Unveiling the Cosmic Web from galaxy redshift surveys



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Standard cosmological paradigm

~13.7 billion years after the Big Bang



By reconstructing the primordial fluctuations!

how can we undo gravity?

Inverse approaches

Time-reversal machines

Linear Lagrangian perturbation theory: Zeldovich approximation

Nusser A. & Dekel A., 1992, ApJ, 391, 443 Kolatt T., Dekel A., Ganon G., Willick J. A., 1996, ApJ, 458, 419 BAO reconstructions! Eisenstein D. J., Seo H.-J., Sirko E., Spergel D. N., 2007, ApJ, 664, 675

Padmanabhan N., Xu X., Eisenstein D. J. et al 2012, MNRAS

Non-linear Lagrangian perturbation theory

Moutard 1991; Buchert & Ehlers 1993; Buchert 1993; Bouchet et al 1995; Catelan 1995 Gramann 1993a, 1993b; Monaco & Efstasthiou 1999; Kitaura & Angulo 2012

Least action principle

Peebles J. 1989, ApJ, 344, L53 Nusser A. and Branchini E. 2000, MNRAS, 313, 587 Branchini E. et al. 2002, MNRAS, 335, 53 Croft & Gaztañaga 1997, MNRAS, 285, 793 Frisch U., Matarrese S., Mohayaee R., Sobolevski A., 2002, Nature, 417, 260 Brenier Y., Frisch U., Hénon M., Loeper G., Matarrese S., Mohayaee R., Sobolevskiĭ A., 2003, MNRAS, 346, 501 Lavaux G., 2010, MNRAS, 406, 1007

Linearisation with cosmological perturbation theory



Kitaura F. S. & Angulo R. E., 2012, MNRAS, 425, 2443, arXiv:1111.6617 (see Gaussianisation: Neyrinck et al 2009, 2011)

Time-reversal machines

Comparison between the reconstructed initial density field and true initial conditions from Millennium Run N-body simulation



Kitaura F. S. & Angulo R. E., 2012, MNRAS, 425, 2443, arXiv:1111.6617

pushing down to scales of 5 Mpc/h



1.00

 $k [h \text{Mpc}^{-1}]$

Velocity field reconstruction

Comparison between the reconstructed normalised divergence of the velocity field and true one from Millennium Run



Kitaura F. S., Angulo R. E., Hoffman Y. & Gottlöber S. 2012, *MNRAS*, 422, 2422, *arXiv*:1111.6629 **remember Adi Nusser's talk**

Velocity field reconstruction

Comparison between the reconstructed x-component of the velocity field and true one from Millennium Run



Kitaura F. S., Angulo R. E., Hoffman Y. & Gottlöber S. 2012, MNRAS, 422, 2422, arXiv:1111.6629

Time-reversal machines

Linear or higher order Lagrangian perturbation theory

needs to smooth the density field on a grid on scales of >~ 5 Mpc/h! lead to an aliased estimate of the primordial density fluctuations!

Least action principle

do not lead to an estimate of the primordial density fluctuations but to a distribution of matter tracers at the initial conditions!

the inverse approach severely suffers from shell crossing!

Statistical approaches

Bayesian approach

$$\mathcal{P}(\mathbf{s}|\mathbf{d},\mathbf{p}) = \frac{\mathcal{P}(\mathbf{s}|\mathbf{p})\mathcal{P}(\mathbf{d}|\mathbf{s},\mathbf{p})}{\int \mathrm{d}\mathbf{s}\,\mathcal{P}(\mathbf{s}|\mathbf{p})\mathcal{P}(\mathbf{d}|\mathbf{s},\mathbf{p})},$$

Gaussian case: Wiener filter:

Bayesian derivation in a cosmological large-scale structure context! Zaroubi S., Hoffman Y., Fisher K. B., Lahav O., 1995, ApJ, 449, 446

