorem:  $2E_k + E_p = 0 \Rightarrow M/R \propto v^2$

if  $M/L = \text{cst}$  then  $L \propto v^2 R$   
if  $L \propto R^2$  (SB definition)

then  $R^2 \propto v^2 R$  (or  $R \propto v^2$ )

finally  $L \propto v^4$

**NB: Classic TF applicable to spiral galaxies (pending normalization for the others)**

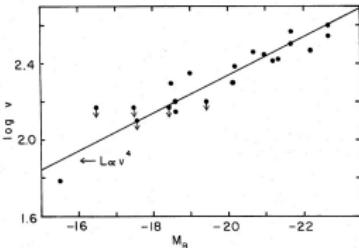


- More: color, sample (Sorce et al. 2013, Sorce & Guo 2016)
- Big projects: e.g. cosmicflows with Spitzer (Sorce et al. 2014)

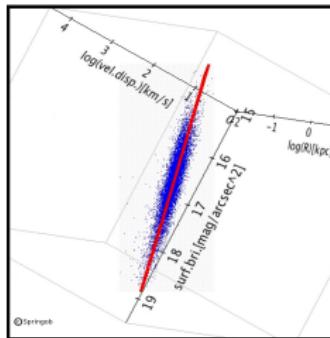
# Faber-Jackson & Fundamental Plane

**Faber-Jackson:** luminosity-velocity dispersion (simple version).

**Fundamental plane:** usually  $R_{\text{eff}}$ ,  $\langle \text{SB} \rangle_{\text{eff}}$ ,  $\sigma$ .



Faber & Jackson 1976 - Credit: 6dFGS collaboration



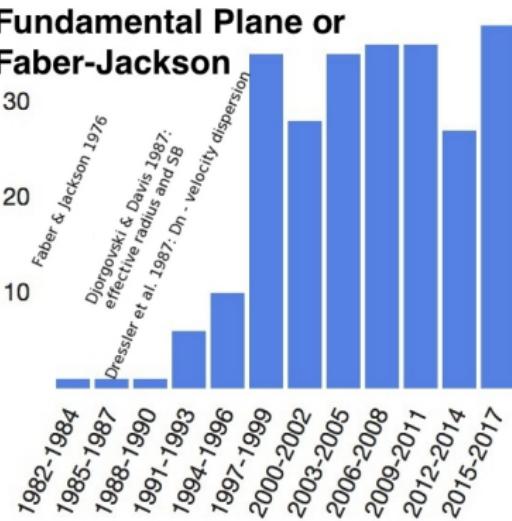
Virial theorem:  $2E_k + E_p = 0 \Rightarrow M/R \propto \sigma^2$

if  $M/L = \text{cst}$  then  $L \propto \sigma^2 R$   
if  $L \propto R^2$  (SB definition)

then  $R^2 \propto \sigma^2 R$  (or  $R \propto \sigma^2$ )

finally  $L \propto \sigma^4$

**NB: Fundamental plane and variants applicable to elliptical galaxies**

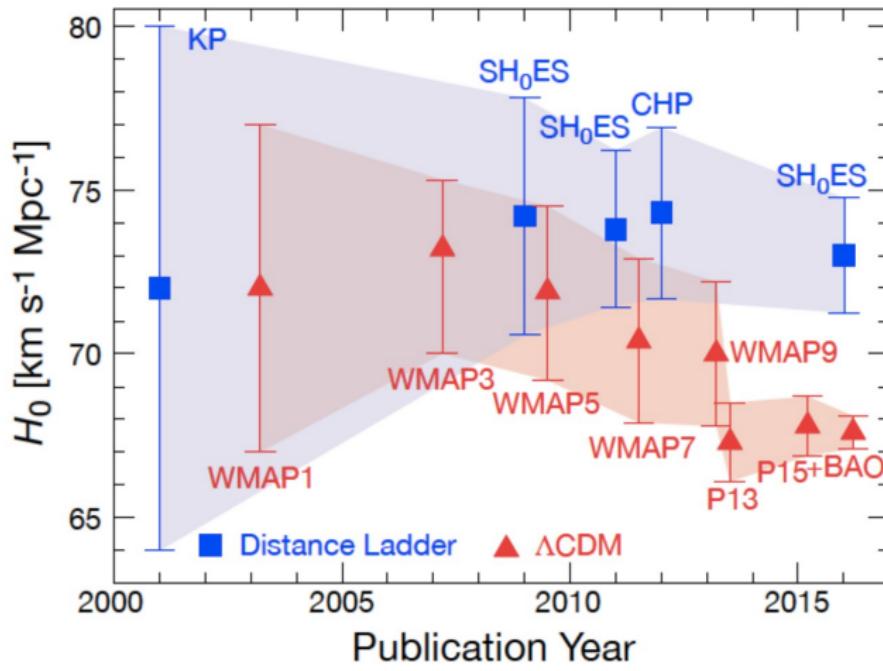


Dn: diameter within which  $\text{SB} = 20.75 \text{mag/arcsec}^2$

- Big projects: 6dF, 6dFGS (Colless et al. 2001, Springob et al. 2014)

# $H_0$ : a never ending history?

distance indicators  $\Rightarrow d \Rightarrow$  When  $d >>$ ,  $H_0 \sim v_{\text{tot}}/d$  but local vs. cosmological (CMB)  $H_0$ :



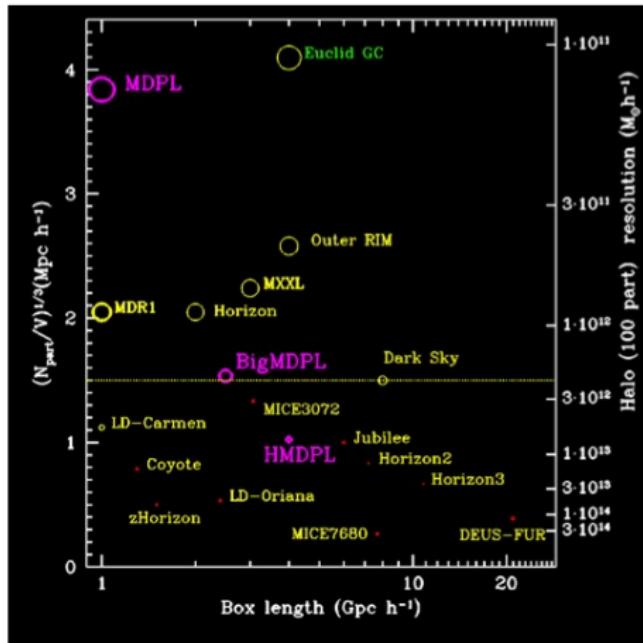
Credit: Freedman 2017

Cosmic variance, another cosmo. model, etc  $\Rightarrow$  test with cosmological simulations

# Numerical

# Simulations: an overview

See Raul Angulo's lecture for detailed information

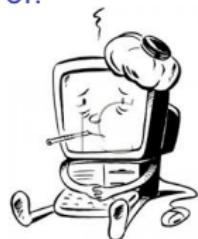


Courtesy of G. Yepes

Need:

- very **large and high resolution** simulations to have **small scales** in all **large scale** environments possible
- even **better with baryons** (e.g Cui & Zhang 2017 for a review )  
⇒ Very challenging / demanding because huge computer resources are required in terms of:

- time
- memory
- storage



# Another option: Constrained Simulations

## PATH INTEGRAL METHODS FOR PRIMORDIAL DENSITY PERTURBATIONS: SAMPLING OF CONSTRAINED GAUSSIAN RANDOM FIELDS

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Center for Theoretical Physics, Center for Space Research, and Department of Physics, Massachusetts Institute of Technology

Received 1987 August 17; accepted 1987 September 10

### ABSTRACT

Path integrals may be used to describe the statistical properties of a random field such as the primordial density perturbation field. In this framework the probability distribution is given for a Gaussian random field subjected to constraints such as the presence of a protovoid or supercluster at a specific location in the initial conditions. An algorithm has been constructed for generating samples of a constrained Gaussian random field on a lattice using Monte Carlo techniques. The method makes possible a systematic study of the density field around peaks or other constrained regions in the biased galaxy formation scenario, and it is effective for generating initial conditions for  $N$ -body simulations with rare objects in the computational volume.



"This identical twin of yours...  
Can you describe him?"

