

Effects of the local features in initial power spectrum on baryon acoustic oscillations

Koichiro Horiguchi (Nagoya University)

Collaborators

Kiyotomo Ichiki (Nagoya University)

Takahiko Matsubara (KEK)

2017Dec14 / COSPA2017 @ Kyoto University

Contents

Introduction

Motivation : CMB & BAO

Features in initial power spectrum

Effects of feature models on BAO

Results & discussion

Parameter estimation using MCMC method

Summary

Contents

Introduction

Motivation : CMB & BAO

Features in initial power spectrum

Effects of feature models on BAO

Results & discussion

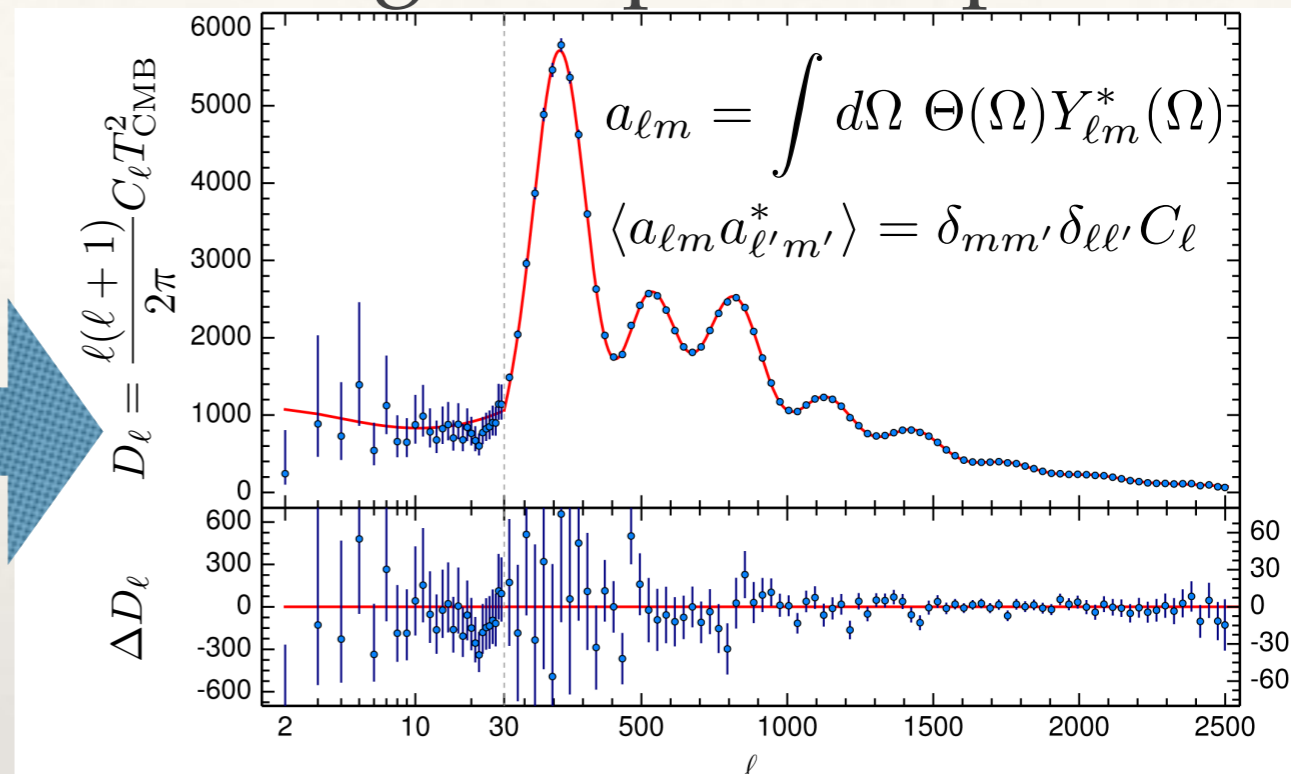
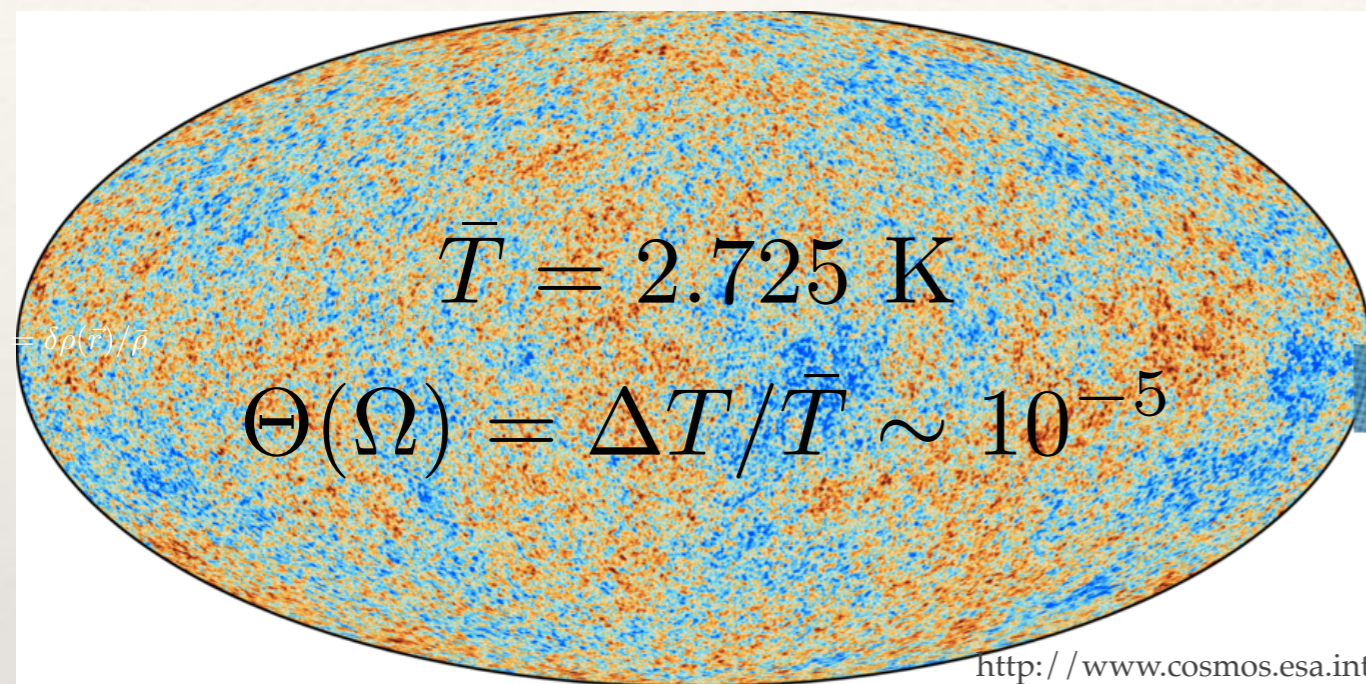
Parameter estimation using MCMC method

Summary

CMB vs BAO

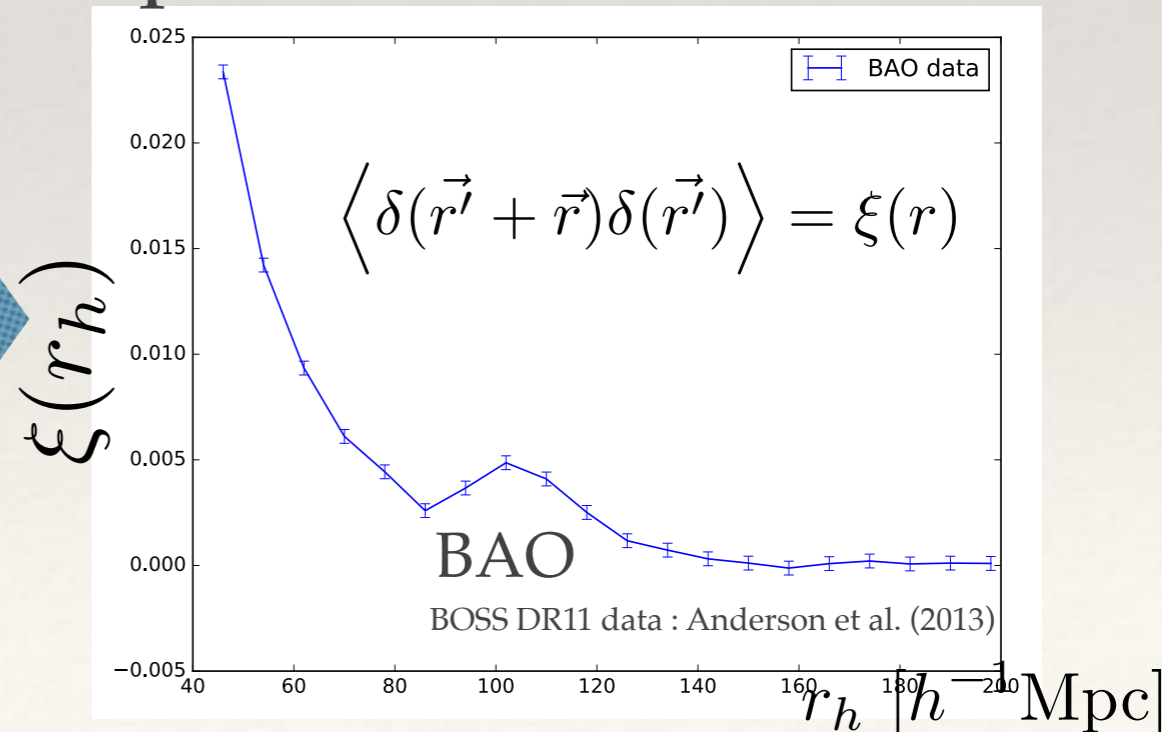
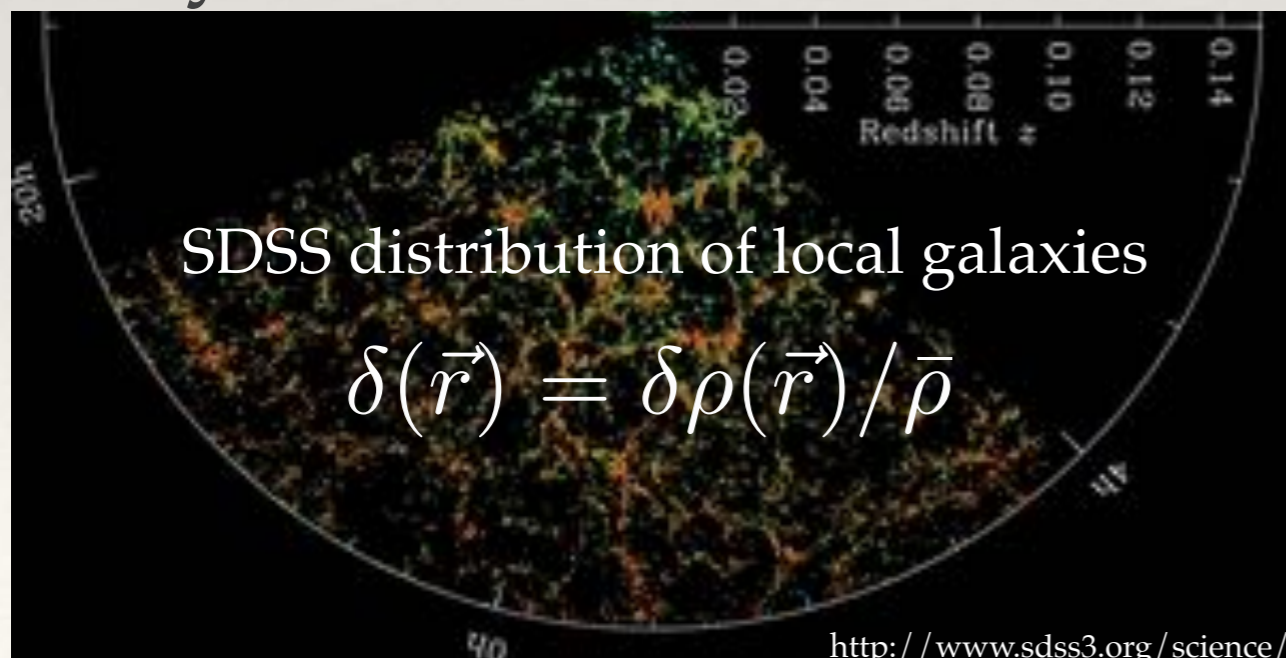
Cosmic Microwave Background

