

SDSS3-BOSS

Baryon Oscillations Spectroscopic Survey What do we learn from it?

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Work with [Xiaoying Xu](#), [Mariana Vargas](#), [Ross O'Connel](#), [Nishant Agarwal](#), Hy Trac, Martin White, Daniel Eisenstein, Antonio Cuesta, Nikhil Padmanabhan, Anze Slosar, Angela Burden, Jeff Schneider (CS), Barnabas Poczós (CS), Junier Oliver (CS), Andrea Klein (CS),
SDSS3-BOSS Collaboration



Outline

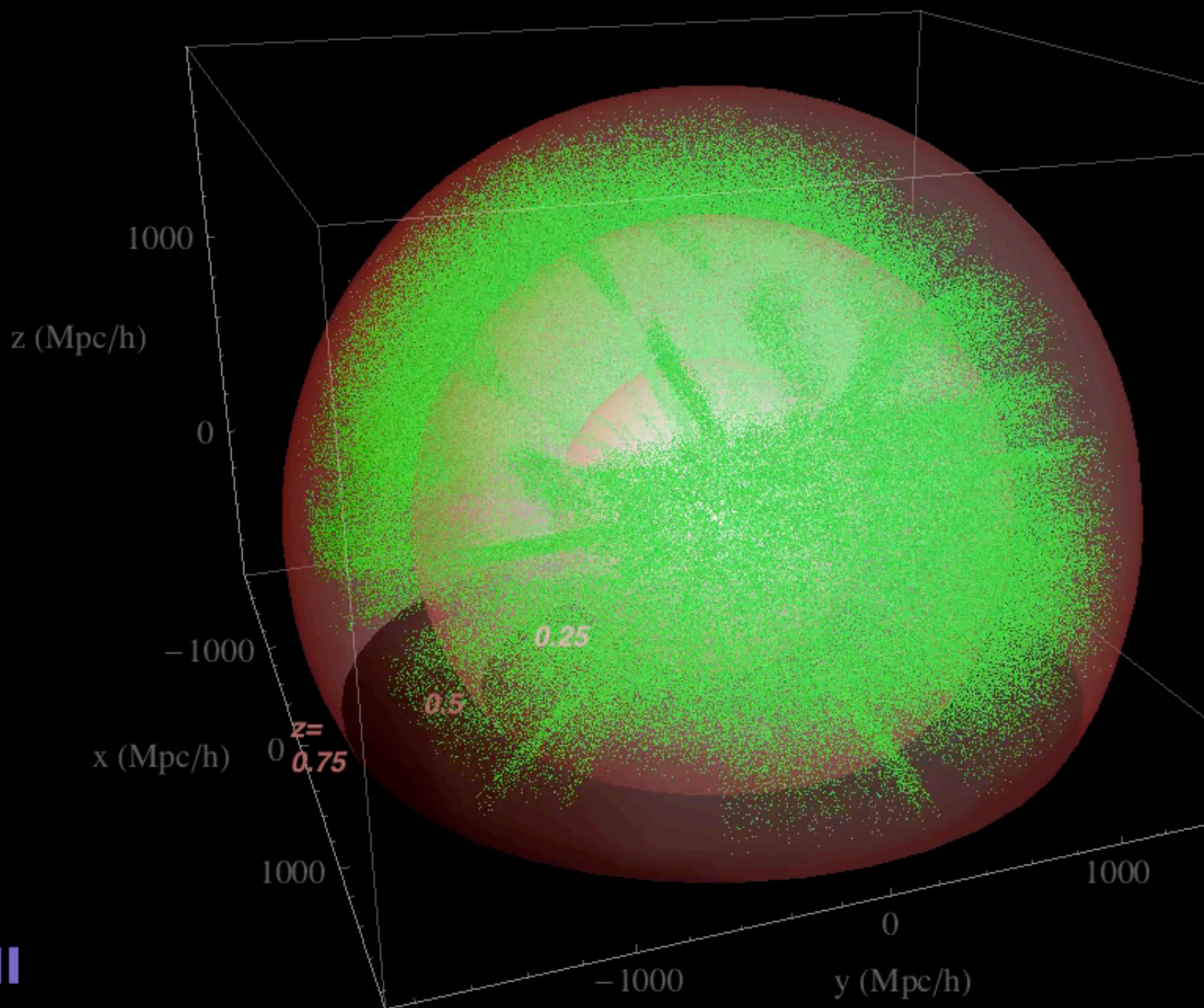
- SDSS III- BOSS :
 - Mini-summary and looking forward
- BAO Reconstruction
- Ultra-fast simulations of the Universe?
- BAO in cross-correlations?

Outline

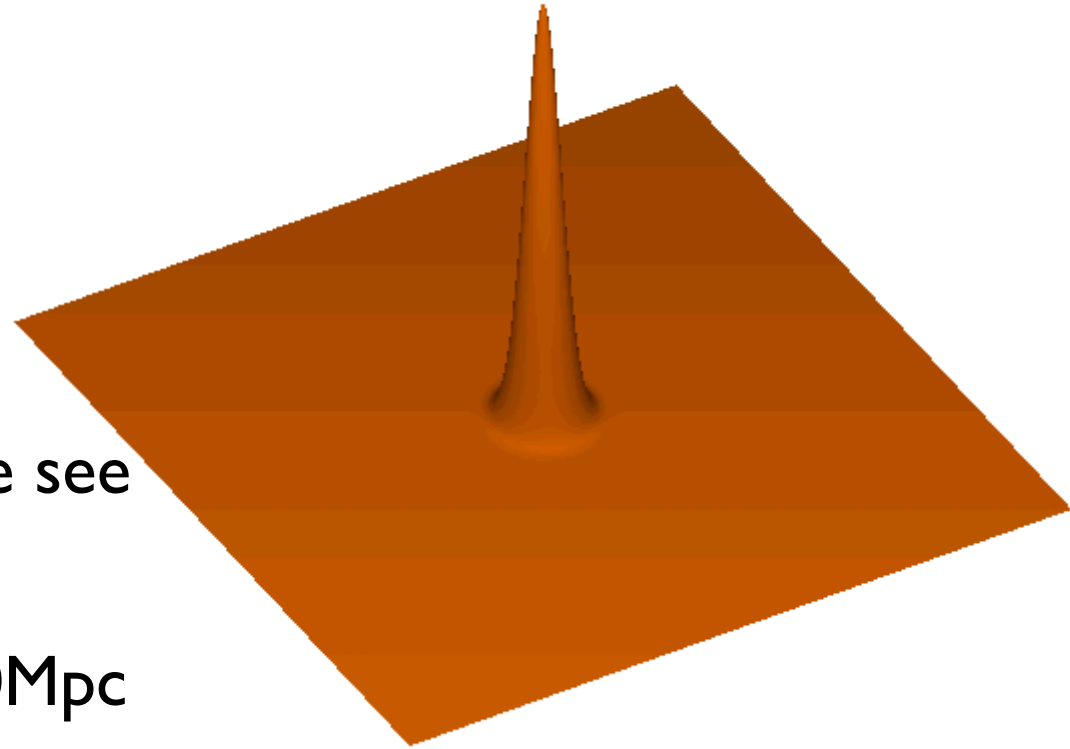
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SDSS3-BOSS Data Release 8 (DR8)

video credit: David Kirkby

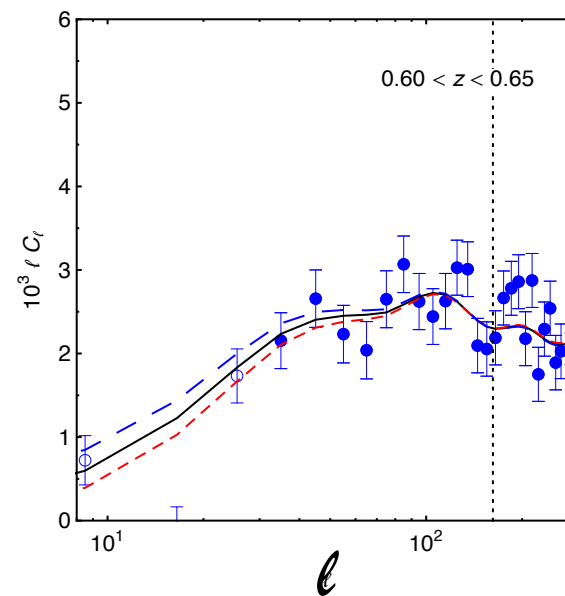
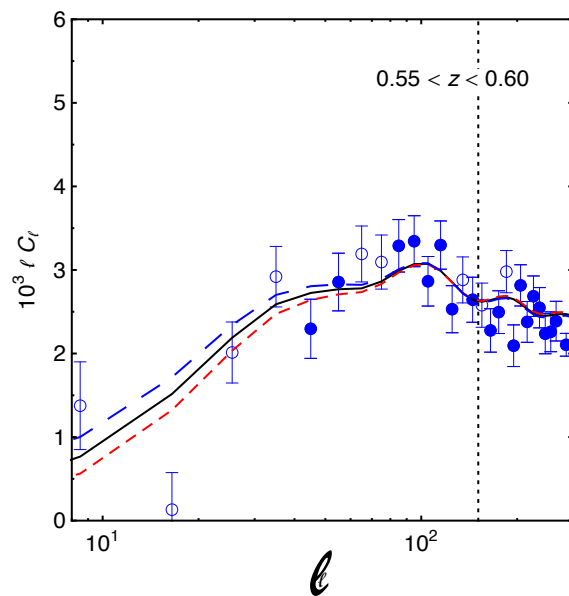
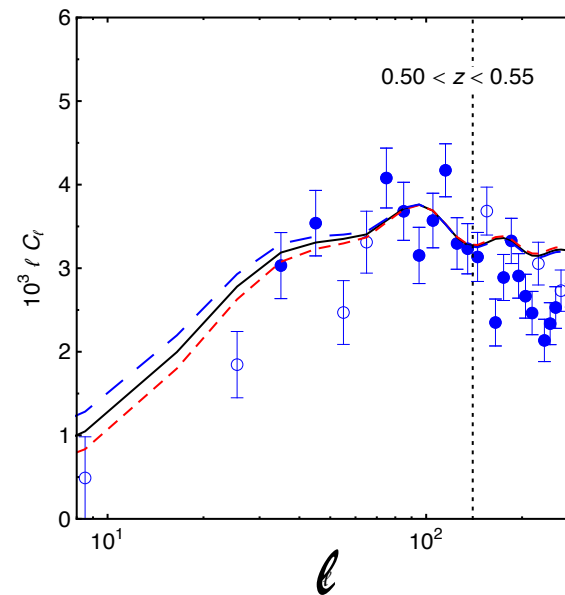
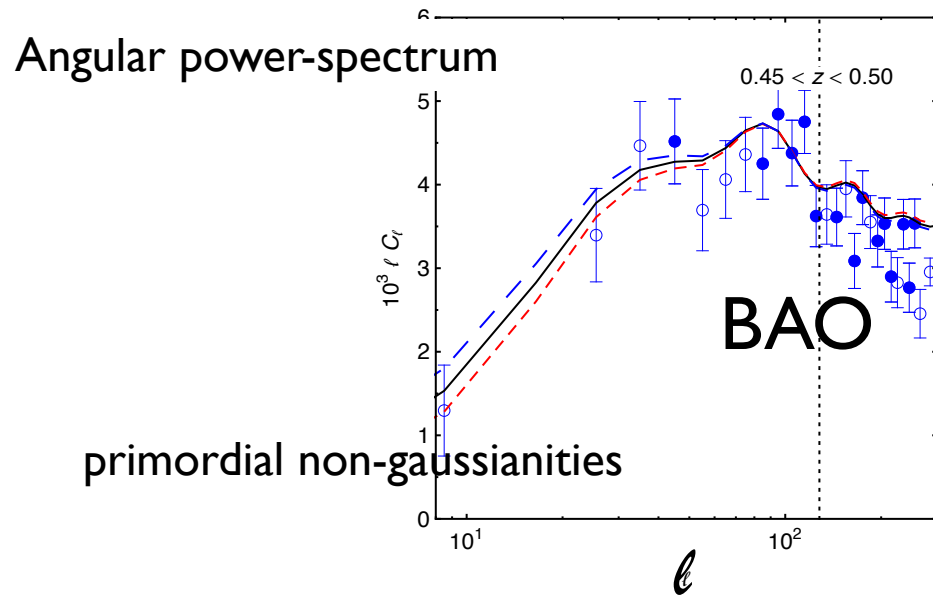


- Each initial overdensity is an over-pressure that launches a spherical sound wave.
- This sound wave travel at 57% speed of light.
- Pressure providing photons decouple at recombination. We see these photons as the CMB
- Wave stalls at a radius of $\sim 150\text{Mpc}$
- Overdensity in shell and center both seed formation of galaxies. Preferred separation of 150Mpc

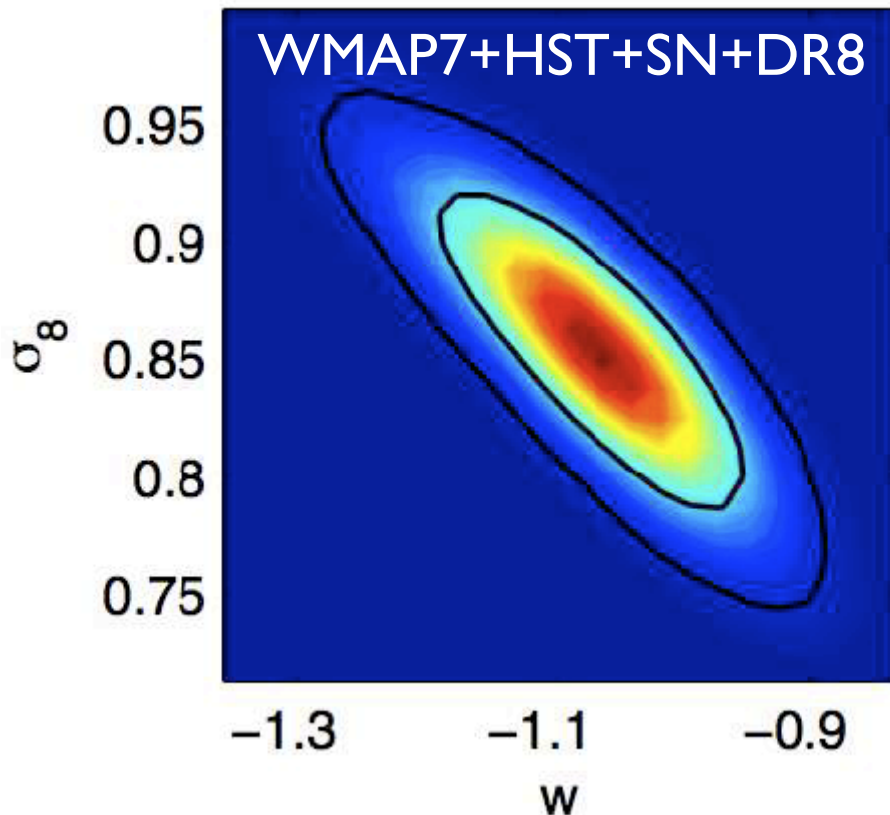


In the beginning, a sound wave ...

What can you do with imaging data?

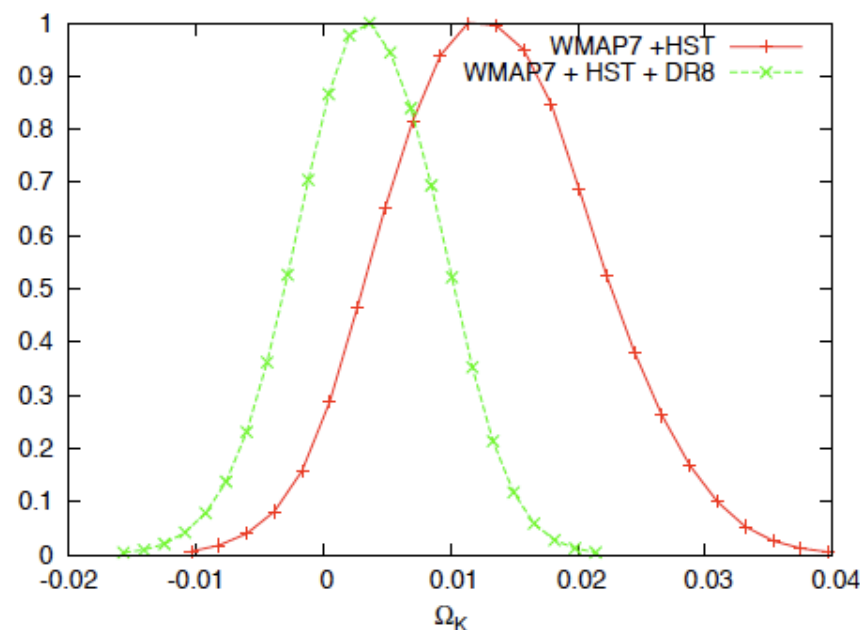
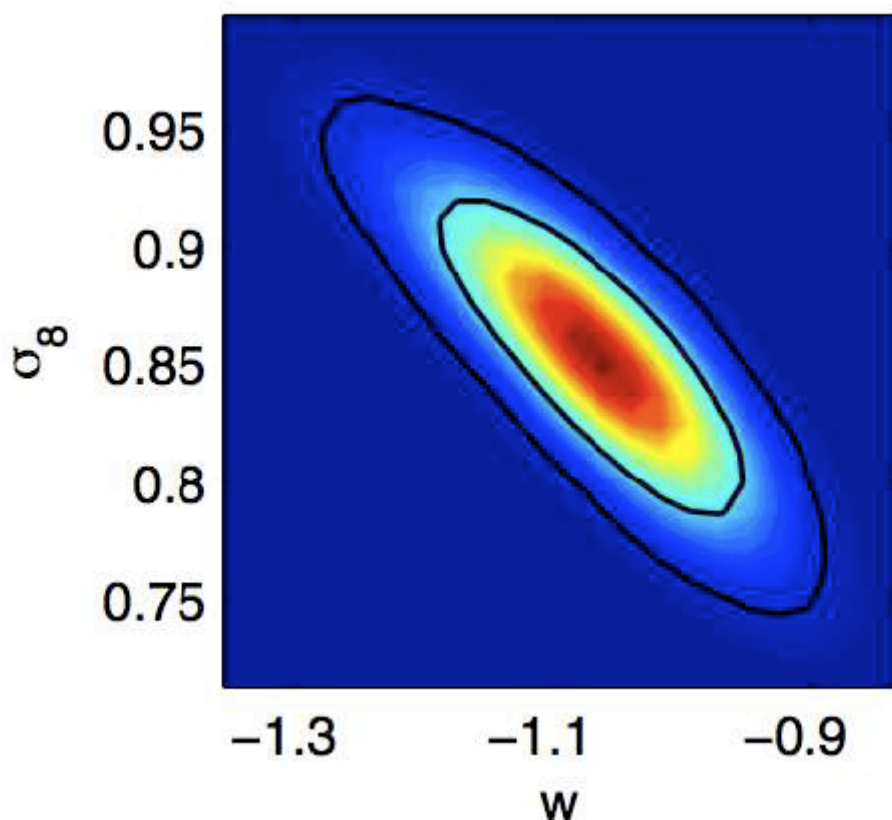


Cosmological Constraints from the overall shape



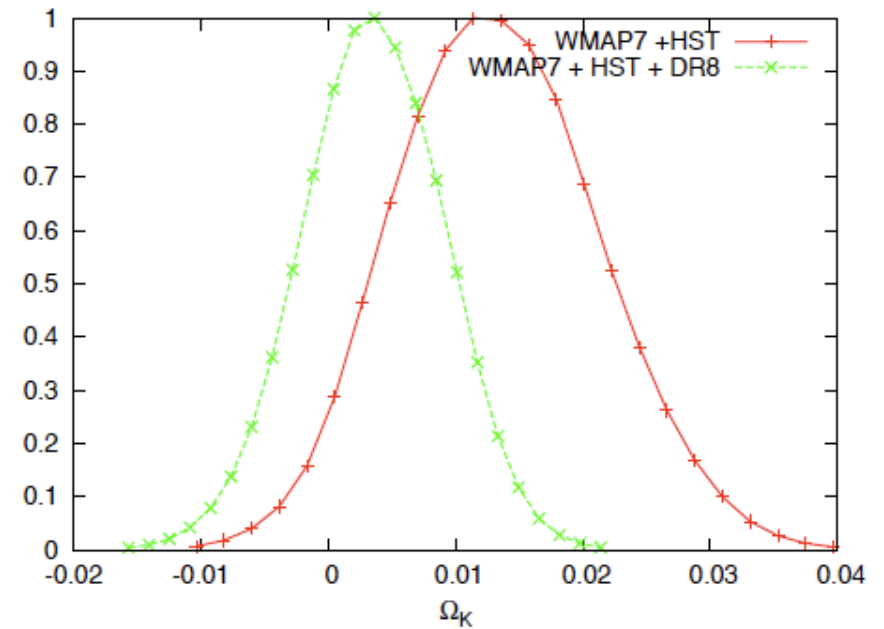
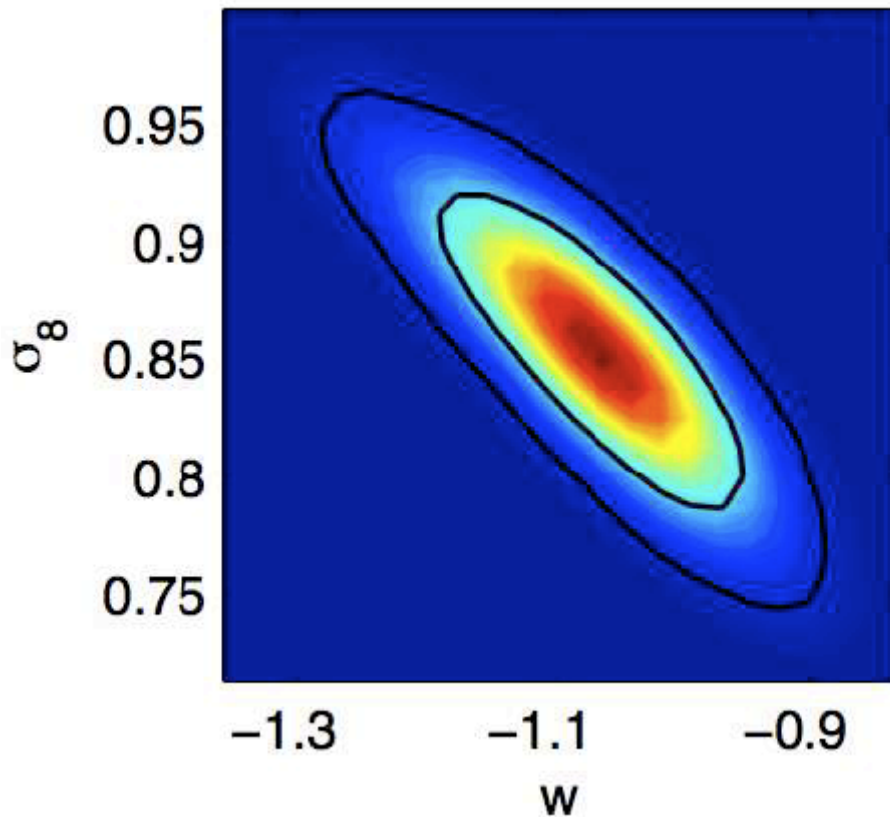
Combining with WMAP7+SN+HST,
Dark Energy equation of state is
constrained to 7% (1-sigma)

Cosmological Constraints from the overall shape

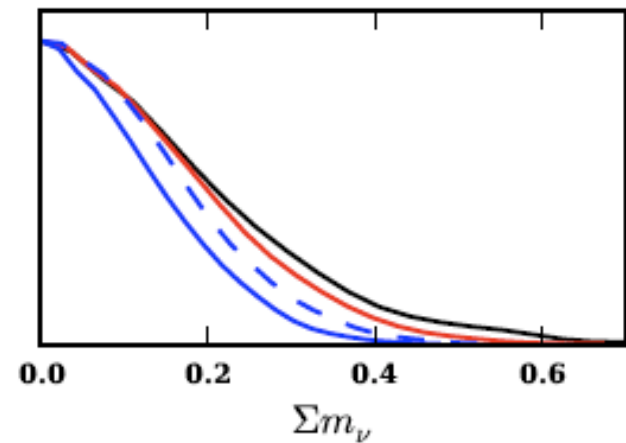


By including DR8 angular clustering (+WMAP+HST), we improve the constraint on flatness of the Universe by **40%** over WMAP7+HST

Cosmological Constraints from the overall shape



The sum of
neutrino masses
is constrained to
less than 0.26eV



What else can you do with imaging data?

