What is the best way to measure baryonic acoustic oscillations?

(arXiv:0804.0233)

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Contents

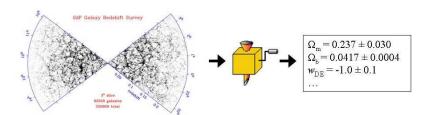
- Introduction
- The evolution of BAO
- 3 The model in practice
- Final remarks

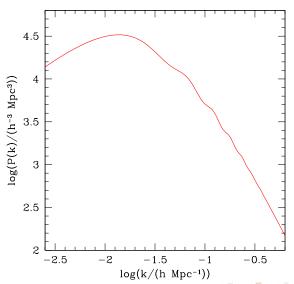
The nature of DE

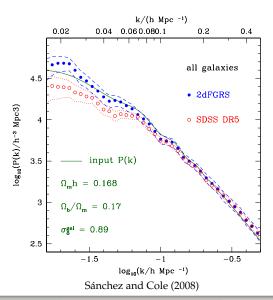
- What is the nature of Dark Energy?
- To distinguish between possible models: analyse

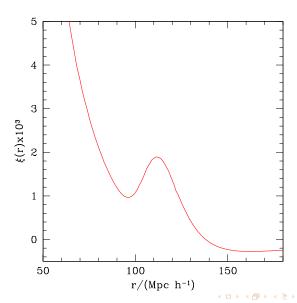
$$w_{\mathrm{DE}} = \rho_{\mathrm{DE}}/p_{\mathrm{DE}}$$

 Precision cosmology: need to control systematics in the analysis pipeline.









 Acoustic oscillations are related to the sound horizon at recombination s.

 Basic idea: use sound horizon scale imprinted in the clustering of galaxies as a standard ruler.

• This test is sensible to w_{DE} trough $D_A(z)$ and H(z).

• Key issue: how does the BAO signature evolves with time?

• According to linear theory the spatial patter does not change.

• Gravitational growth is a non-linear process, even at $r \sim 100 \, {\rm Mpc} \, h^{-1}$.

• Percent level shifts can bias this test and affect its power to constraint w_{DE} .

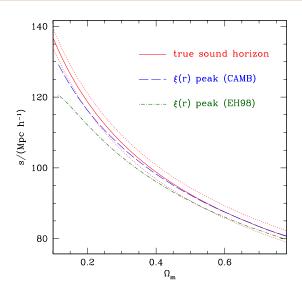
• Numerical simulations show evidence of shifts (Smith et al. 2008, Crocce & Scoccimarro 2008).

• The peak shifts, but there is a bigger problem.

• The position of the peak in $\xi(r)$ does not correspond to the sound horizon.

• The BAO in P(k) have neither a fixed amplitude nor a fixed wavelength.

The peak in $\xi(r)$

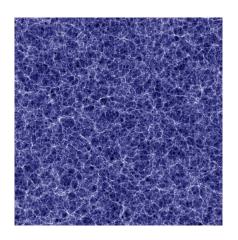


The peak in $\xi(r)$

• Even according to linear theory this leads to biased constraints.

- We need to model the full shape of $\xi(r)$, including:
 - Non-linear evolution.
 - Redshift space distortions.
 - Scale dependent bias.

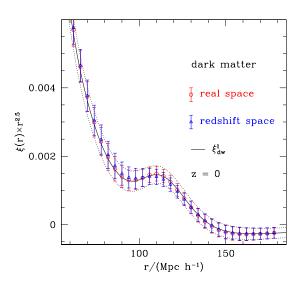
The L-BASICC II run



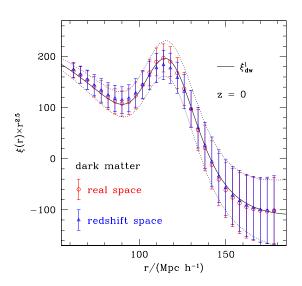
- 50 realizations
- $-L_{\text{Box}} = 1340 \,\mathrm{Mpc} \,h^{-1}$
- $-N_{\rm p}=448^3$
- $m_{\rm p} = 1.75 \times 10^{12} \, {\rm M}_{\odot} \, h^{-1}$
- Cosmological parameters from WMAP+2dF (Sánchez et al.2006)

