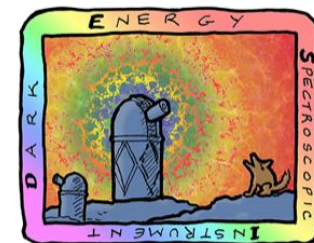


One- and three-dimensional measurements of the matter distribution from eBOSS and first DESI Lyman- α forest samples

Corentin Ravoux

*Euclid France Galaxy Clustering
24 November 2022*

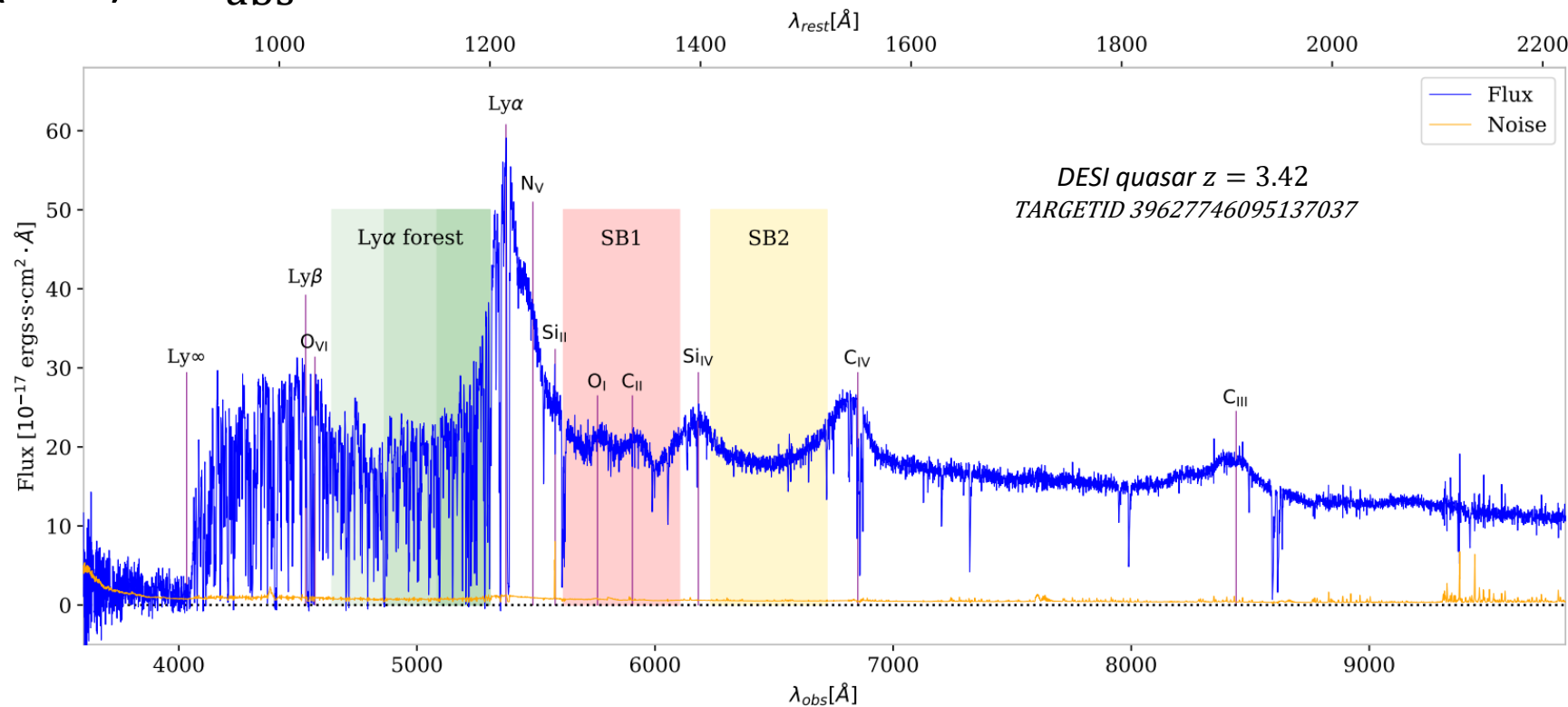


**DARK ENERGY
SPECTROSCOPIC
INSTRUMENT**

U.S. Department of Energy Office of Science

The Lyman- α forest

- Lyman- α : transition of neutral Hydrogen to first excited state $\lambda_{\alpha} = 1215.67 \text{ \AA}$
- Lines in quasar spectra at $\lambda_{\text{obs}} = (1 + z_{\text{abs}})\lambda_{\alpha}$ caused by absorber in the intergalactic medium (IGM) at z_{abs}



Lyman- α forest = Non-linear tracer of the neutral Hydrogen in the IGM

Definition of observational statistics

- Lyman- α contrast: Quasar continuum fitted to normalize absorptions in the flux

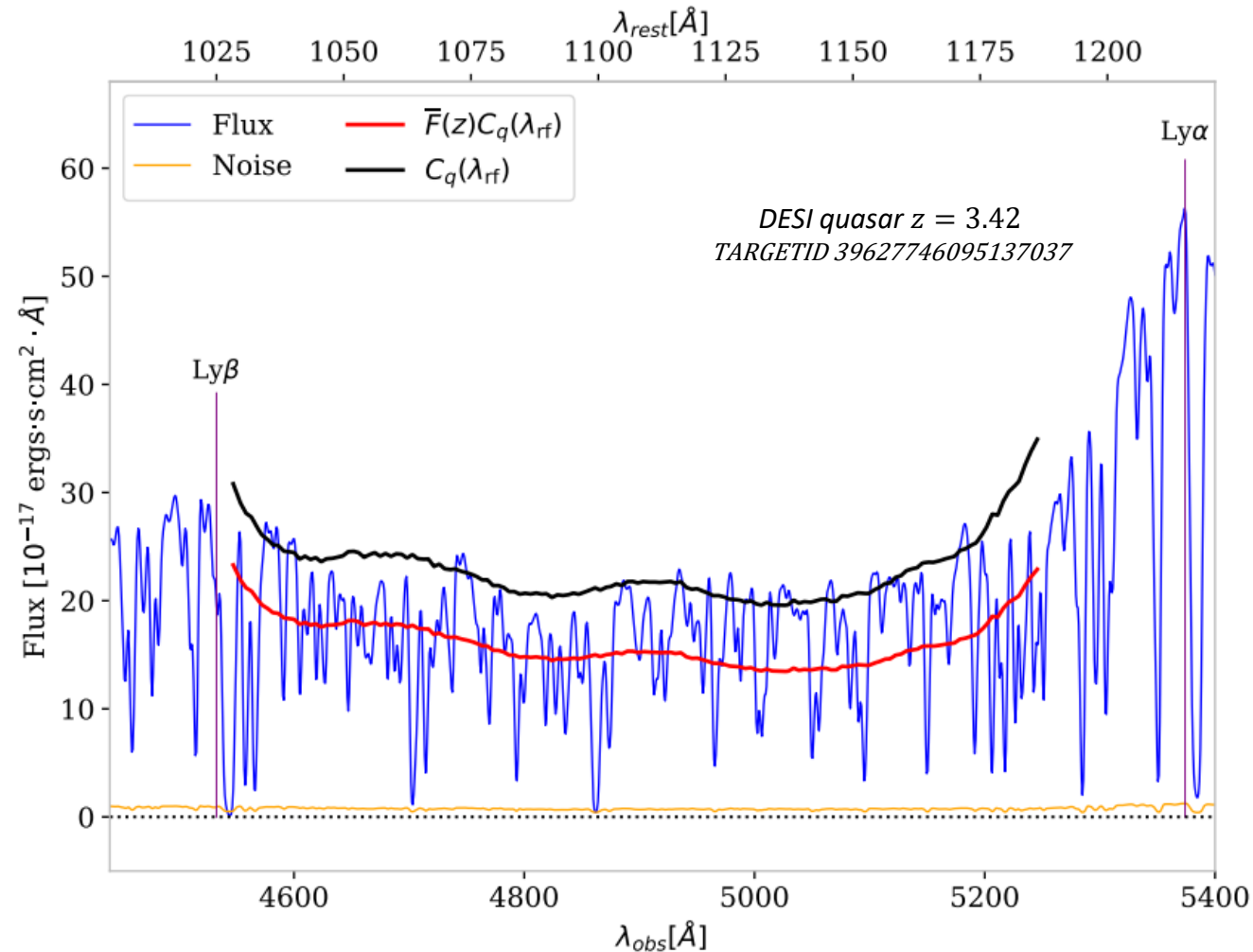
$$\delta_F(\lambda) = \frac{f(\lambda)}{\bar{F}(\lambda)C_q(\lambda, z_q)} - 1$$

- Cross-correlation with a tracer X:

$$\xi_{\alpha X}(\vec{r}) = \langle \delta_F(\vec{x})\delta_X(\vec{x} + \vec{r}) \rangle_x$$

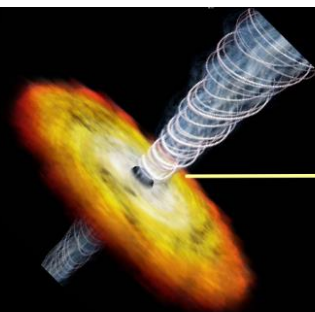
- One dimensional power spectrum:

$$P_{1D,\alpha}(k) = \langle |\delta_F(k)|^2 \rangle$$

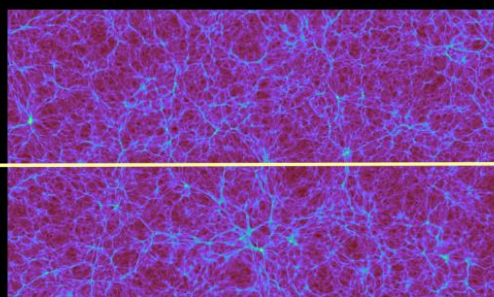


Contaminants

- Near the quasar:
 - Intrinsic continuum
 - Broad absorption line quasars (BAL)
- Along the line-of-sight:
 - Metal absorptions in the IGM
 - Damped Lyman- α systems (DLA)
- Near the telescope:
 - Atmospheric emission lines
 - Instrument noise
 - Spectrograph resolution



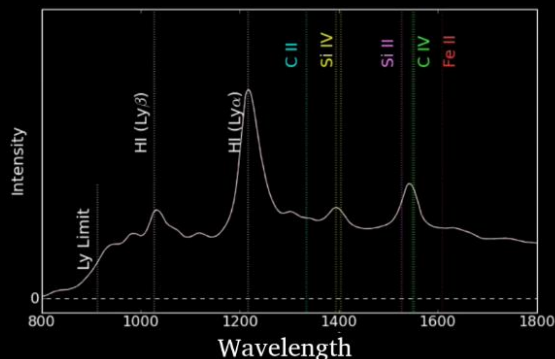
Quasar



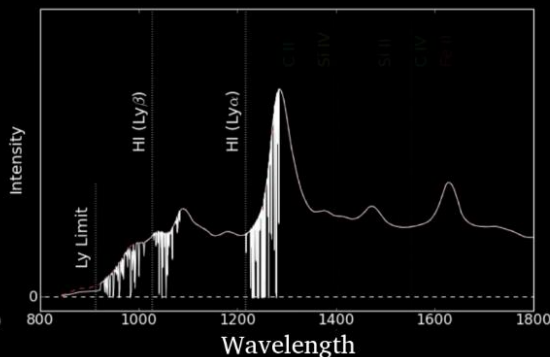
Intergalactic medium



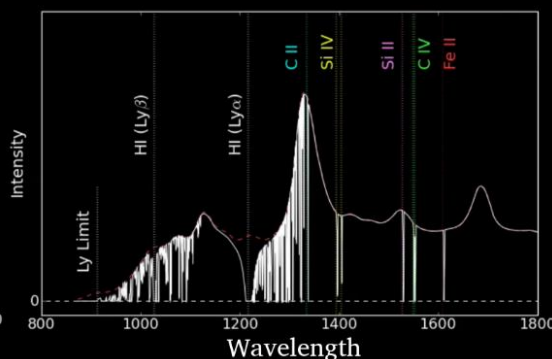
Circumgalactic medium



Quasar emission



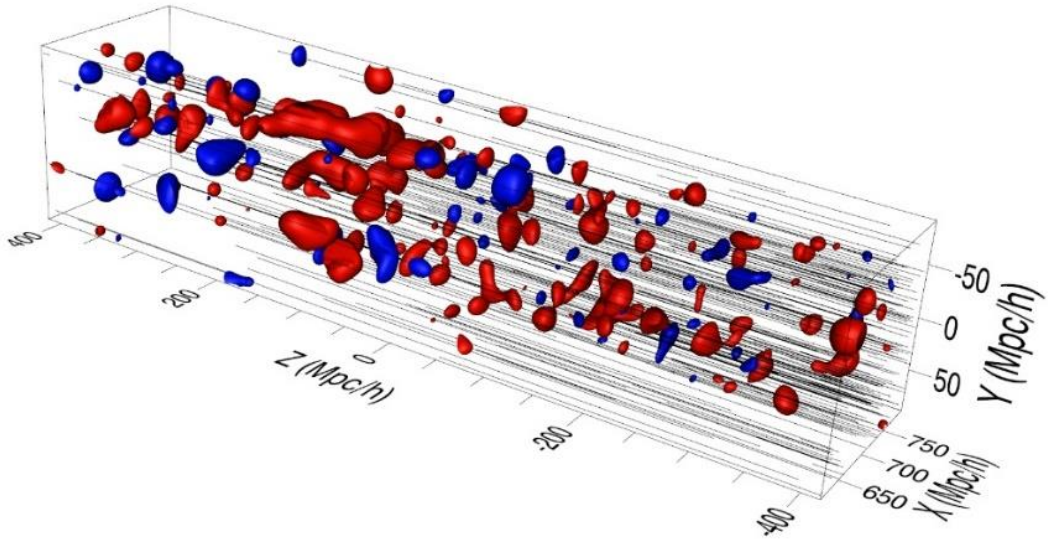
Lyman-alpha absorption



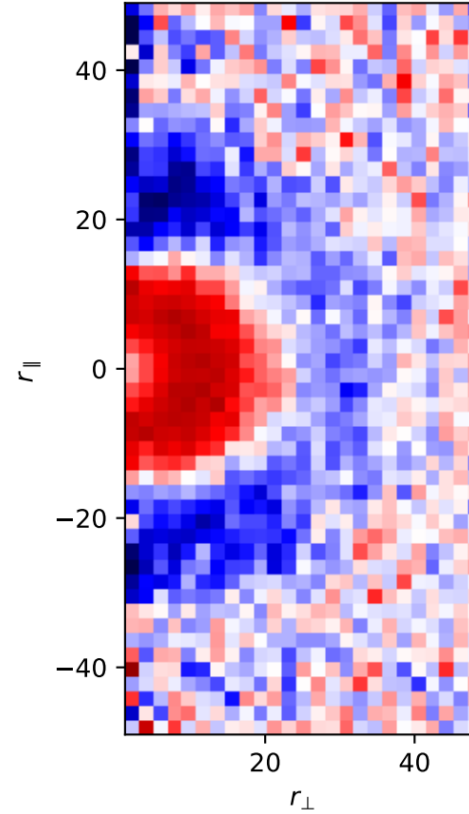
HCD absorption

Adapted from A. Pontzen video

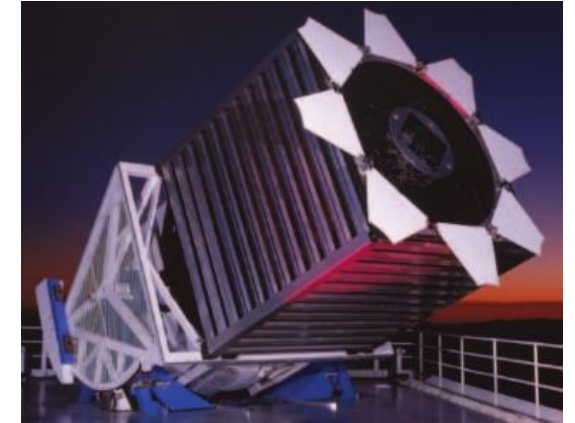
Lyman- α tomography and voids



Ravoux et al. JCAP07(2020)010

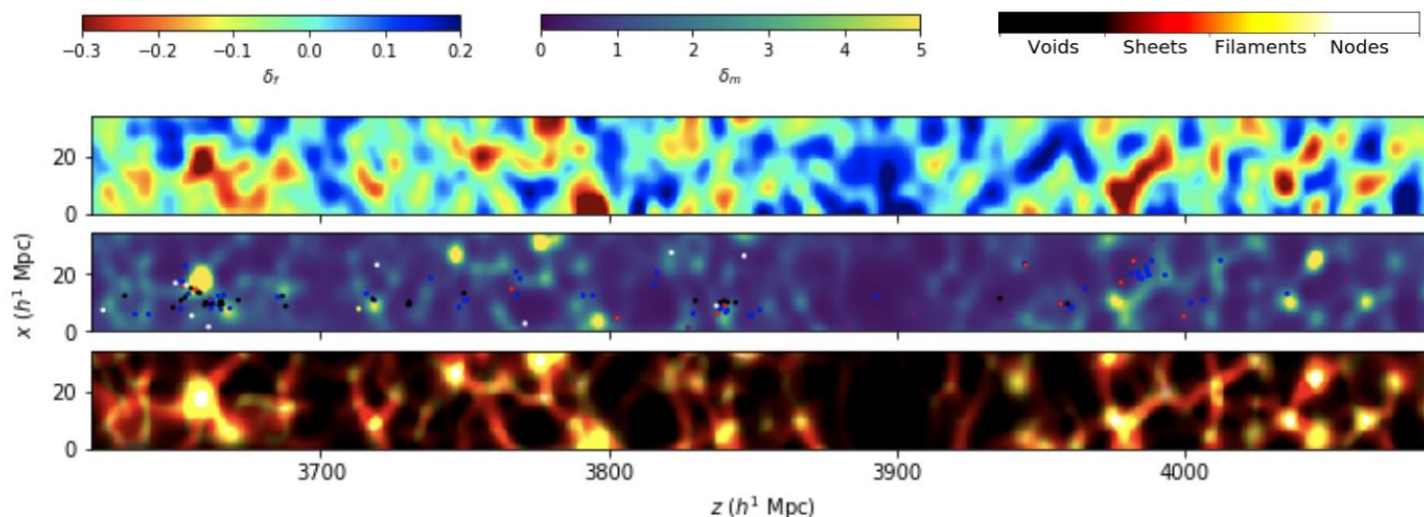


Ravoux et al. 2022



Lyman- α tomography

- 3D map of Lyman- α absorption from 1D spectra (*Pichon et al. 2001*)
- Initial goal: map the cosmic web \sim Mpc scale

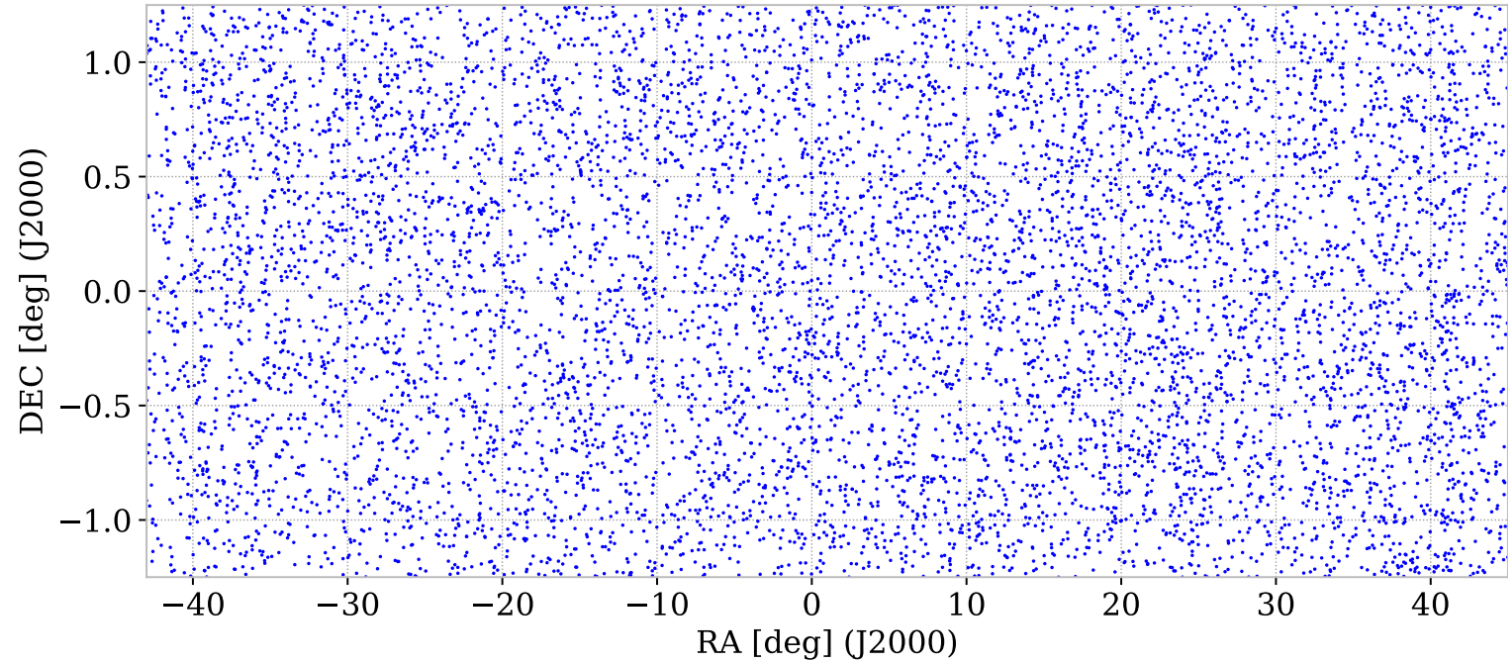


- Example: CLAMATO
(*Lee et al. 2018*)

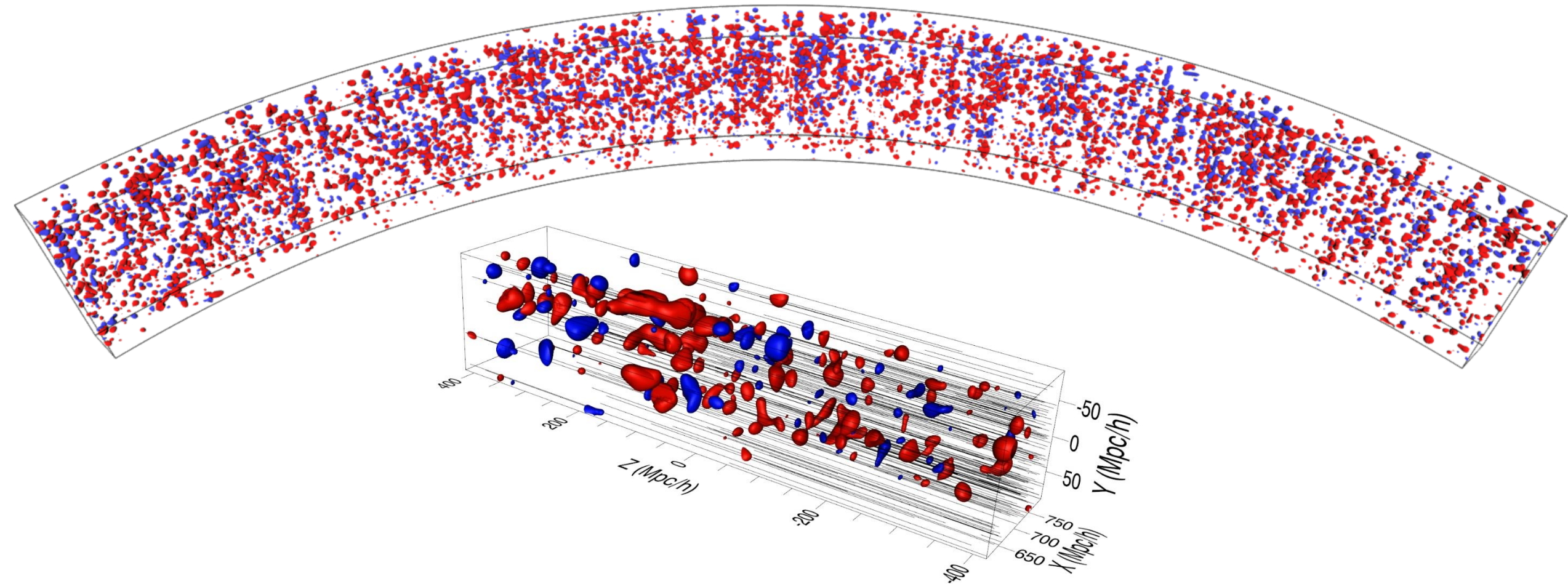
- Our objective: produce 3D map at large scales for large volumes ($\sim \text{Gpc}^3 \cdot h^{-3}$).
 - Need large volume surveys, with lower density of targets
 - Use of eBOSS DR16 data

Large-scale Lyman- α tomography with eBOSS

- Stripe 82 data:
 - Dense and homogeneous field
 - Mean separation $13 h^{-1} \cdot \text{Mpc}$
- Use of CLAMATO Wiener filter algorithm:
 - Noise-dependent interpolation of lines-of-sight with Gaussian kernels
 - Reconstruction length $13 h^{-1} \cdot \text{Mpc}$



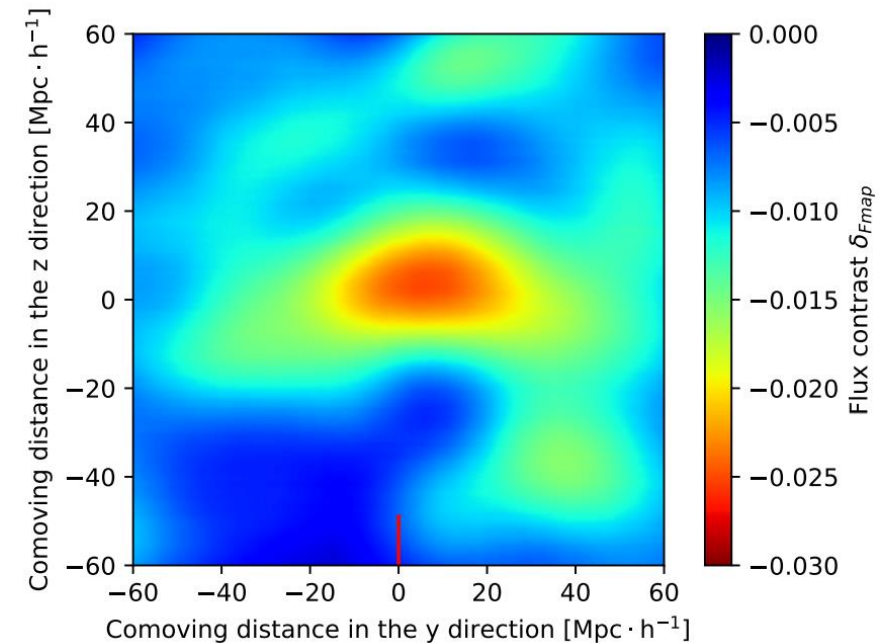
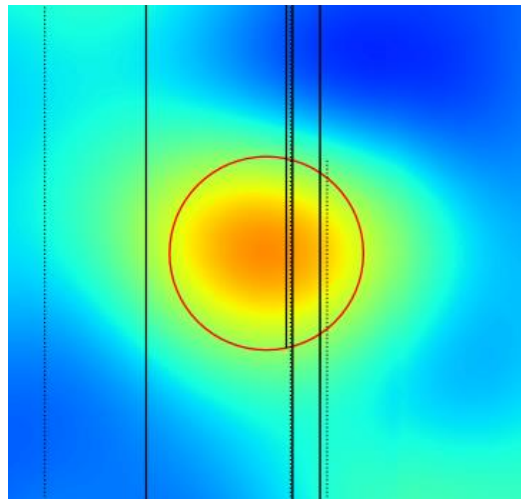
3D representation



Largest 3D map tracing matter at redshift $z > 2$

Applications of Lyman- α tomography

- Average map of the reconstructed Lyman- α contrast around quasar positions:
 - Recast 3D view of cross-correlation between Lyman- α forest and quasars



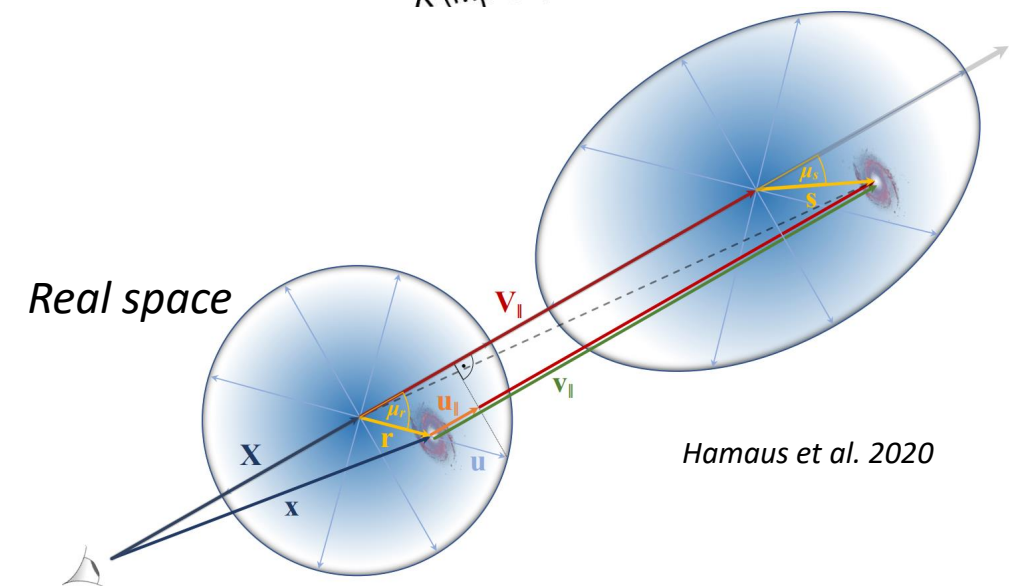
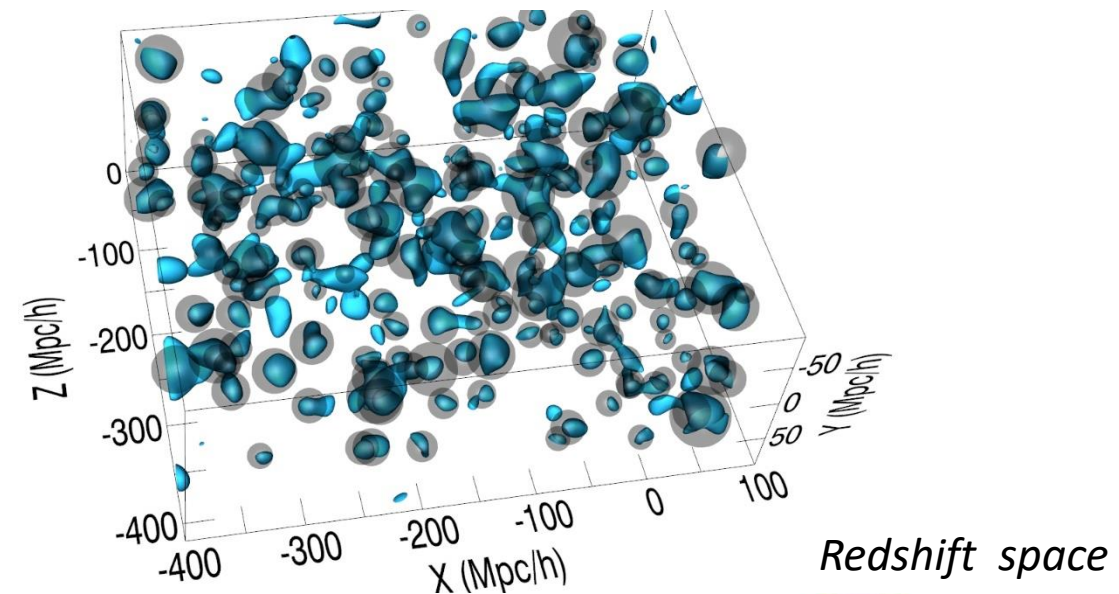
- Proto-cluster candidates identification:
 - Selection on Lyman- α contrast threshold and number of crossed lines-of-sight
 - **Identification of 8 proto-cluster candidates over Stripe 82**

High redshift voids

- Cosmic voids = 80 % in volume of the cosmic web
- Implementation of a 3D multi-threaded spherical void finder

First large void catalog at redshift $z > 2$

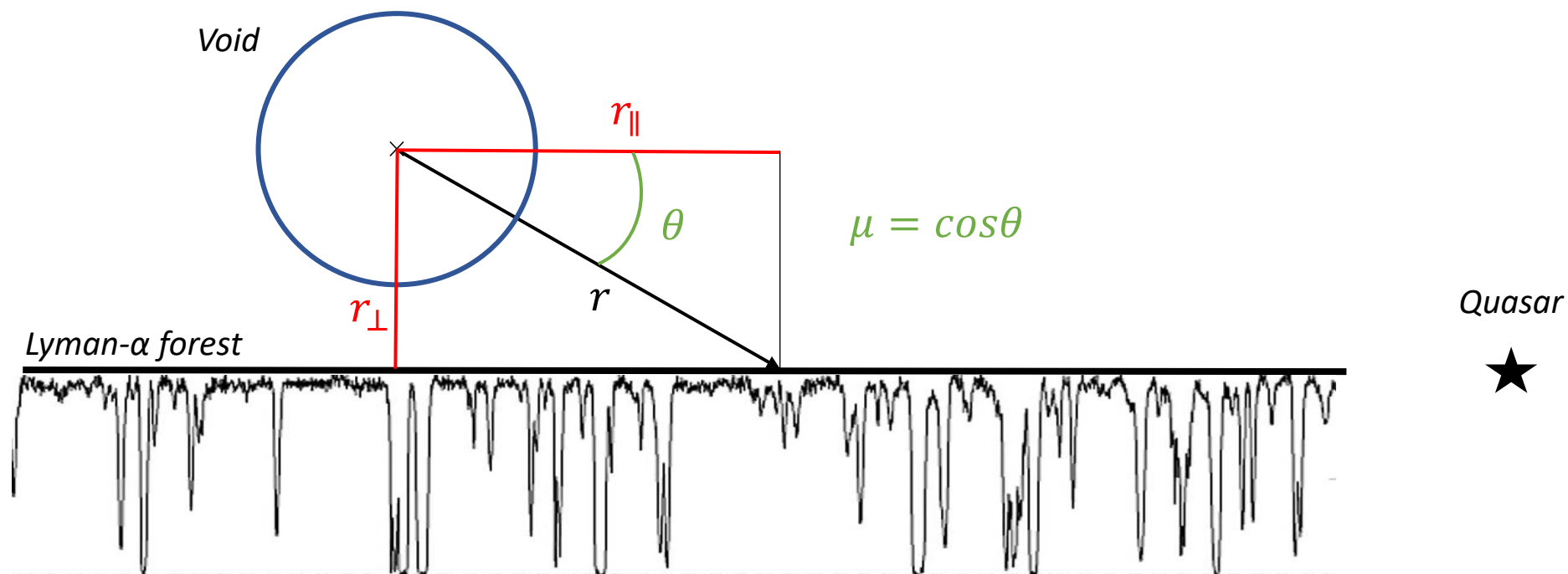
- Our objectives:
 - Study the shape of our voids and extend galaxy void analysis to redshift $z > 2$
 - Measure velocity flow around voids (redshift space distortions)



Lyman- α x void cross-correlation

- Cross-correlation between void centers and Lyman- α flux contrast:

$$\xi_{v\alpha} \left(A \equiv (r_{\perp}, r_{\parallel}) \right) = \frac{\sum_{(i,j) \in A} w_i \delta_{F,i}}{\sum_{(i,j) \in A} w_i}$$

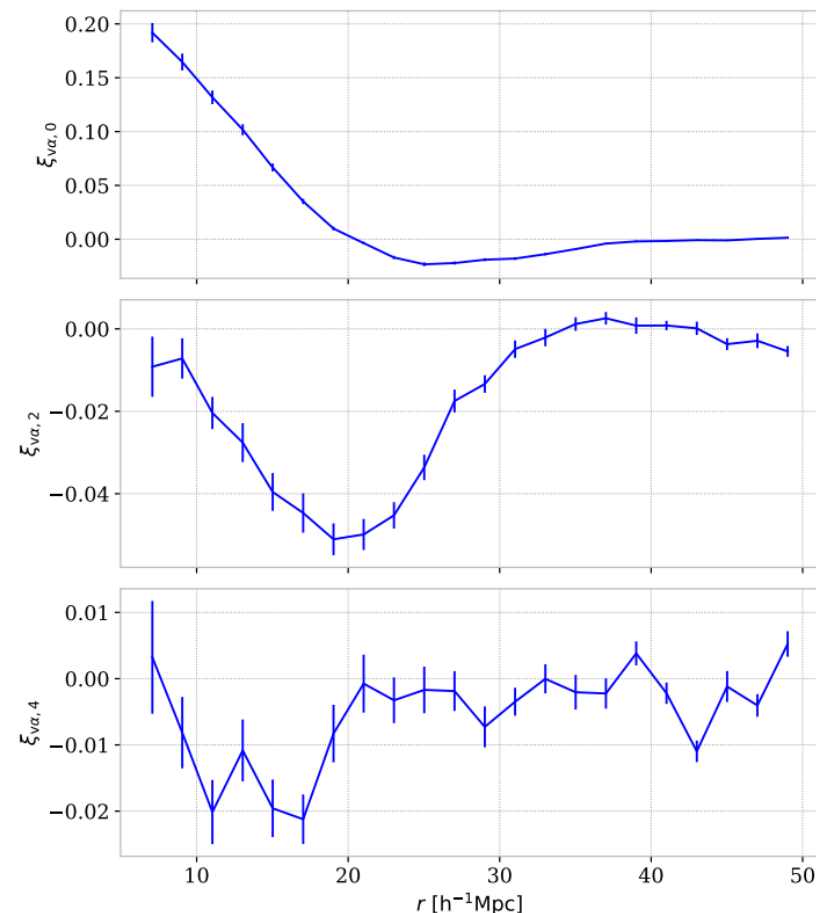
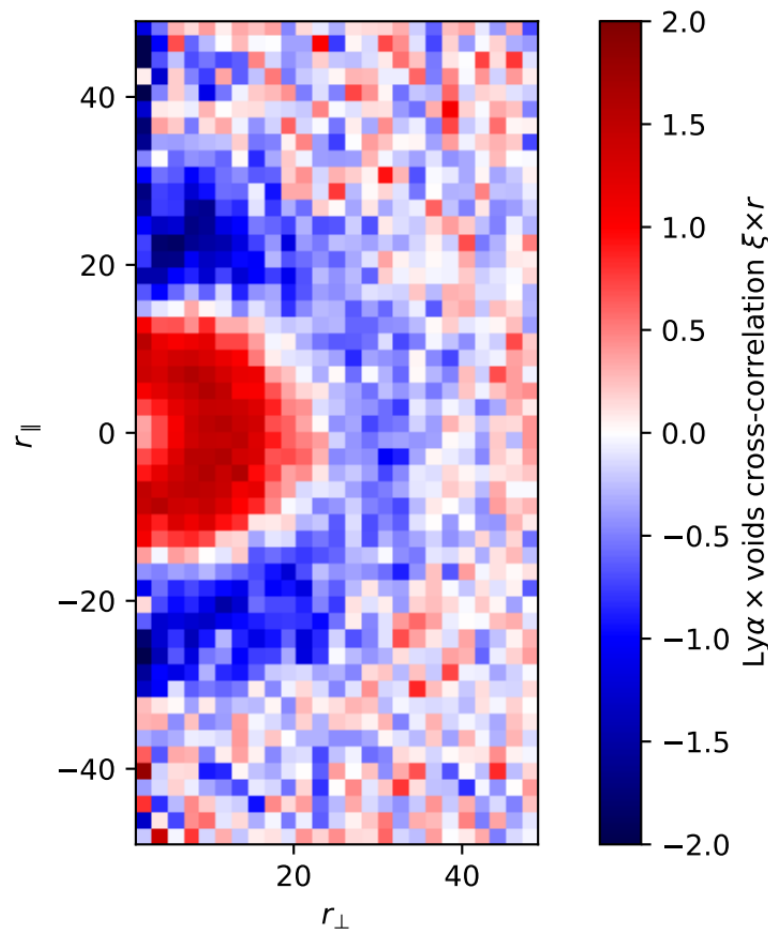