Bunn E. F., Fisher K. B., Hoffman Y., Lahav O., Silk J., Zaroubi S., 1994, ApJ, 432, L75 Fisher K. B., Lahav O., Hoffman Y., Lynden-Bell D., Zaroubi S., 1995, MNRAS, 272, 885 Webster M., Lahav O., Fisher K., 1997, MNRAS, 287, 425 Erdogdu P., Lahav O., Zaroubi S., Efstathiou G., Moody S., Peacock J. A., Colless M., Baldry I. K., et al. 2004, MNRAS, 352, 939 Erdogdu P., Lahav O., Huchra J., et al. 2006, MNRAS, 373, 45

Gaussian prior

improved Wiener filter from Poisson likelihood applied to Sloan



improved Wiener filter from Poisson likelihood applied to Sloan



can we go beyond the Wiener filter?



Lognormal prior + Poisson/Gaussian likelihood

Lognormal-Poisson model

introduced by:

Kitaura F. S. & Enßlin T. A., 2008, MNRAS, 389, 497, arXiv:0705.0429 Kitaura F. S., Jasche J. & Metcalf R. B., 2010, MNRAS, 403, 589, arXiv:0911.1407

 $P(\mathbf{\Phi}|\mathbf{N},\mathbf{S}) \propto G(\mathbf{\Phi})$

$$\times \prod_{k} \frac{\left(w_k \bar{N} \left(1 + b \left(\exp\left(\Phi_k + \mu\right) - 1\right)\right)\right)^{N_k} \exp\left(-w_k \bar{N} \left(1 + b \left(\exp\left(\Phi_k + \mu\right) - 1\right)\right)\right)}{N_k!},$$

applications:

we introduced Hamiltonian-sampling of the LSS:

Jasche J. & Kitaura F. S., 2010, MNRAS, 407, 29, arXiv:0911.2496 Jasche J., Kitaura F. S., Li C. & Enßlin T. A., 2010, MNRAS, 409, 355, arXiv:0911.2498 Kitaura F. S., Gallerani S. & Ferrara A., 2012, MNRAS, 420, 61, arXiv:1011.6233

easy to handle!

other nonlinear reconstruction methods: no additional parameters!

Platen E., van de Weygaert R. Jones B. J. T., Vegter G., Aragon Calvo M. A. 2011 *Schaap & van de Weygaert* 2000; the Groningen and Tartu groups! *Einasto J.*+ 2011; *Saar E.* + 2007; *Martinez* + 2005

Lognormal-Poisson and Lognormal-Gaussian model

Gibbs-sampling for density, peculiar velocity fields and power spectra:

first proposed by: *Kitaura F. S. & Enßlin T. A., 2008, MNRAS, 389, 497, arXiv:0705.0429*

for Gaussian case without RSDs: Jasche J., Kitaura F. S., Wandelt B. & Enßlin T. A., 2010, MNRAS, 406, 60, arXiv:0911.2493

for Lognormal case including RSD:

Kitaura F. S., Gallerani S. & Ferrara A., 2012, MNRAS, 420, 61, arXiv: 1011.6233 $P(a \mid a^{(j)} \mid S \mid d^{z})$

$$\begin{array}{lll} \boldsymbol{s}^{(j+1)} & \leftrightarrow & P(\boldsymbol{s} \mid \boldsymbol{v}^{(j)}, \boldsymbol{S}, \boldsymbol{d}^{z}), \\ \boldsymbol{S}^{(j+1)} & \leftrightarrow & P(\boldsymbol{S} \mid \boldsymbol{s}^{(j+1)}), \\ \boldsymbol{v}^{(j+1)} & \leftarrow & P(\boldsymbol{v} \mid \boldsymbol{s}^{(j+1)}), \end{array}$$

Lyman alpha forest reconstruction on simulations

Let us assume that we have reconstructed the density field along the line-of-sight



Nusser & Haehnelt 1999 Gallerani et al 2011

How do we get the 3D density fields correcting for RSDs and sampling the power spectra?

see Avery Meiksin, Nicolas Tejos & Khee-Gan Lee's talk

Lyman alpha forest reconstruction on simulations (N-body I340 Mpc/h side at z=3 from R.Angulo L-Basic)



(see also Pichon et al 2001 with nonlinear least squares approach!)

