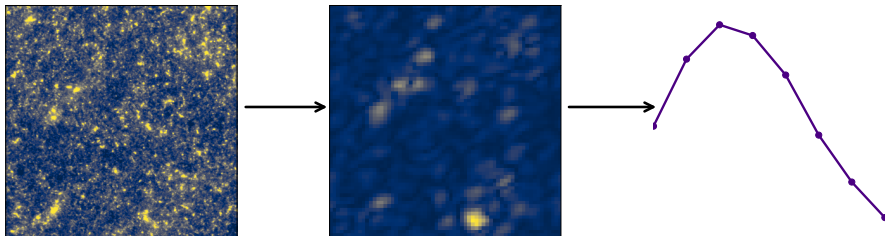


# Scattering transforms for weak lensing cosmological inference



**Mariia Marinichenko**

Leiden University, the Netherlands

Non-Gaussian Universe ■ June 17, 2026



## ① Why ST?

- captures non-Gaussian information
- easy to compute and interpretable

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ST is a fast route from WL maps to comprehensible inference

# Why the power spectrum is not enough

The power spectrum tells us **how much fluctuation power** we have at each angular scale

$$P_{\kappa}(\ell) = \langle |\kappa_{\ell}|^2 \rangle.$$

But it does not keep the Fourier phases, which carry information about **where structures are** and **how they are arranged**

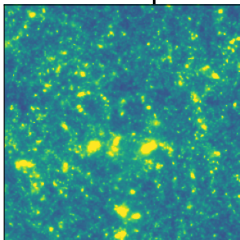
So, we lose part of the information about:

- halos, filaments and voids
- clustering of structures of different sizes
- nonlinear effects from baryons or dark sector physics

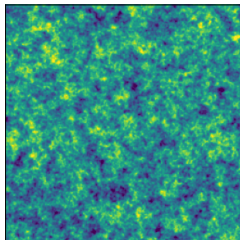
Part of this missing information can be recovered with **higher order statistics**

# Why the scattering transform?

$\kappa$  map

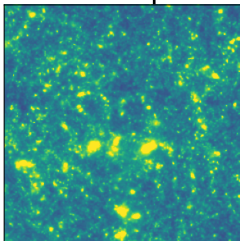


random field

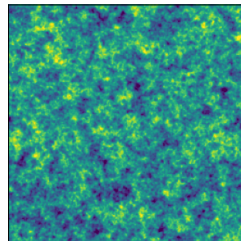


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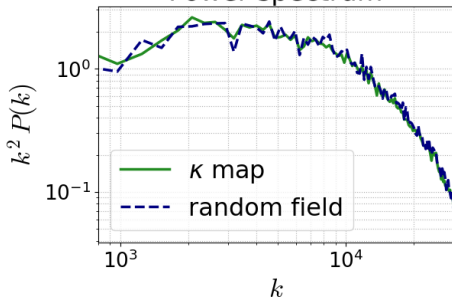
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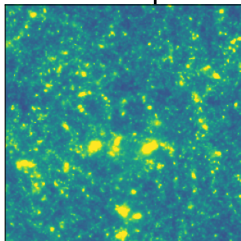


Power spectrum

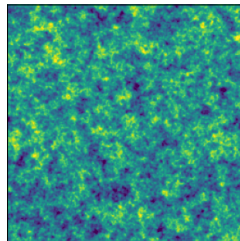


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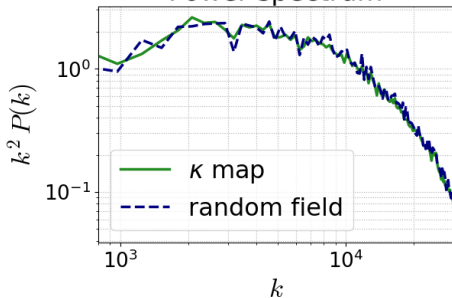
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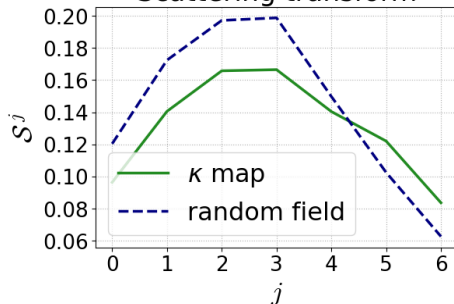
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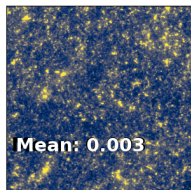
Power spectrum



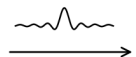
Scattering transform



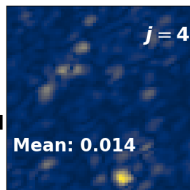
# Scattering Transform



Convolution  
with localized  
kernel  $\psi^j$



A diagram showing a horizontal arrow pointing to the right. Above the arrow is a small sketch of a localized kernel, represented as a wavy line with a central peak. The text "Convolution with localized kernel  $\psi^j$ " is positioned below the arrow.

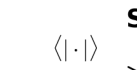


Modulus +  
Mean  
pixelwise

Set of scattering  
coefficients

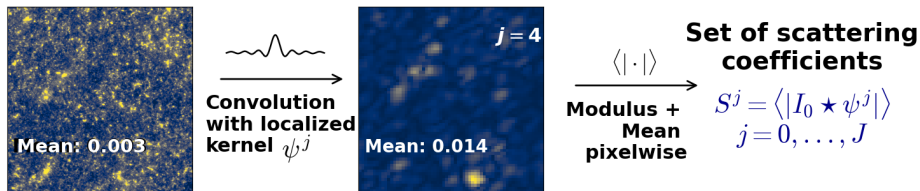
$$S^j = \langle |I_0 \star \psi^j| \rangle$$

$j = 0, \dots, J$



A diagram showing a horizontal arrow pointing to the right. Above the arrow is the notation  $\langle |\cdot| \rangle$ . The text "Modulus + Mean pixelwise" is positioned below the arrow.