Simulations **resembling** the Local Universe (best observed Volume) to make **direct comparisons** on multi-scales (down to the dwarfs)

==

**Reduction** of the **cosmic variance**

Typical vs. Constrained Initial Conditions:

$\sqrt{P(k)}w(k)$  with  $P$ =power spectrum and  $w$ =white noise.

In the second case, particle **velocity and position** are **constrained**.

# Ingredients to get Constrained Simulations



constraints = **observations**

initial conditions for simulations



## Ingredients to get Constrained Simulations

- Redshifts or peculiar velocities

← constraints = observations

initial conditions for simulations



# Ingredients to get Constrained Simulations

- Redshifts or peculiar velocities
  - Method/Technique

← constraints = observations

← initial conditions for simulations



# Redshift vs. Peculiar velocities

## Redshift:

### PROS:

- Easy

### CONS:

- fingers-of-god, kaiser effects, etc

- luminosity bias

## Peculiar velocities:

### CONS:

- Challenging

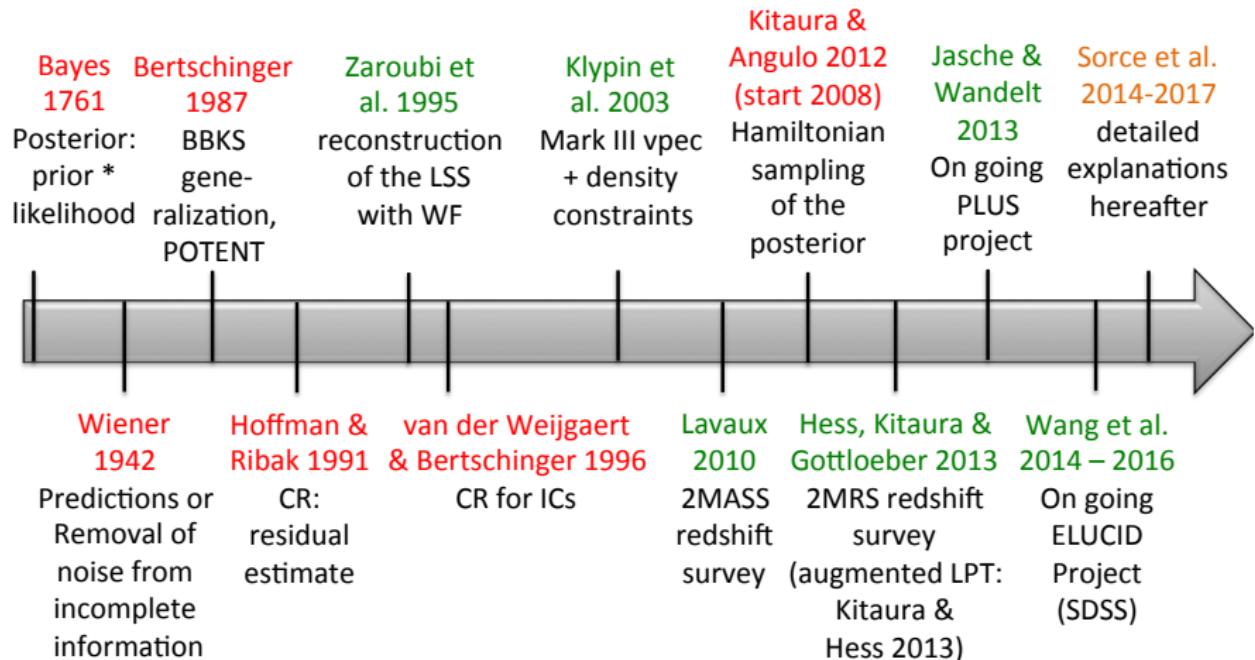
- Homogeneous and heterogeneous Malmquist biases, lognormal error

### PROS:

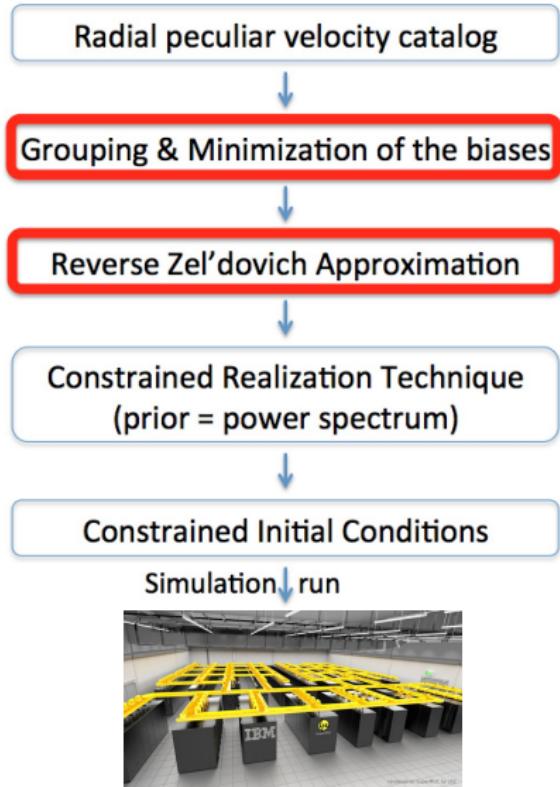
- direct tracers of the underlying gravitational field
- high linearity
- large-scale correlation