CMB angular power spectrum



Baryon Acoustic Oscillations

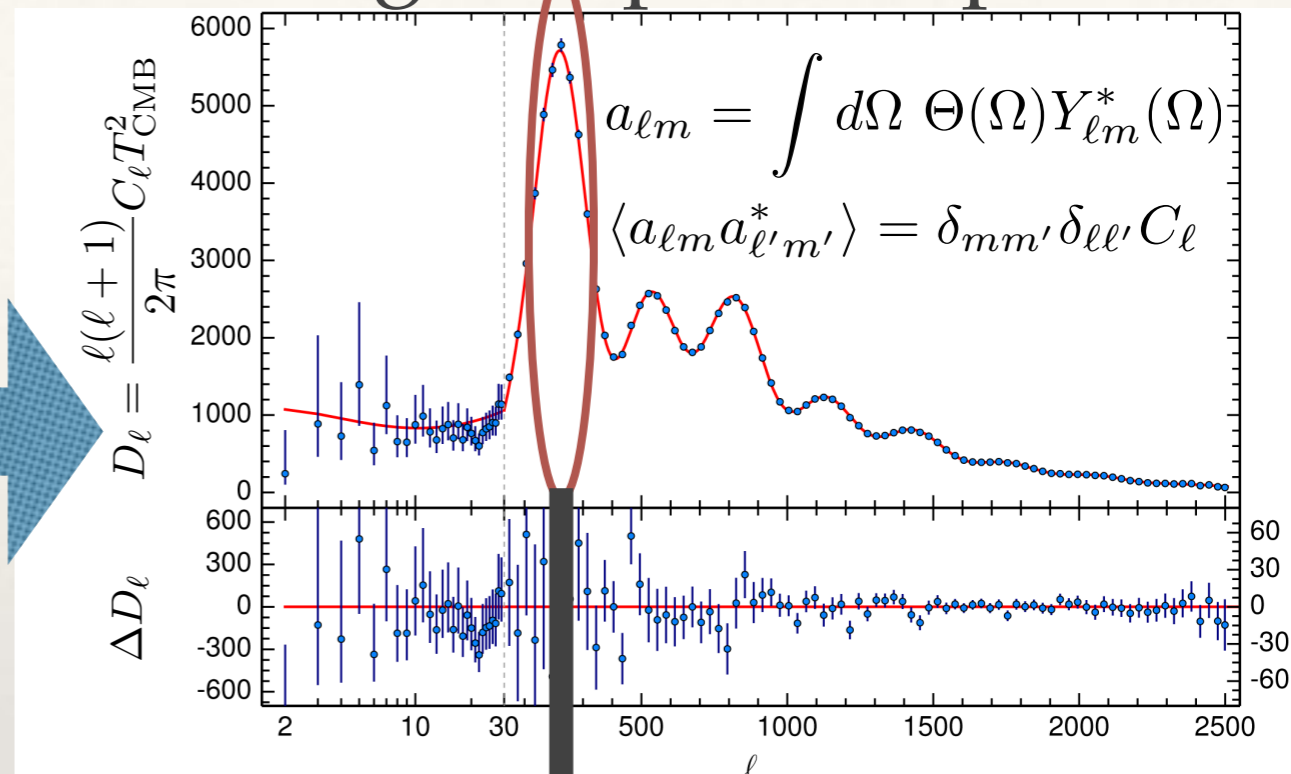
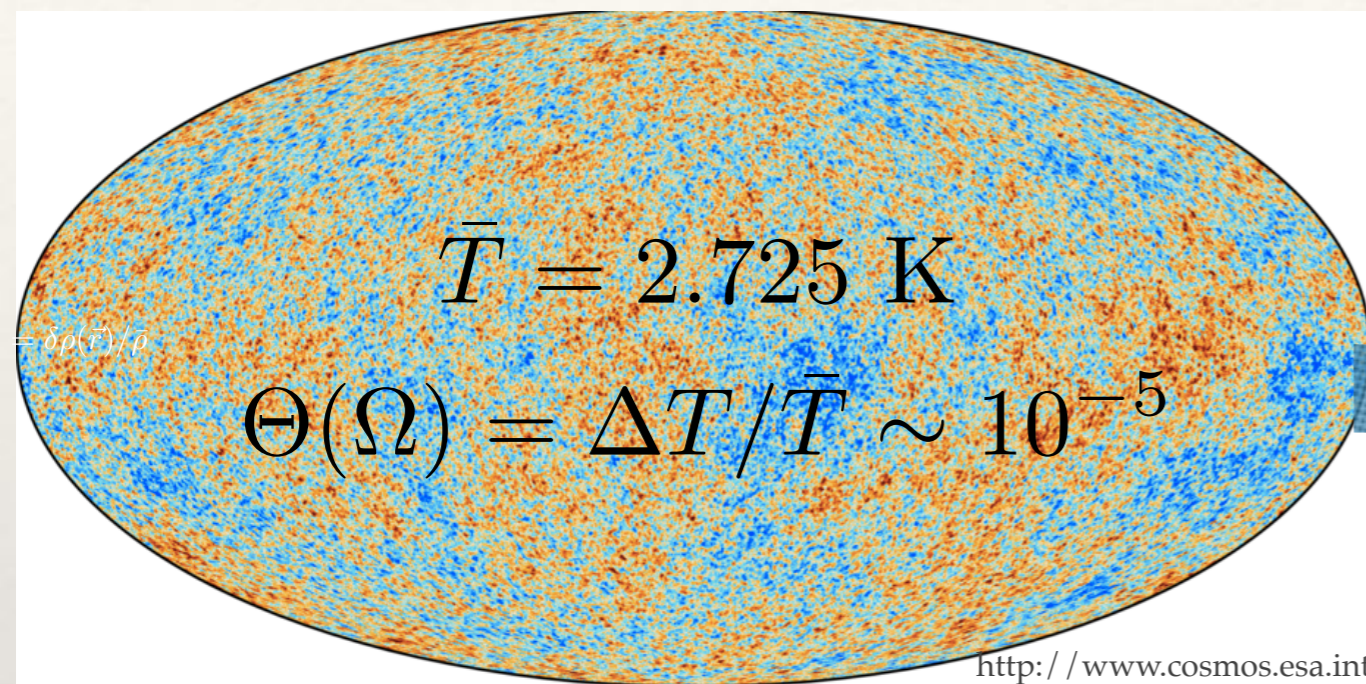
Two points correlation function



CMB vs BAO

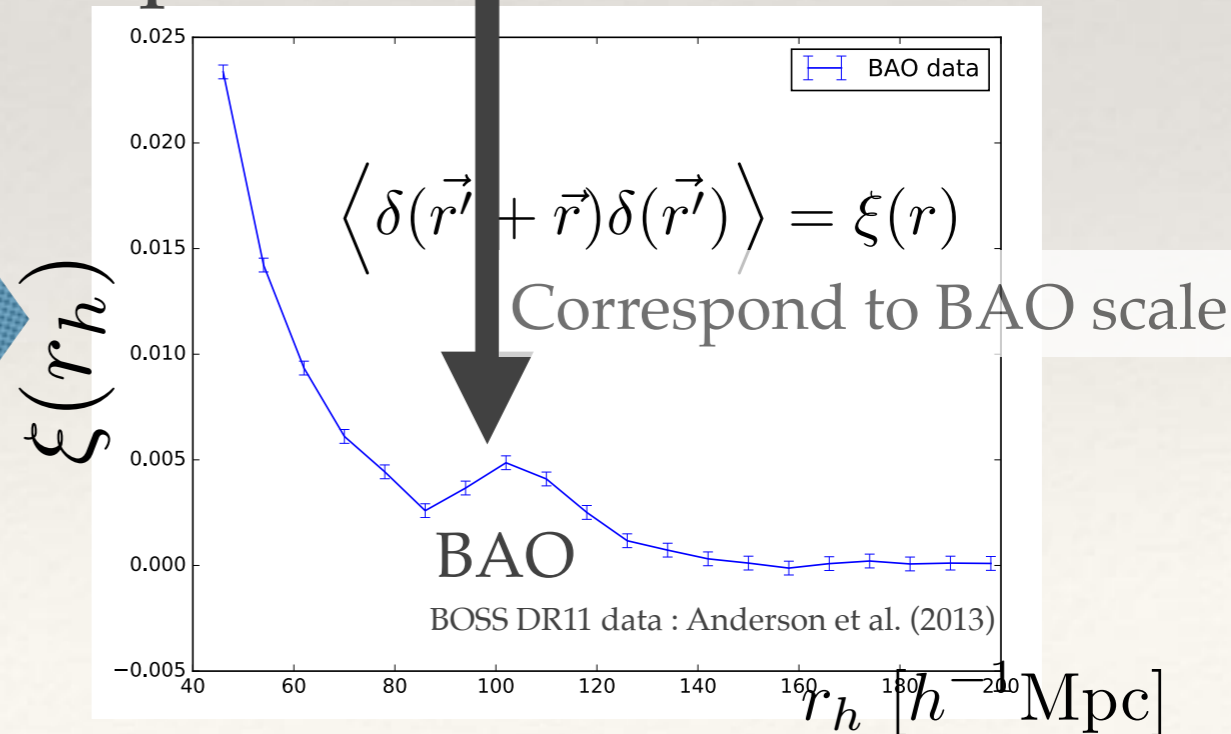
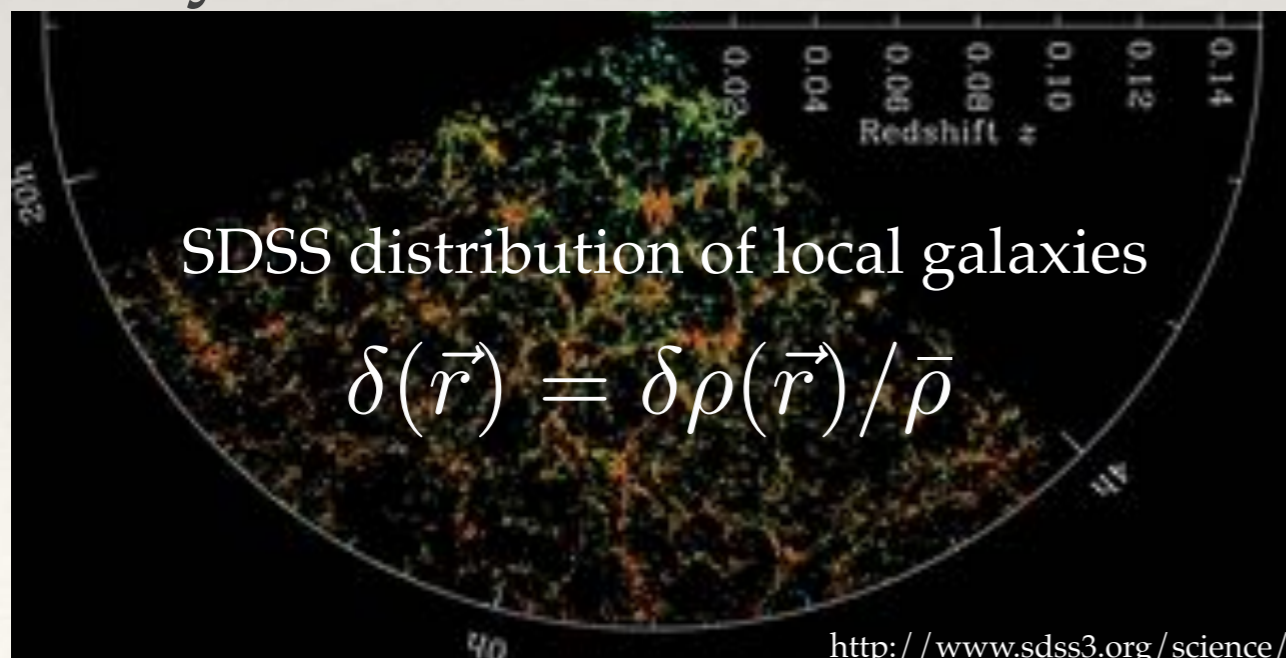
Cosmic Microwave Background