$$\Omega_{\rm m} = 0.237, \ \sigma_8 = 0.77, n_{\rm s} = 0.954$$

The peak in $\overline{\xi(r)}$



The peak in $\overline{\xi(r)}$



The variance in $\xi(r)$

The covariance matrix is given by

$$C_{\xi}(r,r') \equiv \langle (\xi(r) - \bar{\xi}(r))(\xi(r') - \bar{\xi}(r')) \rangle$$
$$= \int \frac{\mathrm{d}k \, k^2}{2\pi^2} j_0(kr) j_0(kr') \sigma_P^2(k)$$

Where the variance in P(k) is given by

$$\sigma_P(k) = \sqrt{\frac{2}{V}} \left(P(k) + \frac{1}{\bar{n}} \right)$$

• Direct application of this eq. overestimates the true covariance.

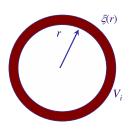
The variance in $\xi(r)$

• An estimate $\hat{\xi}_i$ corresponds to

$$\hat{\xi}_i = \frac{1}{V_i} \int_{V_i} \xi(r) \, \mathrm{d}^3 r,$$

• The covariance of this estimate is given by

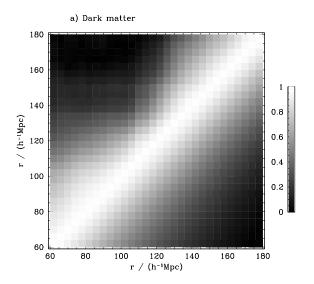
$$\begin{split} C_{\hat{\xi}}(i,j) &= \frac{1}{V_i V_j} \int \mathrm{d}^3 r \int \mathrm{d}^3 r' C_{\xi}(r,r') \\ &= \int \frac{\mathrm{d} k \, k^2}{2\pi^2} \bar{j}_0(k,i) \bar{j}_0(k,j) \sigma_P^2(k), \end{split}$$



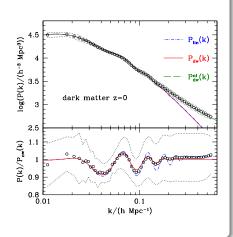
where

$$\bar{j}_0(k,i) = \frac{1}{V_i} \int_{V_i} j_0(kr) \, \mathrm{d}^3 r.$$

The full covariance matrix



- NL evolution distorts the full shape of P(k) (halofit).
- Erases the higher harmonic peaks (Crocce & Scoccimarro 2008).
- A model of the NL *P*(*k*) must include both effects.
- What is the most important effect to model $\xi(r)$?.

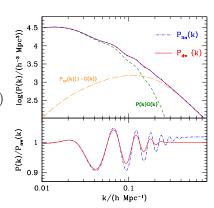


• To model this compute a "dewiggled" power spectrum:

$$P_{dw}(k) = P_1(k)G(k) + P_{nw}(k)(1 - G(k))$$

where

$$G(k) \equiv \exp\left[-(k/\sqrt{2}k_{\star})^{2}\right]$$



• This can be interpreted in terms of RPT (Crocce & Scoccimarro 2006).

$$P_{\rm nl}(k) = P_{\rm lin}(k)G(k) + P_{\rm mc}(k)$$

• The propagator behaves as a Gaussian in the high-k limit, with

$$k_{\star} = \left[\frac{1}{3\pi^2} \int \mathrm{d}k P_{\mathrm{lin}}(k) \right]^{-1/2}$$

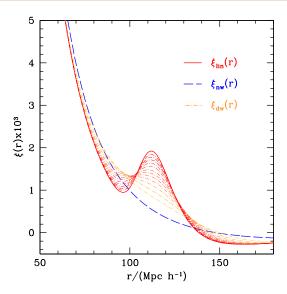
• The term P_{mc} does show acoustic oscillations.