# Techniques and applications: History

See also Francisco Kitaura's lecture



# Building Constrained Initial Conditions



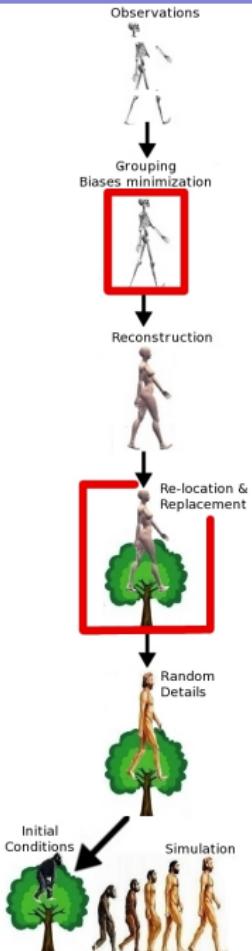
Tully et al.  
2013

Tully 2015,  
Sorce &  
Tempel 2017,  
Sorce 2015

Zaroubi et al.  
1995

Doumler et al.  
2013  
Sorce et al.  
2014b

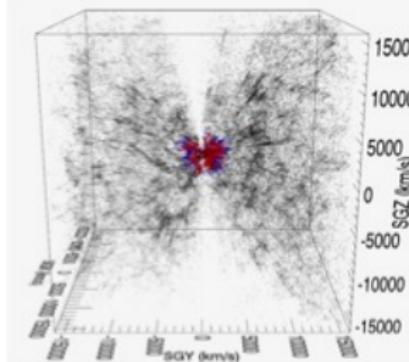
Hoffman &  
Ribak 1991



# Cosmicflows: Observational datasets

CF1

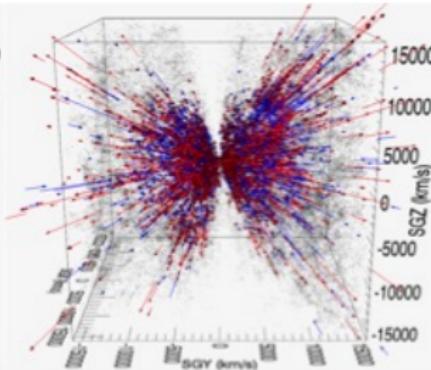
Tully et al. 2008



$\sim 1,8000$

CF2

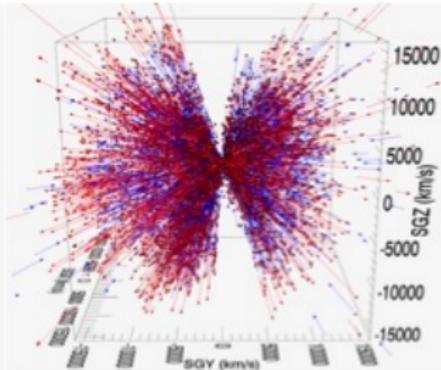
Tully,...,Sorce et al. 2013



$\sim 8,000$   
Radial peculiar velocities

CF3

Tully,...,Sorce 2016



$\sim 17,000$

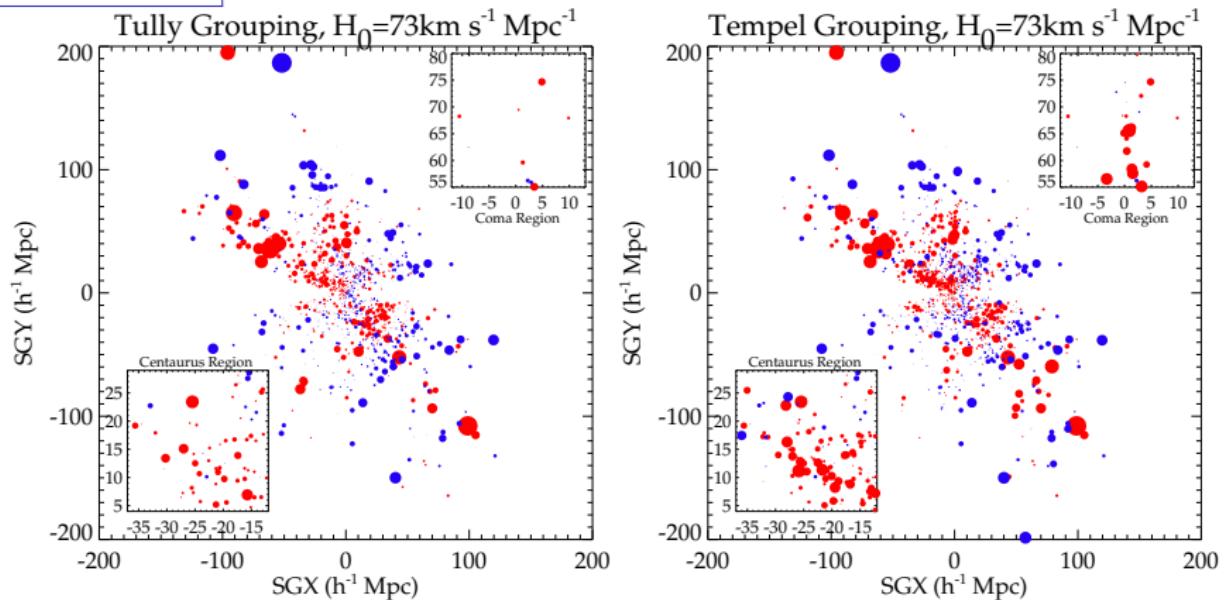
Max. d (Mpc)  $\sim 50$   
Mean d (Mpc)  $\sim 15$

$\sim 300$   
 $\sim 60$

$\sim 370$   
 $\sim 90$

Courtesy D. Pomarède

# Grouping



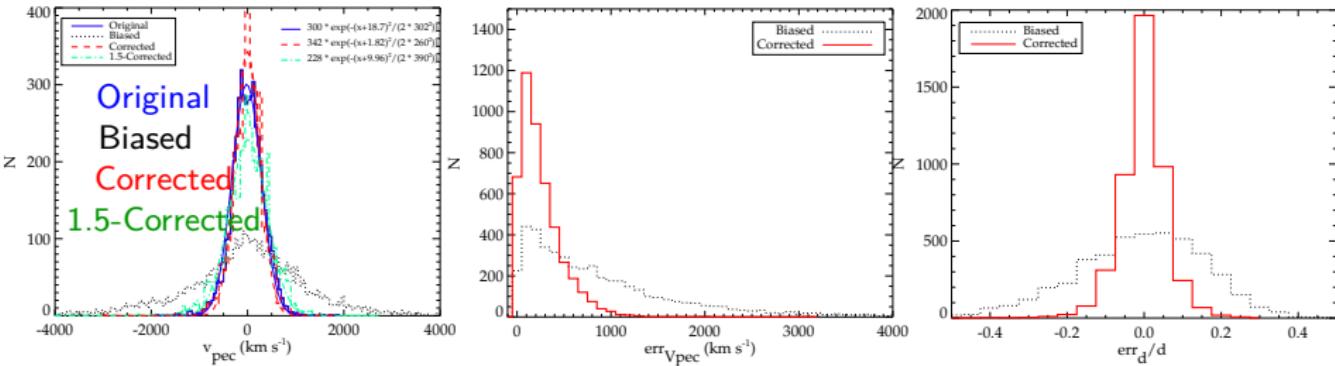
$$\mu_g = \frac{\sum w \times \mu}{\sum w} ; \sigma_{\mu g} = \sqrt{\frac{1}{\sum w}} \text{ where } w = \frac{1}{\sigma_\mu^2},$$

$$d_g = 10^{\frac{\mu_g - 25}{5}} ; \sigma_{dg} = \sigma_{\mu g} \times \frac{\log(10)}{5}$$

$$v_{\text{pec g}} = v_{\text{tot g}} - H_0 \times d_g ; \sigma_{v\text{pec g}} = \sigma_{dg} \times d_g \times H_0$$

# Minimization of the biases

Sorce 2015



## Iterations on:

- if  $v_{\text{pec}} > 0$ ,  $v_{\text{pec c}} = (1 - w)[p(v_{\text{pec}} - \sigma_{v_{\text{pec}}}) + (1 - p)(v_{\text{pec}} + \sigma_{v_{\text{pec}}})] + w v_{\text{pec}}$
- if  $v_{\text{pec}} < 0$ ,  $v_{\text{pec c}} = (1 - w)[p(v_{\text{pec}} + \sigma_{v_{\text{pec}}}) + (1 - p)(v_{\text{pec}} - \sigma_{v_{\text{pec}}})] + w v_{\text{pec}}$

then multiplication by 1.5

- $p$ : probability  $v_{\text{pec}} \notin$  theoretical Gaussian (from the mock) (Sheth and Diaferio, 2001)
- $w$ : weighted uncertainty on  $v_{\text{pec}}$

## After correction:

- distances computed accordingly:  $d_c = (v_{\text{obs}} - v_{\text{pec c}})/H_0$
- 5% fractional error on distances assumed.

# Wiener-Filter

Linear minimum variance estimator between data  $C_j$  and model:  $f_i = \sum_{j=1}^n a_j C_j$

$$f_i = \sum_{j=1}^n \sum_{i=1}^n \langle f_i C_i \rangle \langle C_i C_j \rangle^{-1} C_j$$

$\langle AB \rangle$ : correlation functions involving the prior model P(power spectrum)

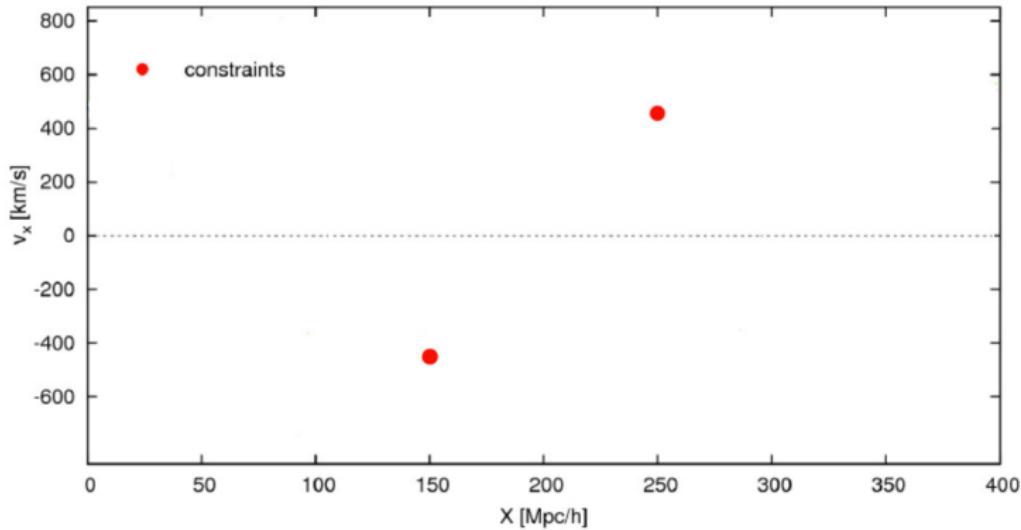
For  $f=\delta$  or  $v$  and  $C=v$  with "j<sub>i</sub>s" the common Bessel functions:

$$\langle \delta(\mathbf{r}') v_\alpha(\mathbf{r}' + \mathbf{r}) \rangle = \frac{\dot{a}f}{(2\pi)^3} \int_0^\infty \frac{ik_\alpha}{k^2} P(k) e^{-i\mathbf{k}\cdot\mathbf{r}} d\mathbf{k} = -\dot{a}f \hat{r}_\alpha \frac{1}{2\pi^2} \int_0^\infty k^2 j_0(kr) P(k) dk$$

$$\langle v_\alpha(\mathbf{r}') v_\beta(\mathbf{r}' + \mathbf{r}) \rangle = \frac{(\dot{a}f)^2}{(2\pi)^3} \int_0^\infty \frac{k_\alpha k_\beta}{k^4} P(k) e^{-i\mathbf{k}\cdot\mathbf{r}} d\mathbf{k} = (\dot{a}f)^2 [\Psi_T \delta_{\alpha\beta}^\kappa + (\Psi_R - \Psi_T) \hat{r}_\alpha \hat{r}_\beta]$$

where  $\Psi_T = \frac{1}{2\pi^2} \int_0^\infty \frac{j_1(kr)}{kr} P(k) dk$  and  $\Psi_R = \frac{1}{2\pi^2} \int_0^\infty [j_0(kr) - \frac{2j_1(kr)}{kr}] P(k) dk$ .