CMB angular power spectrum



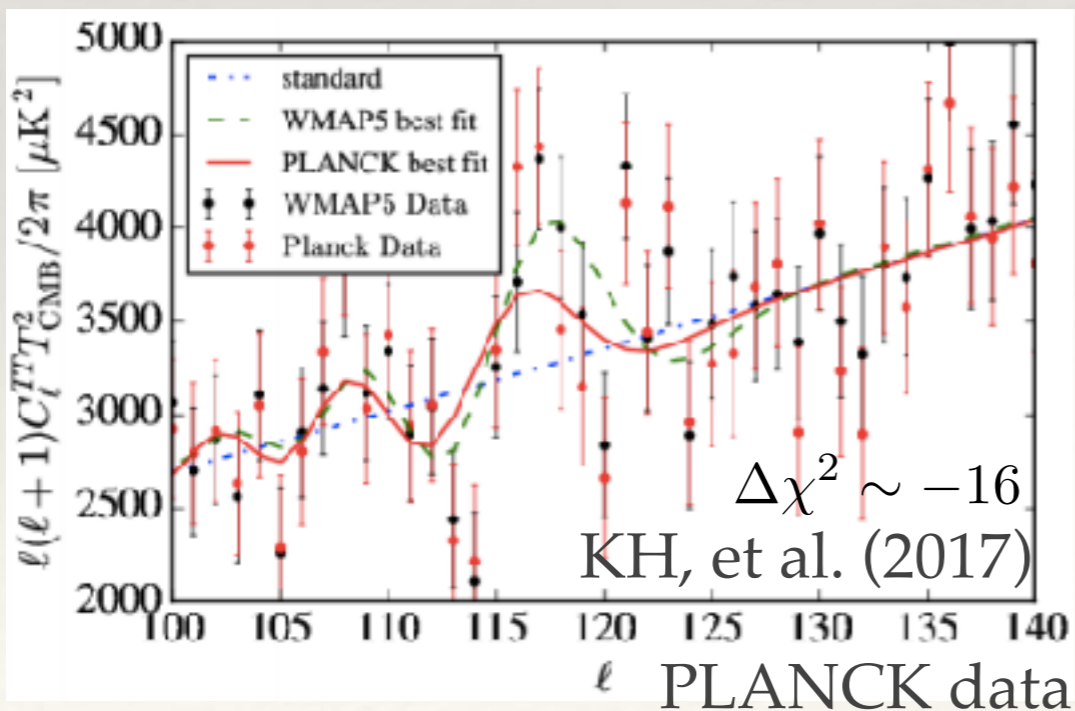
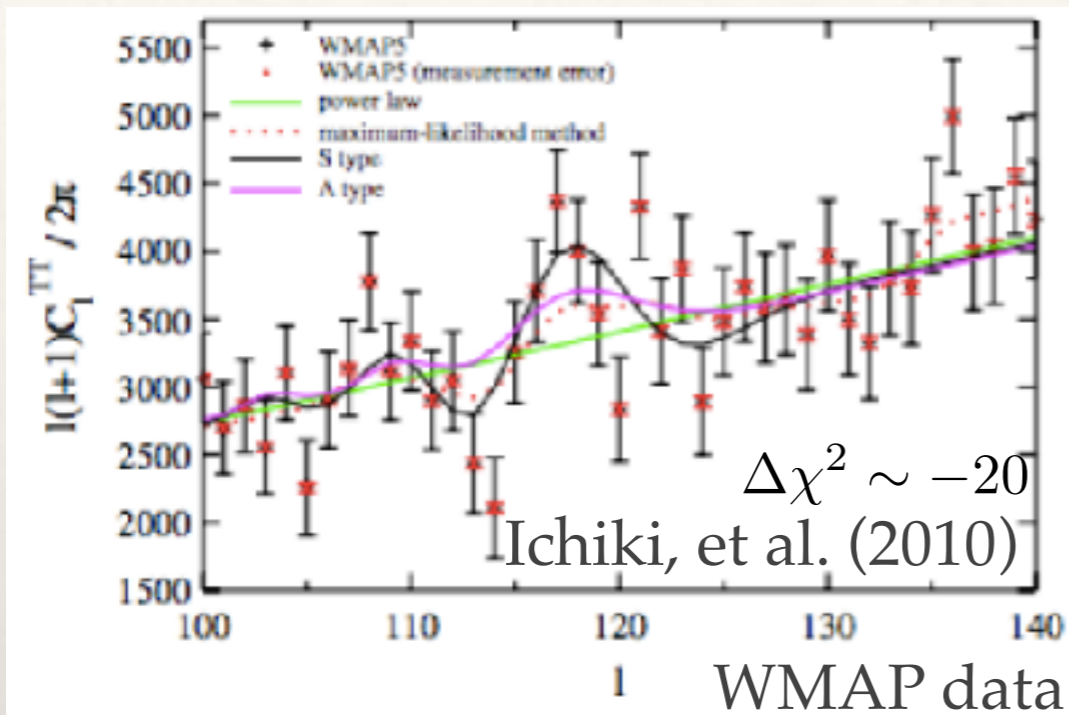
Baryon Acoustic Oscillations