Linear component BAO reconstruction



Kitaura F. S., Gallerani S. & Ferrara A., 2012, MNRAS, 420, 61, arXiv:1011.6233

Redshift-space distortions correction



Limits of the Lognormal approximation

Poisson subsample of dark matter particles with radial selection function from Millennium Run N-body simulation on scales of 3.9 Mpc/h



Springer Series in Astrostatistics, 143, *arXiv:*1112.0492 wrong PDF at underdense regions, but right at high-density regions! right two-point statistics The lognormal approximation has the wrong 3pt statistics! However, in a Bayesian approach the likelihood can weight more in the presence of good enough data!



calculation by Martin White et al 2014

Can we improve the likelihood?

Biasing

- non-linear
- scale dependent
- non-local
- * stochastic

Let us imagine we would know the halo/galaxy density field, i.e. the expected number of halos/ galaxies per finite volume (cell). Stochastic biasing

$$N_h \curvearrowleft \mathcal{P}(N_h | \rho_h)$$

caution! we still need to know the deviation from Poissonity!

over-dispersion modelled by the NB PDF:

$$P(N_i \mid \lambda_i, \beta) = \frac{\lambda_i^{N_i}}{N_i!} \frac{\Gamma(\beta + N_i)}{\Gamma(\beta)(\beta + \lambda)^{N_i}} \frac{1}{(1 + \lambda/\beta)^{\beta}}$$

Kitaura, Yepes & Prada 2014, MNRAS, arXiv:1307.3285

non-Poissonian PDFs: Saslaw W.C., Hamilton A.J.S., 1984, ApJ, 276, 13 Sheth R. K., 1995, MNRAS, 274, 213

stochastic bias: remember Tobias Baldauf's talk Dekel A., Lahav O., 1999, ApJ, 520, 24 Sheth R. K., Lemson G., 1999, MNRAS, 304, 767 and many more see references in e.g. Kitaura et al 2013 like Baldauf+13 Somerville et al 2001, MNRAS, 320, 289 Casas-Miranda et al 2002, MNRAS, 333, 730

remember Miguel's talk! *Neyrinck M et al 2013; Aragon-Calvo M. 2013*



Deterministic biasing parametrization

Fry & Gaztañaga 1993

$$\rho_h = f_h^a \sum_i a_i \delta_{\mathrm{M}}^i$$

see Luigi Guzzo & Sylvain de la Torre's talk *Cen & Ostriker 1993; de la Torre & Peacock 2013*

$$\rho_h = f_h^b \exp\left[\sum_i b_i \log\left(1 + \delta_{\rm M}\right)^i\right]$$

Kitaura, Yepes & Prada 2014 + *Neyrinck et al* 2014

$$\rho_{h} = f_{h} \,\theta(\rho_{\rm M} - \rho_{\rm th}) \,\rho_{\rm M}^{\alpha} \,\exp\left[-\left(\frac{\rho_{\rm M}}{\rho_{\epsilon}}\right)^{\epsilon}\right]$$

thresholding: *Kaiser* 1984

Neyrinck M et al 2013; Aragon-Calvo M. 2013



Perturbative approaches to model BAOs

we are performing mocks for BOSS, 4MOST, JPAS, EUCLID,... collaborators: F. Prada, G. Yepes, V. Müller, C. Scoccola, C.-H. Chuan, H. Gil-Marin, S. Rodriguez

Reference N-body simulation (Gadget): BIGMULTIDARK Volume: (2500 Mpc/h)^3 Hess et al in prep Number of particles: 3840^3 (2M cpu hs) Prada et al in prep halos selected with bdm (density peaks) according to vmax>~350 km/s (LRGs)

-> I consider 8 sub-volumes of (1250 Mpc/h)^3

Simulations with PATCHY: PerturbAtion Theory Catalog generator of Halo/galaxY distributions

Volume: (1250 Mpc/h)^3 Grid number of cells: 512^3 Resolution of the grid: (2.4 Mpc/h)^3 same cosmology (Planck-like) same redshift: z=0.577 on my laptop (quad core i7, 4 cpus+4 virtual cpus, 8 Gb RAM) about 15 mins

53 times lower resolution required!