# Scattering Transform



Scattering transform = wavelet convolution + modulus + mean

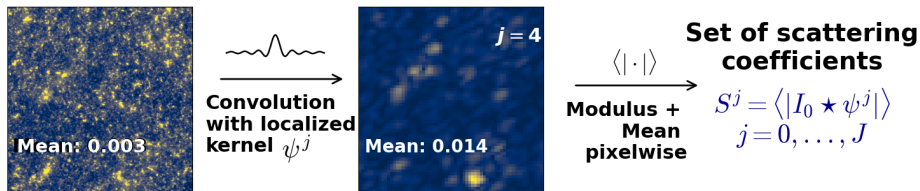
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for a range of dyadic scales

$$j = 0, \dots, J$$

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## Advantages

- **Interpretable** (strength of fluctuations)
- **Efficient** (compact set of estimators)
- **Robust** (stable to noise)

# What information does ST add?

## First order: structures at each scale

$$S_1(j) = \langle |I_0 \star \psi_j| \rangle$$

- measures how much structure appears at some (angular) scale  $j$
- similar idea to the power spectrum, but based on wavelets

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$$S_2(j_1, j_2) = \langle | |I_0 \star \psi_{j_1}| \star \psi_{j_2} | \rangle$$

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- describes the hierarchy of clustering patterns

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ST describes map morphology ordered by scale

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ST can break power spectrum degeneracies using clustering information and redshift evolution

# Why add ST to the Euclid inference pipeline?

## What ST adds

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- but is more compact than a densely binned bispectrum
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ST adds map morphology in a handy way

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## Sensitivity

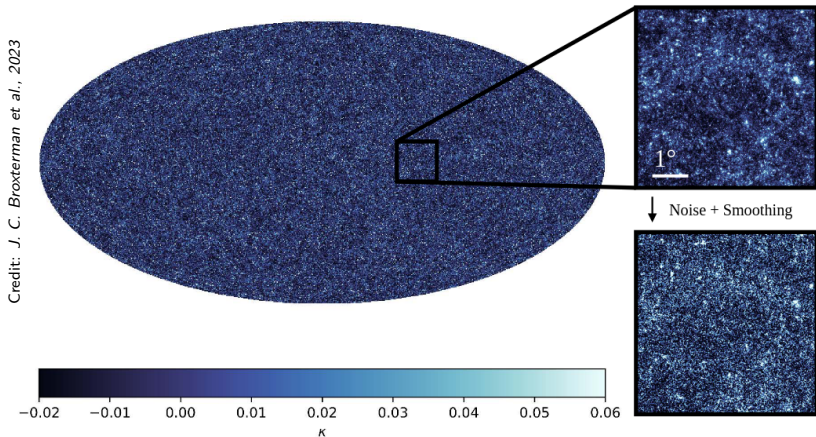
- build ST emulators from FLAMINGO
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## Debiasing

- model baryonic feedback bias with a transfer function
- add Euclid-like galaxy shape noise
- apply filters (Gaussian and Wiener)

# FLAMINGO convergence maps

The transfer functions are derived from **Euclid-like WL mocks** from the **FLAMINGO** simulations



Full-sky weak lensing convergence map

# Emulating scattering coefficients

## Simulation set

- 100 FLAMINGO hypercube simulations
- 13 varied cosmological parameters (10 independent), including  $\Omega_m, \sigma_8, w_0, w_a, M_\nu, \Gamma_{\text{dcdm}}, \dots$
- matched initial phases for paired runs (with and without decaying DM)

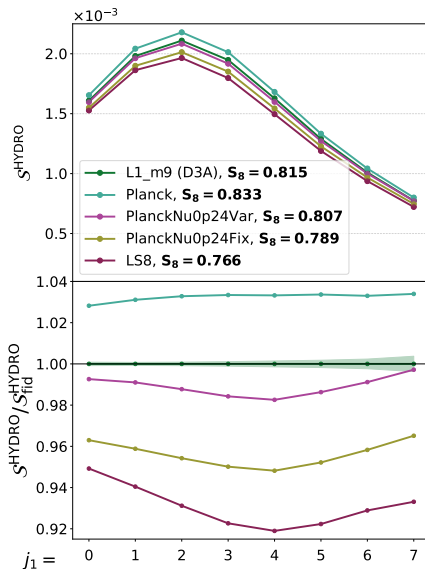
## Data vector

- $\kappa$  maps in 5 tomographic bins
- HEALPix  $N_{\text{side}} = 8192$
- $3.6 \times 3.6 \text{ deg}^2$  non-overlapping patches
- 1st ST coefficients for each patch
- Euclid-like Gaussian galaxy shape noise:
- $n_{\text{gal}} = 30 \text{ arcmin}^{-2}, \sigma_e = 0.26$

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Same feedback, different cosmologies

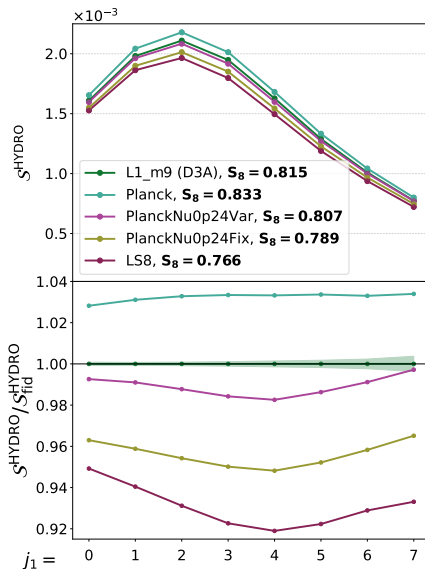


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### Result

Scattering coefficients are  
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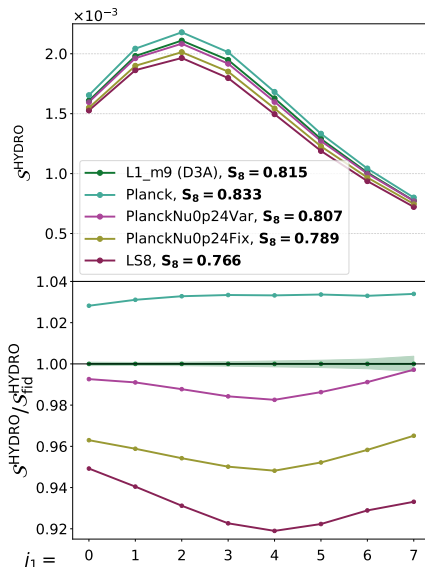


