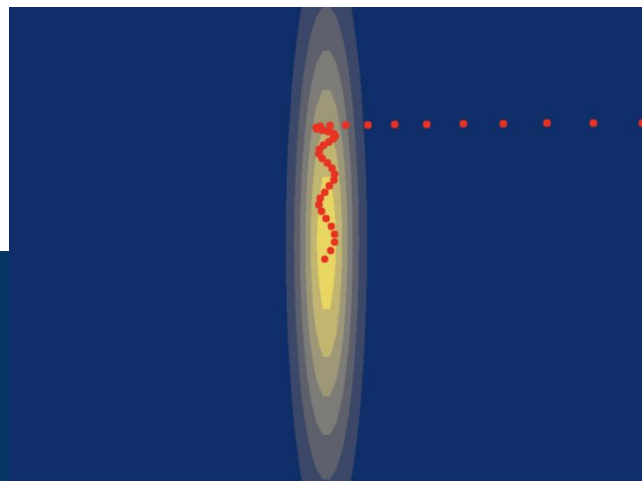
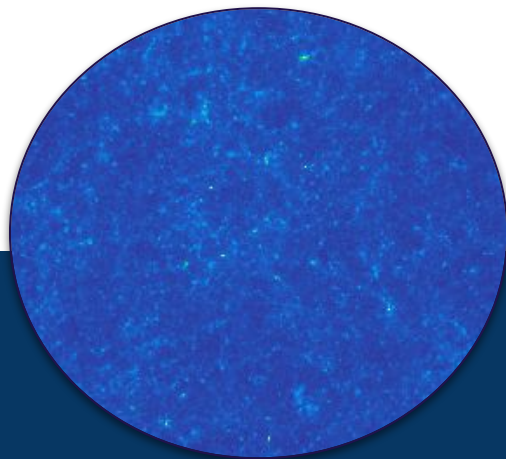
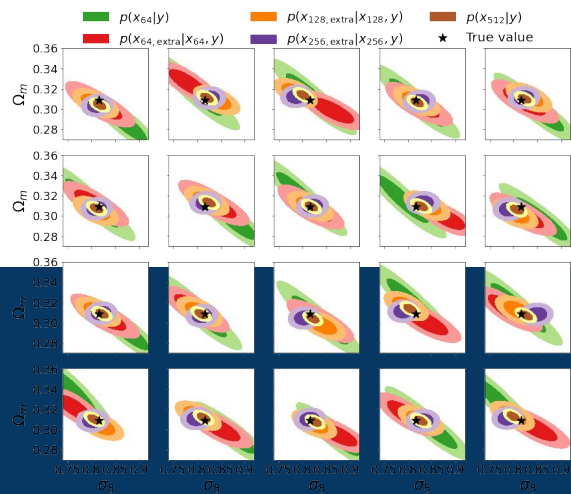


Lambda CDM cosmology model: current status, future prospects



Uros Seljak
UC Berkeley
LBNL

What is Lambda CDM model and how we test it?

- ▶ Lambda CDM model is being tested with a variety of observational probes, primarily Cosmic Microwave Background (CMB) and Large Scale Structure (LSS)
- ▶ Standard model: inflation sets the amplitude and slope of density fluctuation initial conditions (2 parameters), as well as gravity waves amplitude (tensor to scalar ratio r)
- ▶ Some other physics sets the expansion rate (Hubble parameter), matter density, cosmological constant density (dark energy) or equivalently curvature, and baryon density (set by BBN)
- ▶ In total we have about 5-7 parameters that define Lambda CDM in its simplest form
- ▶ Many extensions are possible, but so far no conclusive evidence has been found that they are needed. Examples are dark energy equation of state or other deviations of dark energy from cosmological constant, neutrino mass, non-standard inflation or alternative to it...

Neoclassical tests: use a standard ruler

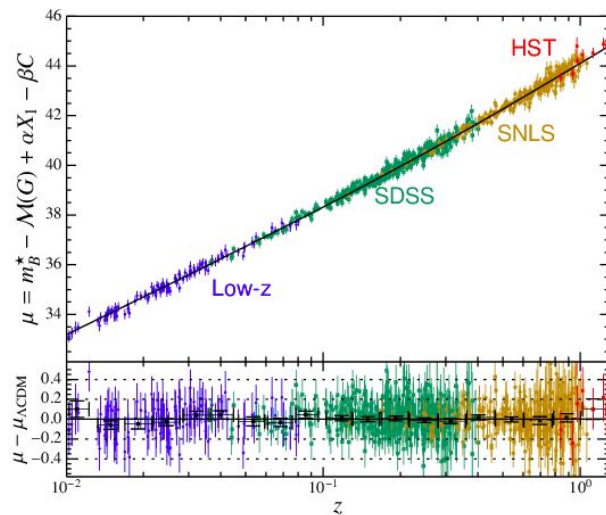
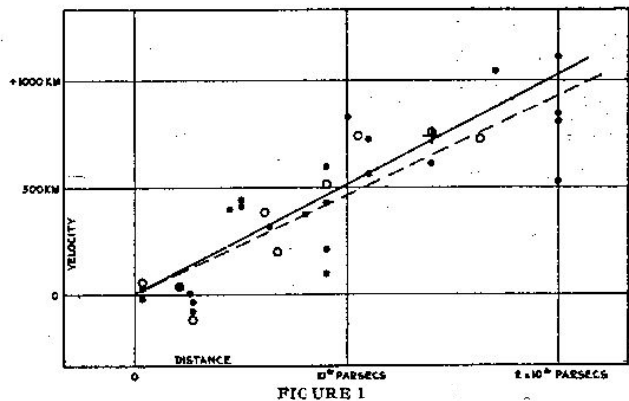
We wish to test Friedmann equation: redshift-distance relation

$$\left(\frac{\dot{a}}{a}\right)^2 = H^2 = \frac{8}{3}\pi G\bar{\rho} - Ka^{-2}$$

$$D_C(z) = \frac{c}{H_0} \int_0^z dz' \frac{H_0}{H(z')}$$

$$\bar{\rho} = \rho_m a^{-3} + \rho_{de} a^{-3(1+w)} + \rho_\gamma a^{-4} + \rho_\nu F(a)$$

Redshift-distance relation has come a long way since the days of Hubble (and people before him): distance ladder



Baryonic Acoustic Oscillations: a standard ruler

Each initial overdensity (in DM & gas) is an overpressure that launches a spherical sound wave.

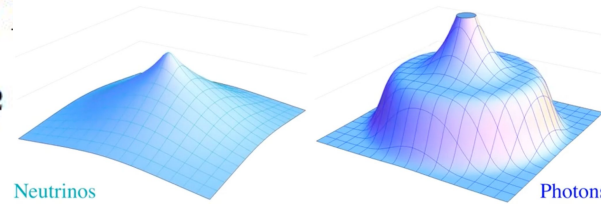
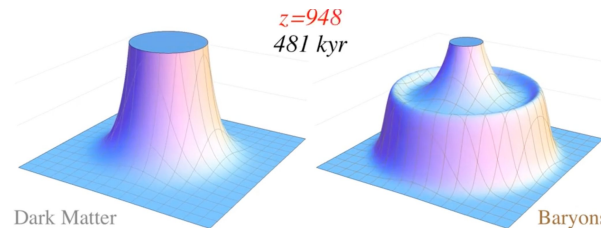
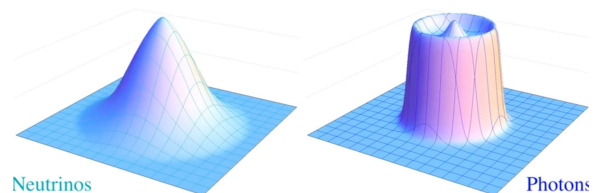
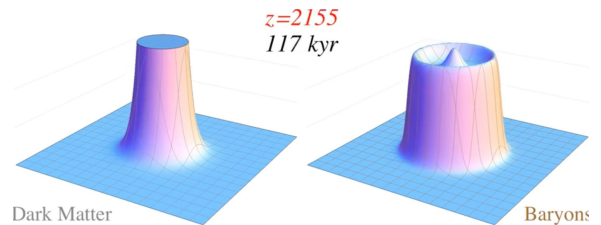
This wave travels outwards at 57% of the speed of light.

Pressure-providing photons decouple at recombination. CMB travels to us from these spheres.