Calibration of PATCHY with N-body simulations

Kitaura, Yepes & Prada 2014, MNRAS, arXiv:1307.3285 and Kitaura+ in prep real-space redshift-space



1st time perturbation theory matches the two-point statistics in the nonlinear regime

Three-point function

calculations by Volker Müller



Triangle sides: s, su, and s(u+v)

The 3-point function is described by the hierarchical ansatz Q(s,u,v) = (s,u,v)/((r12)(r23)+c.c.), i.e. the increase from v=0 to 1 means transition to linear structures.

ALPT in PATCHY



calculations by Hector Gil-Marin

Bispectrum



Can we use this biasing description within Bayesian inference?
Bayesian density reconstruction from biased galaxy data (e.g. eLGs)

 $\rho_{h} = f_{h} \,\theta(\rho_{\rm M} - \rho_{\rm th}) \,\rho_{\rm M}^{\alpha} \,\exp\left[-\left(\frac{\rho_{\rm M}}{\rho_{\epsilon}}\right)^{\epsilon}\right] \quad Ata, \, Kitaura \,\mathcal{E} \, M \ddot{u}ller \, to \, be \, submitted$

See Metin Ata's poster!

$$P(N_i \mid \lambda_i, \beta) = \frac{\lambda_i^{N_i}}{N_i!} \frac{\Gamma(\beta + N_i)}{\Gamma(\beta)(\beta + \lambda)^{N_i}} \frac{1}{(1 + \lambda/\beta)^{\beta}}$$



Bayesian approach with the ARGO-code Kitaura & Enßlin 2008; Kitaura et al 2010; Kitaura et al 2012

Bayesian density reconstruction from biased galaxy data (e.g. eLGs)



Ata, Kitaura & Müller to be submitted

1st time a nonlinear scale-dependent stochastic bias is explicitly implemented in a Bayesian framework

Bayesian density reconstruction from biased galaxy data (e.g. eLGs)

unbiased DM field reconstructions in terms of cell-to-cell correlation and two-point statistics!



Ata, Kitaura & Müller to be submitted

Can we improve the prior?

Expansion of the Lognormal prior

see also talk by Dmitri Pogosyan and Sandrine Codis!

If we know the higher order correlations for the logarithm of the density field, we can perform multidimensional Edgeworth expansions:

Kitaura F. S., 2012, MNRAS, 420, 2737, arXiv:1012.3168

$$P(\Phi) = (\det(\mathbf{S}))^{-1/2} G(\boldsymbol{\nu})$$
(43)

$$\times \left[1 + \frac{1}{3!} \sum_{i'j'k'} \langle \Phi_{i'} \Phi_{j'} \Phi_{k'} \rangle_{c} \sum_{ijk} S_{ii'}^{-1/2} S_{jj'}^{-1/2} S_{kk'}^{-1/2} h_{ijk}(\boldsymbol{\nu})$$
Univariate lognormal expansion introduced by Colombi 1994

$$+ \frac{1}{4!} \sum_{i'j'k'l'} \langle \Phi_{i'} \Phi_{j'} \Phi_{k'} \Phi_{l'} \rangle_{c} \sum_{ijkl} S_{ii'}^{-1/2} S_{jj'}^{-1/2} S_{kk'}^{-1/2} S_{ll'}^{-1/2} h_{ijkl}(\boldsymbol{\nu})$$
Univariate Gaussian expansion Juszkiewicz et al 1995