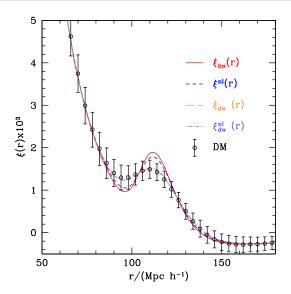


• The correlation function will be given by

$$\xi_{\rm dw}(r) = \xi_{\rm lin}(r) \otimes \tilde{G}(r) + \xi_{\rm nw}(r) \otimes (1 - \tilde{G}(r))$$

- The BAO signal is contained in the first term.
- The convolution broadens and shifts the peak towards smaller scales.





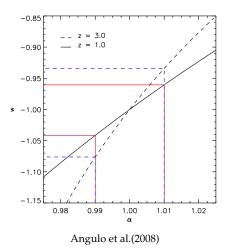
- We tested the efficiency of this model to constrain w_{DE} .
- A very simple case: we know everything except w_{DE} .
- A change in w_{DE} produces a rescaling of the wavenumber from k_{true} to k_{app} .

$$\alpha = \frac{k_{\rm app}}{k_{\rm true}}$$

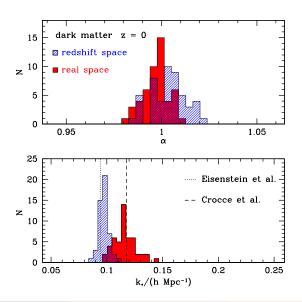
for the correlation function

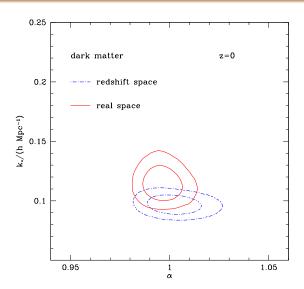
$$r_{\rm app} = r_{\rm true}/\alpha$$

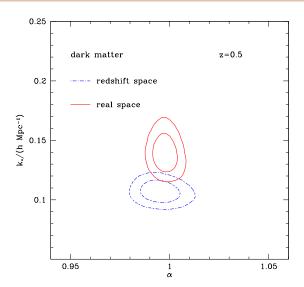


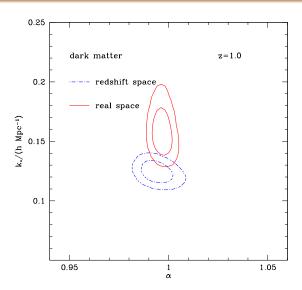


- Error in BAO scale translates to bigger error in $w_{\rm DE}$.
- For fixed $\Omega_{\rm m}$ require distance scale to 0.2% to get 1% in w_{DF} .
- Demands accurate knowledge of systematics





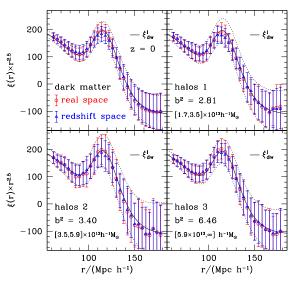




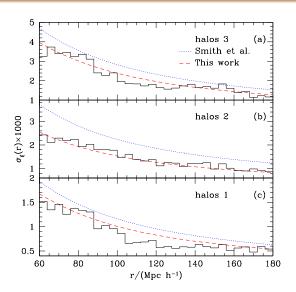
Halo bias

- Galaxies are hosted within dark matter haloes.
- Understanding halo bias is a first step towards galaxy bias.
- Clusters are good alternatives to galaxies in BAO analysis.
- We measured $\xi_{hh}(r)$ for three halo samples.
 - Sample 1: $1.7 \times 10^{13} \ \mathrm{M}_{\odot} \, h^{-1} < m < 3.5 \times 10^{13} \ \mathrm{M}_{\odot} \, h^{-1}$
 - Sample 2: $3.5 \times 10^{13} \ {\rm M}_{\odot} \, h^{-1} < m < 5.9 \times 10^{13} \ {\rm M}_{\odot} \, h^{-1}$
 - Sample 3: $5.9 \times 10^{13} \, \mathrm{M}_{\odot} \, h^{-1} < m$