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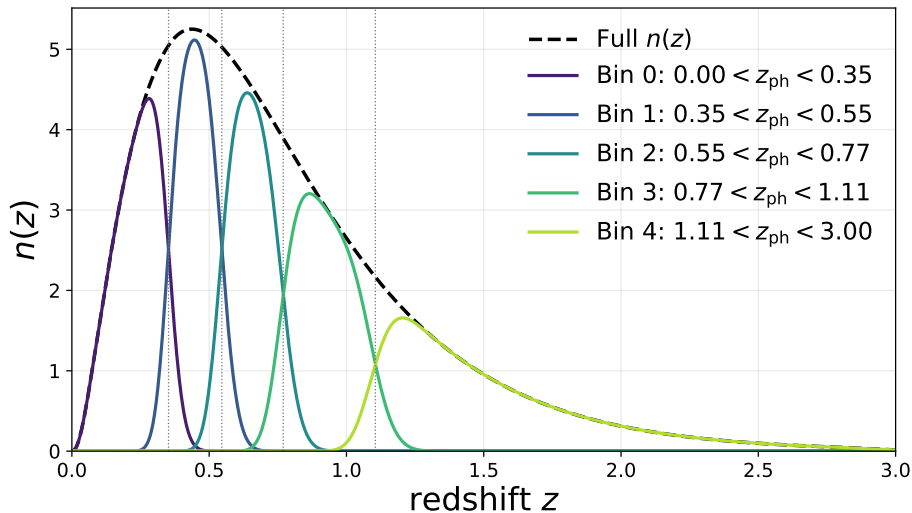
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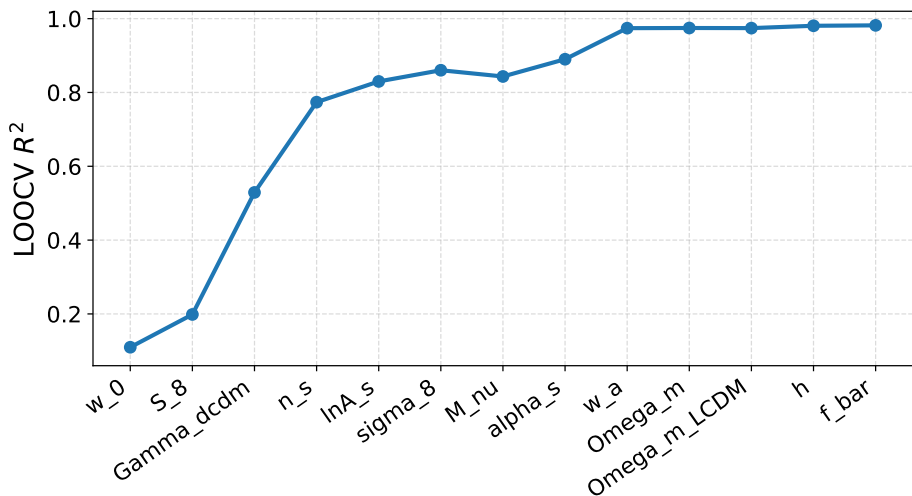
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# Tomographic bins

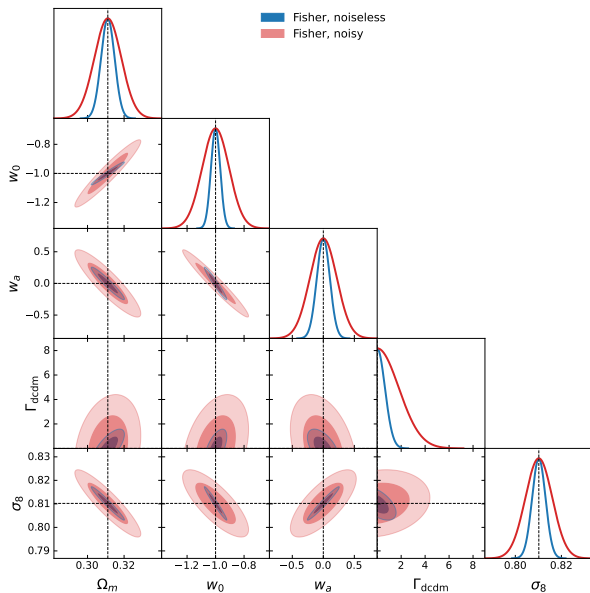


# Which parameters matter for the emulator?

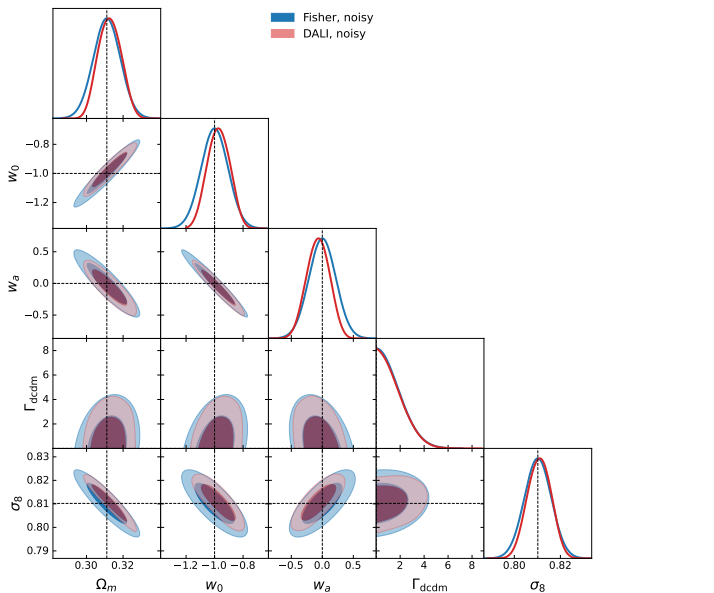


The greedy search selects the physical parameters that ST is sensitive to

# ST forecast: noise



# DALI improvement of the forecast



# Baryonic effects on the weak lensing scattering transform

The effect of baryonic feedback on WL convergence can be quantified through a transfer function, which we extract from **FLAMINGO** simulations.