Sound speed plummets. Wave stalls at a radius of 147 Mpc.

Seen in CMB as acoustic peaks

Overdensity in shell (gas) and in the original center (DM) both seed the formation of galaxies. Preferred separation of 147 Mpc.



Sound horizon at drag epoch (from Planck) : $r_d = 147.49 \pm 0.59$

$$r_d = \int_{z_d}^{\infty} \frac{c_s(z)}{H(z)} dz \quad c_s(z) = 3^{-1/2} c \left[1 + \frac{3}{4} \rho_b(z) / \rho_\gamma(z) \right]^{-1/2}$$

BAO is a Standard Ruler

The acoustic oscillation scale depends on the matter-to-radiation ratio ($\Omega_m h^2$) and the baryon-to-photon ratio ($\Omega_b h^2$)

$$r_d \approx \frac{56.067 \exp[-49.7(\omega_\nu + 0.002)^2]}{\omega_{cb}^{0.2436} \omega_b^{0.128876} [1 + (N_{\text{eff}} - 3.046)/30.60]} \text{ Mpc}$$

The CMB anisotropies measure these and fix the oscillation scale to <1%.

In a redshift survey, we can measure this along and across the line of sight:

BAO along los $\Delta v_{BAO} = \frac{r_d}{1+z} H(z)$

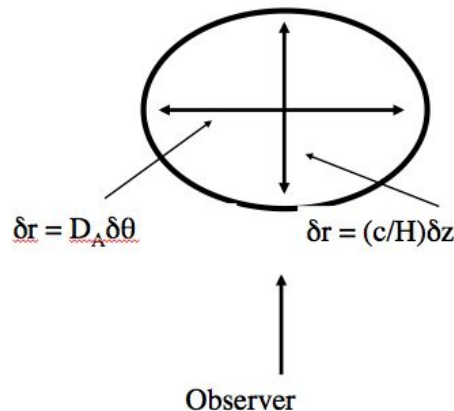
BAO tranverse $\Delta \theta_{BAO} = \frac{r_d}{1+z} \frac{1}{D_A(z)}$

Yields $H(z)$ and $D_A(z)$

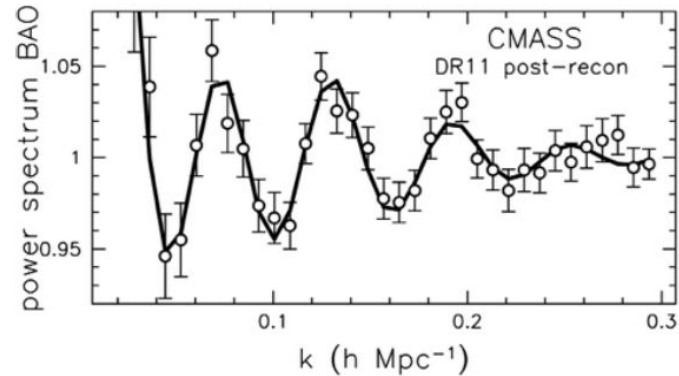
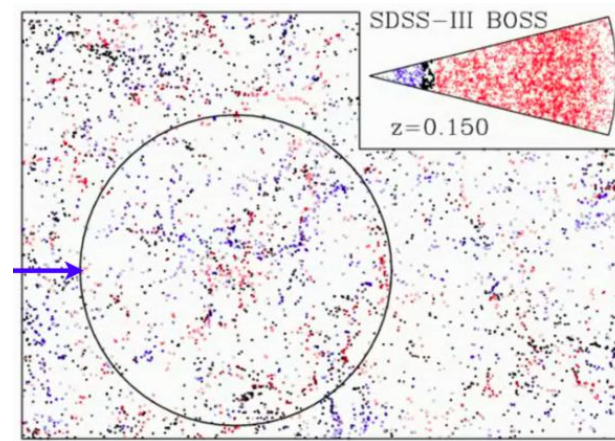
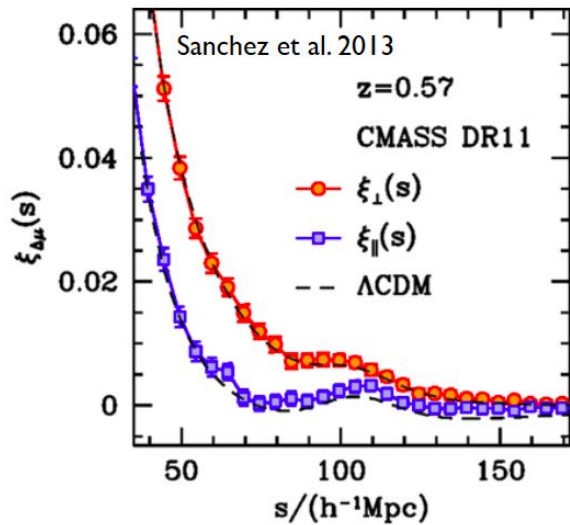
$$D_M(z) = \frac{c}{H_0} S_k \left(\frac{D_C(z)}{c/H_0} \right)$$