$$\times \left[\frac{1}{3!3!2} \sum_{j_1...j_6 \in [1,...,6]} \tilde{\epsilon}_{j_1...j_6} \langle \Phi_{i'_{j_1}} \Phi_{i'_{j_2}} \Phi_{i'_{j_3}} \rangle_{c} \langle \Phi_{i'_{j_4}} \Phi_{i'_{j_5}} \Phi_{i'_{j_6}} \rangle_{c} \right]_{10}$$

$$\times \sum_{ijklmn} S_{ii'}^{-1/2} S_{jj'}^{-1/2} S_{kk'}^{-1/2} S_{mn'}^{-1/2} S_{nn'}^{-1/2} h_{ijklmn}(\boldsymbol{\nu}) + \dots \right],$$

If we know the higher order correlations for the logarithm of the density field, we can perform multidimensional Edgeworth expansions:

Kitaura F. S., 2012, MNRAS, 420, 2737, arXiv:1012.3168

$$P(\boldsymbol{\nu}) = G(\boldsymbol{\nu}) \left[1 + S(\boldsymbol{\nu}) + \mathcal{K}(\boldsymbol{\nu}) + \dots\right]$$

$$\boldsymbol{\Phi} \equiv \ln \boldsymbol{\rho} - \langle \ln \boldsymbol{\rho} \rangle \qquad \boldsymbol{\nu}_{i} \equiv \sum_{j} S_{ij}^{-1/2} \Phi_{j}$$

$$S(\boldsymbol{\nu}) \equiv \frac{1}{3!} \sum_{ijk} \kappa_{ijk} h_{ijk}(\boldsymbol{\nu}) = \frac{1}{3!} \sum_{i'j'k'} \xi_{i'j'k'} \tilde{h}_{i'j'k'}(\boldsymbol{\nu}) \qquad (60)$$

$$= \frac{Q_{3}}{3!} \sum_{i'j'k'} \left[S_{i'j'}S_{i'k'} + S_{i'j'}S_{j'k'} + S_{i'k'}S_{j'k'}\right] \tilde{h}_{i'j'k'}(\boldsymbol{\nu})$$

$$= Q_{3} \left[\frac{1}{2} \sum_{i} \Phi_{i}^{2}\eta_{i} - \frac{1}{2} \sum_{i} S_{ii}\eta_{i} - \sum_{i} \Phi_{i}\right], \qquad \text{complicated + expensive!}$$

$$\eta_{i} \equiv \sum_{j} S_{ij}^{-1} \Phi_{j}$$

Can we improve the prior with a physical model?

The reconstruction problem: Lagrangian to Eulerian problem

primordial density fluctuations can be fully characterized by the 2-point correlation function (neglecting non-Gaussianities) the action of gravity can be summarised by the displacement field

Towards forward approaches see Kitaura 2013, Kitaura et al 2012 Jasche & Wandelt 2013, Wang et al 2013, Hess et al 2013

The posterior distribution function of the primordial fluctuations

$\delta(\boldsymbol{q}) \curvearrowleft P(\delta(\boldsymbol{q})|\{\boldsymbol{s}\}, \mathcal{M}_{\Psi}, \mathcal{M}_{v}, \mathcal{B}_{\mathrm{EUL}}, \{p_{\mathrm{c}}\})$

$$oldsymbol{s} = oldsymbol{q} + oldsymbol{\Psi} + (oldsymbol{v}\cdot\hat{oldsymbol{r}})\hat{oldsymbol{r}}/(Ha))$$

The posterior distribution function of the primordial fluctuations

If we ignore the nonlinear stochastic Bias and RSDs and use a 2LPT-Poisson model $\delta(\boldsymbol{q}) \curvearrowleft P(\delta(\boldsymbol{q})|\{\boldsymbol{s}\}, \mathcal{M}_{\Psi}, \{p_{c}\})$ very complex solution, but beautiful, because it is a one-step posterior expression! **solution by** *Jasche & Wandelt* 2013 see Florent Leclerque's talk It can be improved with data preparation: halo model reconstruction and linear RSDs correction solution within Zeldovich approximation by Wang et al 2013

self-consistent RSDs correction is missing see Patrick Bos' talk

radical simplification of the problem ready for arbitrary structure formation model through Gibbs-sampling splitting approach:

Bayesian Networks Machine Learning (artificial intelligence)

KIGEN-code (KInetic GENeration of the initial conditions, japanese origin)

adaptive, particle based likelihood comparison

 $\delta(\boldsymbol{q}) \curvearrowleft P(\delta(\boldsymbol{q})|\{\boldsymbol{q}\}, \mathcal{B}_{\mathrm{LAG}})$

 $\{\boldsymbol{q}\} \curvearrowleft P(\{\boldsymbol{q}\}|\{\boldsymbol{s}\}, \delta(\boldsymbol{q}), \mathcal{M}_{\Psi}, \mathcal{M}_{v}, \{p_{c}\})$

constrained realisation

Bertschinger 87; Hoffman & Ribak 91; van de Weygaert & Bertschinger 96; Jasche & FK 10, FK+12 2) constrained simulation KIGEN-code: 1st self-consistent phase-space reconstruction code

Hamiltonian sampling combined with Gibbs-sampling

Kitaura F. S., 2013, MNRAS, 429, L84, arXiv:1203.4184 Kitaura F. S., Erdogdu P., Nuza S. E., Khalatjan A., Angulo R. E., Hoffman Y. & Gottlöber S., 2012, MNRAS, arXiv:1205.5560

Top Left: matter PDF z=0 and z=127 Top Right: cell-to-cell between mock galaxy vs N-body ICs fields Millennium Run and Reconstructed ICs vs N-body ICs Bottom: Power-spectra ratio and power-spectra

Top Left: de Lucia mock catalog at z=0 Bottom Left: reconstructed ICs Right: Cosmic Web at z=0

We need accurate and efficient structure formation models!

ALPT: Augmented Lagrangian Perturbation Theory

Kitaura F. S. & Heß S. 2013, MNRAS, 435, L78, arXiv:1212.3514

the instability of higher order LPT Sahni V. & Shandarin S. 1996

see also Tassev S. & Zaldarriaga M., 2012, JCAP, 4, 13 for other LPT improvements with transfer functions (less correlated with the N-body solution than ALPT)

2LPT z=0

ALPT z=0

Now we have a complete method!

Applications to the Local Volume

see Elmo Tempel's talk

fogs compression

Kitaura F. S. & Khalatjan A. using Tegmark M. et al., 2004, ApJ, 606, 702

2002 from the IRAS survey with reverse Zeldovich

Klypin A., Hoffman Y., Kravtsov A. V., Gottlöber S., 2003, ApJ, 596, 19

2013: 1st constrained simulation based on a self-consistent phase-space reconstruction code Hess Steffen, Kitaura F. S. & Gottlöber S., 2013, MNRAS, 435, 2065, arXiv:1304.6565

Red dots represent galaxies from the 2MRS survey (2% are randomly augmented in the galactic plane)

Underlying countour represents the DM constrained simulation

Modelling redshift space distortions

Left: constrained simulation based on 2LPT and compressed fogs Right: constrained simulation based on ALPT modelling fogs

Hess Steffen, Kitaura F. S. & Gottlöber S., 2013, MNRAS, 435, 2065

please ask me how we can solve for the nonlinear, stochastic bias!

we obtain unbiased reconstructions in real-space!

If we have the initial conditions then we have the full phase-space information including the peculiar motions!

see Oliver Hahn, Sergei Shandarin, Mikhail Medvedev, Johan Hidding & Marc Neyrinck's talk *plot by Steffen Hess in collaboration with Ralf Kaehler & Tom Abel from KIGEN reconstructions!*

see Marc Davis, Adi Nusser, Mike Hudson, Martin Feix, Jon Loveday, Christina Magoulas's talk reconstructed peculiar velocity field from ALPT!