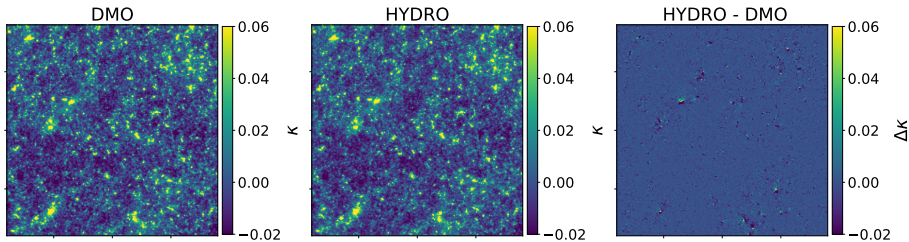


Figure: Convergence, i.e.  $\kappa$ , patches from FLAMINGO

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Baryonic effects in WL **scattering transform** can be **separated from cosmology** through a **transfer function**

# How the baryon model enters inference

## Cosmological parameter inference

Observed  $\kappa$  maps  $\Rightarrow$  ST data vector  $\Rightarrow$  likelihood  $\Rightarrow \Omega_m, \sigma_8, w_0, \Gamma_{\text{dcdm}}, \dots$

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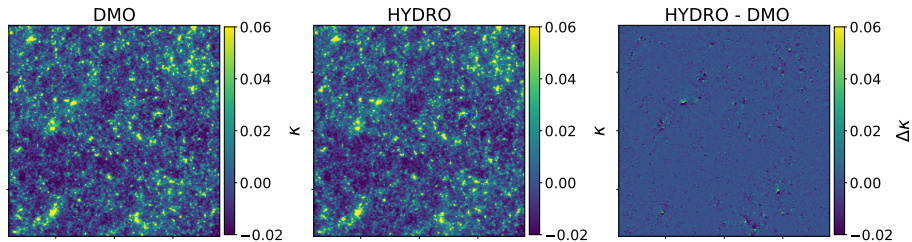
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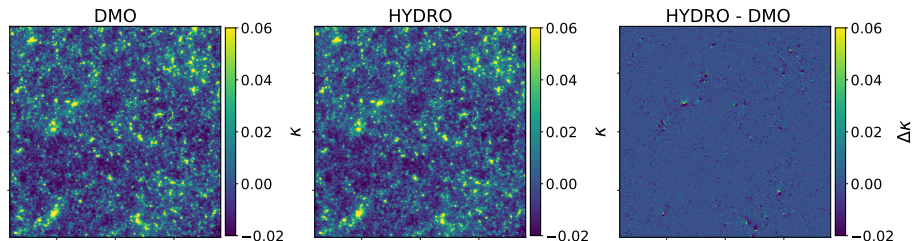
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Algorithm: emulate cosmology, multiply by baryonic response, and infer parameters.

# Transfer function



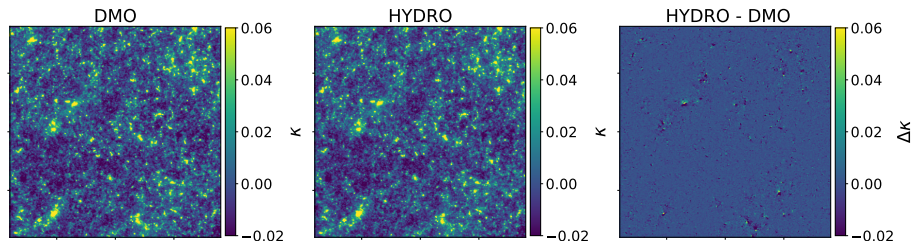
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The baryonic suppression (or boost)  
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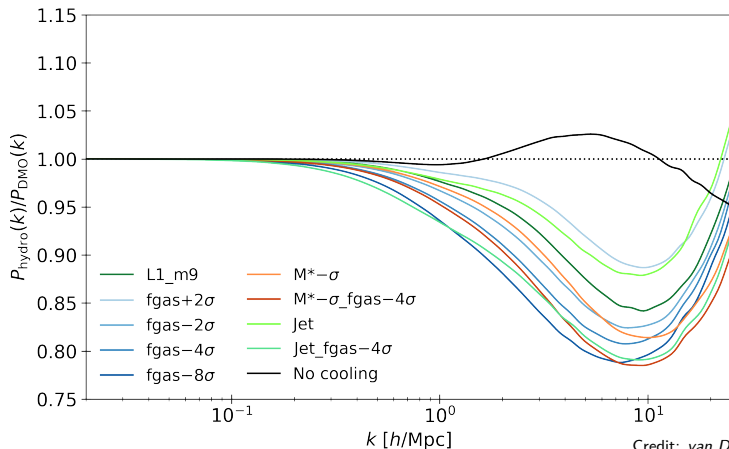
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Similarly, we introduce **the baryonic transfer function for the scattering coefficients**:

$$\mathcal{T} = \frac{\mathcal{S}^{\text{HYDRO}}}{\mathcal{S}^{\text{DMO}}}$$

# Baryonic feedback in FLAMINGO

Calibrated to cluster gas fraction,  $f_{\text{gas}}$ , and stellar mass function,  $M^*$  at  $z \approx 0$

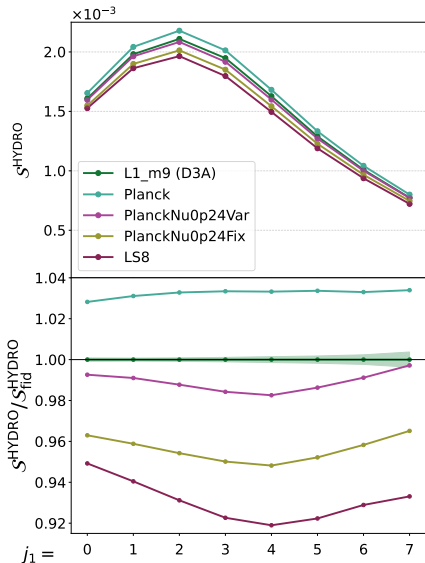


Credit: van Daalen et al. (2025)

