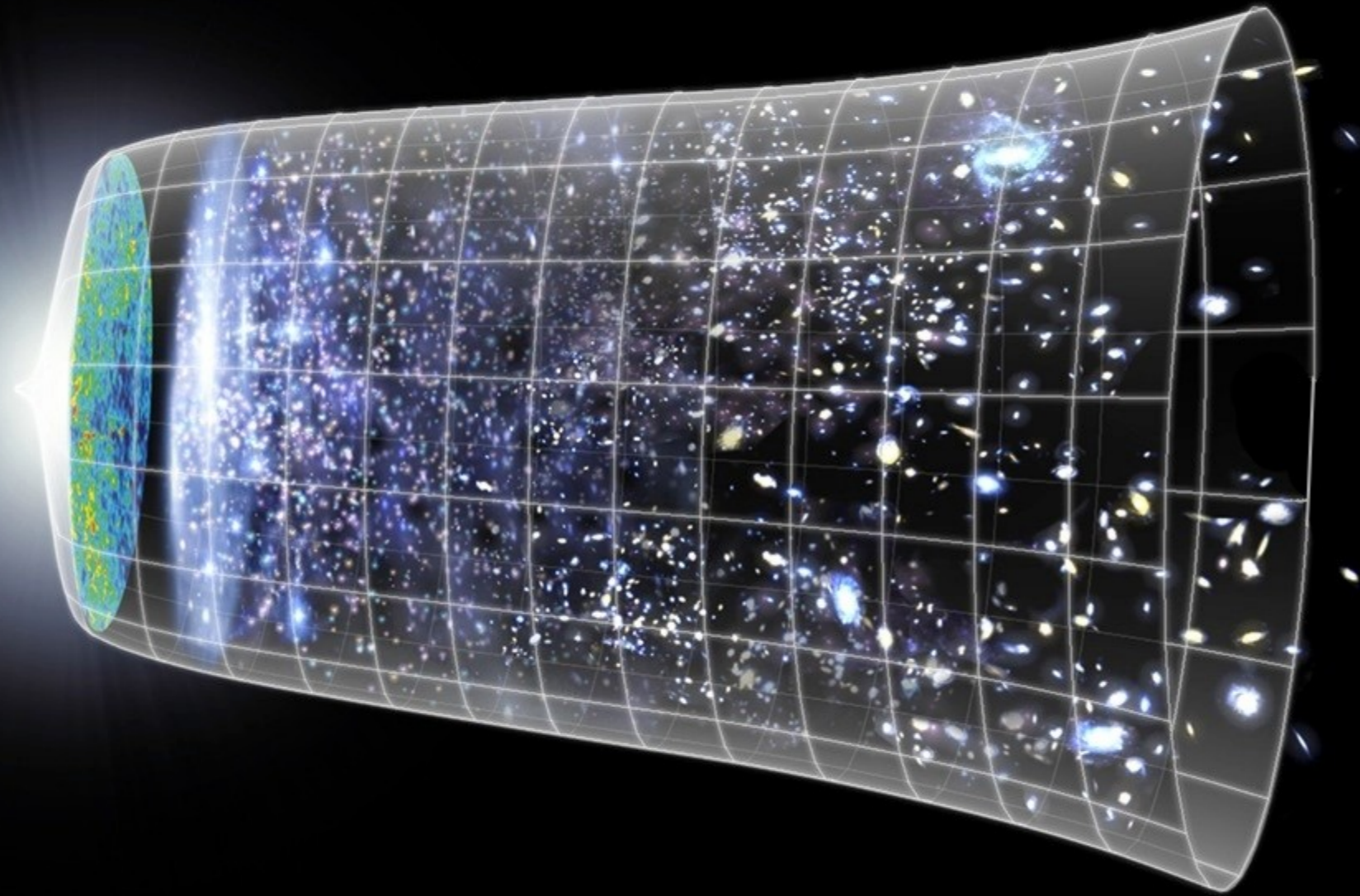


Why we need to care about unobservable modes

Based on

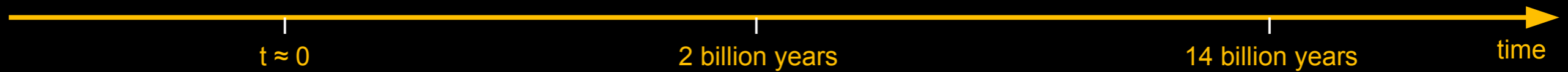
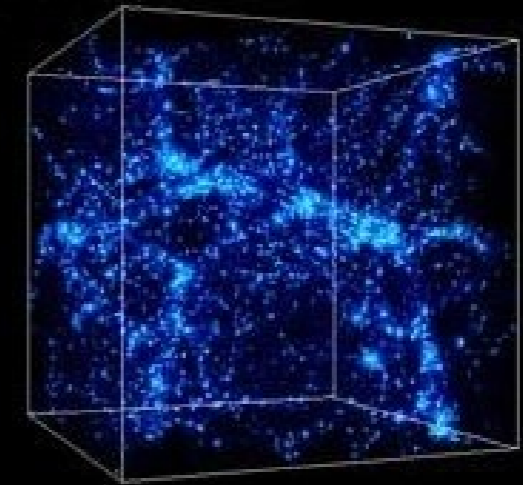
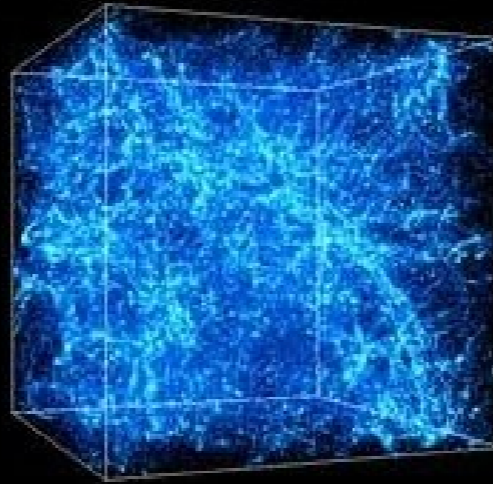
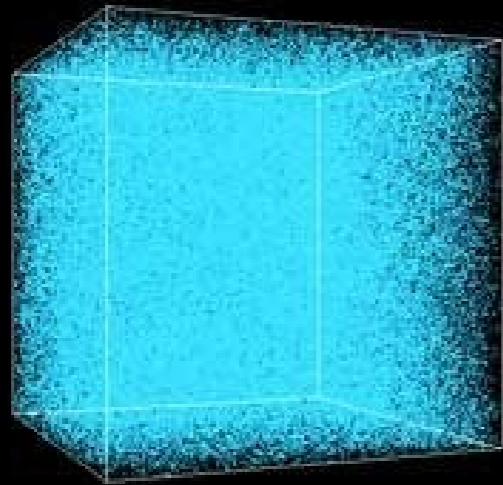
Lacasa & Rosenfeld 2016, Lacasa et al. 2018,
Lacasa & Kunz 2017, Lacasa & Grain 2019,
Gouyou Beauchamps et al. 2022,
Lacasa et al. [arXiv:2209.14421](https://arxiv.org/abs/2209.14421)