$$D_H(z) = c/H(z)$$

$$D_V(z) = [z D_H(z) D_M^2(z)]^{1/3}$$

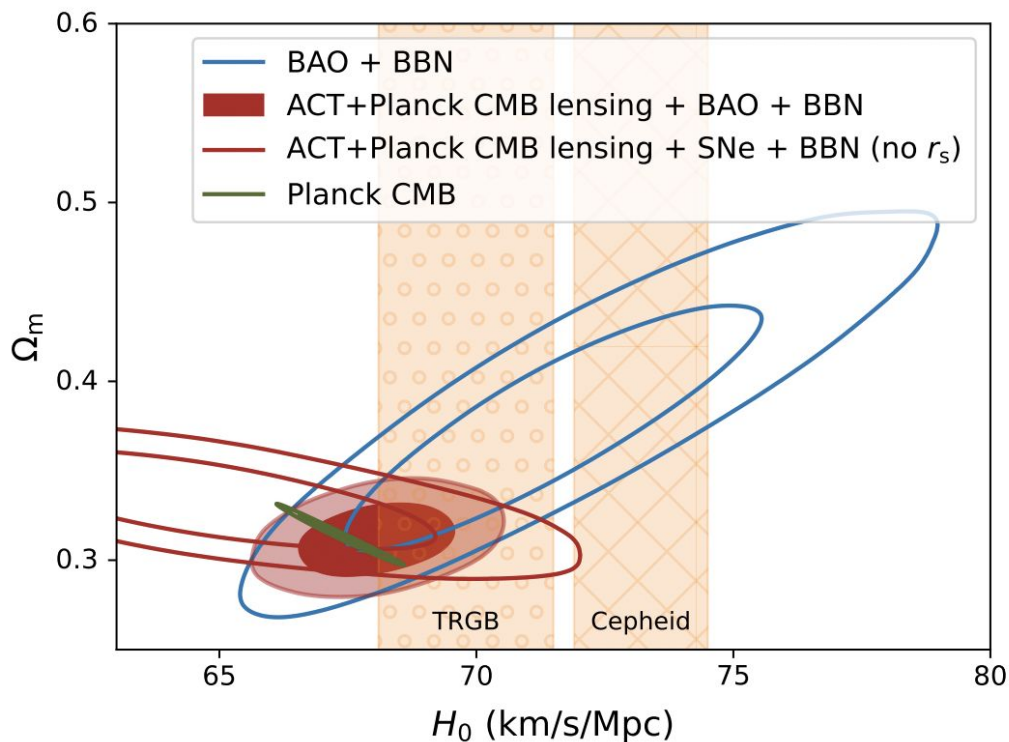


Hunting for BAO: correlation function and power spectrum



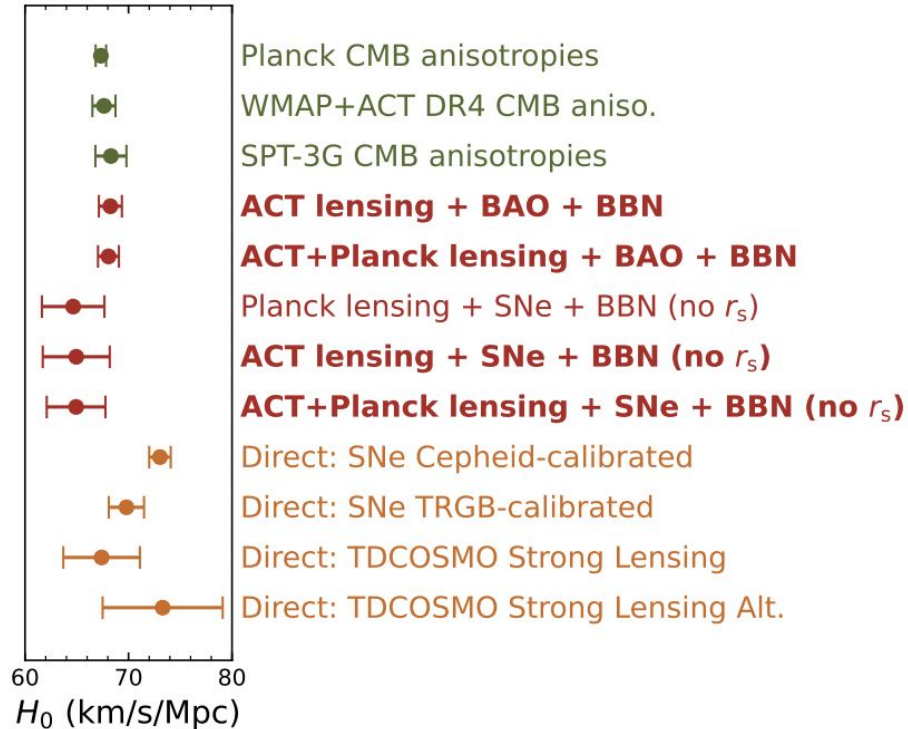
Another distance measure: use matter radiation equality rather than acoustic horizon

Recent CMB lensing
measurement
from ACT (Qu et al,
Madchaveracheril et al)



Current status of Hubble tension

Good consensus between independent probes, with the exception of SNe cepheid calibrated, which is out of line with the rest (also because of its small error)



Initial conditions: inflation

If we have a nearly homogeneous scalar field (inflaton), whose potential is flat somewhere such that its potential energy dominates then we get nearly exponential expansion

The simplest models of inflation predict

- Nearly scale invariant power law power spectrum
- Density fluctuation domination with **subdominant gravity waves**
- Nearly Gaussian
- Adiabatic
- Spatially flat

Open questions:

- What is energy scale of inflation
- How far did the inflation field travel (in Planck units)
- How did inflation begin

Initial conditions: inflation

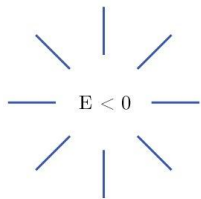
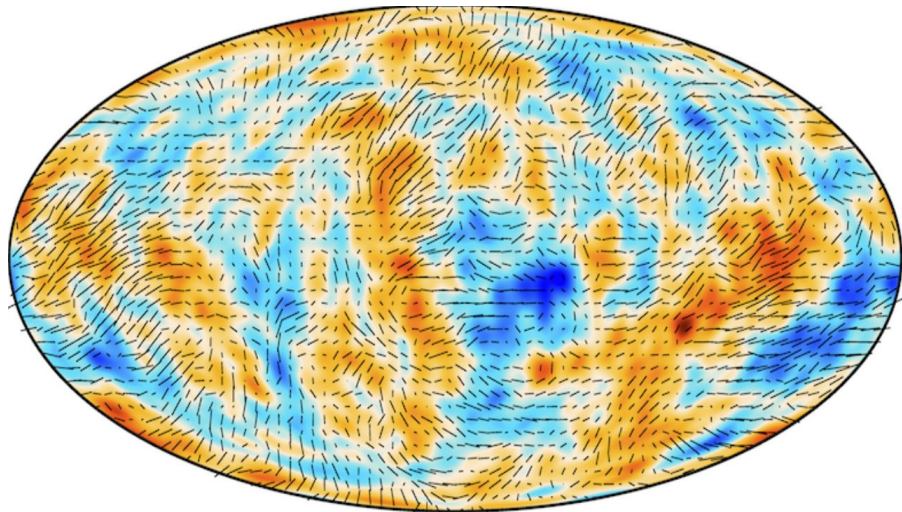
Inflation can be tested with slope of density fluctuations, with tensor to scalar ratio r , with primordial non-Gaussianity (PNG) etc.

Best probe for n_s is CMB+LSS

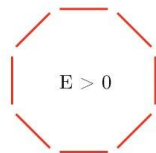
Best probe for r is B mode polarization

Best current probe for PNG is CMB, future will be LSS.

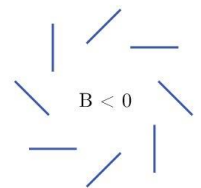
B modes of polarization



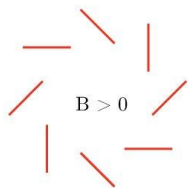
$E < 0$



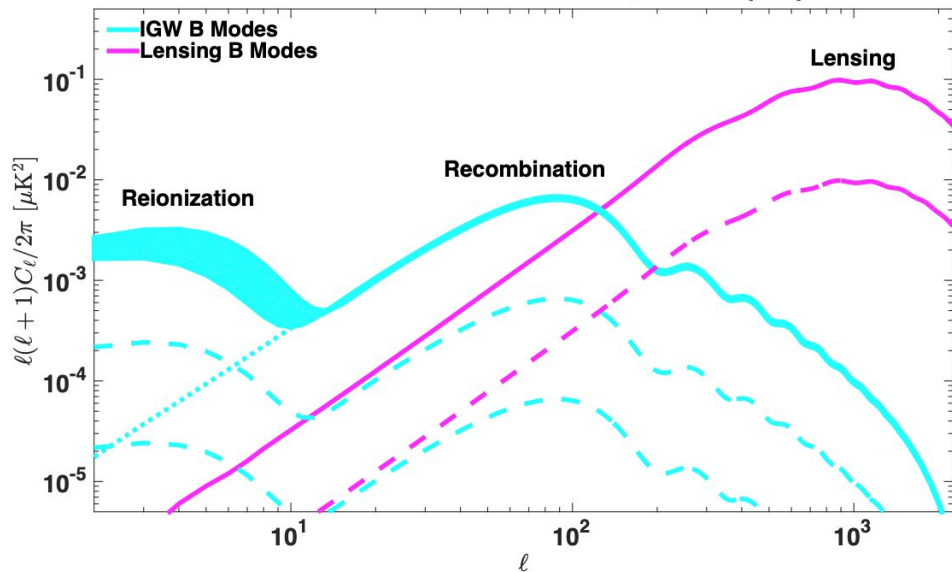
$E > 0$



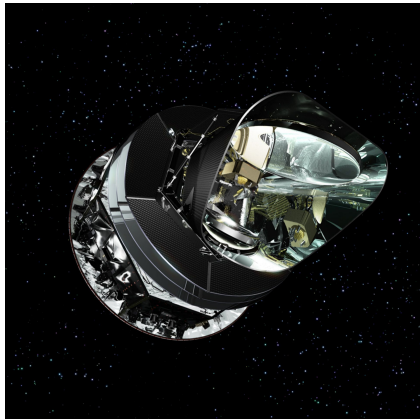
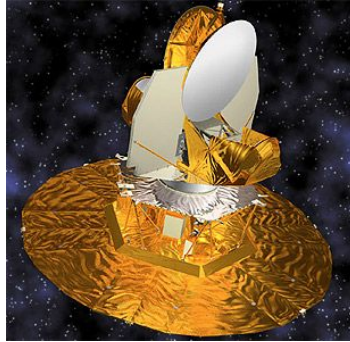
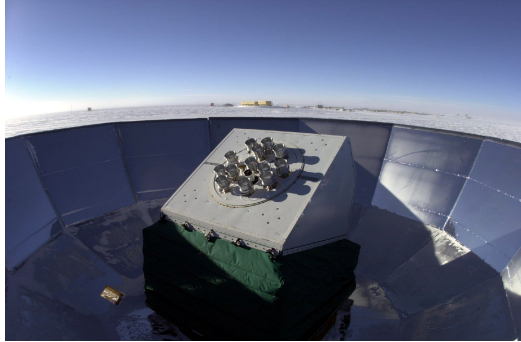
$B < 0$



