# Analysis of the response of void BAO to systematic effects in the SDSS observations using mock datasets

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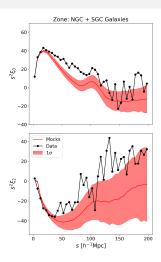
#### Introduction I

- The use of voids + matter ones  $\Rightarrow \sim 10\%$  improvement of error & 20% survey size increase.
- Voids could be less sensitive to systematic effects than matter.
- We want to use mocks to check if there is a difference in the BAO peak shift due to systematical effects between matter tracers and voids.
- Systematics can greatly affect the measurement of the 2PCF in real data.
- Sign of the quadrupole in large-scales.

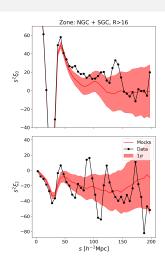




### Introduction II



Galaxy 2PCF from eBOSS ELG sample.



Void 2PCF from eBOSS ELG sample

### Voids I

Definition is controversial, we use a geometrical one.

Delaunay Triangulation  $\longrightarrow$  Voids are circumspheres in the simplices with tracers as vertices.

#### Small voids

- Radius cut:  $R_c < 8 h^{-1} \mathrm{Mpc}$
- $\blacksquare$  Located in regions with high  $\delta_{dm}$
- Correlated with galaxy sample
- Are actually Dark Matter

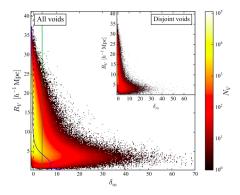
### Big voids

- Radius cut:  $R_c > 15.5 h^{-1} \mathrm{Mpc}$
- $\blacksquare$  Located in regions with low  $\delta_{dm}$
- Anticorrelated with galaxy sample
- Are actually empty





### Voids II



Number of voids,  $N_V$ , as a function of Dark Matter density,  $\delta_{dm}$ , and void radius,  $R_V$ . Taken from (Zhao et al., 2016)

#### How do we know?

- Large voids are constrained to regions with low matter density.
- Small void population spans a wide range of densities.





### Data I

We use 1000 **EZ mocks** Zhao and Chuang (2020).

- Displacement field from Zel'dovich
- PDF extraction from n-body simulation
- Halo position assignment

Table: Fiducial cosmology parameters used to produce the EZmocks used in this work.

Parameter	Value
$\Omega_m$	0.307115
$\Omega_b$	0.048206
h	0.6777
$\sigma_8$	0.8225
$n_s$	0.9611





### Systematics: Fiber collisions I

- Objects too close in the sky  $(r_{cp} < 62'')$  can't be seen at the same time due o width of the fiber.
- One is measured by the plate. The other one(s) could still be measured by other plate.
- Define  $TSR = \frac{\text{measured targets}}{\text{total number of targets}}$ , per sector.
- Compensate this effect with  $w_{cp} \approx \mathrm{TSR}^{-1}$ Actually  $w_{cp} = \frac{\mathrm{total\ number\ of\ targets}}{\mathrm{measured\ targets}}$  but defined per **collision group**.





### Systematics: Redshift Failures I

- Errors in the spectroscopic pipeline  $\Rightarrow$  SSR < 1.
- Two sources of error: Observational conditions & Position of fiber
- This is corrected by the weight

$$w_{noz} \equiv (SSR_{obs}SSR_{pos})^{-1}.$$





## Systematics: Angular Photometric I

- Represented as HEALPIX maps containing different photometric parameters,  $p_i$ .
- Parameters are combined as

$$y^k = \epsilon + \sum_i c_i p_i^k$$

where the model weights,  $c_i$  are optimized (per chunk) such that  $n_{\text{dat},k} \approx n_{\text{ran},k} y^k$ , where k indicates the pixels.

■ The photometric weights designed to partially correct for these effects are defined as

$$w_{\text{systot}} = \left(y^k\right)^{-1}$$
.





### Systematics: Normalization I

Completeness weights are defined as:

$$w_{\rm comp} = w_{\rm systot} w_{cp} w_{noz}.$$

- Some normalization is done (on  $w_{\text{systot}}$  before  $w_{\text{comp}}$  and on  $w_{noz}$  after).
- Invalid objects are set  $w_{cp}$ ,  $w_{noz} = 0$ .
- Only elements with  $SSR \ge 0$ ,  $z \in (0.6, 1.1)$  and completeness > 50% are selected.
- The dependence on n(z) is corrected by

$$w_{\text{FKP}} \equiv \frac{1}{1 + n(z)P_0}; \quad P_0 = 4000 \, h^{-3} \text{Mpc}^3.$$





# Systematics: Application I

■ We compute  $w_{\text{comp}}^{\text{ALLSYST}} = w_{\text{systot}} w_{cp} w_{noz} w_{\text{FKP}}$  and the effective number of tracers in each chunk

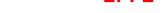
$$n_{\text{eff}} = \sum_{i=1}^{N_{chunk}} w_{\text{comp},i}^{\text{ALLSYST}},$$

where  $N_{chunk}$  is the number of tracers in the chunk considered.