# Wiener-Filter or Reconstruction Technique

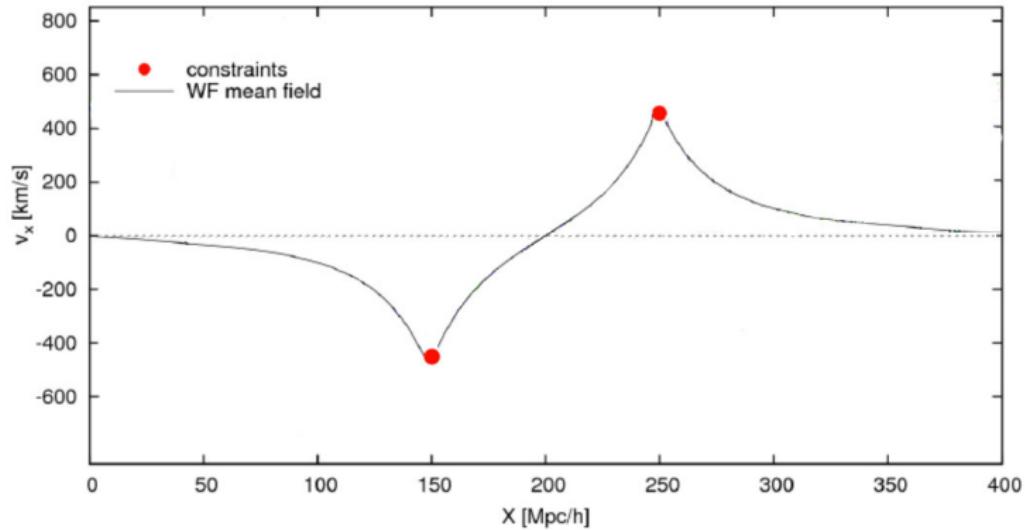


**Wiener-Filter** reconstruction = **Linear Minimal Variance Estimator** (valid down to  $2 \text{ h}^{-1} \text{ Mpc}$ ) using **noisy, sparse data** and a model (Zaroubi et al. 1995)



$$\text{Example : } v_x^{WF}(\mathbf{X}) = \sum_{i=1}^n \langle v_x(\mathbf{X}) C_i \rangle \sum_{j=1}^n \langle C_i C_j \rangle^{-1} (\mathbf{C}_j)$$

# Wiener-Filter or Reconstruction Technique



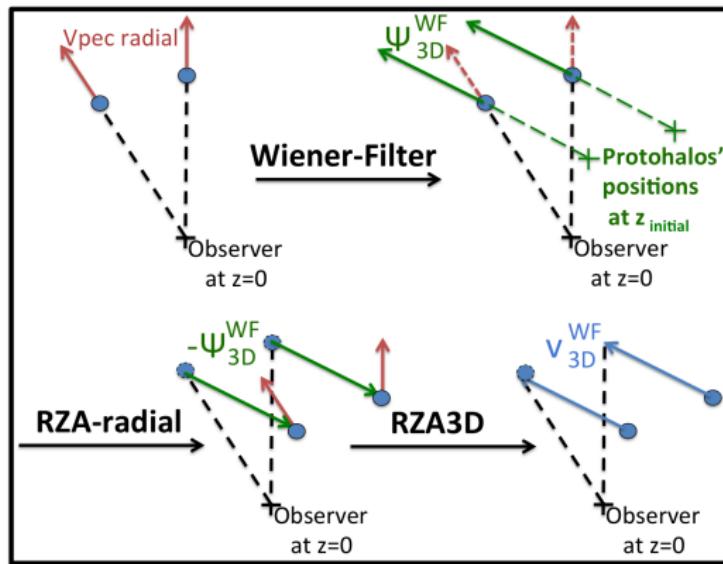
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# Reverse Zel'dovich Approximation

(Doumler et al. 2013 and Sorce et al. 2014 & cf.  
Rien van der Weijgaert's lecture for 'nice' stories)



Reverse Zel'dovich Approximation:

$$\vec{x}_{init}^{RZA} = \vec{r} - \frac{\vec{v}}{H_0 f(t_{init})}$$

growth rate :  $f(t) = \frac{d(\ln D(t))}{d(\ln a(t))}$  growth factor  
scale factor

Linear Theory at 1<sup>st</sup> order valid down to 2 h<sup>-1</sup> Mpc

# Constrained Realization (CR)

1. WF for random constraints  $\tilde{C}_j$  of a random field  $\tilde{f}^{RR}$ :

$$\tilde{f}^{WF} = \sum_{j=1}^n \sum_{i=1}^n \langle f_i \tilde{C}_i \rangle \langle \tilde{C}_i \tilde{C}_j \rangle^{-1} \tilde{C}_j = \sum_{j=1}^n \sum_{i=1}^n \langle f_i C_i \rangle \langle C_i C_j \rangle^{-1} \tilde{C}_j$$

(correlation functions depend only on the prior model)

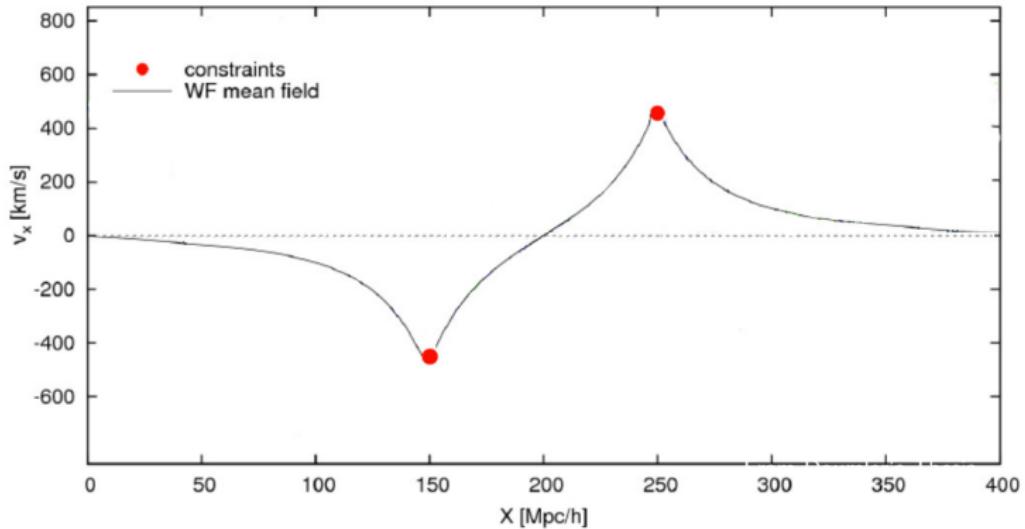
2. Residual:

$$\tilde{R} = \tilde{f}^{RR} - \tilde{f}^{WF} = \tilde{f}^{RR} - \sum_{j=1}^n \sum_{i=1}^n \langle f_i C_i \rangle \langle C_i C_j \rangle^{-1} \tilde{C}_j$$

3. Constrained realization field:

$$f^{CR} = f^{WF} + \tilde{R} = \tilde{f}^{RR} + \sum_{j=1}^n \sum_{i=1}^n \langle f_i C_i \rangle \langle C_i C_j \rangle^{-1} (C_j - \tilde{C}_j)$$