What else can you do with imaging data?

Testing initial conditions of the Universe

- The power spectrum of dark matter halos is given by

$$P_{\text{halo}}(M, k, z, f_{\text{NL}}) = [b_1(M, z) + \Delta b(M, k, z, f_{\text{NL}})]^2 P_{\text{matter}}(k, z)$$

with

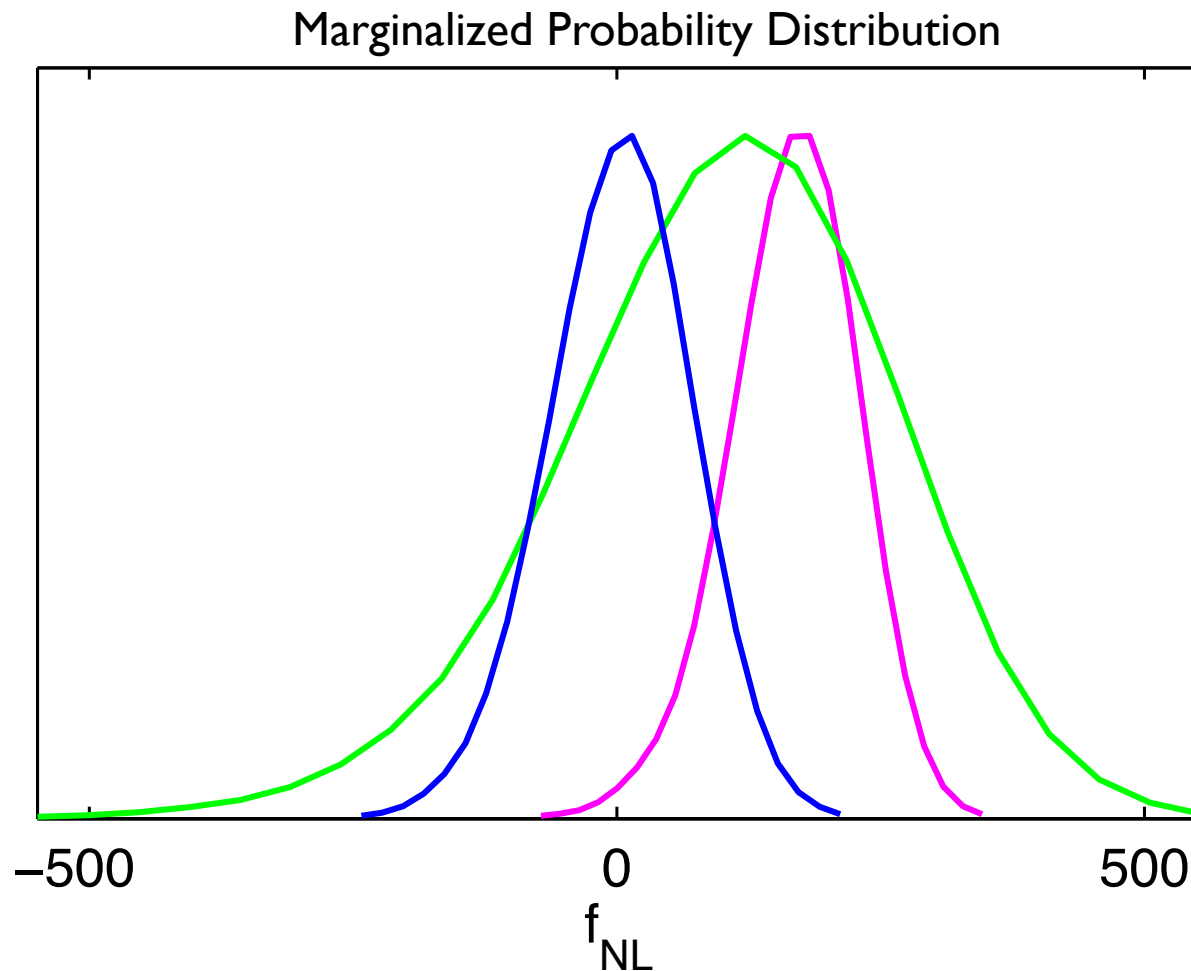
$$\Delta b_{\text{non-Gaussian}} \propto f_{\text{NL}} \frac{(b_1(M, z) - p)}{k^2}$$

S. Matarrese, F. Lucchin, and S. A. Bonometto, 1986; N. Dalal, O. Dore, D. Huterer, and A. Shirokov, 2008; S. Matarrese and L. Verde, 2008; A. Slosar, C. Hirata, U. Seljak, S. Ho, and N. Padmanabhan, 2008

What else can you do with imaging data?

Testing initial conditions of the Universe

Dalal et al., 2008, Slosar et al. 2008...

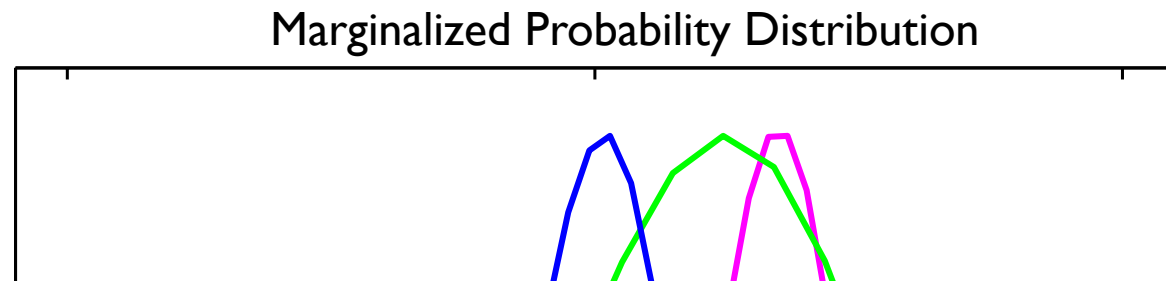


SH, Agarwal, Myers et al. 2013 (under SDSS3 review)

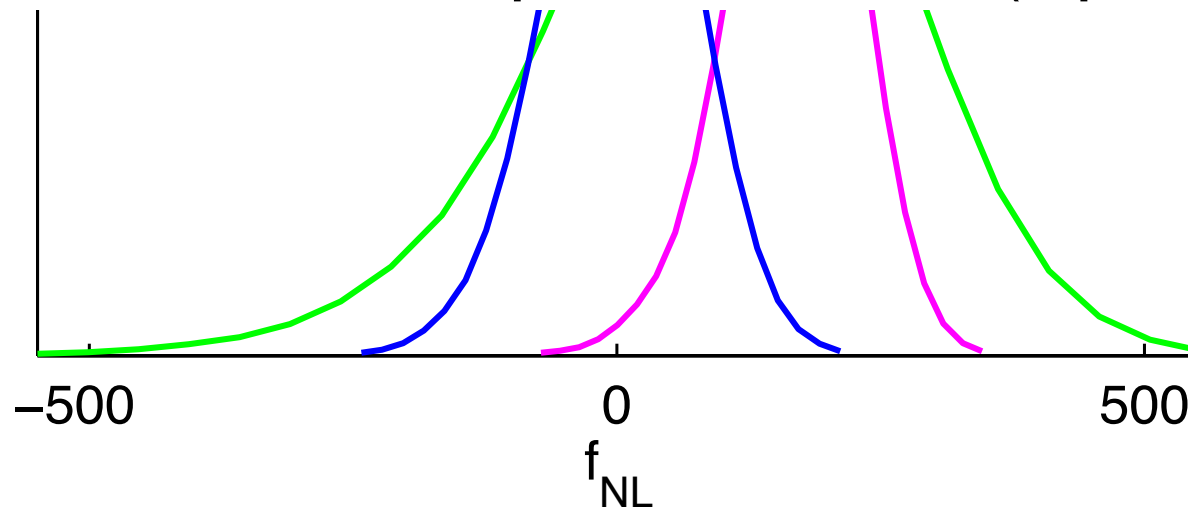
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Dalal et al., 2008, Slosar et al. 2008...



The chronicle of a broken physicist dream....
with 1.6 million quasars over 80 (Gpc/h)^3

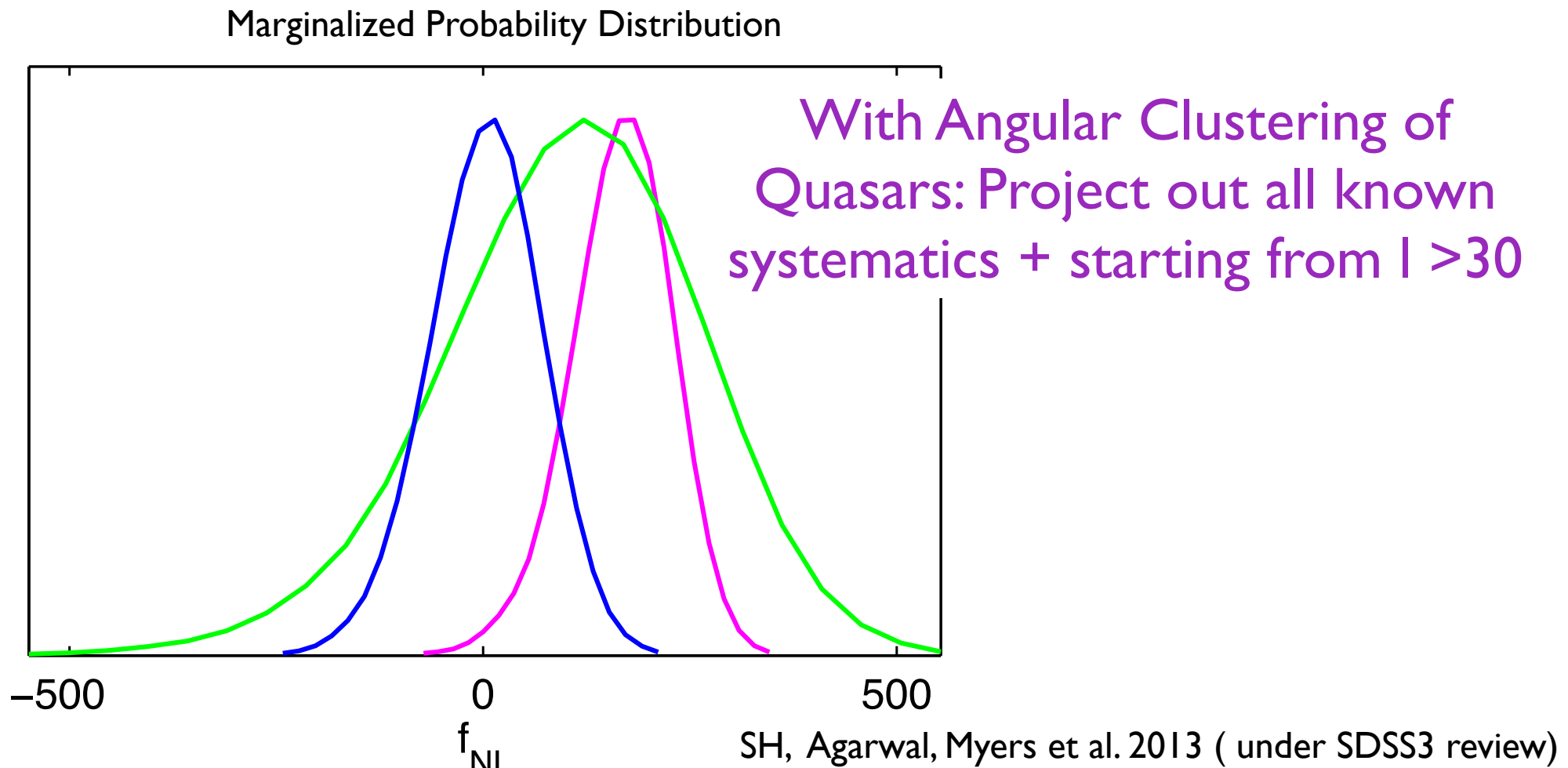


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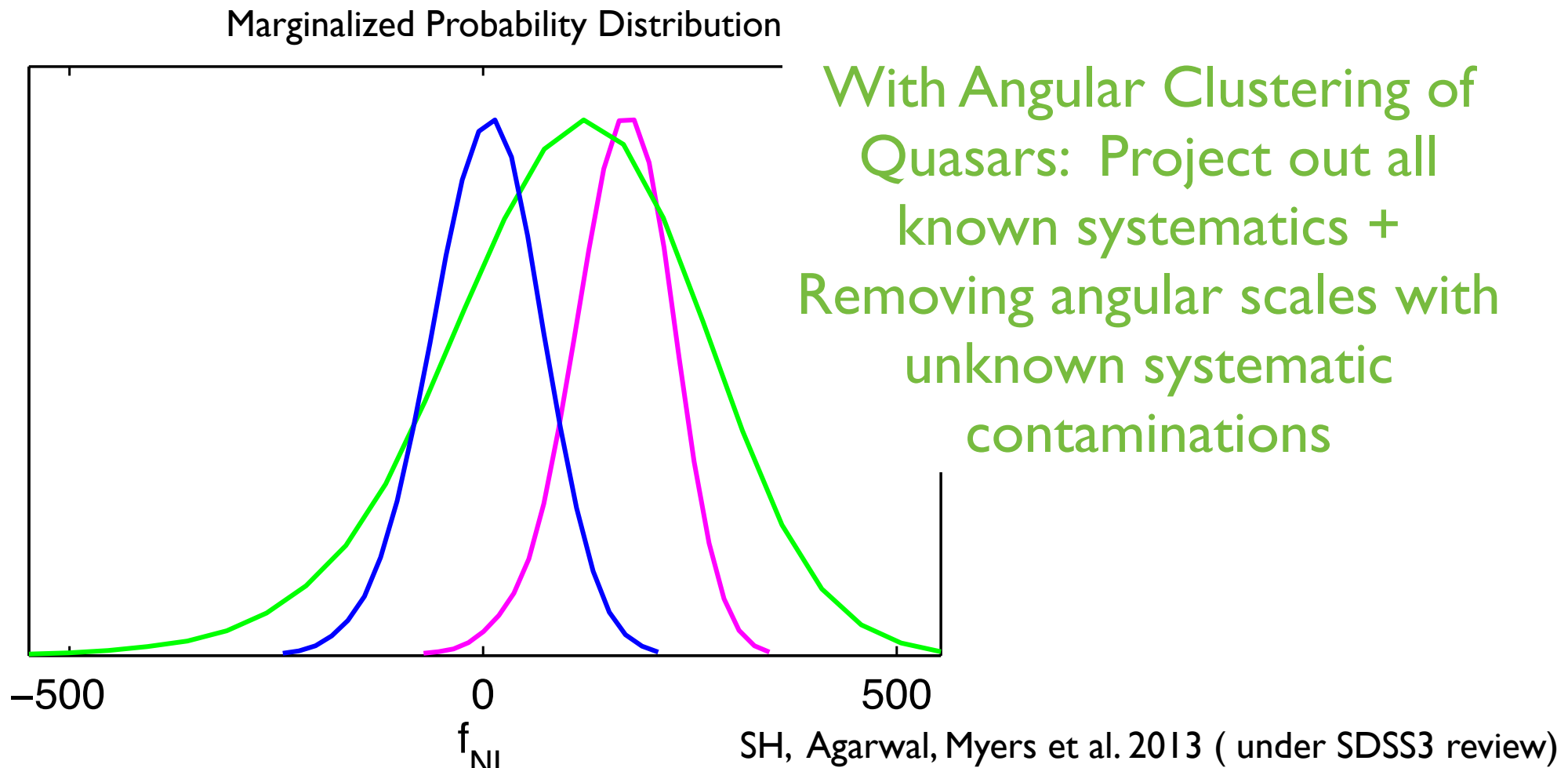
Dalal et al., 2008, Slosar et al. 2008...



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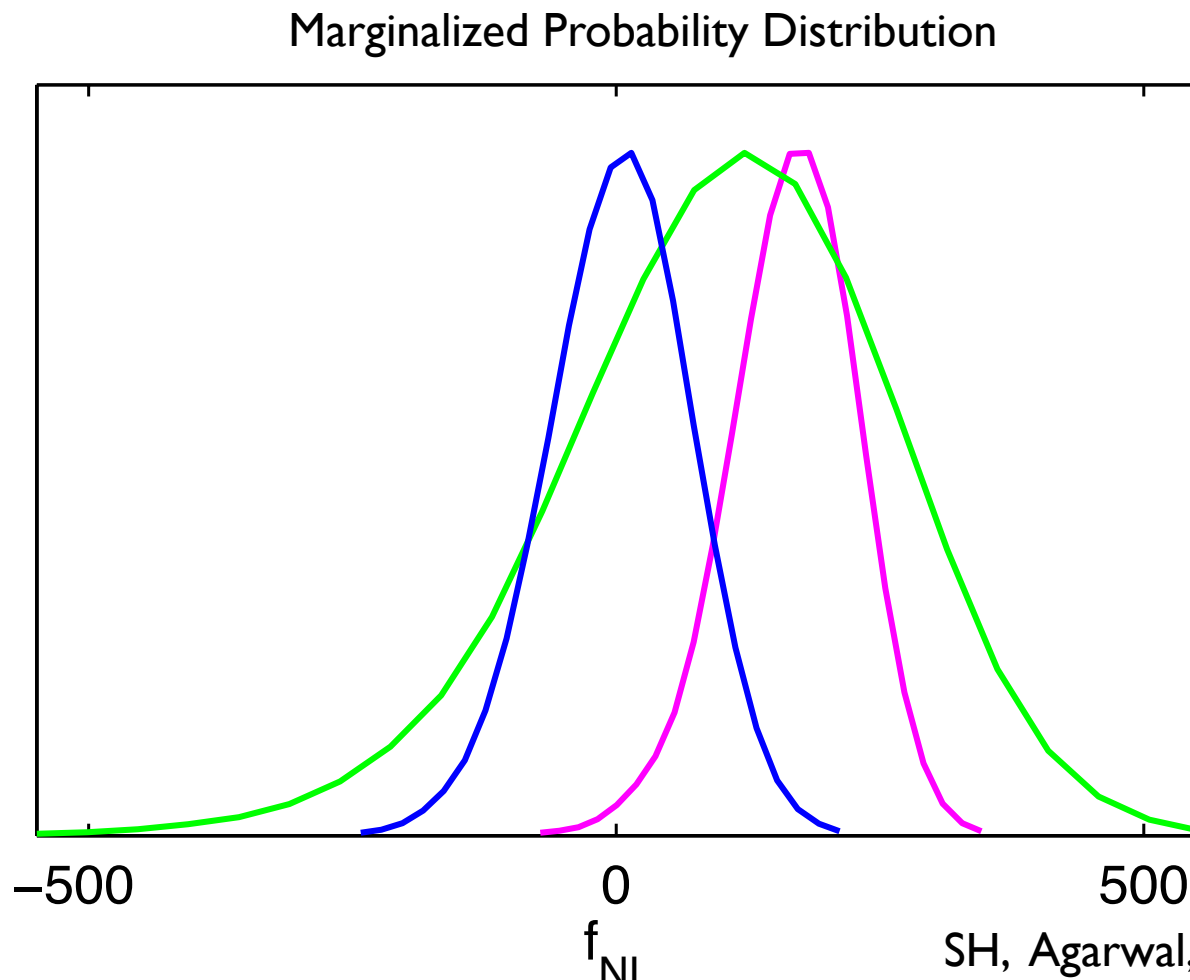
Dalal et al., 2008, Slosar et al. 2008...



What else can you do with imaging data?

Testing initial conditions of the Universe

Dalal et al., 2008, Slosar et al. 2008...



Adding both cleaned
QSOs + LRGs

SH, Agarwal, Myers et al. 2013 (under SDSS3 review)

What else can you do with imaging data?