We then compute

$$w_{\text{comp}}^{\text{PARTIAL}} = w_{\text{FKP}} \prod_{s \in \mathcal{S}'} w_s,$$





## Systematics: Application II

where  $S' \subseteq S$  is the subset of systematic effects to be considered and  $S = \{ \text{systot}, noz, cp \}$ . In this case we compute the new effective number of tracers

$$n'_{\text{eff}} = \sum_{i=1}^{N_{chunk}} w_{\text{comp},i}^{\text{PARTIAL}}.$$

- Objects in the catalog with  $w_{\text{comp}}^{\text{PARTIAL}} = 0$  or veto = 0 (such as objects outside the survey area or masked by bright stars) are removed.
- We then normalize the completeness weights in each chunk by the corresponding  $n_{eff}/n'_{eff}$  to keep the original effective number of tracers.





# Catalog generation I

Export catalogs with

RA, DEC, 
$$z$$
,  $w_{cp}w_{FKP}$ ,  $w_{cp}$ ,  $w_{FKP}$ ,  $n(z)$ 

■ Use DIVE to extract void catalogs with (we use  $\Omega_m = 0.31$ )

- Mask void catalogs
- Create void randoms
  - Combine 100 void mocks
  - Divide into z bins
  - divide each bin in R bins

- Split "vertically" into RA, DEC | z, R.
- Shuffle one of the two halves.
- Recombine halves and all bins.
- Randomly choose 2700 000 with  $R > R_c (= 15.5 \, h^{-1} {\rm Mpc})$

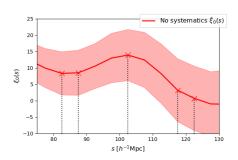




### Radius cut I

Analyze signal-to-noise ratio  $(\mathrm{SNR} \equiv \frac{\langle S \rangle}{\sigma_S})$  for 100 void mocks with different  $R_c$ . Use definition of signal, S, as in Liang, Zhao, Chuang, Kitaura, and Tao (2016).

$$\begin{split} S &= \xi_0(s^{\text{BAO}}) - \\ &\frac{\xi_0(s_1^{\text{dl}}) + \xi_0(s_2^{\text{dl}}) + \xi_0(s_1^{\text{dr}}) + \xi_0(s_2^{\text{dr}})}{4} \end{split}$$

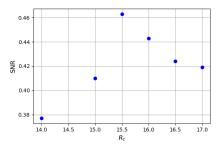


$$\begin{split} s_1^{\rm dl} &= 82.5, \; s_2^{\rm dl} = 87.5, \; s_1^{\rm BAO} = 102.5, \\ s_1^{\rm dr} &= 117.5 \; \text{and} \; s_2^{\rm dr} = 122.5 \, h^{-1} \text{Mpc} \\ &\text{respectively from left to right.} \end{split}$$





### Radius cut II



SNR vs low radius cut  $R_c$ . Upper cut is always set to 50  $h^{-1}{\rm Mpc}$ . Maximum SNR is obtained for  $R_c=15.5\,h^{-1}{\rm Mpc}$ .





### The BAO model I

We use the model in Zhao et al. (2018):

$$\xi_t(s) = \int \frac{k^2 dk}{2\pi^2} \frac{\sin ks}{ks} P_t(k) \exp(-k^2 a^2), \quad a = 1 h^{-1} \text{Mpc}$$
 (2)

$$P_t(k) = \left\{ \left[ P_{\text{lin}}(k) - P_{\text{nw}}(k) \right] \exp\left(\frac{-\Sigma_{\text{nl}}^2 k^2}{2}\right) + P_{\text{nw}}(k) \right\} \frac{P_{\text{t,nw}}(k)}{P_{\text{lin,nw}}(k)}, \quad (3)$$

$$\xi_{\text{model}}(s) \equiv B^2 \xi_t(\alpha s) + A(s), \quad A(s) = \frac{a_1}{s^2} + \frac{a_2}{s} + a_3$$
 (4)

$$\begin{array}{c|c} & \mathsf{Galaxies} & \mathsf{Voids} \\ \frac{P_{\mathrm{t,nw}}(k)}{P_{\mathrm{lin,nw}}(k)} & 1 & 1 + ck^2 \end{array}$$



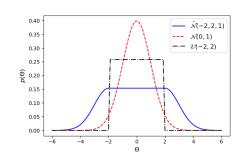


## Parameter fitting I

Bayesian inference for parameter  $\Theta$ :

$$p(\Theta|X) = \frac{p(X|\Theta)p(\Theta)}{p(X)}$$

$$= \frac{\mathcal{L}(X|\Theta)p(\Theta)}{\mathcal{Z}},$$
(5)



Examples of the different kinds of priors used in the fitting.