Measurement on eBOSS data

- Measure $\xi_{v\alpha}(r, \mu)$
- Decomposition into multipoles on Legendre basis:

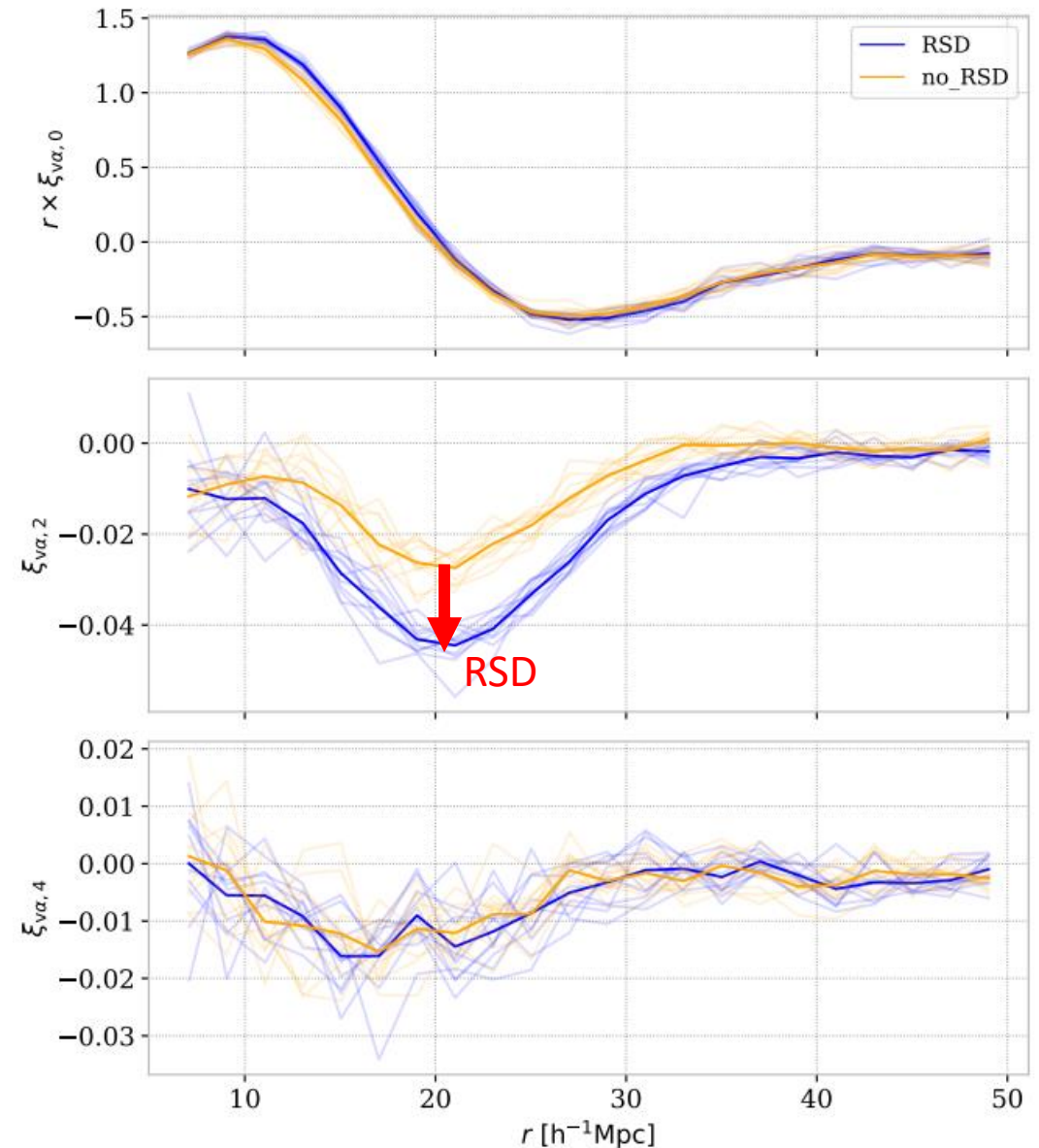
$$\xi_\ell(r) = \int_{-1}^1 \xi(r, \mu) \left(\frac{1 + 2\ell}{2} \right) P_\ell(\mu) d\mu$$

- ξ_0 = Average void profile
- ξ_2 = Measure departure from spherical symmetry, contains RSD signal



Study of cross-correlation with mocks

- Systematic effects:
 - Realizations on the same matter field adding systematics (Continuum fitting, Noise, Metals, DLA)
 - Relatively minor impact on the quadrupole which contains RSD signal
- Impact of RSD:
 - Mocks with and without RSD effect
 - Used 11 realizations to reduce statistical uncertainties
- **RSD impact seen on quadrupole**



RSD parameter measurement

- Adaptation of a linear void model to Lyman-alpha forest:

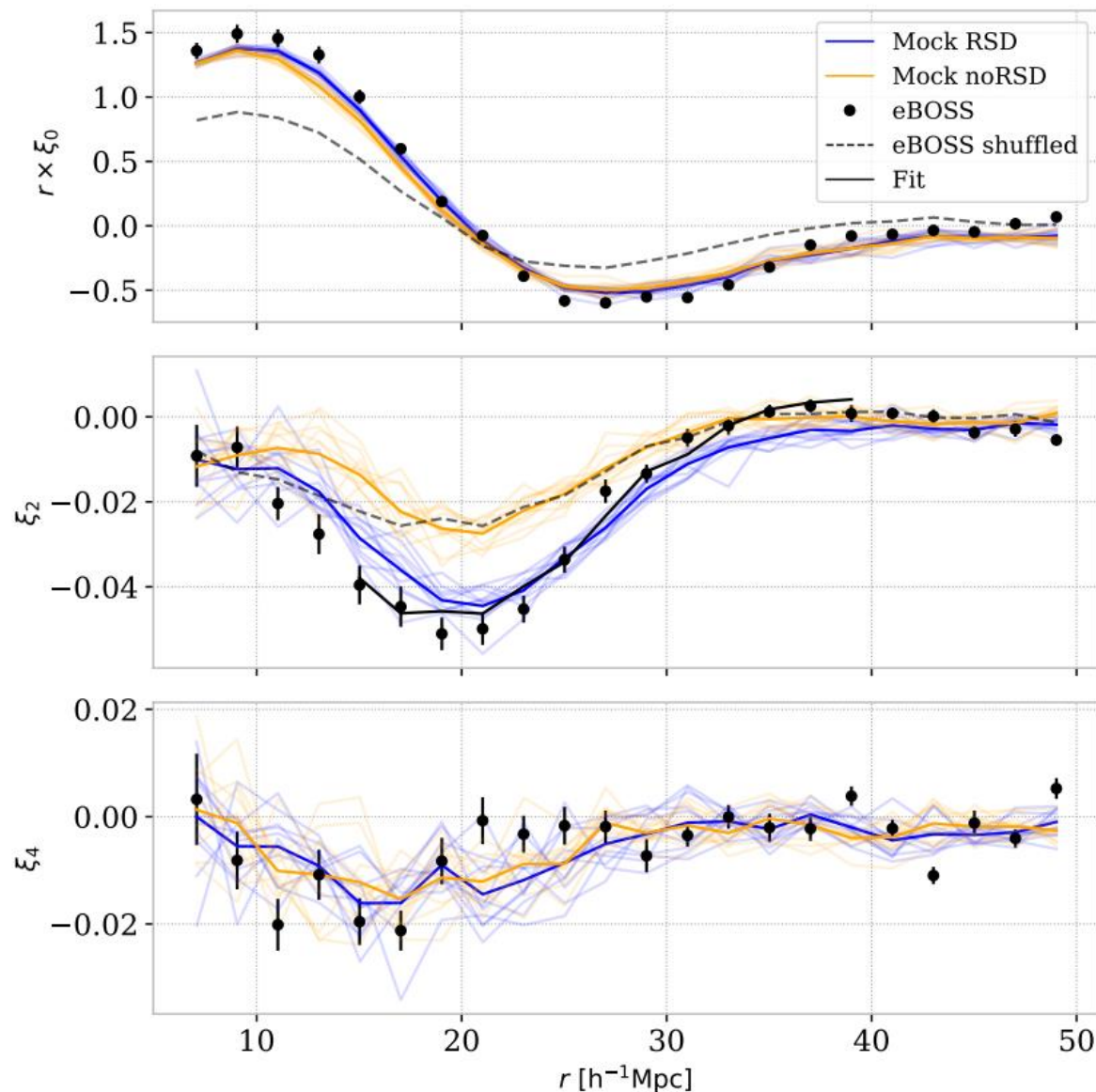
$$\xi_2(r) = \left(\frac{2\beta}{3 + \beta} \right) (\xi_0(r) - \bar{\xi}_0(r))$$

- On eBOSS data:

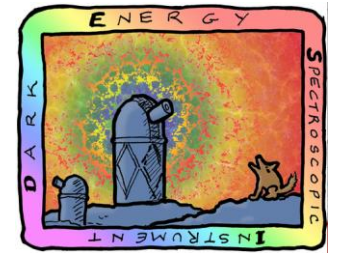
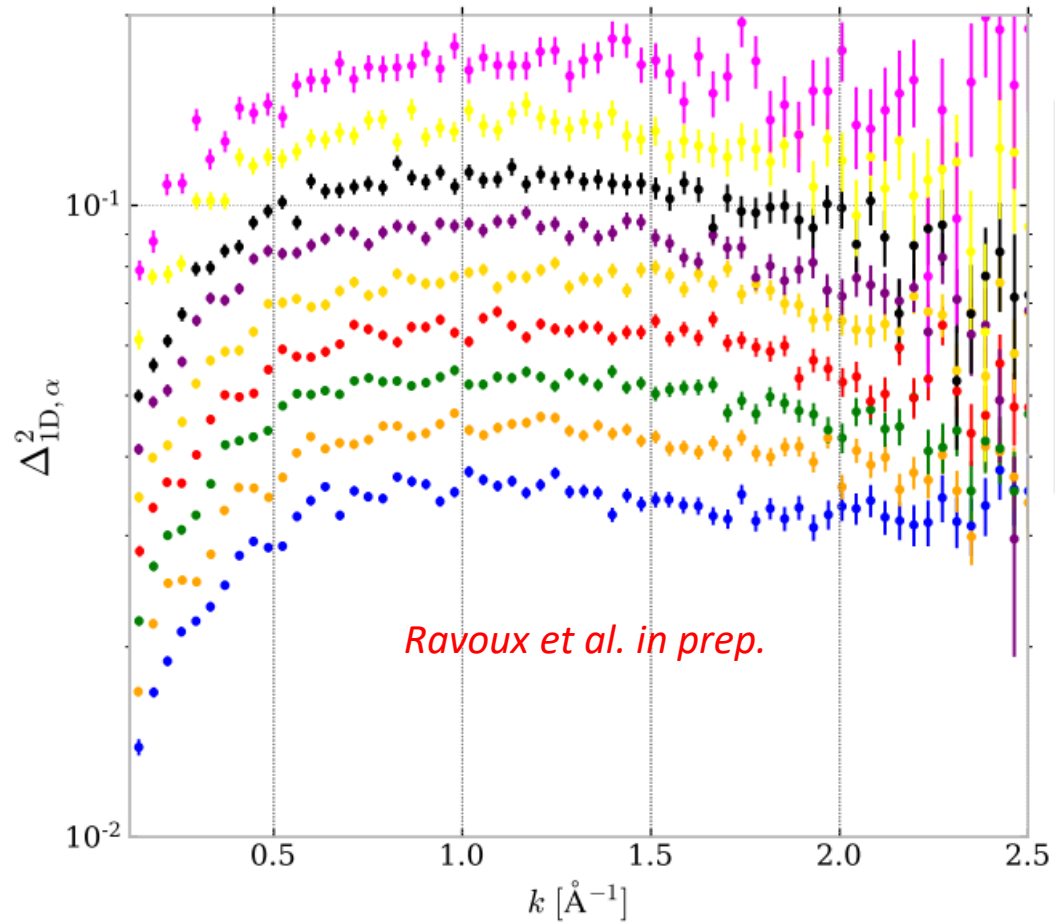
$$\beta = 0.52 \pm 0.05$$

First measurement of velocity flow around voids at $z > 2$

- Full interpretation of this value requires additional studies with cosmological simulations

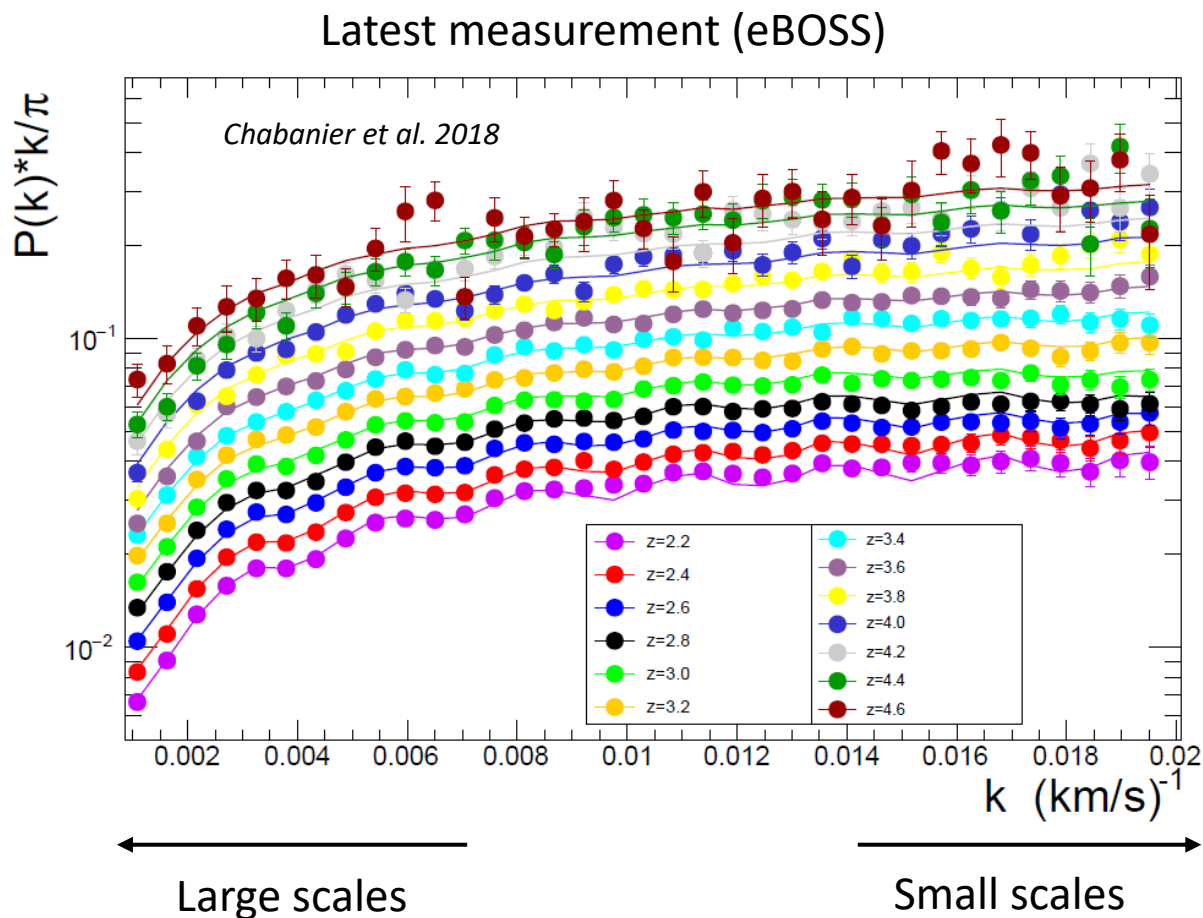


One dimensional power spectrum



$P_{1D,\alpha}$ measurement

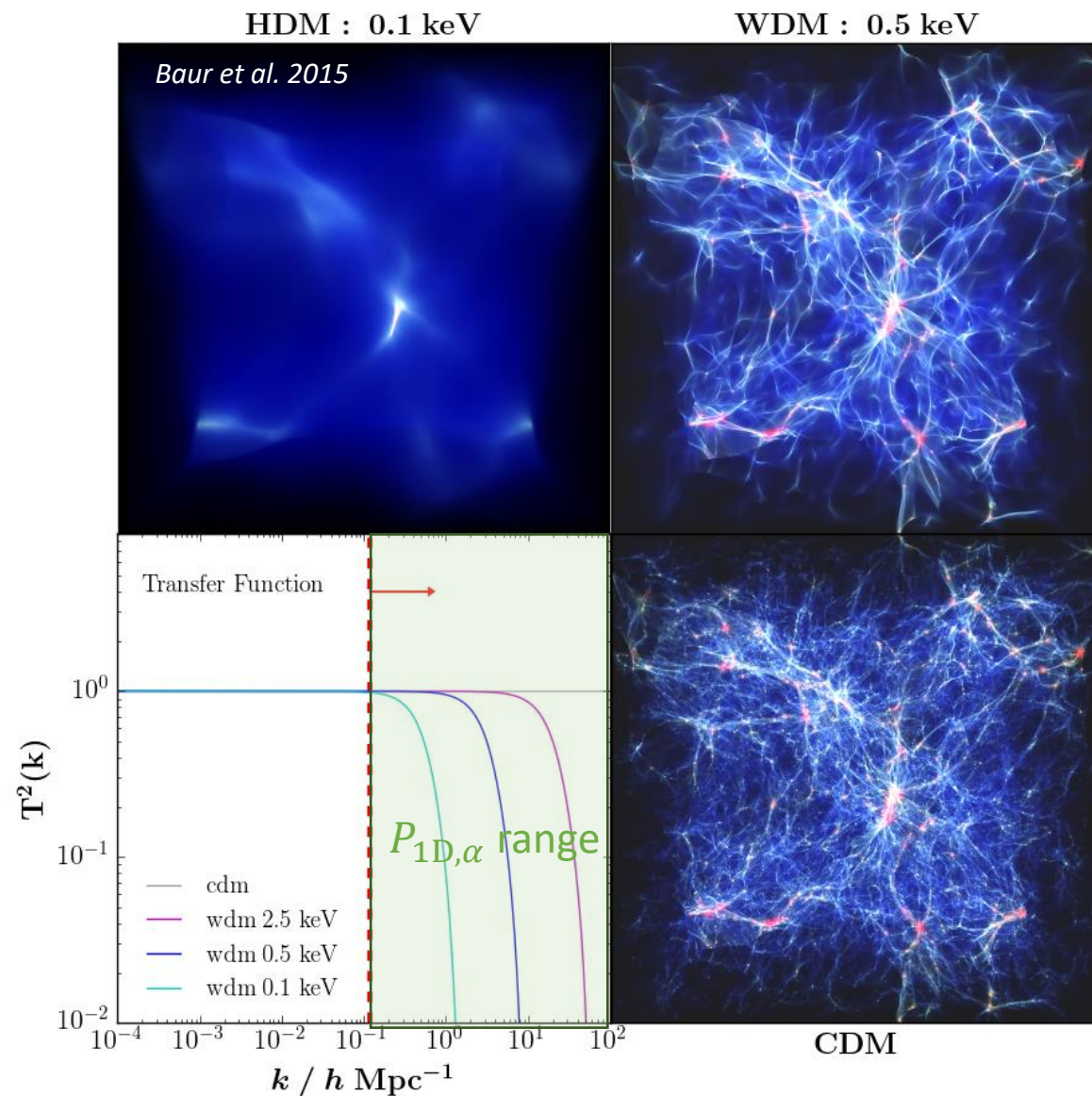
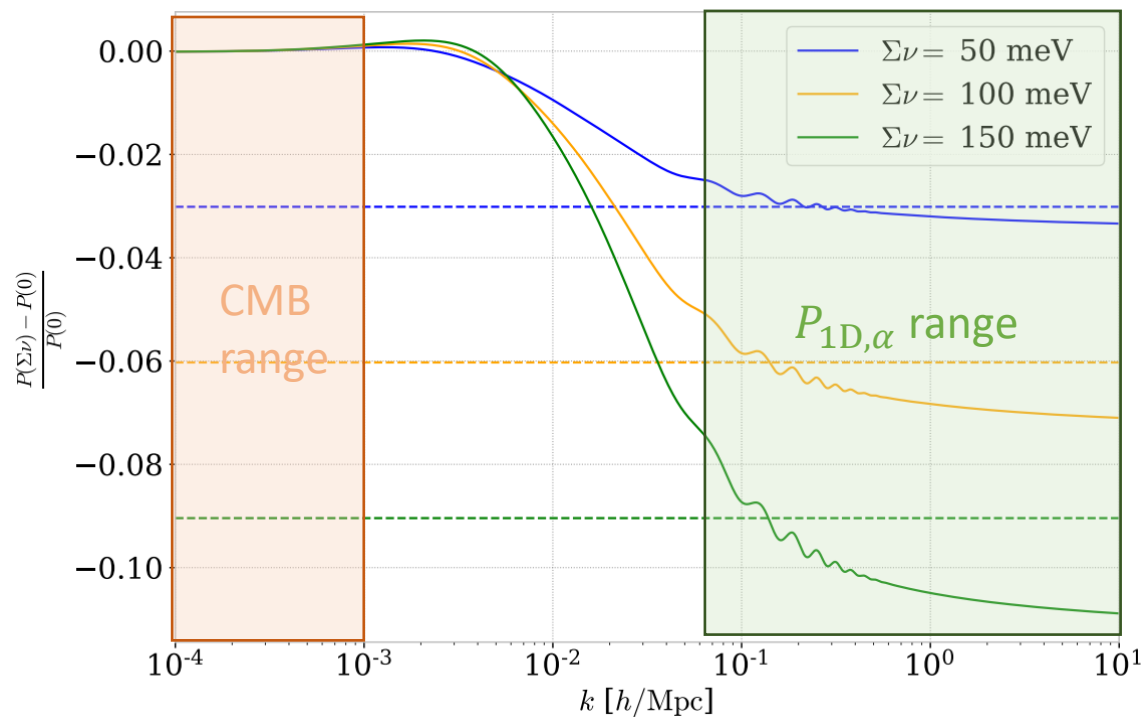
- One dimensional power spectrum ($P_{1D,\alpha}$)
 - Correlation along the line-of-sight
 - Probes small-scale matter clustering
- First measurement with DESI:
 - Increased statistics and resolution
 - Realize first measurement with method similar to eBOSS
- **Use to probe the small-scale matter power spectrum**



Application of $P_{1D,\alpha}$

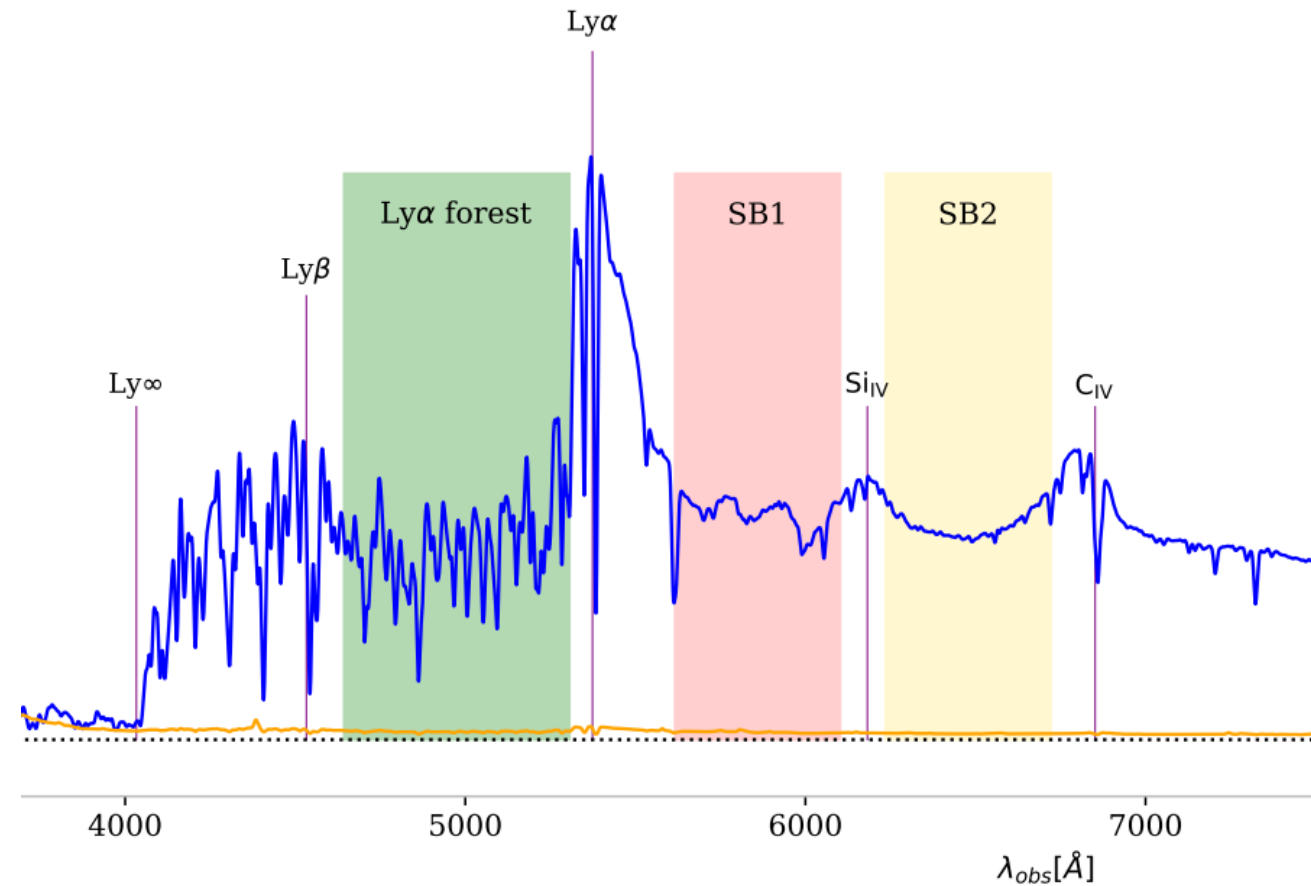
- Matter power spectrum impacted by:
 - Sum of neutrino masses $\sum m_\nu$
 - Dark matter model (e.g. warm dark matter)

$P_{1D,\alpha}$ unique tool to constrain neutrino masses and dark matter properties



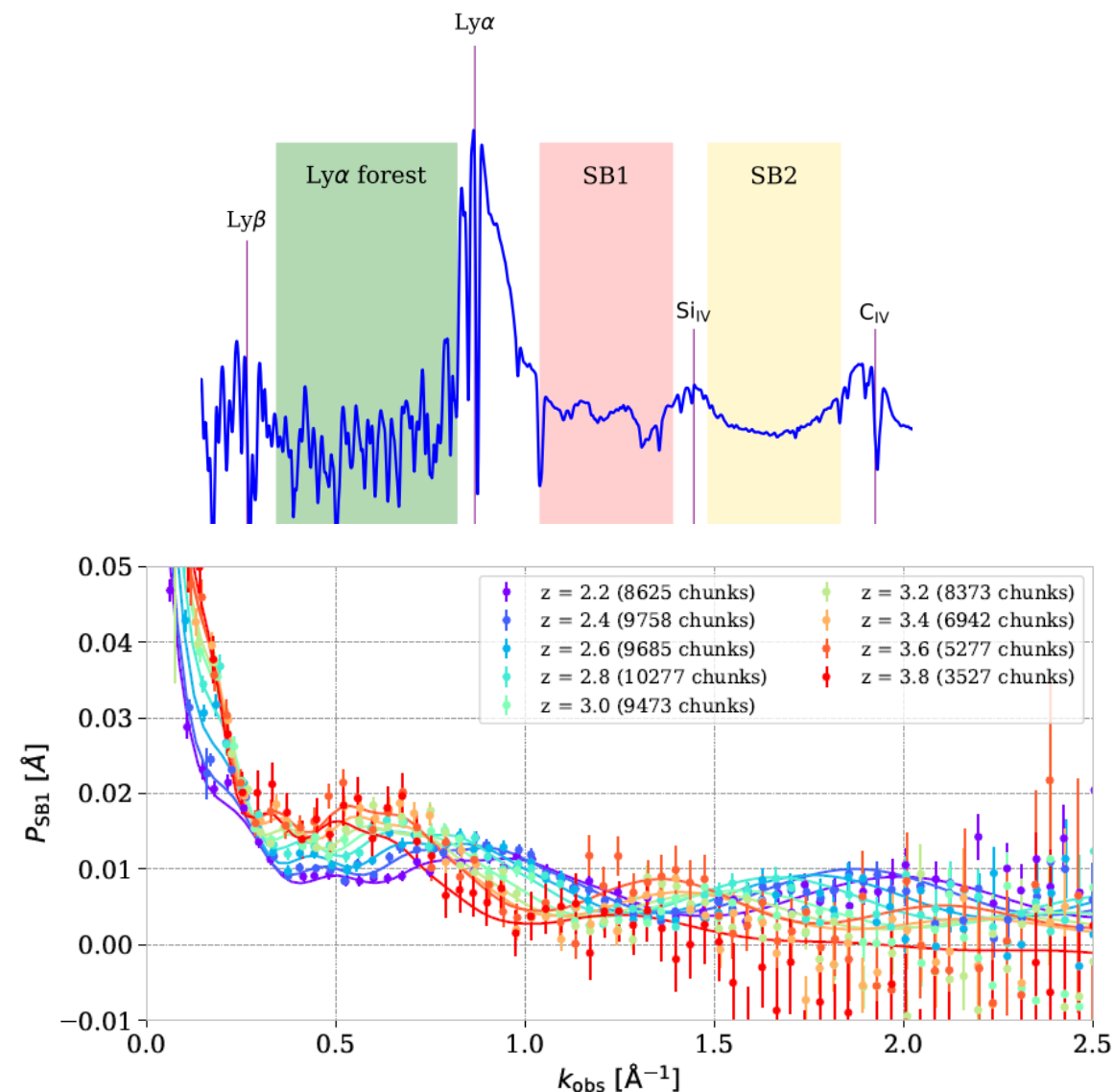
Systematic effects

- Instrumental:
 - Detector (spectrograph) noise
 - Finite spectrograph spectral resolution
- Astrophysical:
 - **Absorption by other IGM elements (metals)**
 - Damped Lyman- α systems (DLAs)
 - Missed broad absorption lines quasars (BALs)
- Analysis:
 - Masking of sky emission lines
 - Continuum fitting error



Example: effect of metals

- Contribution to $P_{1D,\alpha}$ from metals estimated using side bands
- Physically motivated parametrization to closely reproduce side band power spectrum
- Side band power spectrum subtracted to $P_{1D,\alpha}$ measurement

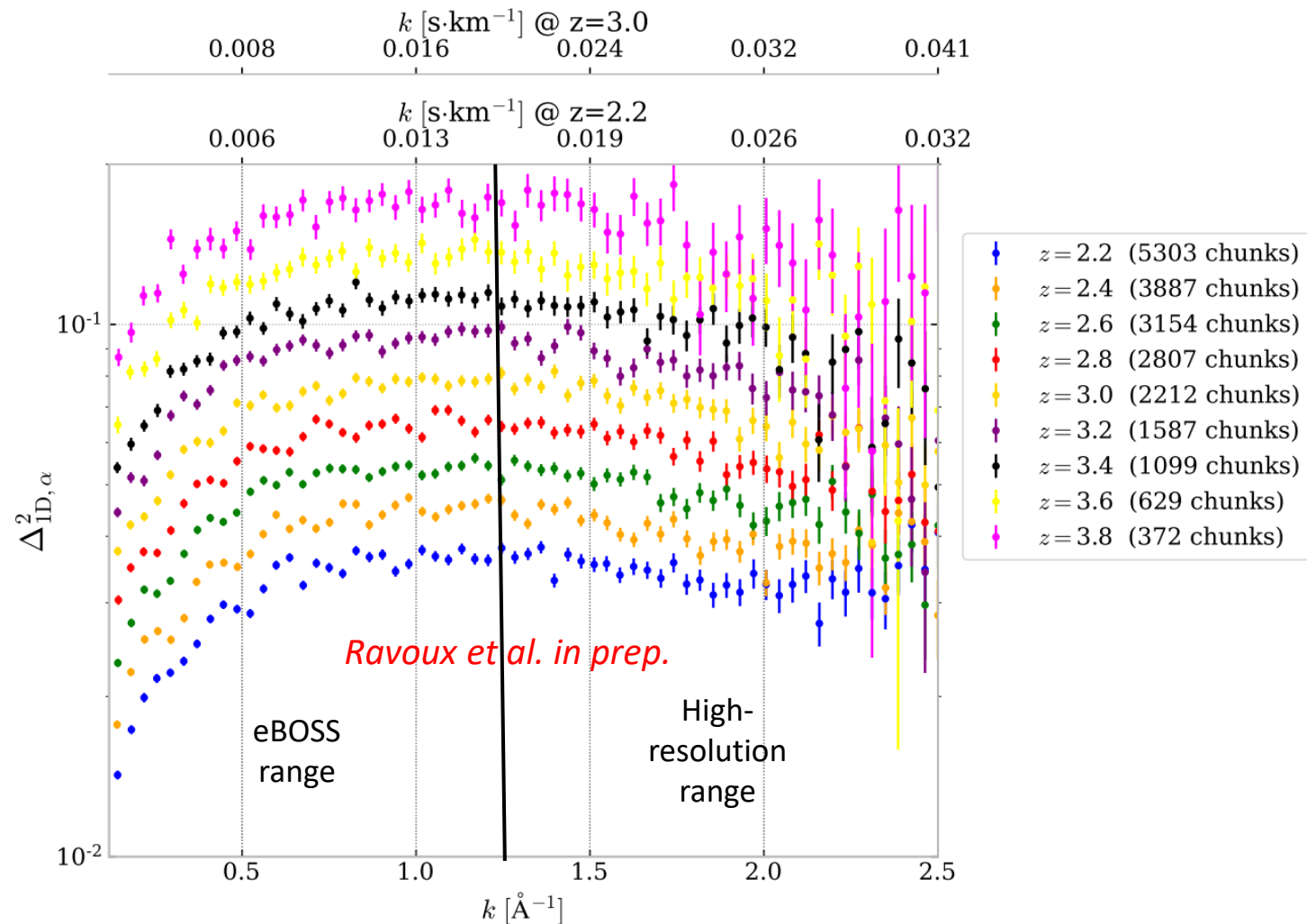


$P_{1D,\alpha}$ measurement with first DESI data

- Data used:

- Quasar spectra from survey validation & first 2 months of main survey
- SNR quality cut applied
- ~ 7000 quasar spectra used

- In agreement with eBOSS and high-resolution measurements on respective wavenumber ranges



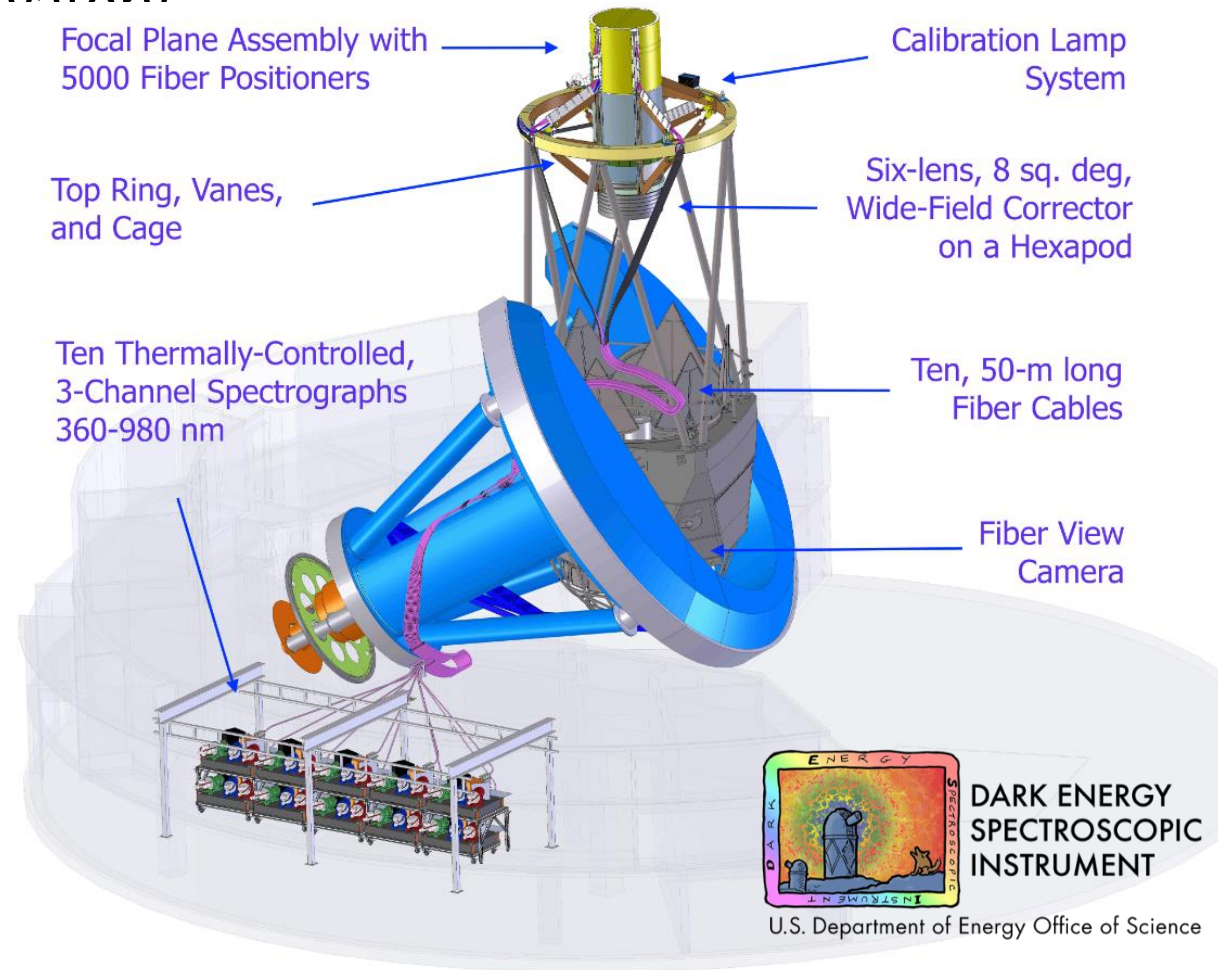
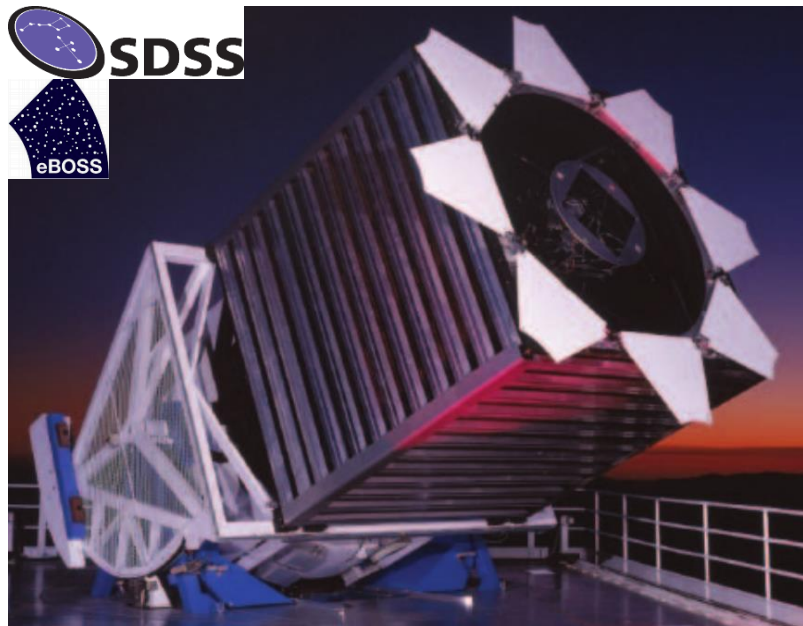
Interpreting this measurement with simulations will improve cosmological constraints

Conclusion

- Lyman- α tomography used to map a portion of the Universe
Ravoux et al. JCAP07(2020)010
- Void cross-correlation, exploratory work to constrain growth of structures
Ravoux et al. 2022
- $P_{1D,\alpha}$ will be used to improve constraints on neutrino mass and DM properties
Ravoux et al. in prep.