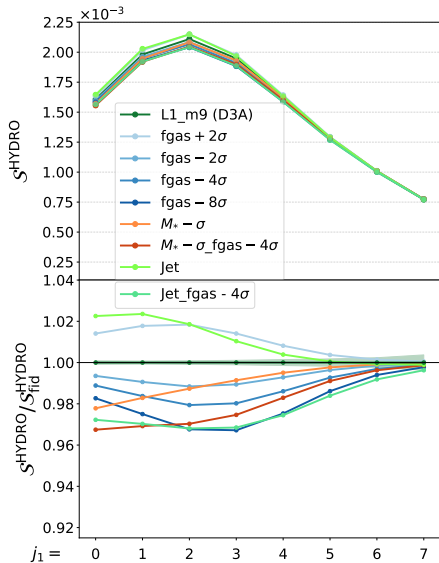


# Impact of baryons in ST

Same feedback, different cosmologies

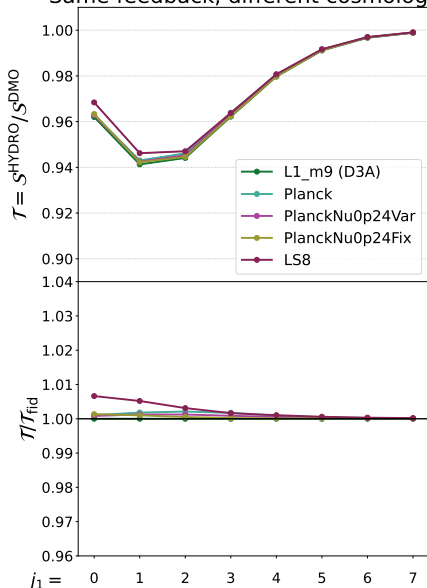


D3A cosmology, different feedbacks



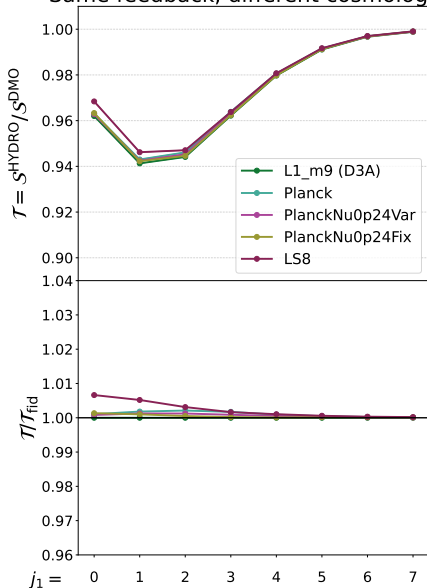
# ST transfer function $\mathcal{T} = \mathcal{S}^{HYDRO} / \mathcal{S}^{DMO}$

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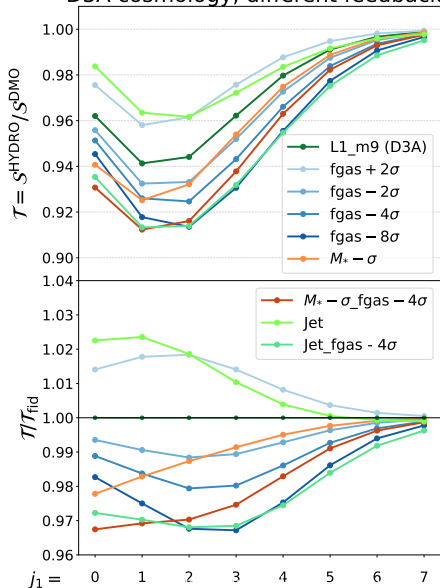


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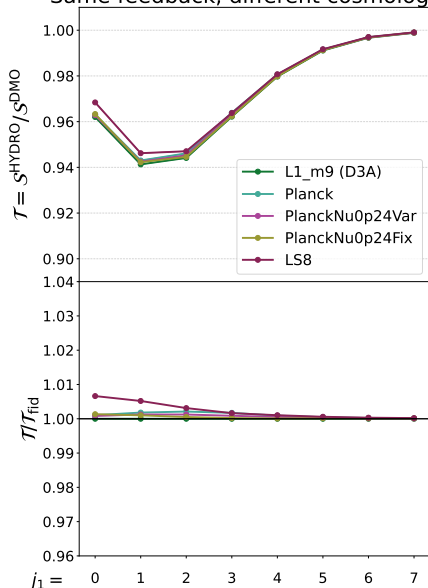


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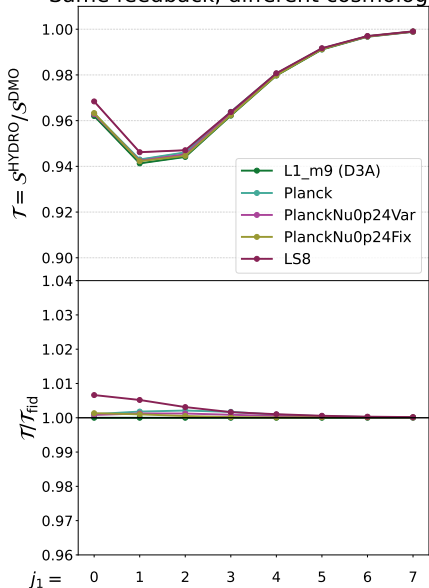
# Noise and smoothing