Two points correlation function

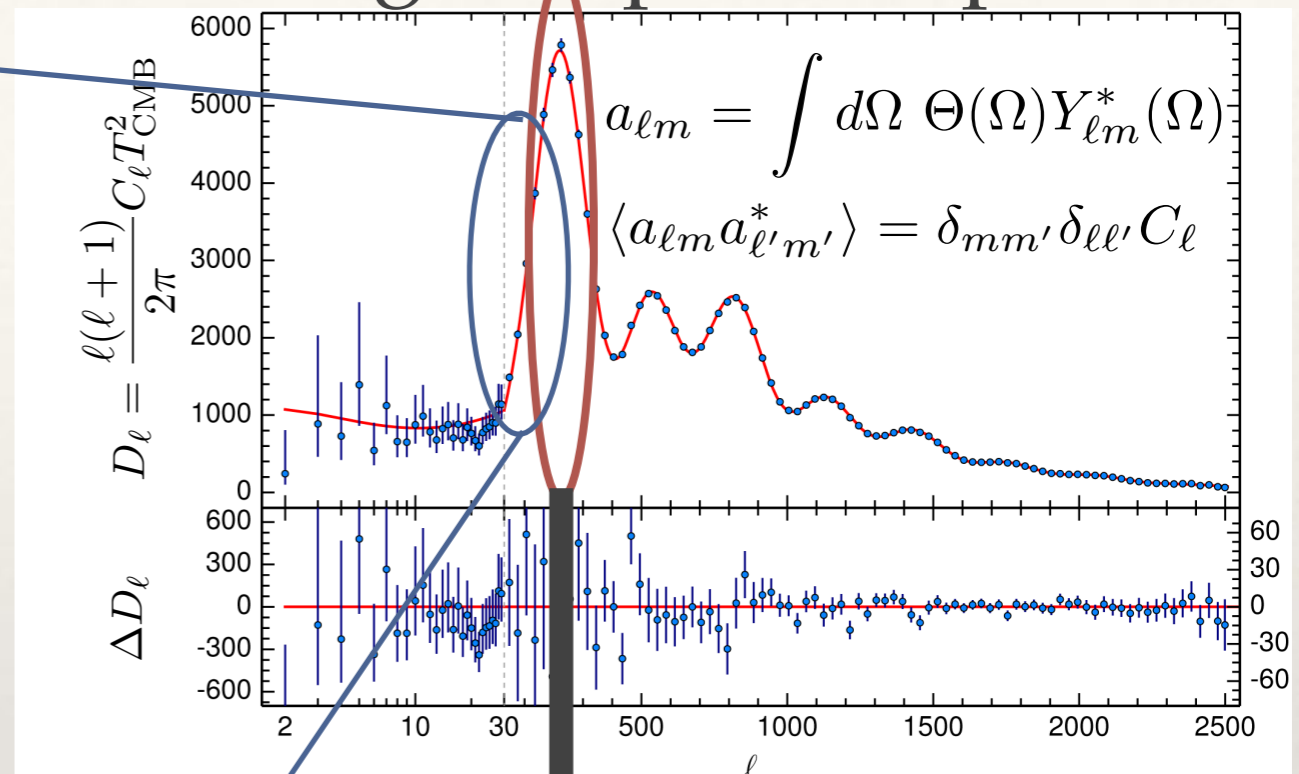


CMB vs BAO

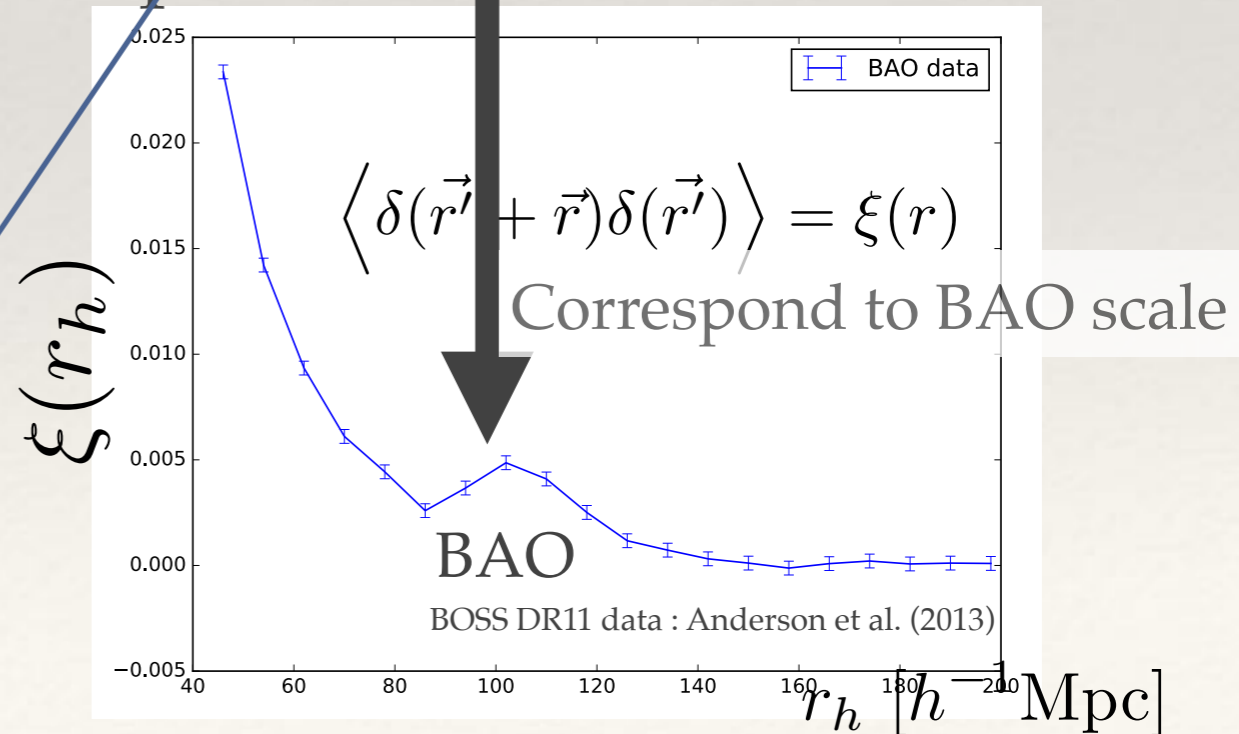
Features at $\ell \sim 120$



CMB angular power spectrum

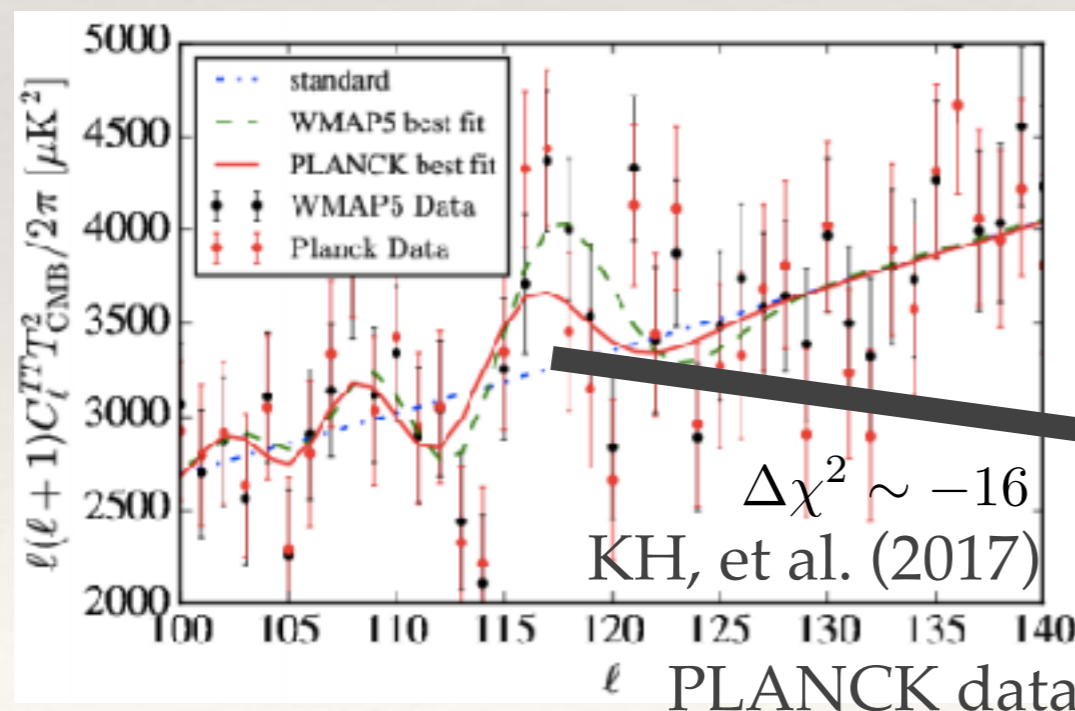
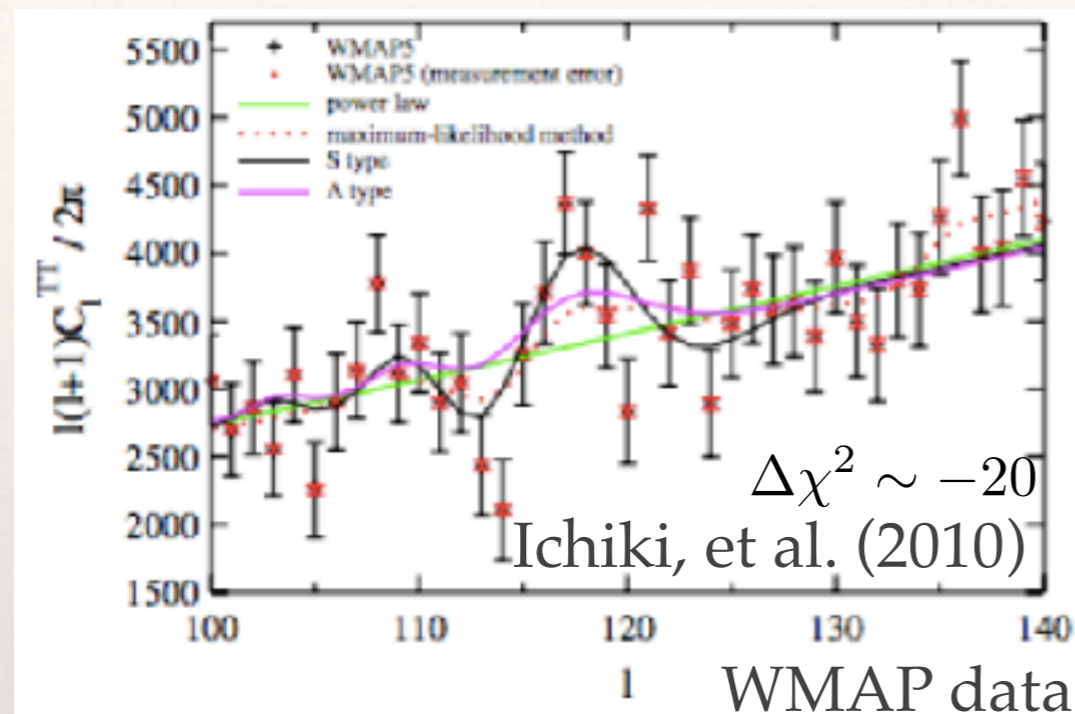