Testing initial state of inflation !

- For non-trivial initial states, the squeezed limit bispectrum is proportional to

$$1 / (k_S^2 k_L^4) = 1 / (k_S^3 k_L^3) \times (k_S / k_L)$$

- This changes the scale-dependence of the non-Gaussian halo bias: $\Delta b_{\text{non-Gaussian}} \propto 1/k^3$

J. Ganc and E. Komatsu, 2012; I. Agullo and S. Shandera, 2012

What else can you do with imaging data?

Testing initial state of inflation !

$$\Delta b(M, k, z, \mathcal{A}_{\text{NL}}, \alpha) = 3\mathcal{A}_{\text{NL}}(b_1(M, z))[b_1(M, z) - p] \frac{\Omega_m H_0^2}{k^2 (k/k_p)^{\alpha-2} T(k) D(z)},$$

- Therefore, we can parameterize the bias correction as

$$\Delta b_{\text{non-Gaussian}} \propto \mathcal{A}_{\text{NL}}(b_1(M, z)) \frac{(b_1(M, z) - p)}{k^\alpha}$$

- Exact local ansatz: $\alpha = 2$

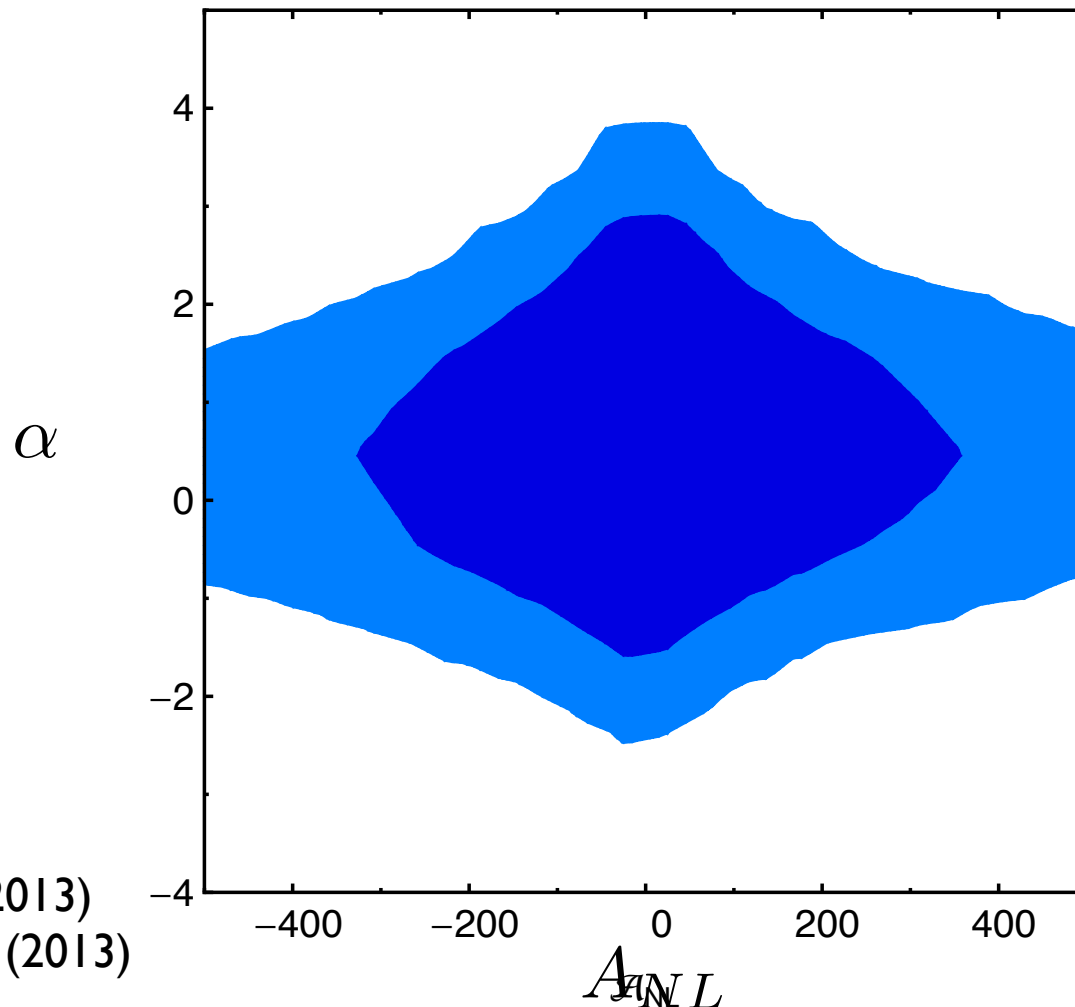
General initial states: $\alpha \sim 3$

Multiple fields: $0 \leq \alpha \lesssim 2 + \mathcal{O}(\epsilon)$ (S. Shandera, N. Dalal, and D. Huterer, 2011; E. Sefusatti, J. R. Fergusson, X. Chen, and E. Shellard, 2012; M. Dias, R. Ribeiro, and D. Seery, 2013)

What else can you do with imaging data?

Testing initial state of inflation !

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Giannantonio et al. (2013)
Agarwal, SH & Shadara (2013)

Testing initial state of inflation !

Why I shouldn't despair yet...

This is what would have happened if we have no systematics!

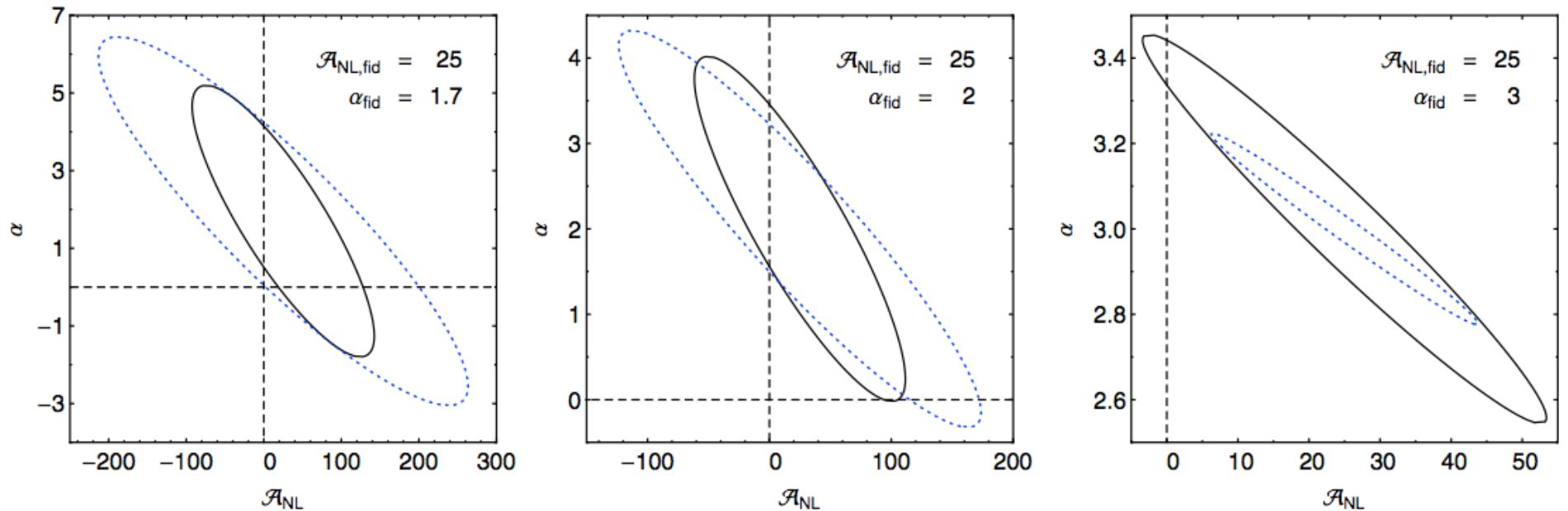


Figure 5. The one-sigma ellipses for LRGs (solid black) and quasars (dotted blue), with fiducial $\mathcal{A}_{\text{NL}}(k_p = 0.1 \text{ Mpc}^{-1}) = 25$ and $\alpha = 1.7, 2, 3$. See also table 5, which lists the results at the pivot points where $\mathcal{A}_{\text{NL}}$ and α are uncorrelated for each choice of α_{fid} and each data set (LRGs or quasars).

What happens when we have
BOSS Spectroscopy ?

What happens when we have BOSS Spectroscopy ?



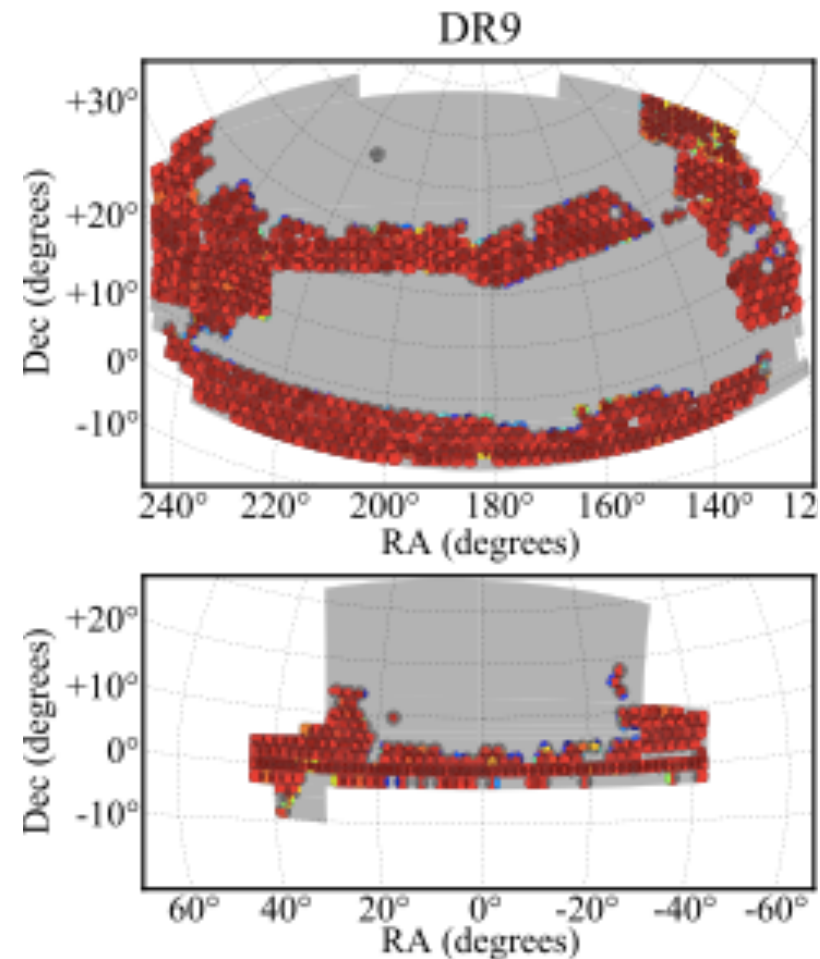
Claudia plugging a BOSS plate and getting it wrong!



SDSS III

Data Release 9-12 Spectroscopy

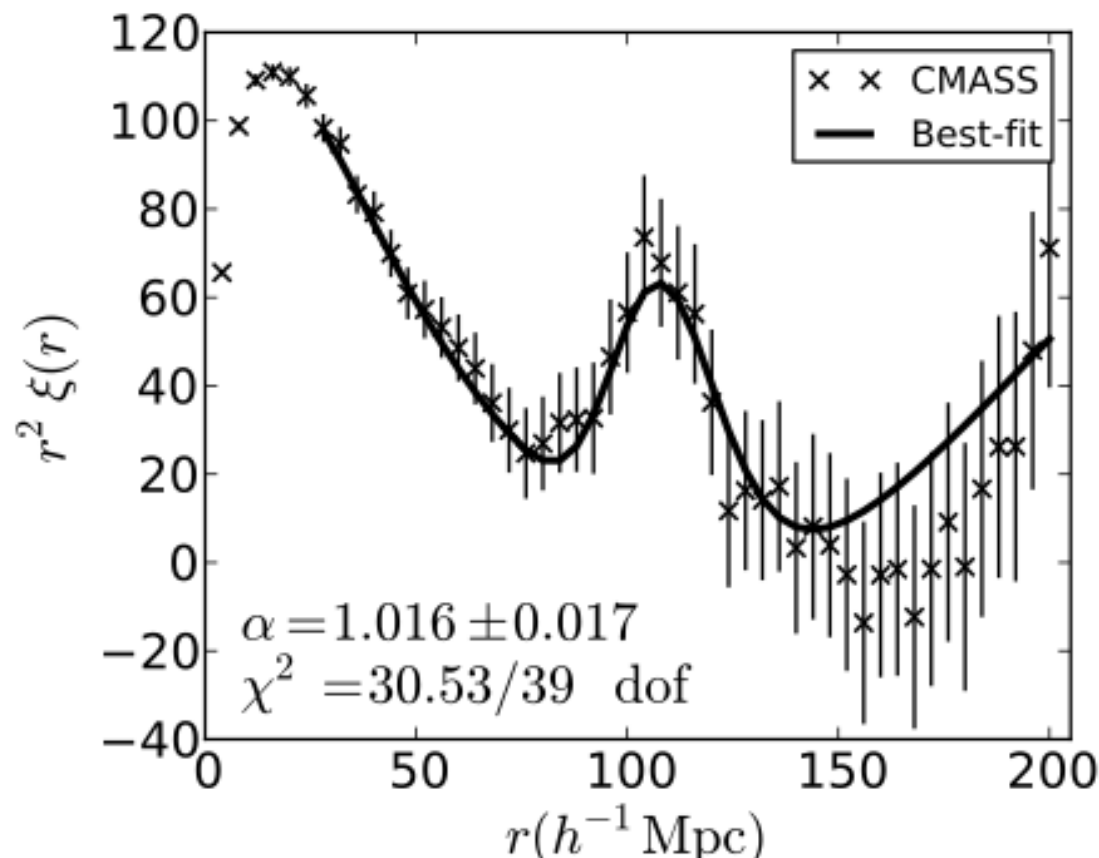
- Public data release:
 - July 2012: 700,000 spectra, 1/3 footprint; + 1.5 million spectra from SDSS I+2 and 14,000 square degrees of imaging
- Our first BAO analysis uses only the higher redshift portion of the sample (called CMASS)
- 264k galaxies over 3275 sqdeg at median $z=0.57$



SDSS III

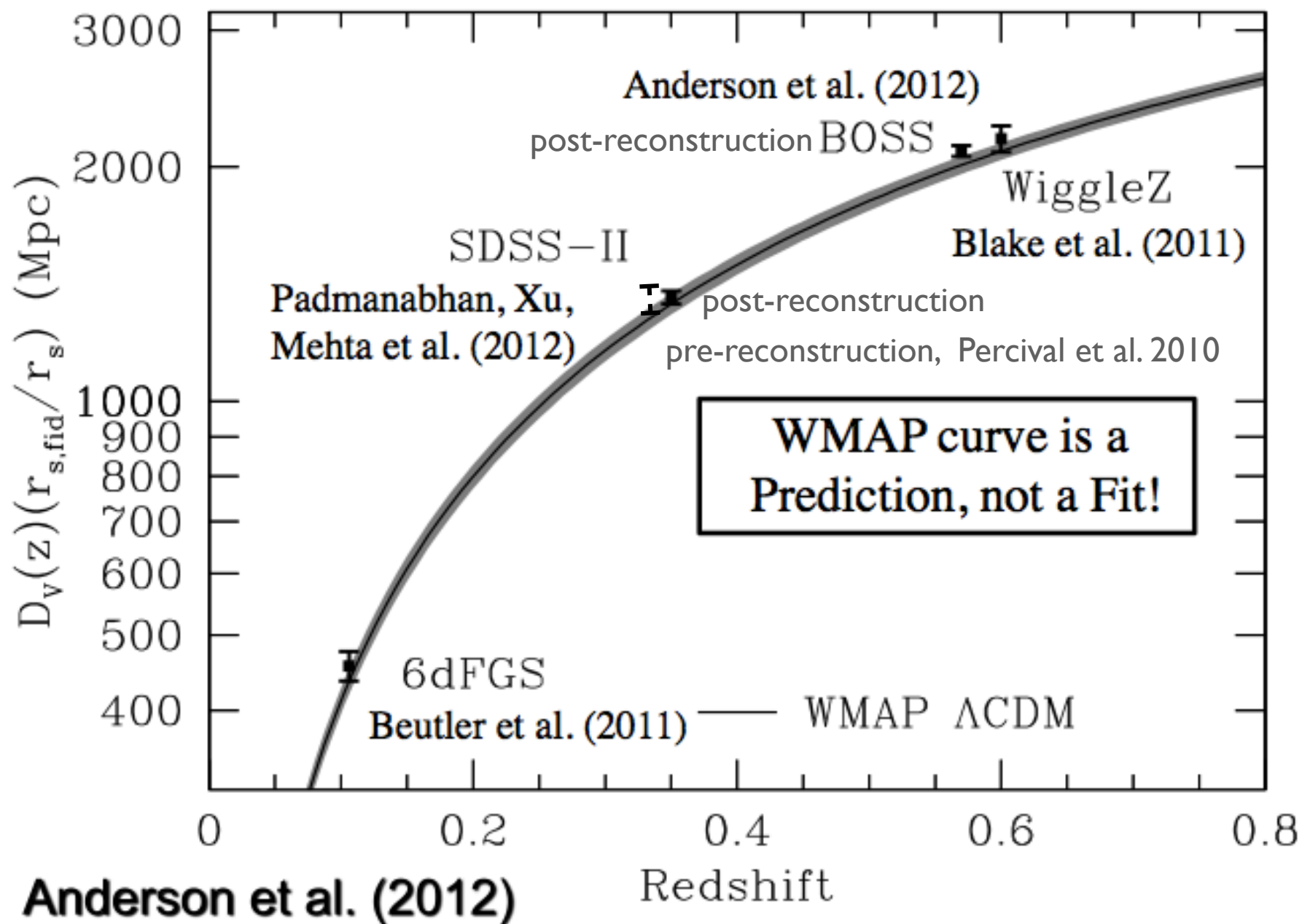
DR9- All Spectroscopy

Extremely clear BAO detection at $z=0.57$;
measure the distance to $z=0.57$ with a 1.7% precision