$B > 0$



Experiments

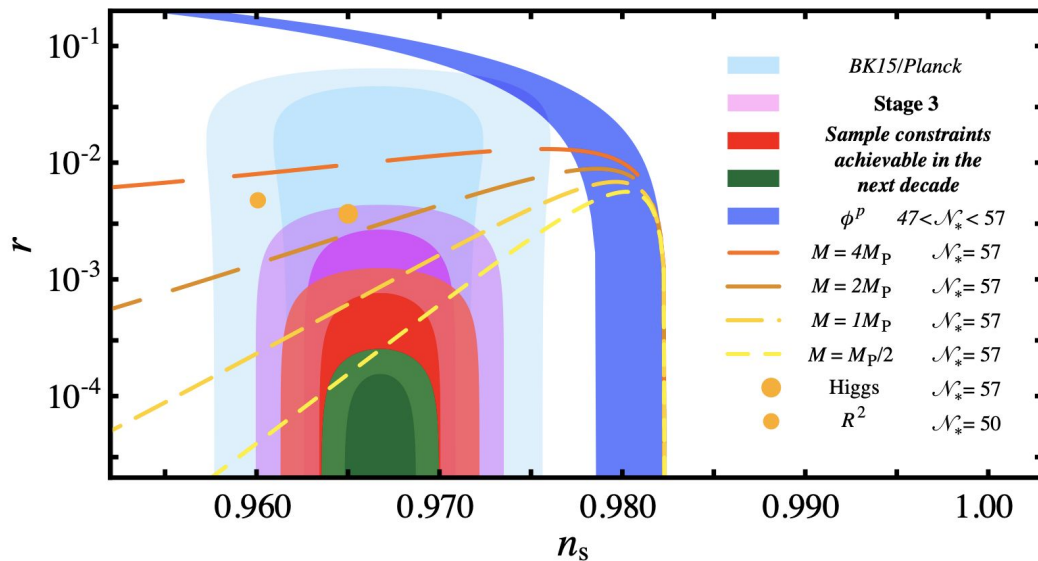
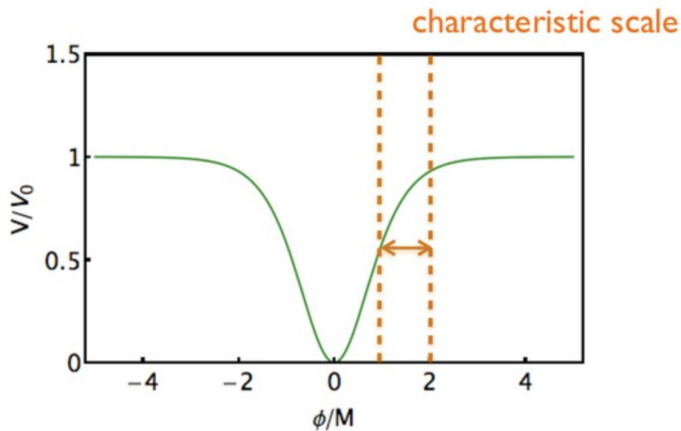


Current and future constraints on single field models

Current limit: $r < 0.06$

Power law models: almost excluded

Plateau and hilltop models: characteristic scale



<https://arxiv.org/abs/1903.04700>

Figure 1: Predictions for the tensor-to-scalar ratio r and spectral index n_s for some representative single-field inflationary models in which $n_s - 1 \propto -1/\mathcal{N}_*$. This class includes monomial models with $V(\phi) \propto \phi^p$ (dark blue), the Starobinsky (R^2) model, and Higgs inflation (orange filled circles). The dashed lines show the predictions of models in this class as function of the scale in the potential. All models with Planckian scale can be detected or excluded in the next decade.

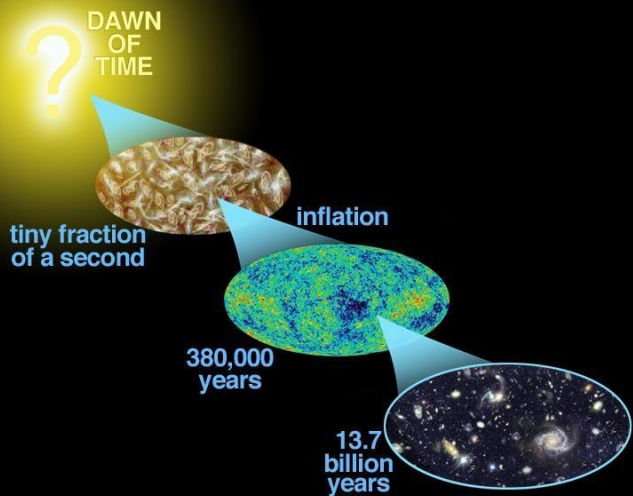
The other tension: amplitude of LSS vs CMB

There have been many claims that LSS amplitude is lower than CMB (Planck etc.)

LSS amplitude can be measured primarily in 2 ways:

- 1) weak lensing (including WL cross-correlation with LSS, and CMB lensing)
- 2) Galaxy clustering via redshift space distortions

Growth of structure by gravity



◆ Perturbations can be measured at different epochs:

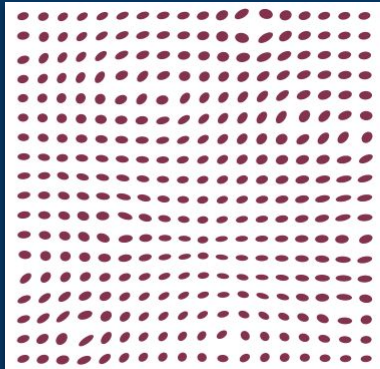
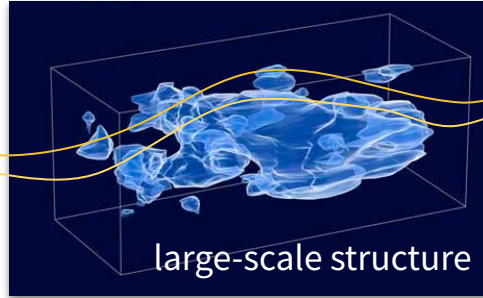
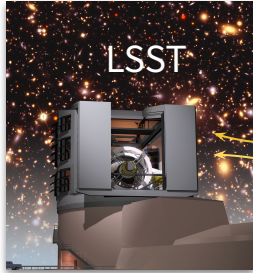
1. CMB $z=1000$
2. Ly-alpha forest $z=2-4$
3. Weak lensing $z=0.3-2$
4. Galaxy clustering, clusters $z=0-2$

Sensitive to matter components, initial conditions...

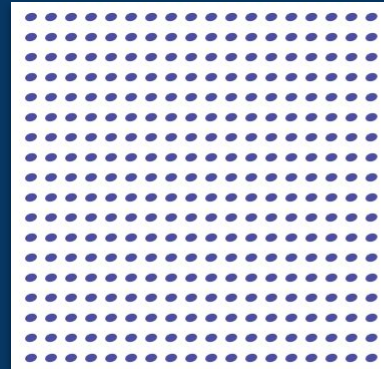
$$\ddot{\delta} + 2H\dot{\delta} = 4\pi G\bar{\rho}\delta \rightarrow \delta(t)$$



Weak Lensing of Galaxies

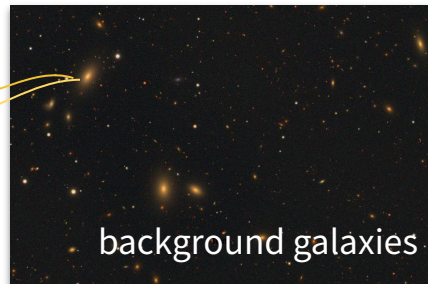
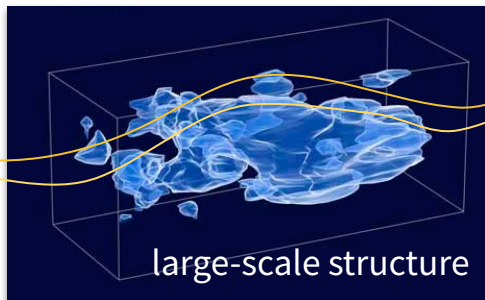
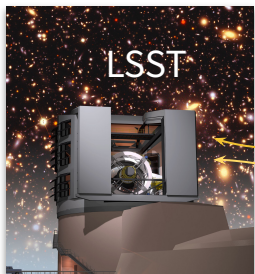