SGX [Mpc/h]

see Brent Tully's, Adi Nusser's & Maciej Bilicki's talk Local Group motion

Kitaura F. S., Erdogdu P., Nuza S. E., Khalatjan A., Angulo R. E., Hoffman Y. & Gottlöber S., 2012, MNRAS, arXiv:1205.5560

CMB: v=627+-22 [km/s], I=276, b=30

we obtain 83% of the CMB measurement in agreement with LCDM considering the matter within a box of 160 Mpc/h side LCDM predicts 80% of the magnitude at distances of 80 Mpc/h

Hubble flow

Boundary effects? Missing large-scale modes?

we use constrained augmented ALPT simulations to correct for these effects!

we correct the Hubble flow from estimates based on local SN and find better agreement with CMB results!

Hess S., Kitaura F. S. & Seljak U. to be submitted

2012: 1st application of a self-consistent forward approach to real data!

galactic plot of the 2MRS survey (shell between 0 to 80 Mpc/h)

Kitaura F. S., Erdogdu P., Nuza S. E., Khalatjan A., Angulo R. E., Hoffman Y. & Gottlöber S., 2012, MNRAS, arXiv:1205.5560

reconstructed cosmic web (shell between 0 to 80 Mpc/h) now that we have the filaments ... remember Elmo Tempel's talk on filaments!

Detection of missing baryons from public PLANCK data!

Suarez-Velásquez I., Kitaura F.-S., Atrio-Barandela F., Mücket J. P. 2013, ApJ, 769, 7, arXiv:1303.5623

results from PLANCK

see Jukka Nevalainen 's talk

in collaboration with Genova-Santos R. T., Atrio-Barandela F., Mücket J. P. in prep correlations up to 20 deg in all channels after subtraction of CMB (217GHz), if it were from clusters correlation should fade away at less than 1 deg!

Reconstructions of the cosmic web

Nuza Sebastian, Kitaura F. S., Heß S., Libeskind N. & Müller V., arXiv:1406.1004

see Marius Cautun, Bernard Jones, Oliver Hahn & Jaime Forero-Romero's talk and talks on voids: R. v.d. Weygaert, P. Sutter, N. Padilla, A. Hawken, Y.C. Cai, N. Hamaus, E. Ricciardelli, A. Pisani

Is our place in the Universe special?

Nuza Sebastian, Kitaura F. S., Heß S., Libeskind N. & Müller V., arXiv:1406.1004

Missing mass in the Local Universe?

Nuza Sebastian, Kitaura F. S., Heß S., Libeskind N. & Müller V., arXiv:1406.1004

Galaxy morphology-environmental study

Nuza Sebastian, Kitaura F. S., Heß S., Libeskind N. & Müller V., arXiv:1406.1004

see Jounghun Lee & Katarina Kovac's talk

in redshift space

for a similar study from the Wiener reconstruction see Lee & Lee 2008
Galaxy morphology-environmental study

Nuza Sebastian, Kitaura F. S., Heß S., Libeskind N. & Müller V., arXiv:1406.1004

in real space





real observations!



mean over an ensemble of ALPT reconstruction!



high resolution constrained N-body simulation!

Conclusions

- * The galaxy distribution in redshift-space encodes information of the full phasespace information given some assumptions: the primordial fluctuations are closely Gaussian distributed with a given power-spectrum and the Universe isotropic. Then we can test the structure formation models with RSDs which match the data in a forward statistical approach.
- The idea of combining LPT on large-scales with small-scale structure formation corrections is very effective: augmented LPT (ALPT)
- The exponential bias with an exponential cut-off models very well the scale dependent nonlinear deterministic bias
- Deviations from Poissonity are important for precision studies of the LSS.
- RSD can be accurately modelled with ALPT + a stochastic dispersion term
- * We are able to incorporate these models in a Bayesian inference framework.
- These ingredients are crucial for precision reconstructions of primordial fluctuations.
- The range of applications is very wide: CMB-dipole, Bulk flows, Hubble constant, WHIM, kSZ, BAO reconstruction, voids, filaments, clusters, walls identification, environmental studies