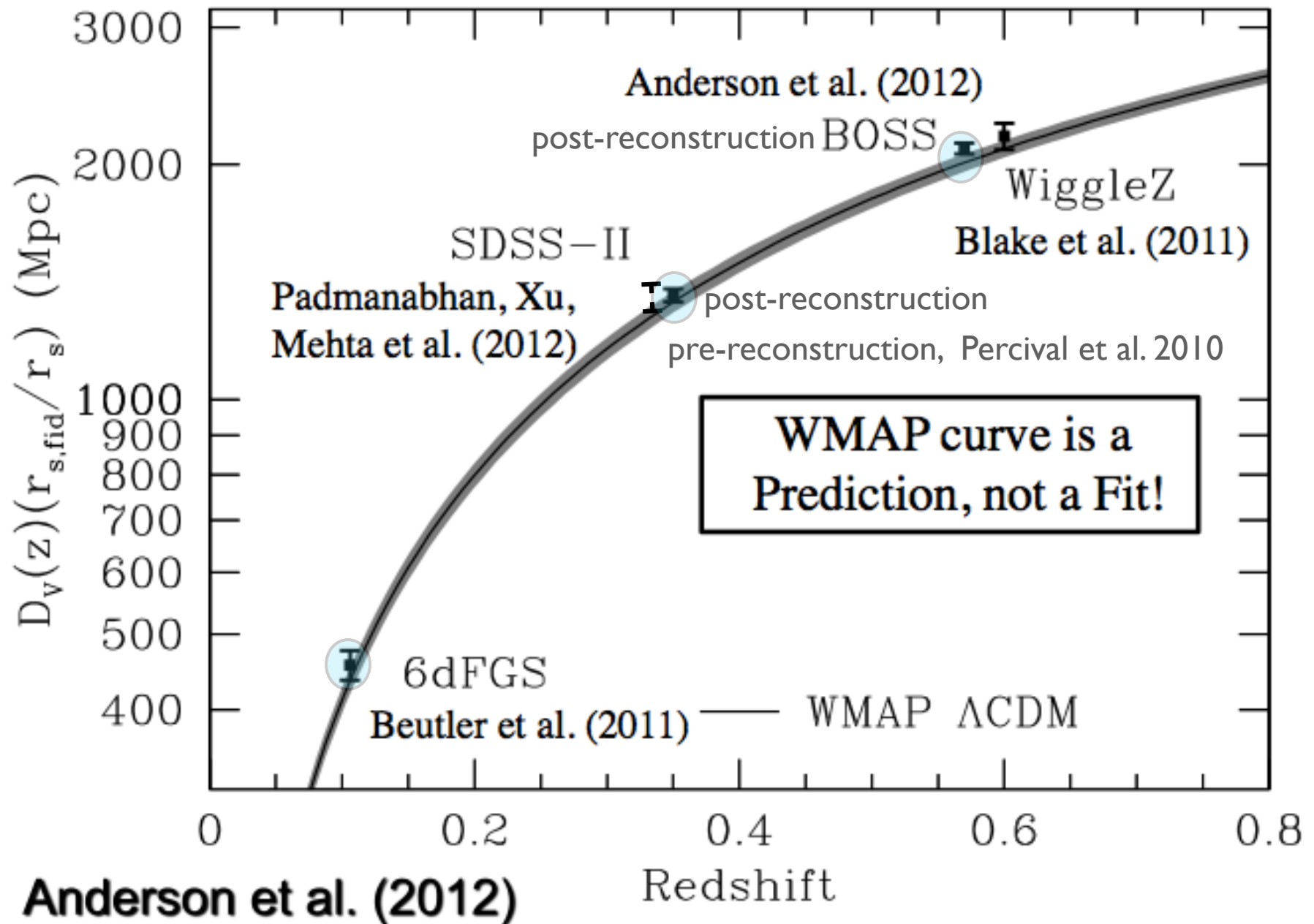


Anderson et al. 2012

BAO Hubble Diagram



BAO Hubble Diagram

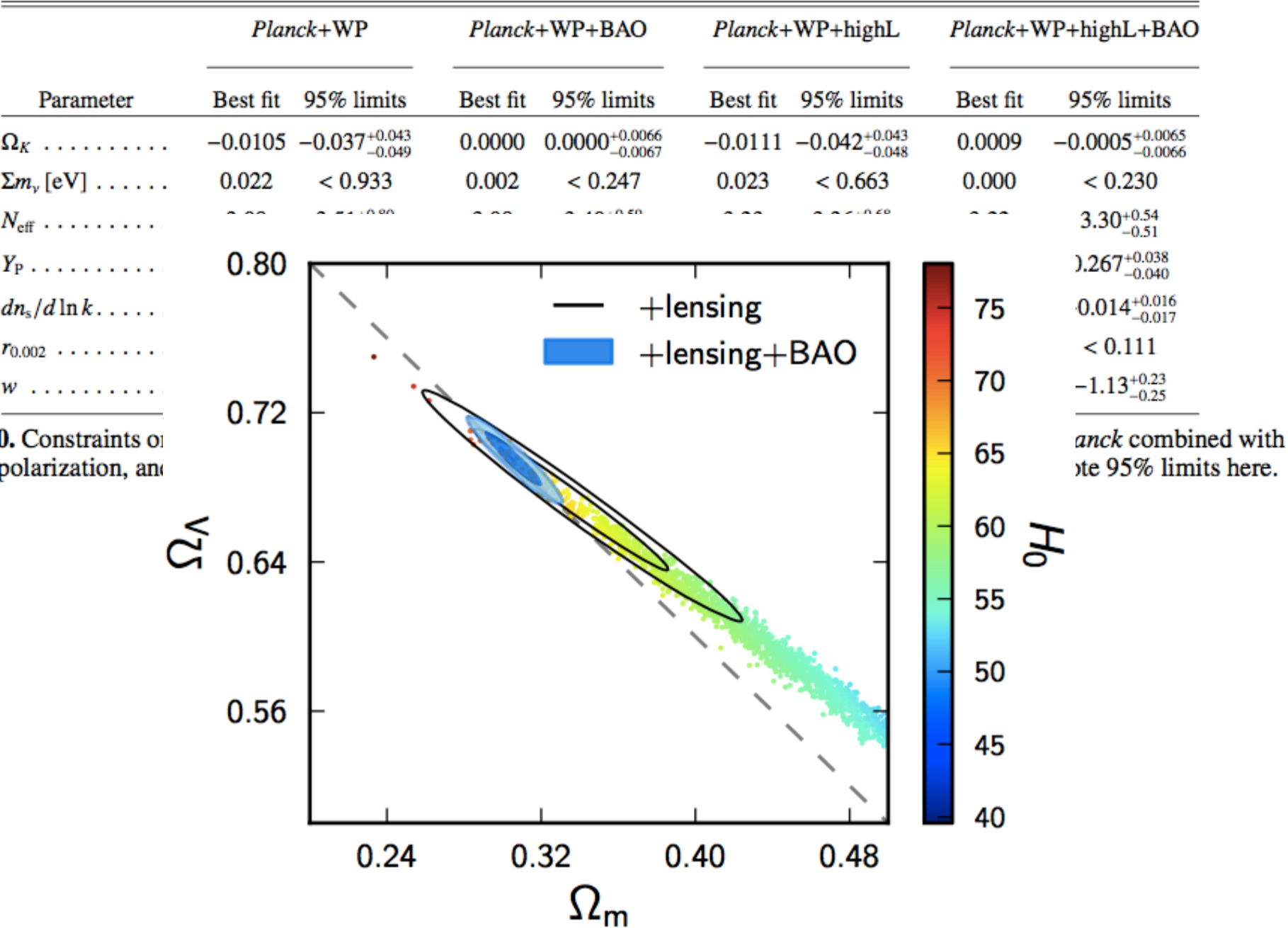


Planck Collaboration: Planck Cosmological Parameters

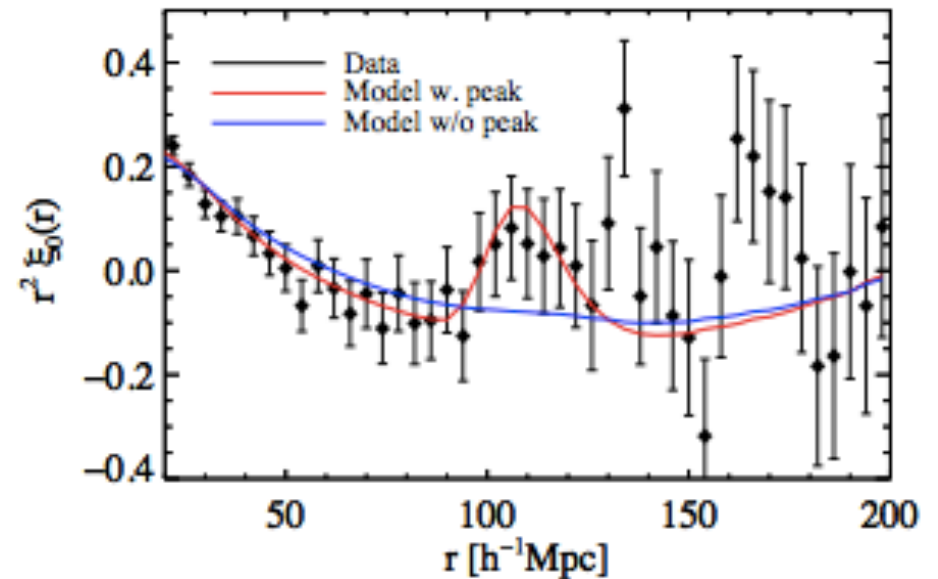
Parameter	<i>Planck</i> +WP		<i>Planck</i> +WP+BAO		<i>Planck</i> +WP+highL		<i>Planck</i> +WP+highL+BAO	
	Best fit	95% limits	Best fit	95% limits	Best fit	95% limits	Best fit	95% limits
Ω_K	-0.0105	$-0.037^{+0.043}_{-0.049}$	0.0000	$0.0000^{+0.0066}_{-0.0067}$	-0.0111	$-0.042^{+0.043}_{-0.048}$	0.0009	$-0.0005^{+0.0065}_{-0.0066}$
Σm_ν [eV]	0.022	< 0.933	0.002	< 0.247	0.023	< 0.663	0.000	< 0.230
N_{eff}	3.08	$3.51^{+0.80}_{-0.74}$	3.08	$3.40^{+0.59}_{-0.57}$	3.23	$3.36^{+0.68}_{-0.64}$	3.22	$3.30^{+0.54}_{-0.51}$
Y_{P}	0.2583	$0.283^{+0.045}_{-0.048}$	0.2736	$0.283^{+0.043}_{-0.045}$	0.2612	$0.266^{+0.040}_{-0.042}$	0.2615	$0.267^{+0.038}_{-0.040}$
$dn_s/d\ln k$	-0.0090	$-0.013^{+0.018}_{-0.018}$	-0.0102	$-0.013^{+0.018}_{-0.018}$	-0.0106	$-0.015^{+0.017}_{-0.017}$	-0.0103	$-0.014^{+0.016}_{-0.017}$
$r_{0.002}$	0.000	< 0.120	0.000	< 0.122	0.000	< 0.108	0.000	< 0.111
w	-1.20	$-1.49^{+0.65}_{-0.57}$	-1.076	$-1.13^{+0.24}_{-0.25}$	-1.20	$-1.51^{+0.62}_{-0.53}$	-1.109	$-1.13^{+0.23}_{-0.25}$

Table 10. Constraints on one-parameter extensions to the base Λ CDM model. Data combinations all include *Planck* combined with *WMAP* polarization, and results are shown for combinations with high- ℓ CMB data and BAO. Note that we quote 95% limits here.

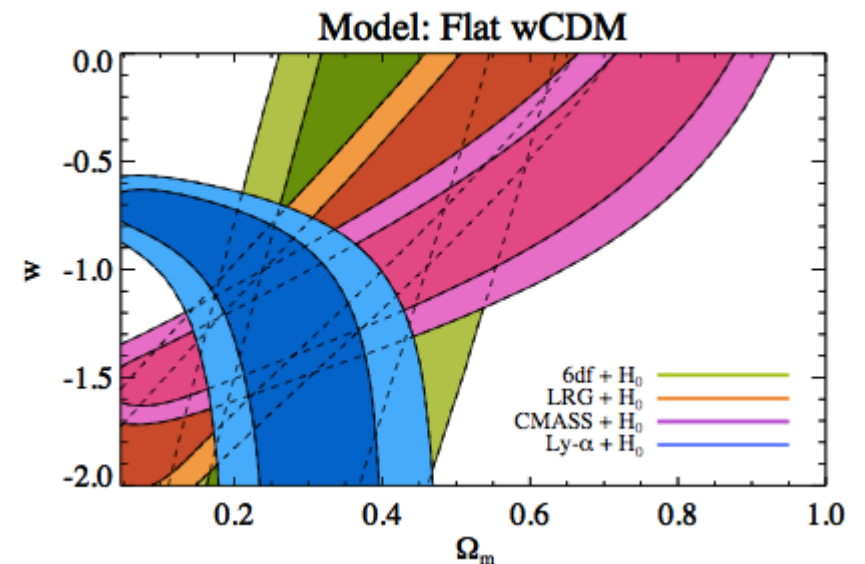
highL = ACT (148,218)+ SPT (95,150,220)
 WP= WMAP polarization
 BAO= 6dF+SDSS(R)+BOSS



- See Anze Slosar's talk!
- BAO in Lyman-alpha forest
- First BAO analysis at $z > 2$



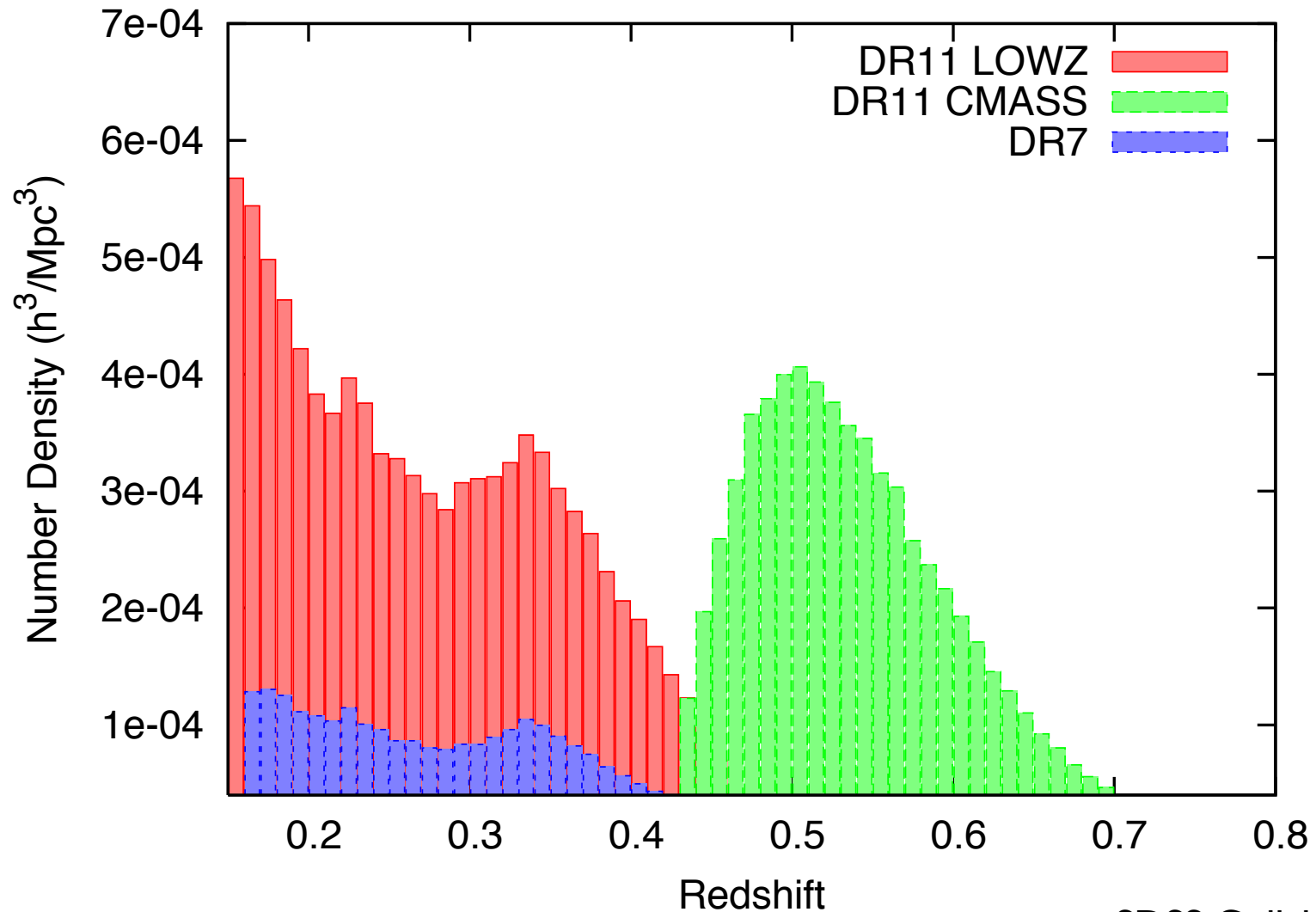
Detected BAO using Lyman alpha forest at $z=2.3$



with NO CMB at all!!

SDSS III

What about after DR9 ?



SDSS III

Data Release 9-12 Spectroscopy

~30K high-z galaxy spectra

~60K

1 million

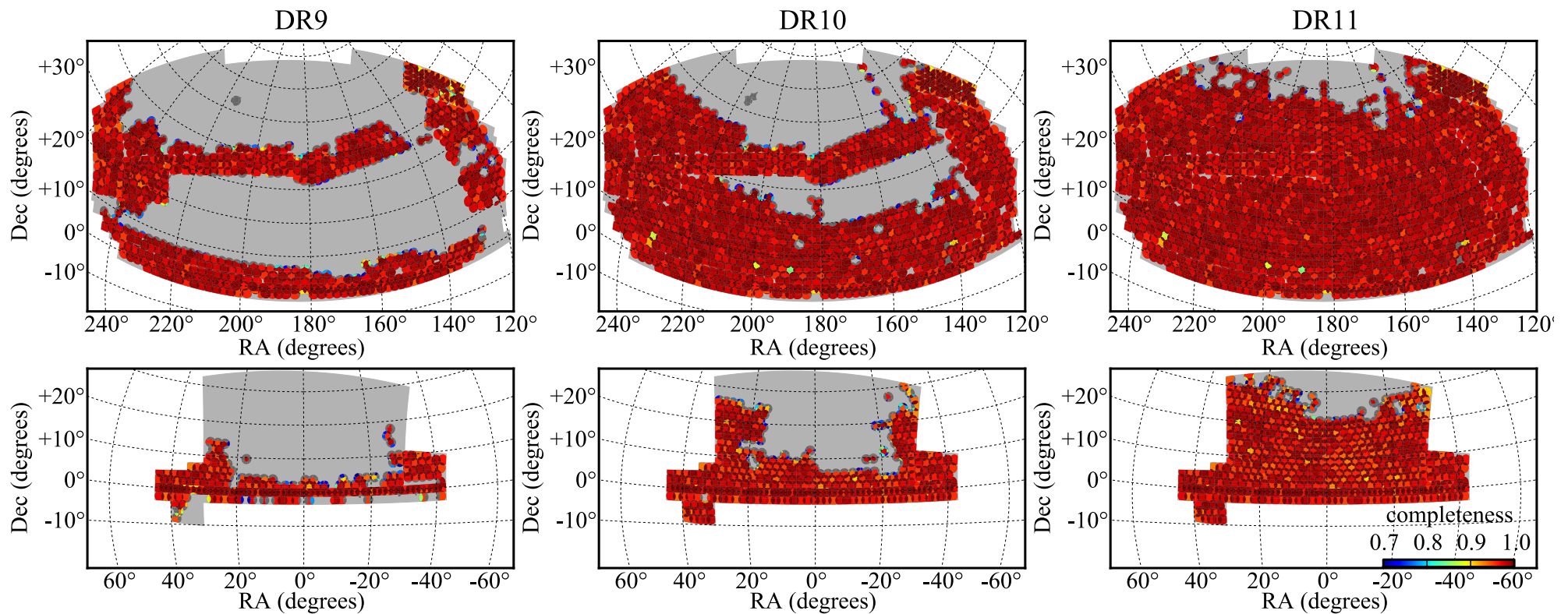
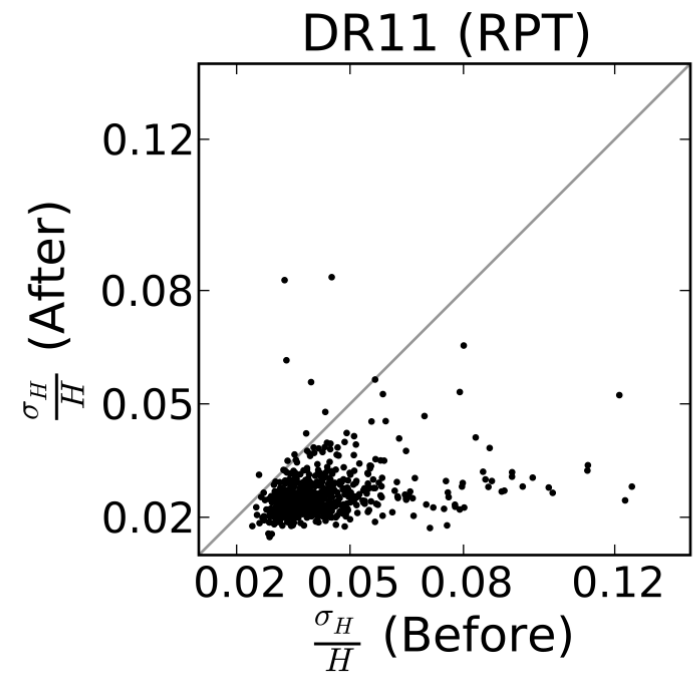
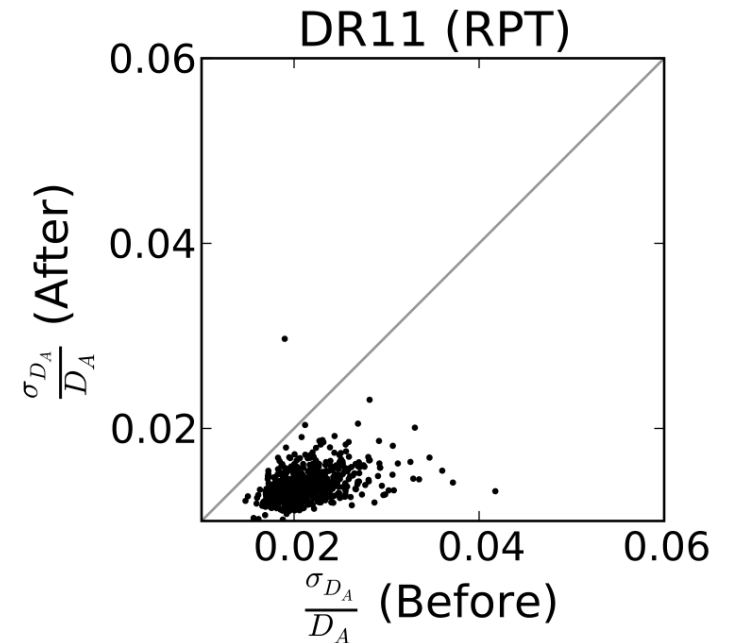


Figure 1. Evolution of the BOSS sky coverage from DR9 to DR11.

- In DR9, reconstruction didn't improve our signal-to-noise of BAO. This is probably due to poor survey window (too much area near a boundary).
- DR11 has a filled geometry. Reconstruction improves mocks dramatically.
- Average Constraints (mocks) on D_A goes from 2.1% to 1.5%.
- Average Constraints (mocks) on $H(z)$ goes from 4.4% to 2.7%
- Expects BAO >8 sigma
- Stay Tuned!



Magana-Vargas, Ho, Xu et al. (in prep)
SDSS Collaboration (in prep)

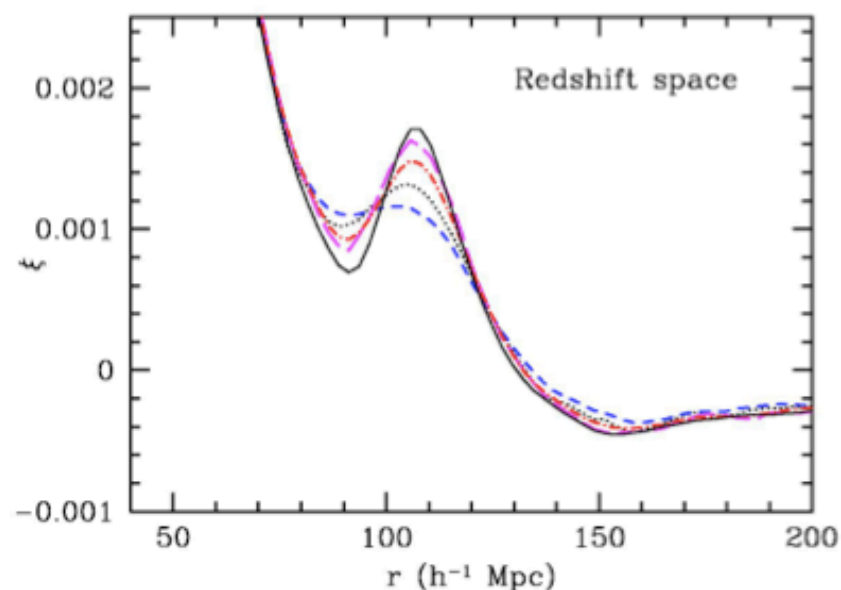
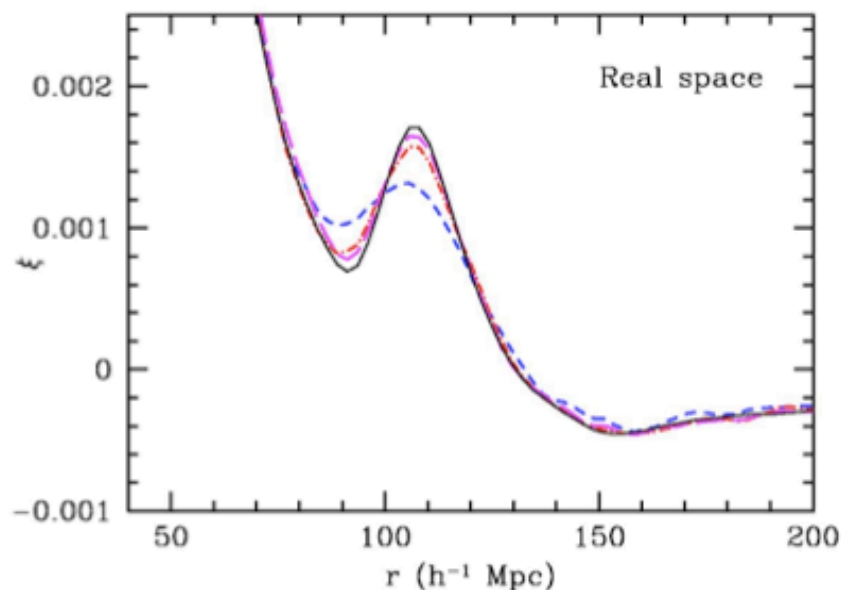
Outline

- SDSS III- BOSS :
 - Mini-summary and looking forward
- BAO Reconstruction

You can reconstruct to significantly different large scale structure, but still get the same/similar BAO post-reconstruction!