Context



NASA/WMAP Science Team

Time and (non-) linearity



Dynamics
of density
fluctuations

linear

non linear



Gravity

induces

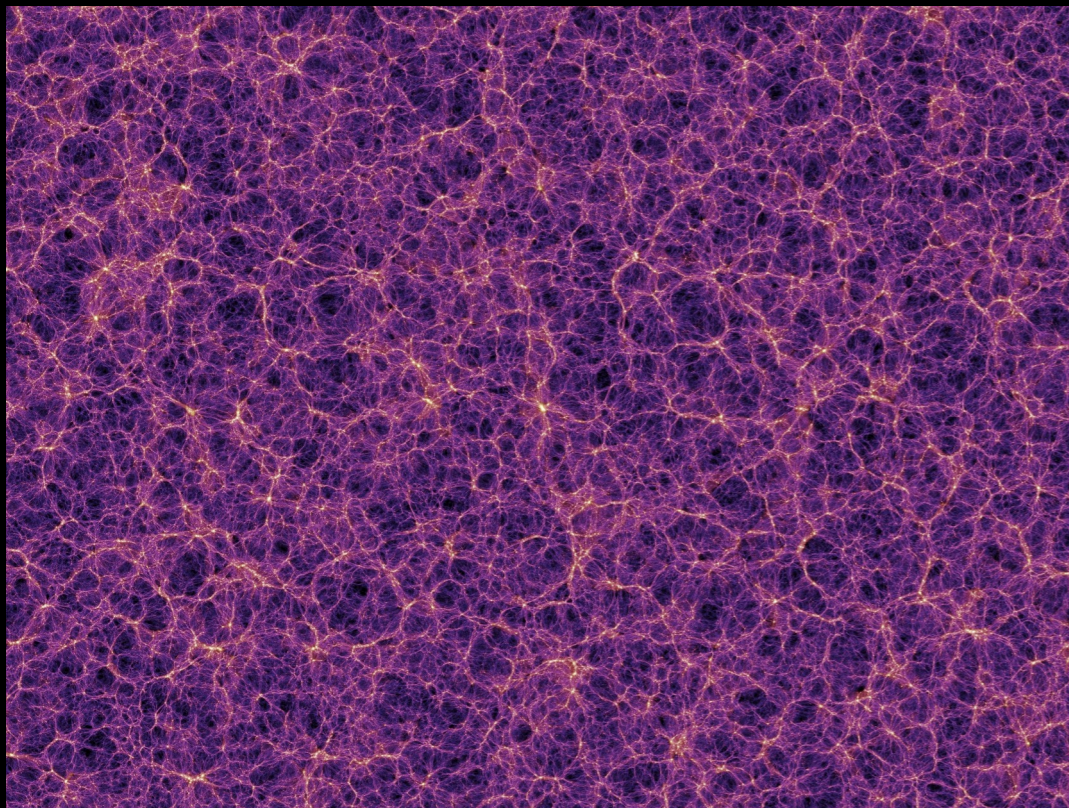


Statistical
distribution

Gaussian

non Gaussian

Scales and (non-) linearity

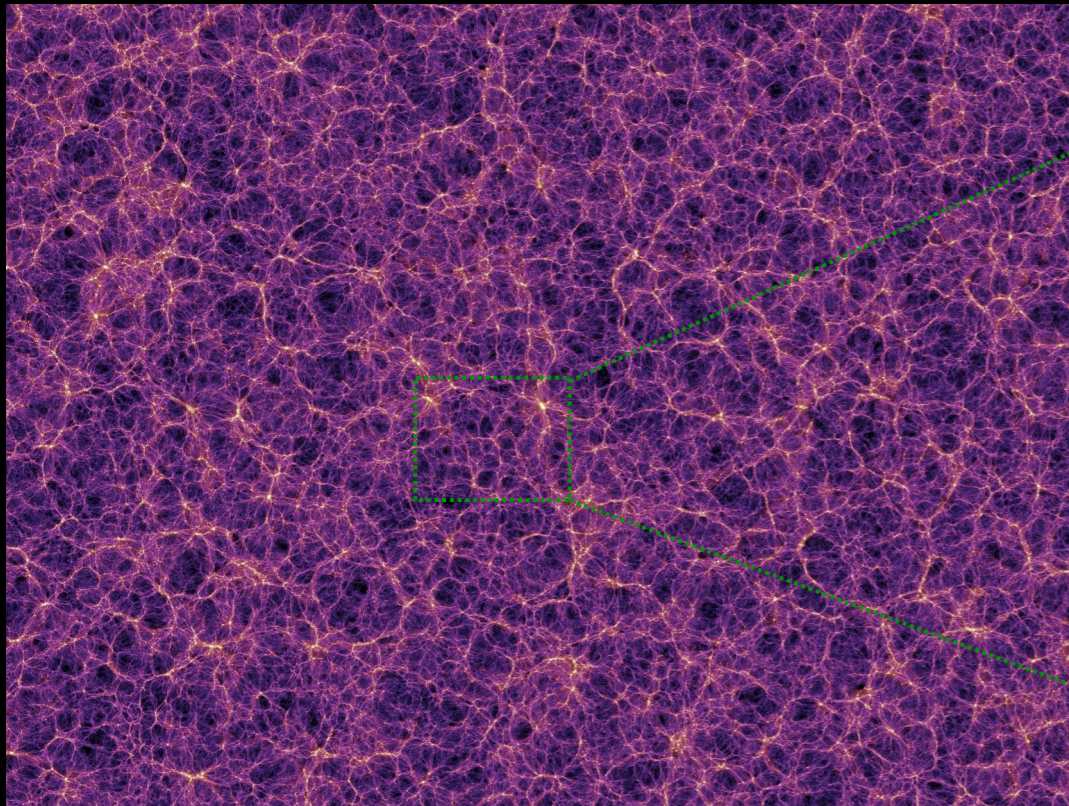


Millenium simulation

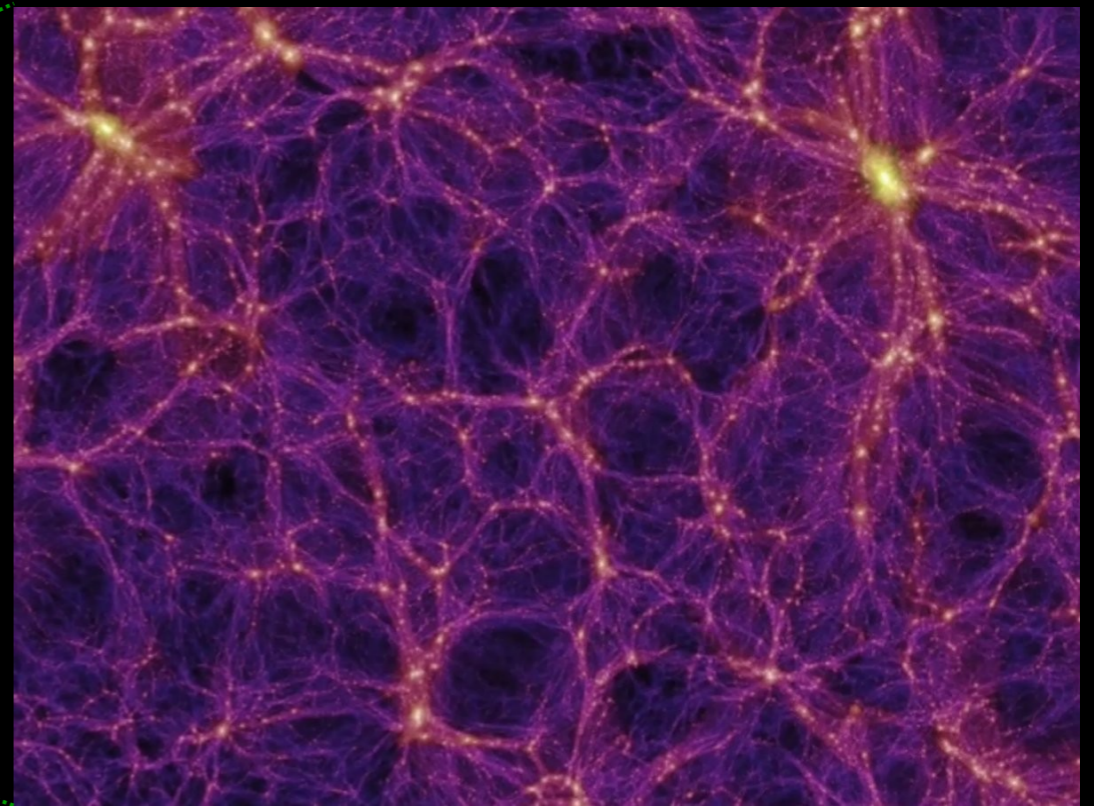
large scales



Scales and (non-) linearity



Millenium simulation



Photometric surveys

Euclid : 30 gals/arcmin²

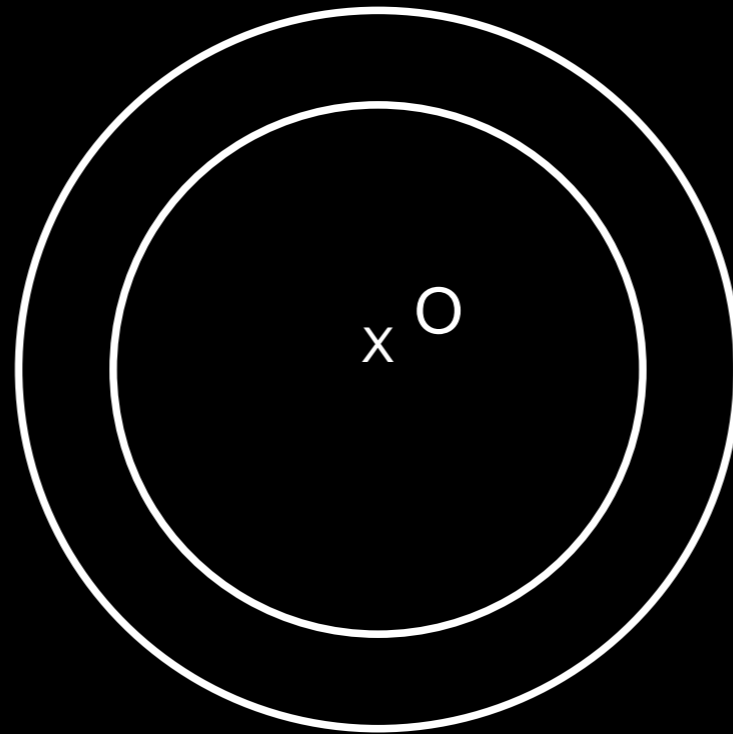
Rubin/LSST : 45 gals/arcmin²

Radio surveys

SKA2 : +precise redshifts

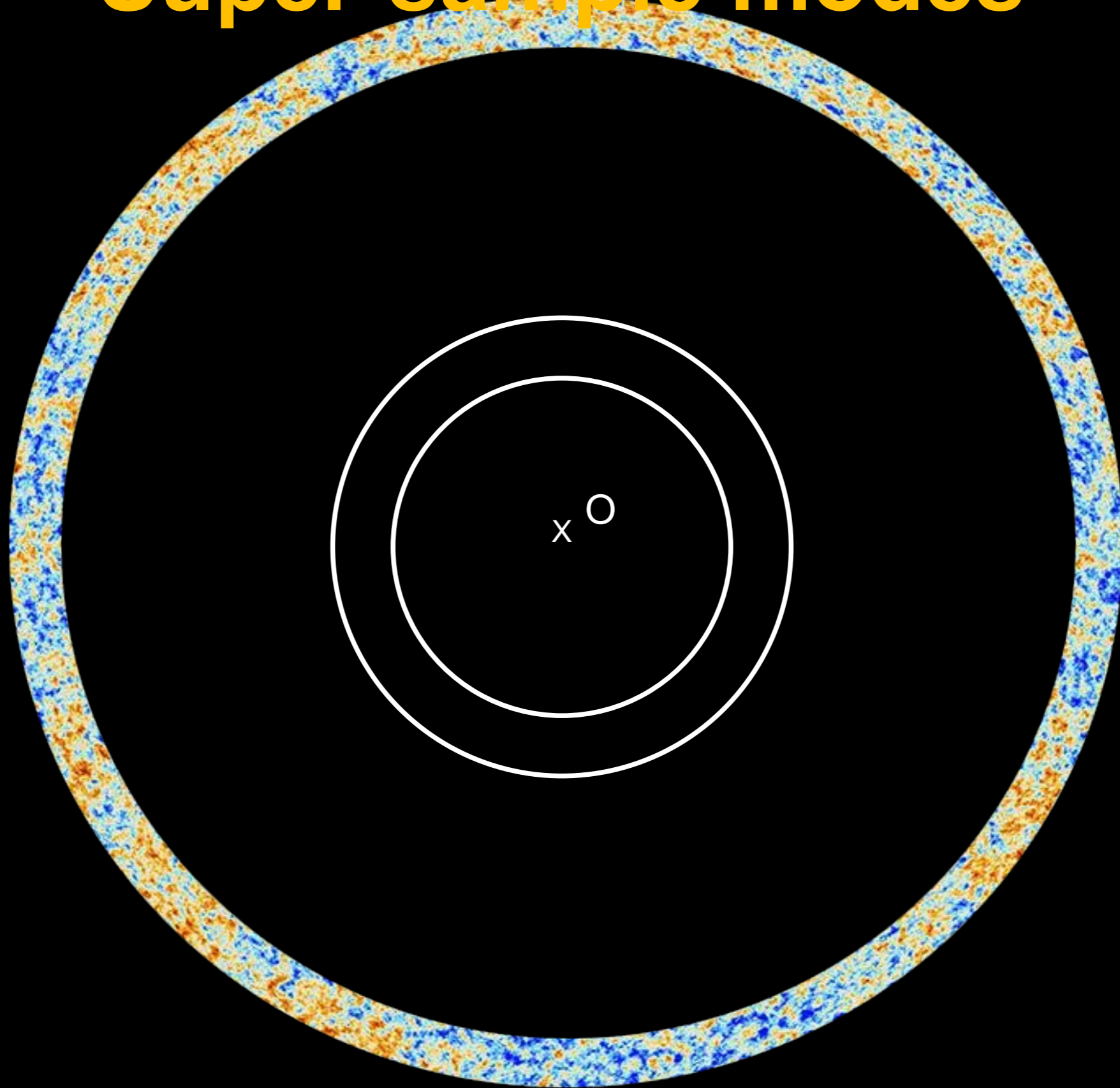
Super-sample modes and their impact

Super-sample modes



Finite z bin = finite volume
(even in full sky)