Same feedback, different cosmologies

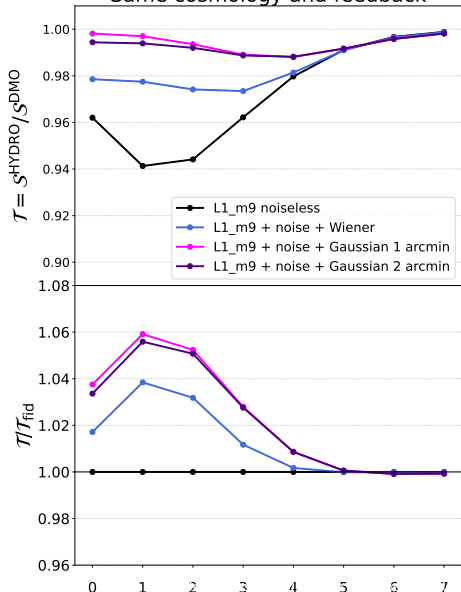


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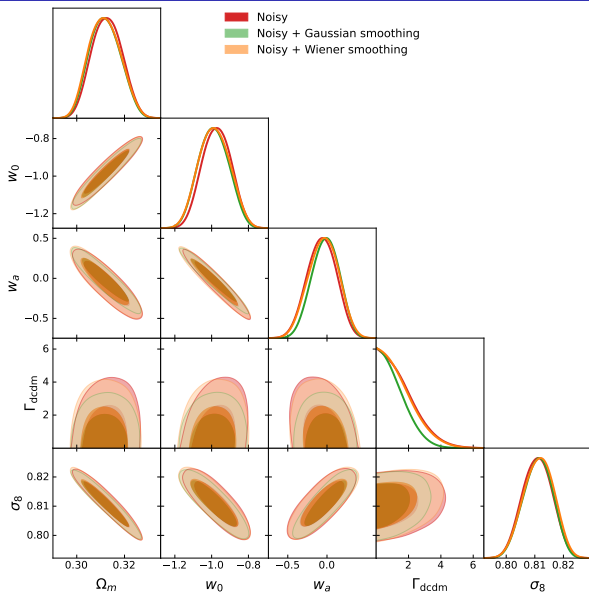
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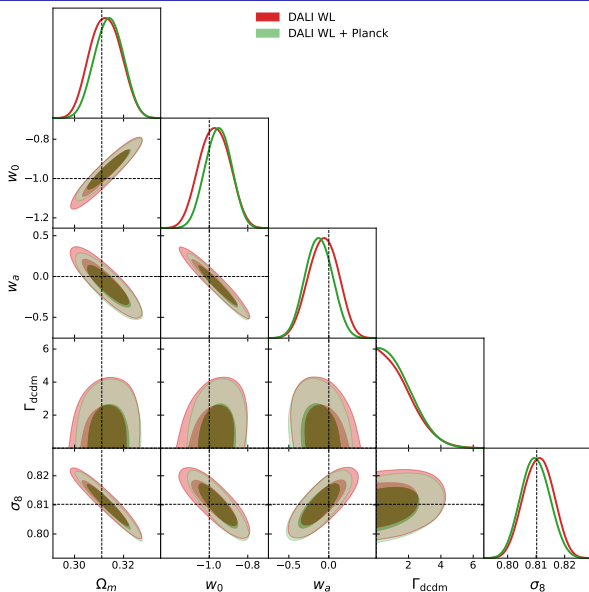
Same cosmology and feedback



# DCDM forecast: smoothing choices



# DCDM constraints from ST + Planck



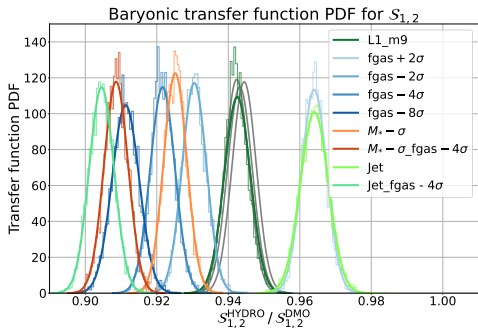
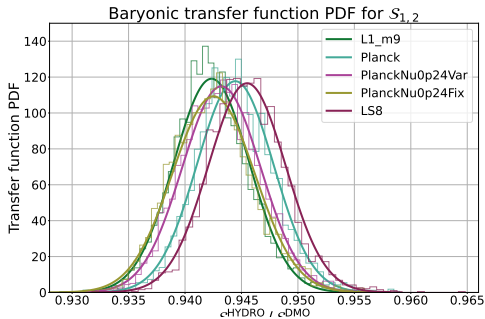
# Take-home messages

- Scattering coefficients add **non-Gaussian information** beyond the power spectrum
- ST is **compact, stable and interpretable**
- ST alone provides tight constraints on cosmological parameters such as  $\Gamma_{\text{dcdm}}$ ,  $\Omega_m$ ,  $\sigma_8$ ,  $w_0$ ,  $w_a$
- Baryonic feedback can be modelled in a handy way by transfer function  $\mathcal{T}$ , but requires smart denoising techniques (e.g., Wiener filtering)

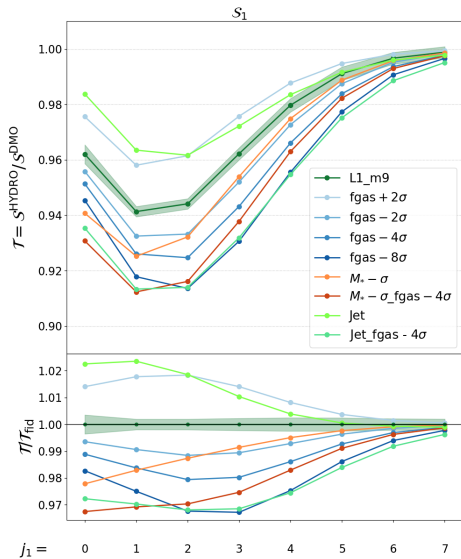
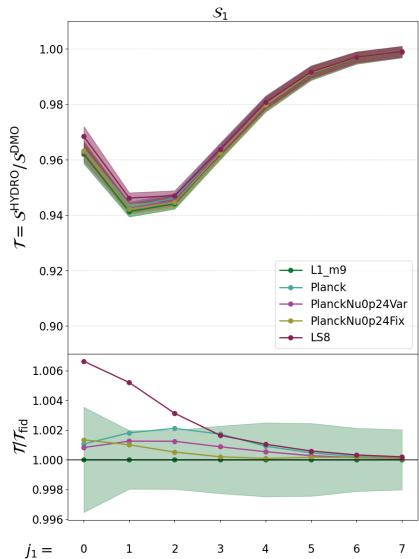
Thank you!