lensing

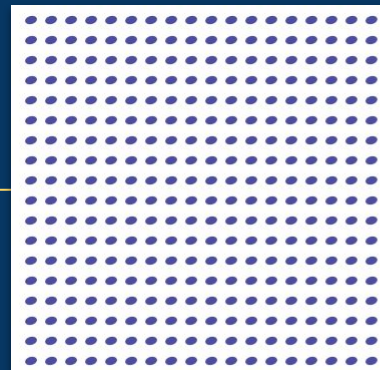
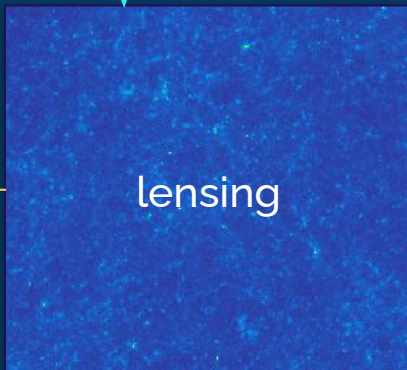
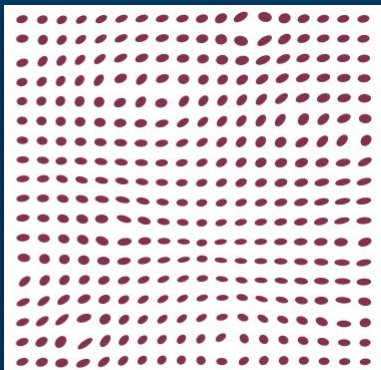




Weak Lensing of Galaxies

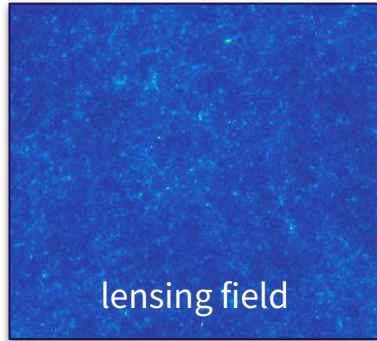


projection





Weak Cosmic Shear



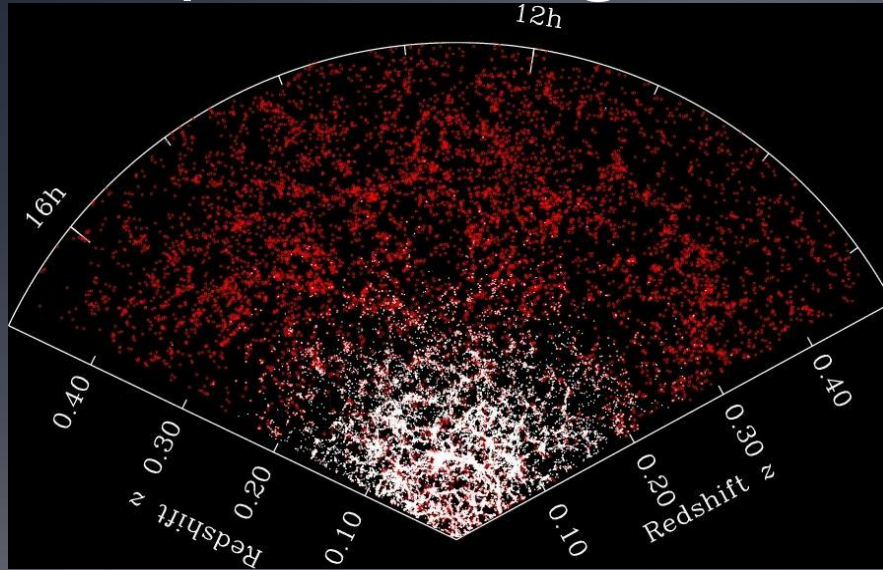
Lensing is sensitive to

- the total matter parameter (Ω_m)
- the amplitude of matter fluctuations (σ_8)
- sum of neutrino masses (M_ν)
- time-varying dark energy (w)

What are advantages and disadvantages of WL?

- + Sensitive to total matter distribution, relatively easy to relate to models
- Measures it in projection ($d=2$ for a single plane, some tomography possible with multiple source redshifts)
- Systematics: Intrinsic alignments (of GI type), baryonic effects, photometric redshift distribution, shear calibration etc
- + WL of CMB does not suffer from most of these systematics!

Galaxy clustering in redshift space



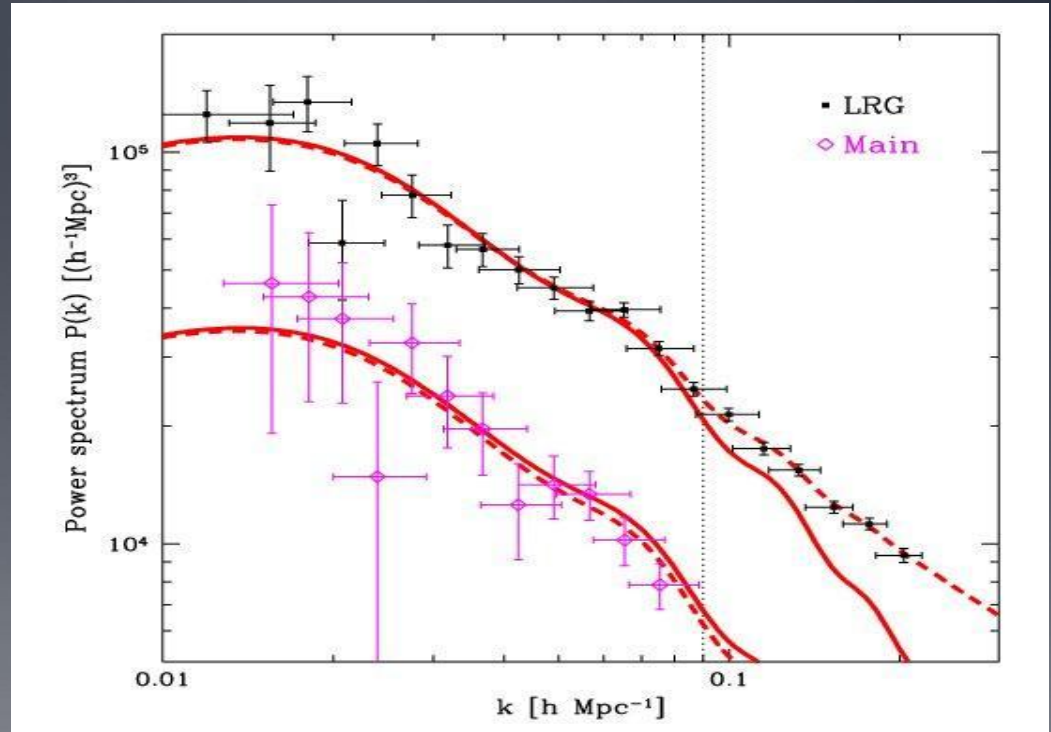
SDSS

Measures 3-d distribution, has many more modes than projected quantities like shear from weak lensing

Easy to measure: effects of order unity, not 1%

Galaxy power spectrum: biasing

- Galaxy clustering traces dark matter clustering
 - Amplitude depends on galaxy type: galaxy bias b
- $$P_{gg}(k) = b^2(k) P_{mm}(k)$$
- To determine bias we need additional information
 - Galaxy bias can be scale dependent: $b(k)$
 - Once we know bias we know how dark matter clustering grows in time

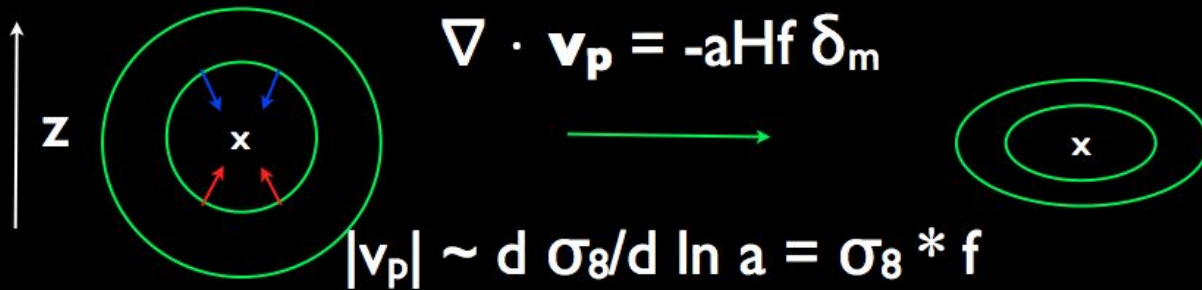


Tegmark et al. (2006)