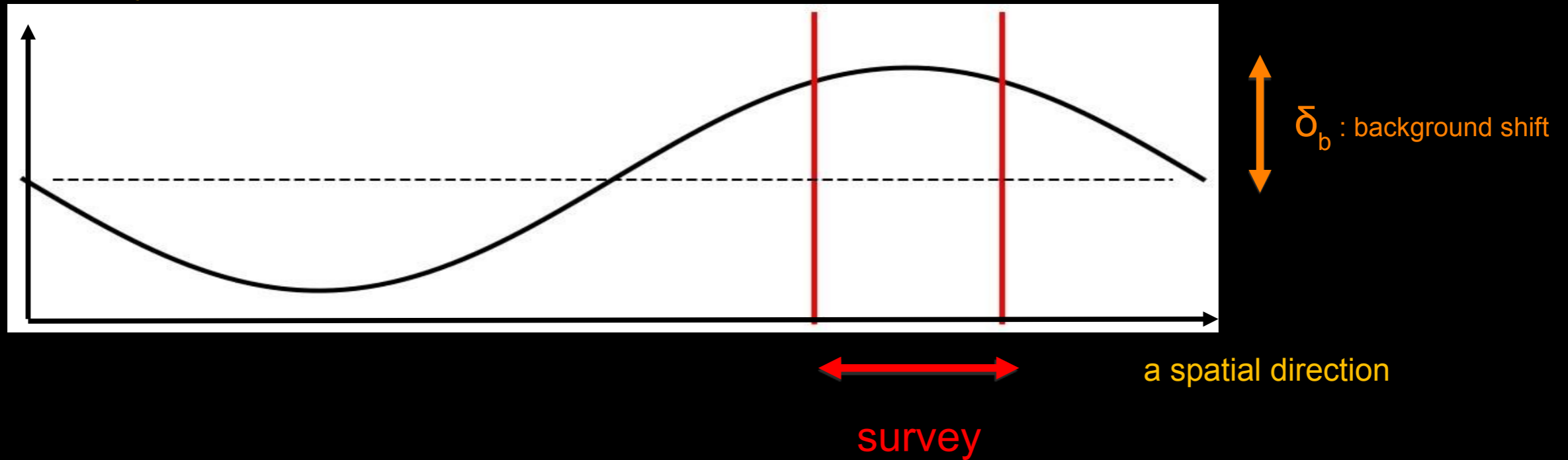
Super-sample modes



Super-sample fluctuations exist

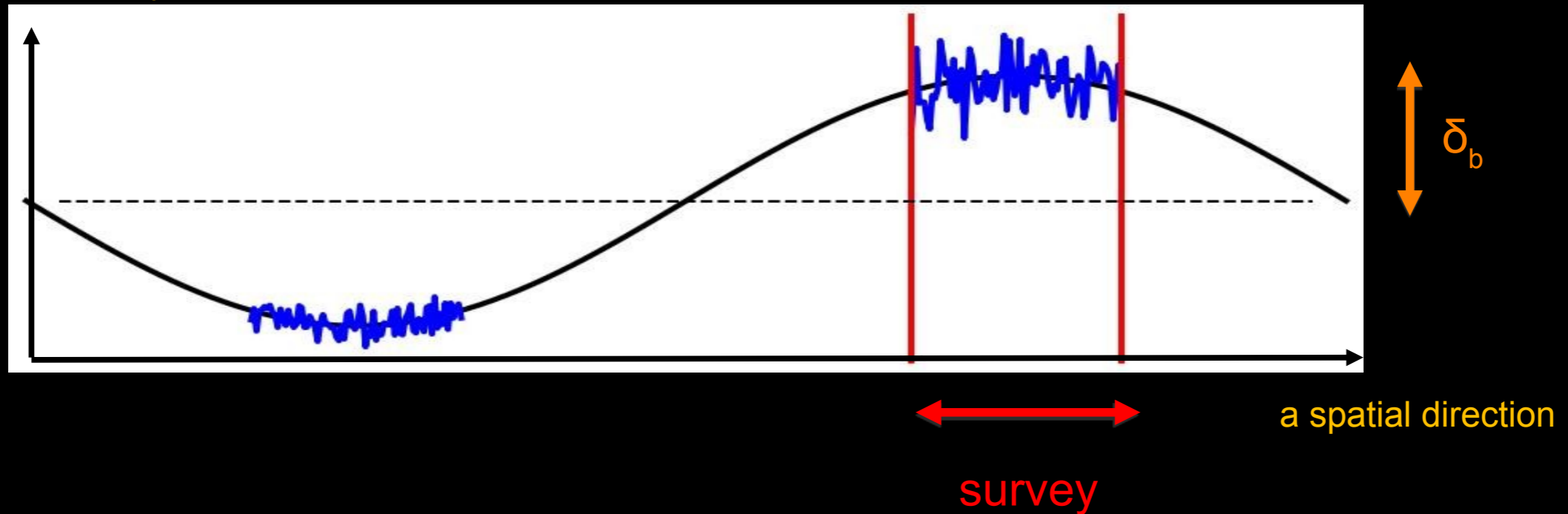
Super-sample covariance (SSC)

Matter density



Super-sample covariance (SSC)

Matter density



Feedback/backreaction of long on short modes

Equivalent to evolution in separate Universe

→ impacts all probes

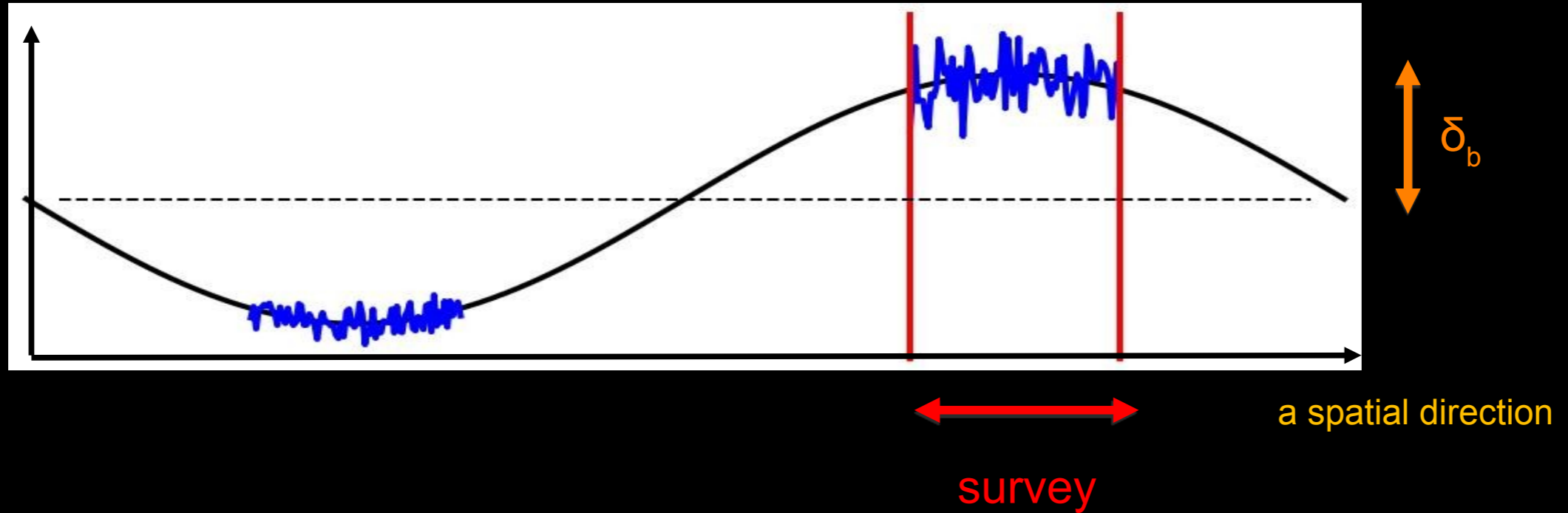
Galaxies, clusters, lensing, HI...

→ impacts all statistics

Power spectrum, bispectrum, machine learning, field level inference...

Super-sample covariance (SSC)

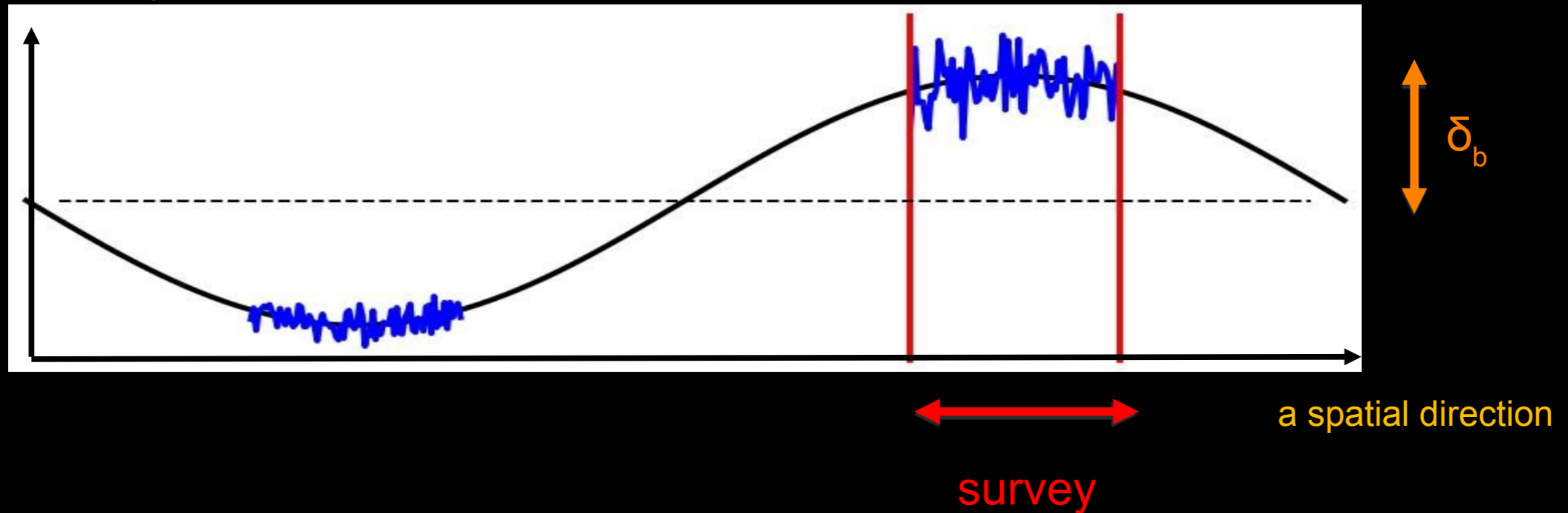
Matter density



$$\text{Cov} = \text{Cov}_G + \text{Cov}_{\text{SSC}} + \text{Cov}_{\text{oNG}}$$

Super-sample covariance (SSC)

Matter density



$$\text{Cov} = \text{Cov}_G + \text{Cov}_{\text{SSC}} + \text{Cov}_{\text{ONG}}$$

$$\text{Cov}_{\text{SSC}}(\mathcal{O}_1, \mathcal{O}_2) = \int dz_1 dz_2 \cdots \frac{\partial \mathcal{O}_1}{\partial \delta_b} \frac{\partial \mathcal{O}_2}{\partial \delta_b} \sigma^2(z_1, z_2)$$