# Wiener-Filter or Reconstruction Technique

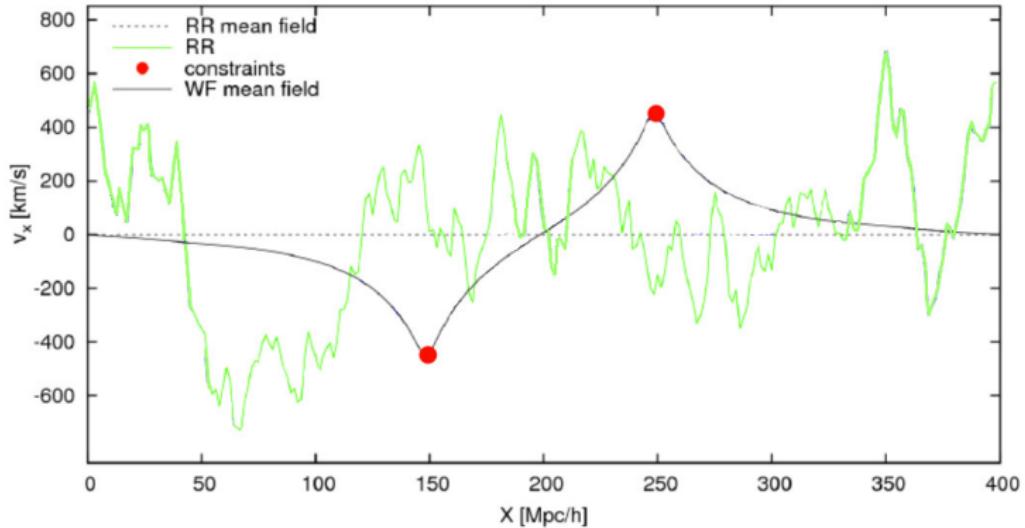


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$$\text{Example : } v_x^{WF}(\mathbf{X}) = \sum_{i=1}^n \langle v_x(\mathbf{X}) C_i \rangle \sum_{j=1}^n \langle C_i C_j \rangle^{-1} (\mathbf{C}_j)$$



# Constrained Realization Technique

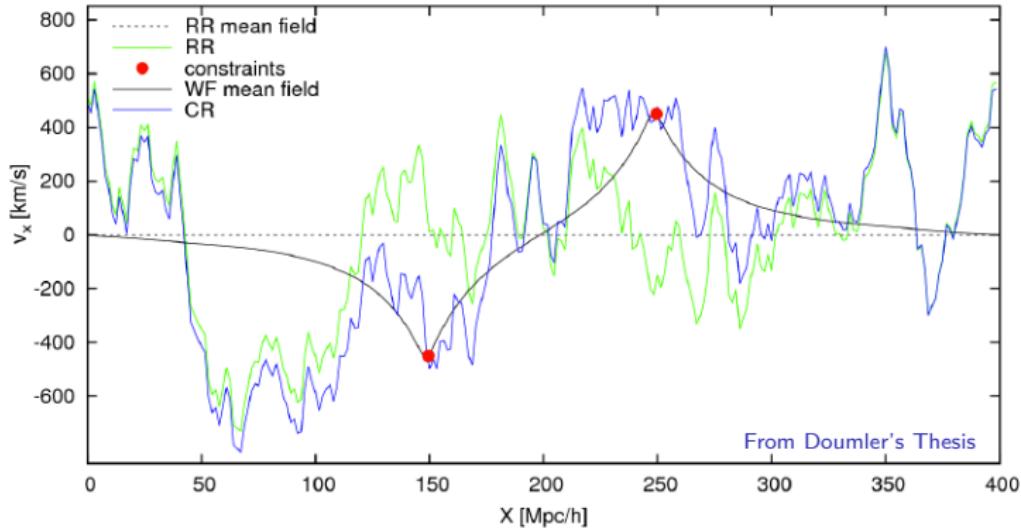


**Constrained Realizations**  $\approx$  Wiener-Filter + Random Realization to compensate for the missing Power Spectrum (Hoffman & Ribak 1991)

$$\text{Example : } v_x^{CR}(\mathbf{X}) = v_x^{RR}(\mathbf{X}) + \sum_{i=1}^n \langle v_x(\mathbf{X}) C_i \rangle \sum_{j=1}^n \langle C_i C_j \rangle^{-1} (\mathcal{C}_j - \bar{C}_j)$$



# Constrained Realization Technique

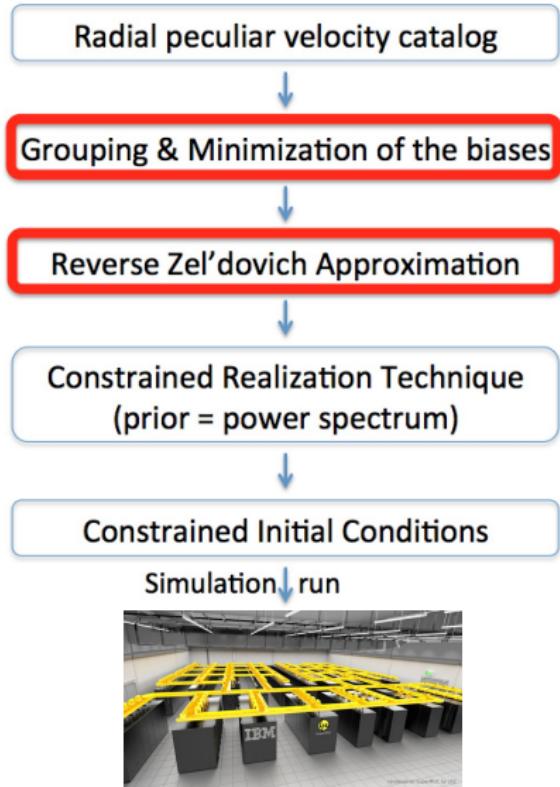


**Constrained Realizations**  $\approx$  Wiener-Filter + Random Realization to  
compensate for the missing Power Spectrum (Hoffman & Ribak 1991)



$$\text{Example : } v_x^{CR}(\mathbf{X}) = v_x^{RR}(\mathbf{X}) + \sum_{i=1}^n \langle v_x(\mathbf{X}) C_i \rangle \sum_{j=1}^n \langle C_i C_j \rangle^{-1} (\mathcal{C}_j - \bar{C}_j)$$

# Building Constrained Initial Conditions



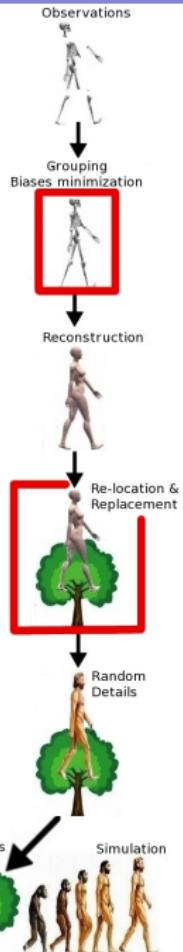
Tully et al.  
2013

Tully 2015,  
Sorce &  
Tempel 2017,  
Sorce 2015

Zaroubi et al.  
1995

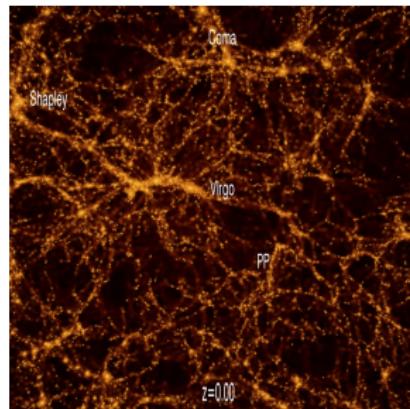
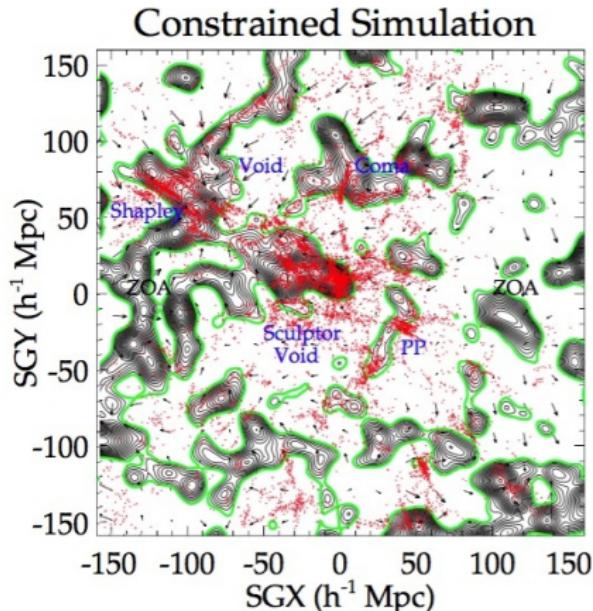
Doumler et al.  
2013  
Sorce et al.  
2014b