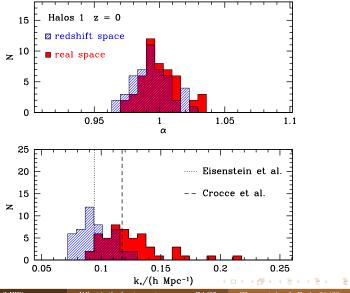
Halo bias



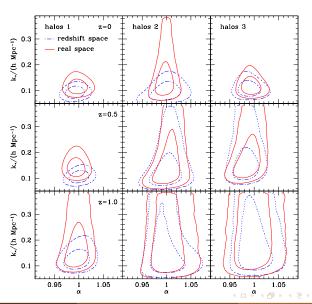
The variance in $\xi(r)$



Halo bias



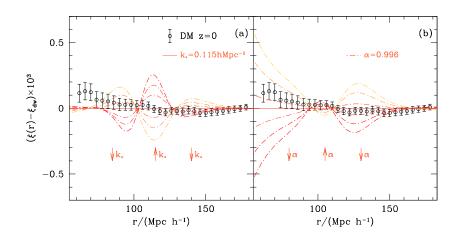
Halo bias

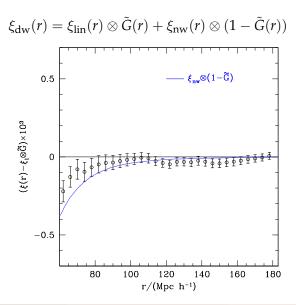


Improving the model

- The model is slightly biased towards α < 1.
- Future surveys will have much smaller sample variances (e.g Euclid or ADEPT)
- This may lead to systematic errors in w_{DE} .
- How can we correct for this small bias?

Improving the model





• In the scales of the accoustic peak (Crocce & Scoccimarro 2008)

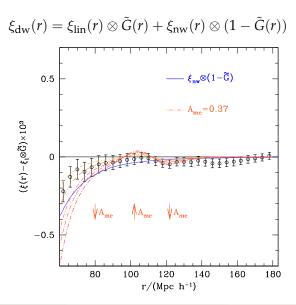
$$\xi_{\rm mc}(r) \propto \xi_{\rm lin}' \, \xi_{\rm lin}^{(1)}(r)$$

where

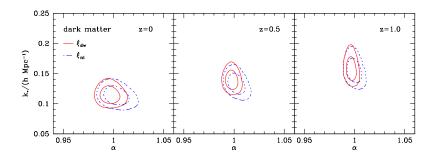
$$\xi_{\text{lin}}^{(1)}(r) \equiv \hat{r} \cdot \nabla^{-1} \xi_{\text{lin}}(r) = 4\pi \int P_{\text{lin}}(k) j_1(kr) k \, dk$$

• Based on this result they proposed an approximate model where

$$\xi_{\rm nl}(r) = \xi_{\rm lin}(r) \otimes \tilde{G}(r) + A_{\rm mc} \, \xi_{\rm lin}'(r)$$

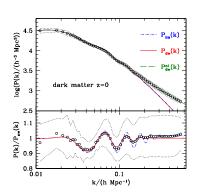


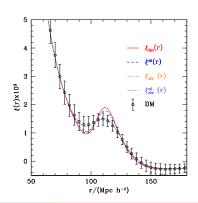
- The new model helps to alleviate the small bias towards $\alpha < 1$.
- RPT could help to improve the constraints even further.



P(k) vs $\xi(r)$

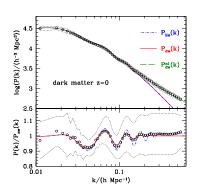
- P(k) and $\xi(r)$ are affected in different ways by NL evolution, redshift space distortions and bias.
- Which one offers more advantages?

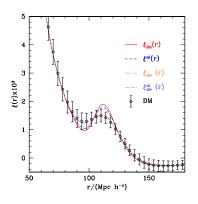




NL evolution

- The shape of P(k) is strongly affected by NL evolution.
- $\xi(r)$: k_{\star} accurately describes the damping of the oscillations.