Reaction of observable to background shift

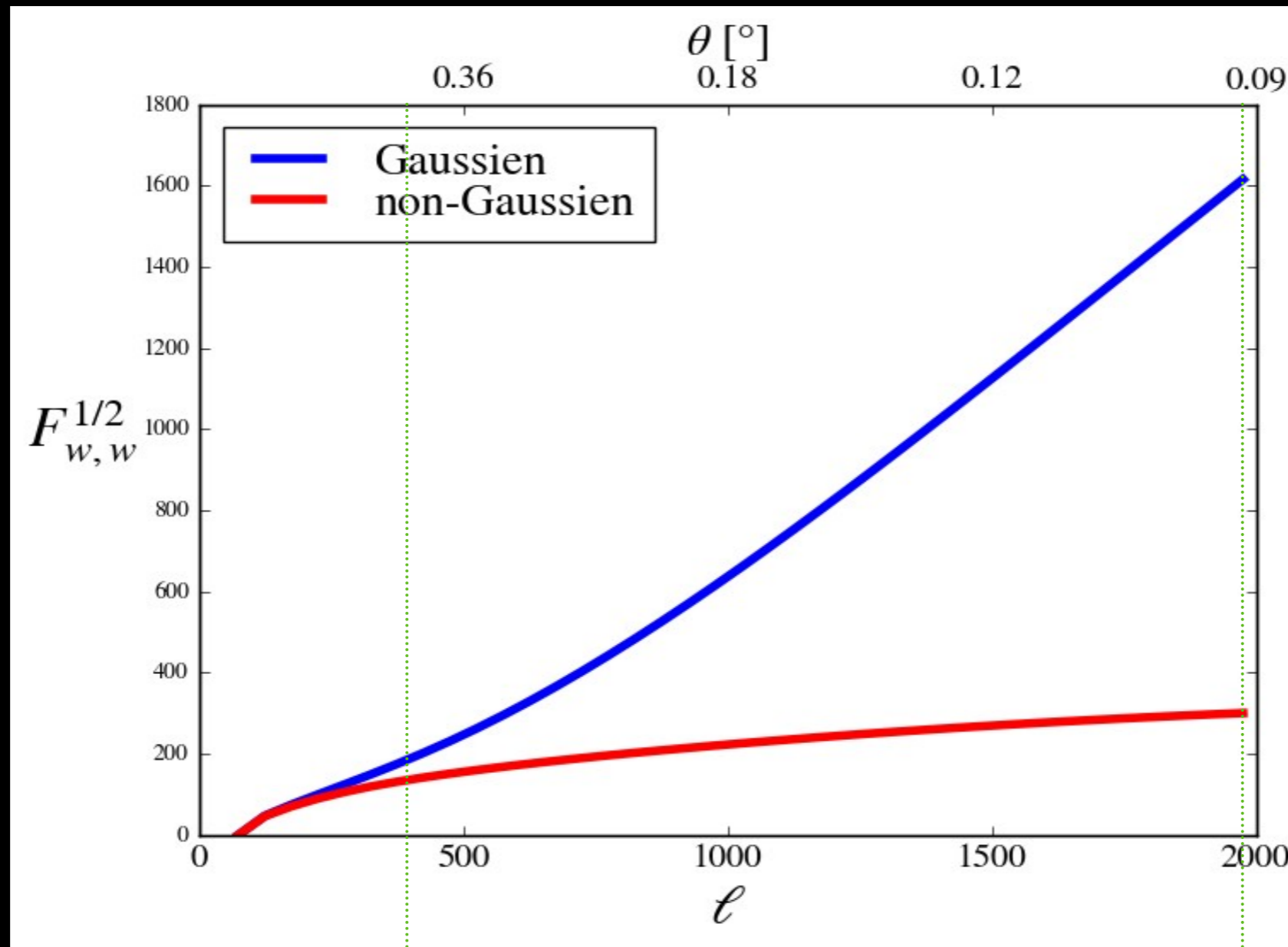
Variance of background shift

Euclid

and making SSC popular

Impact of SSC for Euclid : first look

Fisher information on Dark Energy (unmarginalized)
forecasts for photometric galaxy C_l



↓ ÷ 5

angular resolution of analysis

~ previous DES analysis

Euclid target

Fast and easy SSC

$$\text{Cov}_{\text{SSC}}(\mathcal{O}_1, \mathcal{O}_2) = \int dz_1 dz_2 \frac{\partial \mathbf{o}_1}{\partial \delta_b} \frac{\partial \mathbf{o}_2}{\partial \delta_b} \sigma^2(z_1, z_2)$$



$$\text{Cov}_{\text{SSC}}(\mathcal{O}_1, \mathcal{O}_2) \approx R_1 \mathcal{O}_1 R_2 \mathcal{O}_2 S_{i,j}$$

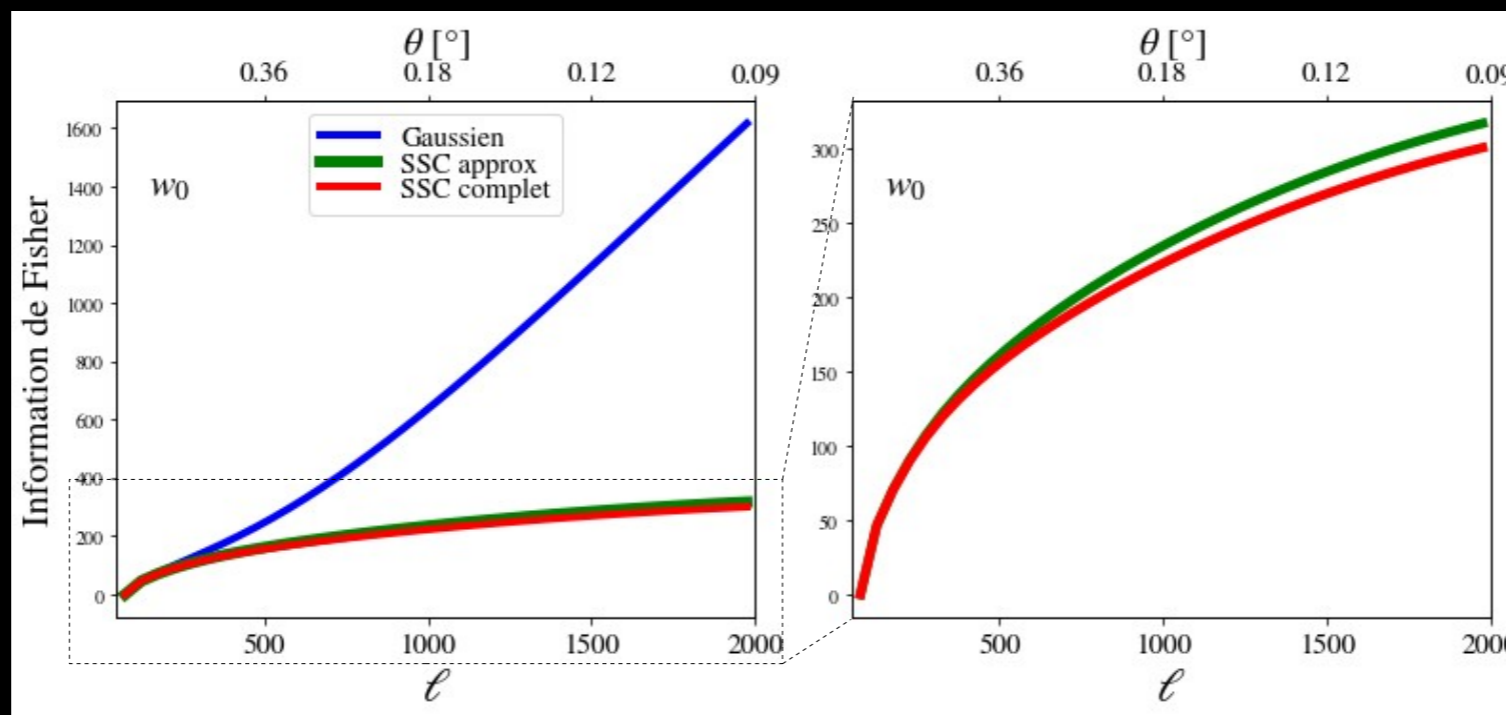
Lacasa & Grain 2019

dimensionless response

can be computed with PySSC

<https://github.com/fabienlacasa/PySSC>

<https://pyssc.readthedocs.io>



Validation of the SSC approximation: DE Fisher information

Impact on constraining Dark Energy

Euclid-like forecast for Dark Energy Figure of Merit (FoM)

with weak lensing (WL), photometric galaxy clustering (Gcph)
and their cross-correlation (XC)

| Probe | Survey area [deg ²] | Gaussian | Full-sky SSC | |
|------------|---------------------------------|----------|--------------|-------|
| WL | 5 | 0.014 | 0.009 | |
| | 15000 | 43.12 | → 26.329 | - 40% |
| GCph | 5 | 0.035 | 0.029 | |
| | 15000 | 103.71 | → 88.636 | - 15% |
| GCph+WL+XC | 5 | 0.346 | 0.153 | |
| | 15000 | 1038.13 | → 454.59 | - 55% |

“full-sky” = rescaled by sky fraction
(as tradition for Gaussian covariance)

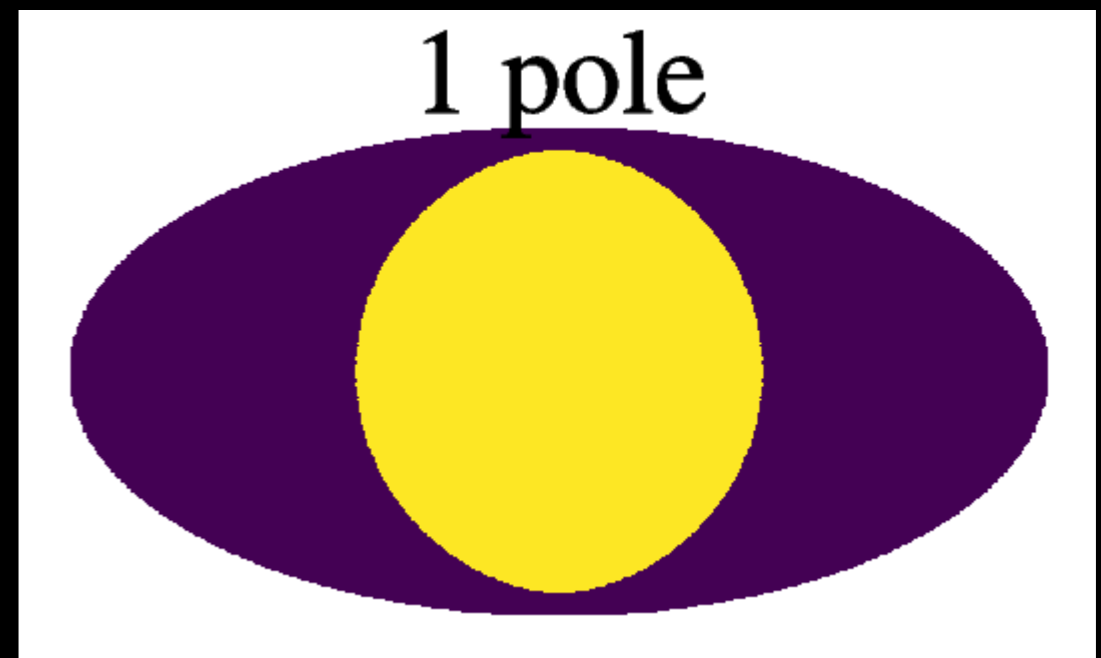
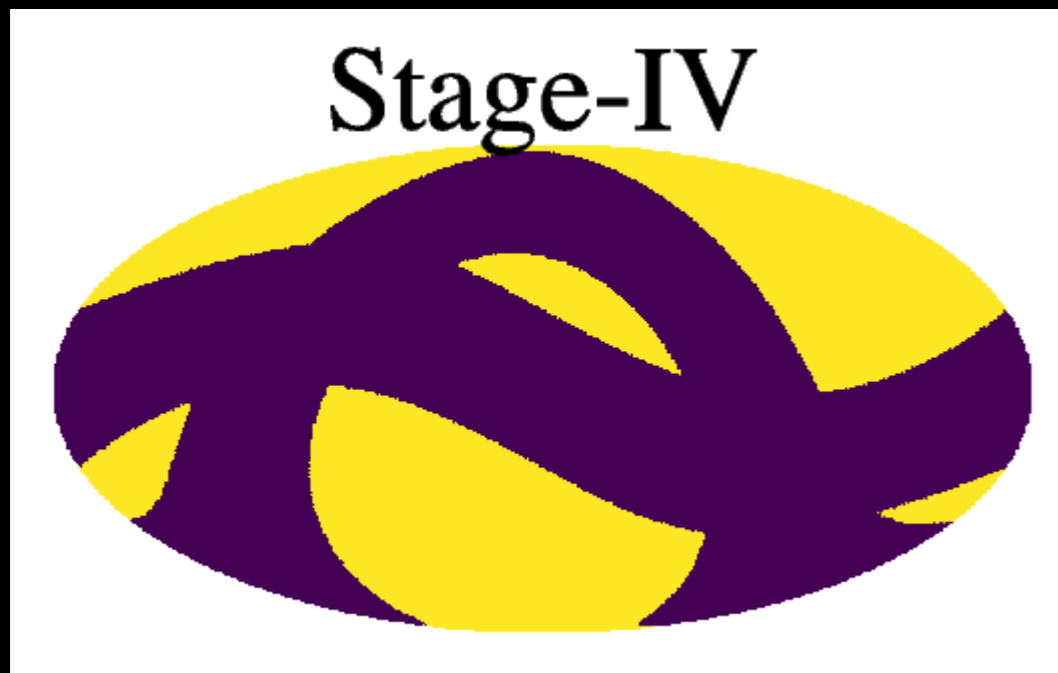
Gouyou Beauchamps et al. 2022