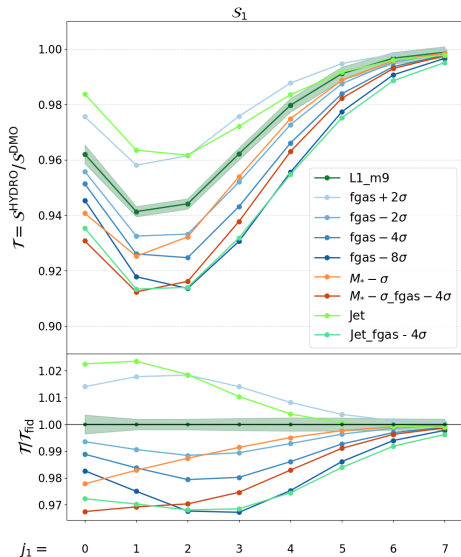
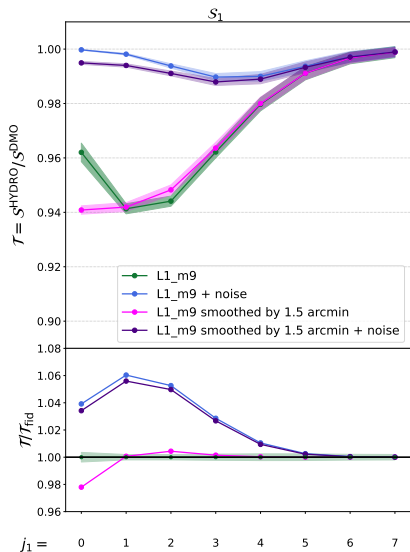
# Backup slides