Two points correlation function

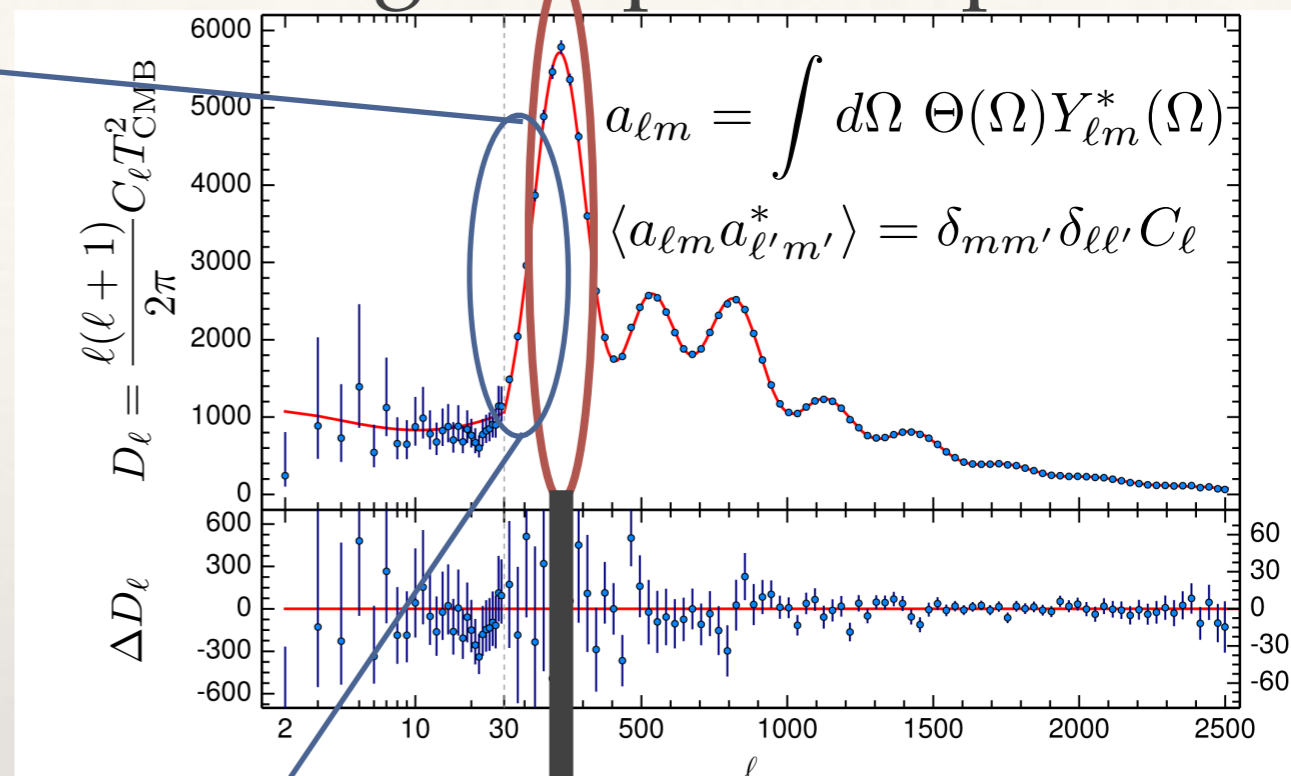


CMB vs BAO

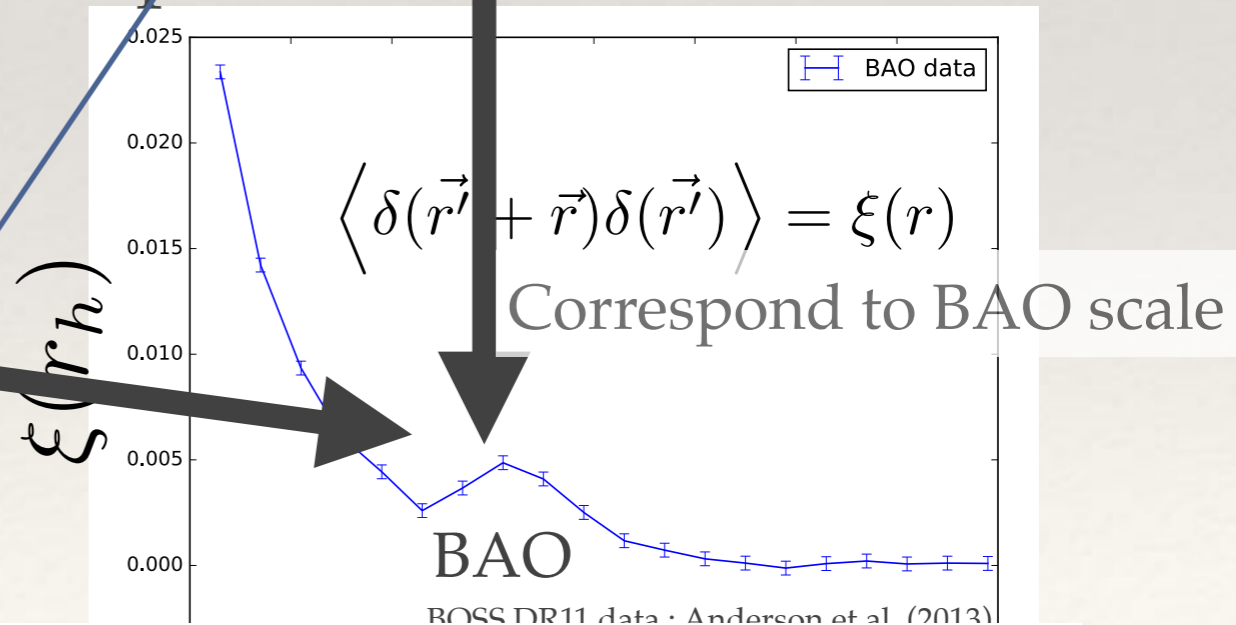
Features at $\ell \sim 120$



CMB angular power spectrum



Two points correlation function



We want to know the effects of these features on BAO [pc]

Contents

Introduction

Motivation : CMB & BAO

Features in initial power spectrum

Effects of feature models on BAO

Results & discussion


Parameter estimation using MCMC method

Summary

Feature models in initial power spectrum

We assume that these features are originated from the initial power spectrum $P(k)$

$$C_\ell = \int T_\ell^2(k) P(k) d\ln k$$

$$\xi(r) = \int T_B^2(k) P(k) \frac{\sin(kr)}{kr} d\ln k$$


$$P(k) = P_s(k) + P_F(k)$$

$$P_s(k) = A \left(\frac{k}{k_0} \right)^{n_s - 1}$$

$$k_0 = 0.05 \text{ [Mpc}^{-1}\text{]}$$

(i) Delta type feature

$$P_F(k) = Bk\delta(k - k_*)$$

(ii) Oscillating type feature

$$P_F(k) = B \left(\frac{k}{k_0} \right)^{n_s - 1} \times \cos \left(\frac{\pi(k - k_*)}{\kappa} \right) \exp \left(-\frac{(k - k_*)^2}{\kappa^2} \right)$$

(K.H. et al 2017)

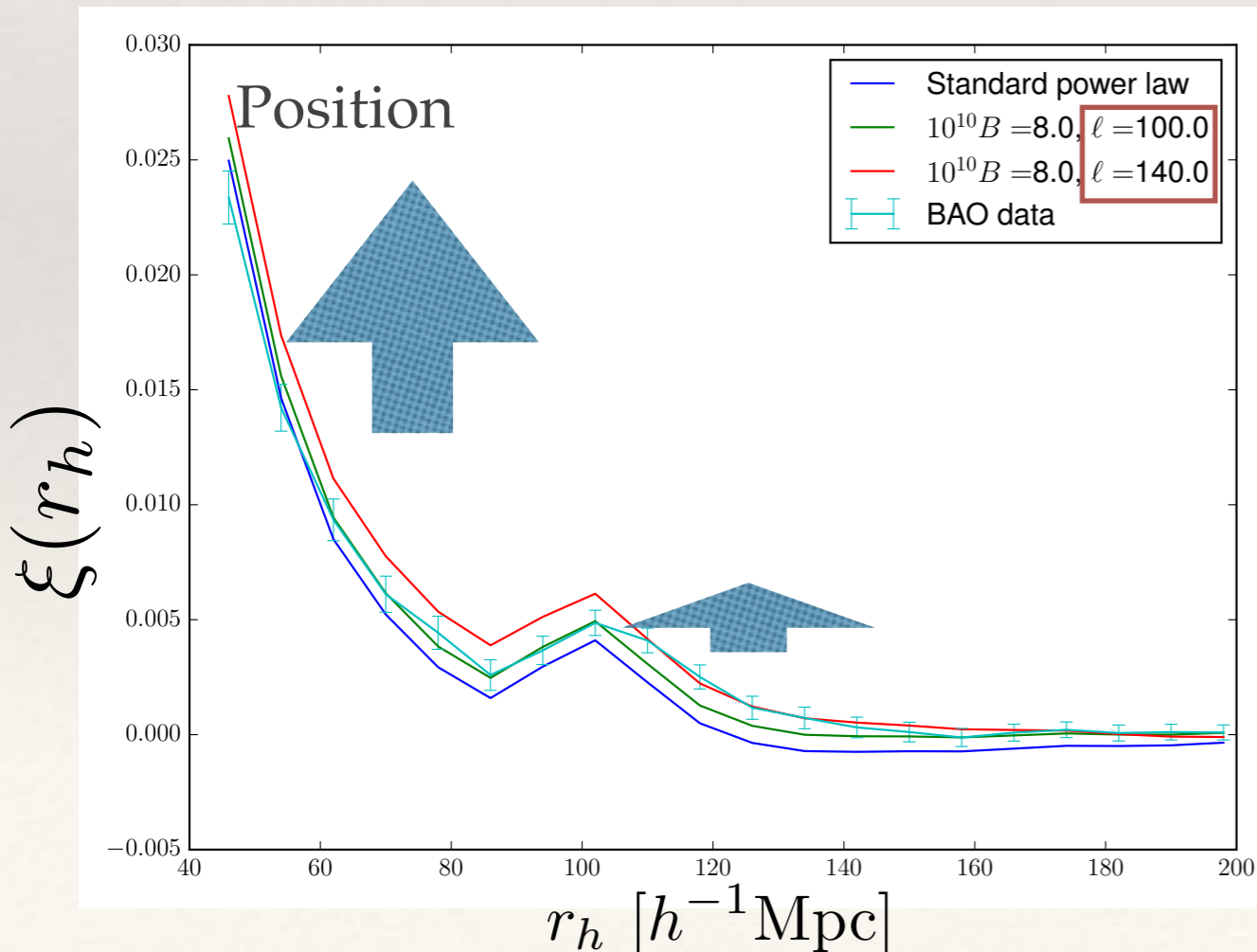
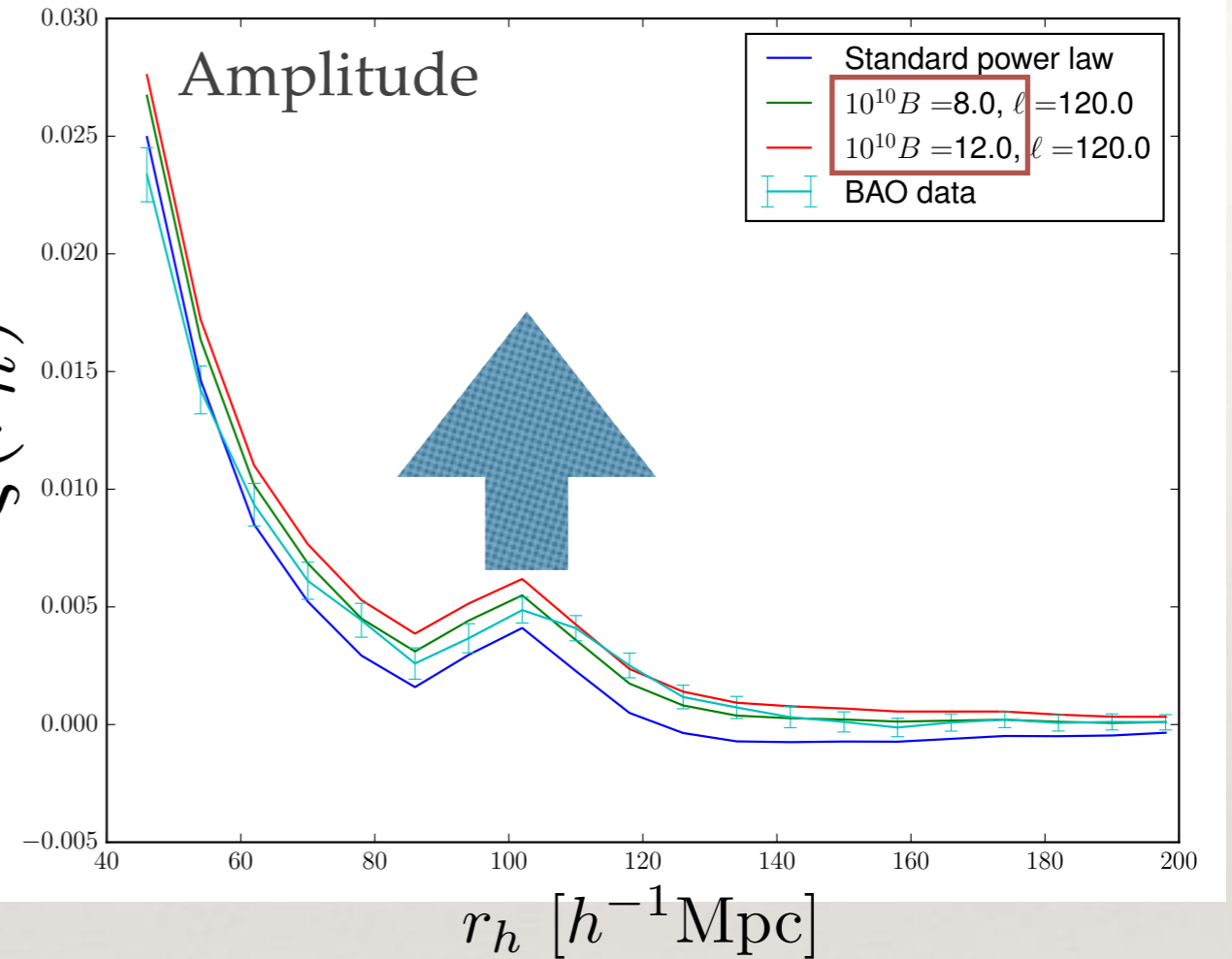
B : Amplitude κ : Width [10^{-4}Mpc^{-1}] k_* : Position