Fast and easy SSC, in partial sky

Team formed during Euclid France summer school 2019
S. Gouyou Beauchamps, I. Tutusaus, M. Aubert, P. Baratta, A. Gorce

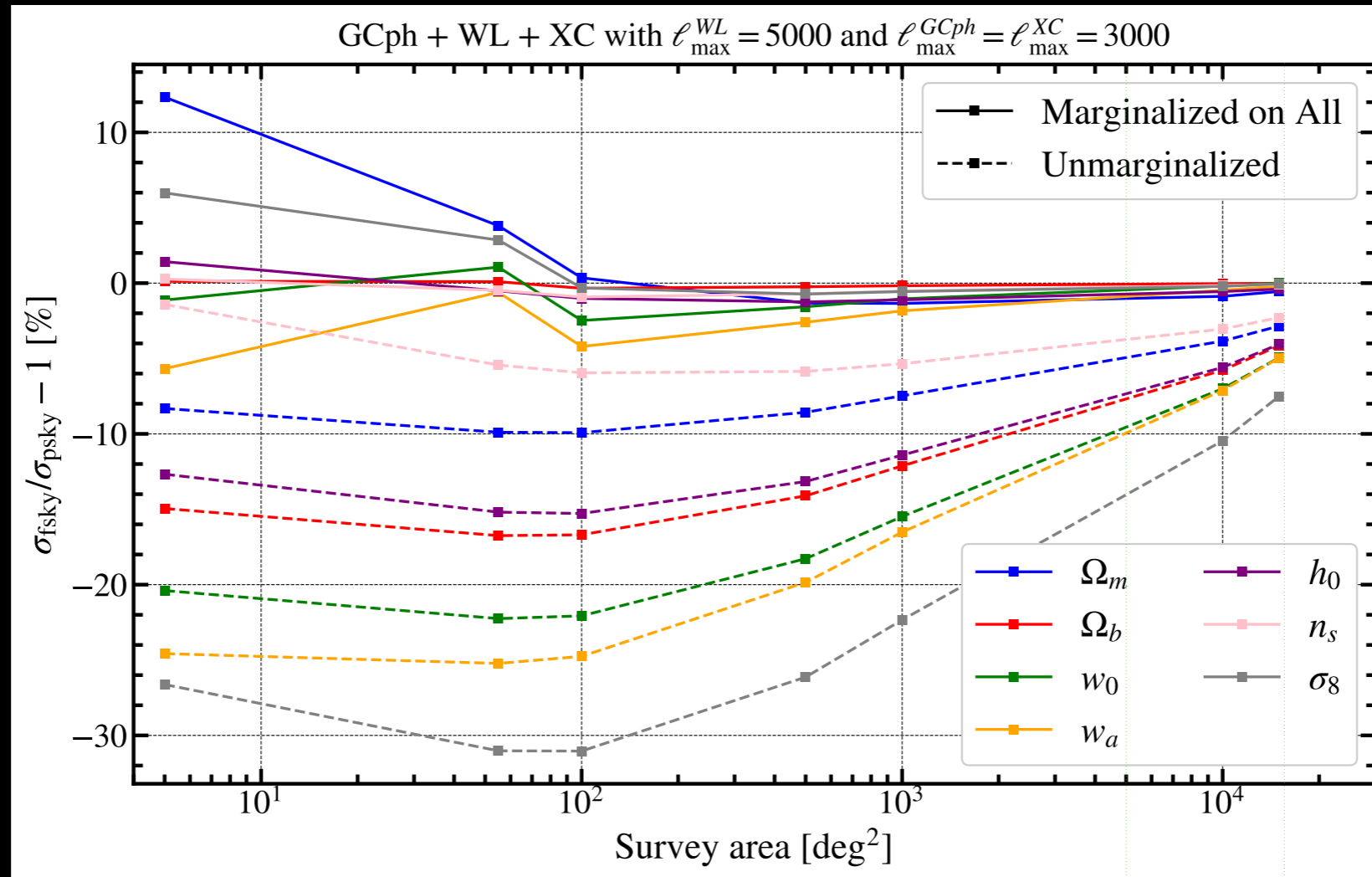
Formalism to account for survey mask in SSC

New PySSC release
<https://pyssc.readthedocs.io>



Fast and easy SSC, in partial sky

Comparison of cosmo errors:
rigorous partial sky vs traditional rescaling



DES-like Euclid-like

Gouyou Beauchamps et al. 2022

A new approach to SSC

Problems of the trad approach (=numerical covariance)

Compared to Gaussian covariance, SSC matrix

- is highly off-diagonal
- has low rank

→ **Problems**

- memory and CPU cost is higher
- numerical inversion can be unstable (ill-conditioned problem)
- does not take advantage of a diagonal covariance

Solution: invert total covariance analytically (Woodbury identity)

→ analytical formulae for likelihood and Fisher

SSC directly for likelihood or Fisher

Likelihood is a correction to the noSSC case

$$\ln \mathcal{L} = \ln \mathcal{L}_{\text{noSSC}} + \delta \ln \mathcal{L}$$

Can also get Fisher or inverse Fisher matrices

$$F = F_{\text{noSSC}} - Y^T (1 + I(\delta_b) S)^{-1} Y$$

$$F^{-1} = F_{\text{noSSC}}^{-1} + Z^T \Sigma^2(\delta_b) Z$$

$I(\delta_b)$: Fisher info on δ_b (unmarginalised). Y : contains probes responses.

Interests:

- updating a pipeline at low cost
- fast if noSSC is diagonal

Conclusions

Conclusions

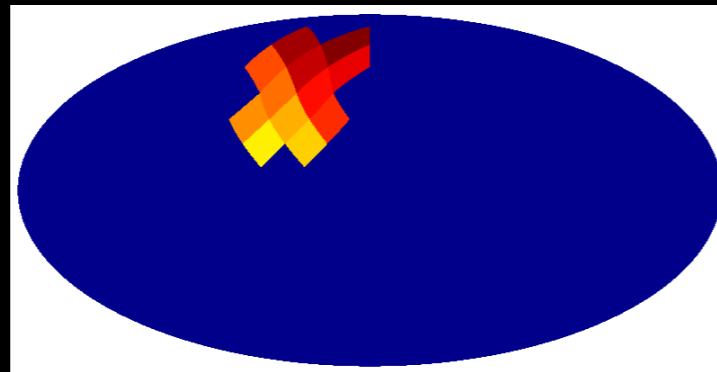
- Long wavelength modes are unobservables but backreact on observed modes
→ super-sample covariance (SSC)
- SSC correlates all probes, even if independent otherwise
- SSC importance increases with statistical power
Ex: push to high resolution, combine probes, combine surveys.
=> Impact for stage IV constraints on Dark Energy, σ_8 , Ω_m ... (modified gravity?)
- Fast & easy computation possible: public code `PySSC`
including mask effect
<https://github.com/fabienlacasa/PySSC>
<https://pyssc.readthedocs.io>
- New method directly at level of likelihood or Fisher
→ easier, faster, robust
Currently implementing it for Euclid + CMB

Thank you

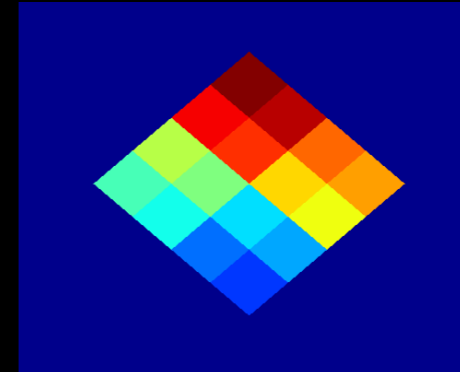
Additional slides

SSC and traditional covariance methods

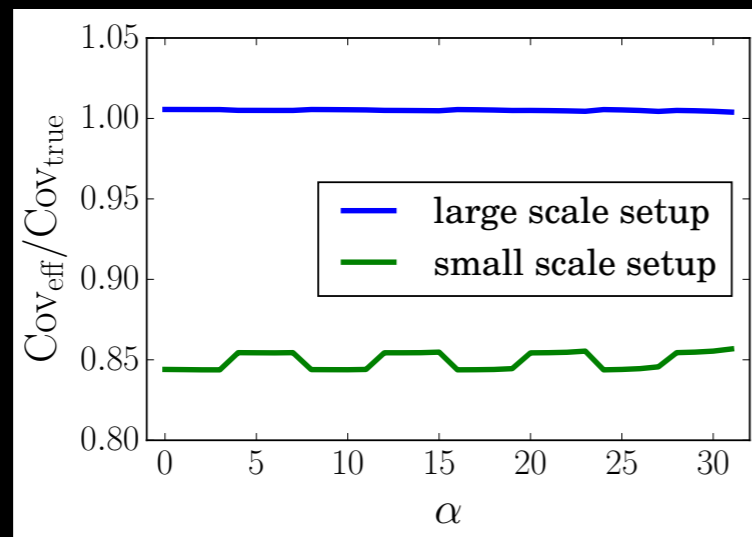
Method 1: survey footprint replicated at different points of a large simulation