Hoffman &  
Ribak 1991



# How did the Local Universe form?

At  $z=0$

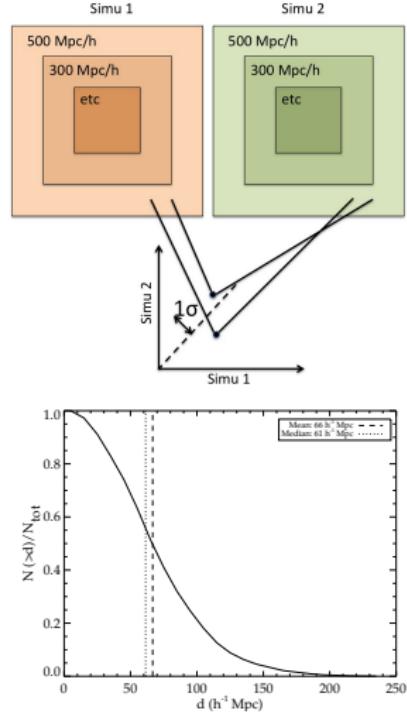
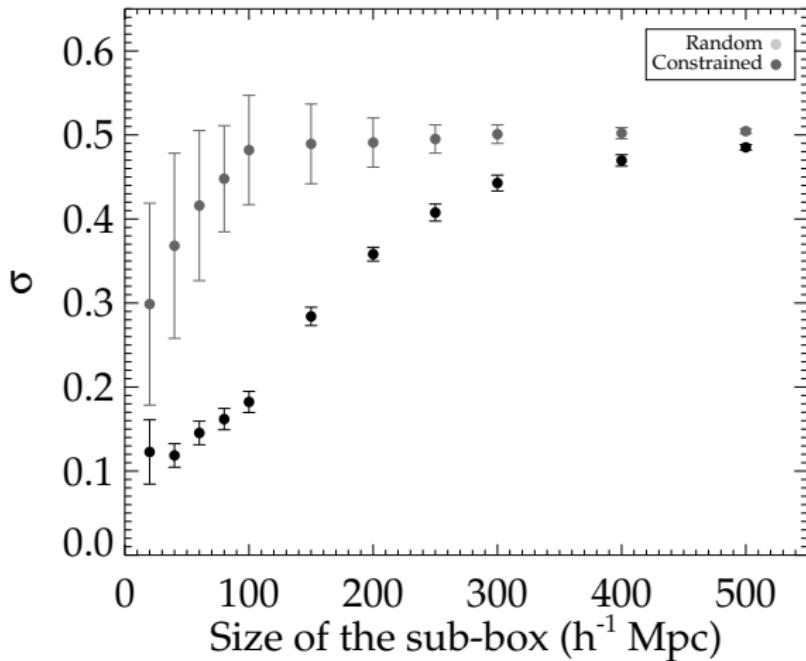


Observations for comparisons: redshift catalog •

Observations to constrain = Peculiar Velocities: CF2 catalog

Simulation:  $L=500 \ h^{-1} \text{Mpc}$ ,  $n=512^3$ , full field (contours, arrows)

# Robust Large-Scale Environment



Smoothing:  $5 h^{-1}$  Mpc

Mean and scatter of  $1-\sigma$  scatters in cell-to-cell comparisons

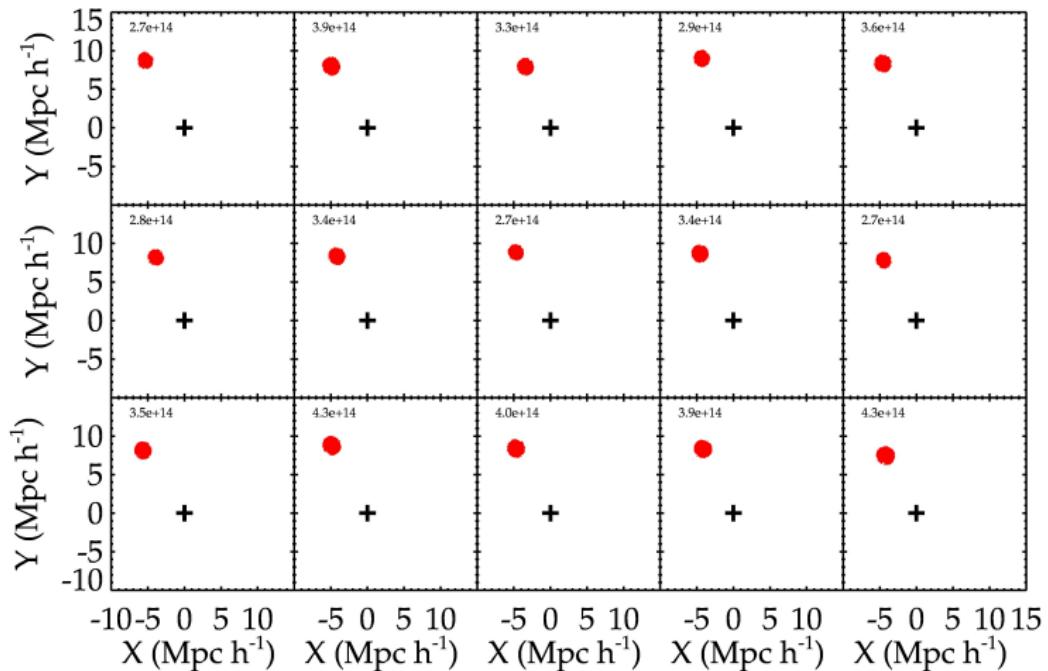
Robust Large-Scale Environment → to study local structures and objects

Can we zoom-in?

What about the clusters?



# How did the Virgo cluster form?

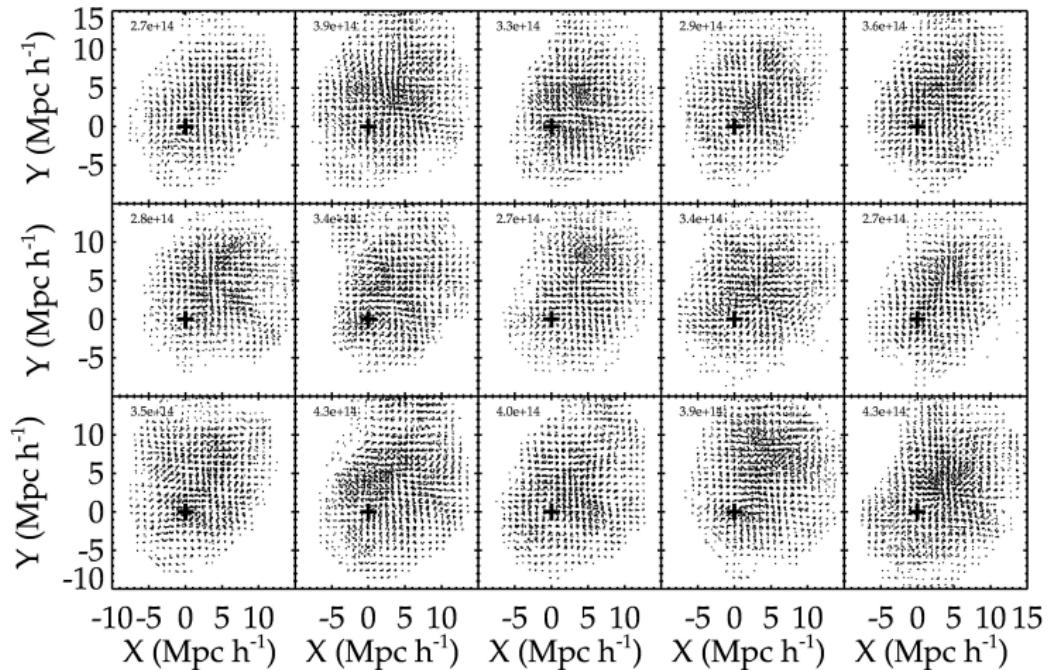


Dark Matter Haloes - Virgo Candidates: Particles at  $z=0$

- Shift  $\sim 3\text{-}4 \text{ } h^{-1} \text{ Mpc}$
- Mass within  $\sim [0.5, 2]$  estimated mass (Ludlow & Porciani 2011)