Redshift space distortions

redshift $cz = aHr + v_p$

real to redshift space separations



isotropic

squashed along line of sight

$$f = d \ln \sigma_8 / d \ln a$$

Linear and nonlinear effects

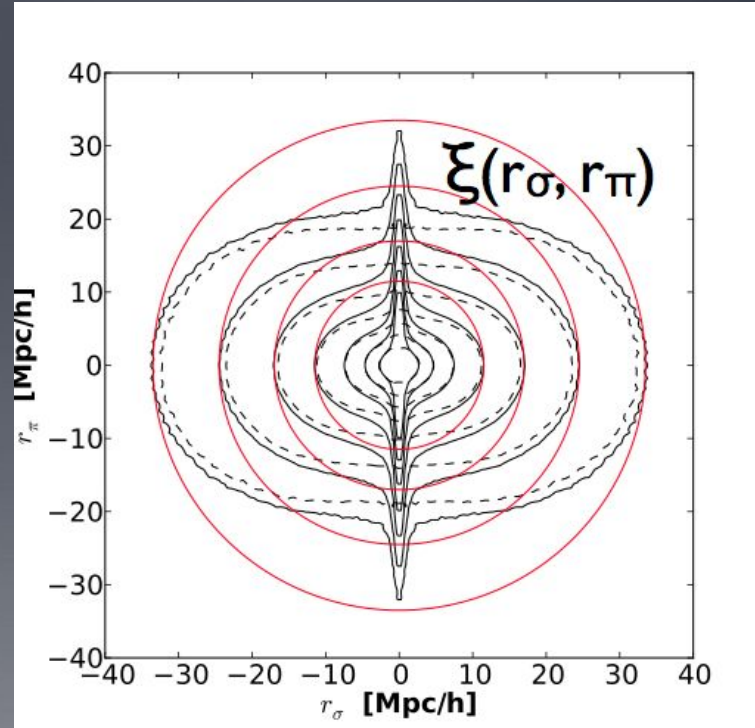
On very large scales linear RSD distortions: Kaiser formula

$$\delta_g = (b + f\mu^2)\delta = b(1 + \beta\mu^2)\delta$$

$\mu = \vec{k} \cdot \vec{n} / k$
From angular dependence we can determine velocity power proportional to $f\sigma_8$

$$\beta = f/b$$

On small scales: virialized velocities within halos lead to FoG, extending radially 10 times farther than transverse

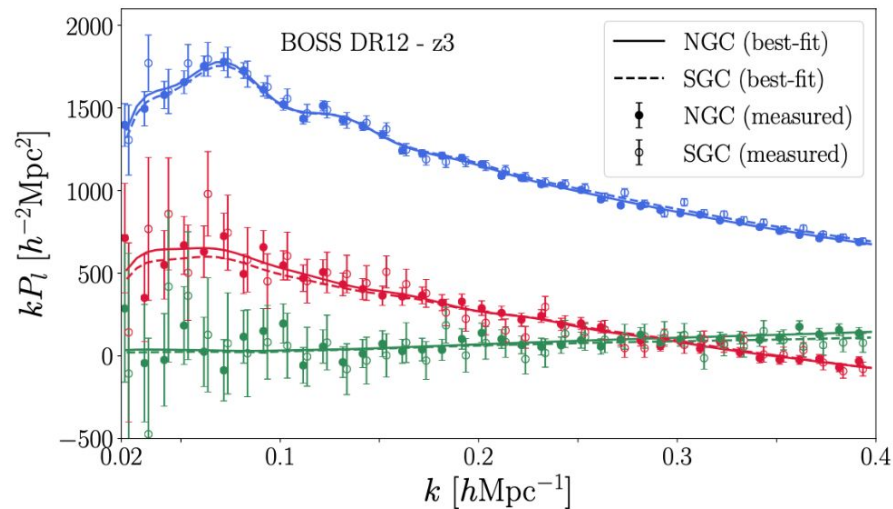
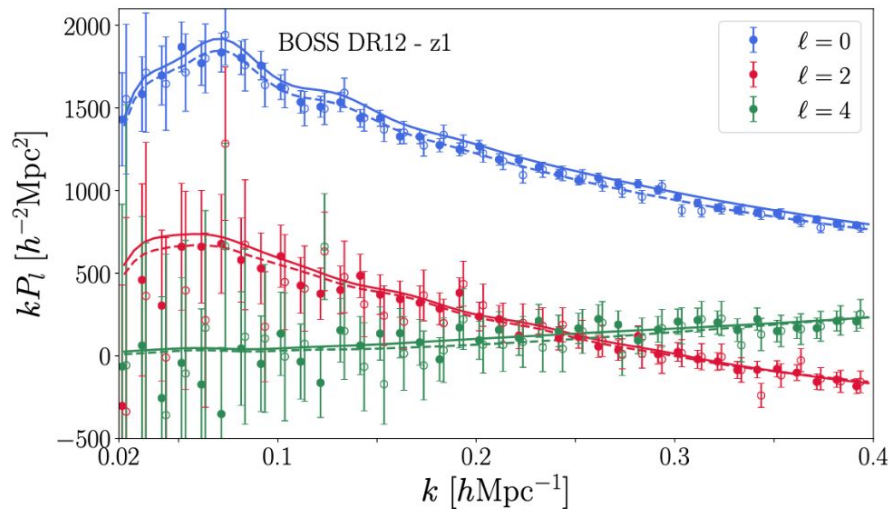


What are advantages and disadvantages of RSD?

- + Sensitive to velocities, which trace total matter
- + 3d data and high signal means a lot of information
- Systematics: modeling of nonlinearity in bias and in velocities requires a lot of free parameters, and their priors
- Significant disagreement between different analyses on the same data, and same analyses truncated at a different minimal scale. Unclear how to choose this scale cut.

HaloPT SDSS analysis: SGC, NGC: CMASS, LOWZ

Yu, Seljak, Li, Singh 2022

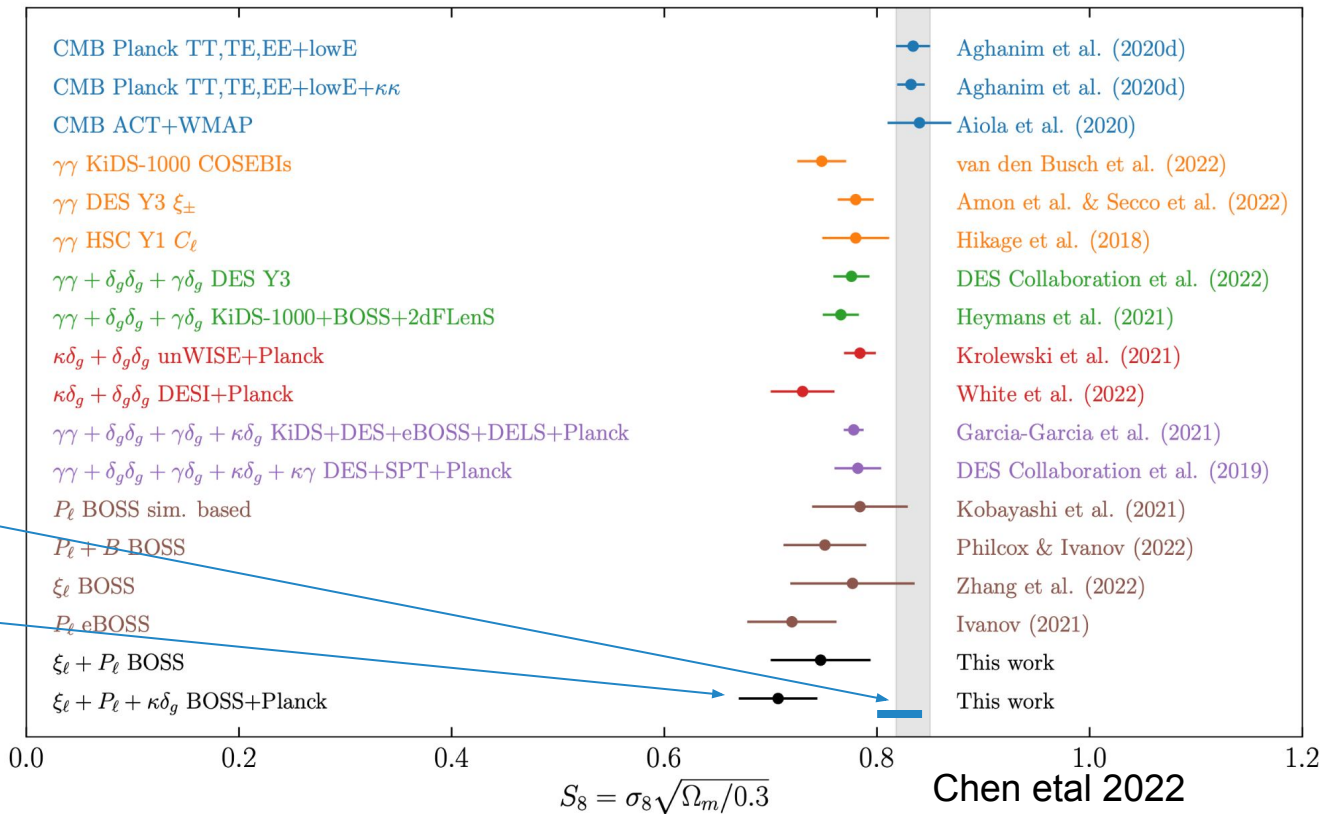


10 or 11 parameter fit (3 halo mass components)