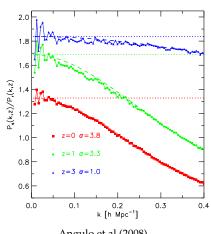
Redshift space distortions

• *P*(*k*): Kaiser factor only applicable on large scales

$$S \equiv \left(1 + \frac{1}{5}\beta + \frac{2}{5}\beta^2\right)$$

 Distortions require extra parameter

$$\frac{P_{\rm s}(k)}{P(k)} = S\left(1 + \sigma^2 k^2\right)$$



Angulo et al.(2008)

Redshift space distortions

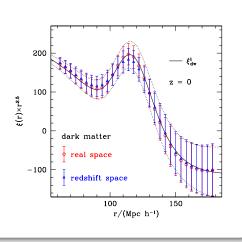
• $\xi(r)$: well described by Kaiser factor

$$S = \frac{\xi_s(r)}{\xi(r)} \equiv \left(1 + \frac{1}{5}\beta + \frac{2}{5}\beta^2\right)$$

• The effect of σ is masked in the value of k_{\star}

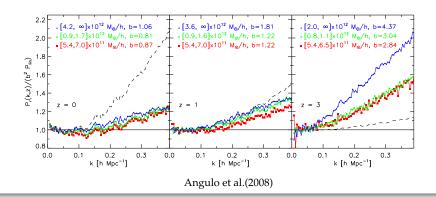
$$k_{\star} = 0.115 \pm 0.009$$

$$k_{\star} = 0.097 \pm 0.004$$



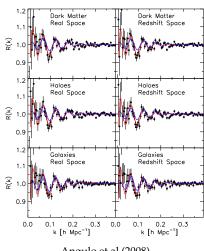
Halo bias

- P(k): halo bias is scale dependent.
- $\xi(r)$: halo bias can be treated as constant.



Which one is better?

- $\xi(r)$ is less sensitive to these problems.
- Other methods propose to divide P(k) by a smooth function.
- These approaches can not be easily generalized to $\xi(r)$.
- How do these approaches perform in our case?.



Angulo et al.(2008)

Which one is better?

- The shape of $\xi(r)$ provides extra information.
- This gives better constraint on the distance scale (and hence on w_{DE}).
- Large-scale shape of $\xi(r)$ useful to constraint other parameters.

 $\bullet z = 0$

$$p(k): \alpha = 1.006 \pm 0.008$$

 $\xi(r): \alpha = 1.003 \pm 0.008$

• z = 0.5

$$p(k): \alpha = 1.003 \pm 0.007$$

$$\xi(r): \ \alpha = 1.002 \pm 0.005$$

• z = 1.0

$$p(k)$$
: $\alpha = 1.000 \pm 0.006$

$$\xi(r): \ \alpha = 1.000 \pm 0.003$$

Which one is better?

- The results from P(k) show a bias towards $\alpha > 1$.
- Future surveys will cover much larger volumes.
- The results obtained from $\xi(r)$ are more consistent with $\alpha = 1$.

$$z = 0$$

$$p(k): \alpha = 1.006 \pm 0.008$$

 $\xi(r): \alpha = 1.003 \pm 0.008$

•
$$z = 0.5$$

$$p(k): \alpha = 1.003 \pm 0.007$$

 $\xi(r): \alpha = 1.002 \pm 0.005$

•
$$z = 1.0$$

$$p(k): \ \alpha = 1.000 \pm 0.006$$

$$\xi(r)$$
: $\alpha = 1.000 \pm 0.003$

Conclusions

- Peak position is *not* equivalent to *s*, even in linear theory.
- It is necessary to model the full shape of $\xi(r)$.
- A simple model for $\xi(r)$: NL evolution, redshift-space distortions and halo bias.
- Systematics are simpler to deal with in $\xi(r)$ than in P(k).
- Full shape of $\xi(r)$ provides tighter constraints on w_{DE} than P(k).
- Both statistics will allow to reach the full potential of BAO as DE probes.

Pan-STARRS

"Data! Data! Data!, he cried impatiently. I can't make bricks without clay".

Sherlock Holmes, The Adventure of the Copper Beeches (1892)

Thank you!



Pan-STARRS





