$k_* d_{ang} \sim \ell$ d_{ang} : Angular diameter distance to LSS

Effects of features on BAO : Delta type feature

(i) Delta type : Amplitude & Position

$$P_F(k) = A \left(\frac{k}{k_0} \right)^{n_s - 1} + \underline{Bk} \delta(k - \underline{k_*}) \xi(r_h)$$



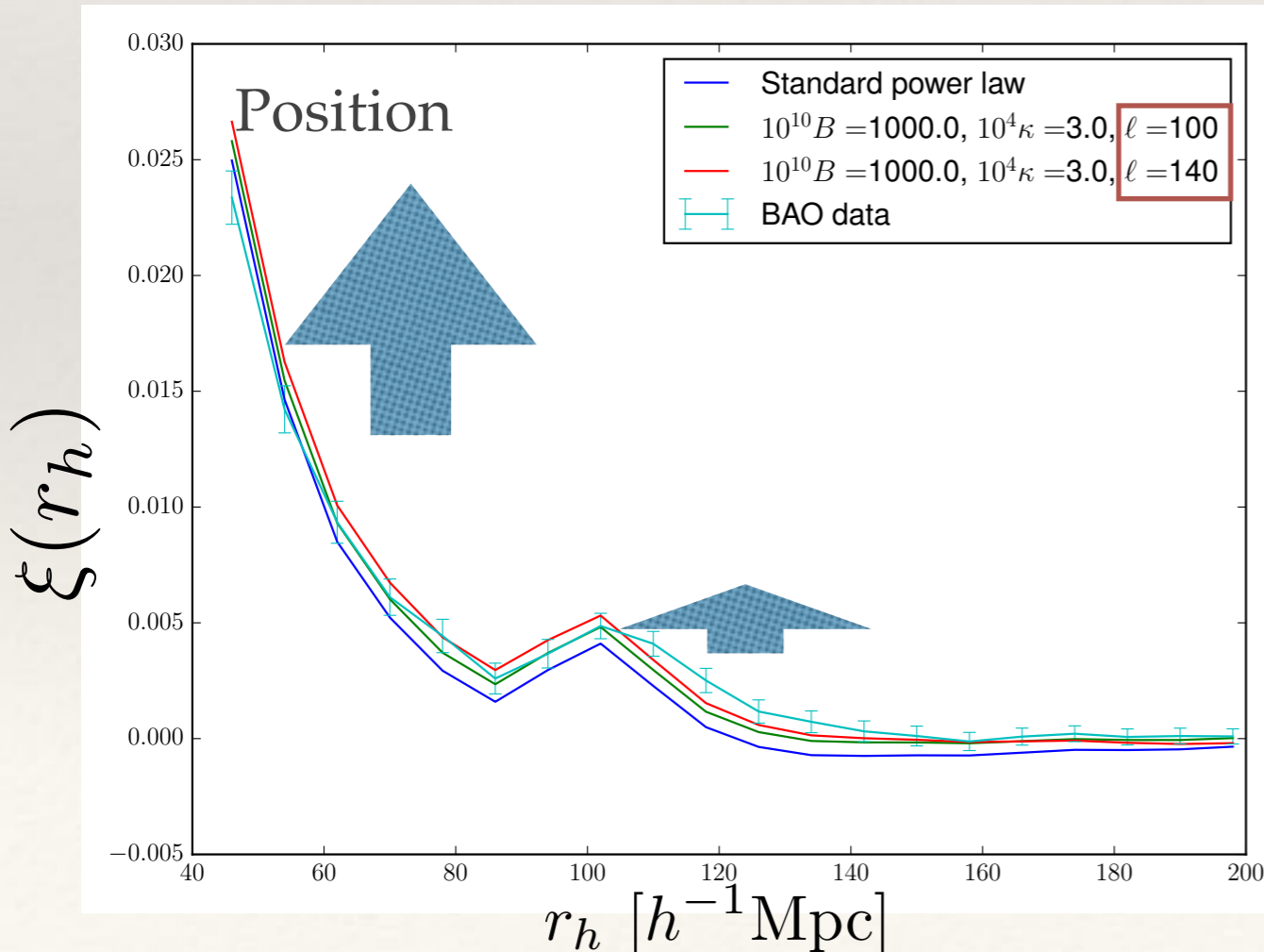
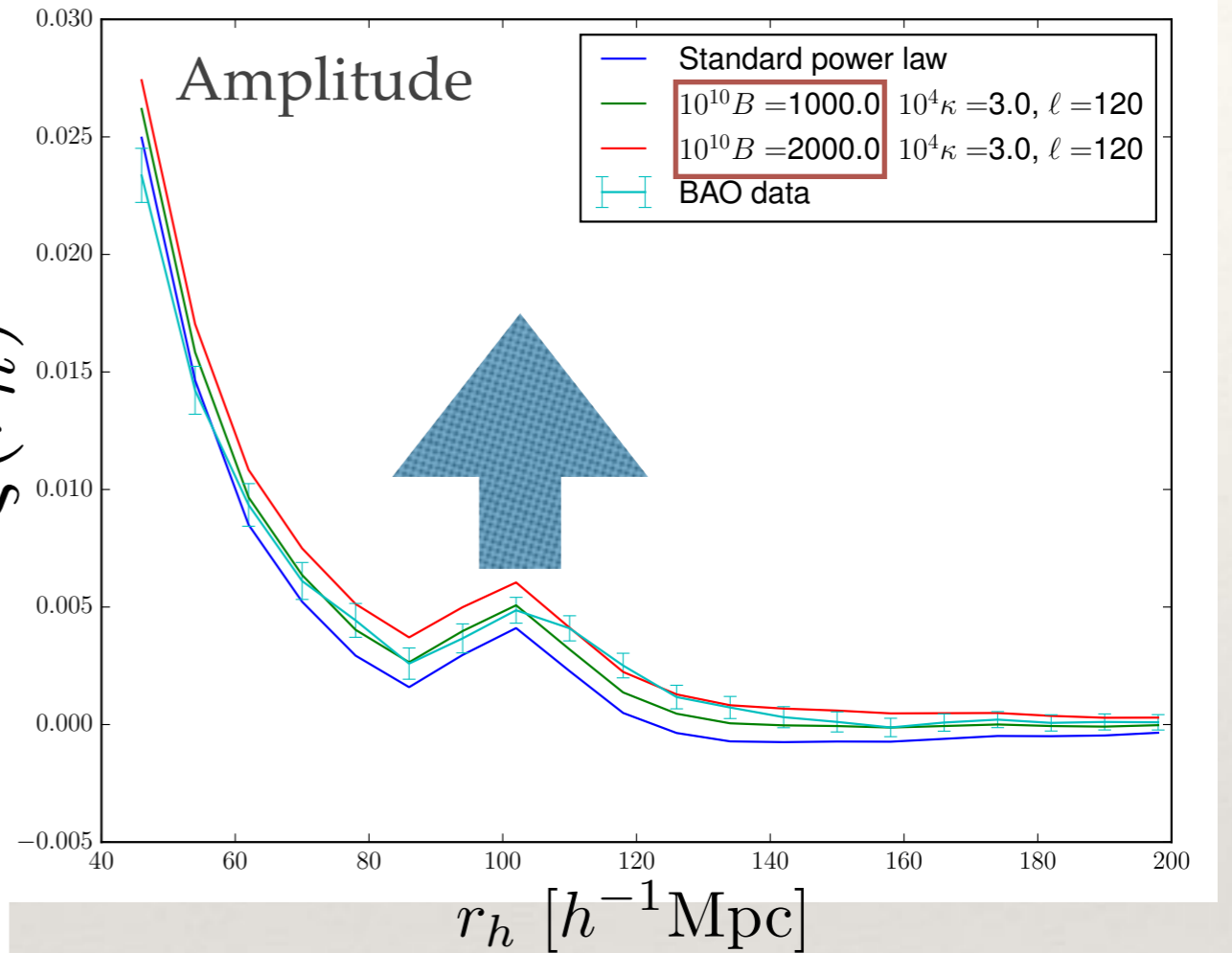
Center ell becomes larger

Amplitude at small scales becomes larger

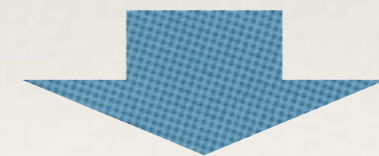
Effects of features on BAO : Oscillating feature

(ii) Oscillating type : Amplitude & Position

$$P(k) = A \left(\frac{k}{k_0} \right)^{n_s-1} + \underline{B} \left(\frac{k}{k_0} \right)^{n_s-1} \times \cos \left(\frac{\pi(k - \underline{k_*})}{\underline{\kappa}} \right) \exp \left(-\frac{(k - \underline{k_*})^2}{\underline{\kappa}^2} \right) \xi(r_h)$$