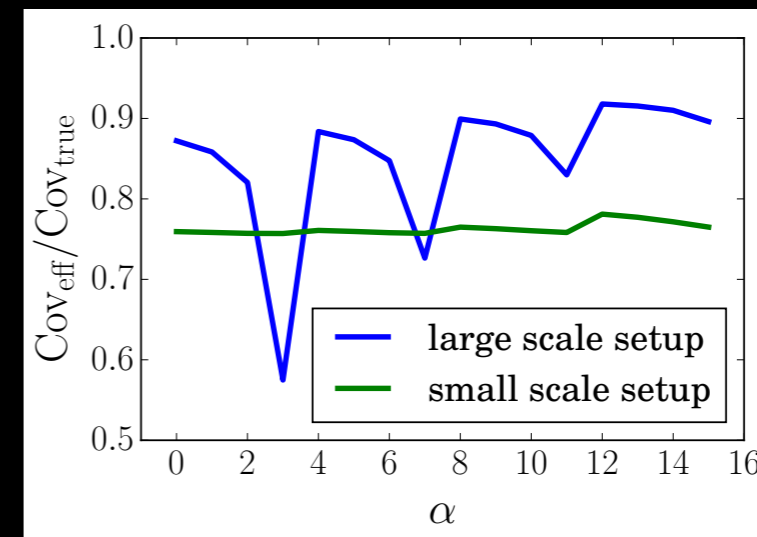
large scale setup



small scale setup



same redshifts



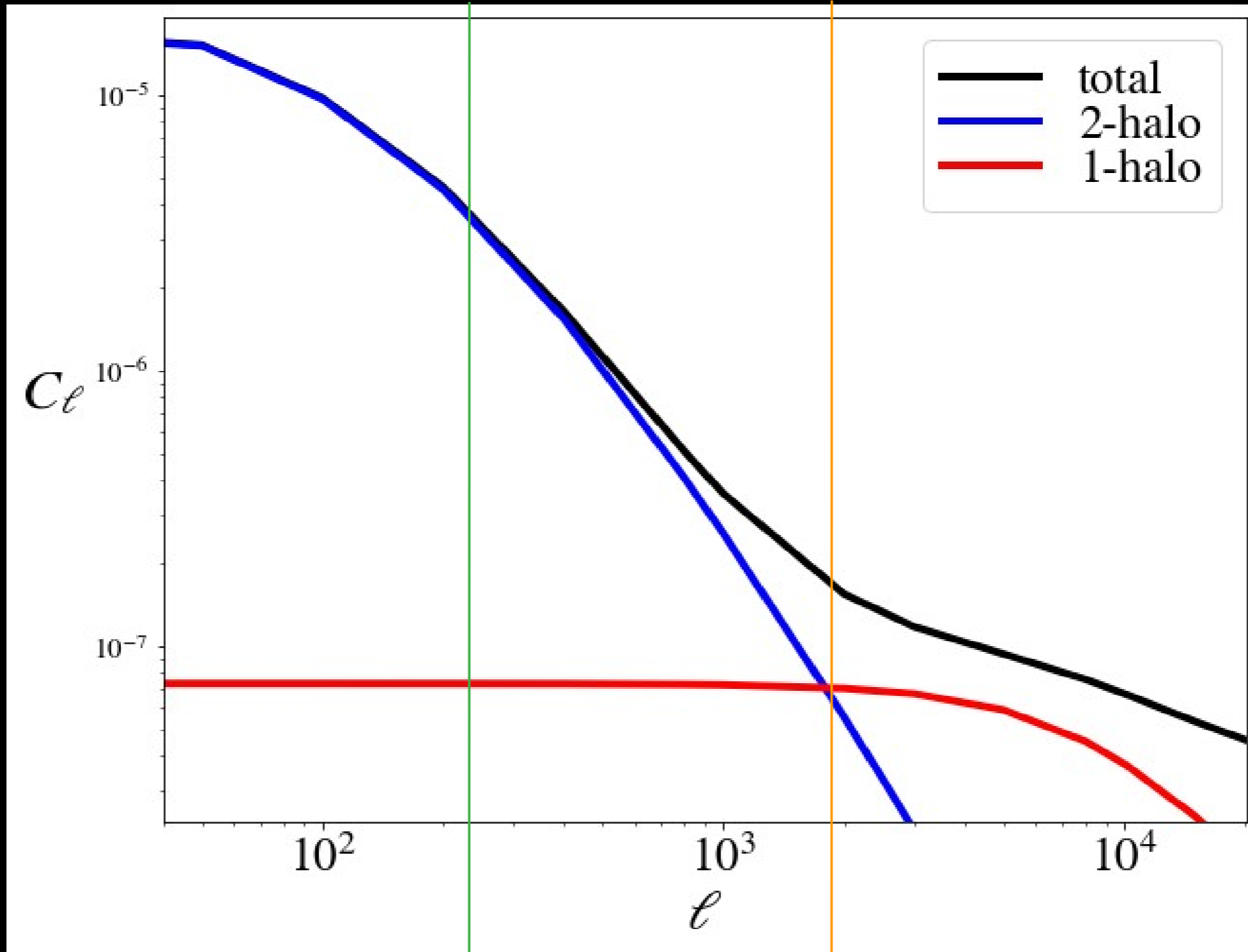
two different redshifts

Method 2 (jackknife):

- Random subsamples removed from data → fake data realisations
- Apply canonic covariance estimator and rescale to survey area

Works for classic covariance terms. For SSC: biased by up to factor 5.

Scales



$z=1$

Regime

Linear

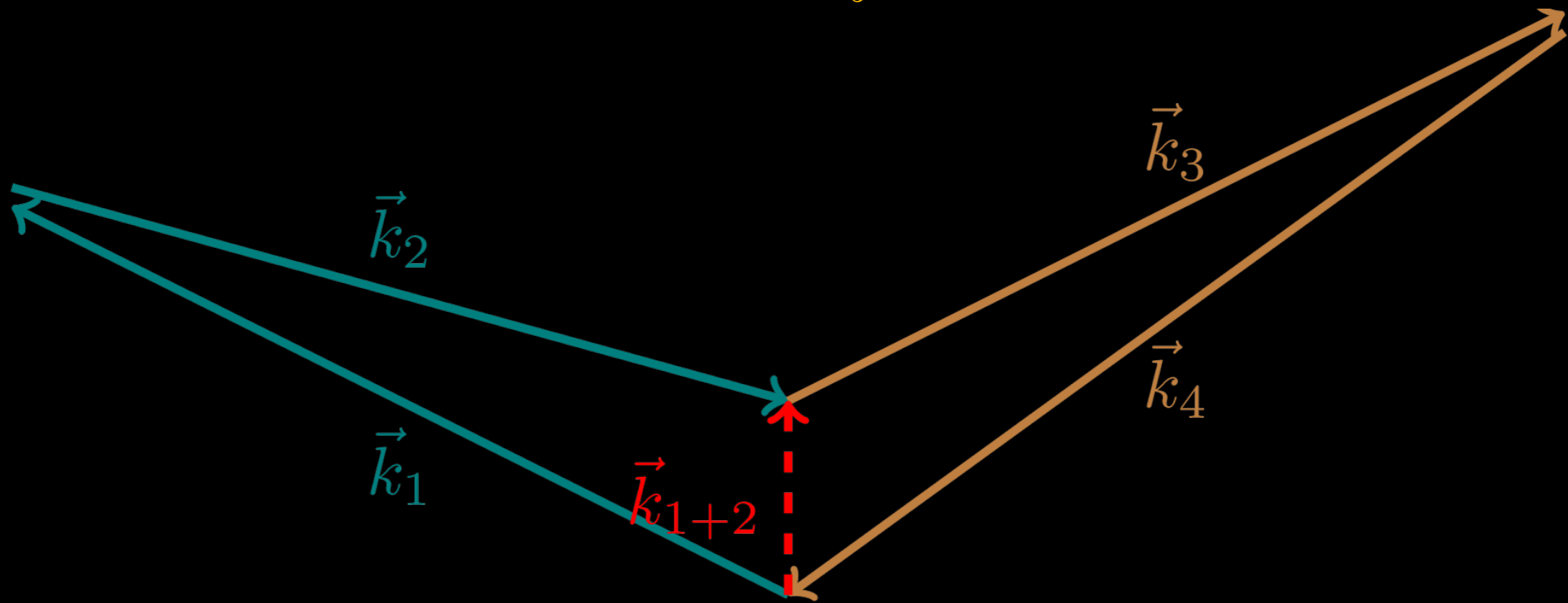
Mildly
non-linear

Highly
non-linear

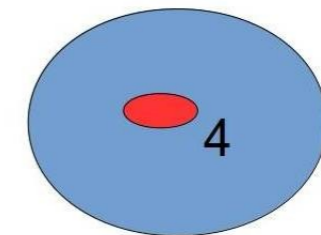
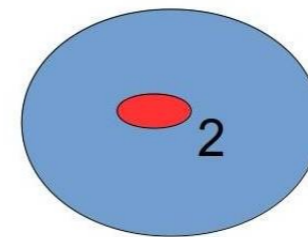
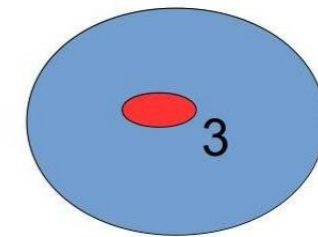
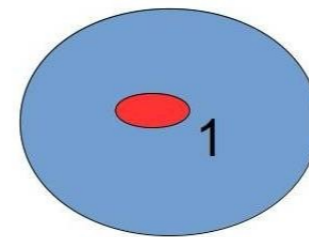
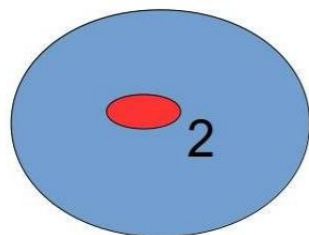
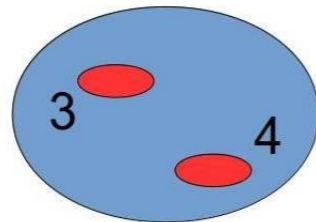
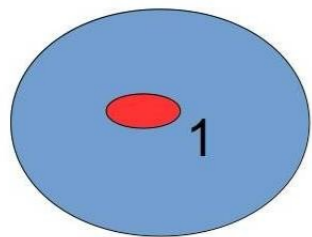
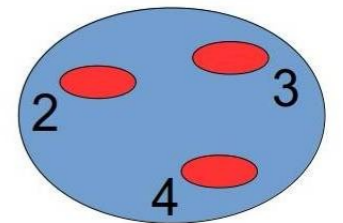
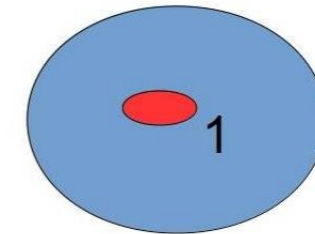
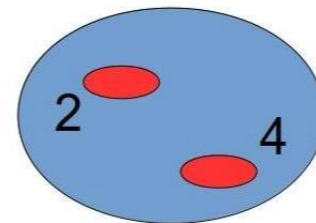
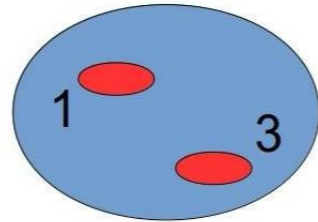
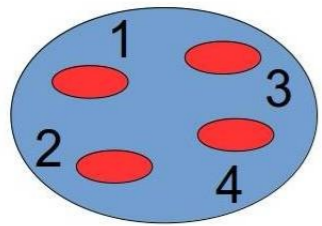
Covariance and 4-point function

$$\text{Cov} (C_\ell, C_{\ell'}) = \text{Cov}_G + \text{Cov}_{NG}$$

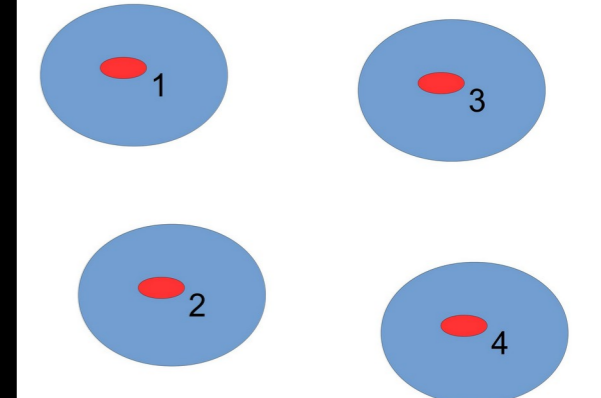
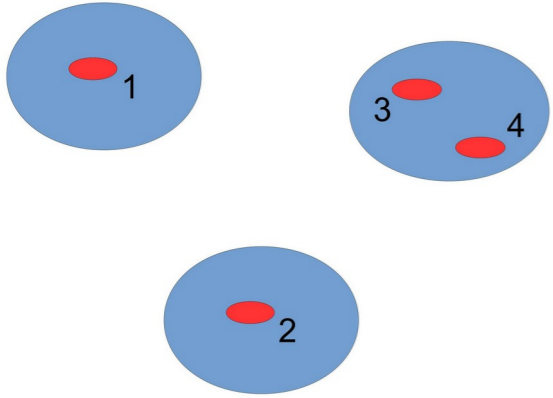
$$\text{Cov}_{NG} \propto T_{\ell, \ell'}^{\ell', \ell'} (L = 0) \propto \int T_{\text{gal}}(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4)$$



Diagrams for Cov(CI)