Spectroscopic surveys for cosmology

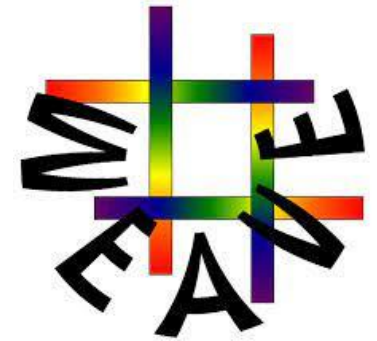
- **eBOSS**: 2.5 m telescope at Apache Point Observatory
 - Observations end March 2019
- **DESI**: 4 m telescope at Kitt Peak Observatory
 - Automated targeting, 5000 spectra / observation
 - Survey validation (SV) early 2021
 - Main survey started in May 2021



Lyman- α observations

- Moderate-resolution quasar surveys: Use of multi-object spectrographs

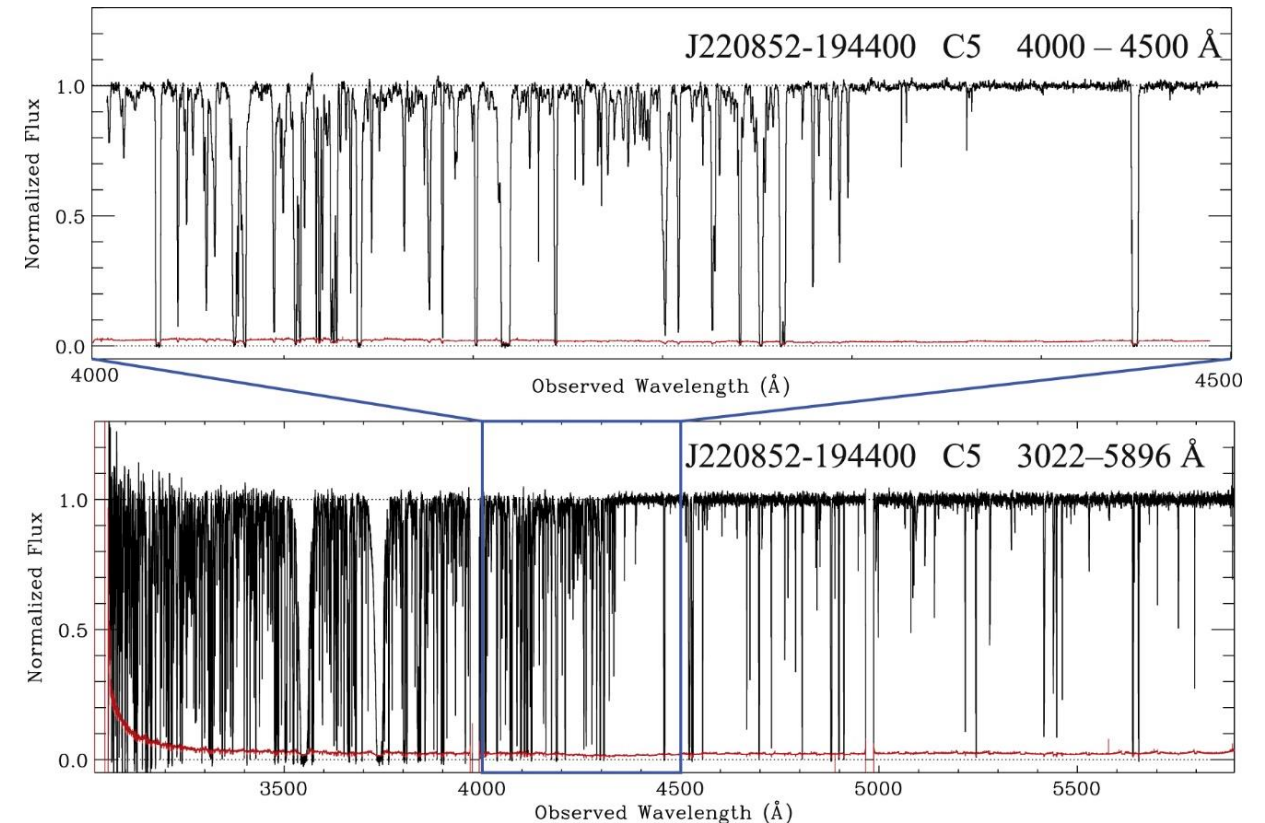
- SDSS/eBOSS
- DESI
- WEAVE-QSO



- High-resolution quasar observations: SQUAD (VLT), KODIAQ (Keck), COS (HST), ANDES (ELT)
- Other target: Lyman- α forest from Lyman-Break Galaxies (CLAMATO, DESI-LBG)

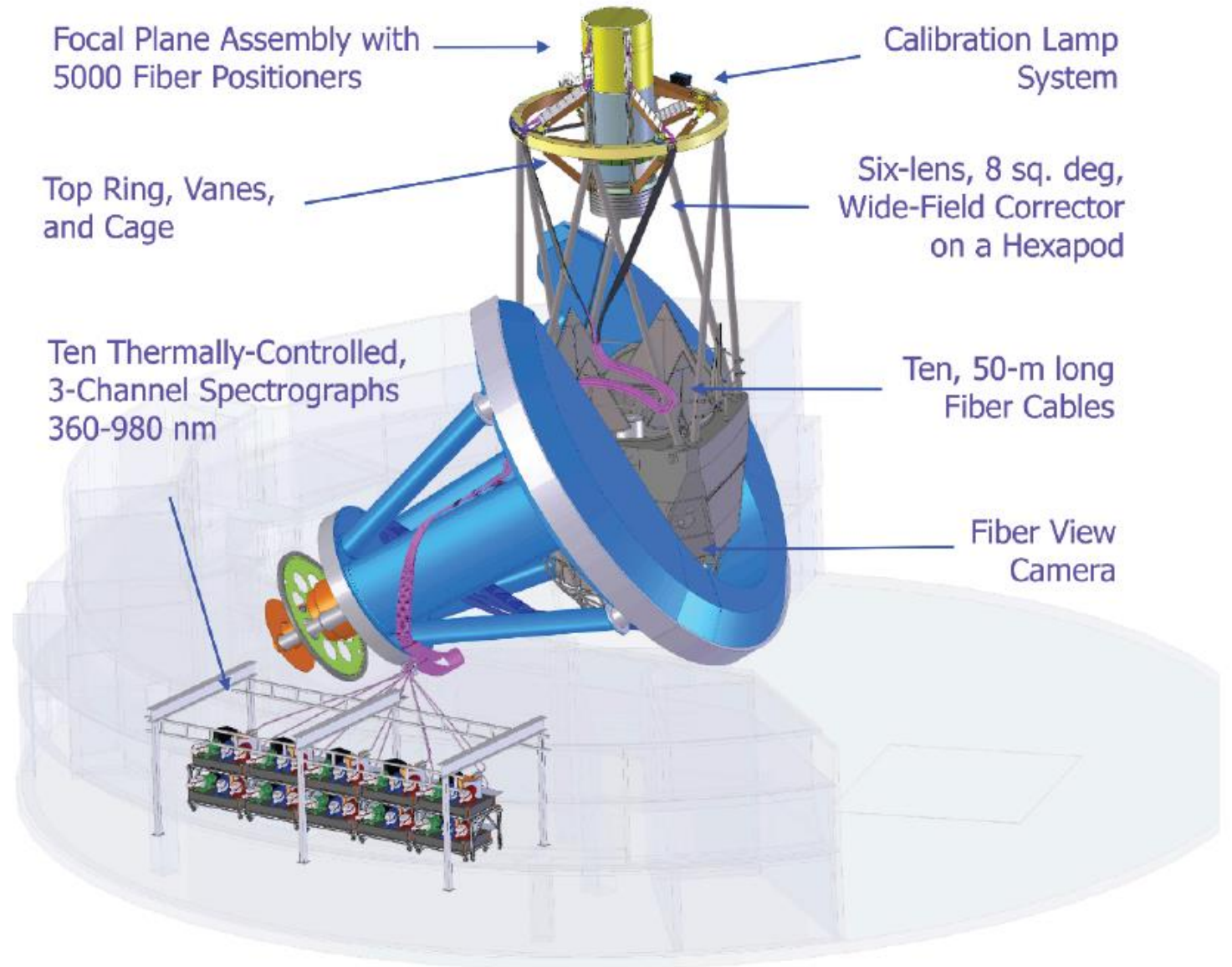
High resolution observations

- Currently : $R = 30\,000 - 100\,000$, mean SNR per pixel ~ 20
 - SQUAD survey (UVES,VLT), Murphy et al. 2019, 467 quasars
 - KODIAQ survey (HIRES,Keck), O’Meara et al. 2017, 300 quasars
 - COS instrument (HST), Danforth et al. 2016, 87 quasars
- Future instruments:
 - ESPRESSO instrument (VLT)
 - 4MOST Cosmology redshift survey (VISTA)
 - ANDES (ELT)

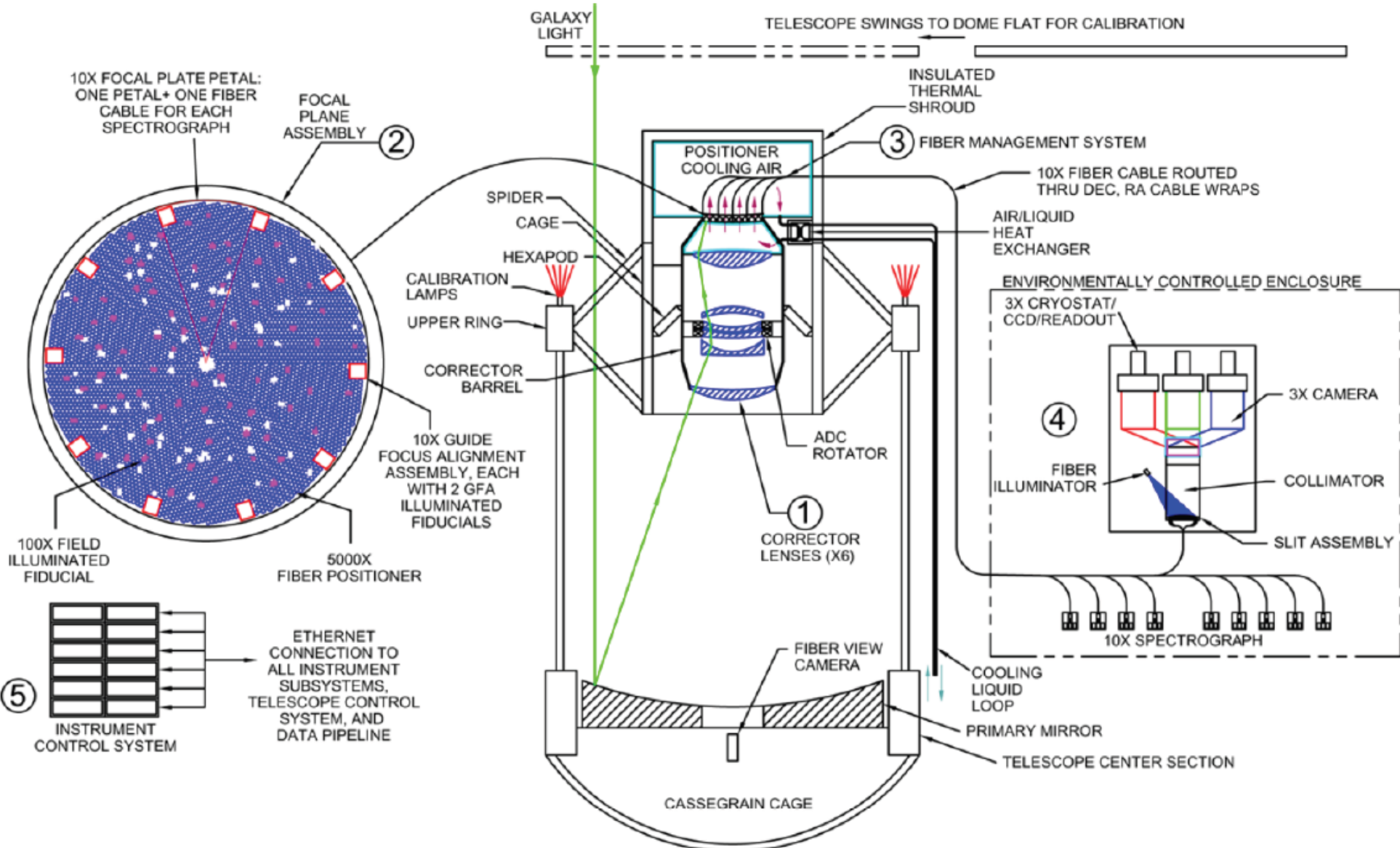


DESI Instrument

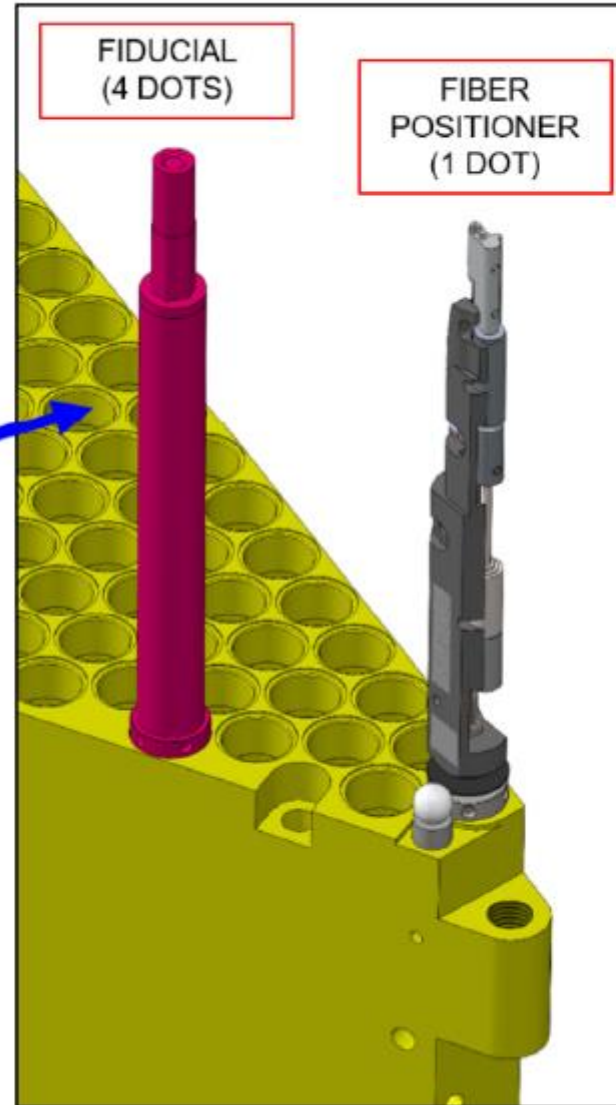
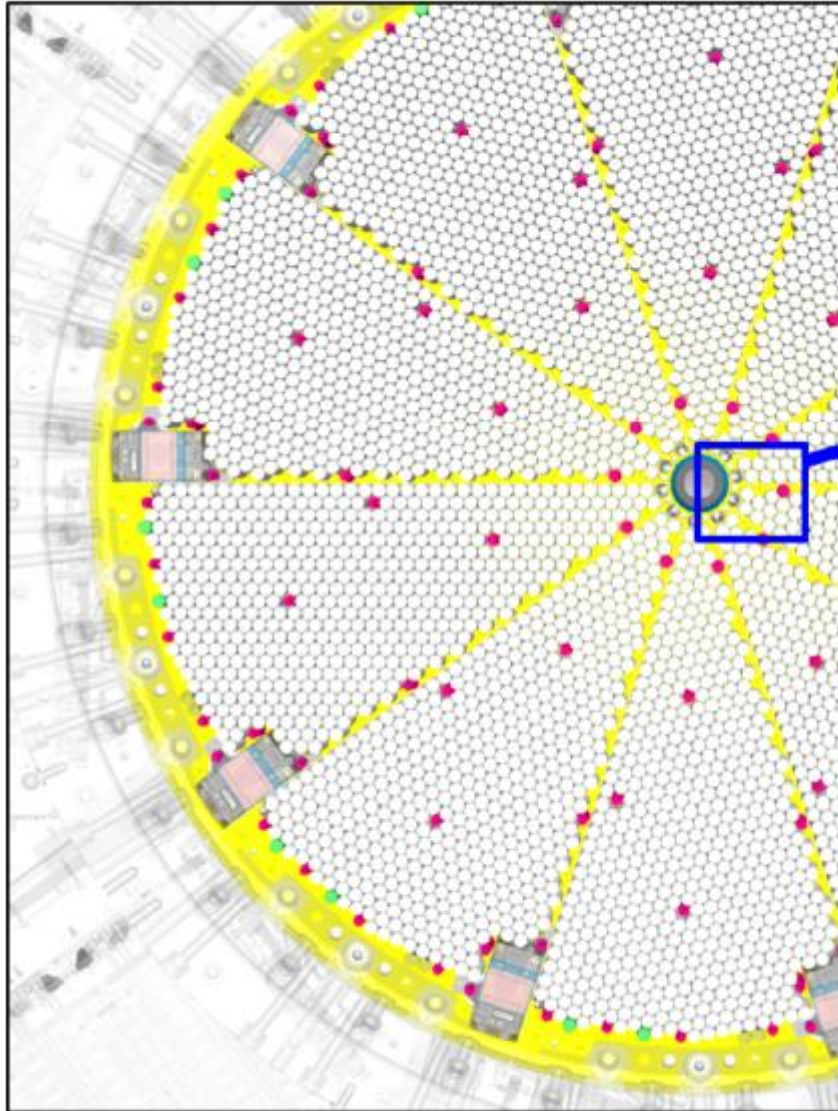
- Multi-object spectrograph
- Optical system redirect light from 5000 targets to 10 spectrographs
- Targeting done with a focal plane system composed of automated positioners
- Spectrographs composed of 3 spectral band each, receive 500 fiber light



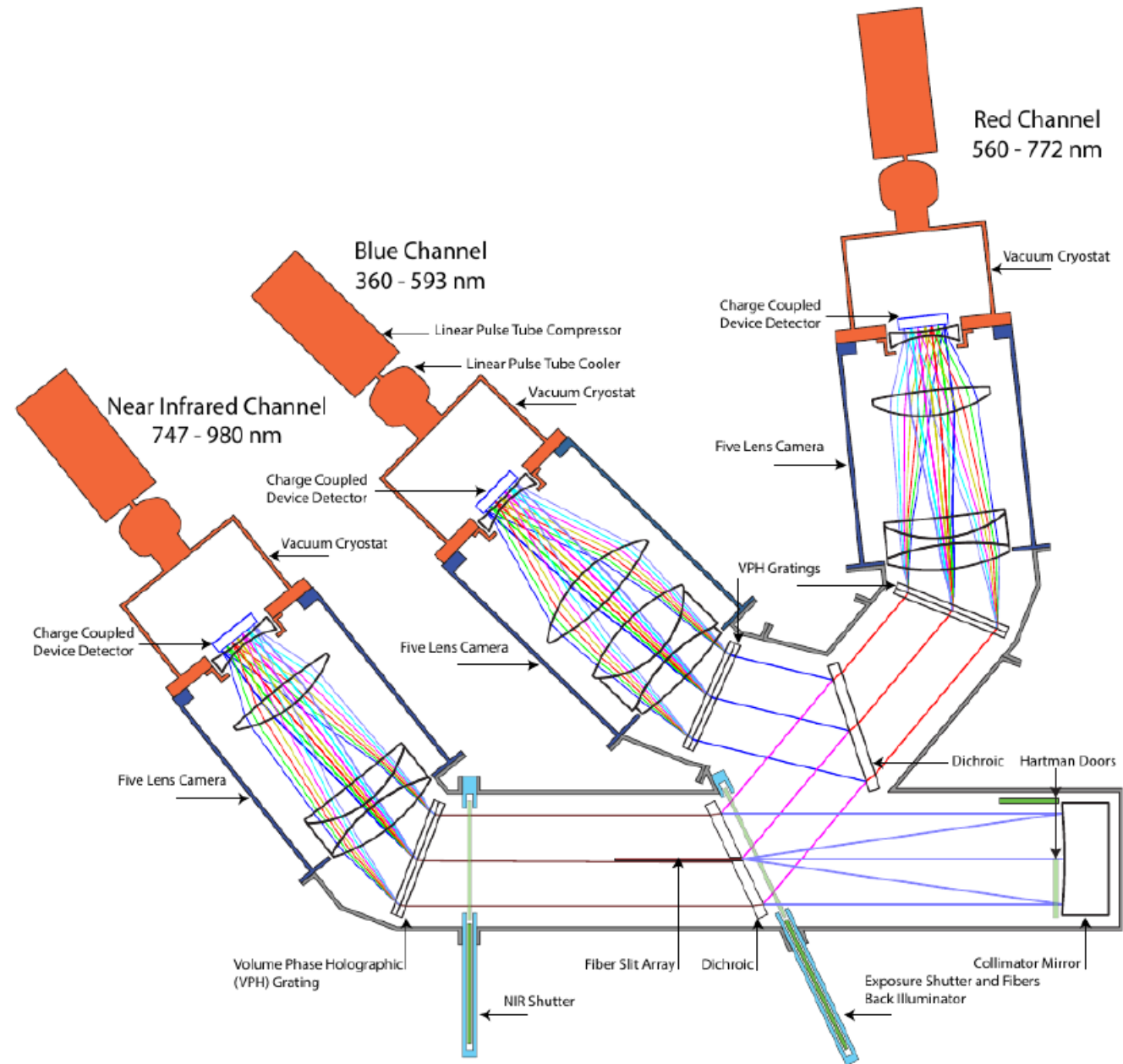
DESI Instrument



DESI Focal plan

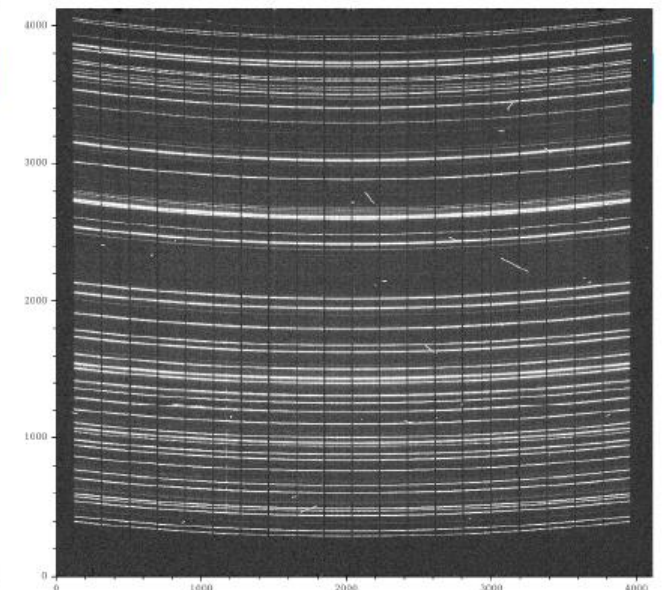
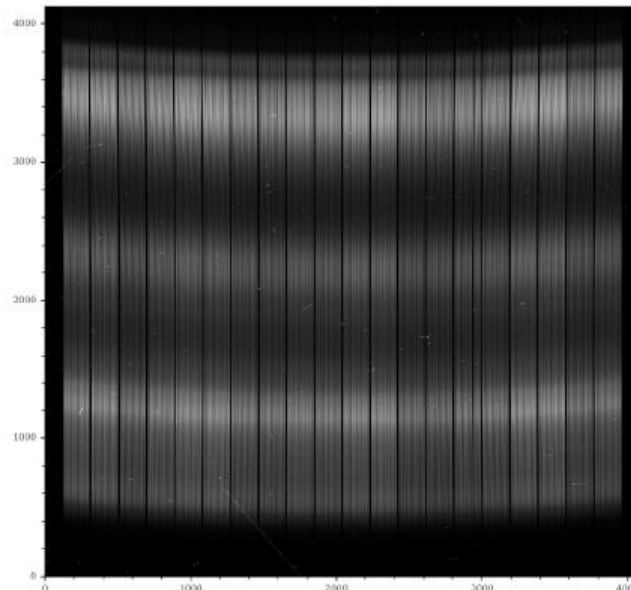
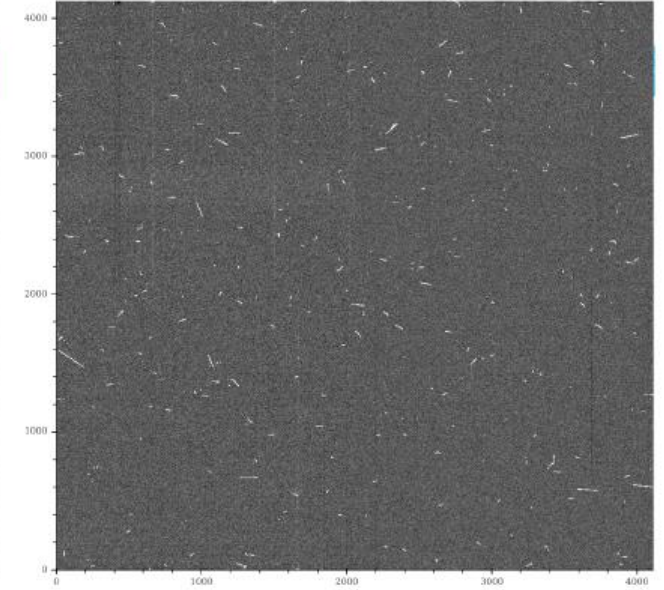
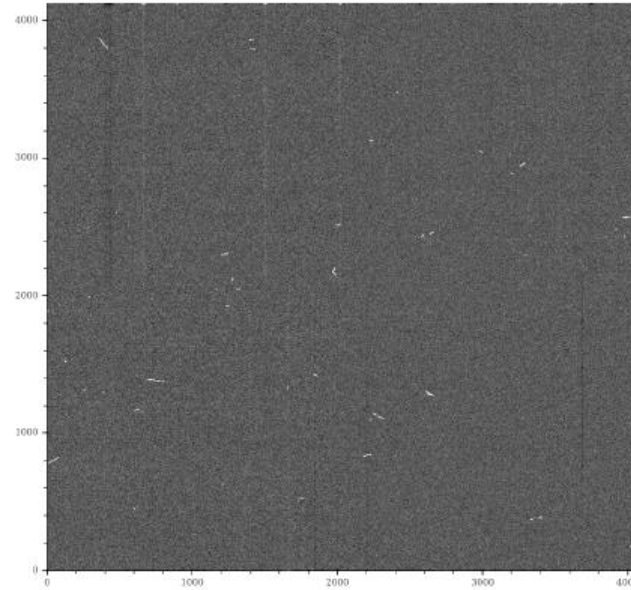


DESI Spectrograph

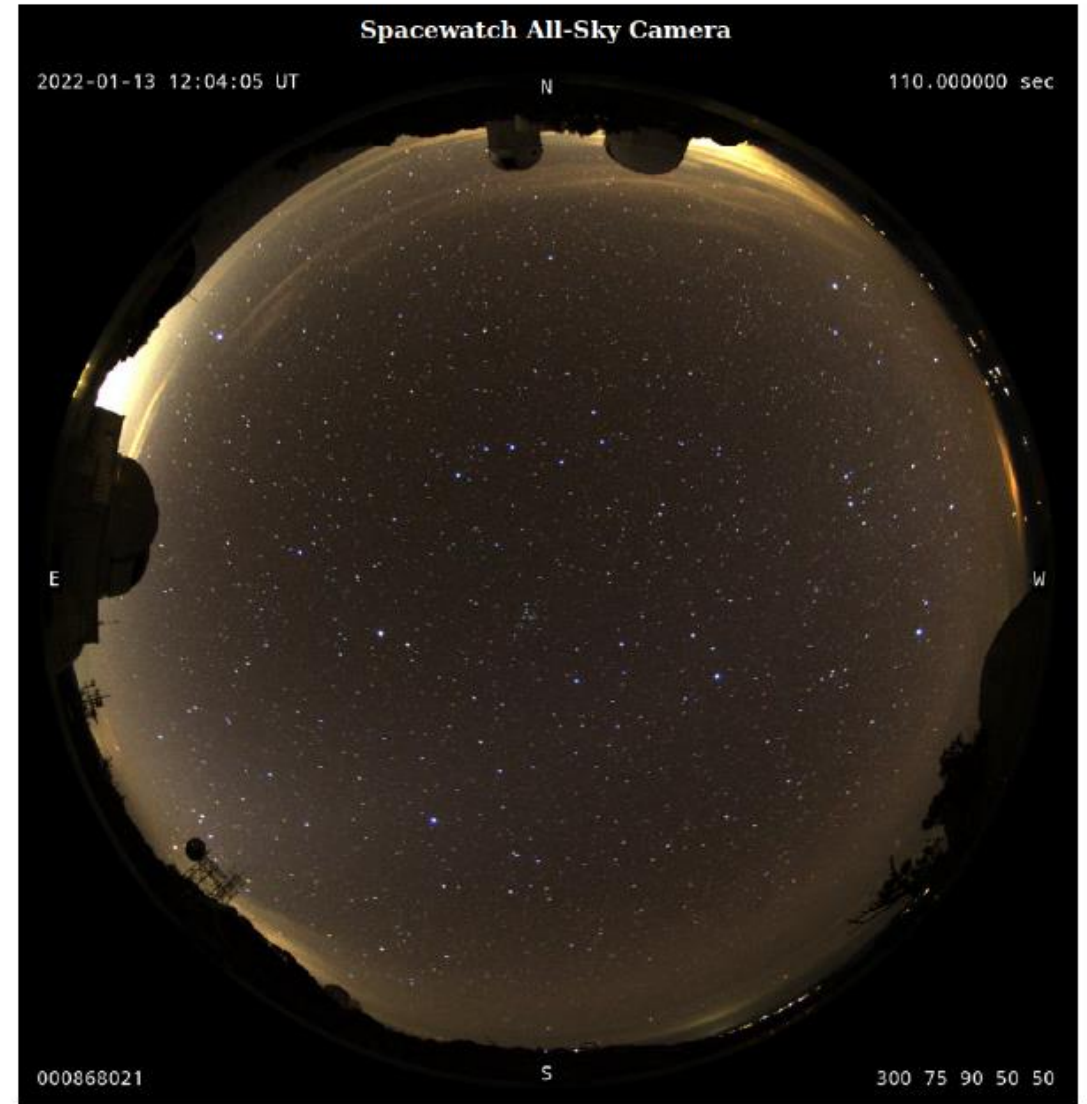
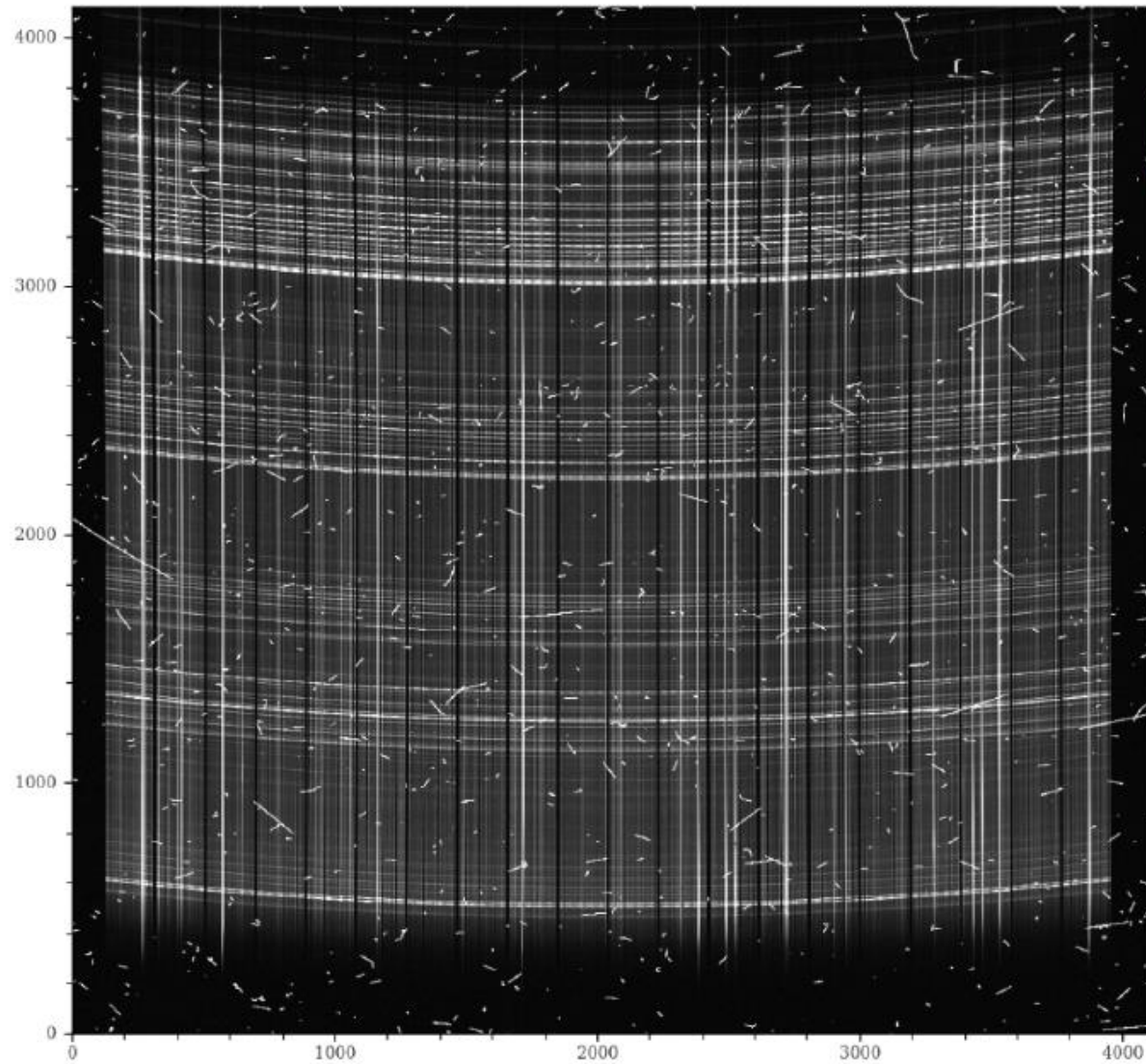


DESI shift

- DESI zero, dark, flat and arc frames on my observing night (from left to right, top to bottom).
- Used for calibrating spectrograph CCD, sky level, sky lines, wavelength grid...