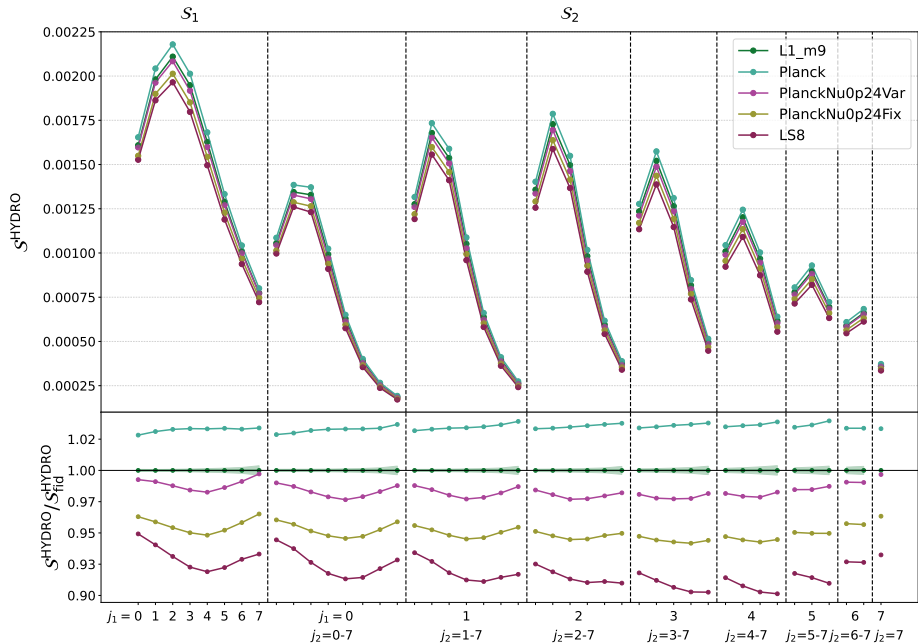


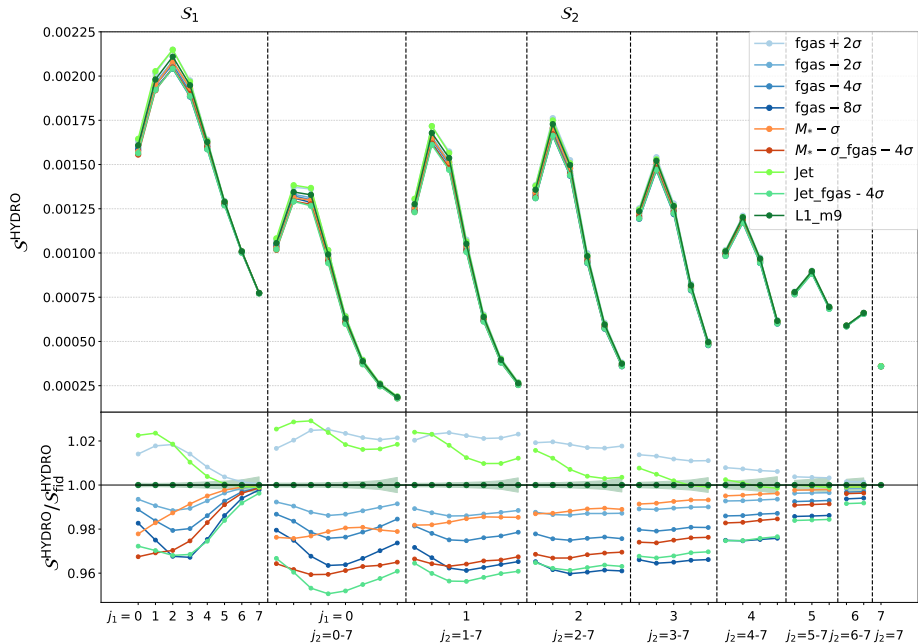
# With std

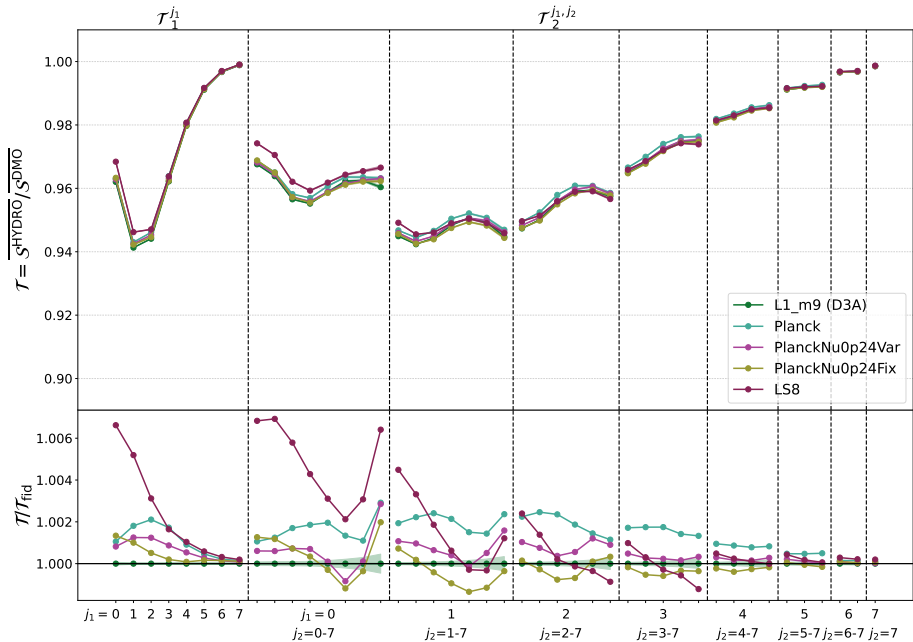


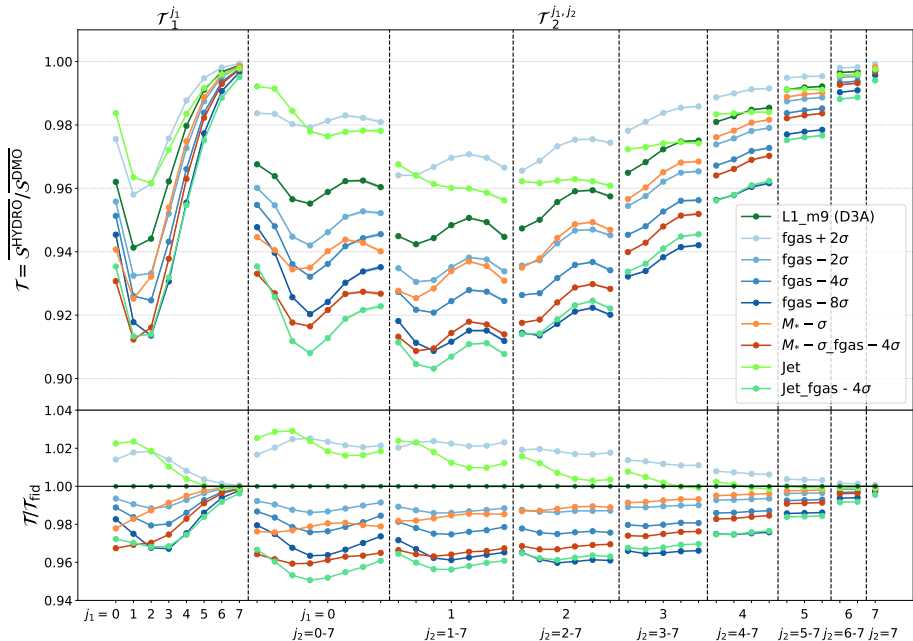
# With std

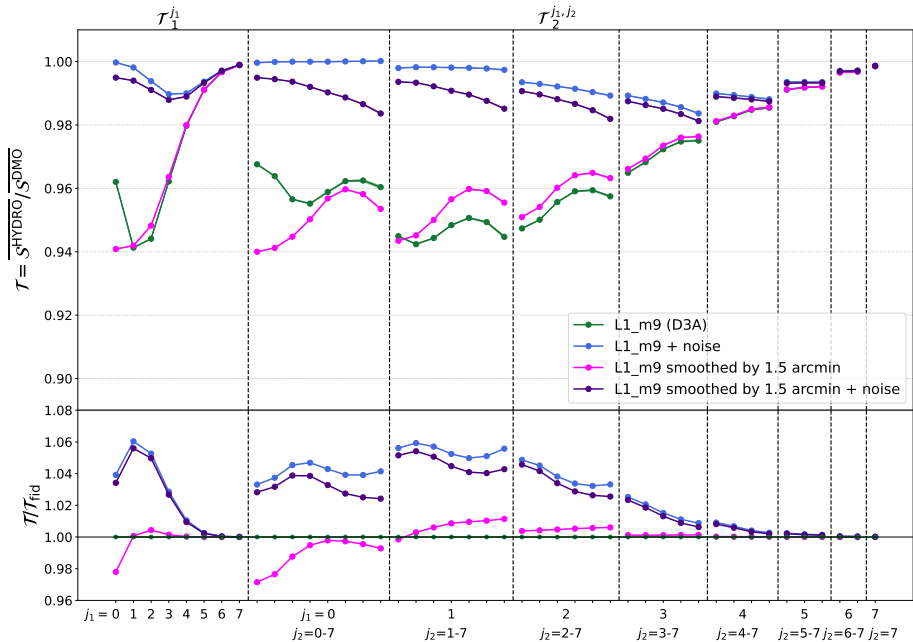


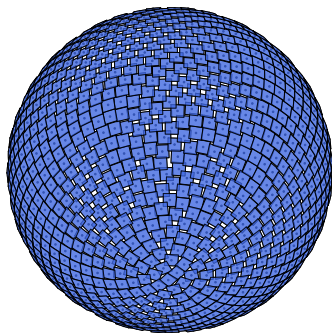
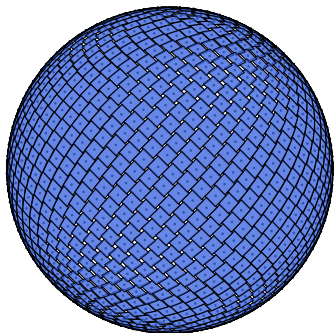












**Table 2.** Scale definitions used throughout the paper, with corresponding diadic scales  $j$ , Fourier wavenumbers  $k$ , and real-space scales  $D$  at  $z \approx 1$ .

Name	$j$	$k$ ( $h \text{ Mpc}^{-1}$ )	$D$ ( $\text{Mpc } h^{-1}$ )
Smallest	0	3.6	0.3
Small	1 – 2	1.8 – 0.9	0.6 – 1
Intermediate	3 – 5	0.4 – 0.1	2 – 9
Large	6 – 7	0.06 – 0.03	18 – 36