Bringing in Reconstruction ... (Going back in time)

- Most of the non-linear degradation is due to large scale flows. These are produced by the same large scale structure that we are measuring for the BAO signature.
- Map of galaxies tells us where the mass is that sources the gravitational forces that created the bulk flows
- Can run this backwards and undo most non-linearity
- Restore the statistical precision available per unit volume



Eisenstein, Seo, Sirko, Spergel, 2007

Reconstruction: general idea

- Smooth the density field to filter out high k modes, which are difficult to model.
- Compute the negative Zel'dovich displacement, $\mathbf{s}$, from the smoothed density field: $\mathbf{s}(\mathbf{k}) = -i(\mathbf{k}/k^2)\delta(\mathbf{k})\mathcal{S}(k)$, where $\mathcal{S}$ is the smoothing kernel (see below).
- Shift the original particles by $\mathbf{s}$ and compute the “displaced” density field, δ_d .
- Shift an initially spatially uniform distribution of particles by $\mathbf{s}$ to form the “shifted” density field, δ_s .
- The reconstructed density field is defined as $\delta_r \equiv \delta_d - \delta_s$ with power spectrum $P_r(k) \propto \langle |\delta_r^2| \rangle$.

Seo, Eisenstein, Sirki & Spergel 2008

Noh, White & Padmanabhan 2009

Reconstruction: general idea

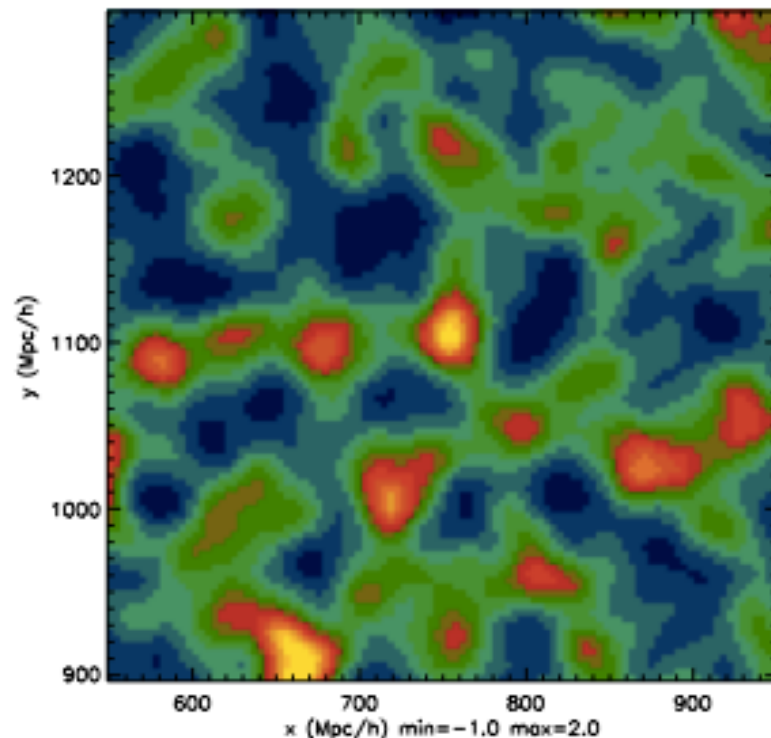
- Smooth the density field to filter out high k modes, which are difficult to model.

- Compute $\delta_s(\mathbf{k})$ and $s(\mathbf{k})$ kernel

- Shift places

- Shift particles

- The δ_s will



displacement field: smoothing

the “dis-

n of particles.

$\delta_r \equiv \delta_d -$

Seo, Eisenstein, Sirki & Spergel 2008

Noh, White & Padmanabhan 2009

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$$\mathcal{S}(k) = e^{-(kR)^2/2}.$$

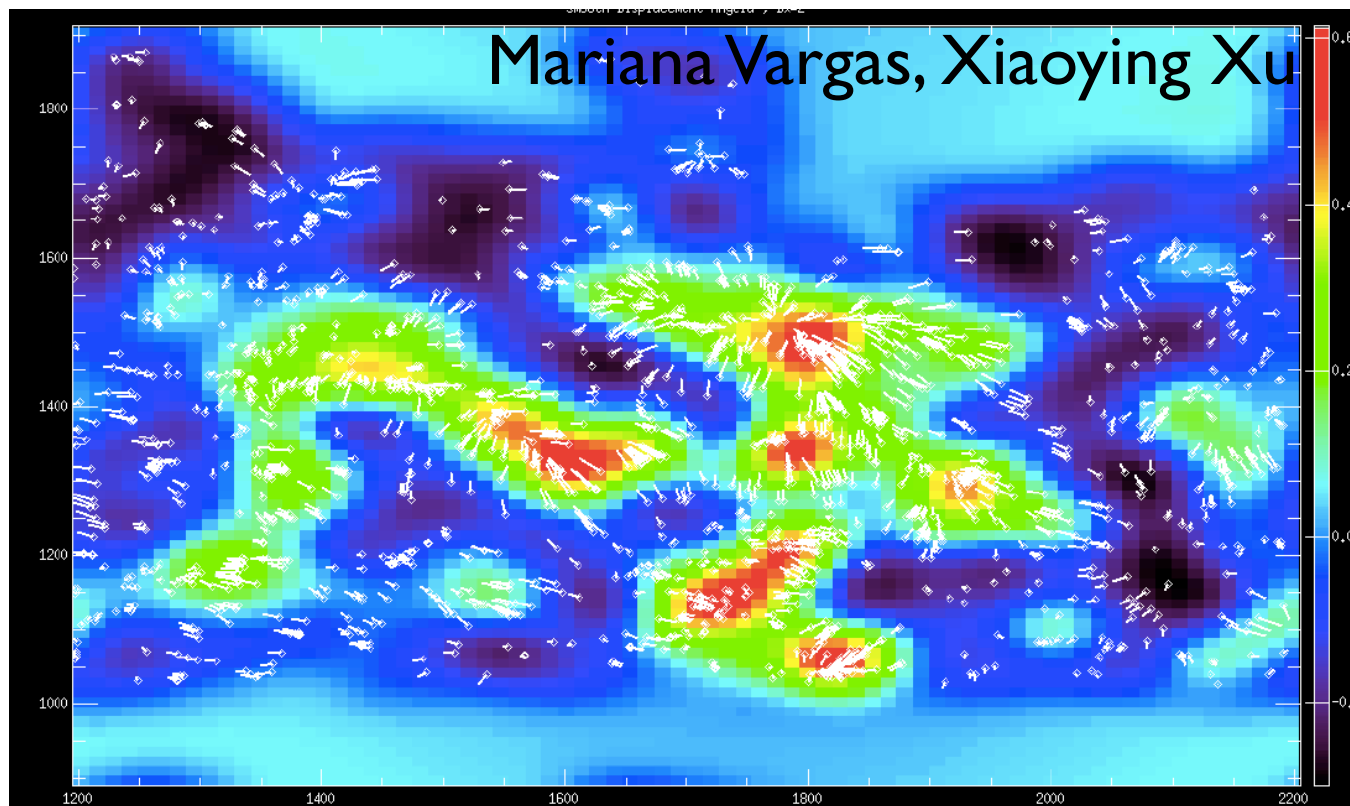
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contours: density field

white lines: displacement field

Seo, Eisenstein, Sirki & Spergel 2008

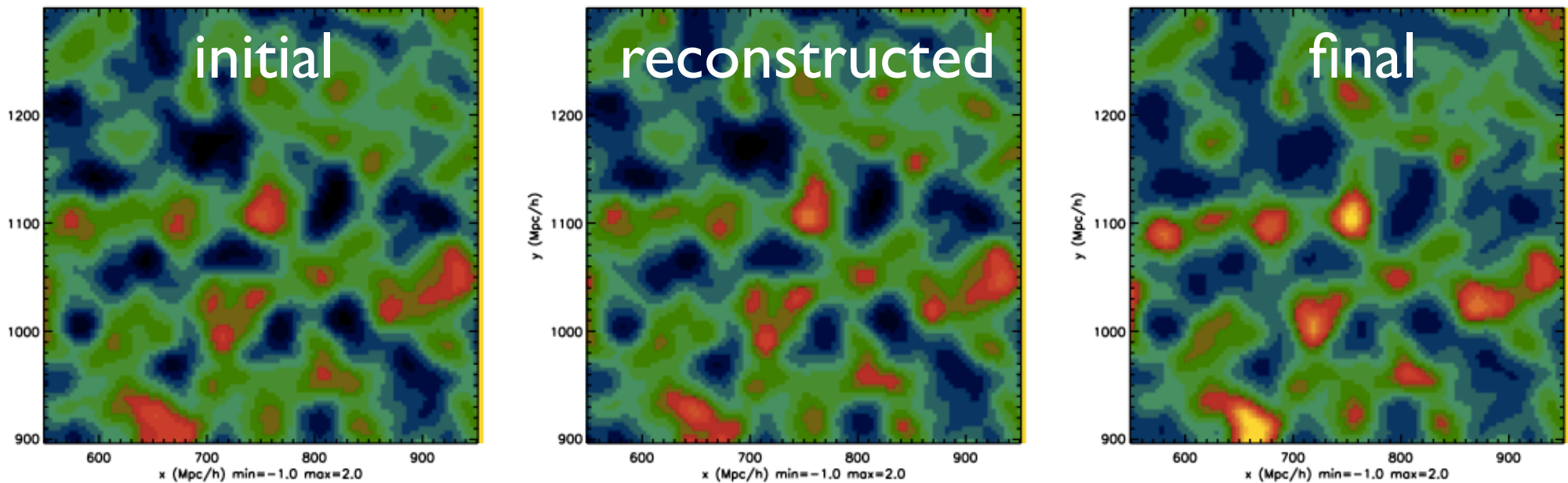
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Seo, Eisenstein, Sirki & Spergel 2008

Noh, White & Padmanabhan 2009



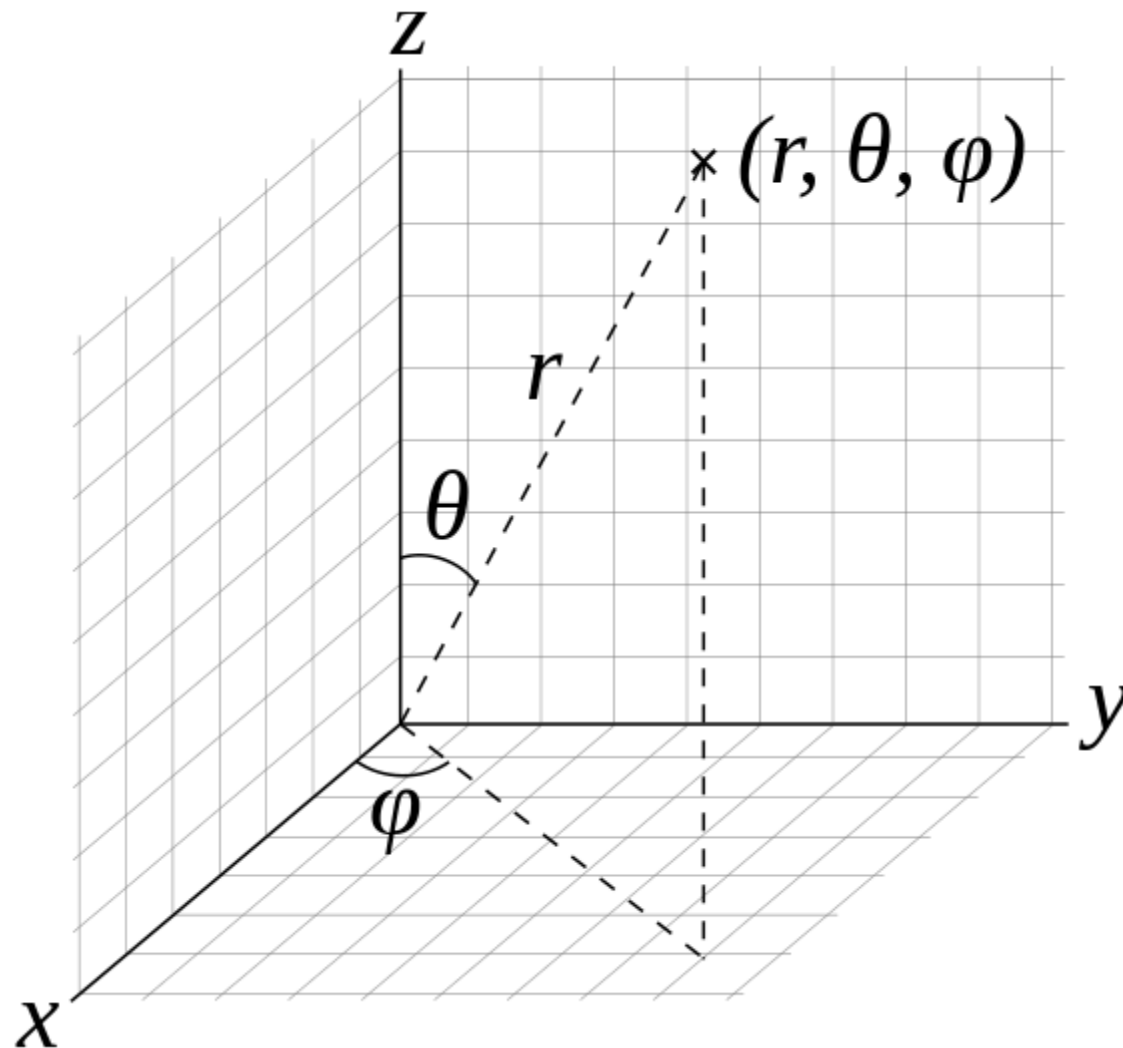
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Seo, Eisenstein, Sirki & Spergel 2008

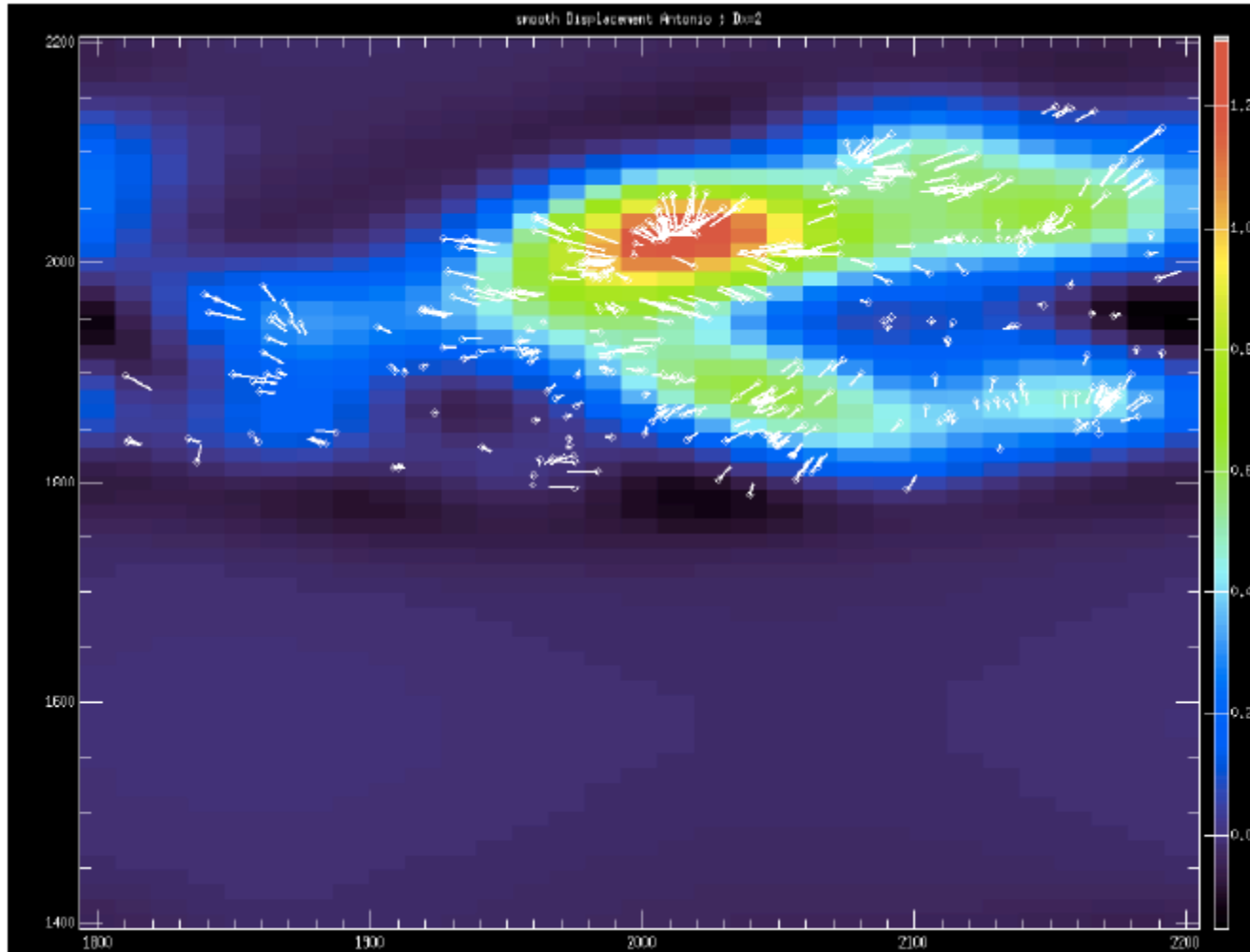
Noh, White & Padmanabhan 2009

As simple as the algorithm sounds like...
three different codes following similar methodology generate
different displacement field given the same input.