Higher order terms



Context recall

$$\delta_h(\mathbf{x}) = b_1 \delta_m(\mathbf{x}) + b_2 (\delta_m(\mathbf{x})^2 - \langle \delta_m^2 \rangle) + b_{s^2} s^2(\mathbf{x}) + \dots = \sum_{\mathcal{O}} b_{\mathcal{O}} \mathcal{O}(\mathbf{x})$$

$$\delta_m = \delta_m^{(1)} + \delta_m^{(2)} + \delta_m^{(3)} \quad \delta_m^{(n)}(\mathbf{k}) = \int d^3 \mathbf{q}_{1\dots n} \delta_D(\mathbf{k} - \mathbf{q}_{1+\dots+n}) F_n(\mathbf{q}_{1\dots n}) \delta_m^{(1)}(\mathbf{q}_1) \dots \delta_m^{(1)}(\mathbf{q}_n)$$

3-halo term

$$T_{3h}(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) \propto B_{\text{halo}}(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3 + \mathbf{k}_4) + 5 \text{ perm.}$$

$$B_{\text{halo}} = B_{2\text{PT}} + B_{b_2} + B_{s^2}$$

$$B_{2\text{PT}}(k_{123} | M_{123}) = b_1(M_1) b_1(M_2) b_1(M_3) 2! F_2(\mathbf{k}_1, \mathbf{k}_2) P(k_1) P(k_2) + 2 \text{ perm.}$$

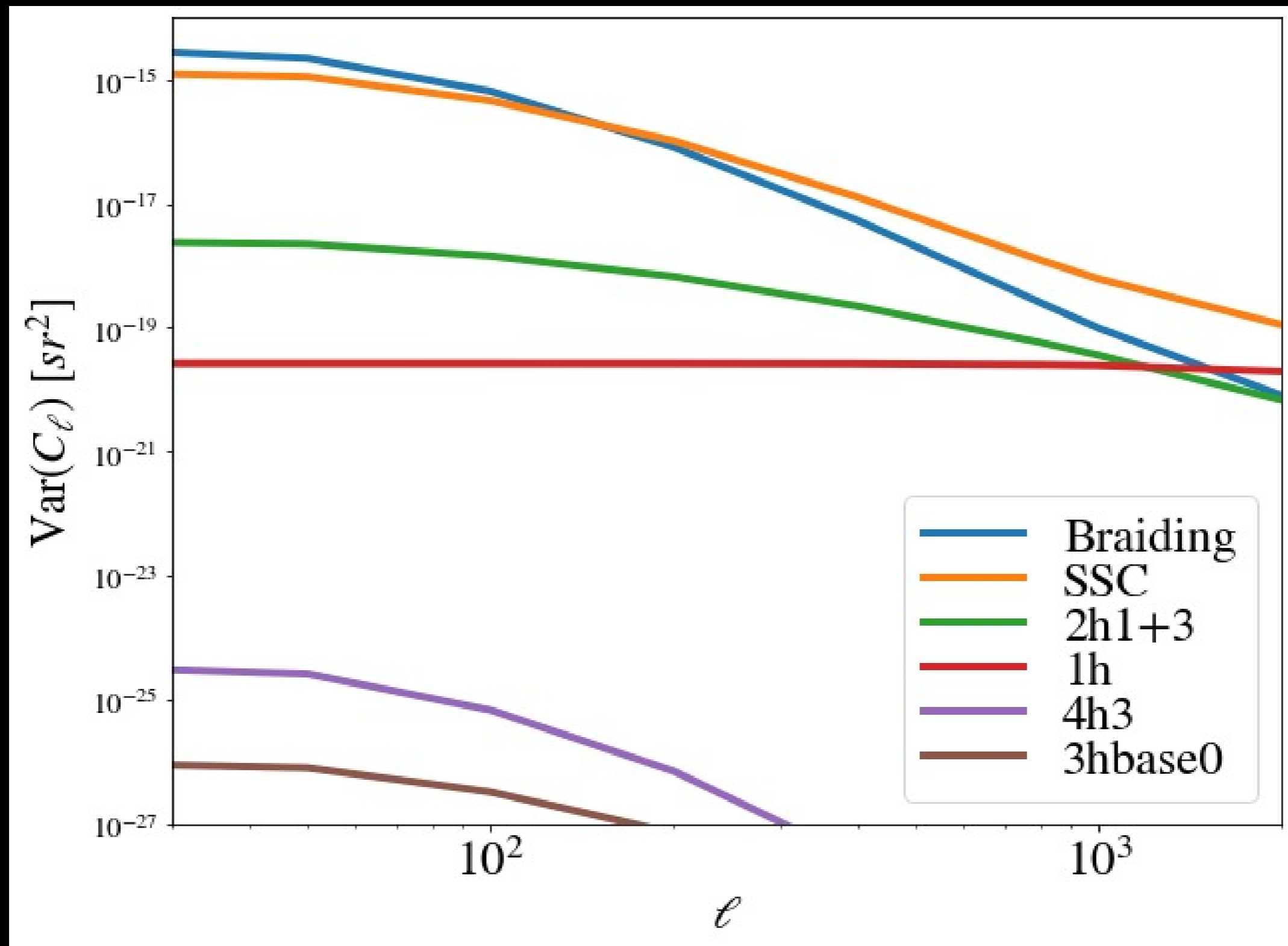
4-halo term

$$T_{4h}(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) \propto T_{\text{halo}}(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) = T^{2 \times 2} + T^{1 \times 3}$$

$$T^{2\text{PT} \times 2\text{PT}}(\mathbf{k}_{1234} | M_{1234}) = b_1(M_1) b_1(M_2) b_1(M_3) b_1(M_4) 4 F_2(\mathbf{k}_{1+3}, -\mathbf{k}_1) F_2(\mathbf{k}_{1+3}, \mathbf{k}_2) \\ \times P(k_{1+3}) P(k_1) P(k_2) + 11 \text{ perm.}$$

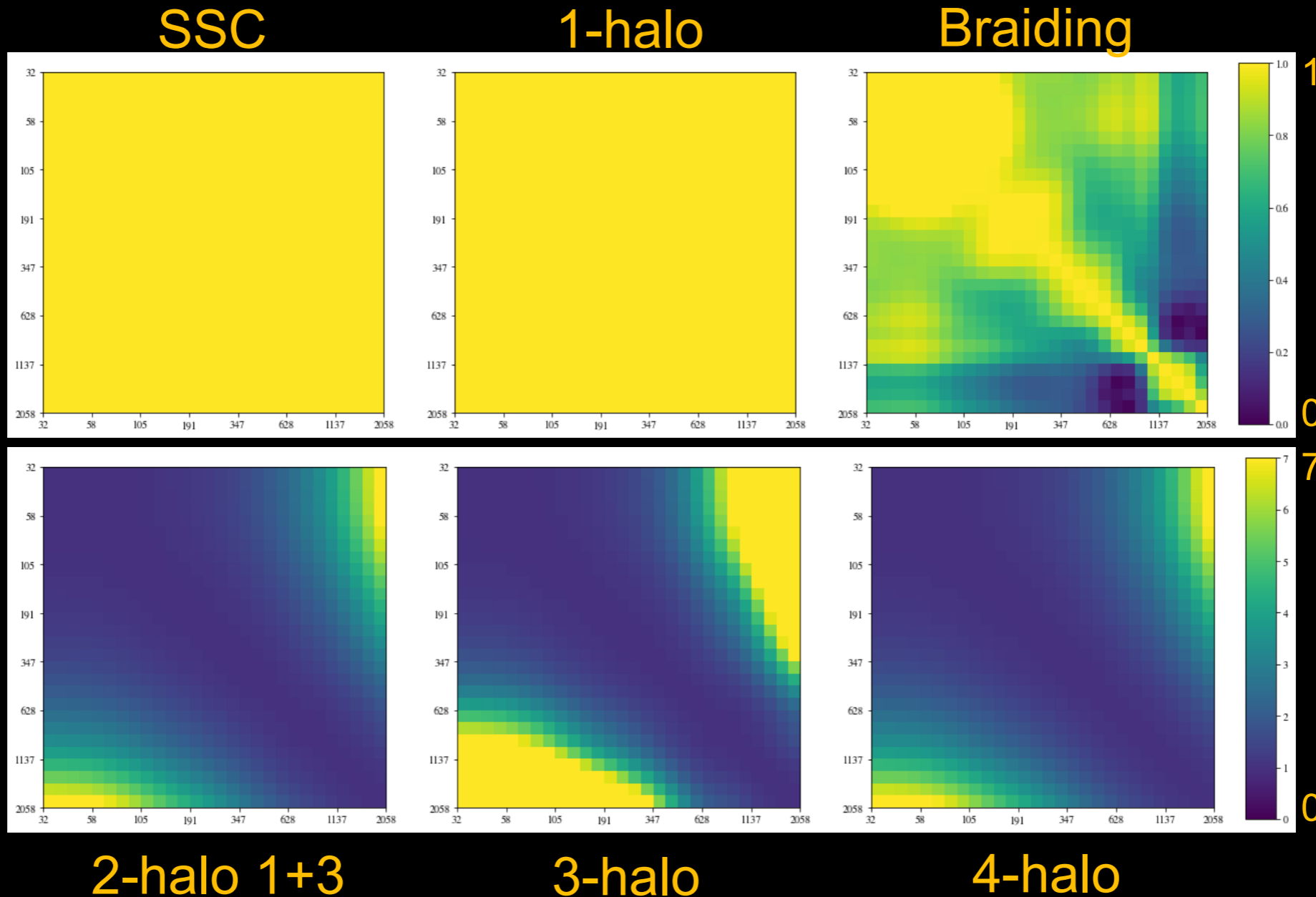
$$T^{3\text{PT}}(\mathbf{k}_{1234} | M_{1234}) = b_1(M_1) b_1(M_2) b_1(M_3) b_1(M_4) 3! F_3(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3) P(k_1) P(k_2) P(k_3) + 3 \text{ perm.}$$