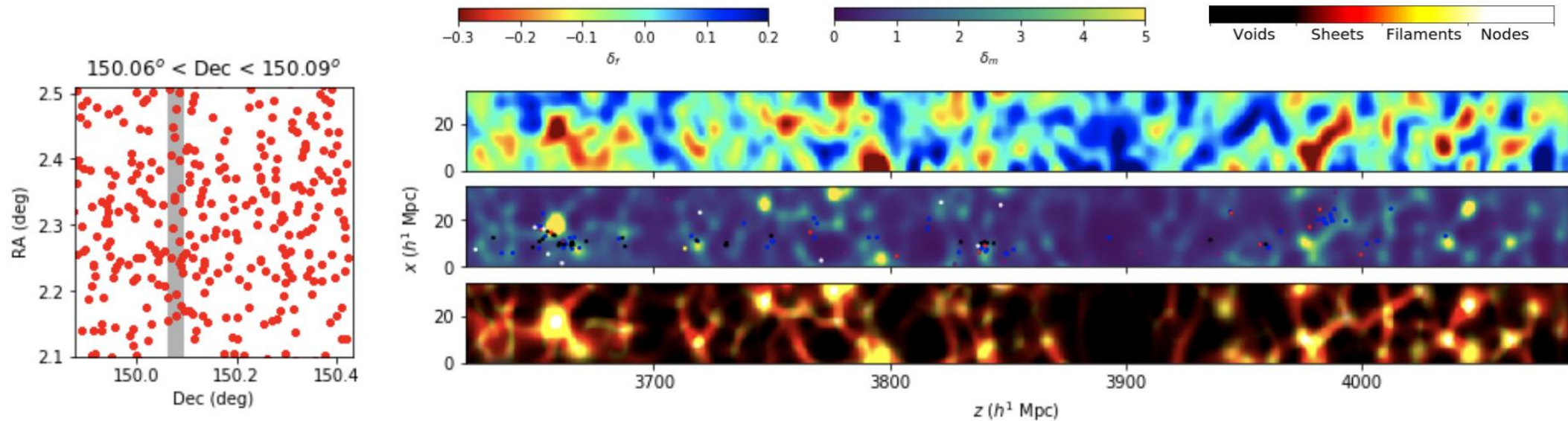
DESI shift



500 spectra distributed on the spectrograph CCD

Small-scale Lyman- α tomography

- 3D map of Lyman- α absorption from 1D spectra (*Pichon et al. 2001*)
- CLAMATO (*Lee et al. 2018*):
 - Small and dense field in COSMOS, 0.157 deg^2 , $1455 \text{ objects/deg}^2$
 - Use LBG and quasar spectra



- Goal: Map the cosmic web, reach Mpc scale
 - Might be achievable with ELT (*Japelj et al. 2019*)

Wiener filter

- Input data:

$$\vec{d} = \vec{s}_p + \vec{n}$$

- Minimization of

$$\epsilon = E \left[\left| \vec{s}_m - \hat{s} \right|^2 \right]$$

sm = « true » map signal
s = estimator
(Assumption on sm)

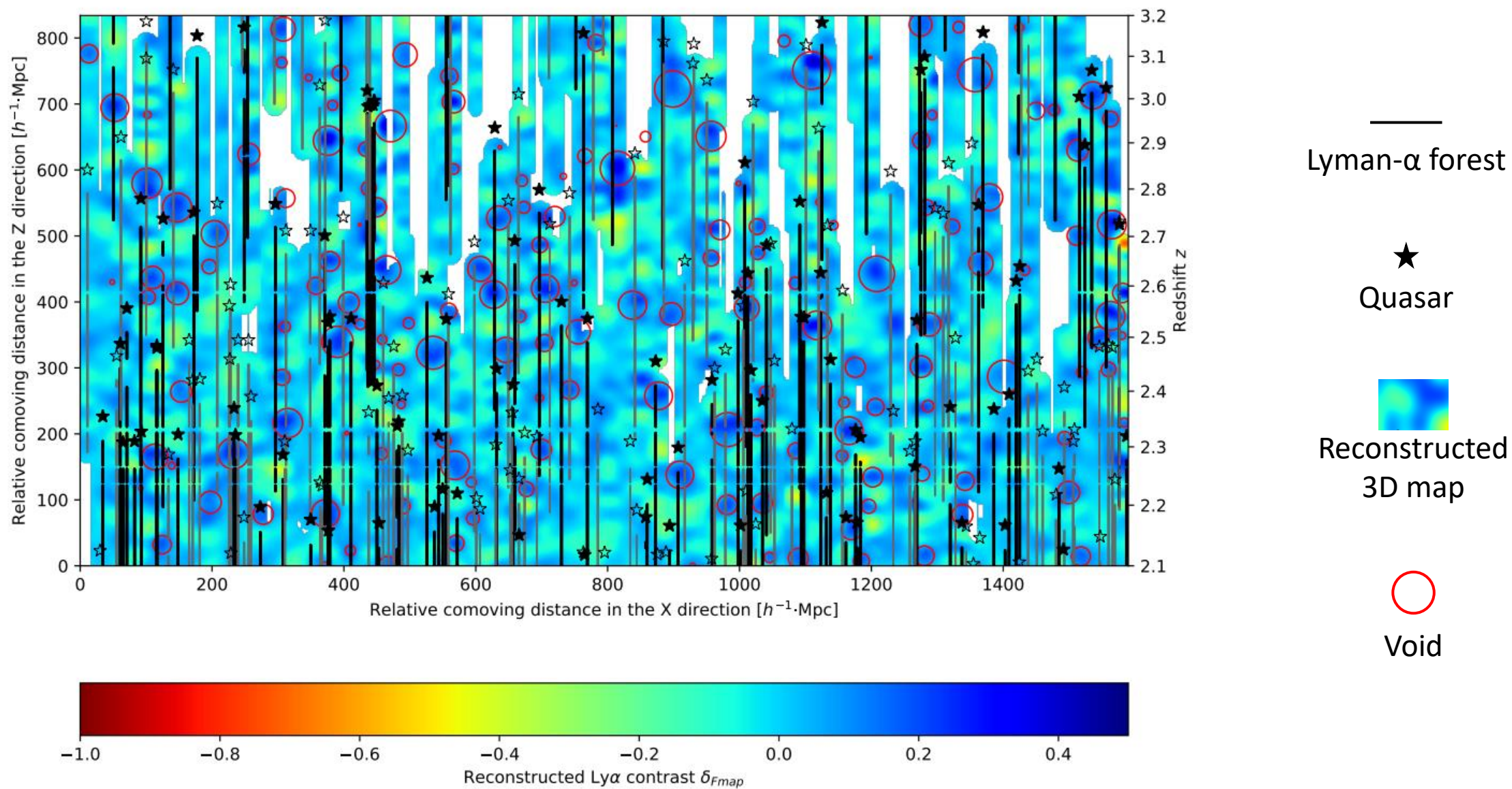
- Minimal error estimator:

$$\hat{s} = \left(\mathbf{S}_{mp} (\mathbf{S}_{pp} + \mathbf{N})^{-1} \right) \cdot \vec{d}$$

- Gaussian kernels:

$$\mathbf{S}_{ij} = \sigma_F^2 \exp \left(-\frac{(r_{i\parallel} - r_{j\parallel})^2}{2L_{\parallel}^2} \right) \exp \left(-\frac{(r_{i\perp} - r_{j\perp})^2}{2L_{\perp}^2} \right)$$

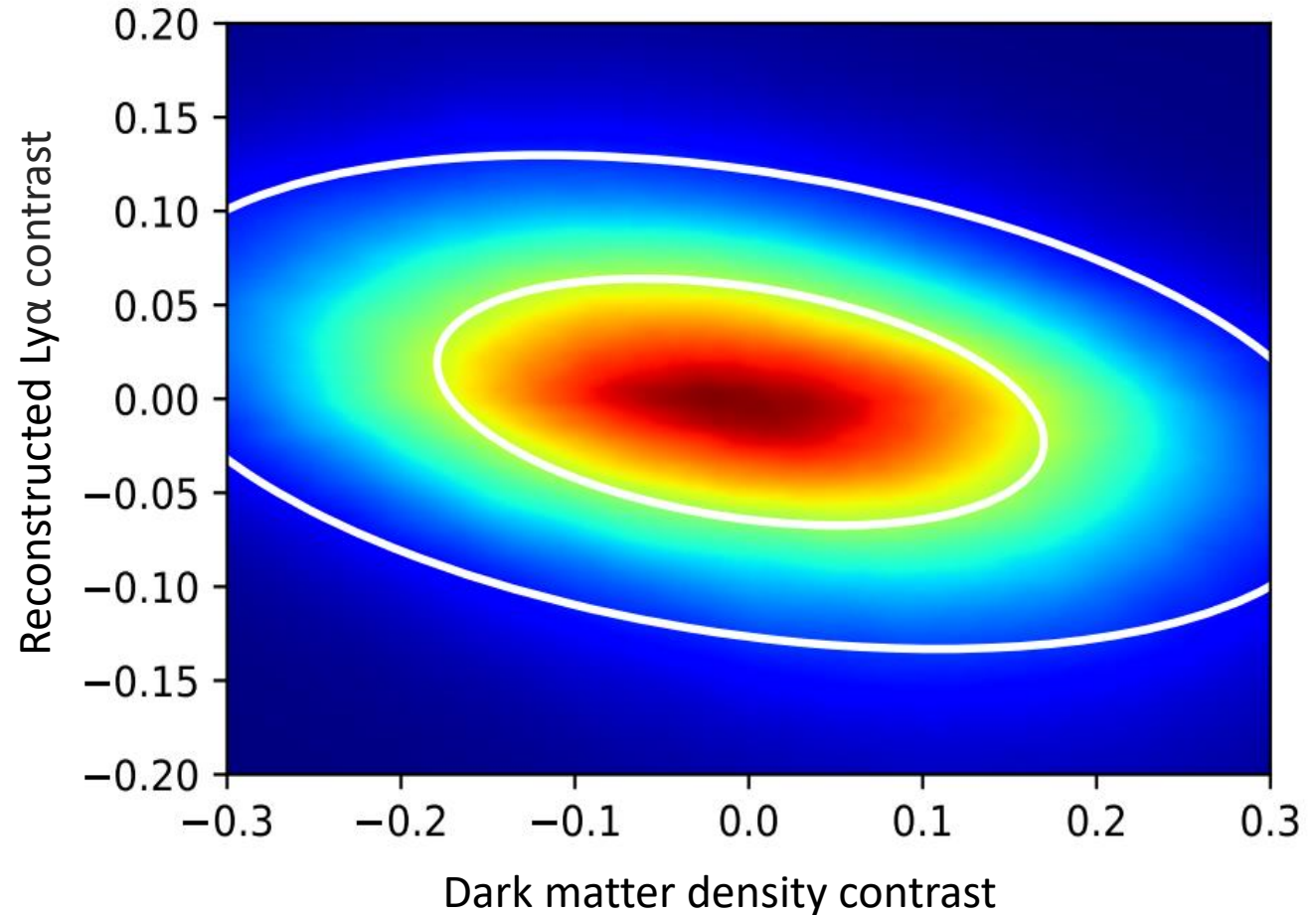
Large-scale Lyman- α tomography with eBOSS



Correlation with matter density

- Use dedicated simulations (mocks)
- Production of Lyman- α forest samples with the same properties as Stripe 82
- Test of tomographic algorithms
- Comparison of underlying matter field with reconstructed Lyman- α contrast

$$r = 34\%$$

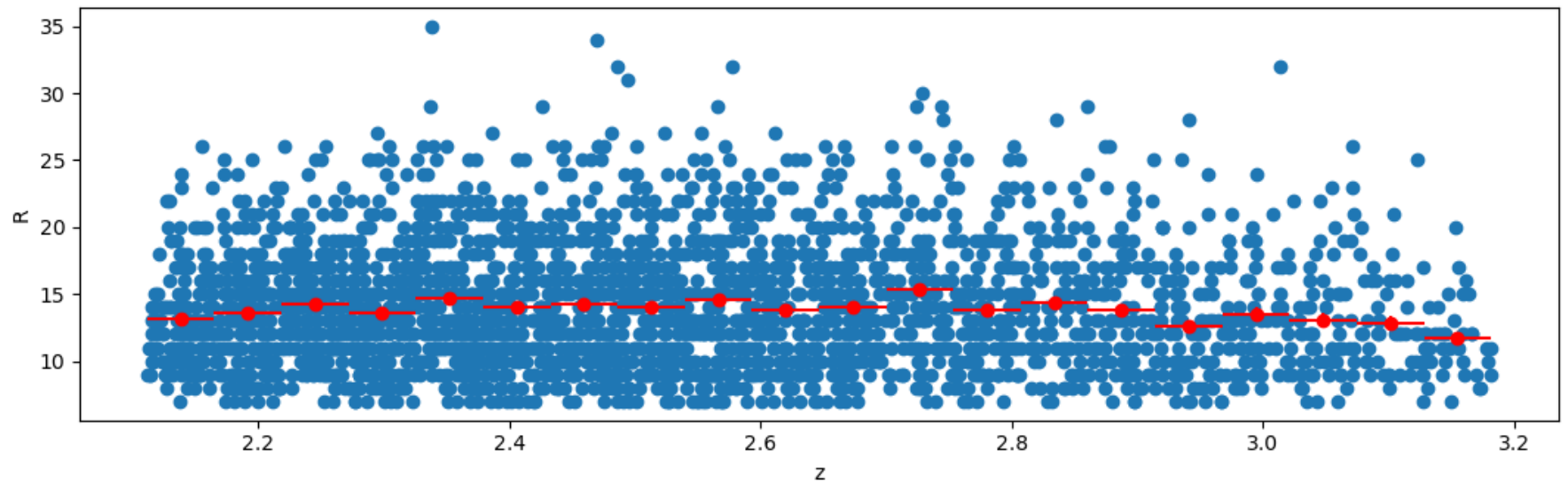
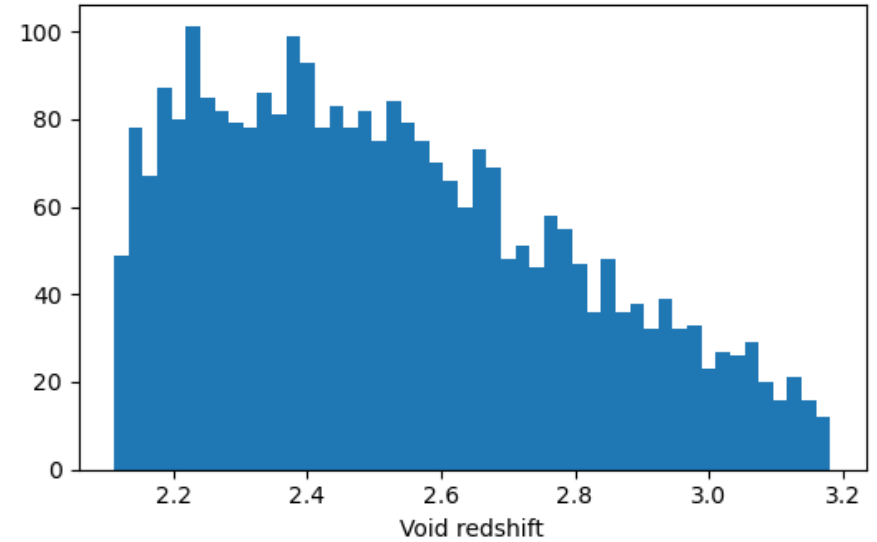
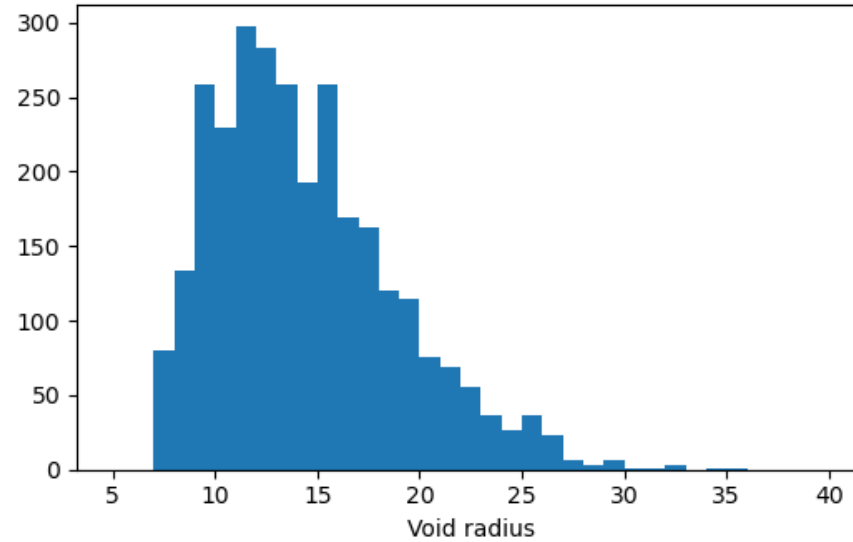


Void finder details

- Spherical void finder:
 - Select all pixels with reconstructed Ly α contrast larger than $\delta_{\text{th}}=0.14$
 - Sphere grown around this pixel until the mean reconstructed Ly α contrast inside reaches $\delta_{\text{av}}=0.12$
- Watershed void finder:
 - All pixels with a reconstructed Ly α contrast larger than are selected and sorted into groups of neighboring pixels.
 - Centers defined with the largest contrast
 - Radius defined by the total volume of pixels $R = (3N_{\text{pix}}V_{\text{pix}}/4\pi)^{1/3}$
- Voids with radius lower than R_{min} removed
- Overlapping void removed by iteration or clustering

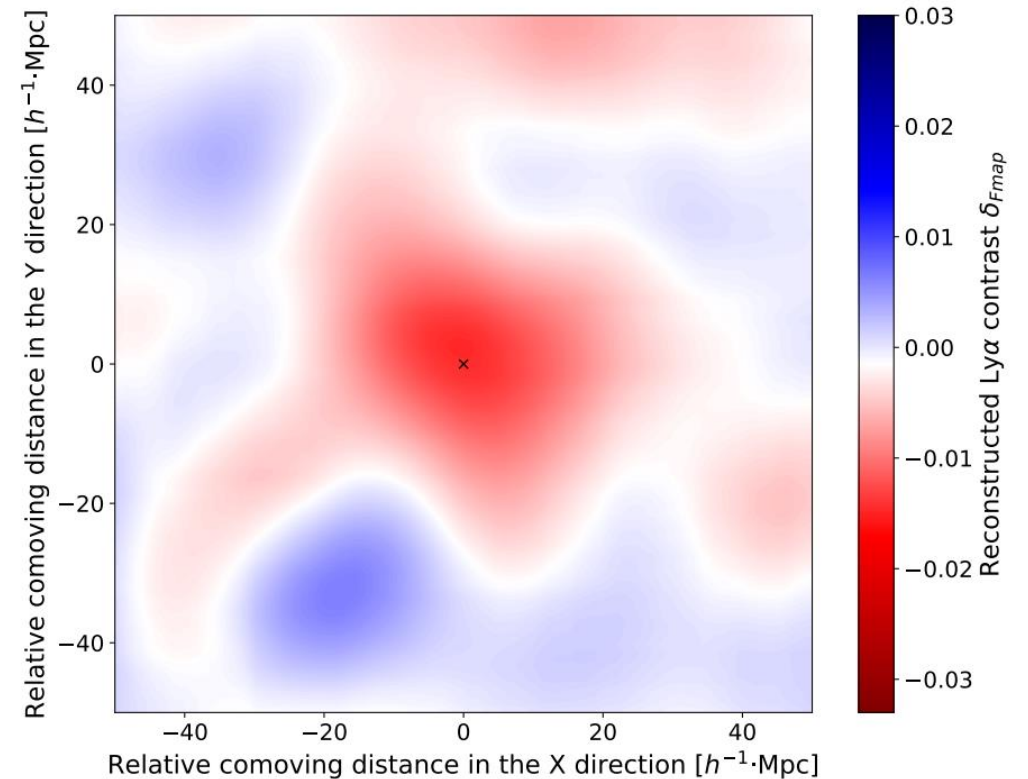
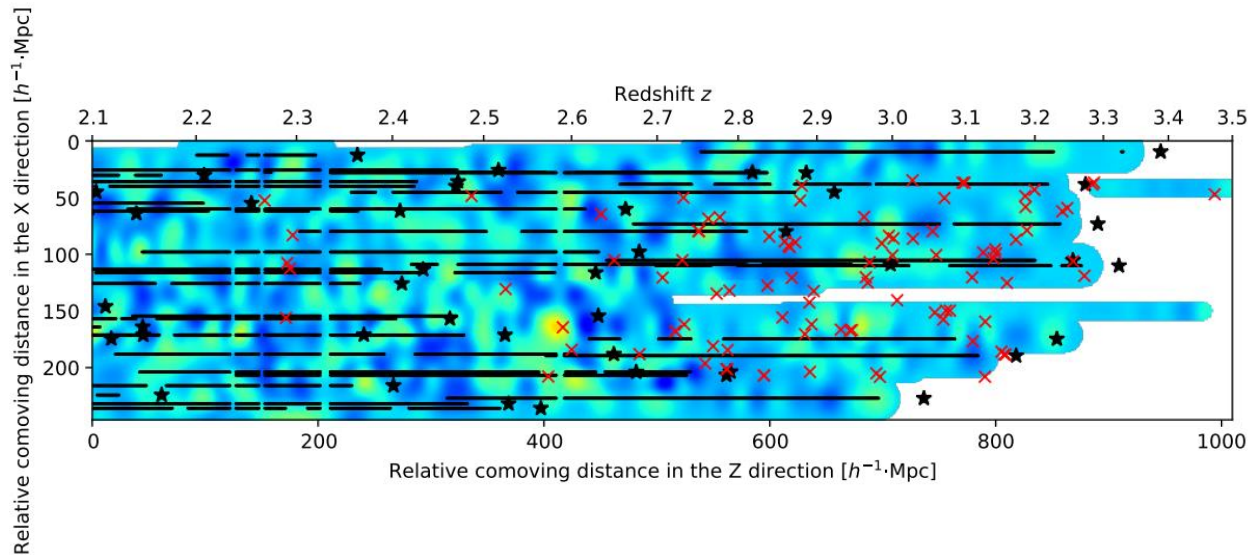
Void statistics

- Radius and redshift histograms
- Distribution of radius as a function of redshift rather stable.



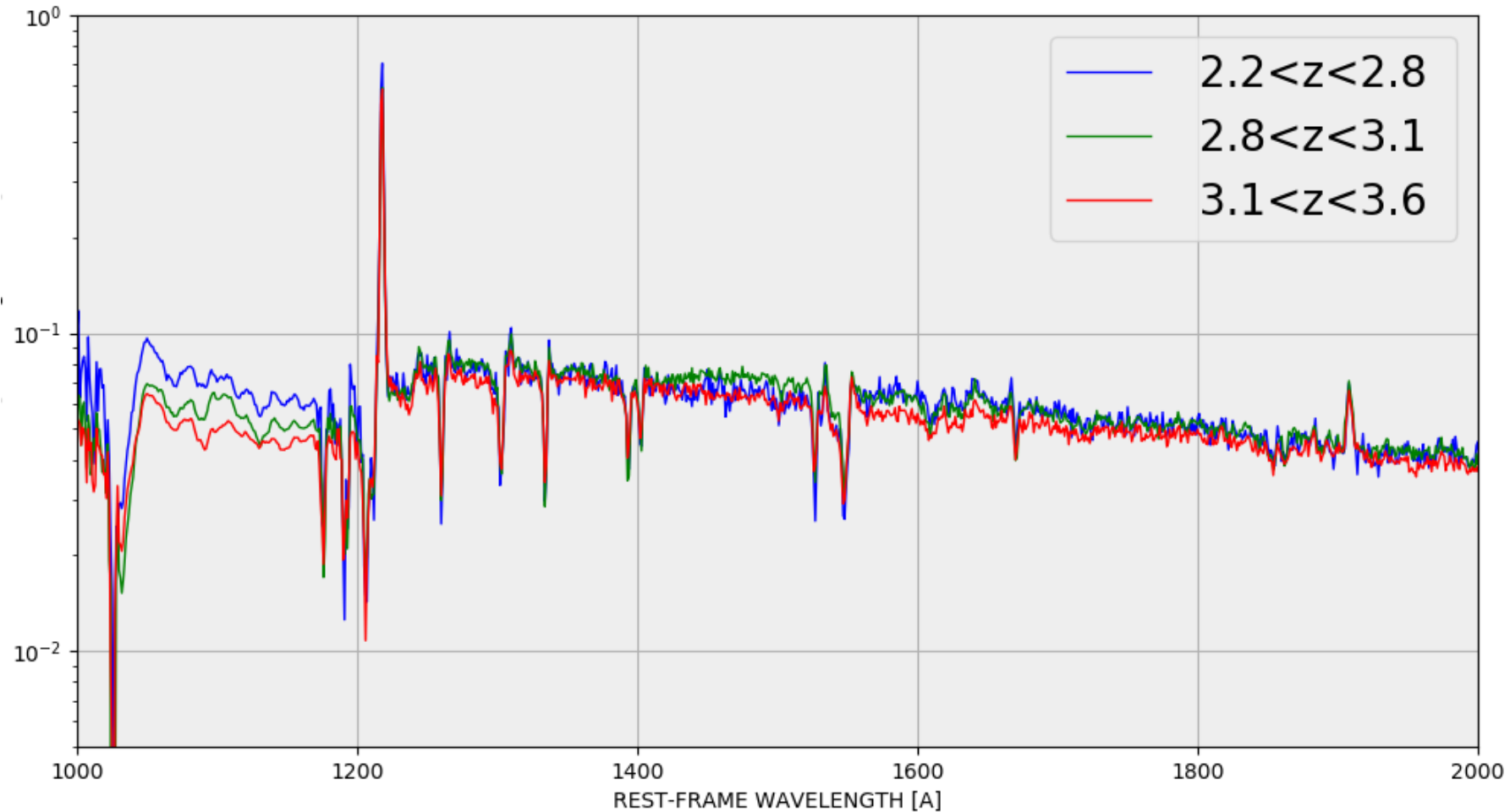
DESI-LBG: Secondary target program

- Secondary project to test DESI ability to observe Lyman-break galaxies:
 - Tracers
 - Lyman-alpha absorption
- Tomographic map with quasars only on COSMOS field
- Stacking on LBG positions on the tomographic map yields an over-dense signal with 3σ statistical significance.



DESI-LBG: Secondary target program

- Stacked spectra of 2200 DESI LBGs (credits: Christophe Yèche)



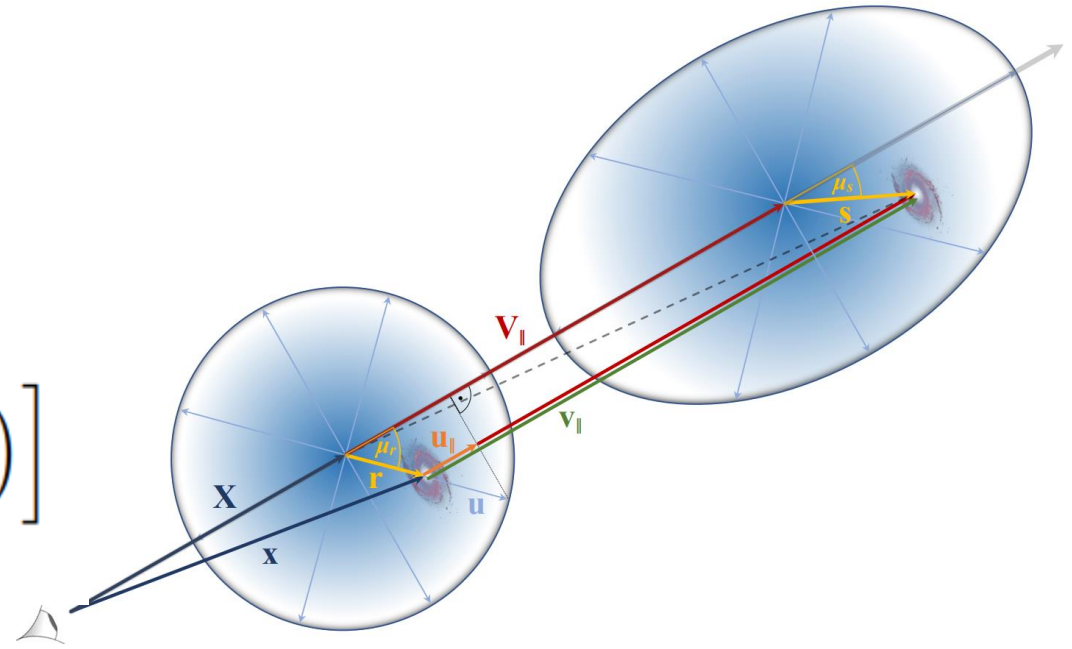
RSD void model

- Velocity:
$$\vec{u} = -\frac{1}{3} \frac{fH}{1+z} \bar{\delta}(r) \vec{r} \quad \bar{\delta}(r) = \frac{3}{r^3} \int_0^r \delta(r') r'^2 dr'$$

- RSD transformation:
$$\vec{s} = \vec{r} + (1+z) \frac{\hat{X} \cdot (\vec{u} - \vec{U})}{H(z)} \hat{X}$$

- Optical depth conservation:

$$\begin{aligned} \tau_{\alpha}^s(\vec{r}) &= \tau_{\alpha}(r) \left| \frac{d^3 \vec{r}}{d^3 \vec{s}} \right| \\ &= \tau_{\alpha}(r) \left[1 + \frac{f}{3b} \bar{\delta}(r) + \frac{f}{b} \mu^2 (\delta(r) - \bar{\delta}(r)) \right] \end{aligned}$$



RSD void model

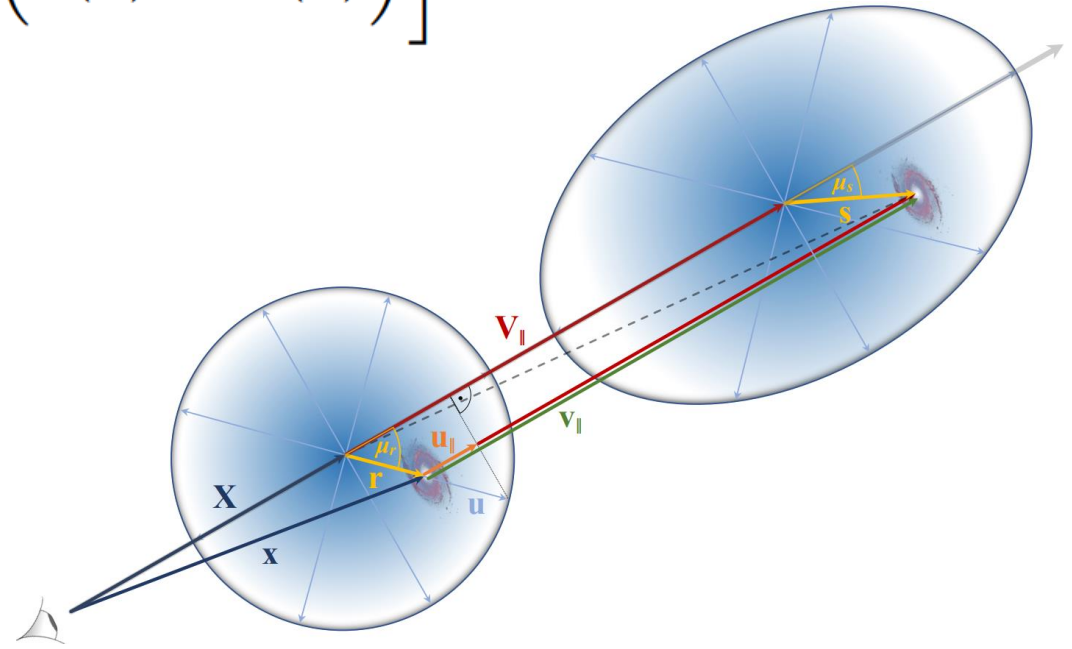
- Taylor expansion:

$$F[\tau_\alpha^s(\vec{k})] = \sum_{n=0}^{\infty} \frac{F^{(n)}[0] \left(\tau_\alpha^s(\vec{k}) - 0 \right)^n}{n!}$$

- At linear order and introducing velocity bias:

$$F[\tau_\alpha^s(\vec{r})] = F[\tau_\alpha(r)] + b_\eta f \bar{F} \left[\frac{1}{3} \bar{\delta}(r) + \mu^2 \left(\delta(r) - \bar{\delta}(r) \right) \right]$$

$$b_\eta = \tau_\alpha \left. \frac{\partial \delta_F}{\partial \tau_\alpha} \right|_{\tau=0} = \tau_\alpha \frac{F^{(1)}[0]}{\bar{F}}$$



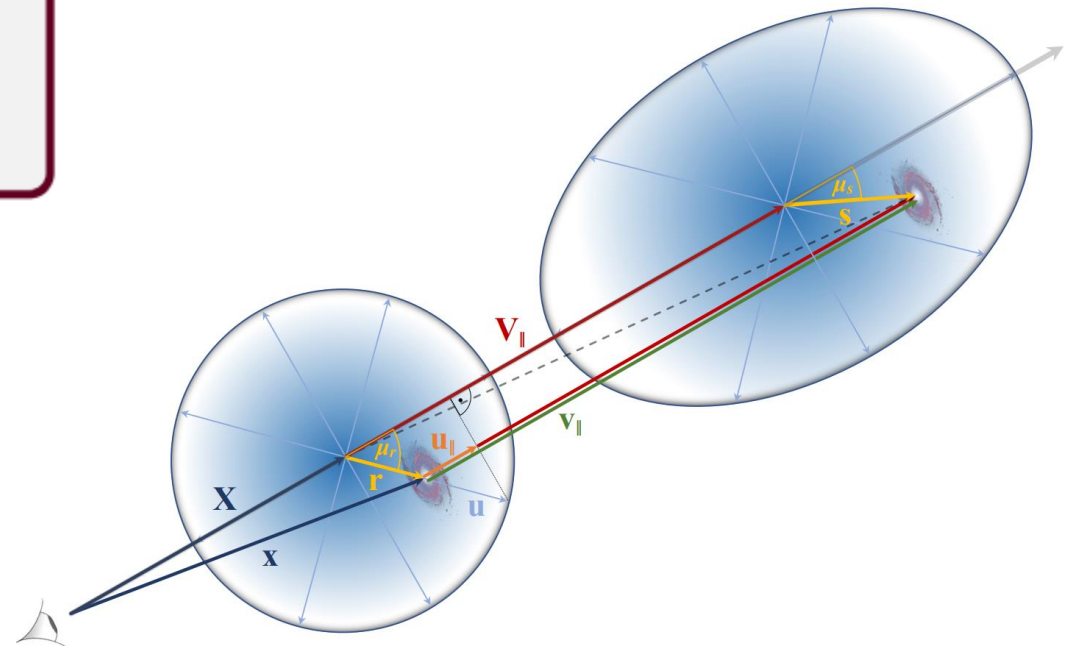
RSD void model

- Multipoles:

$$\xi_{\nu\alpha,0}^s(r) = \left(1 + \frac{\beta}{3}\right) \xi_{\nu\alpha}(r)$$

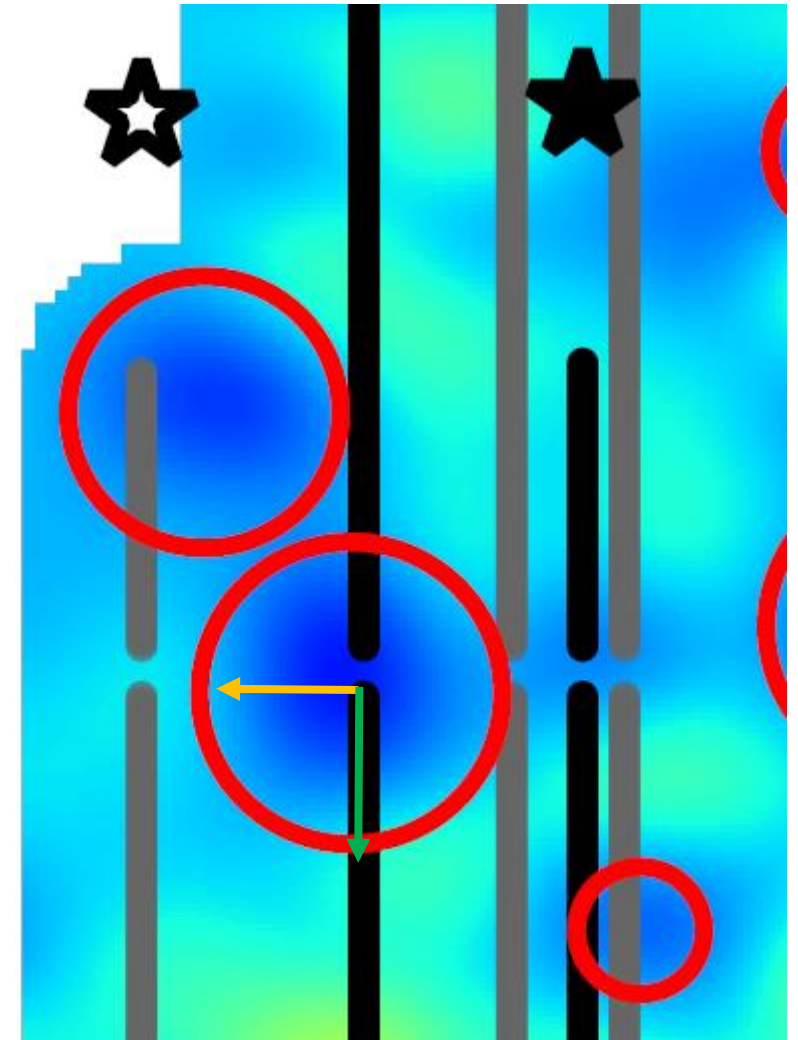
$$\xi_{\nu\alpha,2}^s(r) = \left(\frac{2\beta}{3}\right) (\xi_{\nu\alpha}(r) - \bar{\xi}_{\nu\alpha}(r))$$

$$\xi_{\nu\alpha,2}^s(r) = \left(\frac{2\beta}{3 + \beta}\right) (\xi_{\nu\alpha,0}^s(r) - \bar{\xi}_{\nu\alpha,0}^s(r))$$



Possible sources of tomographic effect

- First source:
 - With sparse LOS geometry, the void position is biased toward the LOS which generates it on the transverse plan
- Second source:
 - Map flux-contrast signal higher near to the LOS and decrease following a Gaussian kernel along transverse direction
- Along the LOS, field is statically over-dense after an under-density (void). Not necessary the case for transverse direction
- Difference between these two effects generate a quadrupole



Tomographic effect model

- Two toy models created:

- Biasing of void position $\vec{s} = s_{\parallel}\hat{X} + s_{\perp}\hat{Y} = r_{\parallel}\hat{X} + (r_{\perp} - \epsilon(r_{\perp}))\hat{Y}$

- Void finder efficiency $\xi_{\text{v}\alpha}^s(\vec{r}) = \xi_{\text{v}\alpha}(r) \times \frac{n_{\text{los}}(r_{\perp})}{\overline{n_{\text{los}}}}$

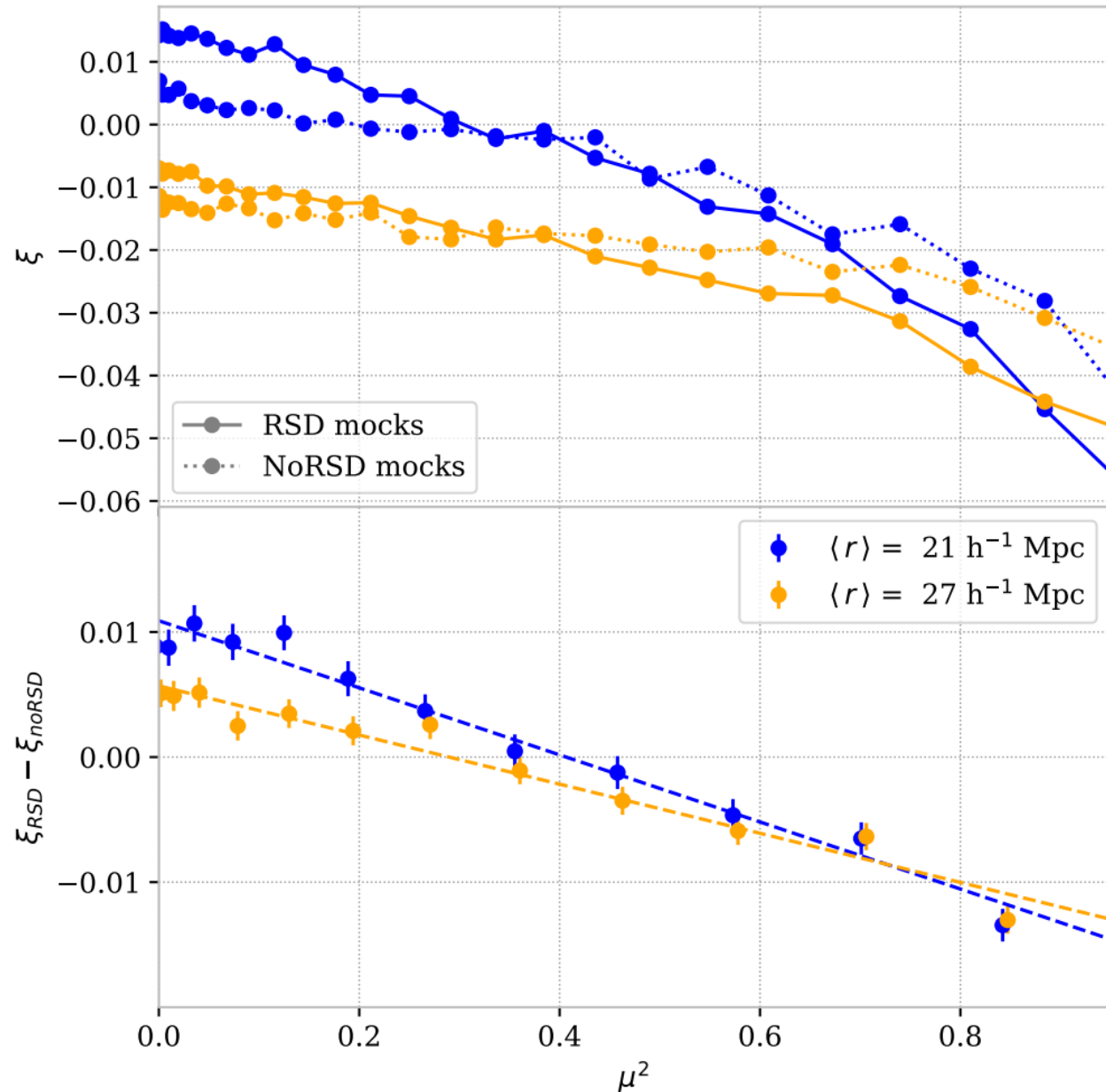
- Similar results:

- Create monopole, quadrupole and hexadecapole
 - All poles proportional to a function which only depends on r . The proportionality coefficient depends on the toy model considered

$$\xi_{\text{tomo},\ell} = A_{\ell}f(r)$$

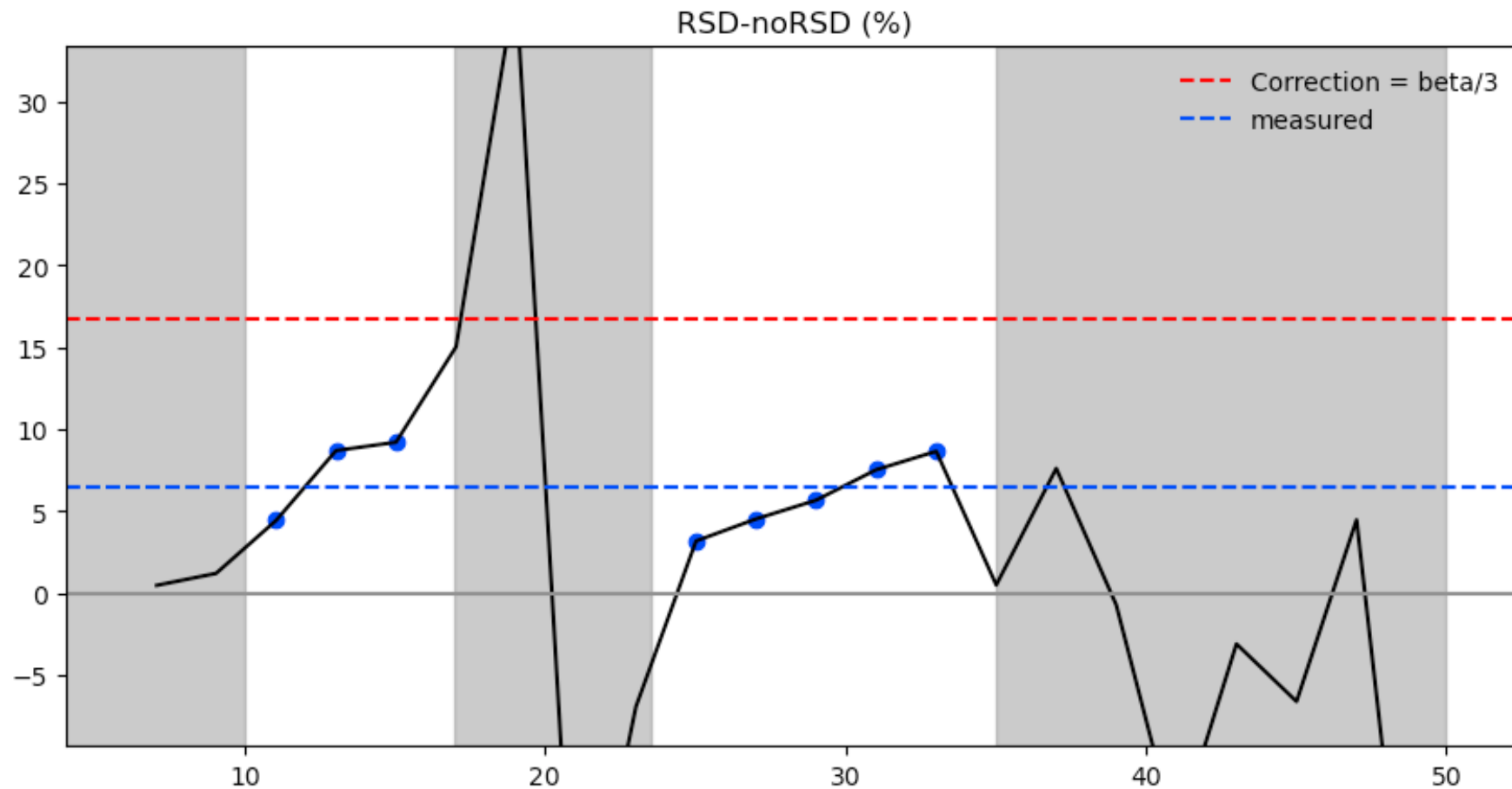
Comparison mocks with and without RSD

- Removal of the tomographic effect with no RSD mock
- Linear relation as in the Kaiser model



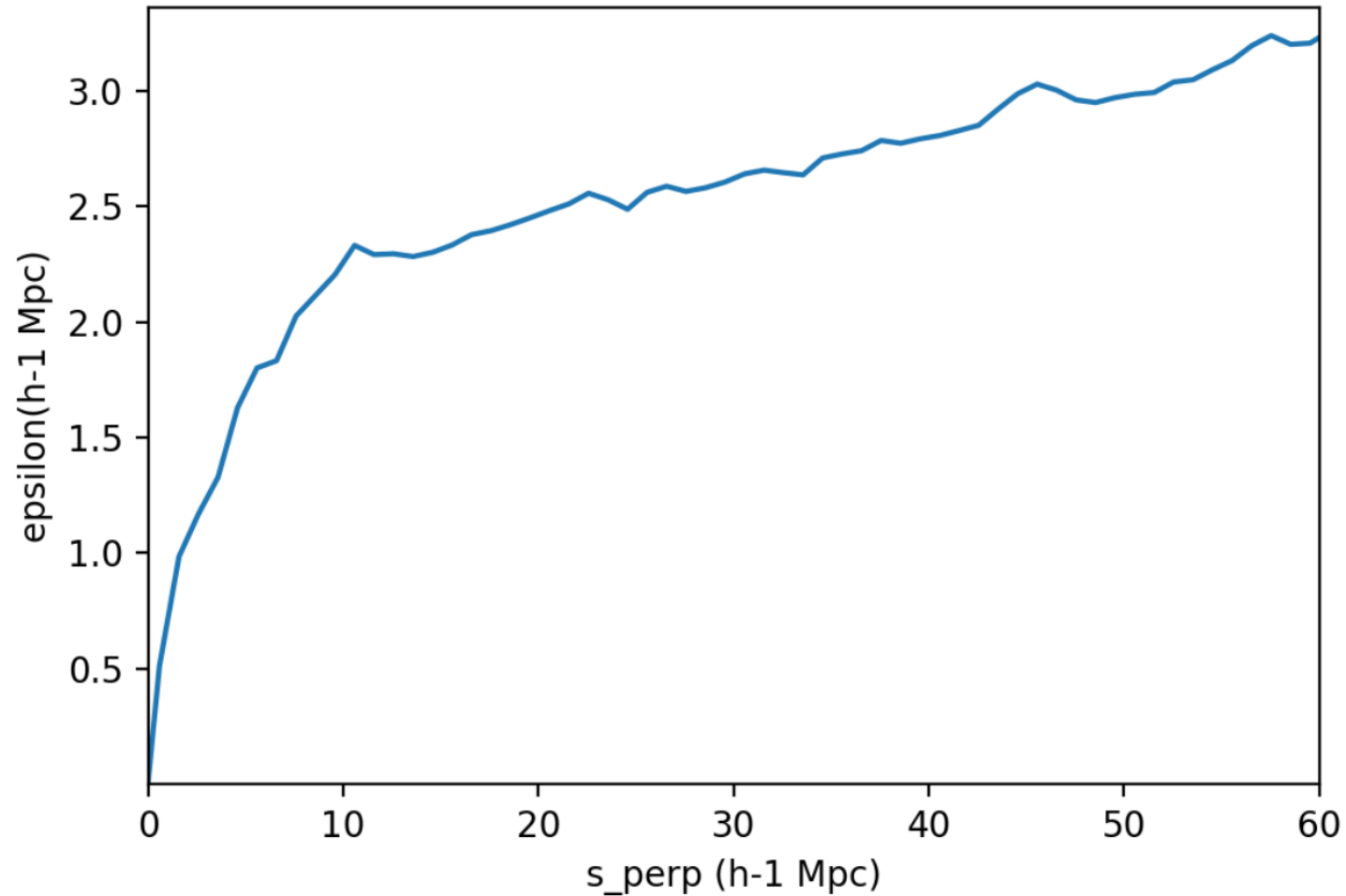
Comparison mocks with and without RSD

- Impact of RSD on the monopole: 6% difference, 17% expected with the RSD model
- Several effect can change this difference (void finder, tomographic mapping)



Measuring the tomographic effect

- Cumulative histogram ratio between randomly placed voids and voids obtained on the tomographic map



Effect of void and tomographic parameters

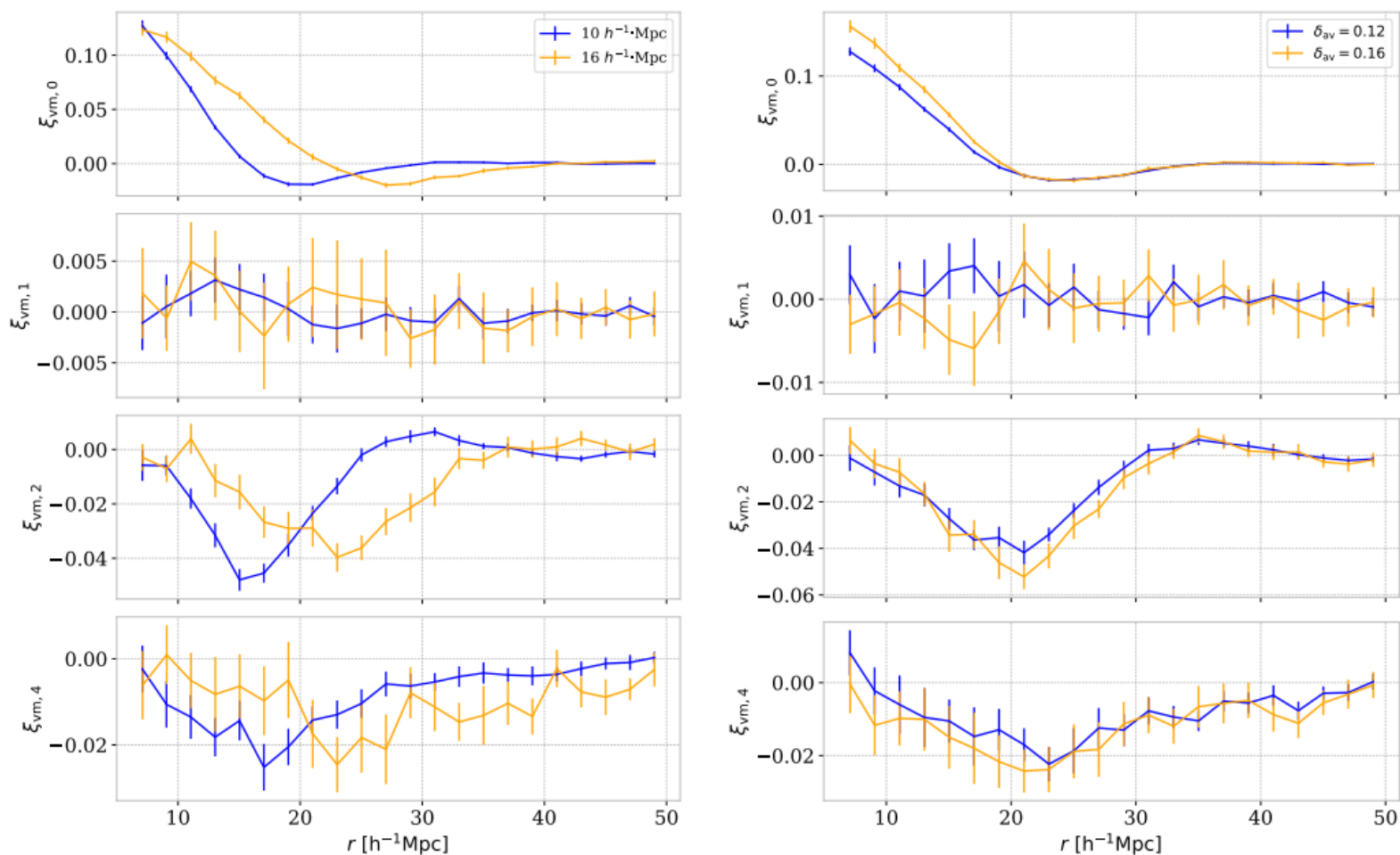
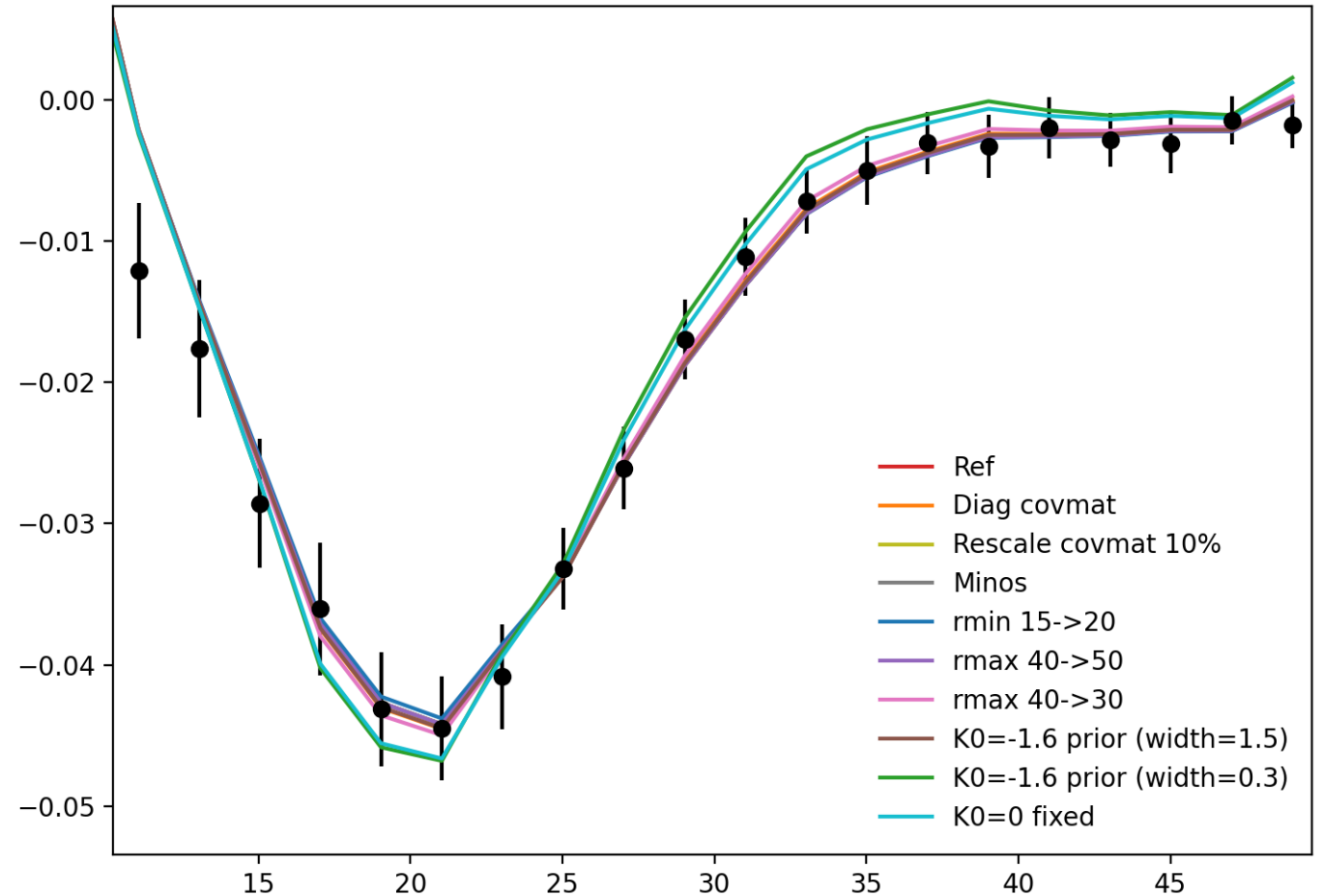


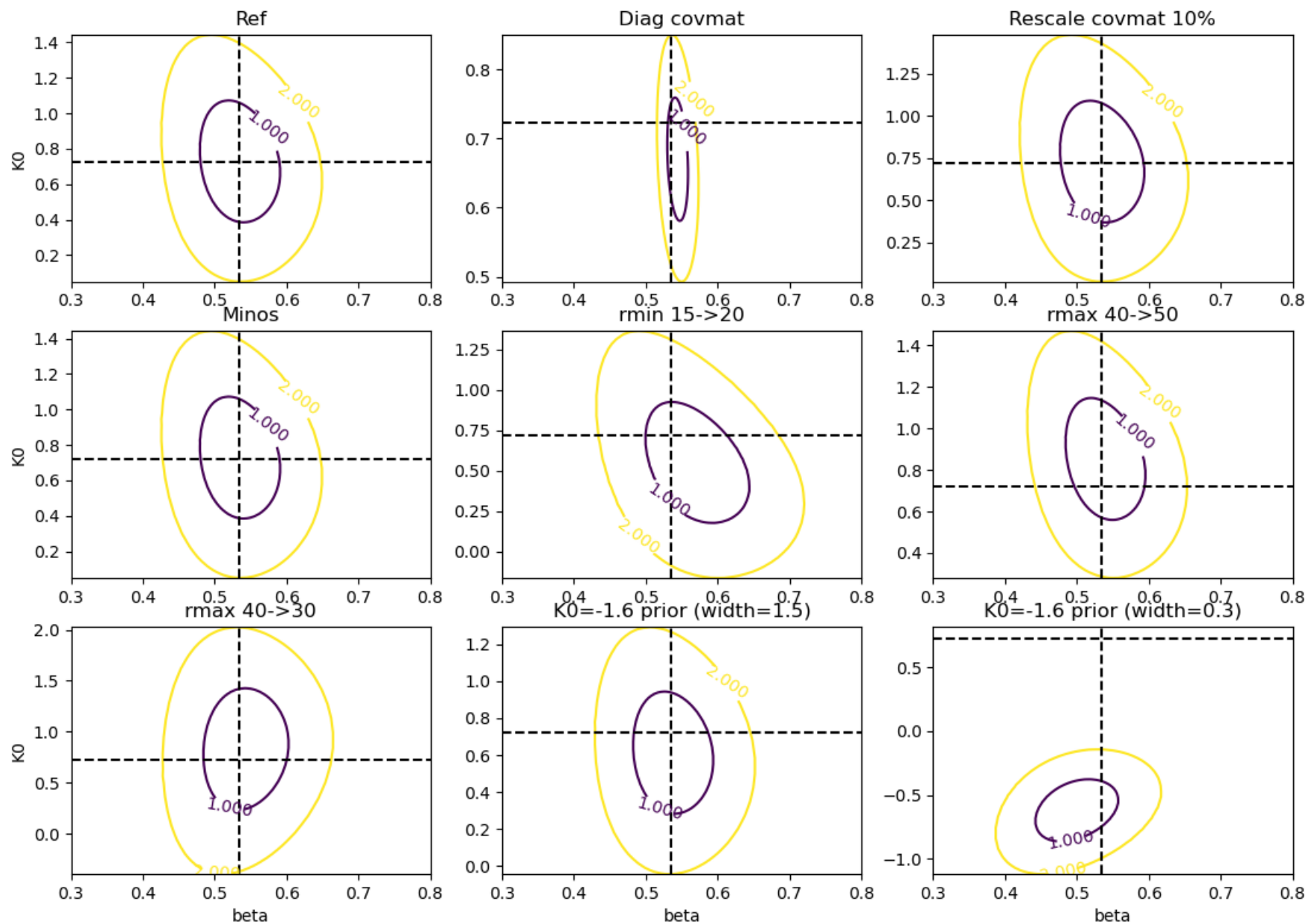
FIGURE 7.13 – Impact of analysis parameters on the *raw-noRSD* mock multipoles. (left) Variation of both L_{\perp} and L_{\parallel} correlation lengths from 10 to 16 $h^{-1}\text{Mpc}$. (right) Variation of spherical void finder parameter δ_{av} from 0.12 to 0.16.

Fit variations

- Different estimates of the error bars (picca, diagonal covariance matrix, picca corrected by mocks)
- Change of fit interval
- Change of nuisance parameter



Fit contours

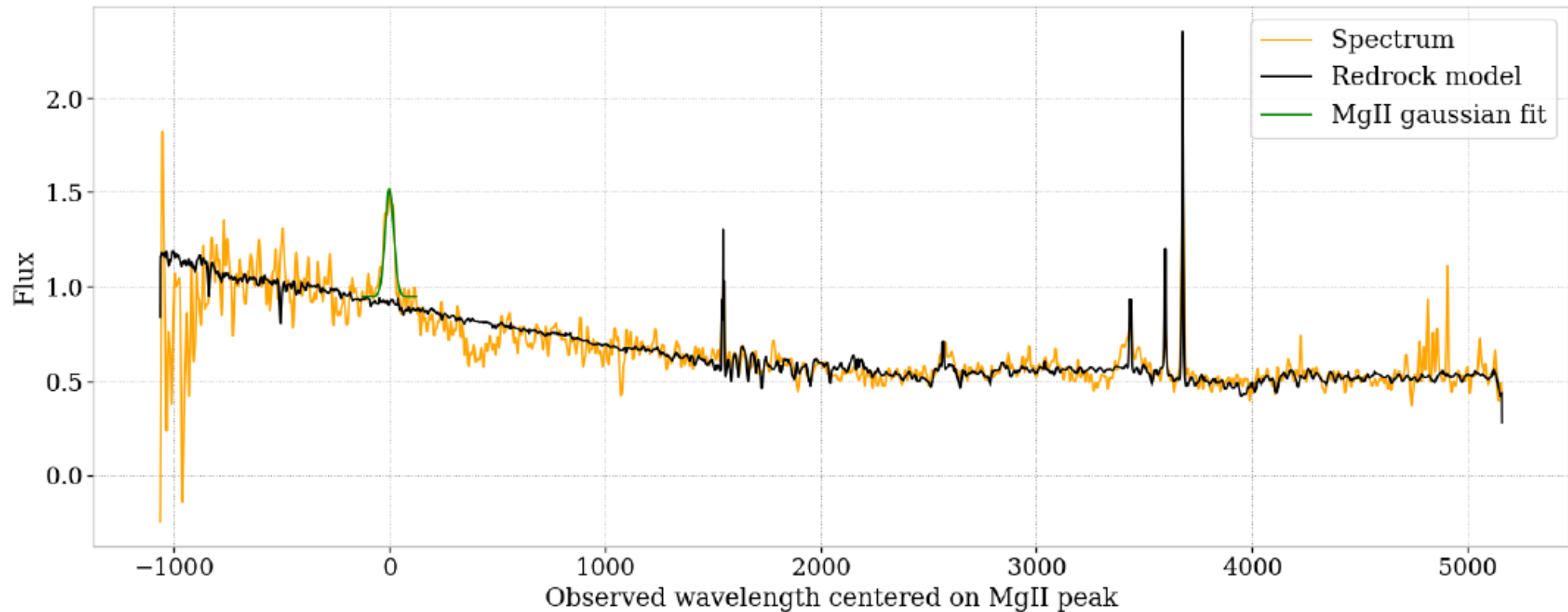


Other tests

- Using shuffle instead of mocks to correct tomographic effect quadrupole result in a 1σ bias of the β value (increase of 0.06)
- Measurement of β on raw mocks decreases β by 1σ (0.05)

Mgii afterburner

- Algorithm to retrieve low-z quasar missed by the main redshift algorithm of DESI
- Included in DESI pipeline
- Increasing completeness of quasars up to 98%



Quasar continuum fitting procedure

- Continuum model

$$C_q(\lambda, z_q) = (a_q + b_q \Lambda) C \left(\lambda_{\text{rf}} = \frac{\lambda}{(1 + z_q)} \right)$$

$$\Lambda = \begin{cases} \log(\lambda) & \text{logarithmic binning (SDSS)} \\ \lambda & \text{linear binning (DESI)} \end{cases}$$

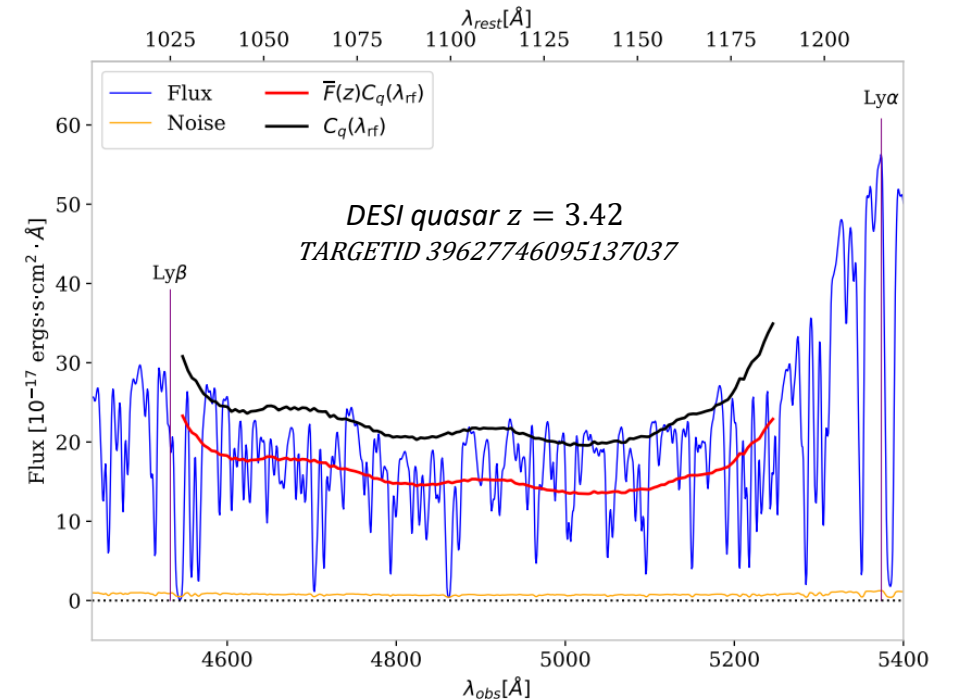
- Likelihood minimization

$$\mathcal{L} = - \sum_i \frac{[f_i - \bar{F}(\lambda_i) C_q(\lambda_i, z_q, a_q, b_q)]^2}{\sigma_q^2(\lambda_i)} - \ln [\sigma_q^2(\lambda_i)]$$

$$\frac{\sigma_q^2(\lambda)}{(\bar{F}(\lambda) C_q(\lambda))^2} = \eta(\lambda) \frac{\sigma_{\text{pip},q}^2(\lambda)}{(\bar{F}(\lambda) C_q(\lambda))^2} + \sigma_{\text{lss}}^2(\lambda) + \epsilon(\lambda) \frac{(\bar{F}(\lambda) C_q(\lambda))^2}{\sigma_{\text{pip},q}^2(\lambda)}$$

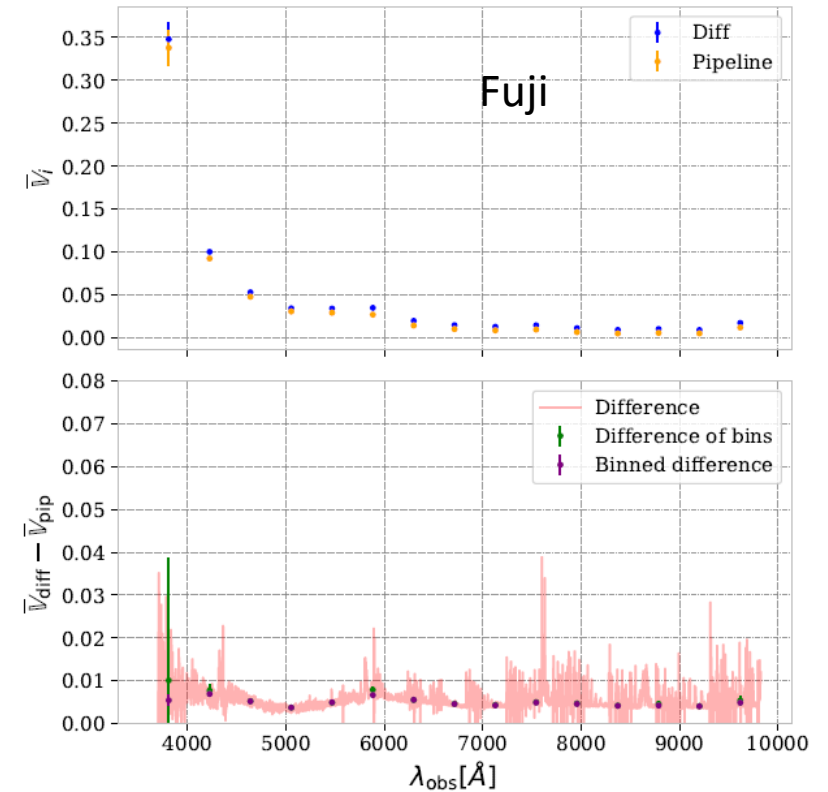
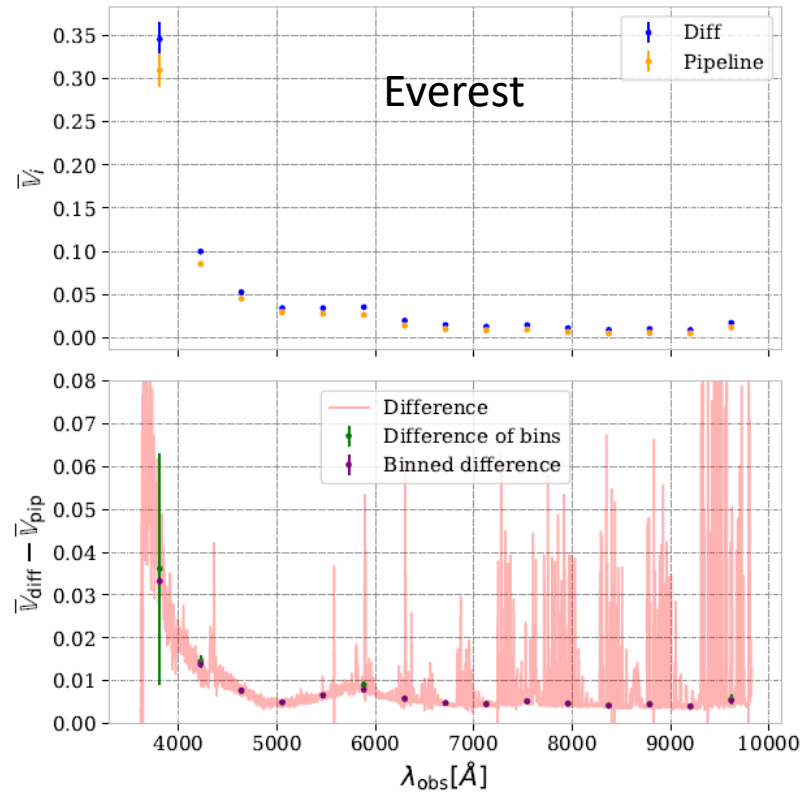
- Noise associated:

$$\sigma_{\delta_F}(\lambda) = \eta(\lambda) \frac{\sigma_{\text{pip},q}(\lambda)}{\bar{F}(\lambda) C_q(\lambda)}$$



Diff noise on spectra

- Another way to help assessing noise level
- Use quasar and LRG spectra with several exposures:
 - Calculate average difference between exposures
- Improvement between Everest (left) and Fuji (right) data reduction due to accounting CCD position dependence



$P_{1D,\alpha}$ diff noise

- Diff noise estimator

$$\Delta f_j = \frac{1}{2} \left(\frac{\sum_{k=1}^{N_{\text{even}}} (\mathbb{V}_{\text{pip},k})^{-1} f_k}{\sum_{k=1}^{N_{\text{even}}} (\mathbb{V}_{\text{pip},k})^{-1}} - \frac{\sum_{k=1}^{N_{\text{odd}}} (\mathbb{V}_{\text{pip},k})^{-1} f_k}{\sum_{k=1}^{N_{\text{odd}}} (\mathbb{V}_{\text{pip},k})^{-1}} \right)$$

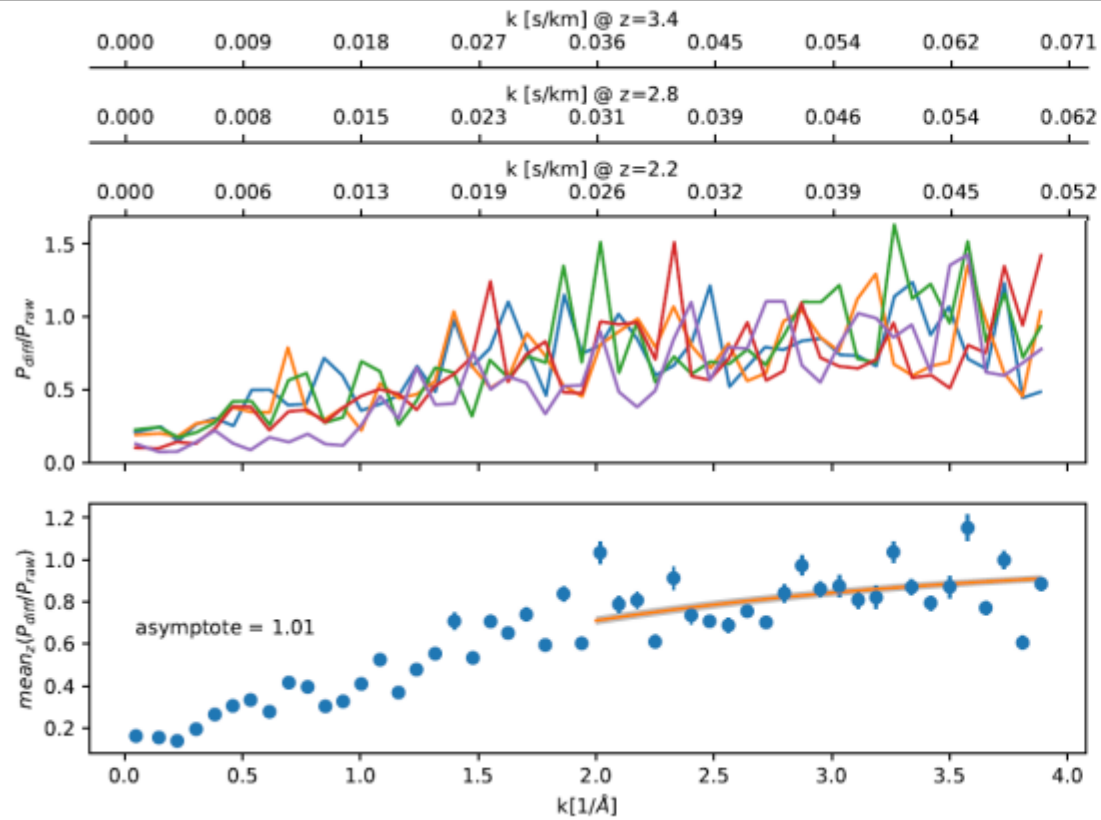
$$\sigma_{\Delta f_j} = \frac{1}{2} \sqrt{\frac{1}{\sum_{k=1}^{N_{\text{even}}} (\mathbb{V}_{\text{pip},k})^{-1}} + \frac{1}{\sum_{k=1}^{N_{\text{odd}}} (\mathbb{V}_{\text{pip},k})^{-1}}}$$

$$\Delta f_j^{\text{corr}} = 2 \frac{\frac{1}{\sqrt{\sum_{k=1}^{N_{\text{tot}}} (\mathbb{V}_{\text{pip},k})^{-1}}}}{\sqrt{\frac{1}{\sum_{k=1}^{N_{\text{even}}} (\mathbb{V}_{\text{pip},k})^{-1}} + \frac{1}{\sum_{k=1}^{N_{\text{odd}}} (\mathbb{V}_{\text{pip},k})^{-1}}}} \Delta f_j$$

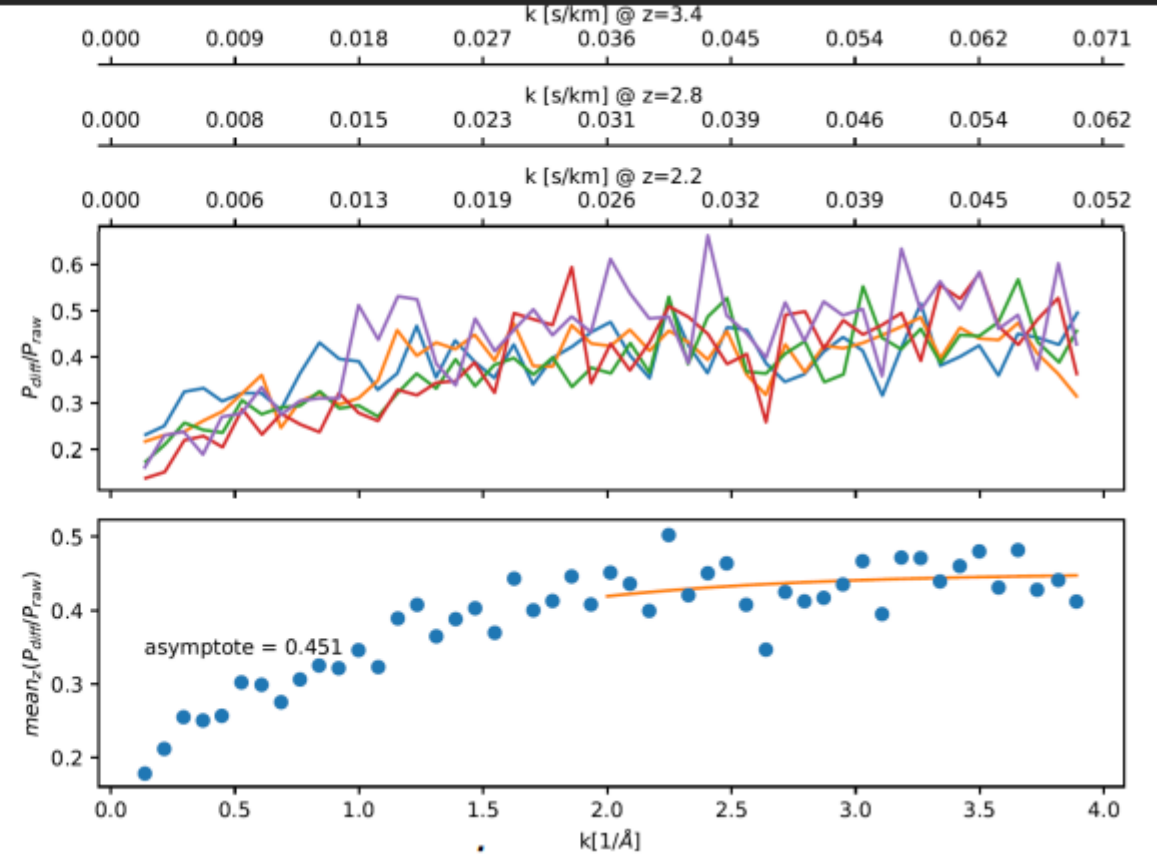
$$P_{\text{diff}}(k) = \left| \mathcal{F} \left[\frac{\Delta f_j^{\text{corr}}(\lambda)}{\overline{F}(\lambda) C_q(\lambda)} - 1 \right] \right|^2$$