$$[f(z_{\text{eff}}), \sigma_8(z_{\text{eff}}), b_{1,C_A}, b_{1,S_A}, b_{1,S_B}, f_s, f_{s_B}, \langle N_{>1,s} \rangle, \sigma_c, \sigma_{s_A}, f_{s_B s_B}^{1h}]$$

Sigma8 tension?

- ▶ WL discrepancy between Chen etal and Singh etal almost certainly due to the modeling
- ▶ Chen combined RSD + WL is strongly below Planck
- ▶ Corresponding analysis of Halo PT (Yu etal) is consistent with Planck: same data as Chen etal
- ▶ **Modeling crisis?**
- ▶ **Too early to claim LSS is low compared to Planck?**

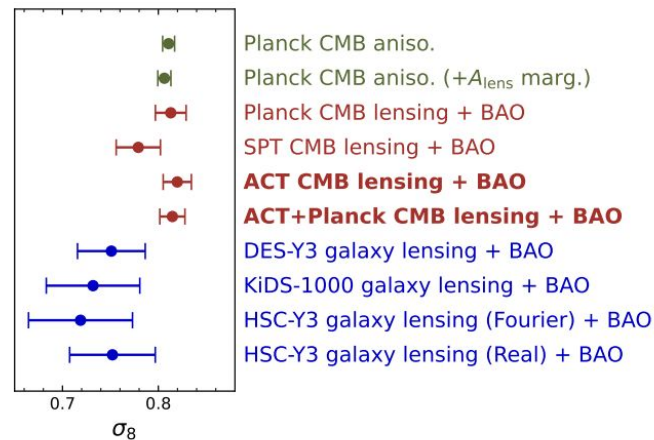
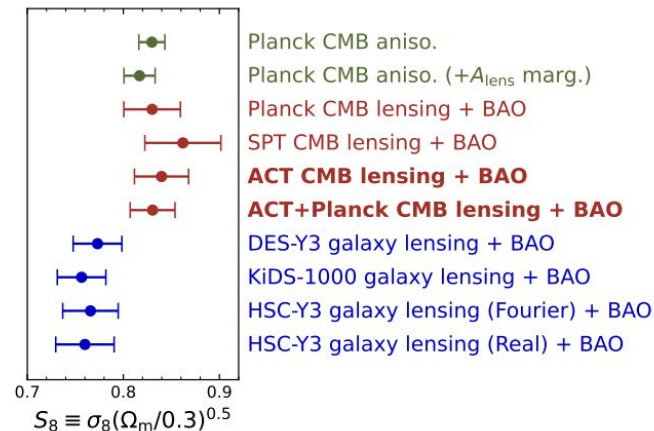


Lensing of CMB to the rescue

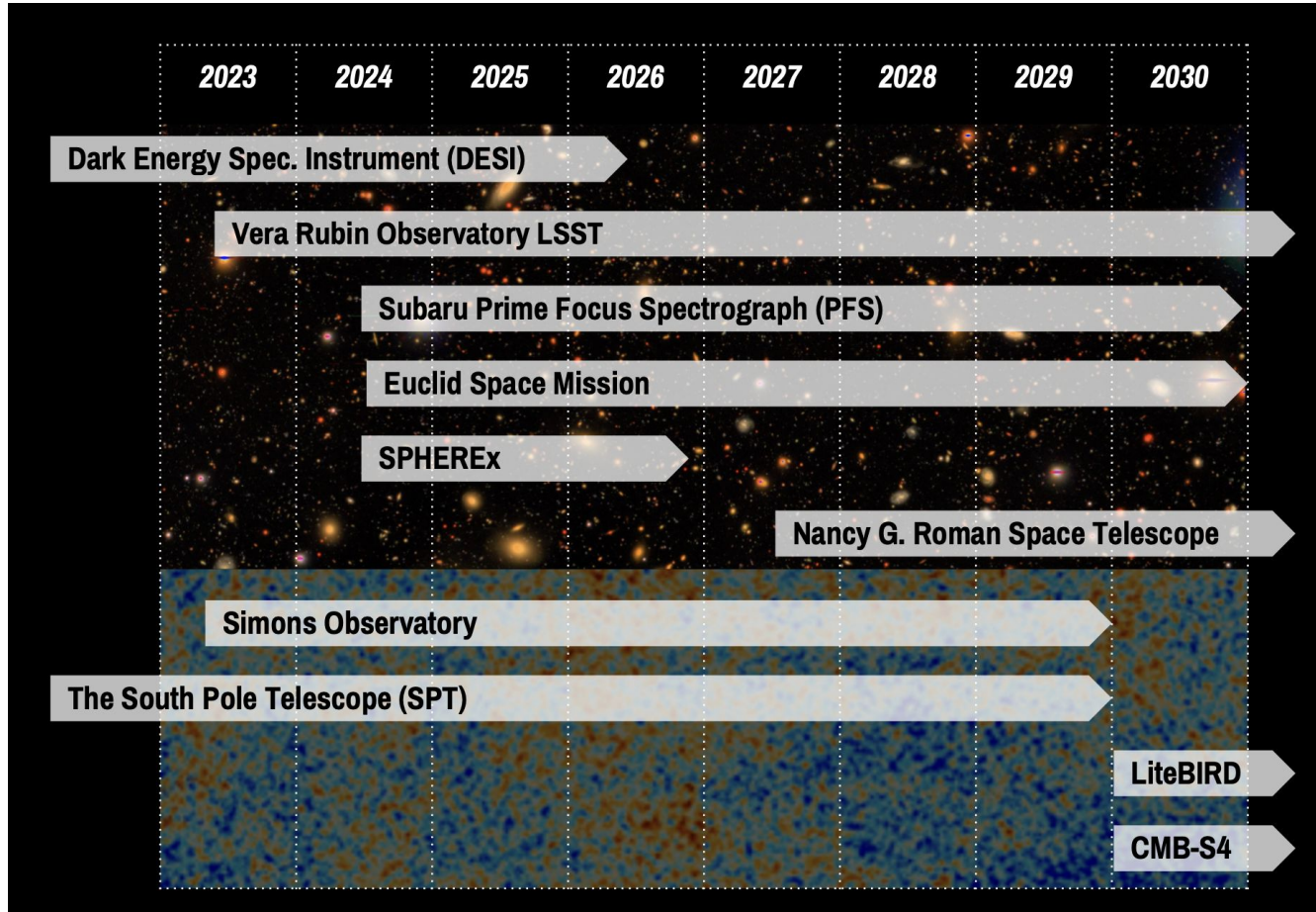
Lensing of CMB has very few systematics, and Planck lensing is consistent with Planck CMB

Recent ACT results are striking in their quality and show no tension with Planck

Similar results with CMB+Wise reanalysis (Krolewski et al)



The future is bright!