Center ell becomes larger

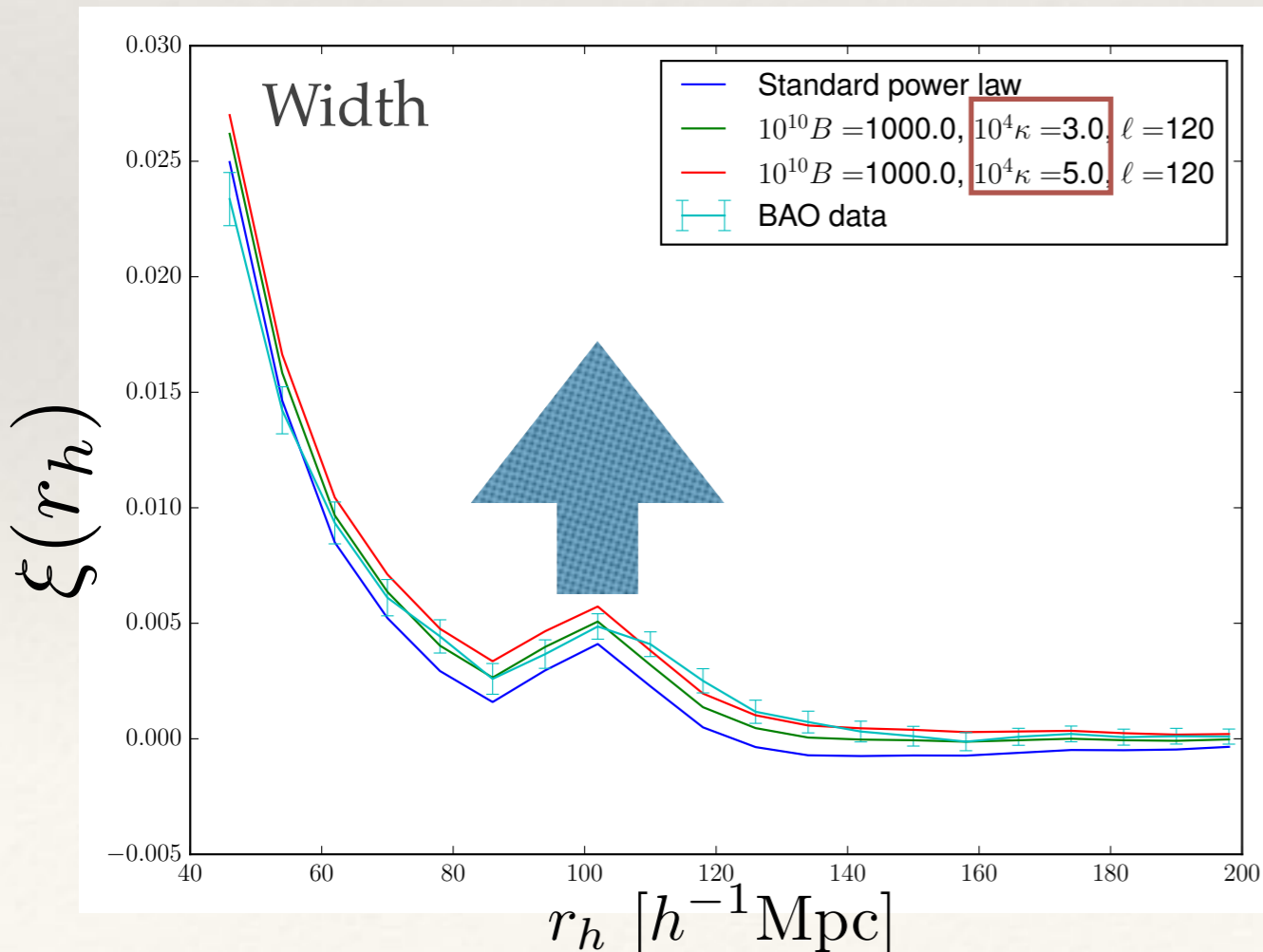
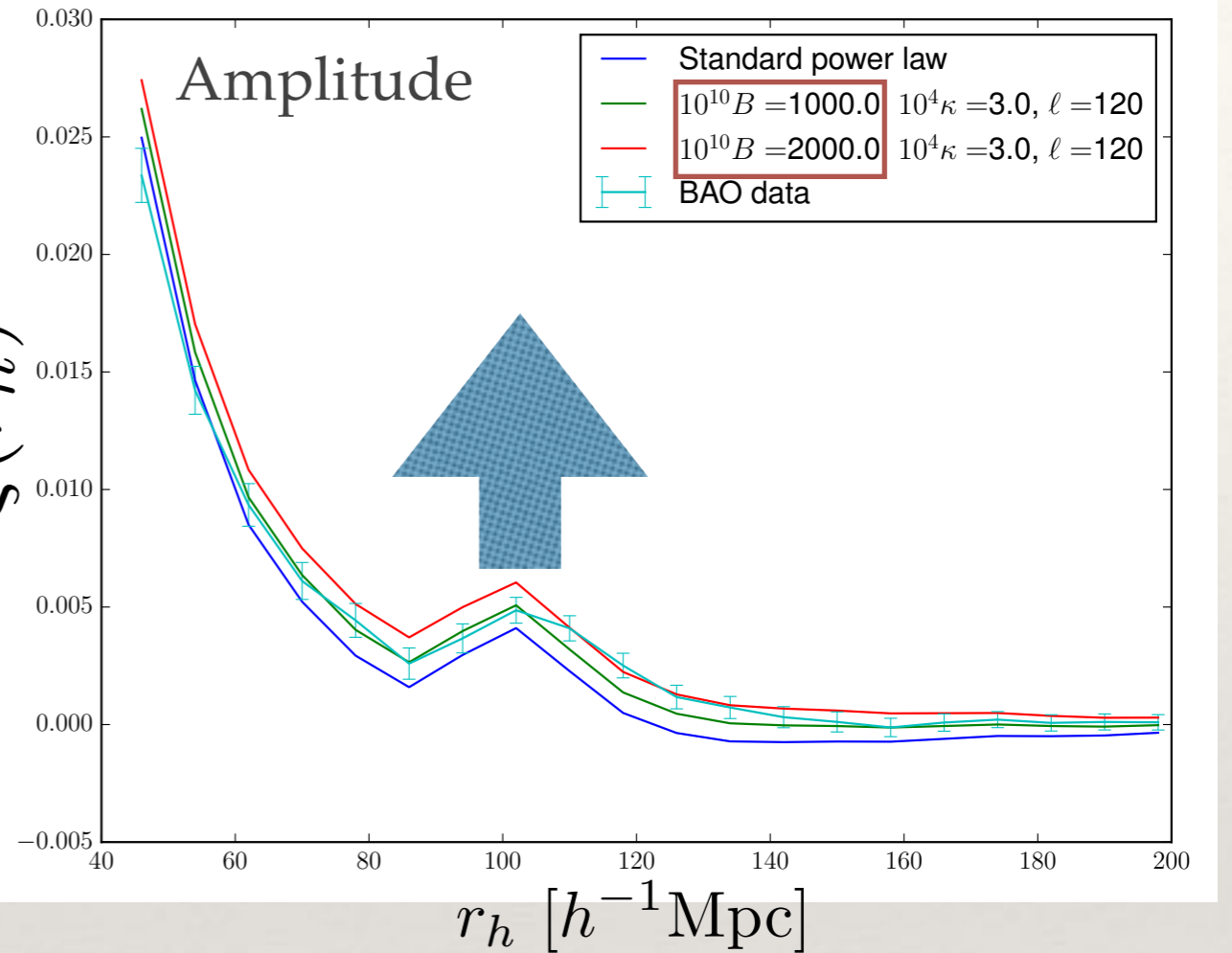


Amplitude at small scales becomes larger

Effects of features on BAO : Oscillating feature

(ii) Oscillating type : Amplitude & Width

$$P(k) = A \left(\frac{k}{k_0} \right)^{n_s-1} + B \left(\frac{k}{k_0} \right)^{n_s-1} \times \cos \left(\frac{\pi(k - k_*)}{\kappa} \right) \exp \left(-\frac{(k - k_*)^2}{\kappa^2} \right) \xi(r_h)$$



Almost the same effect
 Amplitude A : larger
 Width κ
 BAO amplitude : larger

Contents

Introduction

Motivation : CMB & BAO

Features in initial power spectrum

Effects of feature models on BAO

Results & discussion

Parameter estimation using MCMC method

Summary

Markov-Chain Monte-Carlo (MCMC) analysis

We analyze the feature parameters
by performing MCMC analysis.
Cosmological parameters : Planck Best fits.

MCMC analysis

Data :

BAO : two-points correlation function data (Anderson 2013 : CMASS DR11)



(i) Delta type feature

$$P_F(k) = \underline{B} k \delta(k - \underline{k_*})$$

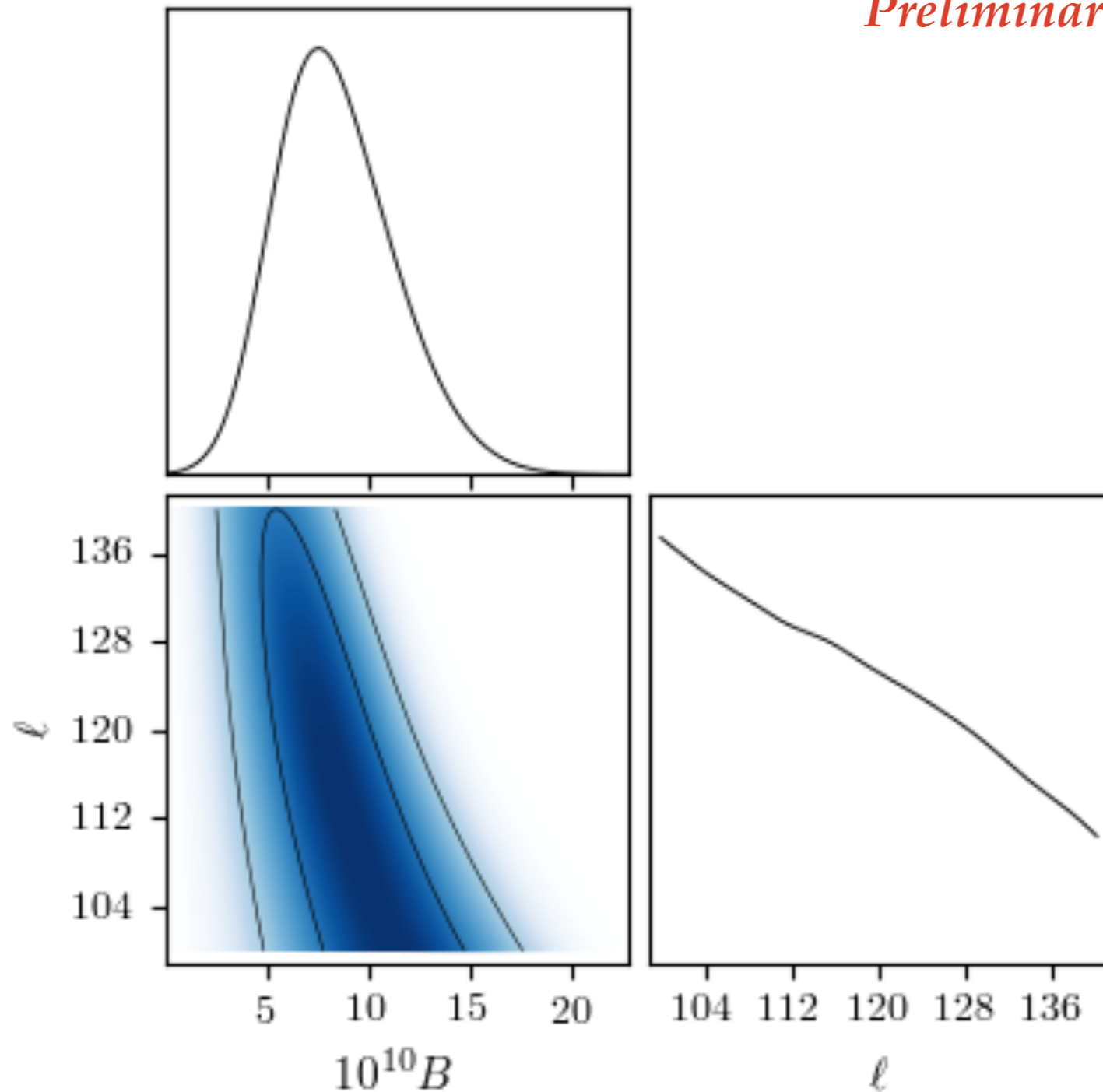
(ii) Oscillating type feature

$$P_F(k) = \underline{B} \left(\frac{k}{k_0} \right)^{n_s-1} \times \cos \left(\frac{\pi(k - \underline{k_*})}{\underline{\kappa}} \right) \exp \left(-\frac{(k - k_*)^2}{\kappa^2} \right)$$