Detection of DESI pipeline noise issue

Andes

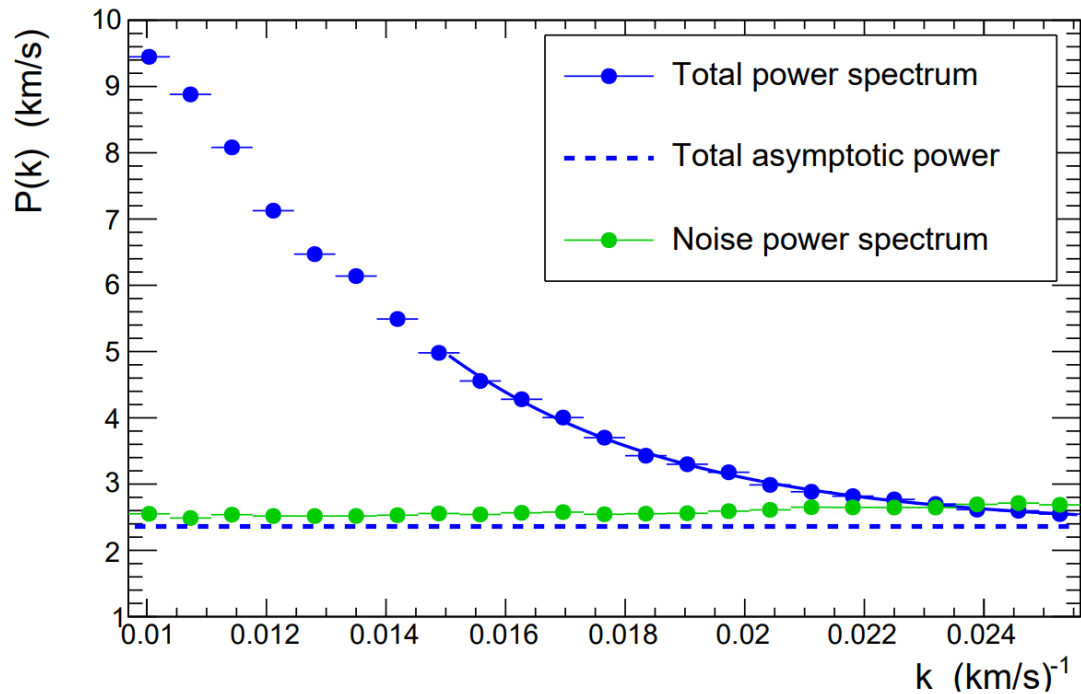


Blanc

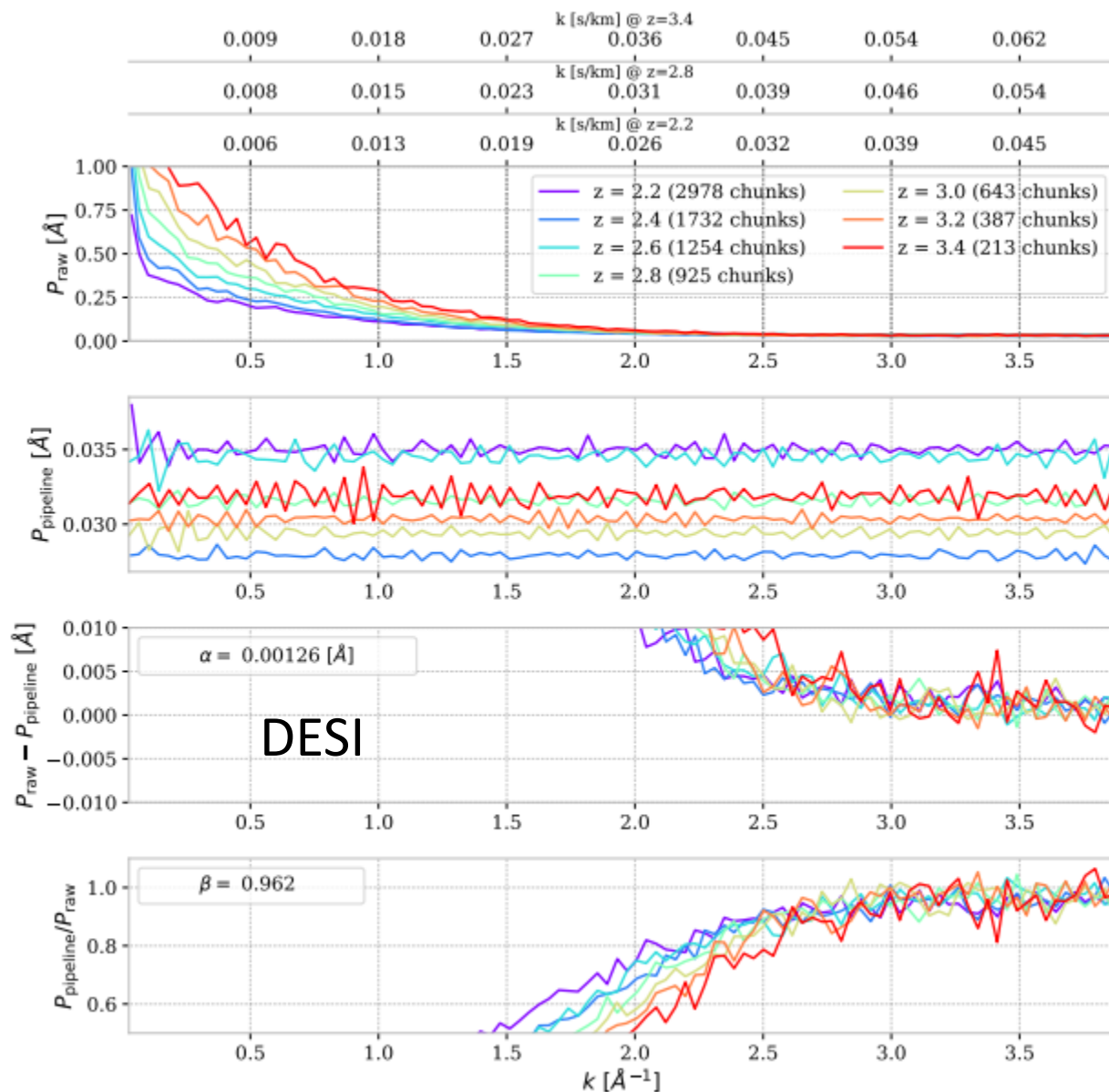


Asymptotic measurement

- Example of measurement on SV3 vs eBOSS

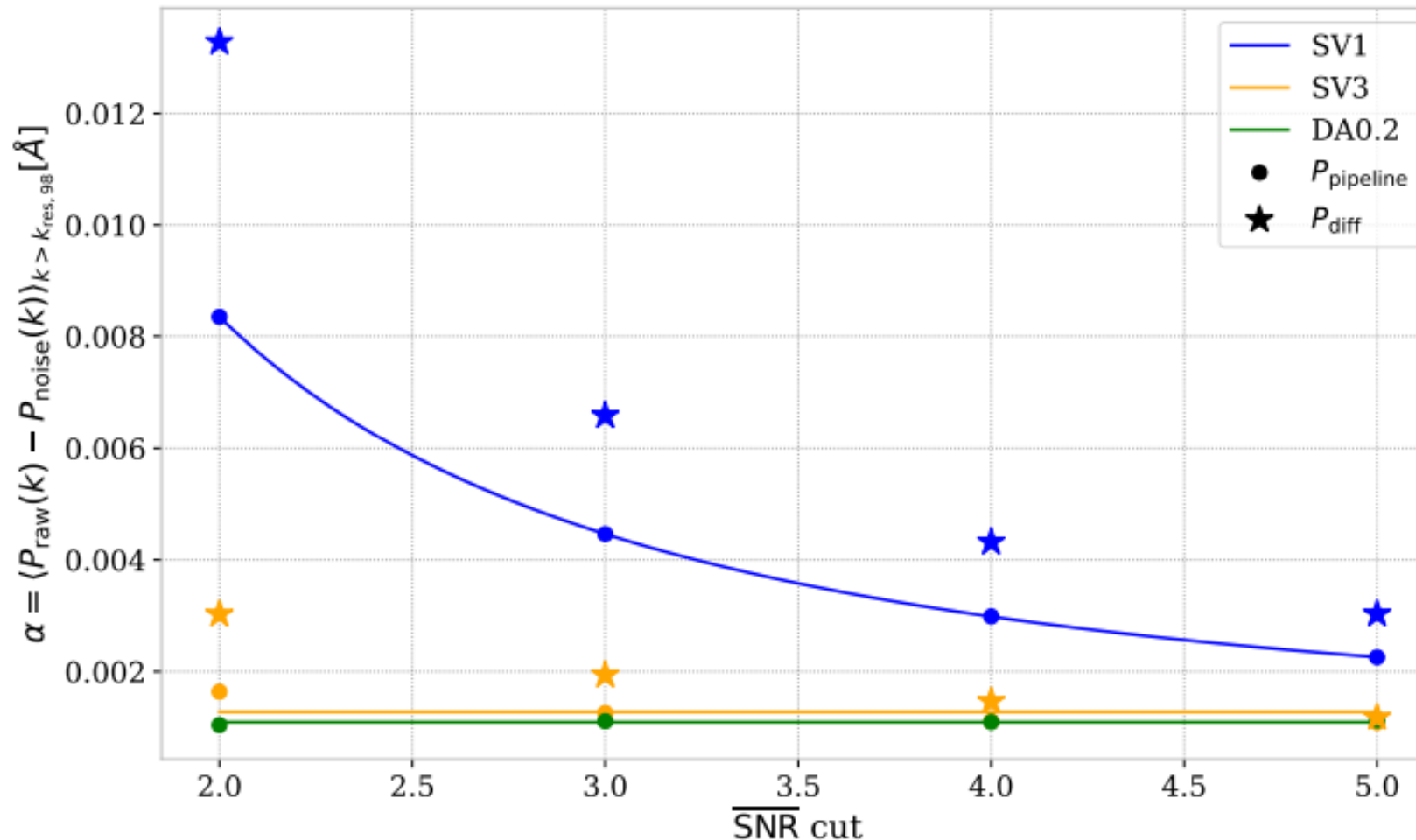


eBOSS



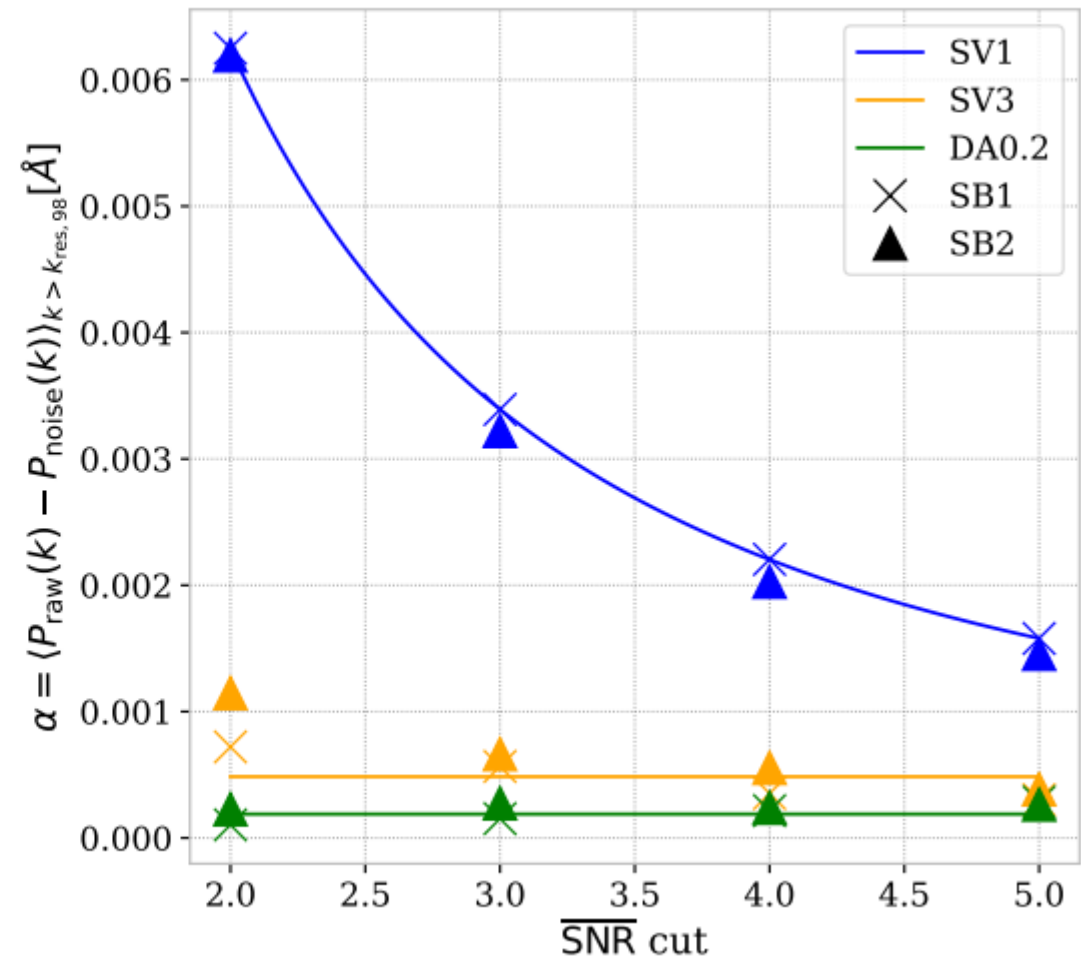
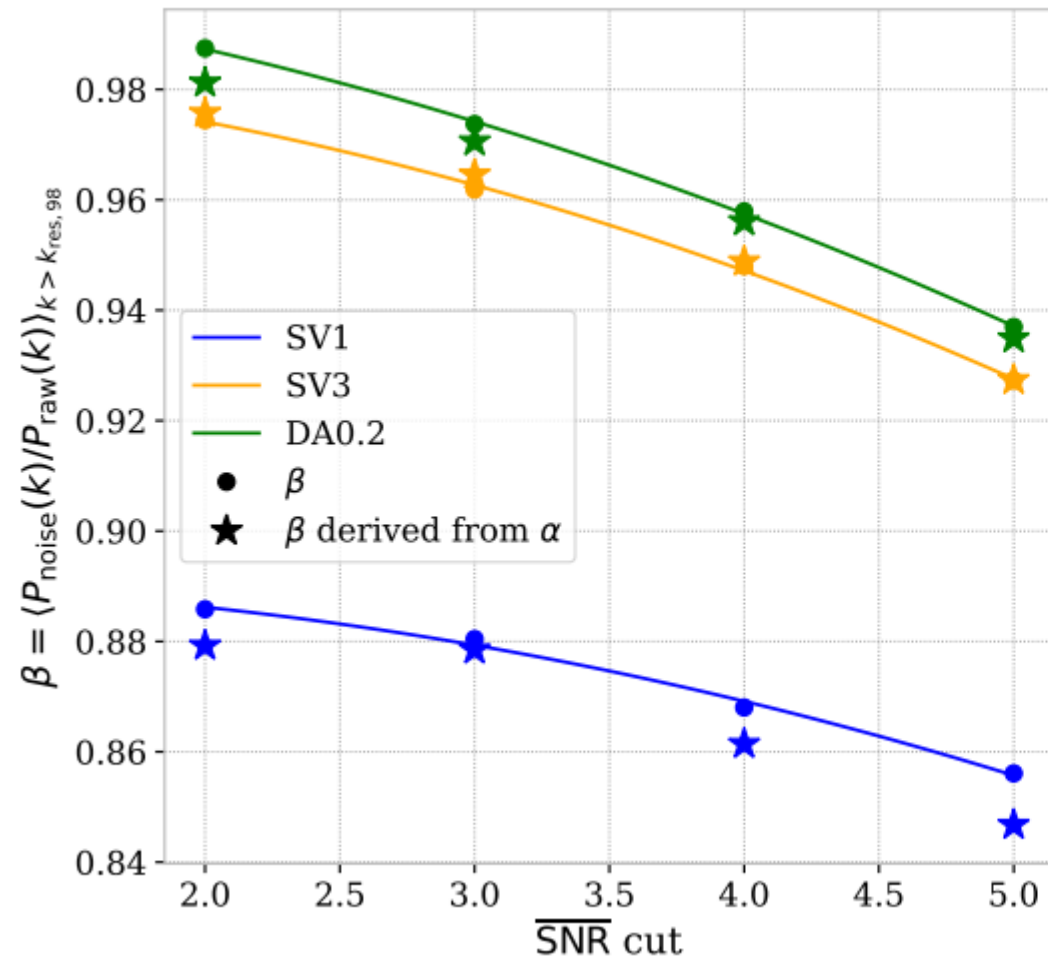
Comparison diff and pipeline noise on last reductions

- Power spectrum difference (raw and noise)



Comparison diff and pipeline noise on last reductions

- Power spectrum ratio and side bands measurements

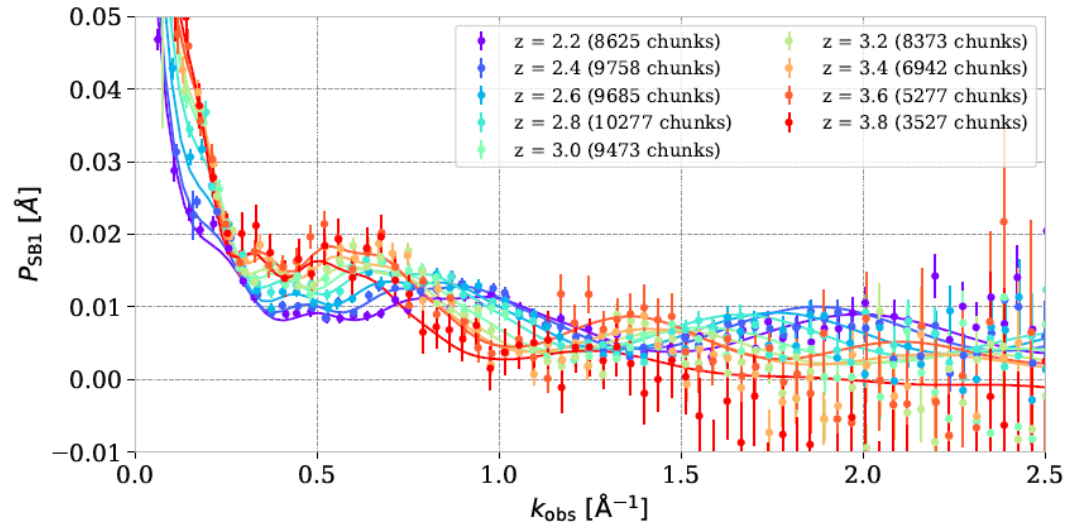
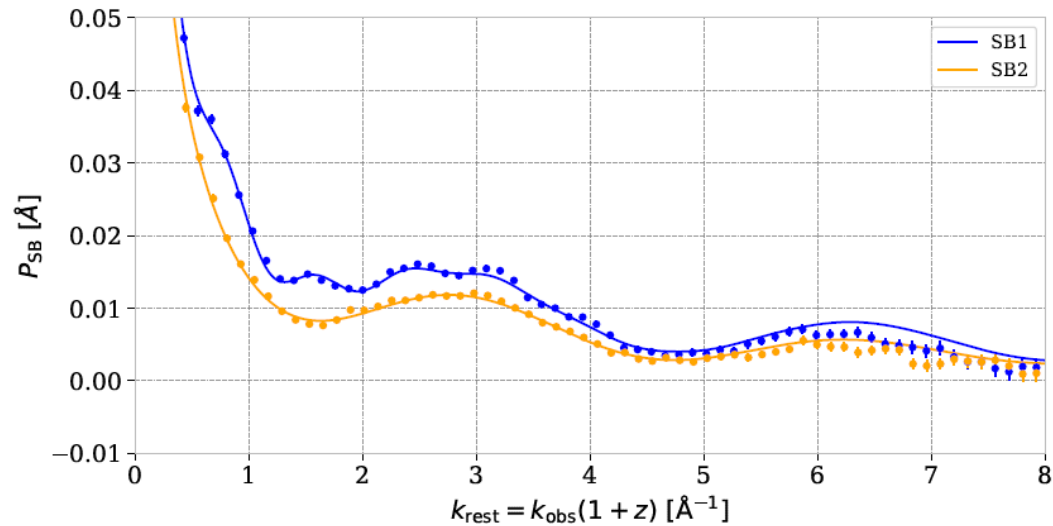


Noise correction: conclusion

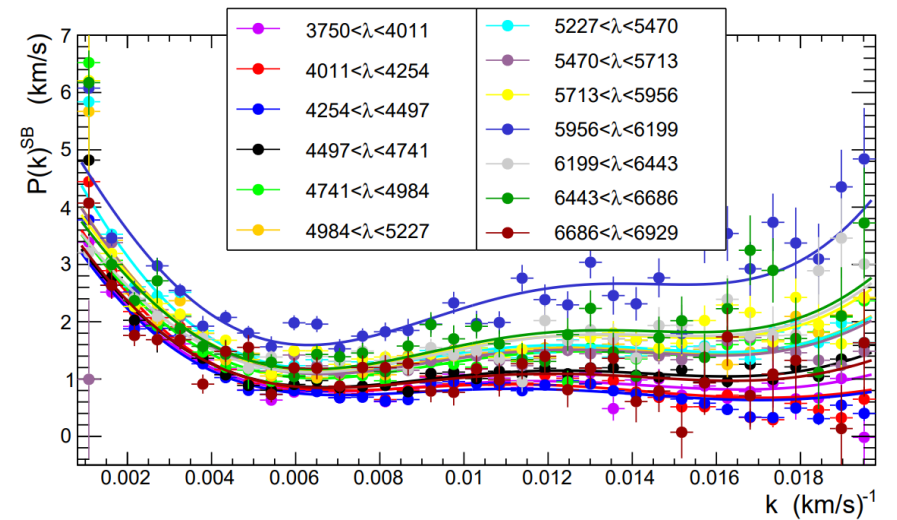
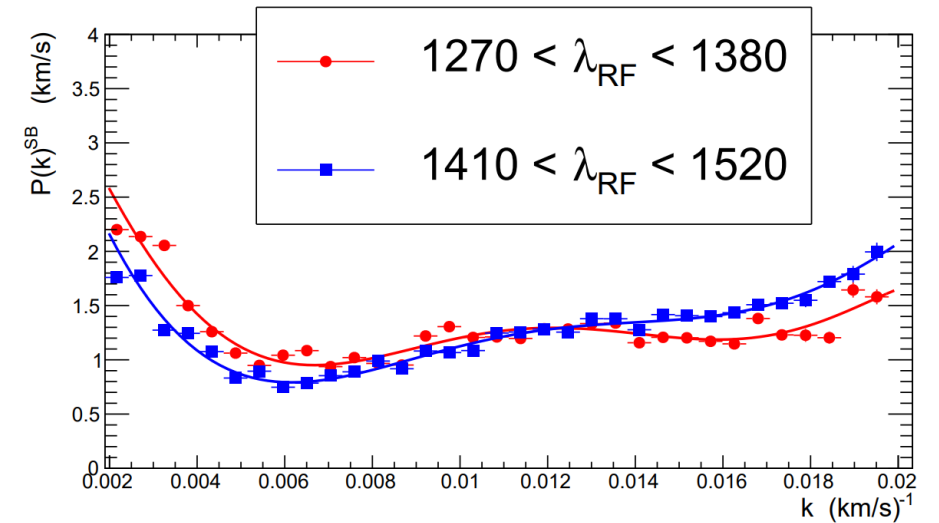
- Asymptote results suggest that the correction is:
 - Mostly additive
 - Survey dependent (related to different number of exposures?)
 - SNR cut-dependent for SV1
- Correction applied:

Band	Data	$P_{\text{noise,miss}} = \alpha \text{ [\AA]}$
Ly α	SV1	$0.026 \times (\overline{\text{SNR}})^{-1.77} + 0.00076$
	SV3	0.00127
	DA0.2	0.00109

$P_{1D,\alpha}$ side bands



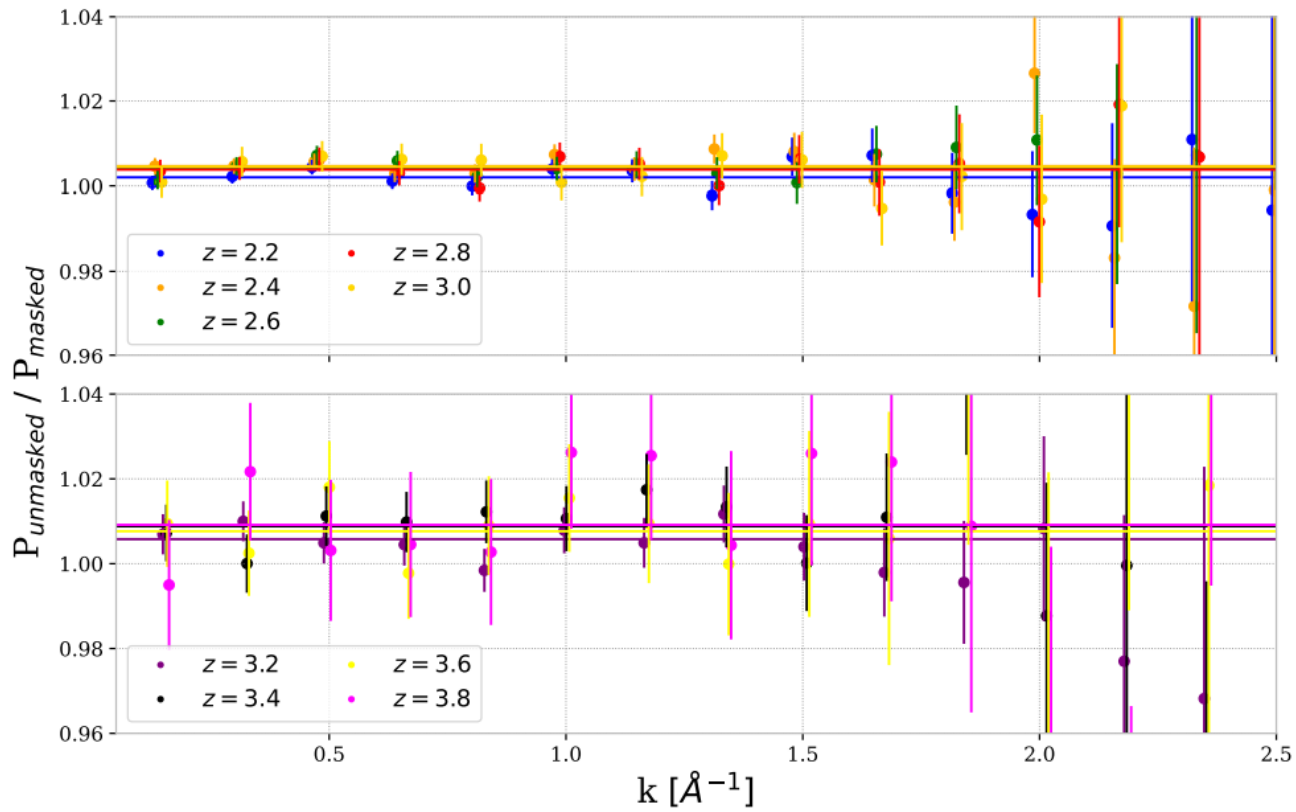
DESI



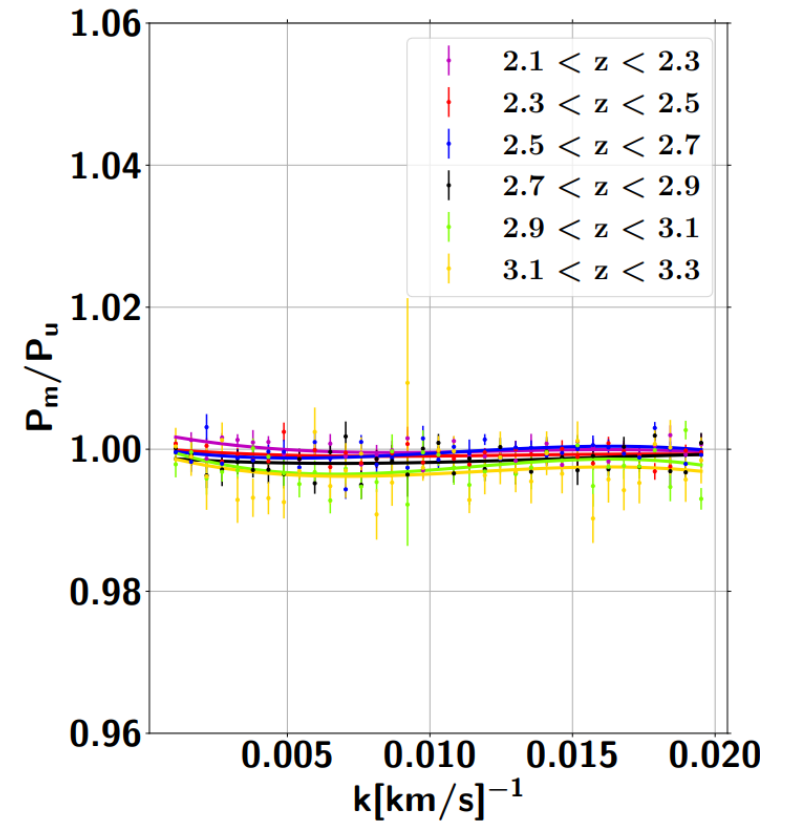
eBOSS

$P_{1D,\alpha}$ mocks

- DLA masking



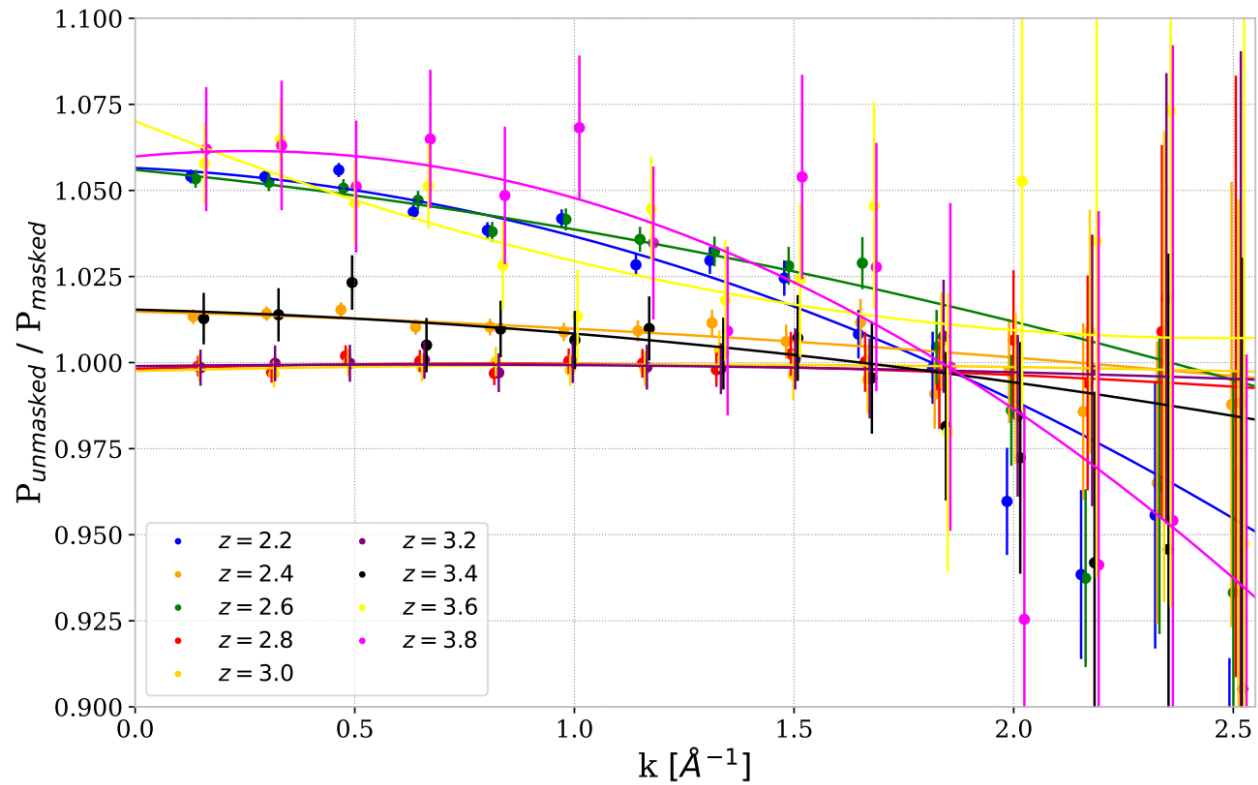
DESI



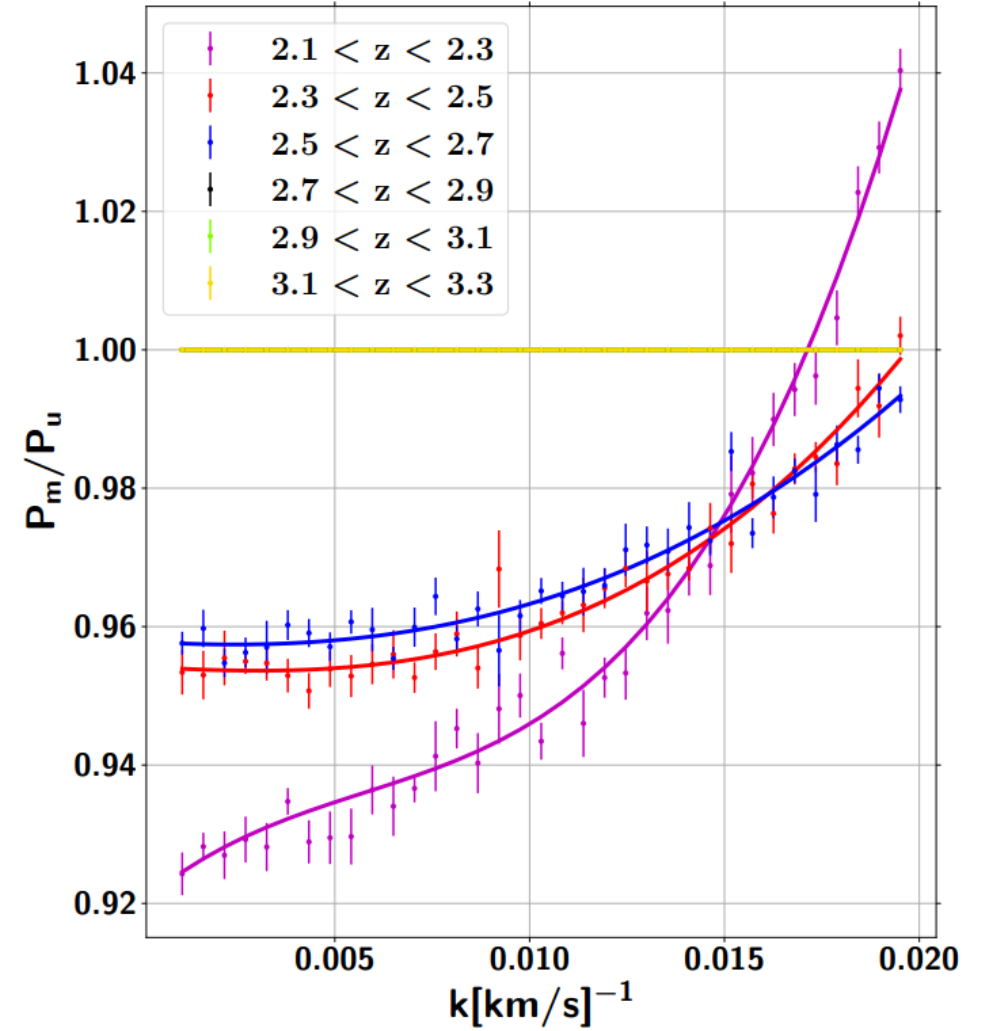
eBOSS

$P_{1D,\alpha}$ mocks

- line masking



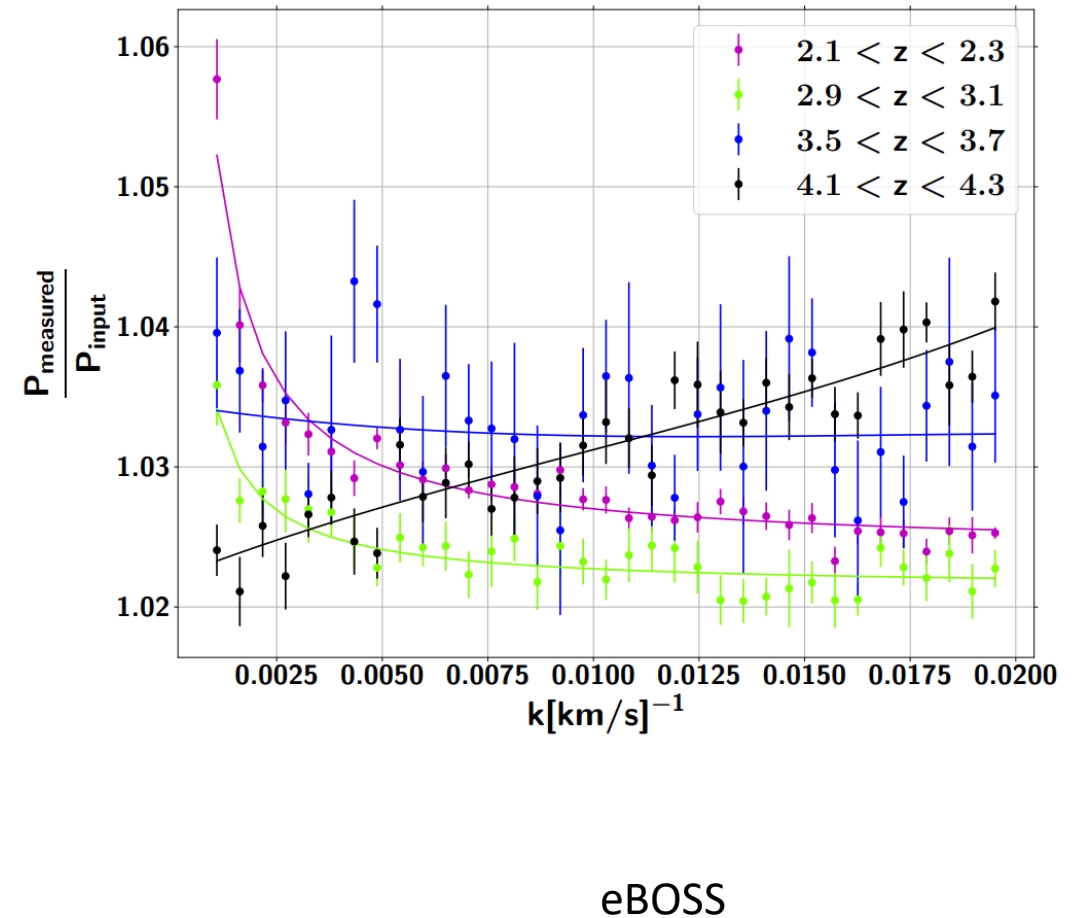
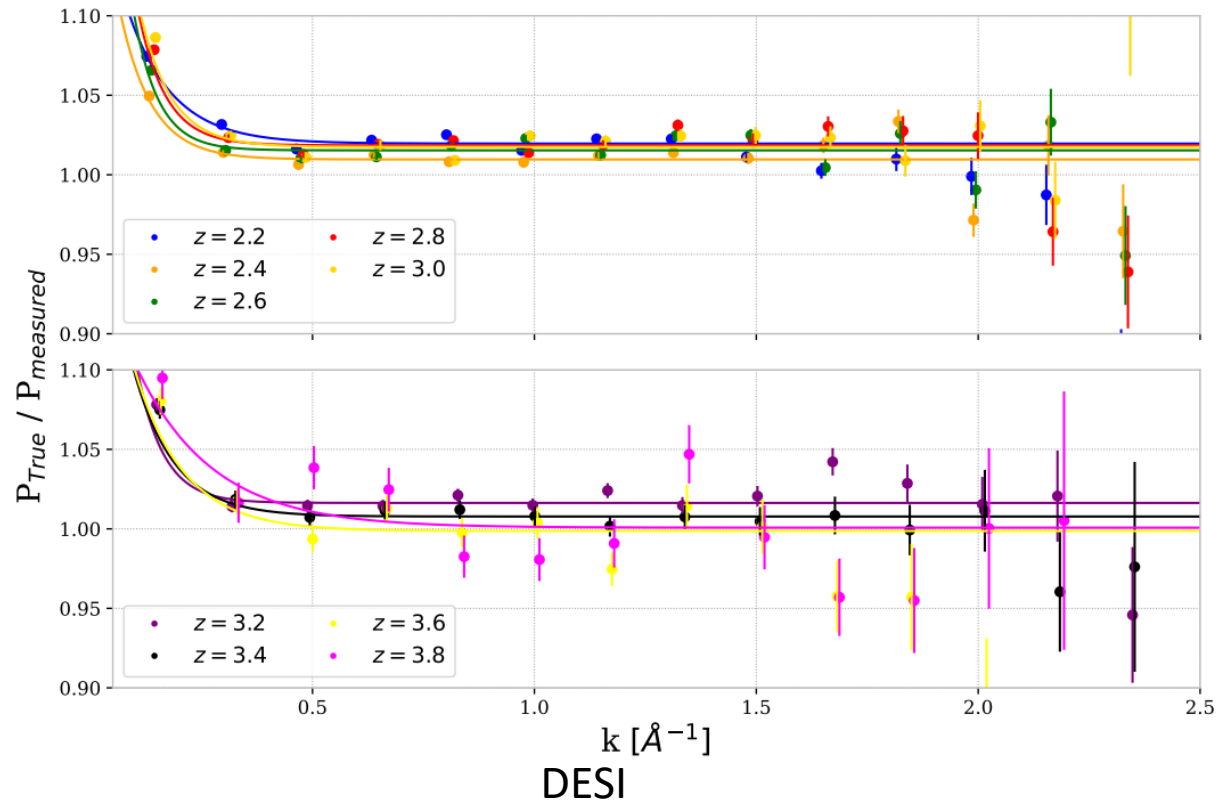
DESI



eBOSS

P1D mocks

- Residual correction (work in progress) which includes:
 - Include possible effect of wrong modeling of resolution matrix (mocks)
 - Continuum fitting errors
 - Possible FFT biases



Final $P_{1D,\alpha}$ model

- Final model implemented:

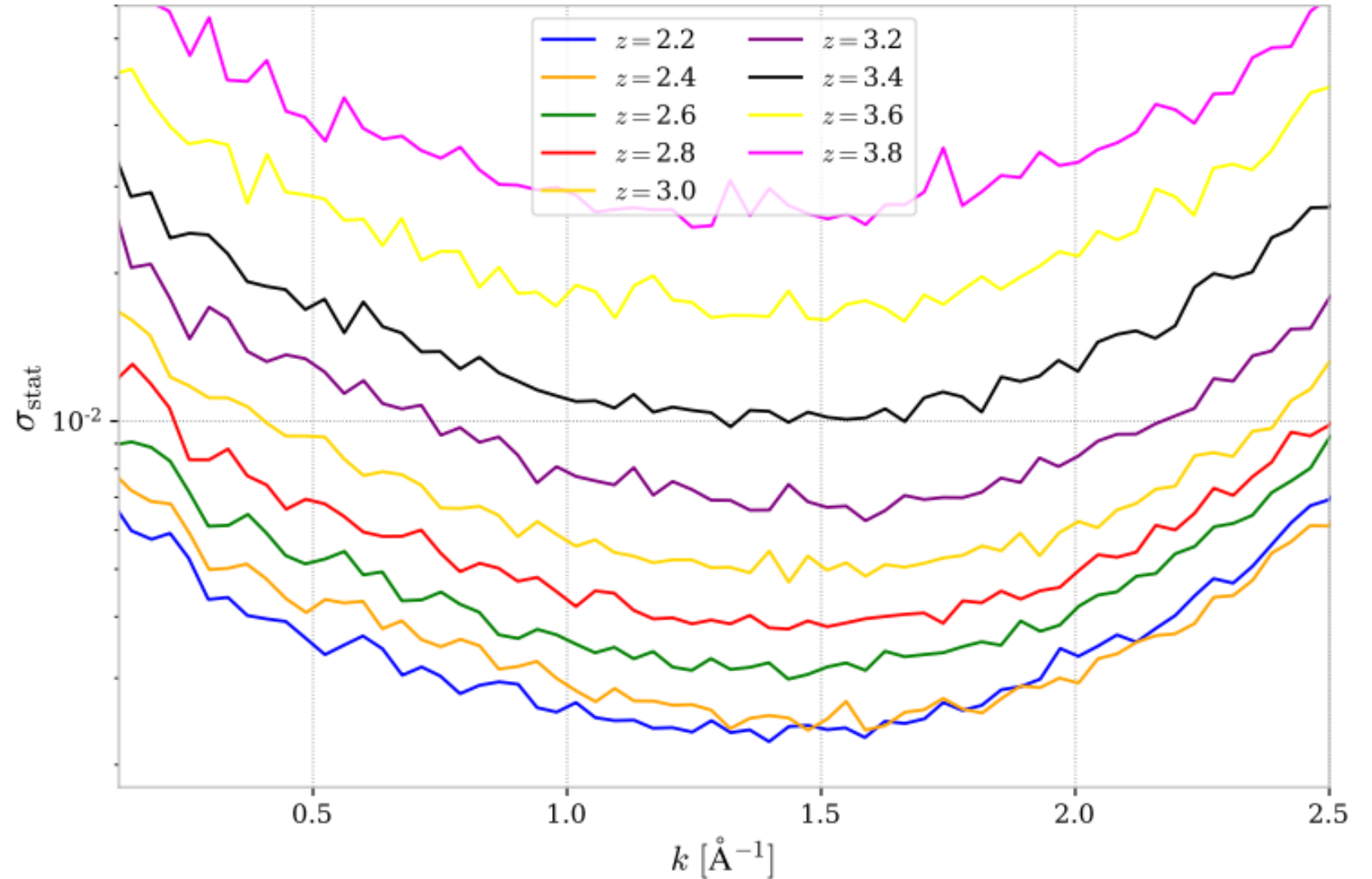
$$P_{1D,\alpha}(k) = A_{\text{sky}}(k, z)A_{\text{residual}}(k, z)A_{\text{hcd}}(k, z) \left(\left\langle \frac{P_{\text{raw}}(k) - P_{\text{noise}}(k) - \alpha(\text{SNR})}{W^2(k, R, \Delta\nu)} \right\rangle - P_{\text{SB1}}(k) \right)$$

- SNR cut chosen redshift-dependent: used of the eBOSS P1D cuts for now.