Just reminding us the coordinate system



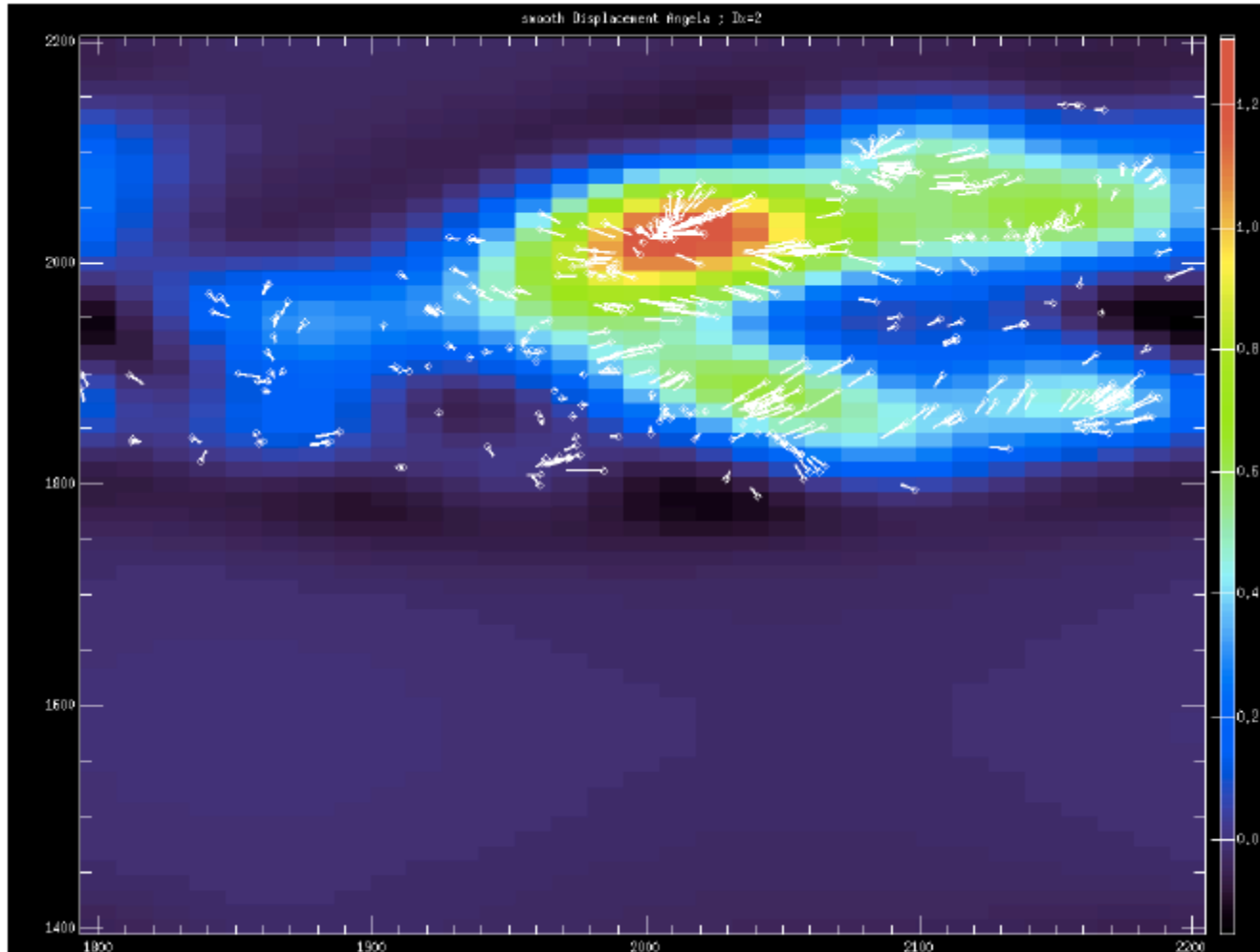
Displacement in r



Method I

Magana-Vargas, SH , Xu + BOSS Clustering WG (in prep)

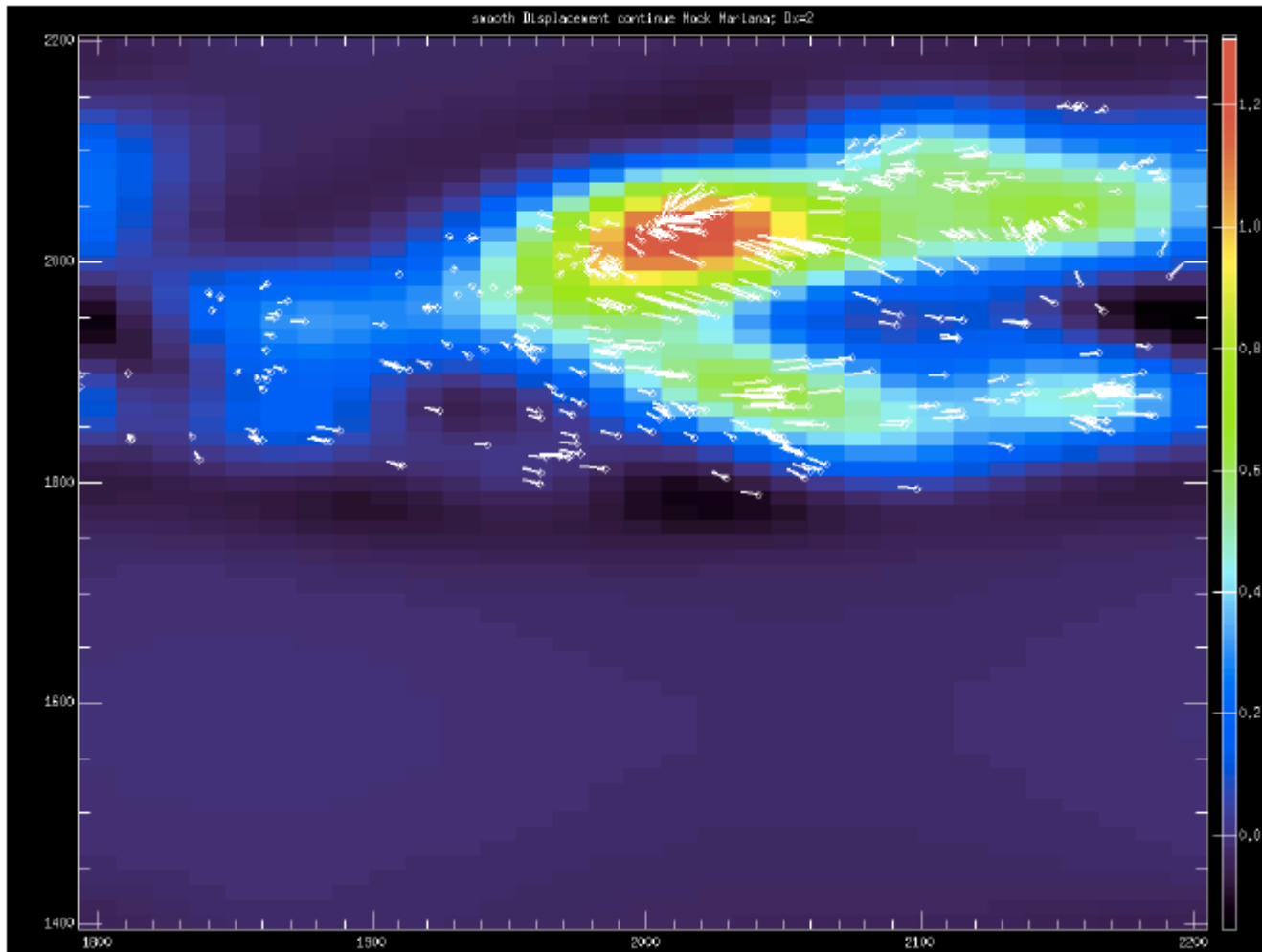
Displacement in r



Method 2

Magana-Vargas, SH , Xu + BOSS Clustering WG (in prep)

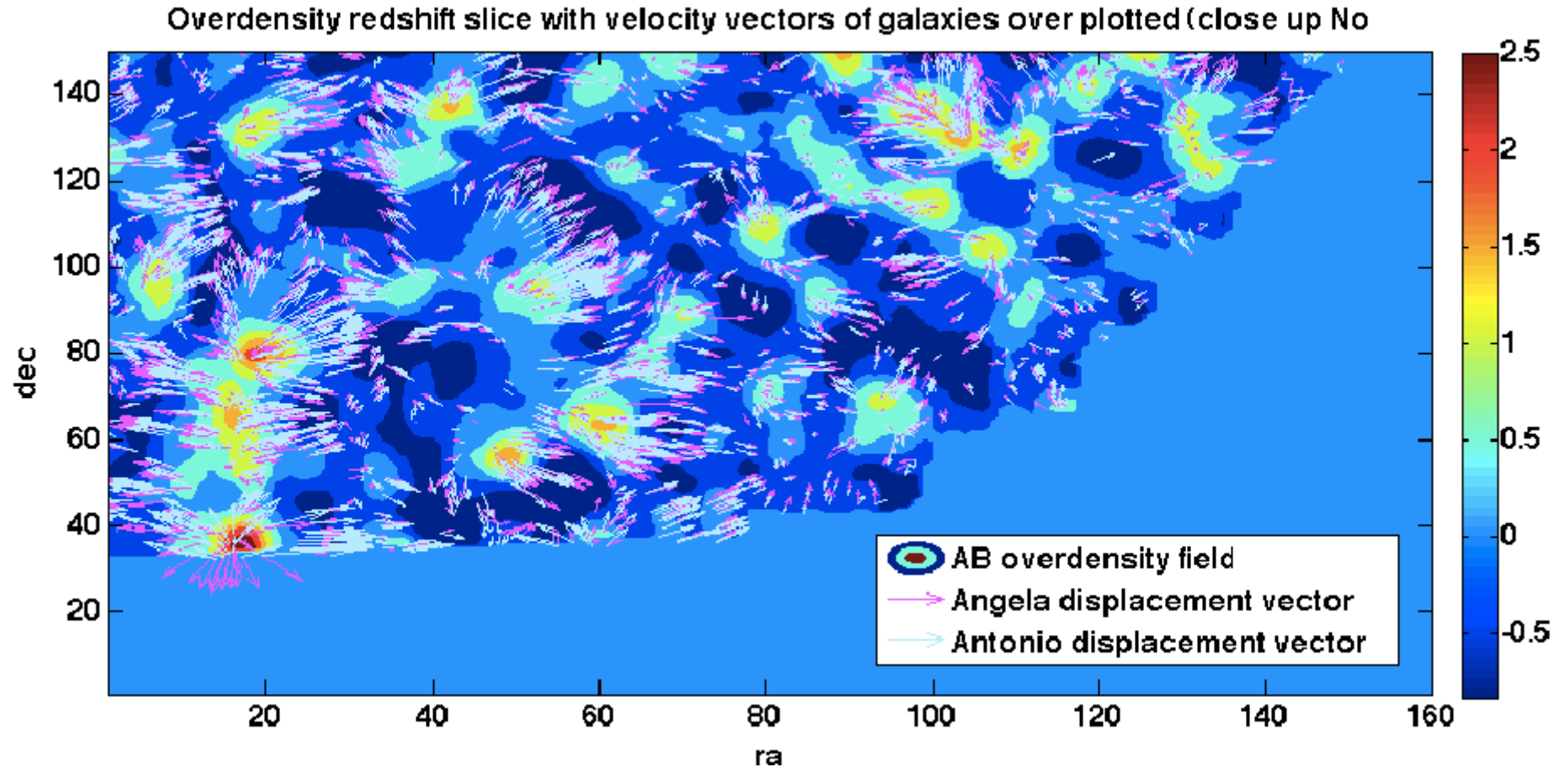
Displacement in r



Method 3

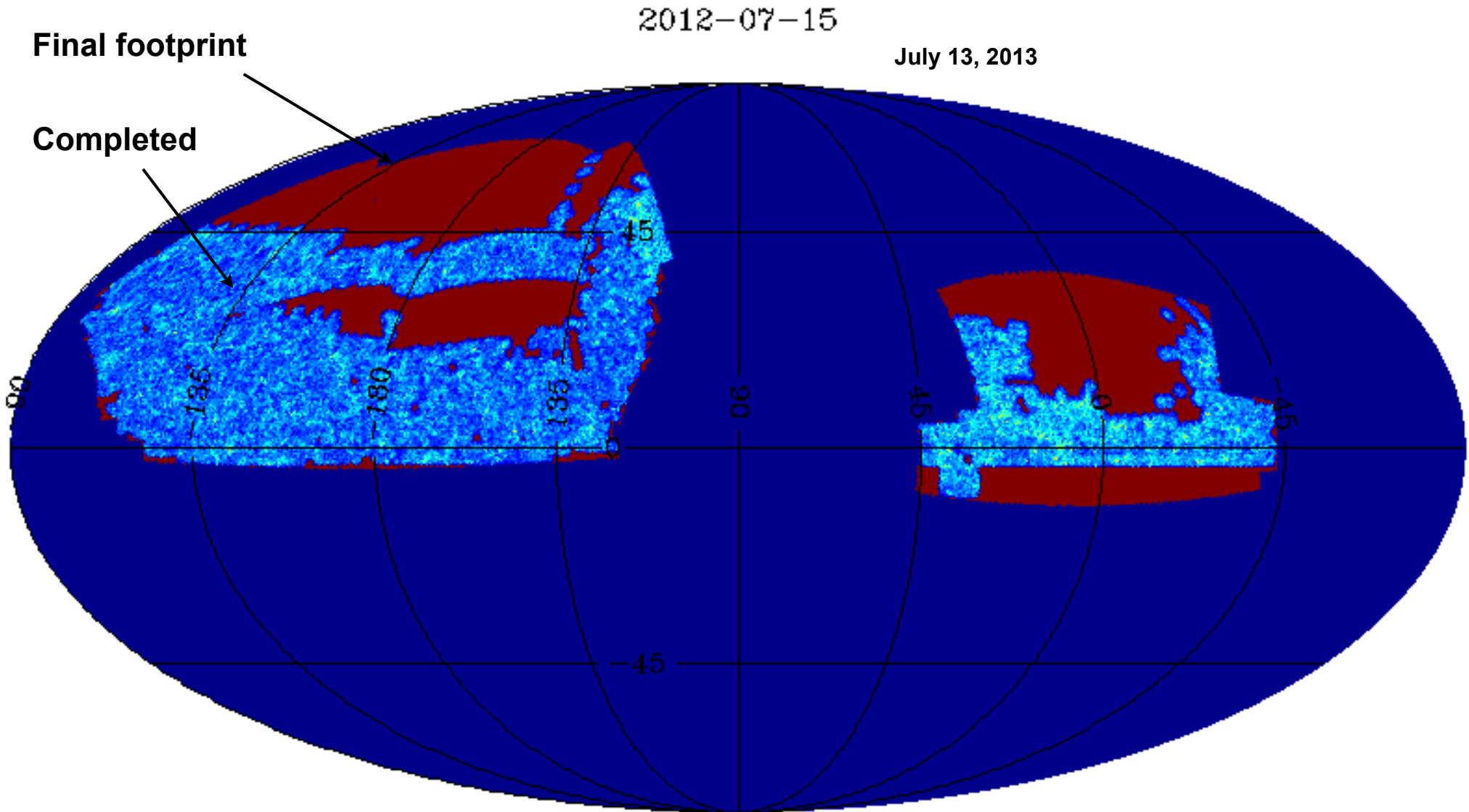
Magana-Vargas, SH , Xu + BOSS Clustering WG (in prep)

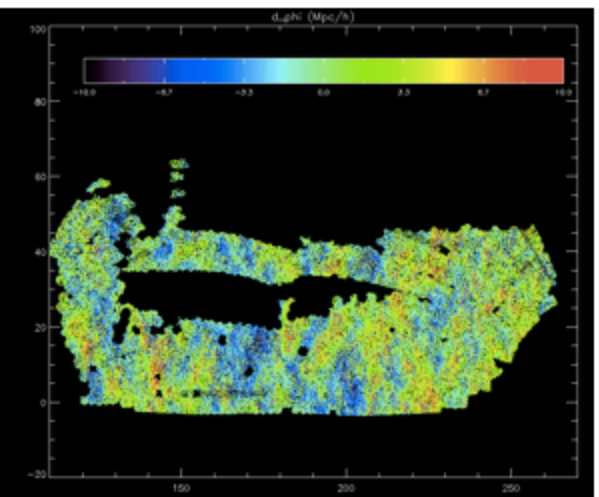
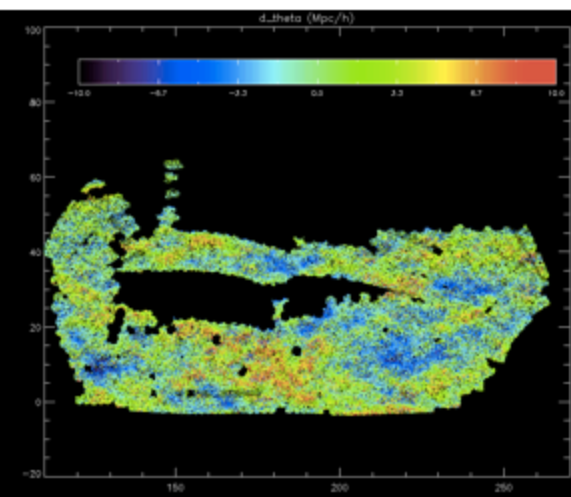
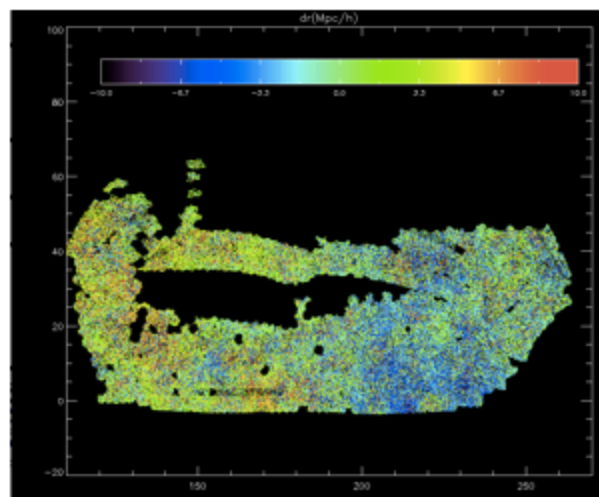
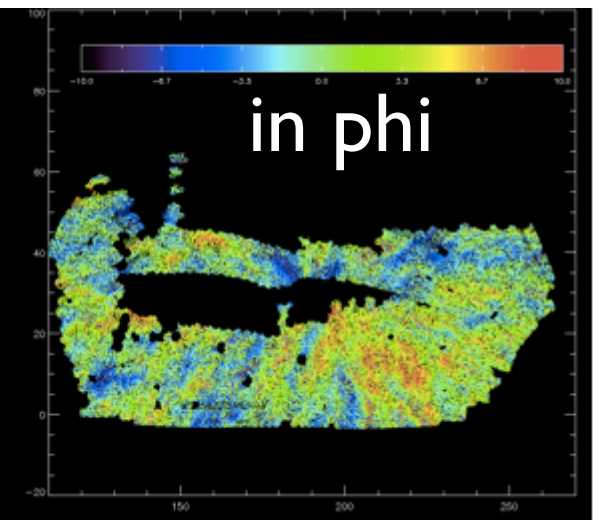
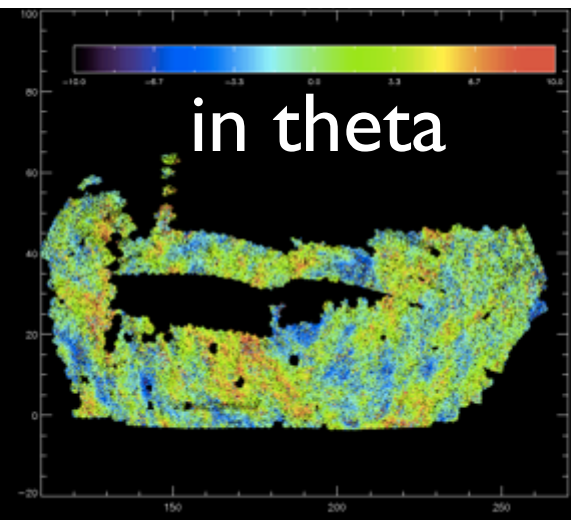
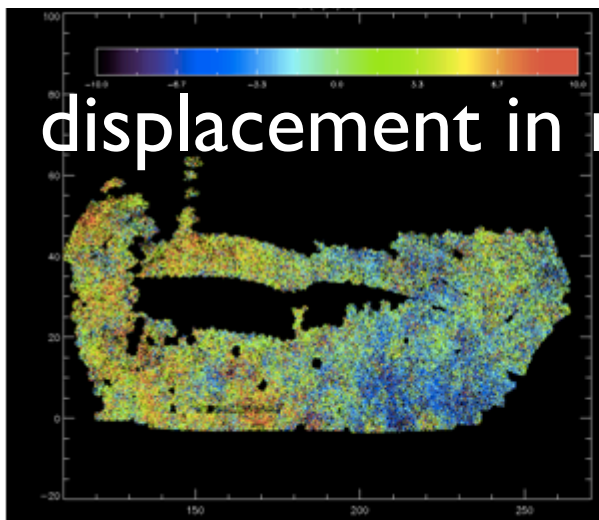
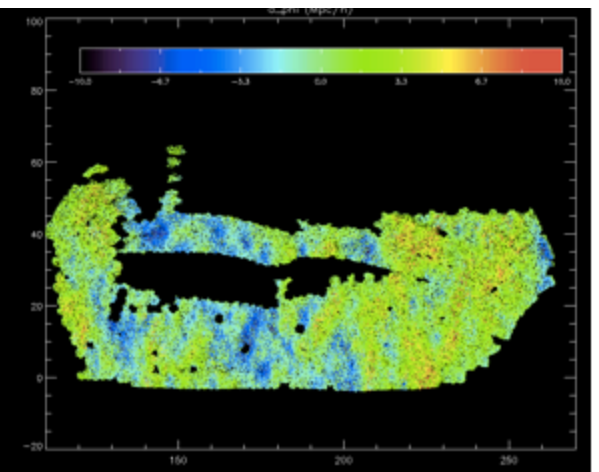
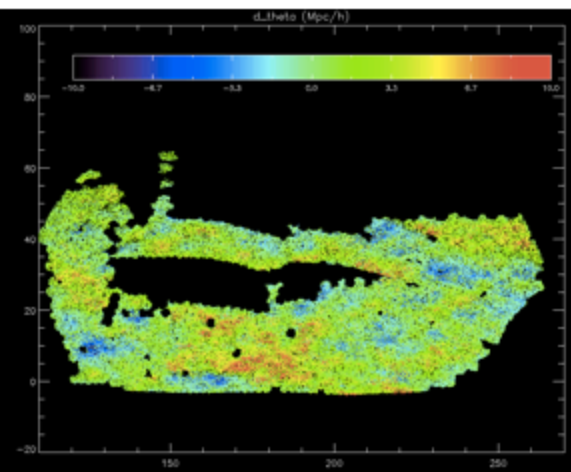
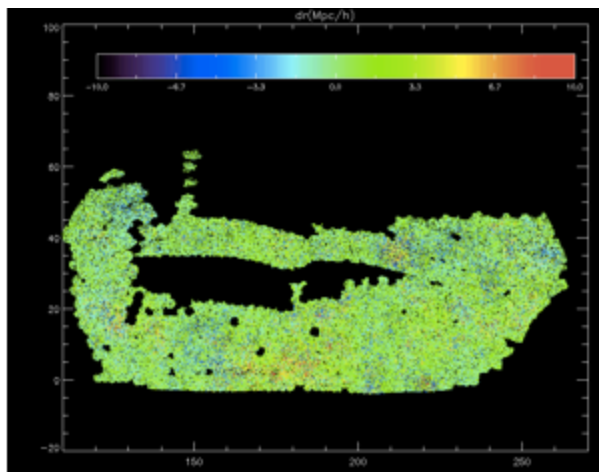
Effects of different reconstruction method choices



Courtesy plots from Angela Burden

How about applying 3 methods to the same BOSS dataset?