Markov-Chain Monte-Carlo (MCMC) analysis

1. Delta type feature

Preliminary



Best fit parameters

$$10^{10}B = 8.03$$

$$k_* d_{ang} = 118.0$$

Best fit

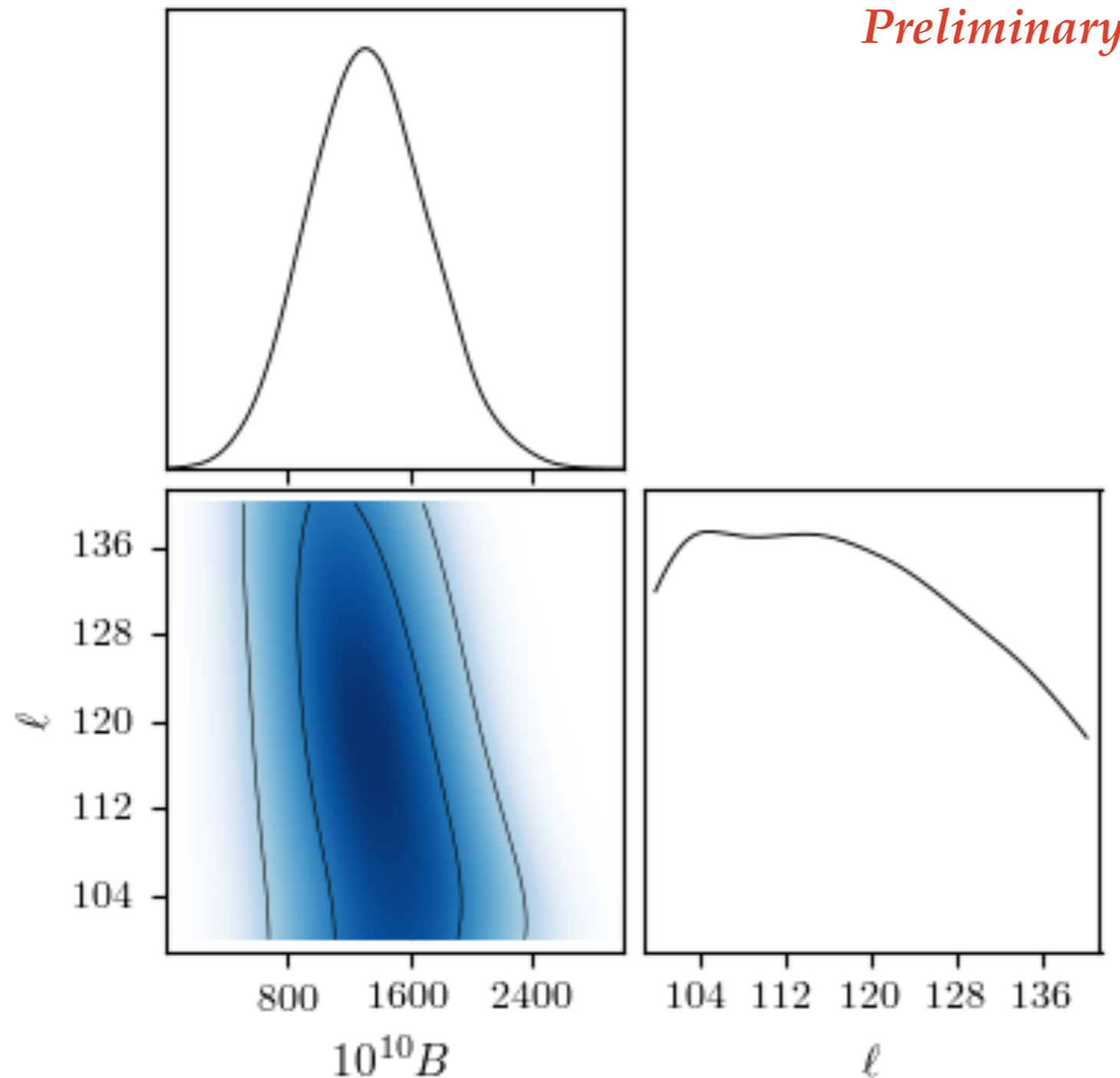
$$\chi^2 = 20.3$$

Ref :

Planck Best $\chi^2 = 32.8$

Markov-Chain Monte-Carlo (MCMC) analysis

2. Oscillating feature



Best fit parameters

$$10^{10} B = 1331$$

$$\kappa = 3.14 \text{ (Fixed)}$$

$$k_* d_{ang} = 118.6$$

Best fit

$$\chi^2 = 20.3$$

Ref :

$$\text{Planck Best } \chi^2 = 32.8$$

Comments

$$10^{10} B_{BAO} \sim 10^3 \gg 10^{10} B_{CMB} \sim 50$$

Ichiki, et al (2010)

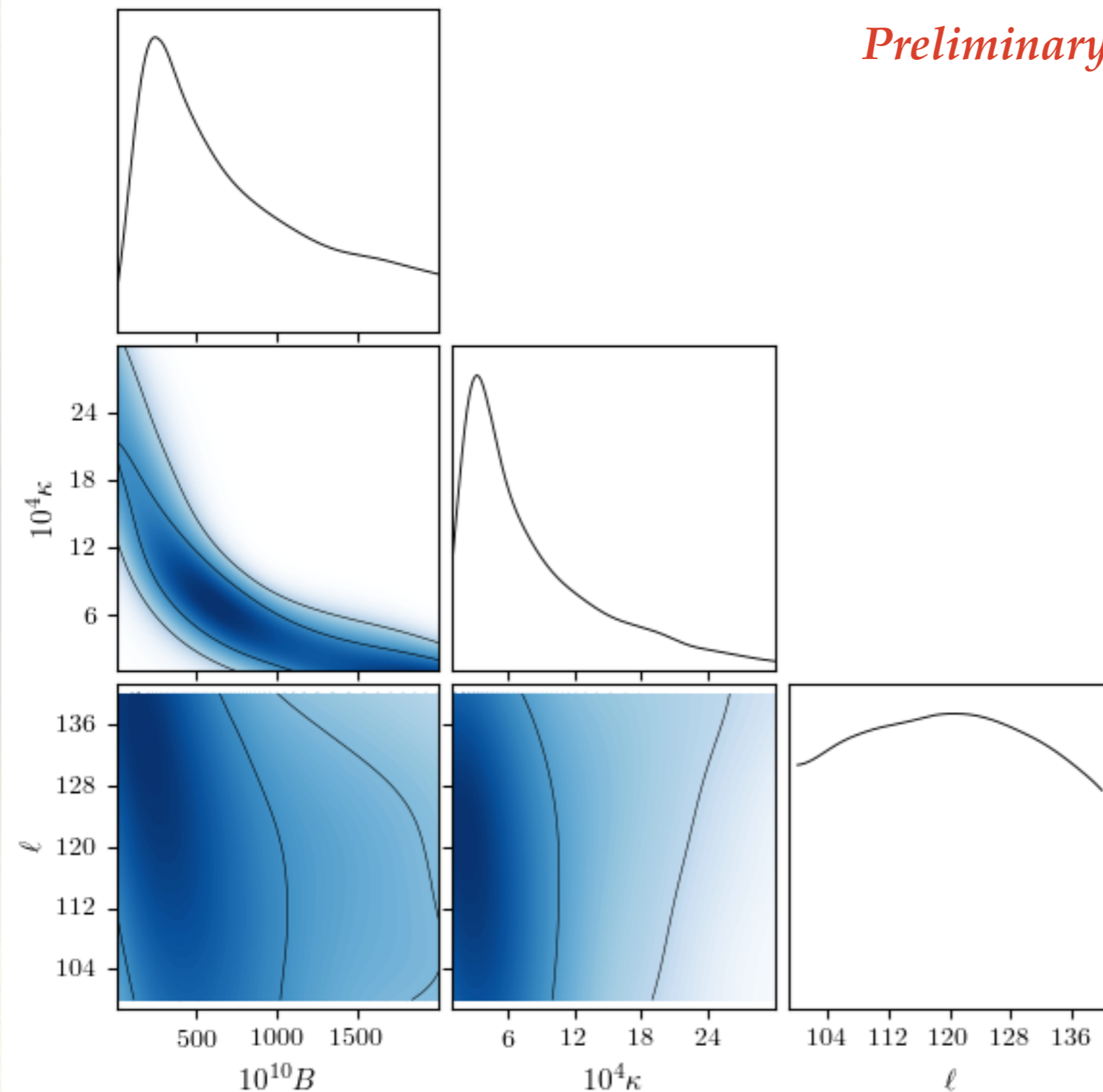
KH, et al (2017)

From CMB

Markov-Chain Monte-Carlo (MCMC) analysis

2. Oscillating feature

Preliminary



Best fit parameters

$$10^{10} B = 2000$$

$$\kappa = 2.11$$

$$k_* d_{ang} = 118.3$$

Best fit

$$\chi^2 = 20.3$$

Ref :

Planck Best $\chi^2 = 32.8$

Comments

$$10^{10} B_{BAO} \sim 10^3 \gg 10^{10} B_{CMB} \sim 50$$

Ichiki, et al (2010)

KH, et al (2017)

From CMB

Contents

Introduction

Motivation : CMB & BAO

Features in initial power spectrum

Effects of feature models on BAO

Results & discussion

Parameter estimation using MCMC method

Summary

Summary

- The effects of features in initial power spectrum on BAO
- Delta and oscillating type initial power spectrum models
- MCMC analysis using BAO data : features are preferred

Resulting parameters

Delta type

$$10^{10} B = 8.03$$

$$k_* d_{ang} = 118.0$$

Oscillating type (fixed width)

$$10^{10} B = 1331 (\gg 10^{10} B_{CMB} \sim 50)$$

$$k_* d_{ang} = 118.6$$

Next, we will focus on galaxy bias $P(k) \rightarrow (1 + b(k))P(k)$
and we will analyze the spectrum with the bias parameters