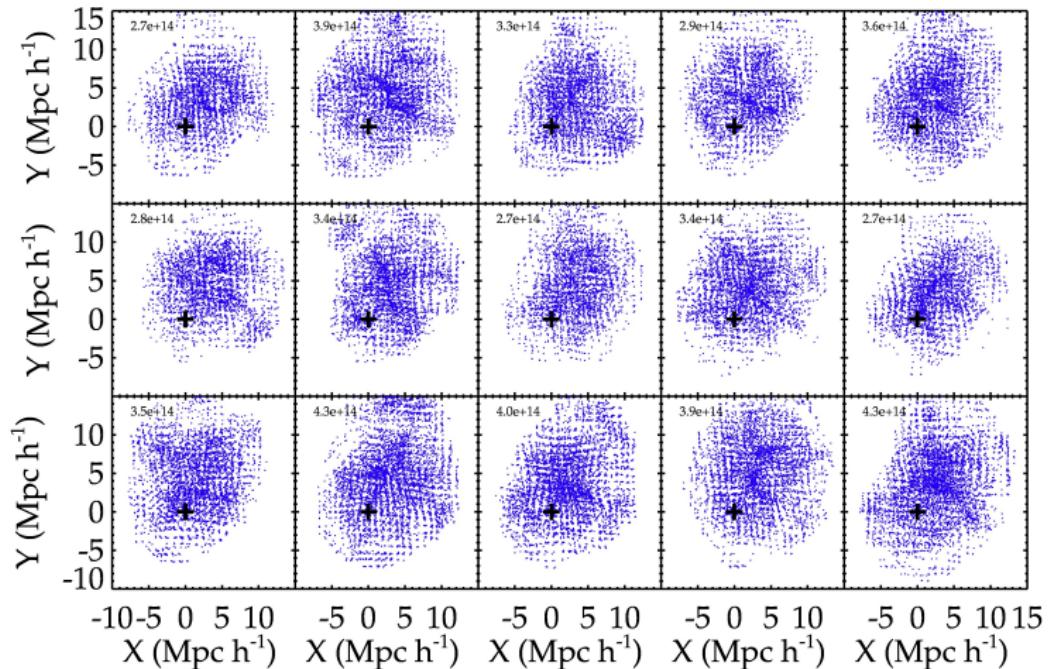
$M_{200}$

# How did the Virgo cluster form?



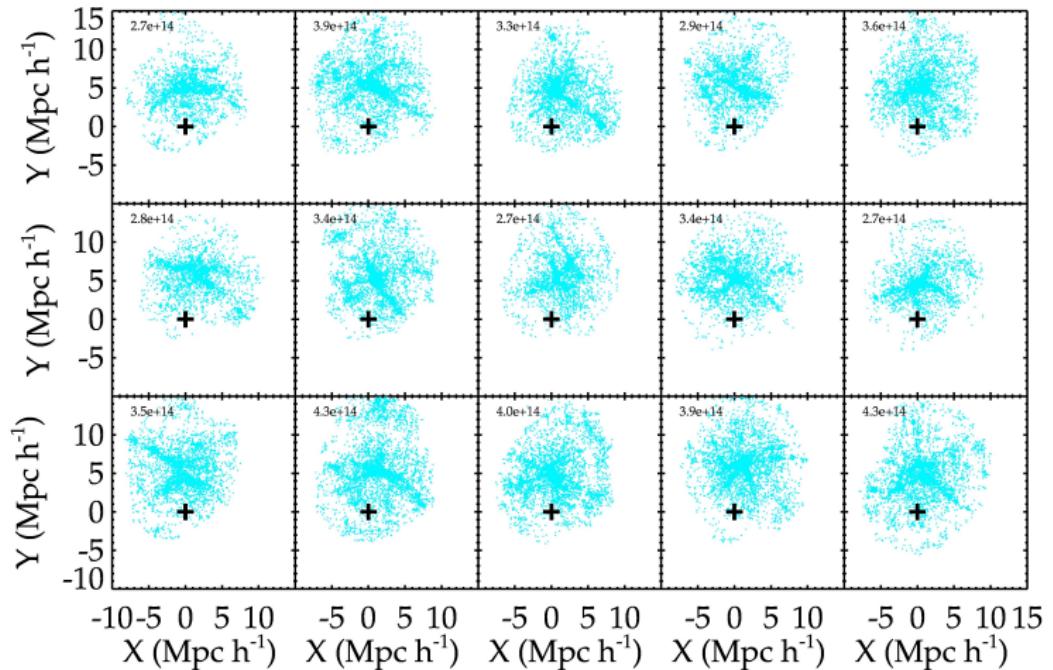
Dark Matter Haloes - Virgo Candidates: Particles at  $z=10$ .

# How did the Virgo cluster form?



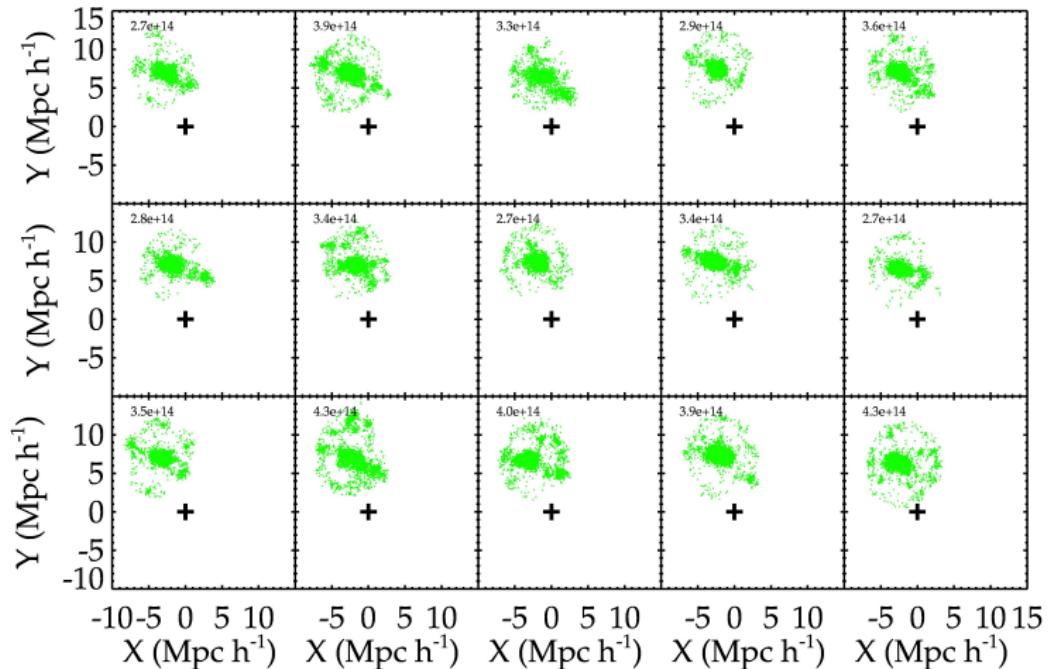
Dark Matter Haloes - Virgo Candidates: Particles at  $z = 5$ .

# How did the Virgo cluster form?



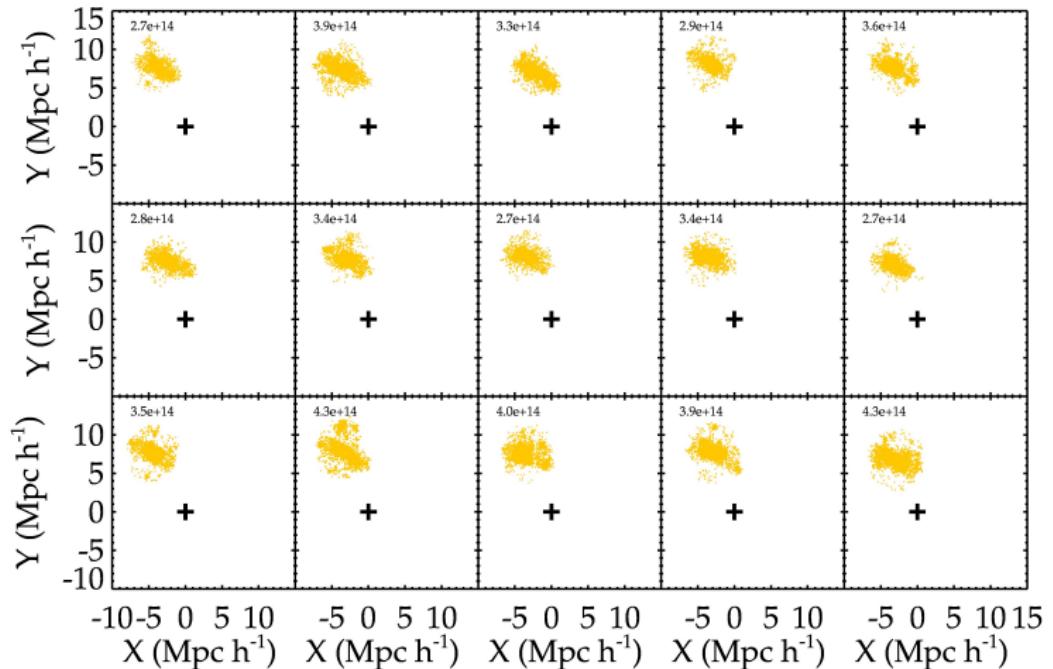
Dark Matter Haloes - Virgo Candidates: Particles at  $z=2$ .