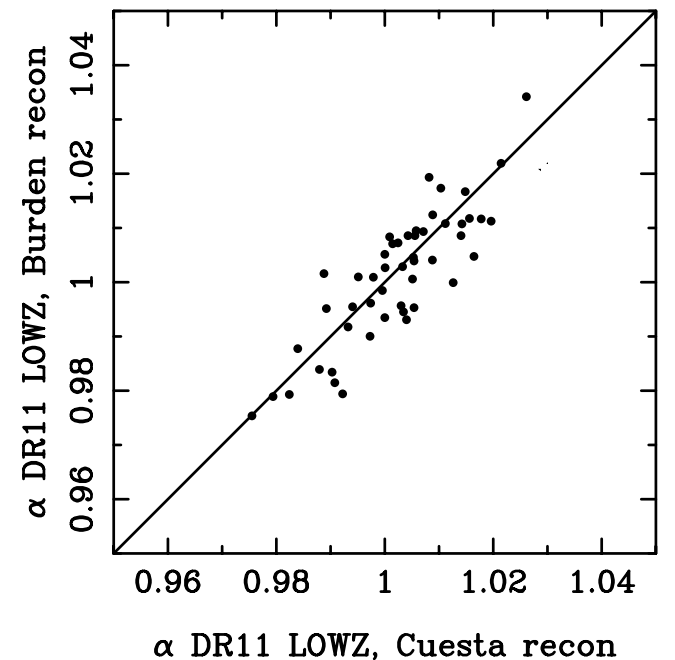
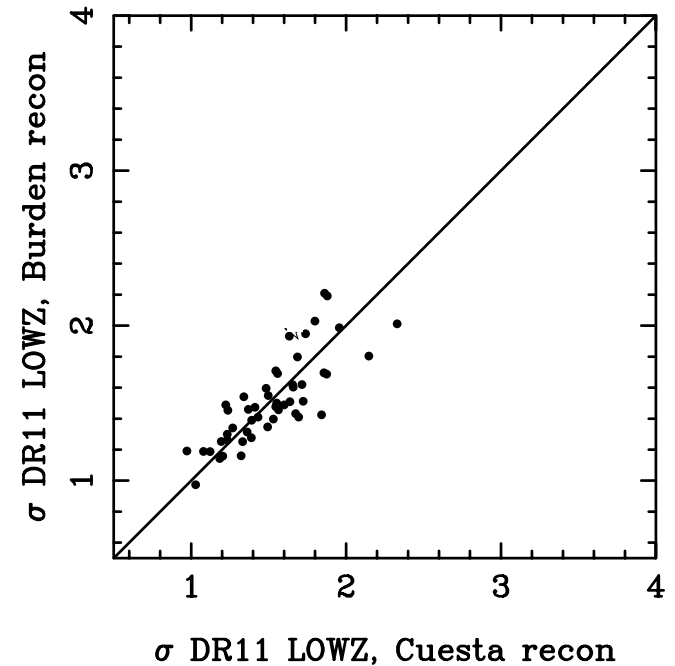
Test results from two methods on 50 mocks

$$\bar{\alpha} = 1.00289 \pm 0.00161$$

$$\bar{\alpha} = 1.00169 \pm 0.00172$$

- Conclusions?
- All methods are relatively unbiased.
- The scatter plots show that they correlate with each other very well.
- But there is a scatter between the methods

Courtesy plots from Will Percival



Outline

- SDSS III- BOSS :
 - Mini-summary and looking forward
 - BAO isotropic and anisotropic fitting
 - BAO Reconstruction

You can reconstruct to significantly different large scale structure, but still get the same/similar BAO post-reconstruction!

Now we need to understand and include the systematic errors we are introducing when using reconstruction.

Outline

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- BAO Reconstruction
- Ultra-fast simulations of the Universe?
- BAO in cross-correlations?

Aim: Efficiently make lots of mocks relatively accurately. BOSS galaxy clustering working group used ~600 mocks using PTHalos method.
Can we do better?

Aim: Efficiently make lots of mocks relatively accurately. BOSS galaxy clustering working group used ~600 mocks using PTHalos method.

Can we do better?

New Direction: Use a non-parametric Machine learning algorithm that does distribution-to-distribution regression to learn the non-linear evolution of the Universe.

What the heck is Machine Learning?

- Machine Learning algorithms **learns trends from the data itself**, it does not impose pre-assumed models on the data.
- The advantage of ML is that it is fully non-parametric:
 - The only assumption necessary is that some relationship **does** exist between halo properties (features) and the number of galaxies that will reside in it and that this relationship is continuous.

Cool Examples of ML applications (Courtesy Slide from Kayvon Fatahalian)

[Shrivastava 2011]

“Find images that are similar to a query image (even if not similar in individual pixel values).”

Query image
(snowy day)



Matches



[Doersch 2012]

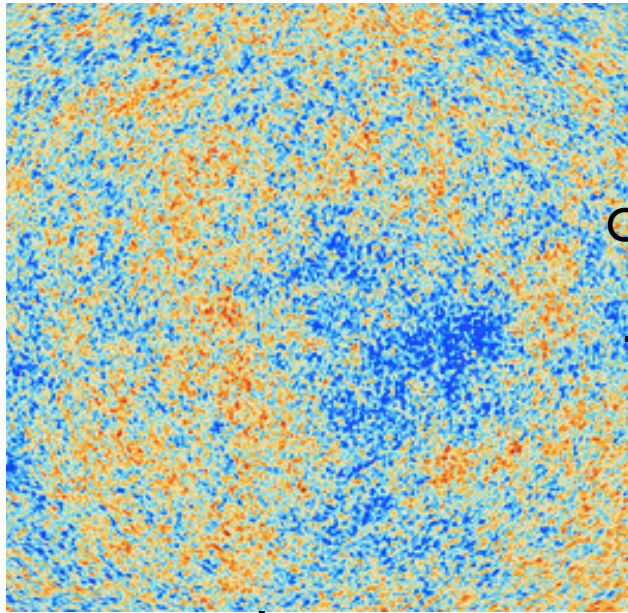
“Find meaningful visual elements that are unique to Paris”



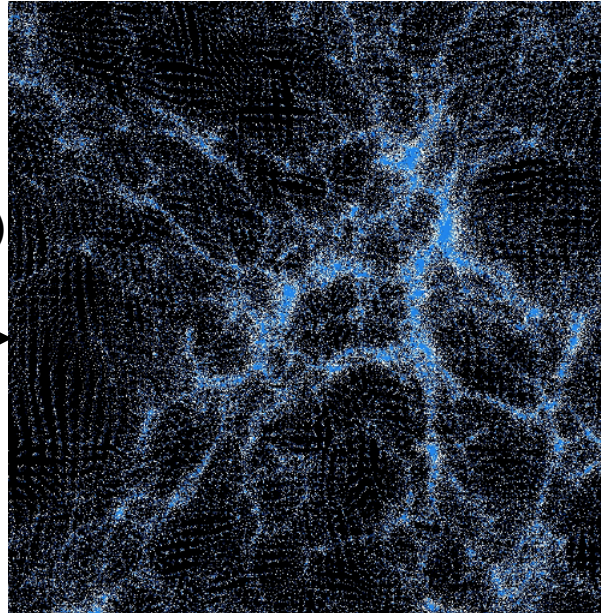
Initial condition of the Universe

Approximate Simulation of the
Universe, P0

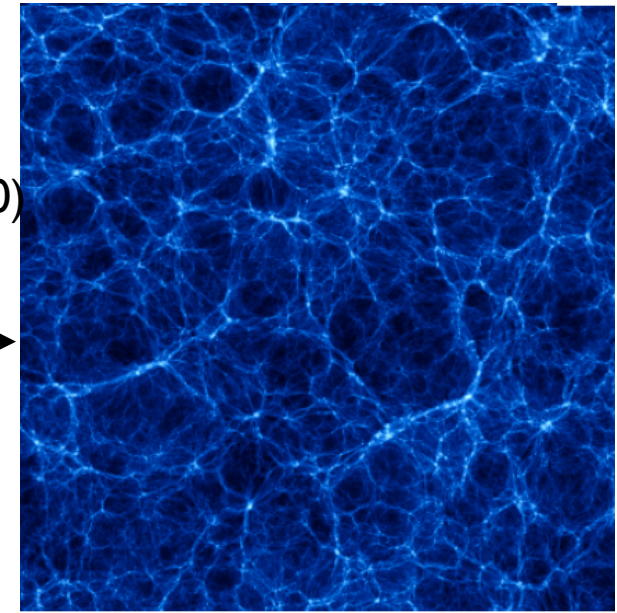
Full Simulation of the Universe,
Q0



$O(10)$
sec



$O(10)$
sec



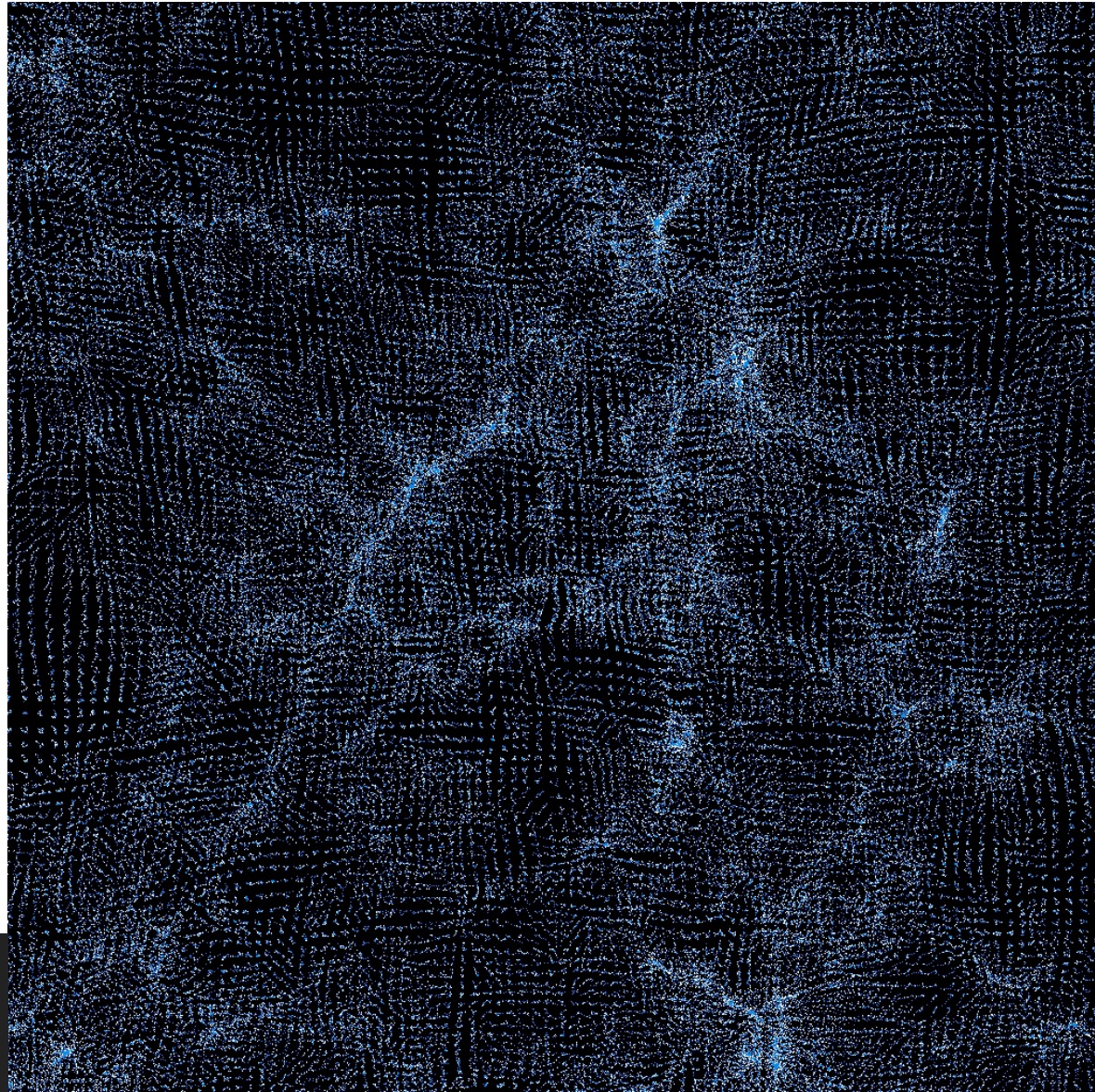
Approximate
evolution of Universe

Distribution-to-
distribution regression

$O(100,000)$ CPU seconds

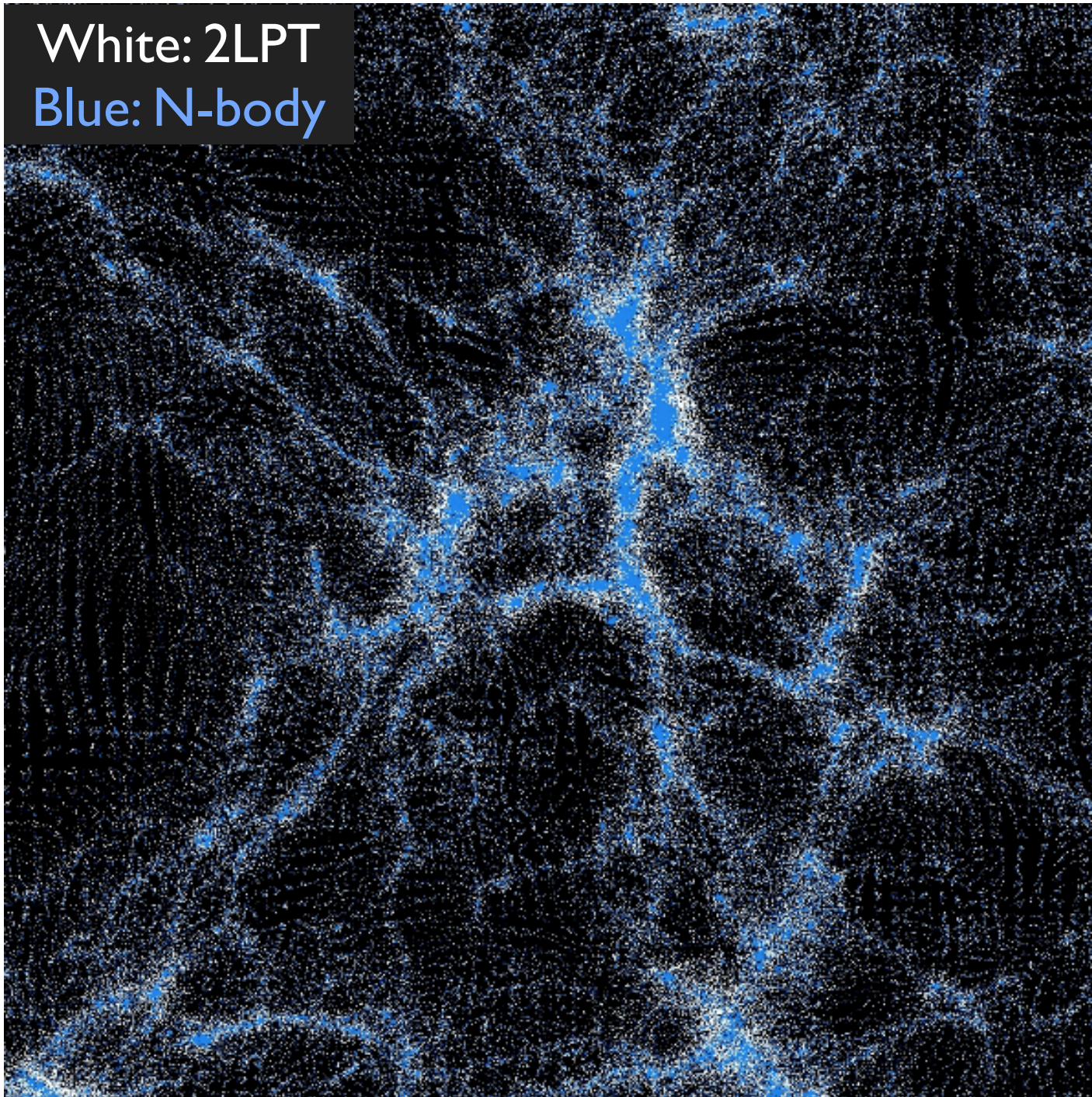
Exact evolution of the Universe

How does approximate (2LPT) field compares to N-body

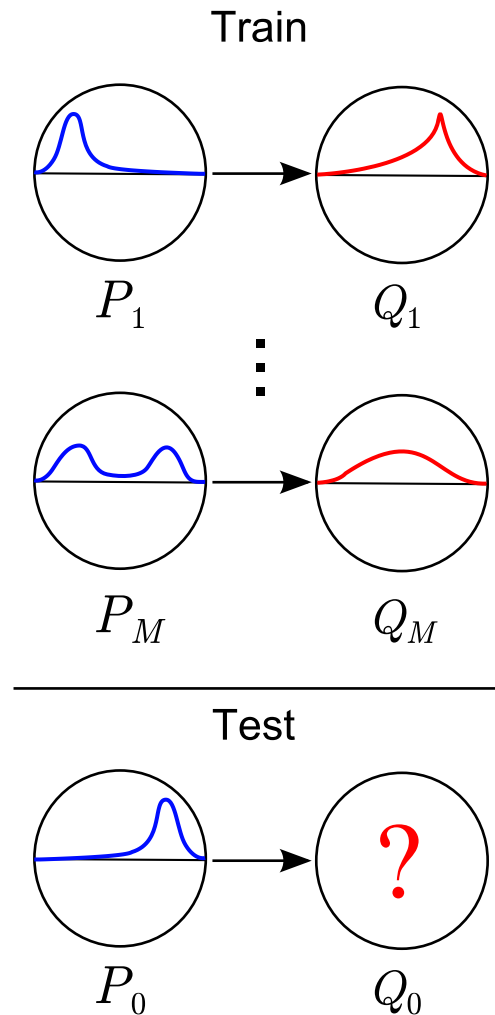


White: 2LPT
Blue: N-body

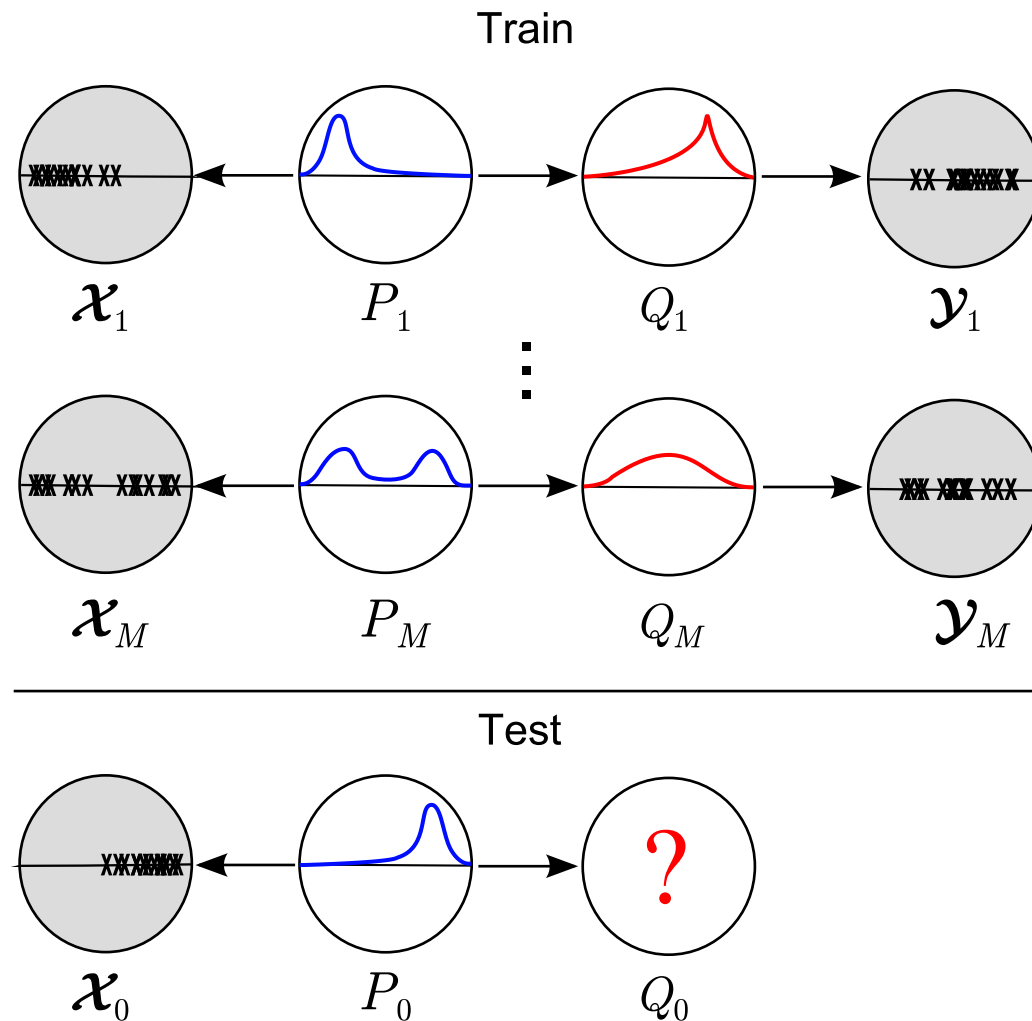
White: 2LPT
Blue: N-body



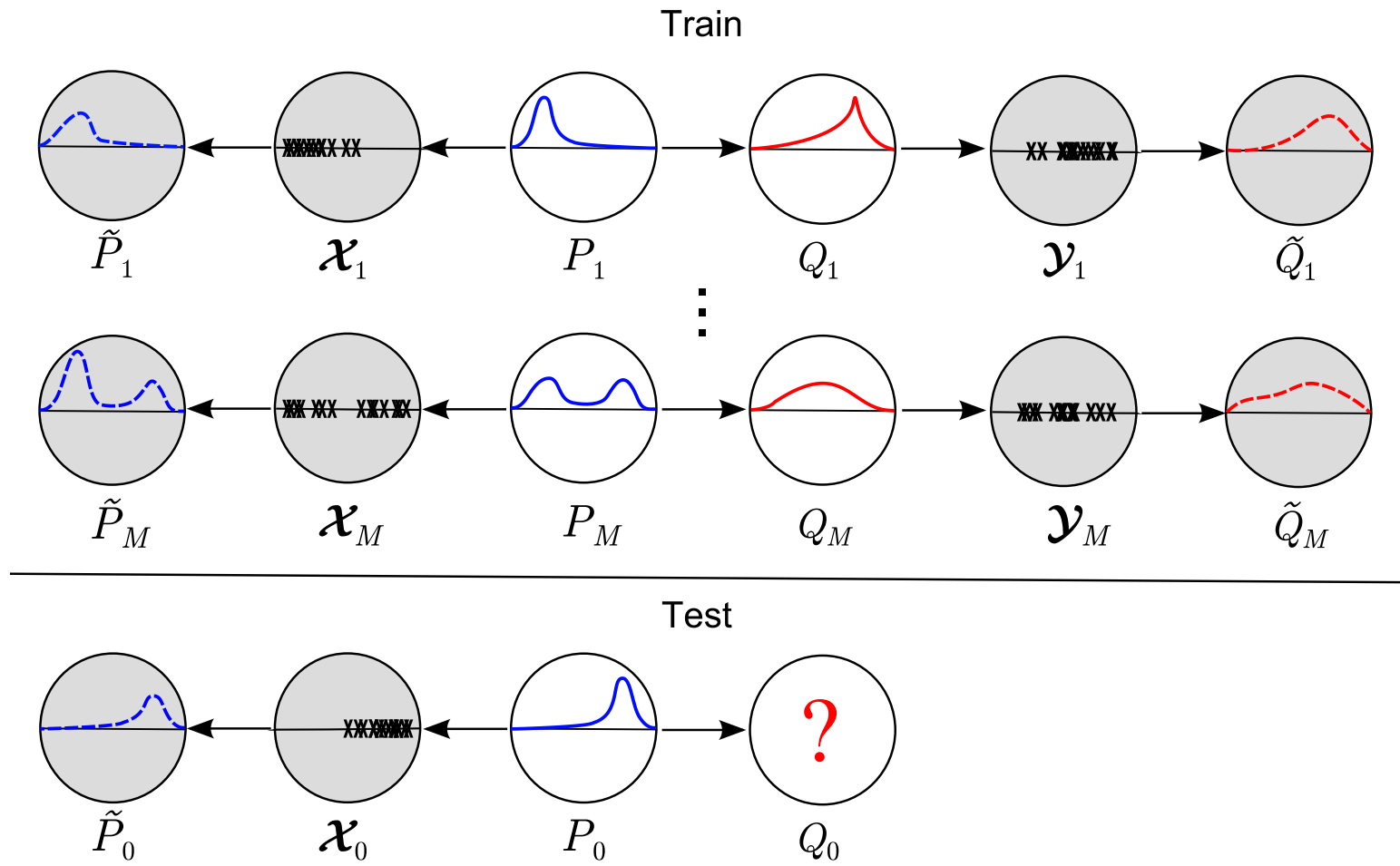
(2LPT) Distribution to (N-body) Distribution Machine Learning algorithms