# Parameter fitting II

#### Voids

$$p(\Sigma_{\rm nl}) = \mathcal{U}(0, 20) \tag{6}$$

$$p(B) = \mathcal{N}(2, 0.15) \tag{7}$$

$$p(\alpha) = \mathcal{U}(0.8, 0.12) \tag{8}$$

$$p(c) = \bar{\mathcal{N}}(-500, 1000, 100)$$
 (9)

#### **Galaxies**

$$p(\Sigma_{\rm nl}) = \mathcal{U}(5, 17) \tag{10}$$

$$p(B) = \bar{\mathcal{N}}(1.4, 1.6, 0.12)$$
 (11)

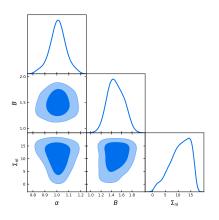
$$p(\alpha) = \mathcal{U}(0.8, 0.12) \tag{12}$$



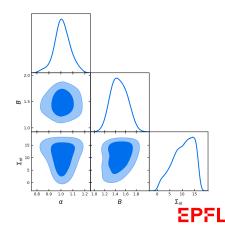


### Results: Mean 2PCF Galaxies I

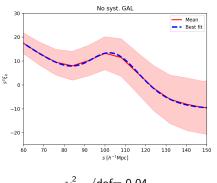
### No systematics



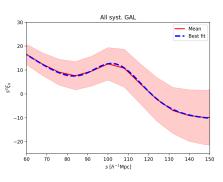
#### **All systematics**



### Results: Mean 2PCF Galaxies II



$$\chi^2_{
m best}/{
m dof} = 0.04$$



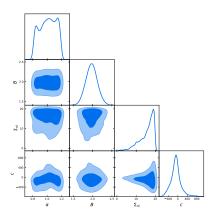
$$\chi^2_{
m best}/{
m dof}=0.06$$



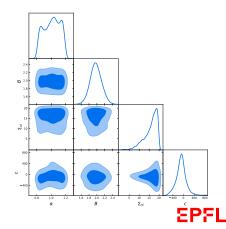


### Results: Mean 2PCF Voids I

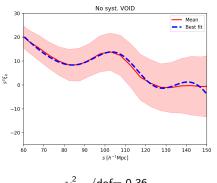
### No systematics



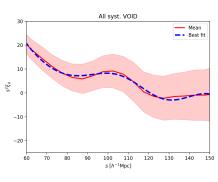
#### All systematics



### Results: Mean 2PCF Voids II



$$\chi^2_{\rm best}/{\sf dof} = 0.36$$



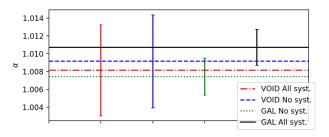
$$\chi^2_{\rm best}/{\sf dof} = 0.42$$





## Results: Mean2PCF Comparison I

$$\begin{array}{cc} \text{Galaxies} & \text{Voids} \\ \alpha_{\rm all}-\alpha_{\rm none} & (3.34\pm2.89)\times10^{-3} & (-9.9\pm72.9)\times10^{-4} \end{array}$$

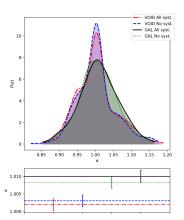


Fit results for the dilation parameter  $\alpha$  when using the mean of the mocks.



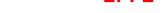
### Results: Individual 2PCF I

$$\alpha_{\rm all}-\alpha_{\rm none} \quad \mbox{(1.89} \pm 2.39) \times 10^{-3} \quad \mbox{(-1.07} \pm 2.39) \times 10^{-3}$$



Fit results for the  $\alpha$  parameter from fitting each of the mocks individually. Top panel shows the distributions obtained. Bottom panel shows the mean values and the (scaled) standard deviation as error.





### Conclusions I

- The fit to the mean 2PCF shows a smaller shift due to systematics in the void case.
- The posteriors  $p(\alpha|X)$  in the void case have large widths that make it difficult to be confident in the conslusion above.
- The individual fits show smaller uncertainty in the void measurement and still show the difference in the peak shift.
- Priors can be tweaked to get a better  $p(\alpha|X)$  for voids but reduced  $\chi^2$  is already close to 1.
- Dataset is noisy as shown by the low SNR
- The analysis should be repeated with other tracers with better SNR (e.g. LRG)





### References I

- Liang, Y., Zhao, C., Chuang, C.-H., Kitaura, F.-S., & Tao, C. (2016). Measuring baryon acoustic oscillations from the clustering of voids. *Monthly Notices of the Royal Astronomical Society*, 459(4), 4020–4028. doi:10.1093/mnras/stw884
- Zhao, C. & Chuang, C.-h. (2020). The SDSS-IV Extended Baryon Oscillation Spectroscopic Survey: mock catalogues for the eBOSS final Data Release. *in preparation*, 10(December 2019), 1–10.
- Zhao, C., Chuang, C.-h., Kitaura, F.-s., Liang, Y., Pellejero-Ibanez, M., Tao, C., ... Yepes, G. (2018). Improving baryon acoustic oscillation measurement with the combination of cosmic voids and galaxies. 19(February), 1–19. arXiv: 1802.03990. Retrieved from http://arxiv.org/abs/1802.03990



### References II

Zhao, C., Tao, C., Liang, Y., Kitaura, F.-S., & Chuang, C.-H. (2016). DIVE in the cosmic web: voids with Delaunay triangulation from discrete matter tracer distributions. Monthly Notices of the Royal Astronomical Society, 459(3), 2670–2680. doi:10.1093/mnras/stw660