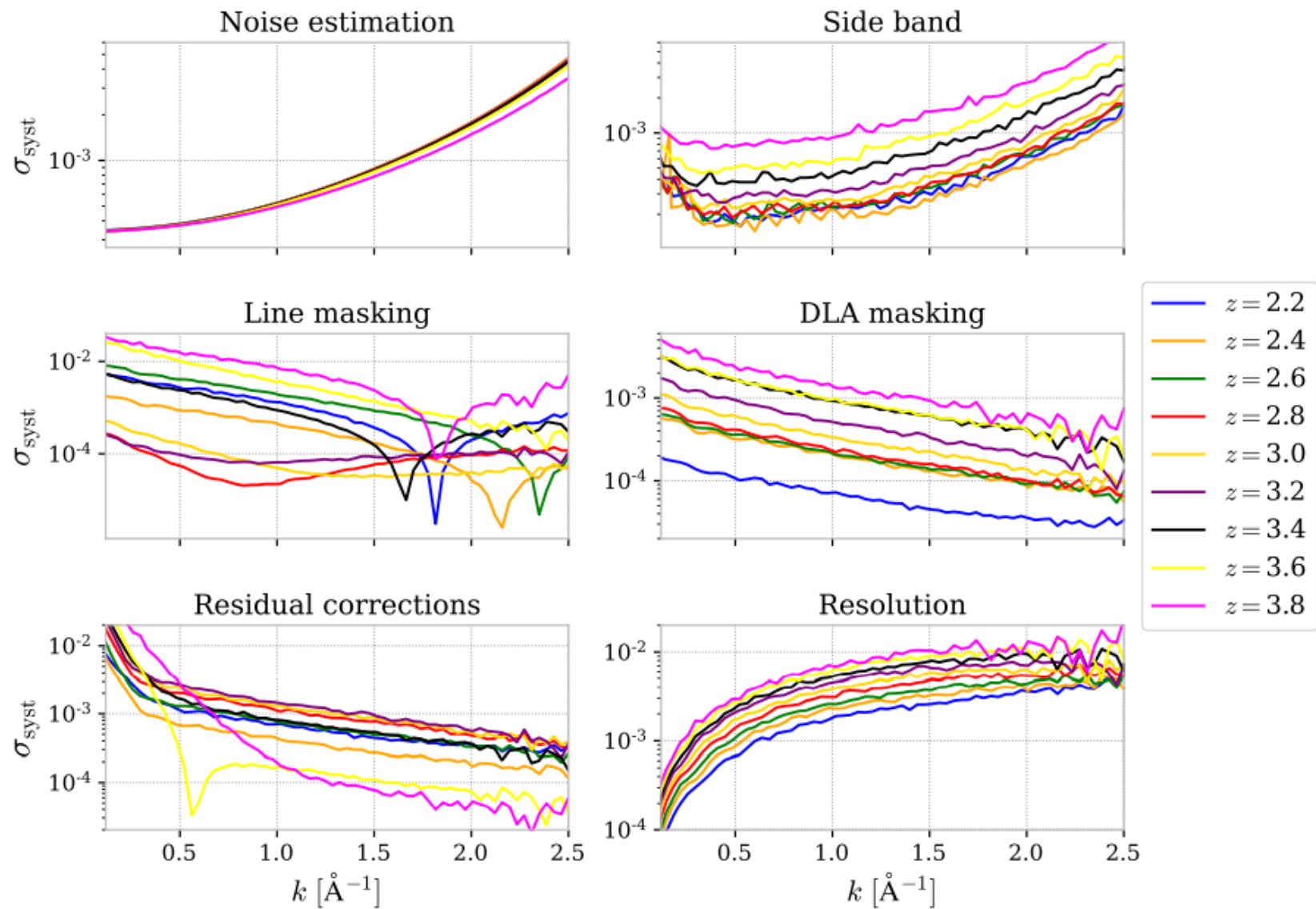
Redshift bin	2.2	2.4	2.6	2.8	3.0	3.2	3.4	> 3.6
SNR threshold	4.1	3.9	3.6	3.2	2.9	2.6	2.2	2.0

$P_{1D,\alpha}$ statistical uncertainties

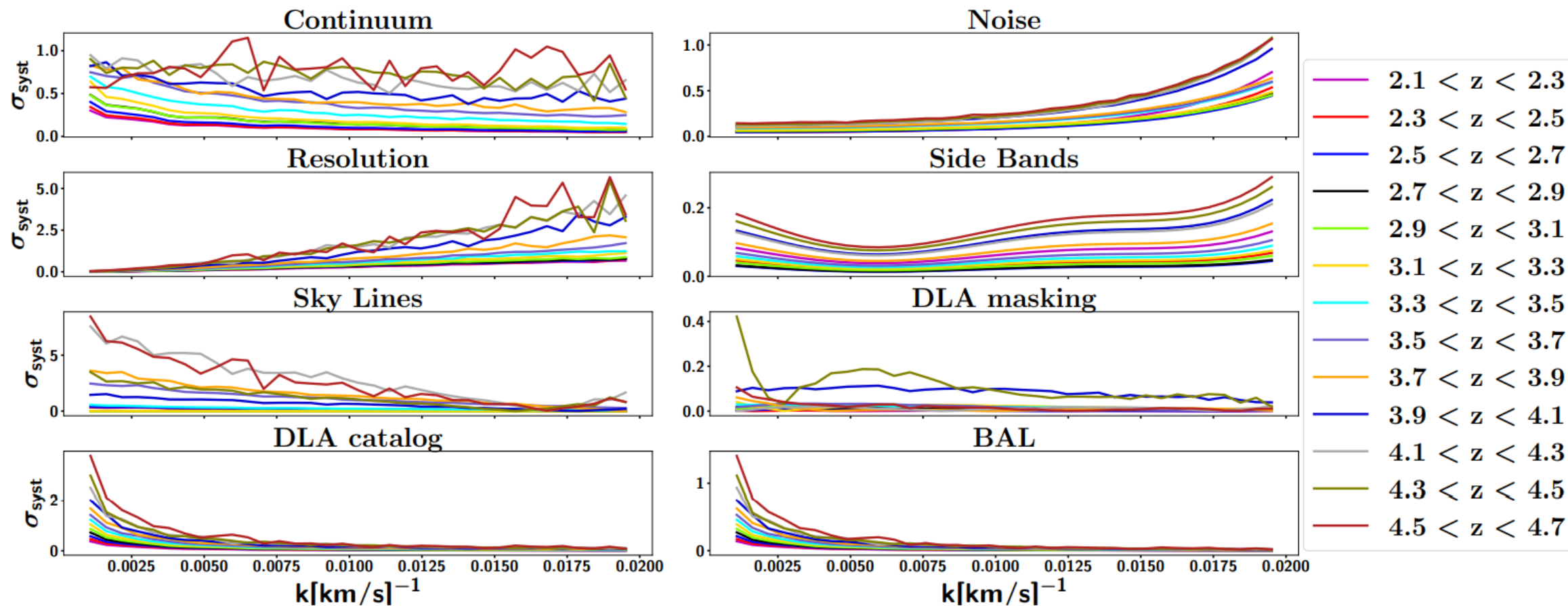
- Larger than eBOSS
- Shape in agreement:
 - Large scales = increasing due to lack of mode with the size of the sub-forest
 - Small scales = noise and resolution



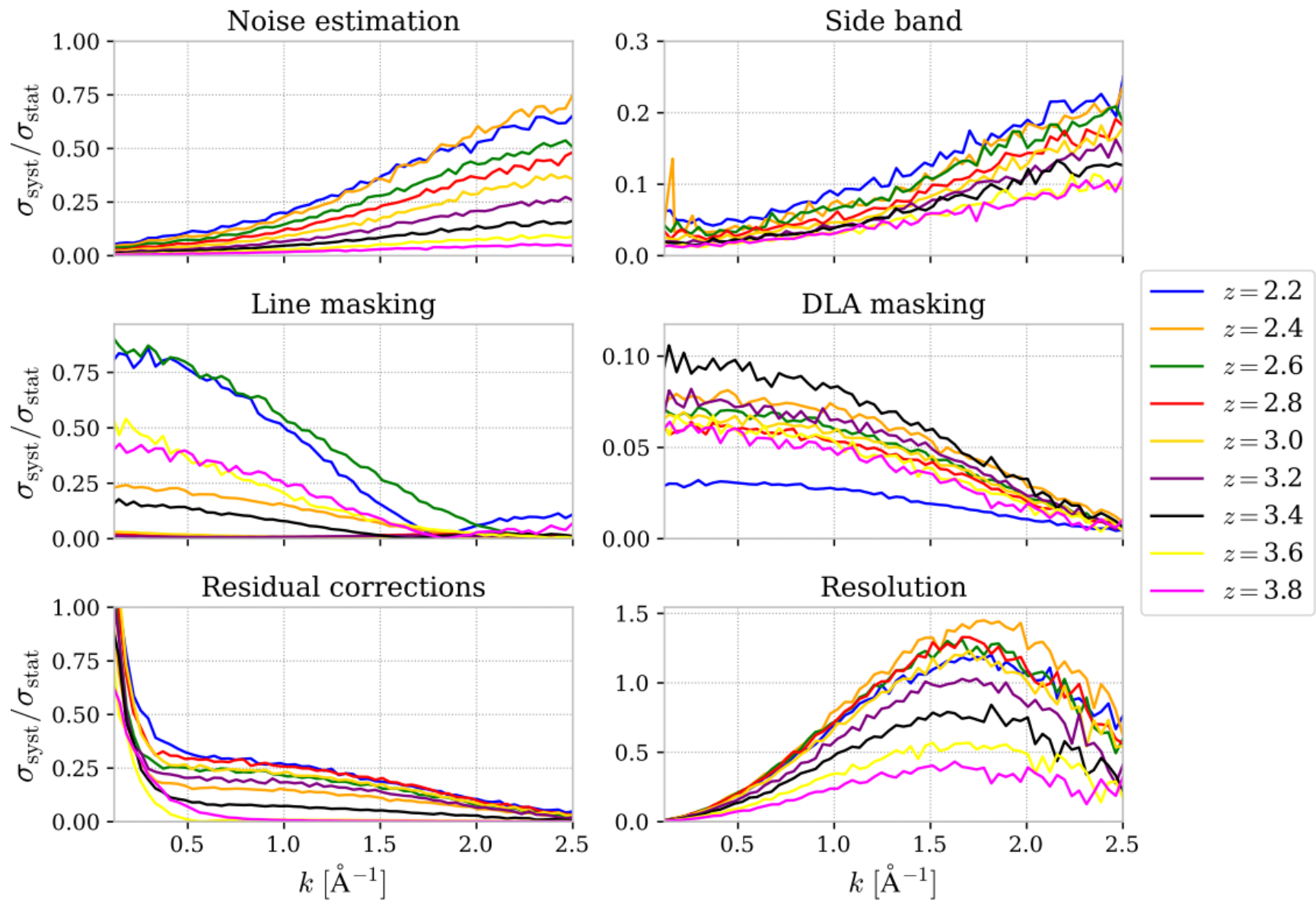
$P_{1D,\alpha}$ DESI systematics and statistics



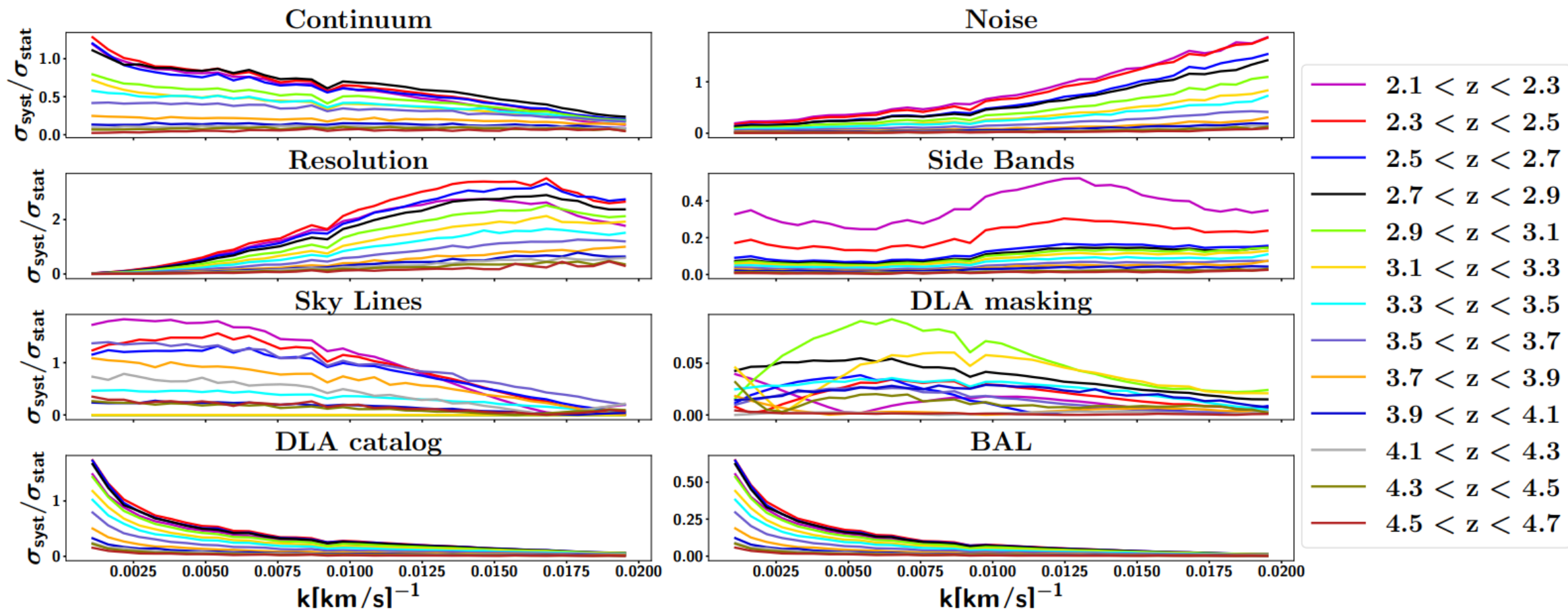
$P_{1D,\alpha}$ eBOSS systematics and statistics



$P_{1D,\alpha}$ DESI systematics and statistics

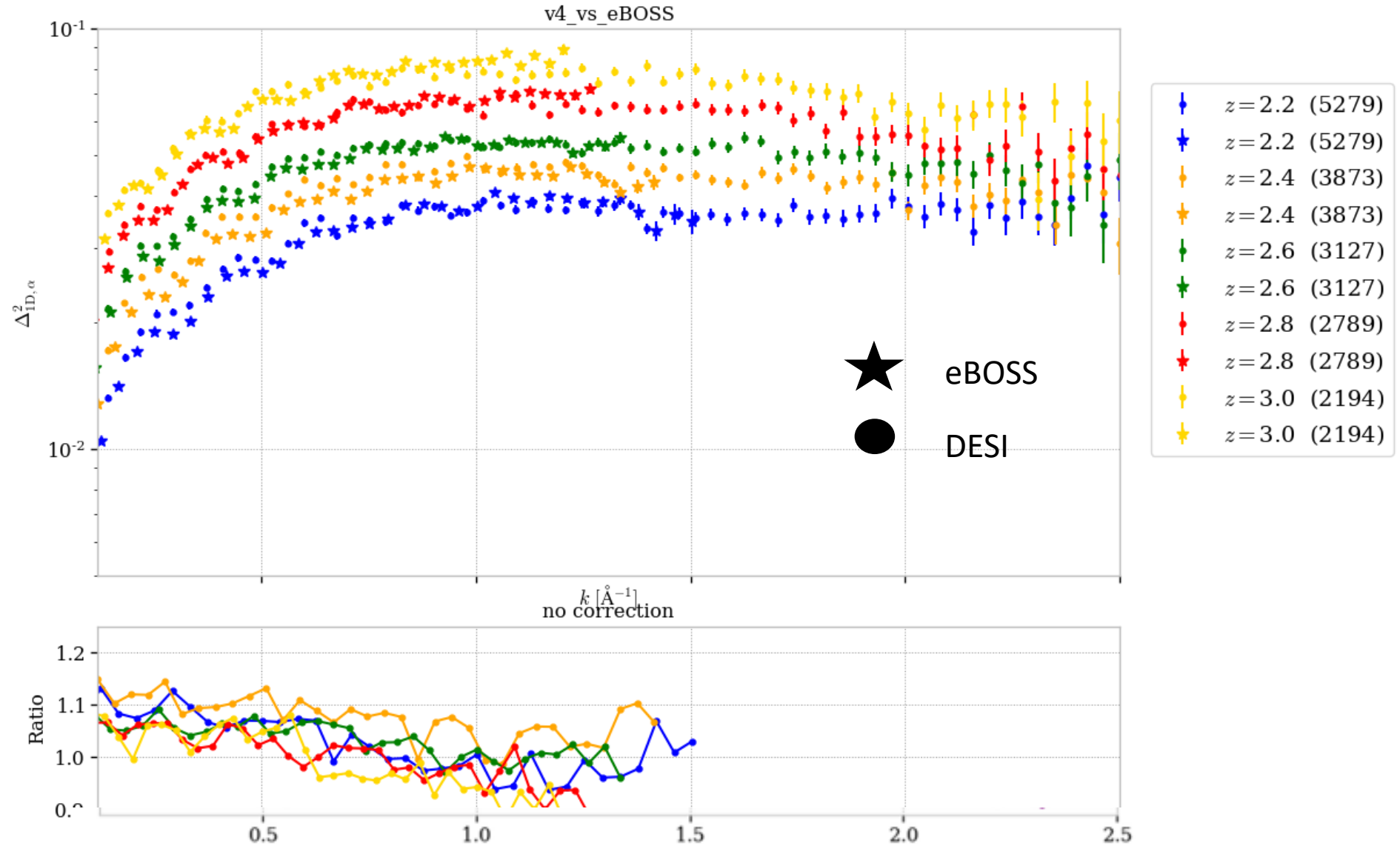


$P_{1D,\alpha}$ eBOSS systematics and statistics



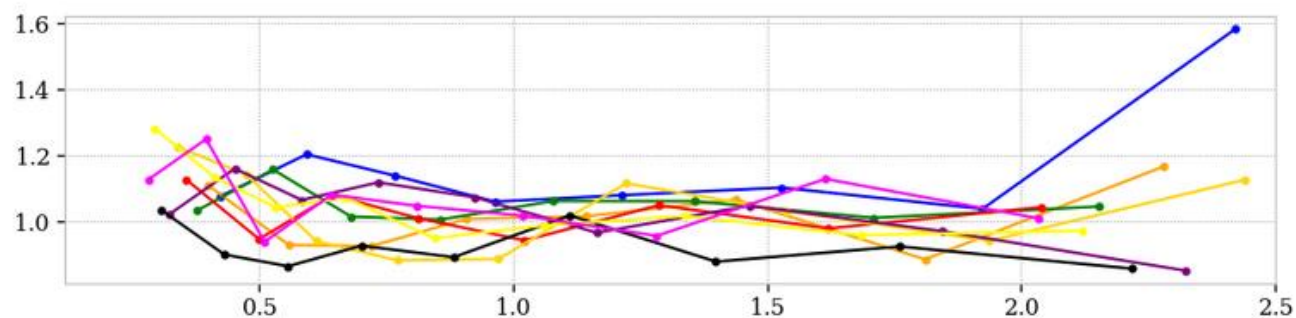
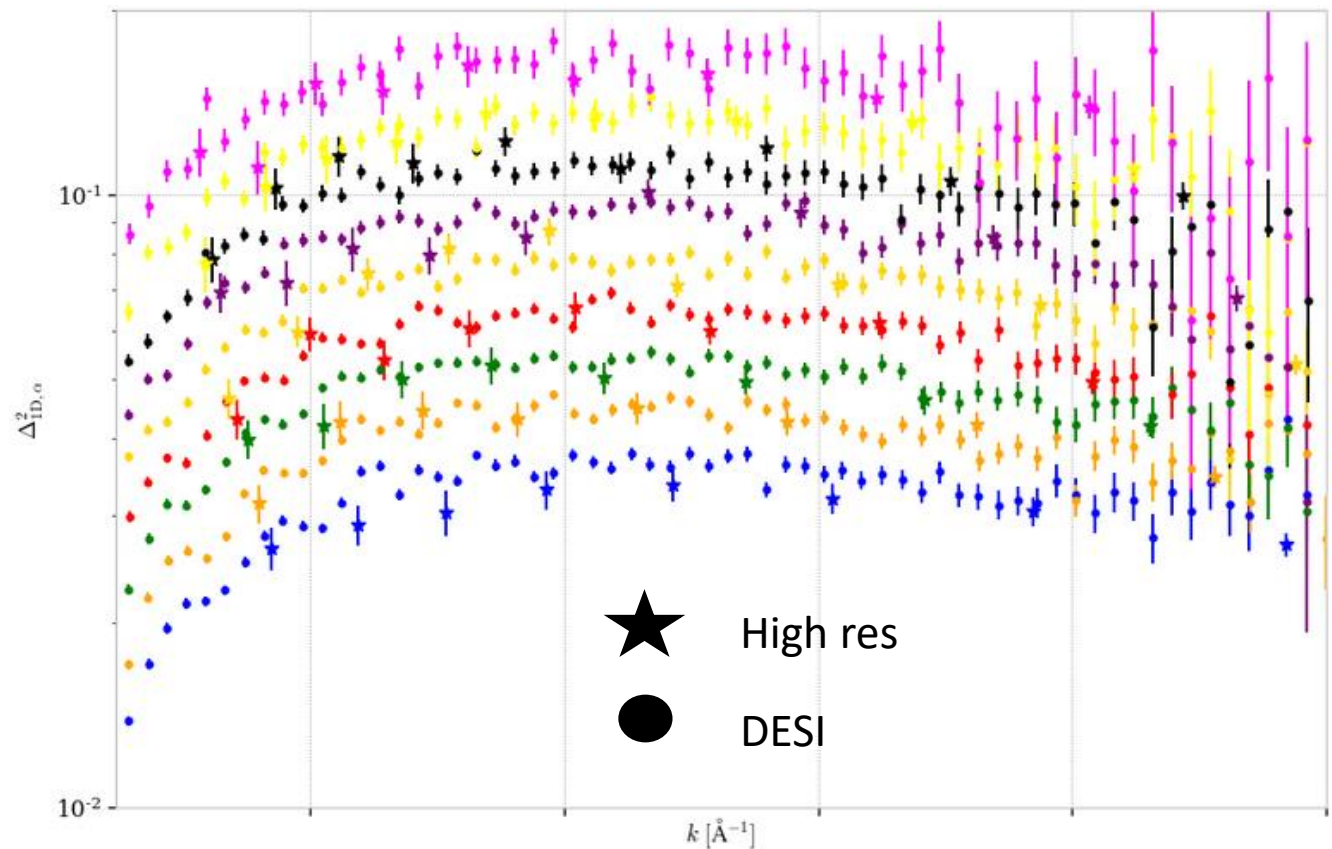
$P_{1D,\alpha}$ comparison eBOSS

- Noise and resolution corrected power spectra
- No other corrections (Metals, masking, ...)



$P_{1D,\alpha}$ comparison high resolution

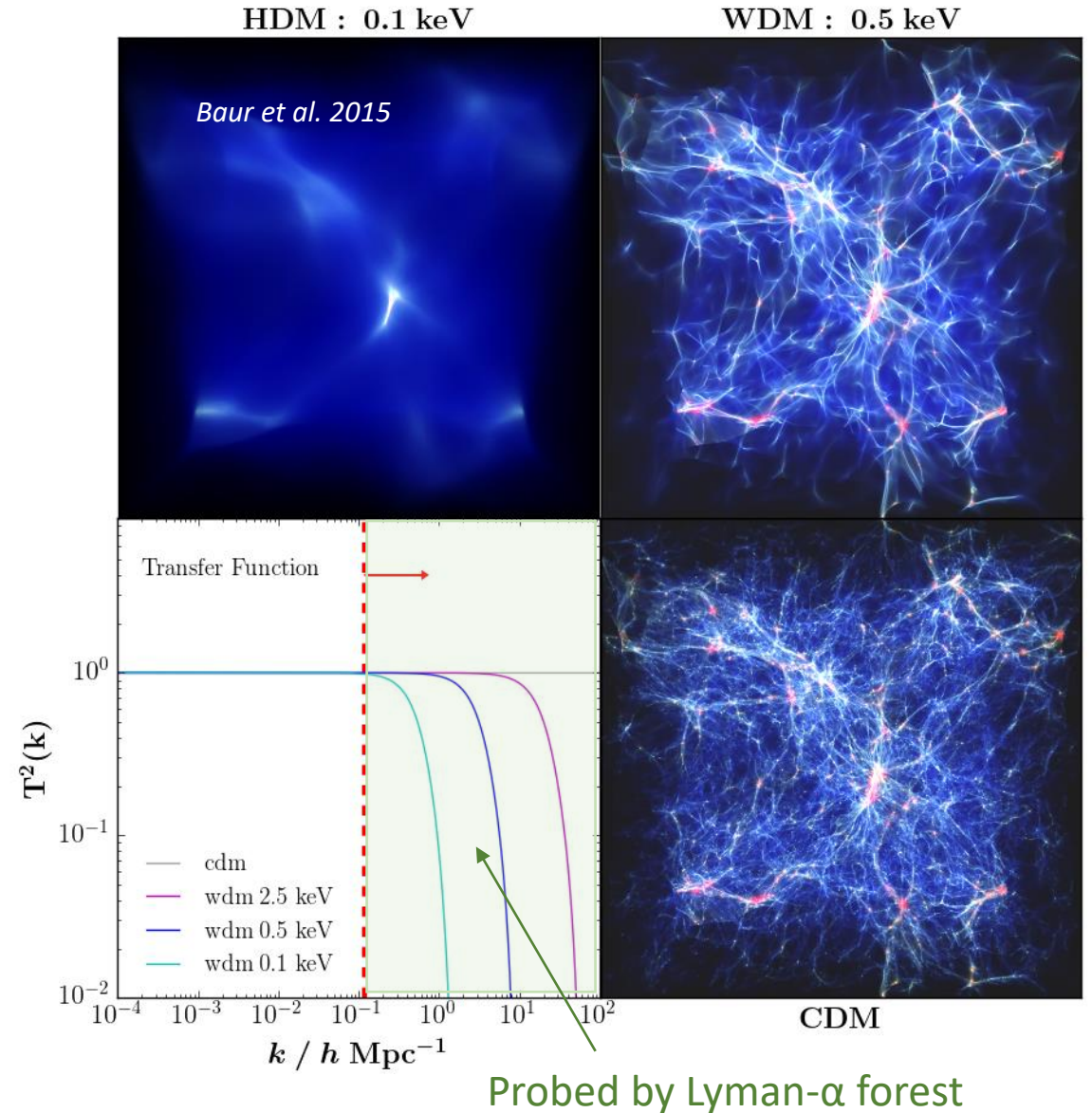
- High-resolution data from Karaçaylı et al. 2022
- KODIAQ, SQUAD, and XQ-100 data



Dark matter properties

Q: Can we test specific dark matter models?

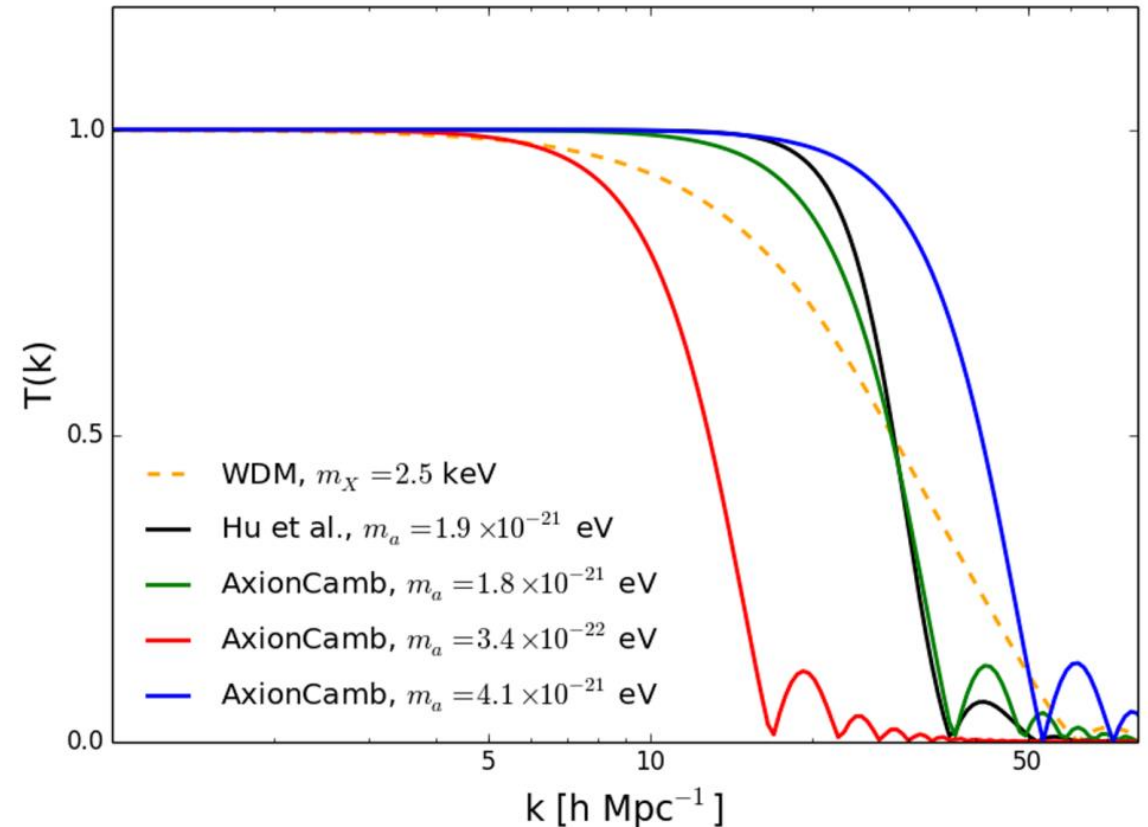
- Matter clustering may be impacted by dark matter properties
- Warm dark matter:
 - Thermal relics from cosmic microwave background
 - Mass-dependent power spectrum cut-off on small scales
- Other models:
 - Fuzzy dark matter
 - Self-interacting dark matter
 - Primordial black holes



Fuzzy Dark Matter

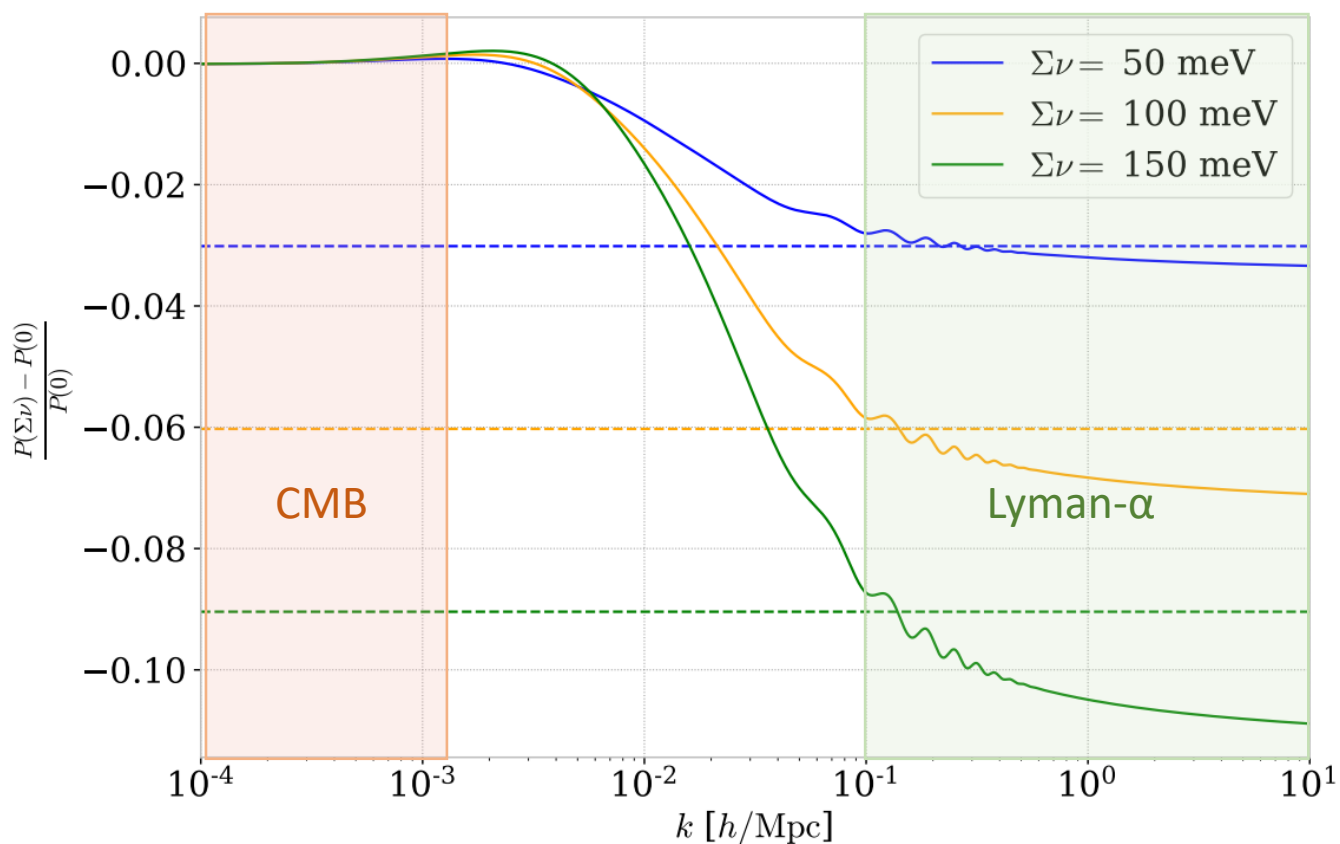
- Fuzzy Dark Matter (Armengaud et al. 2017, Irsic et al. 2017):
 - De Broglie length close to structure formation and DM halo dynamics
$$\frac{\lambda_{dB}}{2kpc} \sim \left(\frac{10^{-22} eV}{m} \right) \left(\frac{10 km/s}{v} \right)$$
 - Smooth the density fluctuation by quantum wave effects
 - Constraint by P1D:

$$m_a > 2 - 3 \times 10^{-21} eV$$



$P_{1D,\alpha}$ simulations

- For BOSS/eBOSS: Taylor expanded grid
- For DESI: Emulated simulation grid with Gaussian Processes (Walther et al. 2021)