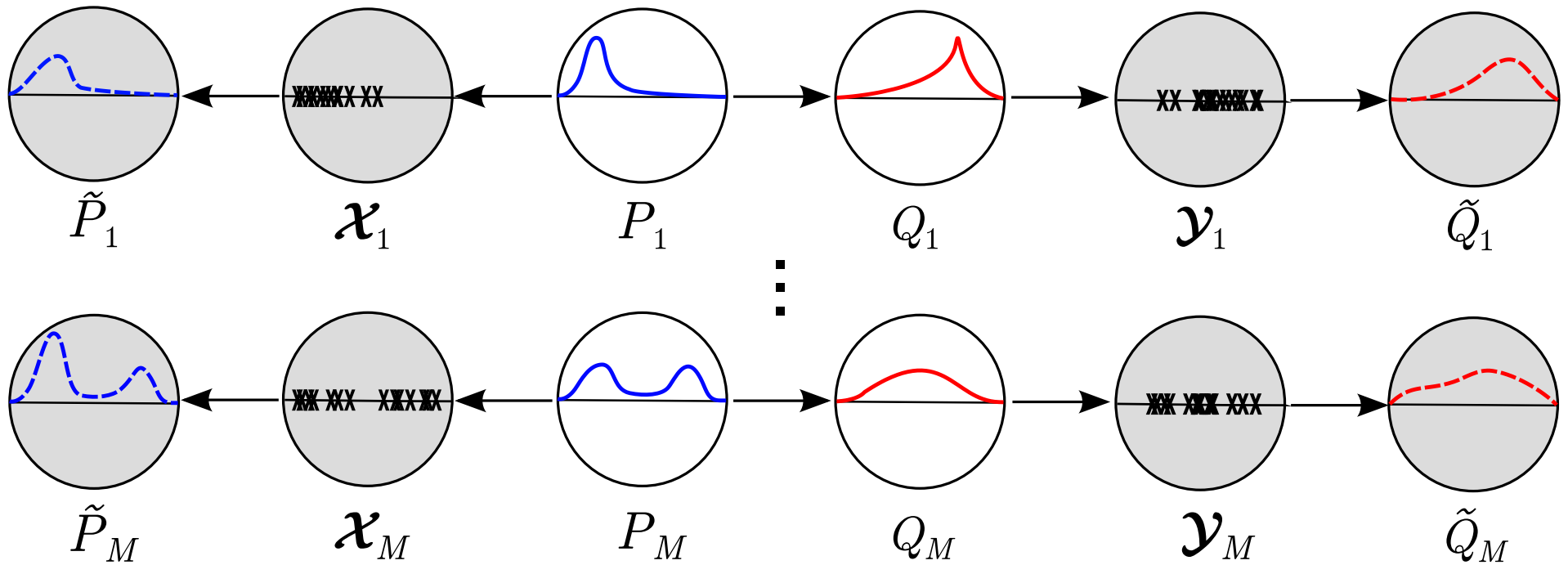
(2LPT) Distribution to (N-body) Distribution Machine Learning algorithms



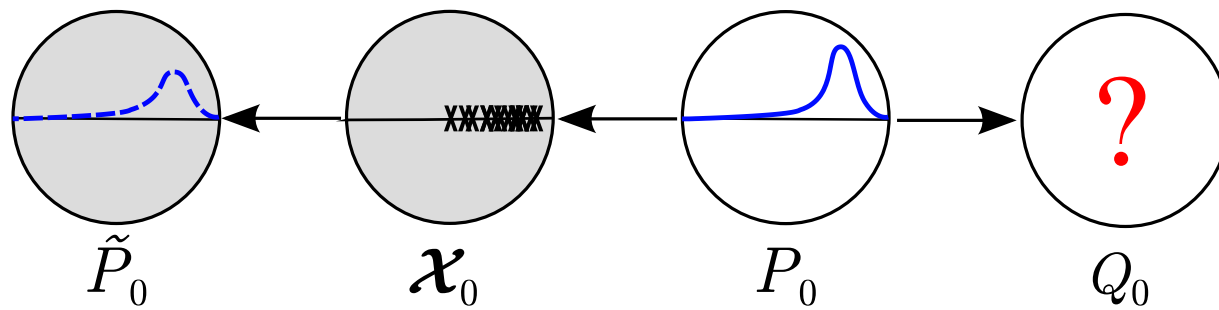
(2LPT) Distribution to (N-body) Distribution Machine Learning algorithms



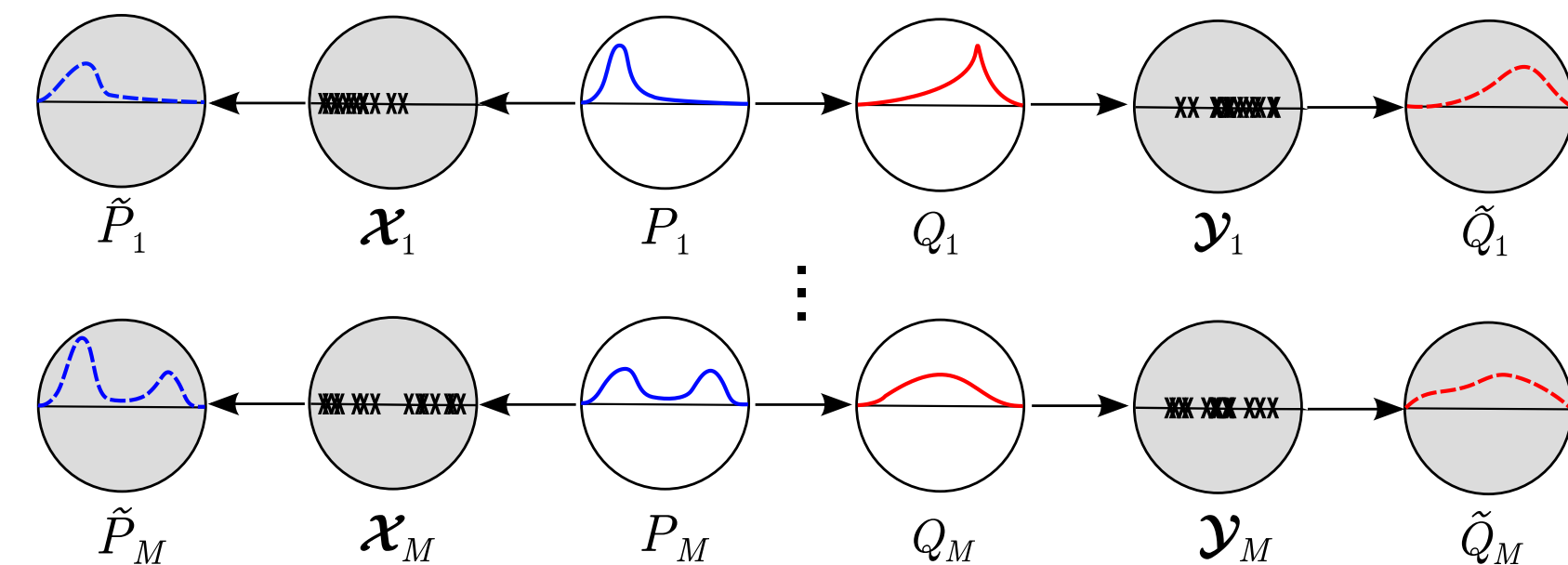
Train



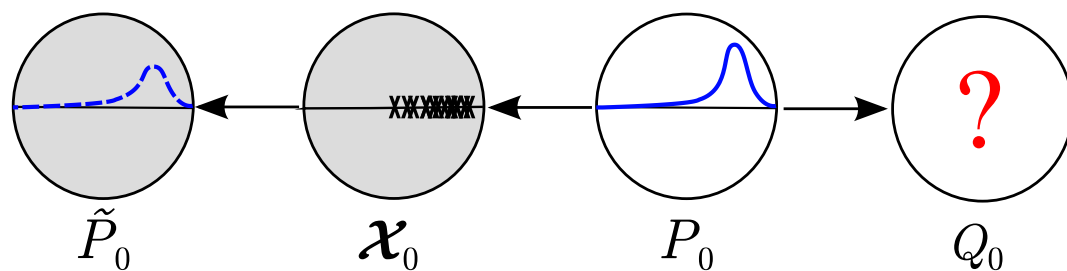
Test



Train



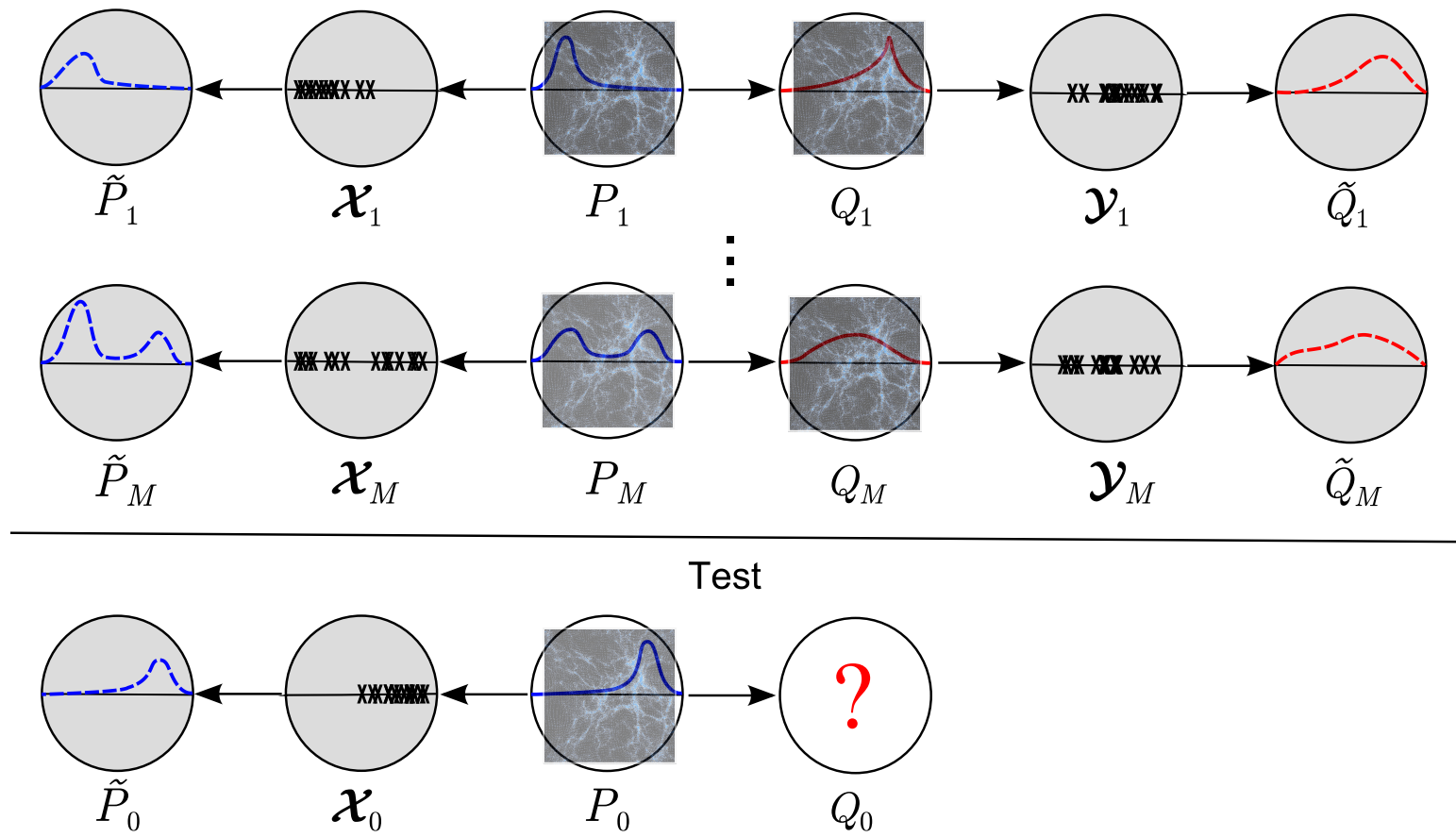
Test



$$W(\tilde{p}_1, \tilde{p}_0) \times \tilde{q}_1 + \dots + W(\tilde{p}_M, \tilde{p}_0) \times \tilde{q}_M = \hat{f}(\tilde{p}_0)$$

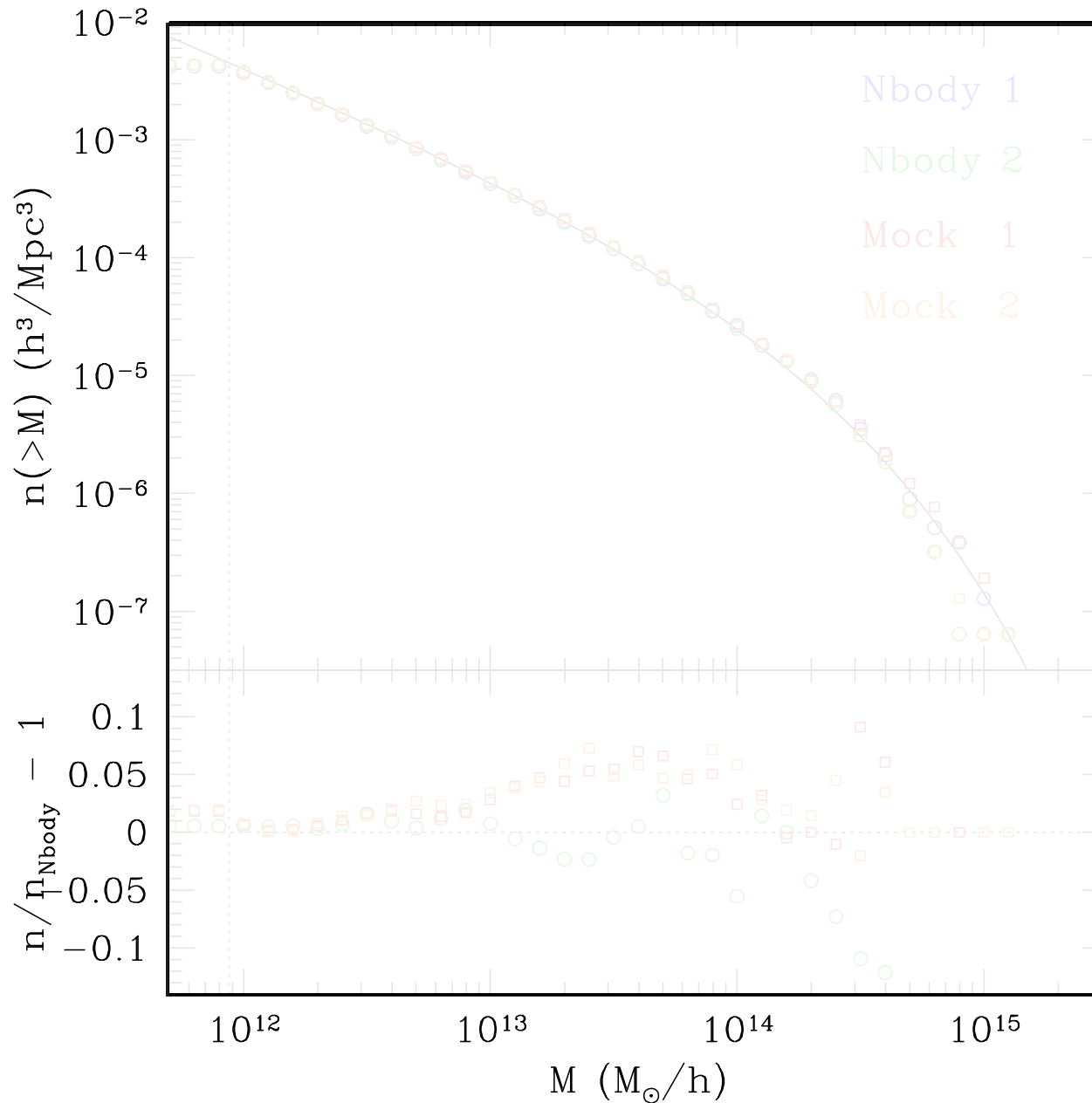
(2LPT) Distribution to (N-body) Distribution Machine Learning algorithms

2LPT (P) $\rightarrow$ N-body(Q)



Very much in prep and speculation

Very preliminary: let's compare the halos statistics



Outline

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You can pull out a BAO signal
even when one of the tracers is really sparse

BAO in QSO x Ly α Cross-Correlations

BOSS at High Redshift

- $\sim 100,000$ $z = 2.15 - 3.5$ quasar spectra
- ~ 45 million Ly α pixels
- First anisotropic BAO measurement at $z > 2$ done this year using Ly α autocorrelations.
- Quasar autocorrelation not yet good enough for BAO, but we can cross-correlate with Ly α !

BAO in QSO x Ly α Cross-Correlations

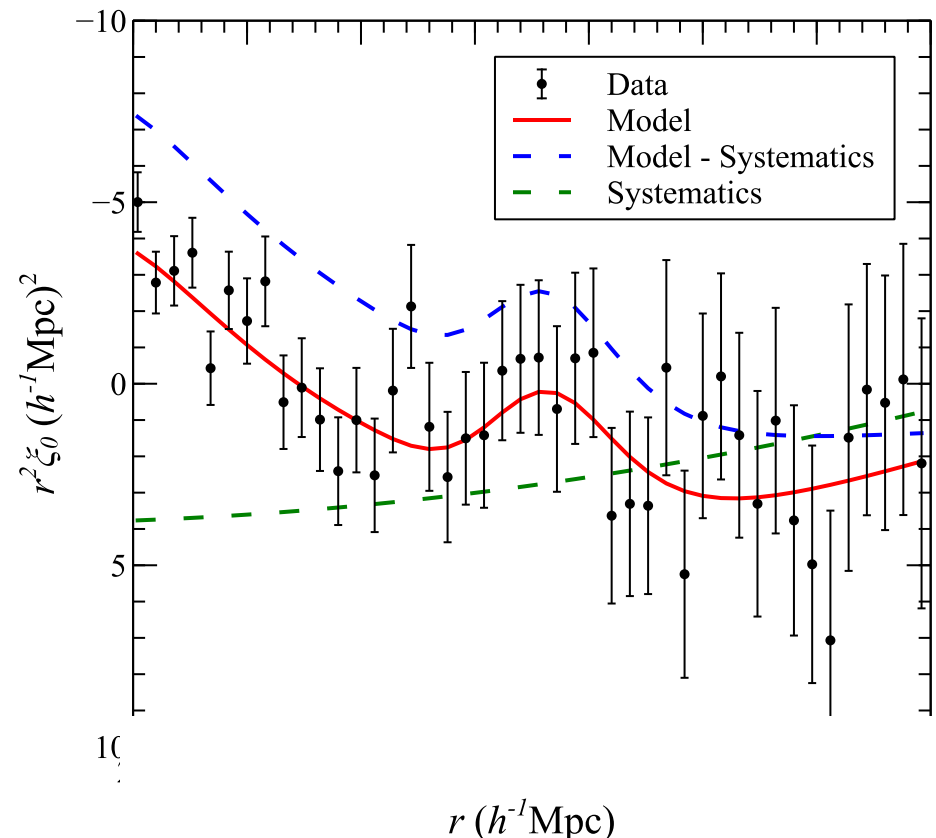
BOSS at High Redshift

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O'Connell, SH , et al. (in prep)
Font-Ribera, Kirkby et al. (in prep)

Our Measurement

- First BAO detection in quasars (via Cross-correlations). Developed new estimator, method for determining covariance matrix, systematic corrections for Ly α forest.



BOSS: What did we learn?

- SDSS III- BOSS : Lots of cosmological constraints and other cool things I didn't get to talk about...
- BAO Reconstruction : Different methods give rise to similar (but not the same) BAO. Possible improvement?
- Ultra-fast simulations of the Universe! A way forward to make many synthetic Universes quickly?
- BAO in cross-correlations! We can look for BAO in the most unlikely places, such as very sparsely sampled quasars. [Other applications of correlating neutral hydrogen and halos?]



The not-so-new anymore assistant faculty Shirley Ho



Machine Learning

Exo-planets

Dark Energy

Inflation