# How did the Virgo cluster form?



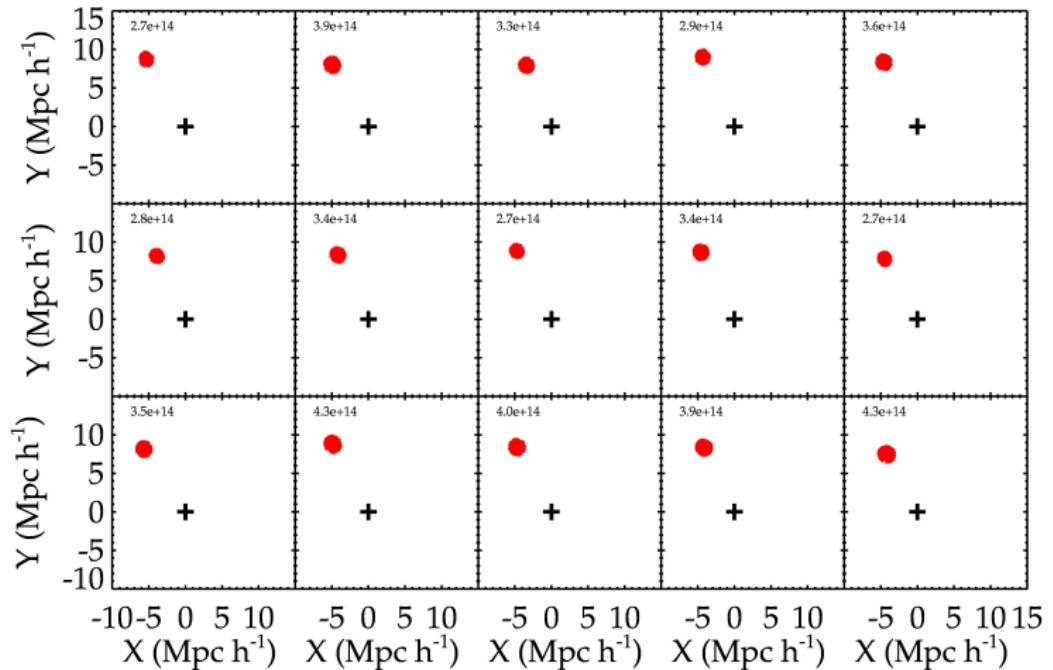
Dark Matter Haloes - Virgo Candidates: Particles at  $z=0.5$

# How did the Virgo cluster form?



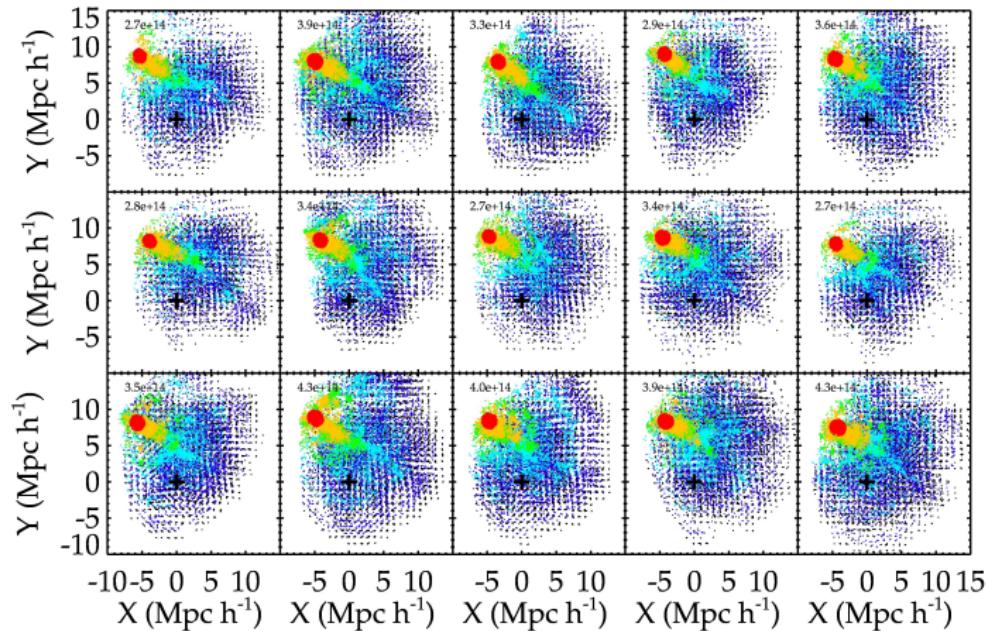
Dark Matter Haloes - Virgo Candidates: Particles at  $z=0.25$

# How did the Virgo cluster form?



Dark Matter Haloes - Virgo Candidates: Particles at  $z=0$ .

# How did the Virgo cluster form?



Dark Matter Haloes - Virgo Candidates:

- Similar formation / evolution

One color per redshift:

10, 5, 2, 0.5, 0.25, 0

## Want more?

- Zone of Obscuration (Sorce et al. 2017): e.g. **Vela Supercluster** (Kraan-Korteweg et al. 2017)
- **Virgo** (Sorce et al. 2016b, Sorce et al. in prep.): e.g. **preferential direction of infall, merging history, etc**
- Local Group (e.g. Carlesi, Sorce et al. 2016) & **Reionization** (Ocvirk et al in prep., Sorce et al. in prep.): e.g. **mass ratio, tangential velocity**
- 3<sup>rd</sup> catalog: **preliminary results**

⇒ Come to discuss & see poster & more movies

# Near Field Cosmology: summary

## Observational side:

- Redshifts (pros/cons)
- Peculiar velocities (via distance indicators, pros/cons)



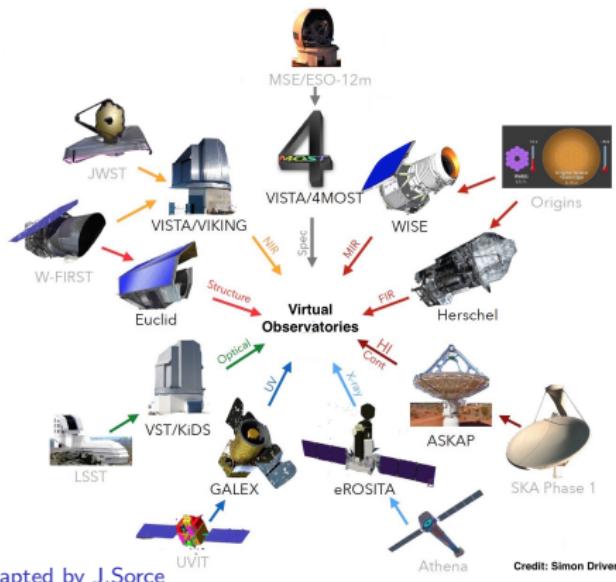
## Numerical side:

- Reduction of the cosmic variance (1<sup>st</sup> Bertschinger 1987)
- Several techniques & applications (backwards/forwards, redshifts/velocities/both)

⇒ wide range of studies possible through comparisons & statistics

# Near Field Cosmology and beyond: prospectives

## Observational side:



## Numerical side:

- hydrodynamical constrained cosmological simulations (**full or zoom** Bertschinger 2001): detailed comparisons with **galaxy** populations to **improve models**
- foreground effect (SZ & SW): **de-bias** large surveys, reach **precision cosmology**

## Acknowledgements

Thank you, Merci, Danke,

Gracias, Grazie, Spasibo,

Mahalo, Xièxie, Arigatô,

Toda, Tak, Dank u,

Obrigada, Cám Ôn, Dziękuję, ...