Measurement error bars



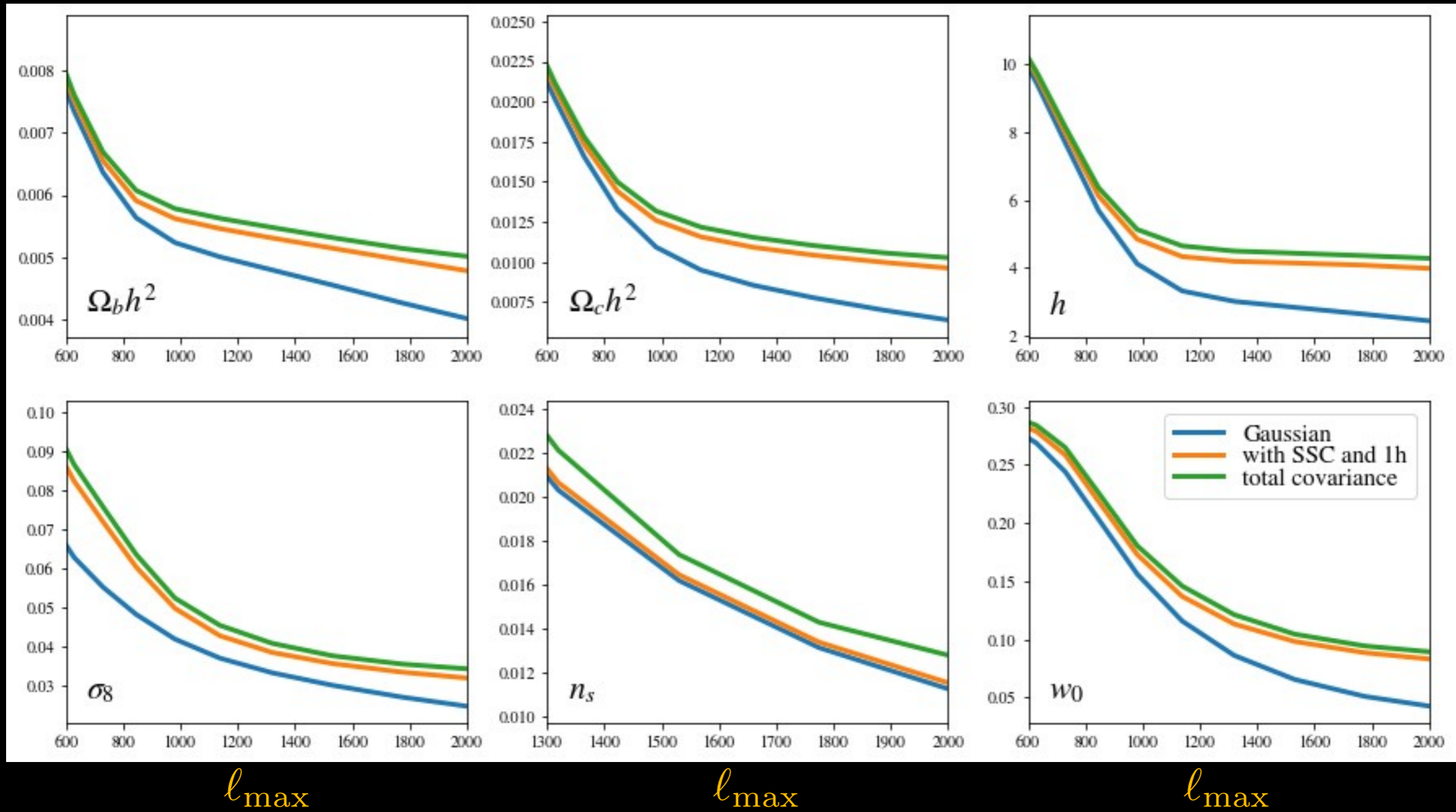
$z=1$

Covariance matrices : off-diagonal importance



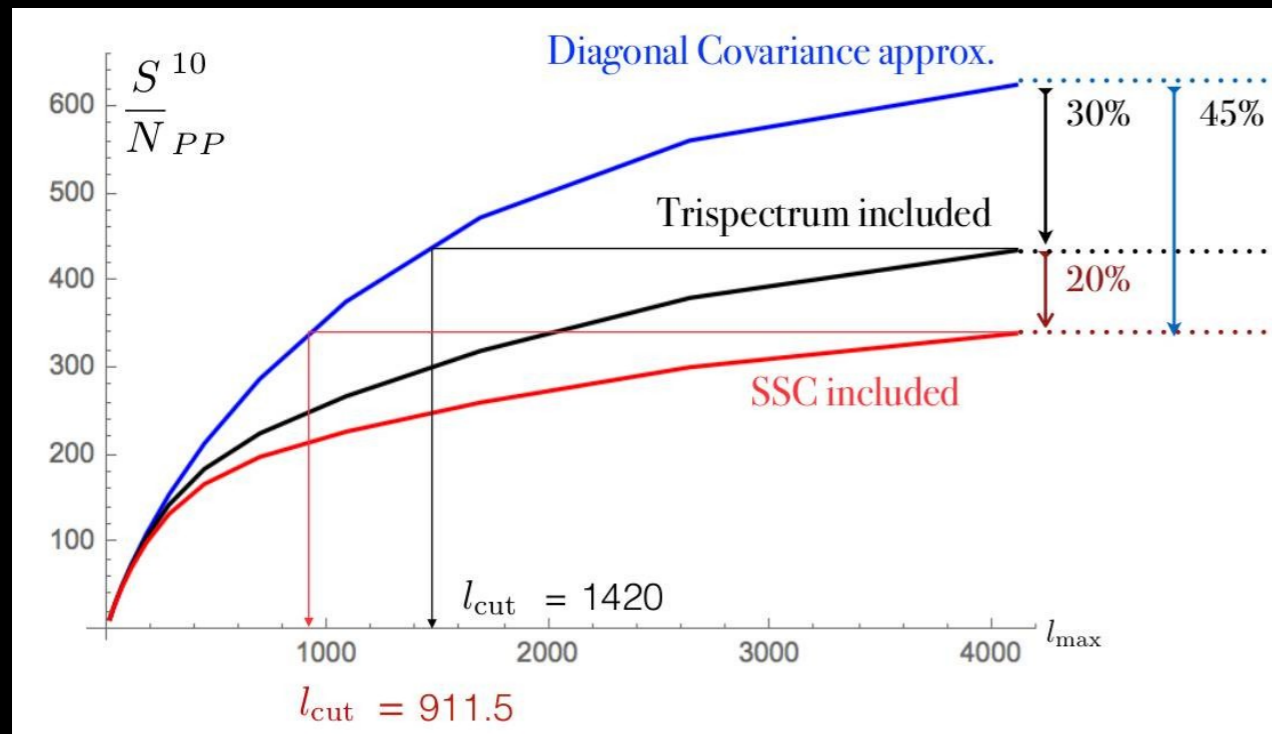
$$\frac{C_{ij}}{\sqrt{C_{ii}C_{jj}}}$$

Cosmological error bars



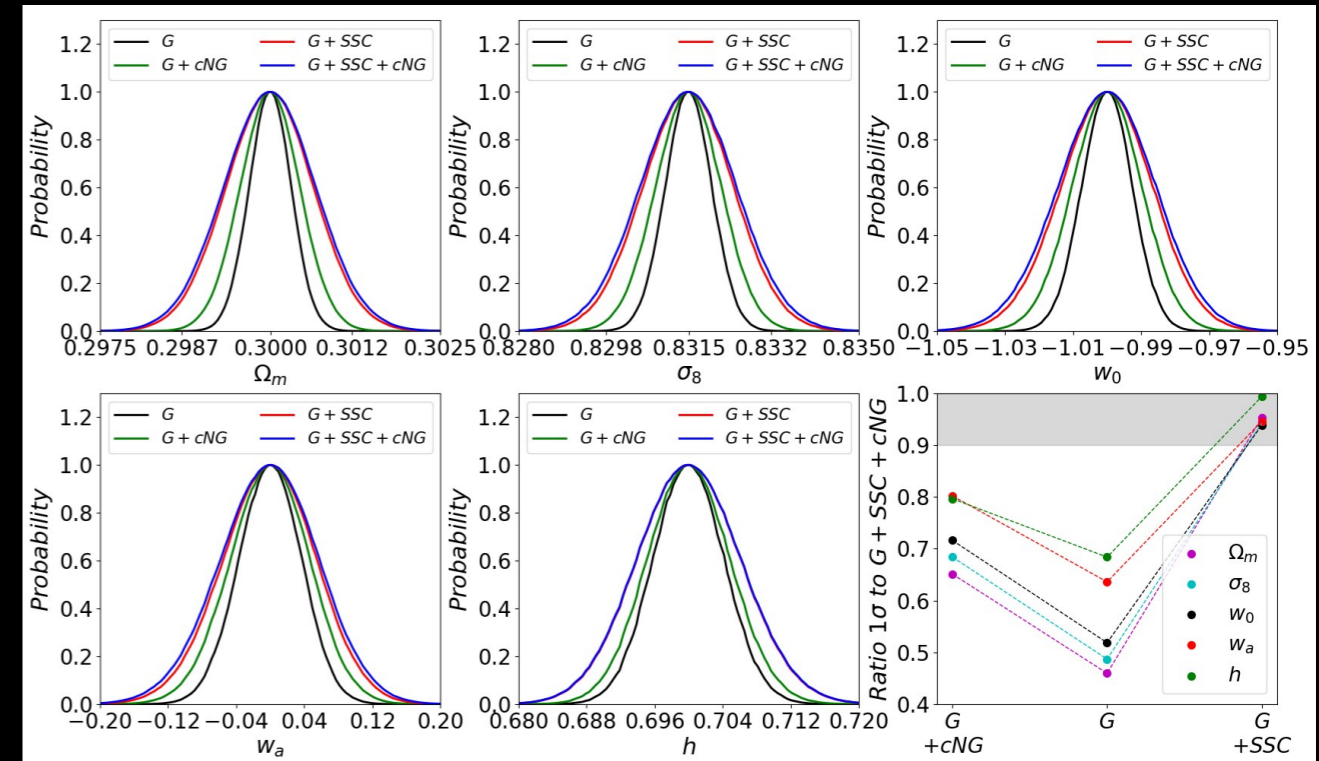
Is SSC important ?

Weak lensing : yes



Rizzato et al. 2018

Euclid : decrease of S/N by factor ~ 2

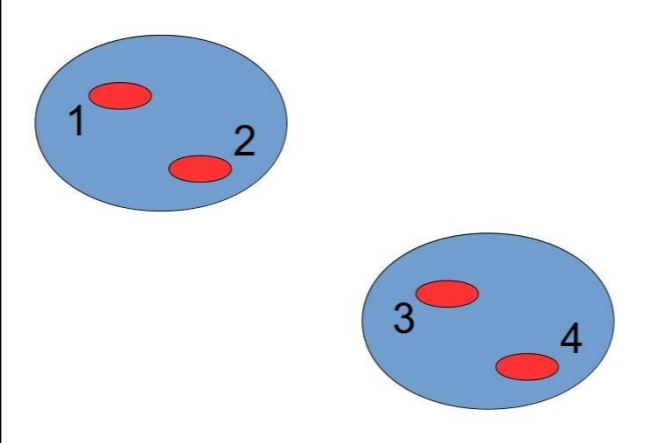
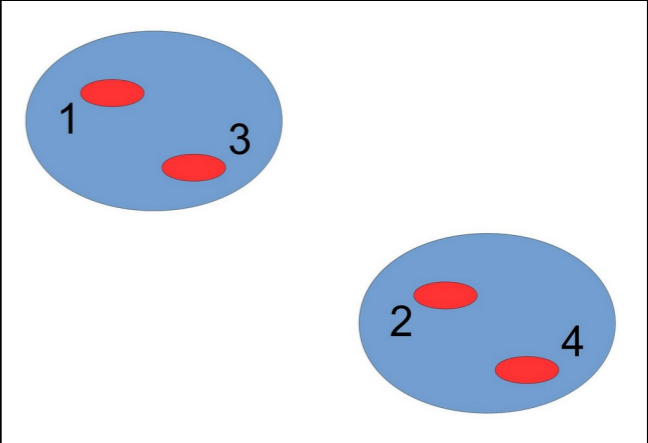

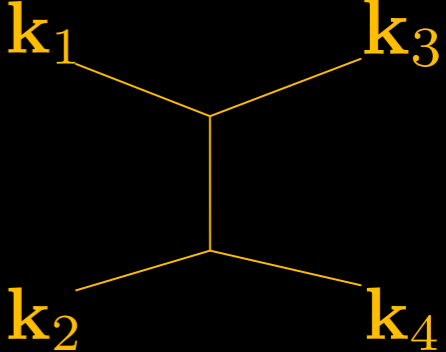


Barreira et al. 2018

Euclid : error bars increase +30% to +110%

DE, σ_8 and Ω_m particularly affected

SSC vs Braiding

| | SSC | Braiding |
|-------------------------------|--|---|
| Terms in the trispectrum | $\propto P(\mathbf{k}_1 + \mathbf{k}_2)$ | $\propto P(\mathbf{k}_1 + \mathbf{k}_3)$ |
| Diagram example (2-halo) |  |  |
| Feynman diagram (4-halo, 2x2) |  |  |
| Physical interpretation | Modulation of the two spectra by a long wavelength (super-survey) matter mode | Modulation of each mode by an intra-survey matter mode ? |