Cosmology

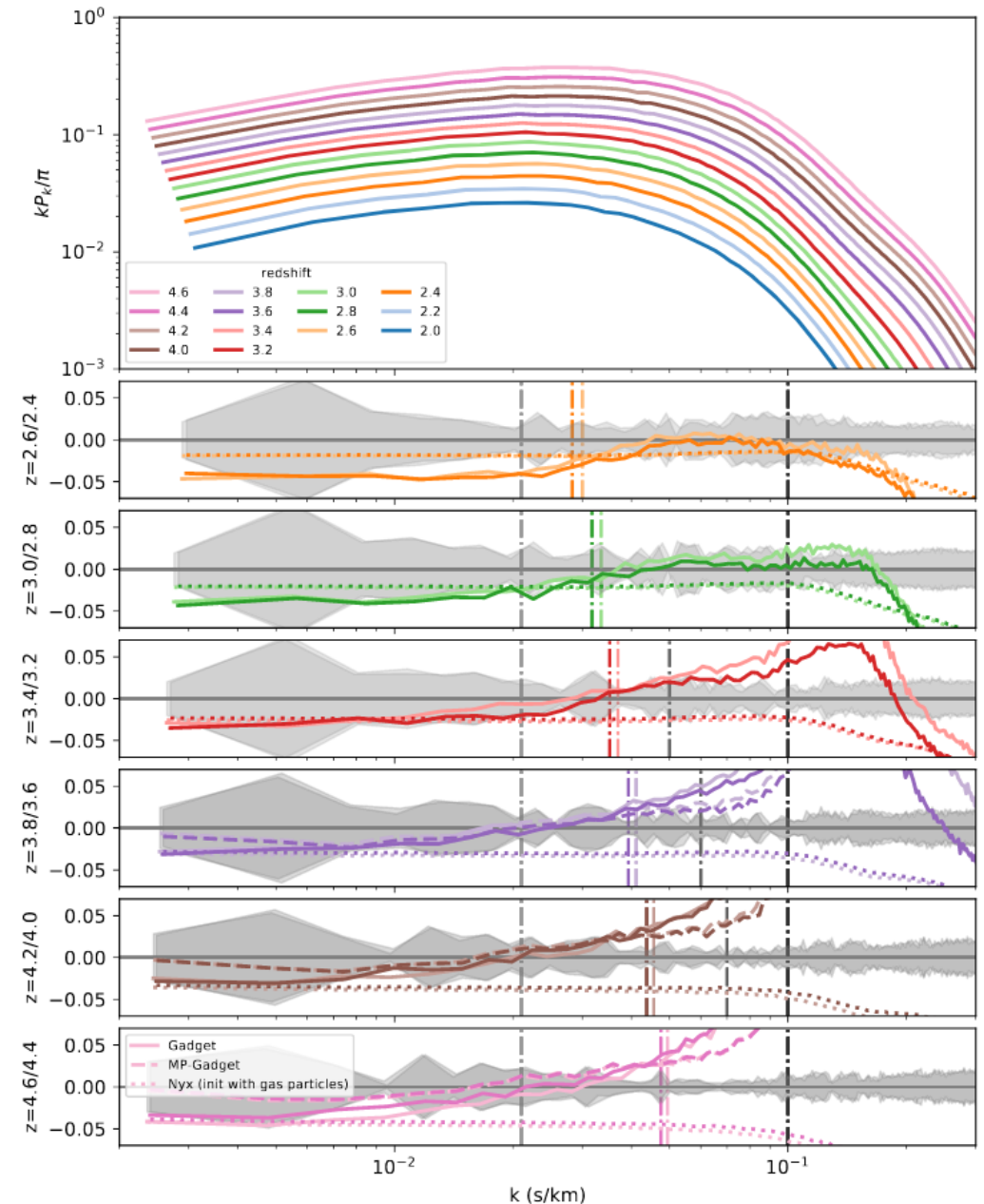
<i>parameter</i>	<i>central</i>	<i>range</i>
keV / m_x	0.0	+0.2 +0.4
$\Sigma m_\nu / eV$	0.0	+0.4 +0.8
h	0.675	± 0.05
Ω_M	0.31	± 0.05
σ_8	0.83	± 0.05
n_s	0.96	± 0.05
$dn_s / d \ln k$	0.00	± 0.04
z_{reio}	12	± 4
N_{eff}	3.046	± 1
$T_0^{z=3} / K$	14,000	$\pm 7,000$
$\gamma^{z=3}$	1.3	± 0.3
A^τ	0.0025	± 0.0020
η^τ	3.7	± 0.4

Intergalactic Medium

Optical Depth

$P_{1D,\alpha}$ simulations

- Interpretation of $P_{1D,\alpha}$ measurement with simulation:
 - At the scales considered, high-resolution hydrodynamical simulations required.
 - Nyx grid $4096^3/120$ Mpc (2M CPU hours)
- Gaussian processes emulator:
 - Covers cosmological parameter space
 - Reduce number of simulations
- Contributed to run simulations and compare codes

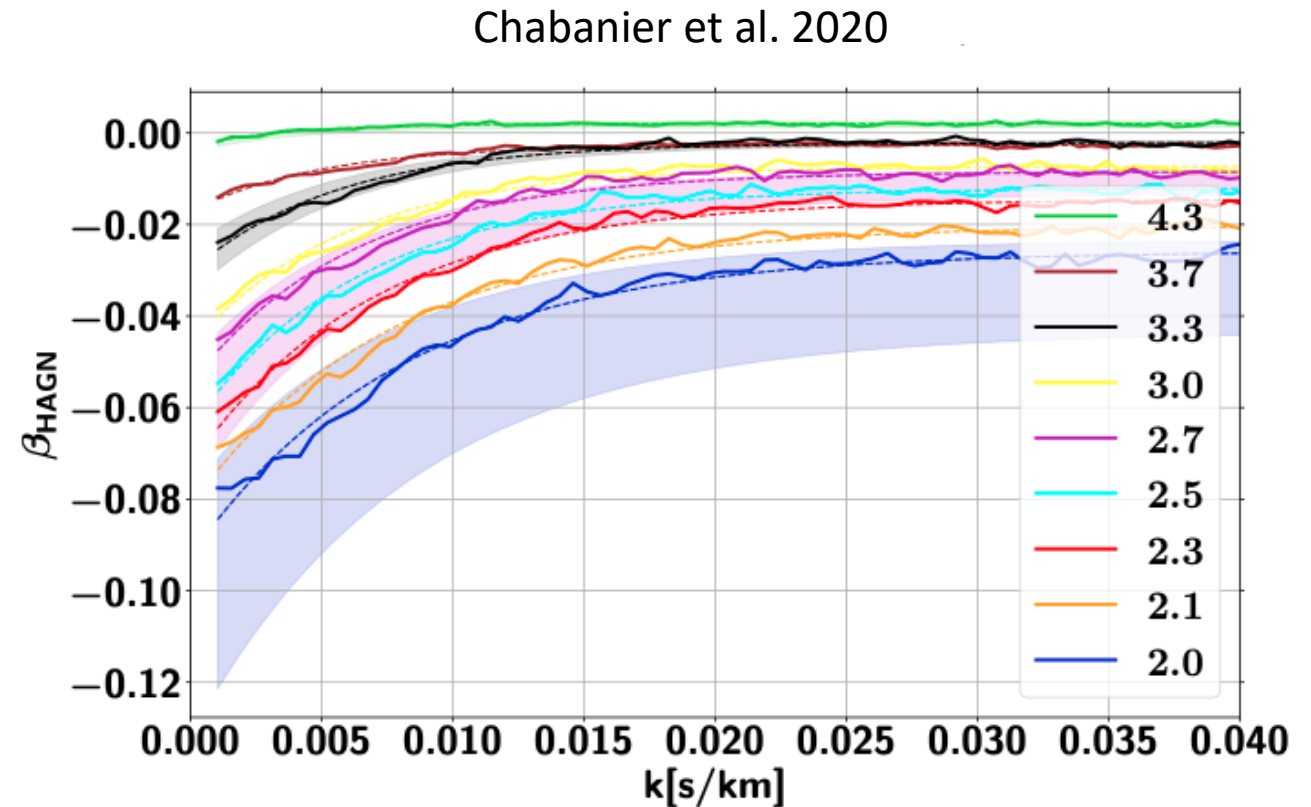
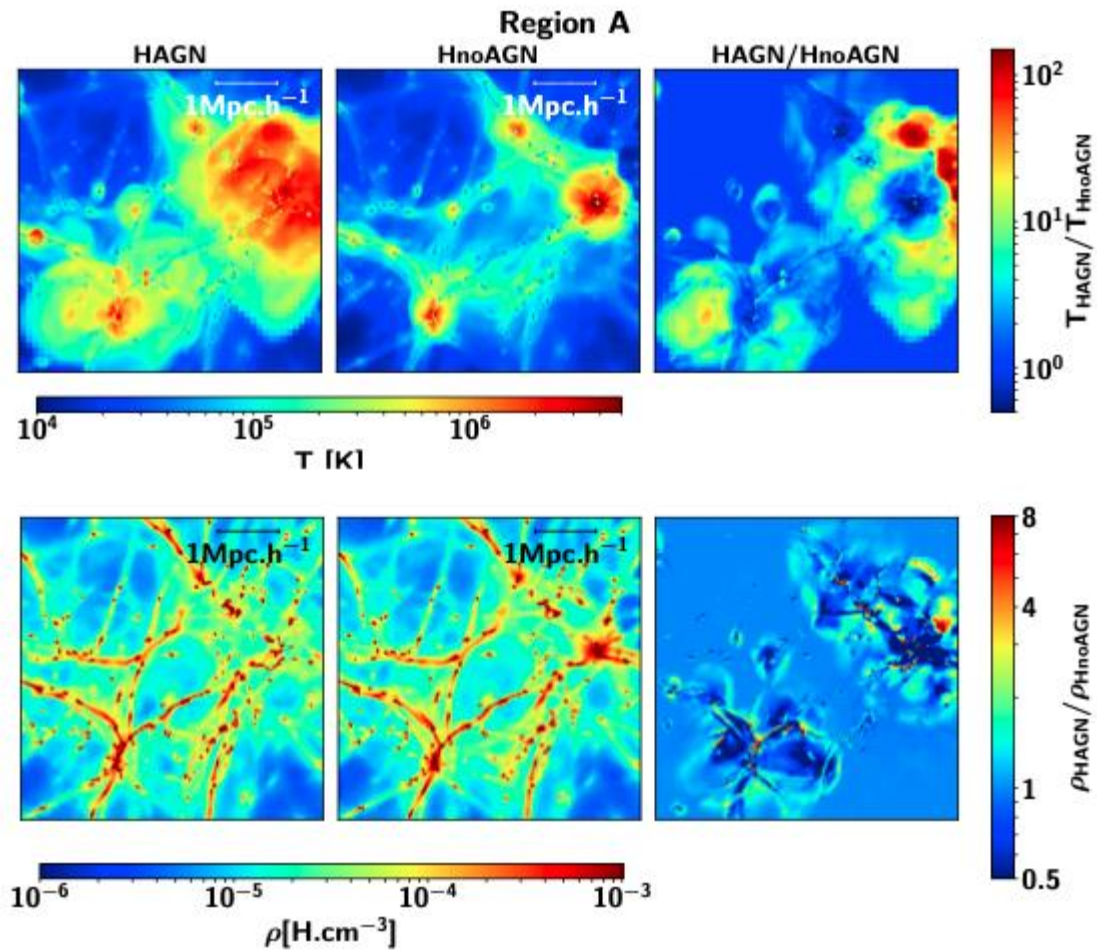


Nyx physics

- Nyx = Hydrodynamical code on grid + Dark matter particles on PM scheme
- Lyman- α forest not very sensitive to very dense IGM regions
 - AMR is not adapted
- Other physical processes modeled in Nyx:
 - Gas chemistry = fixed composition with H and He abundance
 - Inverse Compton + atomic collisional processes
- Effects not included:
 - Thermal feedback from AGN or supernovae
 - Inhomogeneous radiative background (UV)
 - High redshifts: full reionization history (assumed homogeneous)
- Choice: No explicit simulation of these effects but taken into account as a nuisance at the fitting stage
 - Example: AGN effect on P1D accounted for (Chabanier et al. 2020, Horizon-AGN simulation)
 - More modeling effort needed to take into account other effects.

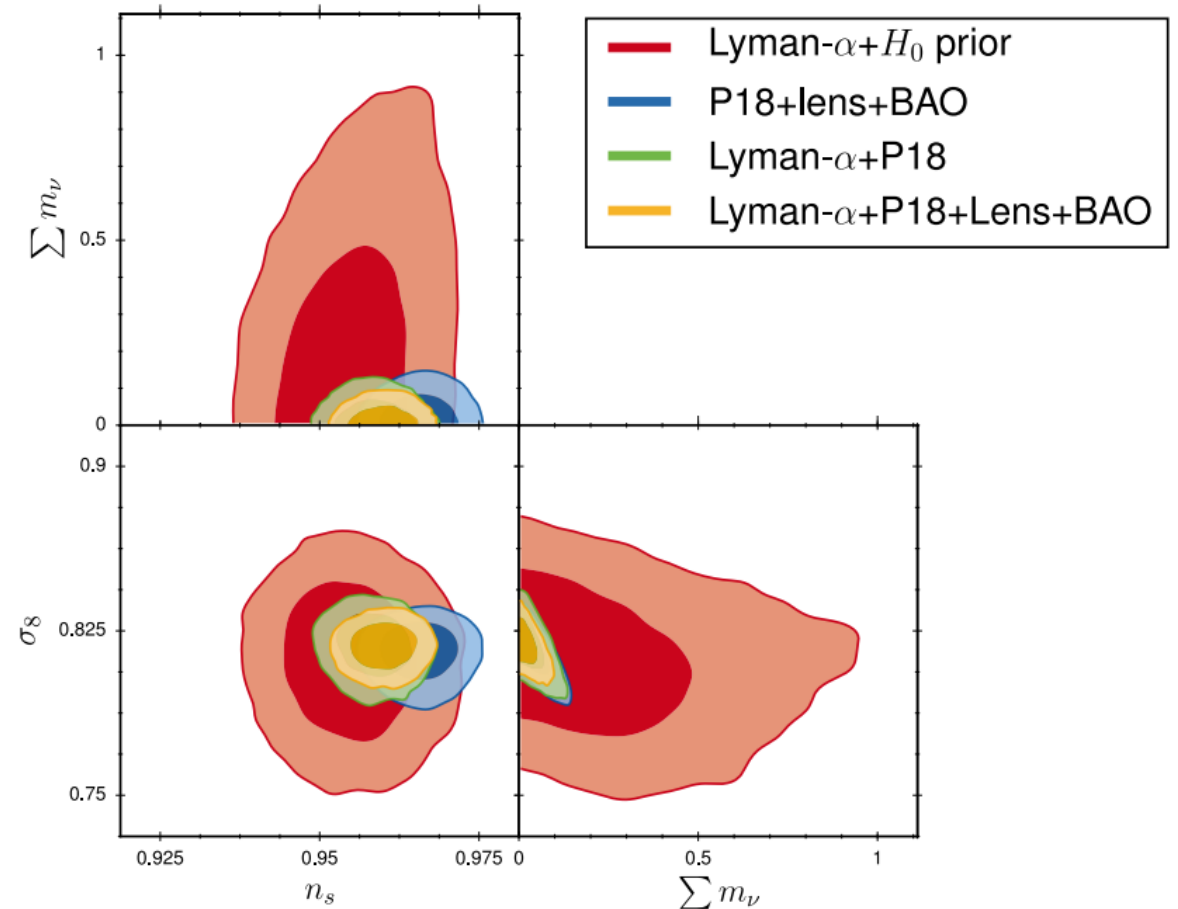
AGN feedback on P1D

- Physical effect = baryons and temperature redistribution in the IGM
- P1D correction, using different feedback parameters with HorizonAGN simulations



Current neutrino and WDM constraints

- Cosmological constraints using P1D computed on data and simulations (Palanque-Delabrouille et al. 2020)
- Loose constraint by Lyman- α only (Neutrinos simulated)
- Strong constraints combining with CMB data
- DESI: Emulated grid and higher data statistics will improve constraints and their robustness



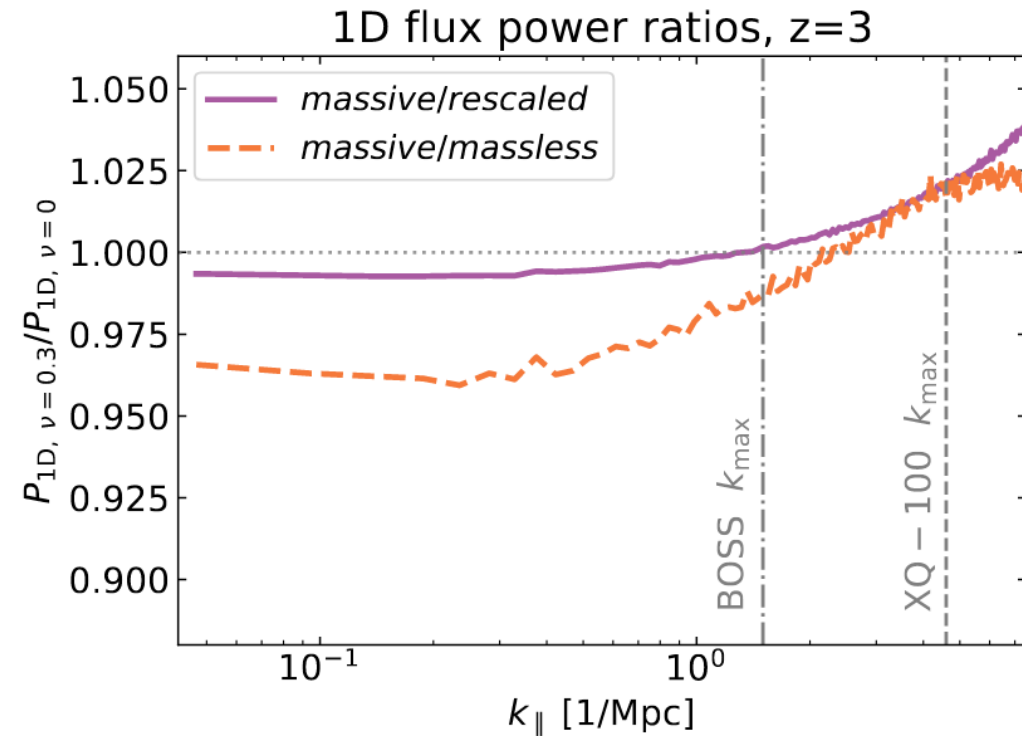
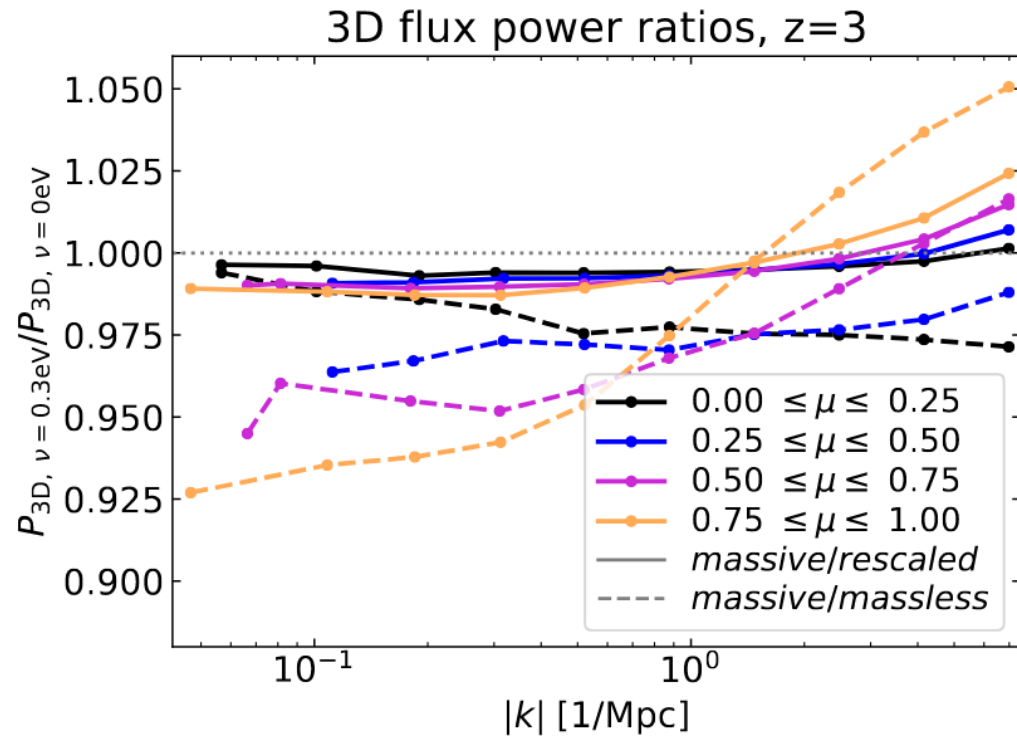
Current neutrino and WDM constraints

- Neutrino mass:
 - P1D: $\sum m_\nu < 0.58 \text{ eV}$
 - P1D + CMB: $\sum m_\nu < 0.11 \text{ eV}$
 - P1D + CMB + BAO + WL: $\sum m_\nu < 0.09 \text{ eV}$

(Palanque-Desabrouille et al. 2020) 95 % CL
- WDM mass (eBOSS + XQ100):
 - $m_X > 5.3 \text{ keV}$ 95 % CL
- Sterile neutrinos (Baur et al. 2016):
 - Equivalence relation with thermal relics mass (WDM)
 - P1D constraint: $m_s > 34 \text{ keV}$ (Non-resonantly produced)
 - X-ray signal at $m_s = 7 \text{ keV}$ → in strong tension

Rescaling method for neutrino constraints

- Neutrinos can be accounted into simulation by exploiting the power spectrum level degeneracy



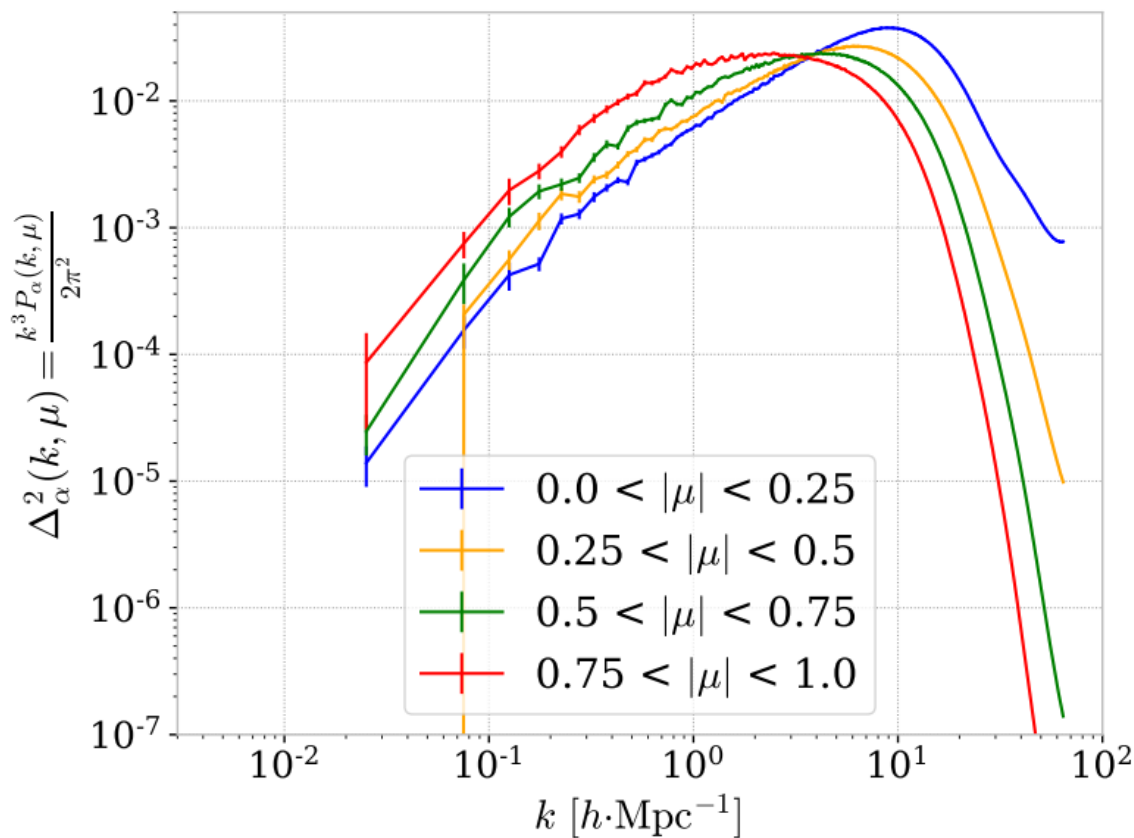
DESI forecast

- DESI collaboration 2016

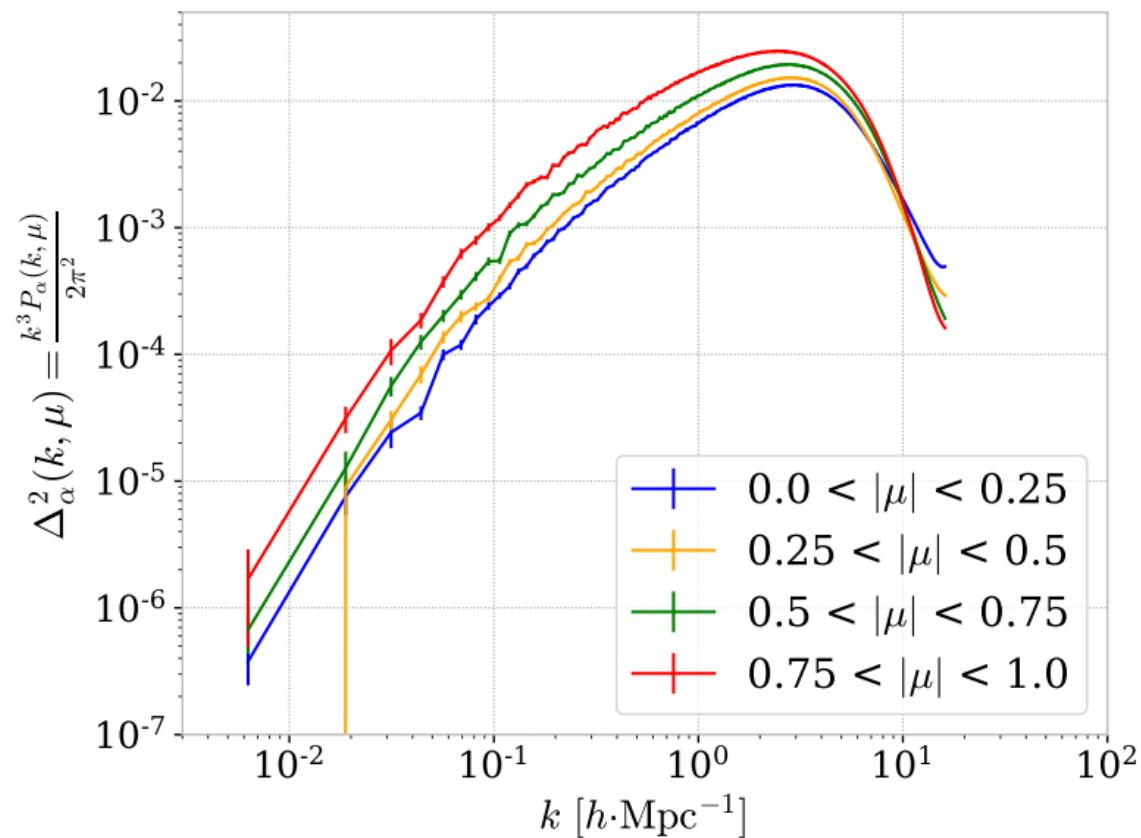
Data	$\sigma_{\Sigma m_\nu}$ [eV]	$\sigma_{N_{\nu,\text{eff}}}$
Planck	0.56	0.19
Planck + BAO	0.087	0.18
Gal ($k_{\text{max}} = 0.1h \text{ Mpc}^{-1}$)	0.030	0.13
Gal ($k_{\text{max}} = 0.2h \text{ Mpc}^{-1}$)	0.021	0.083
Ly- α forest	0.041	0.11
Ly- α forest + Gal ($k_{\text{max}} = 0.2$)	0.020	0.062

Details on $P_{3D,\alpha}$ simulations

- $P_{3D,\alpha}$ predictions on Jean-Zay “grand challenge” simulations
- Variation of resolution and box size



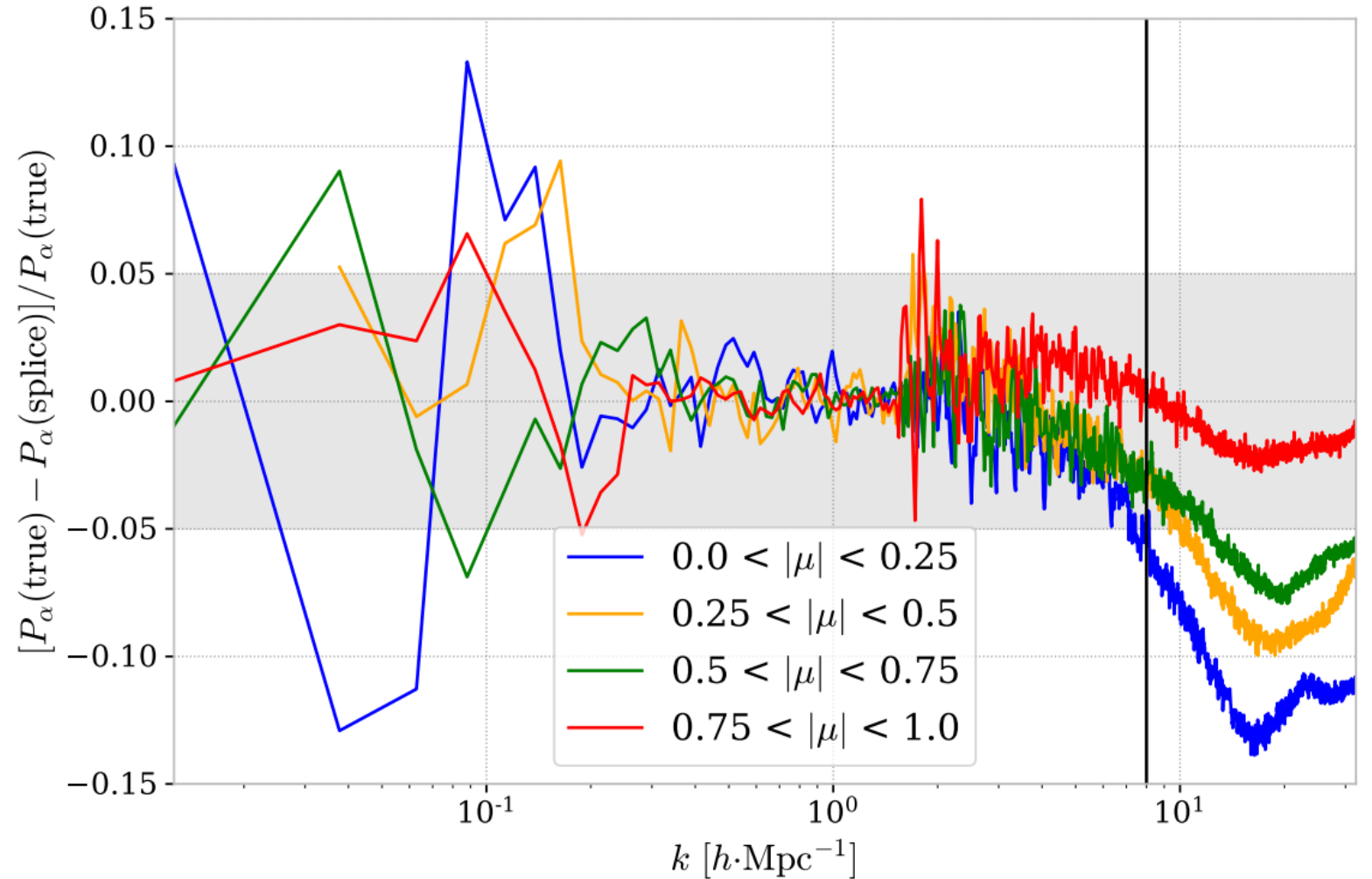
L125



L500

Details on $P_{3D,\alpha}$ simulations

- Splicing: use of high resolution and small boxes, with low resolution and large boxes.
- Verification on boxes for which high resolution/large size box is available.



Details on $P_{3D,\alpha}$ simulations

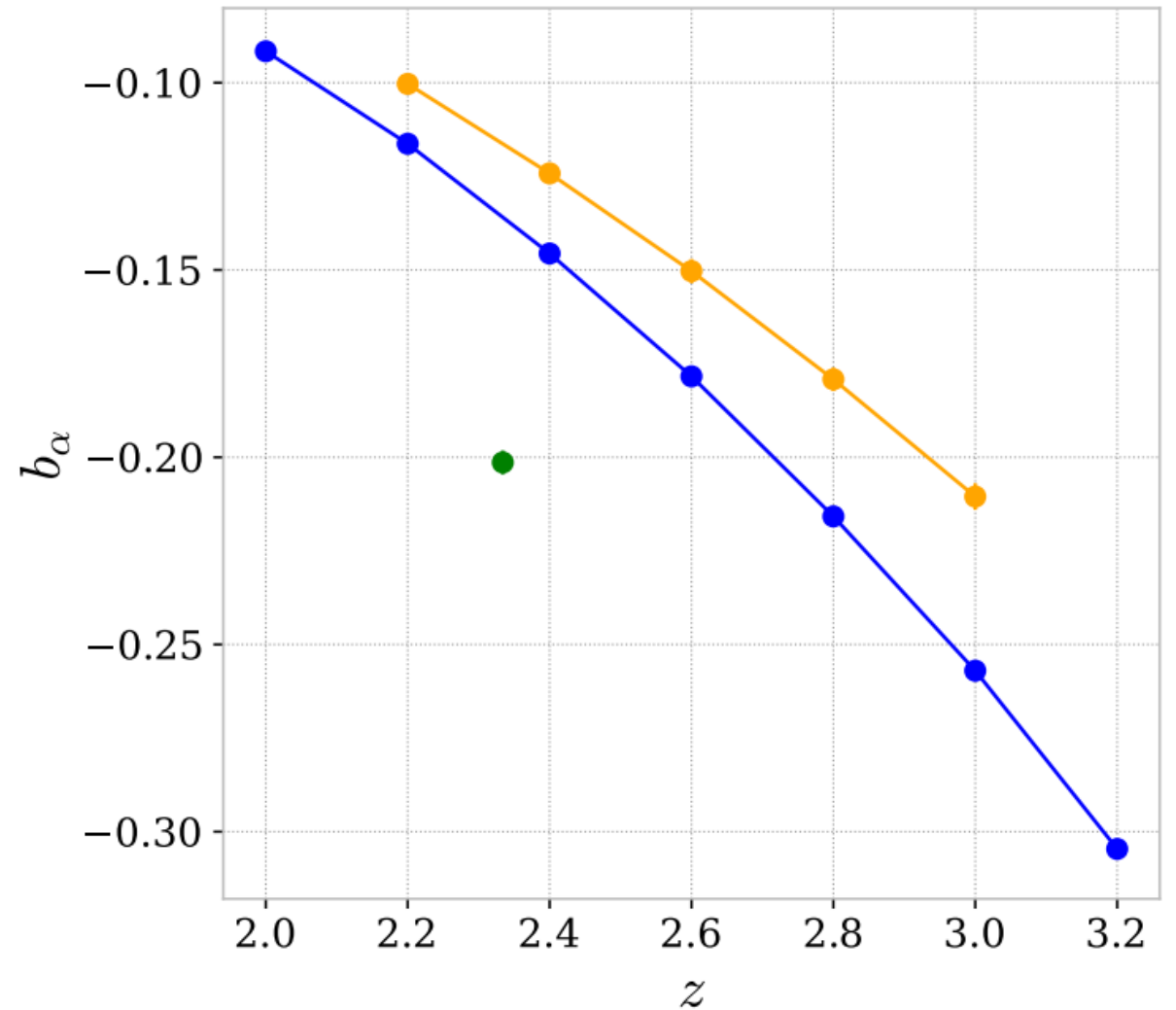
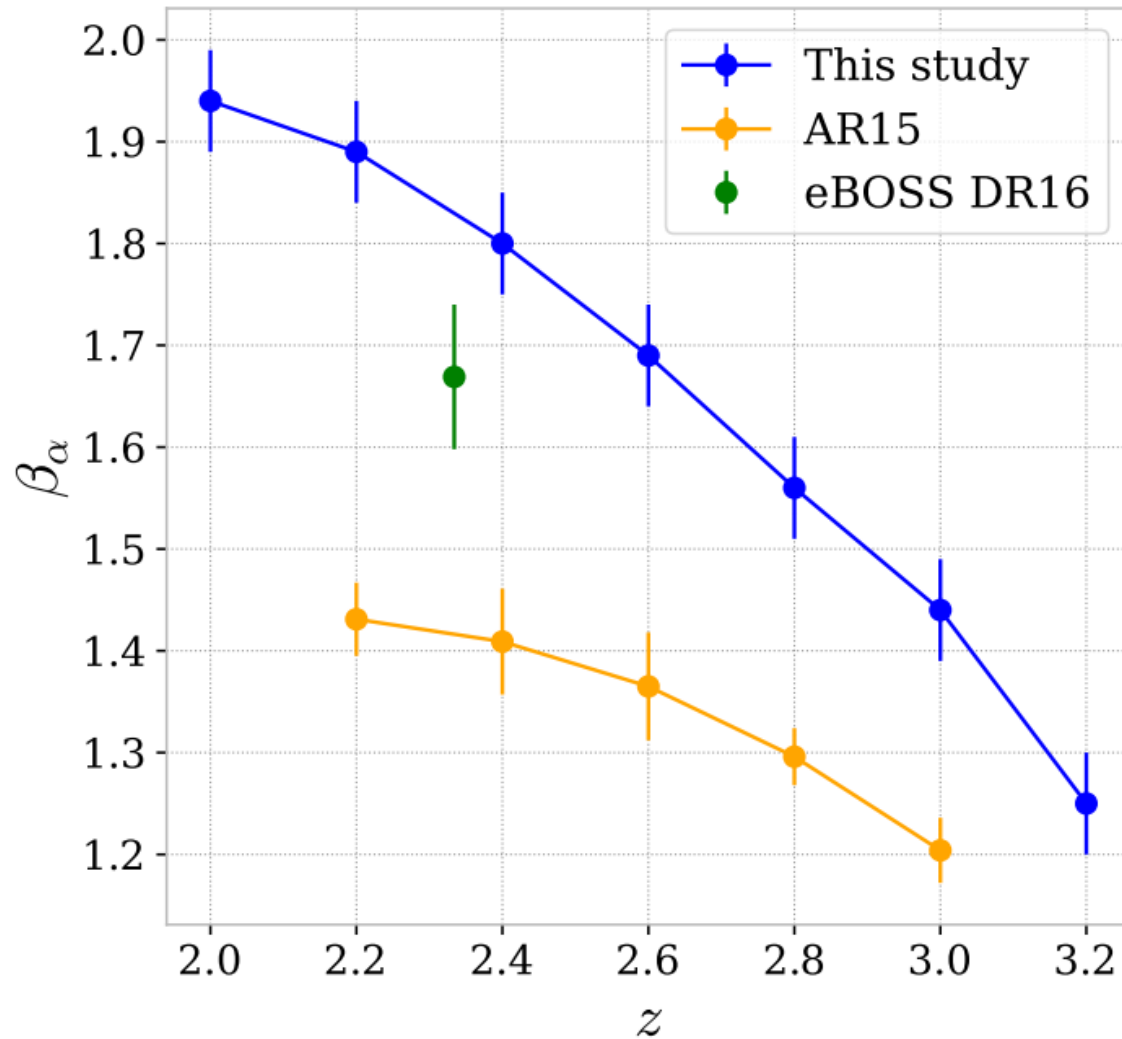
- $P_{3D,\alpha}$ model

$$P_{\text{model},\alpha}(k, \mu) = b_{\alpha}^2 \left(1 + \beta_{\alpha} \mu^2\right)^2 P_{\text{m}}(k) D(k, \mu)$$

$$D_1(k, \mu) = \exp \left\{ \left[q_1 \frac{k^3 P_{\text{m}}(k)}{2\pi^2} + q_2 \left(\frac{k^3 P_{\text{m}}(k)}{2\pi^2} \right)^2 \right] \left[1 - \left(\frac{k}{k_{\text{v}}} \right)^{a_{\text{v}}} \mu^{b_{\text{v}}} \right] - \left(\frac{k}{k_{\text{p}}} \right)^2 \right\}$$

Details on $P_{3D,\alpha}$ simulations

- Fitting of linear parameters



Other $P_{3D,\alpha}$ simulations

- Givans et al. 2022