New surveys

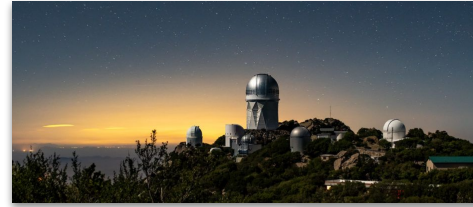
Vera Rubin Observatory (LSST)



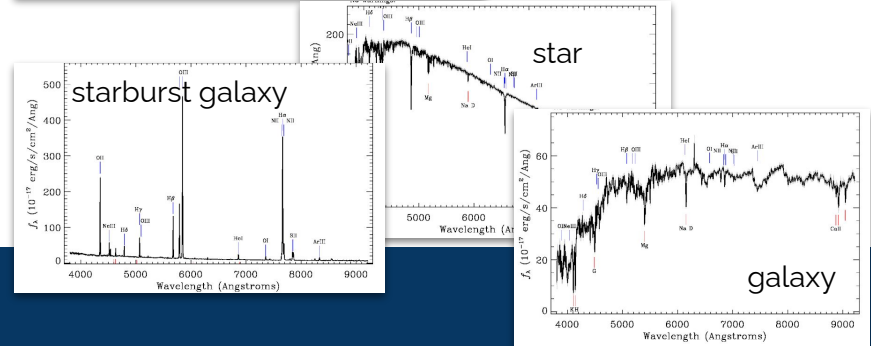
20 billion galaxies
17 billion stars
20 terabyte data/day



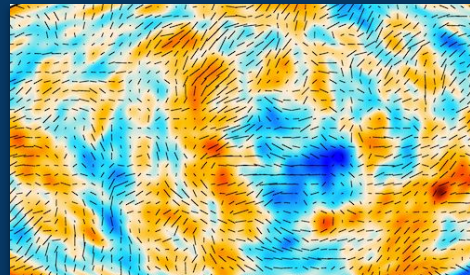
Dark Energy Spectroscopic Instrument (DESI)



35 million galaxies
10 million stars



CMB-S4



50 PB total database

Cosmic Microwave Background

New Methods: Machine Learning and more

LSS is very nonlinear, and power spectrum analyses miss a lot of information

Sometimes we can do a full explicit likelihood analysis, e.g. CMB lensing and RSD (expensive to sample)

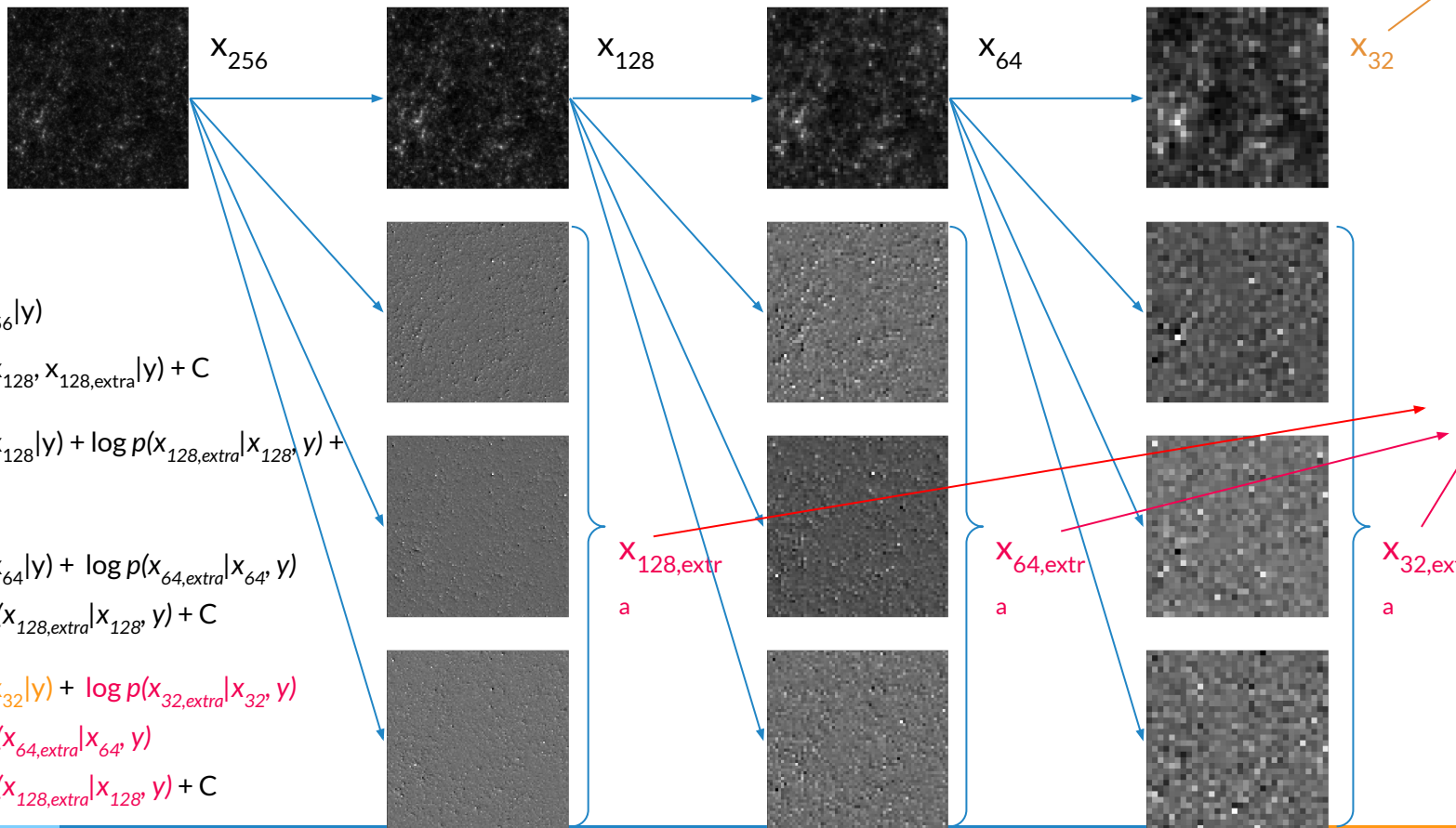
Otherwise we can use implicit likelihood analysis using ML, eg WL

These methods show promise of a large information gain relative to power spectrum

Multiscale flow: a new hierarchical flow (Dai & Seljak, 2023)

▷ Consider a cosmological field with 256^2 resolution:

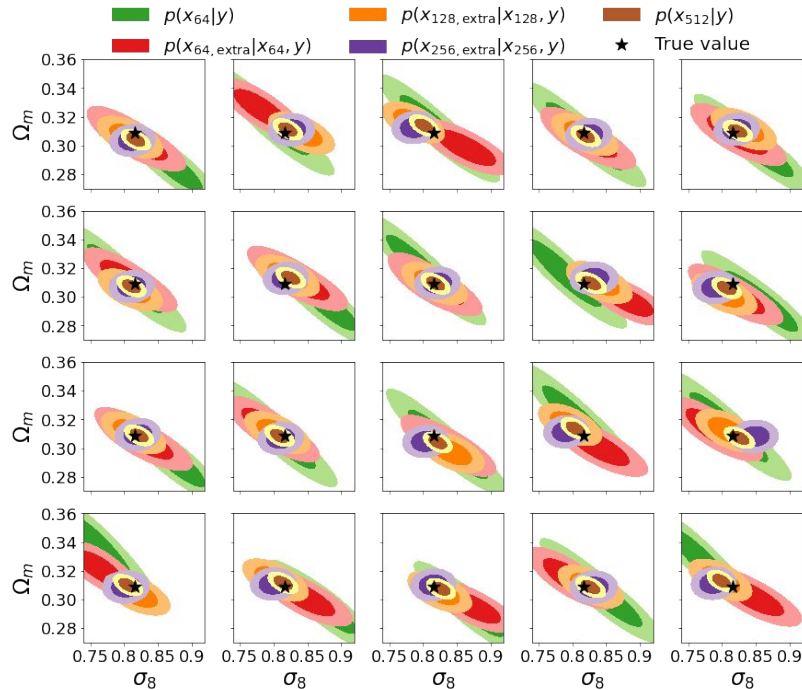
TRENF



$$\begin{aligned}
 & \log p(x_{256}|y) \\
 &= \log p(x_{128}, x_{128,extra}|y) + C \\
 &= \log p(x_{128}|y) + \log p(x_{128,extra}|x_{128}, y) + C \\
 &= \log p(x_{64}|y) + \log p(x_{64,extra}|x_{64}, y) \\
 & \quad + \log p(x_{128,extra}|x_{128}, y) + C \\
 &= \log p(x_{32}|y) + \log p(x_{32,extra}|x_{32}, y) \\
 & \quad + \log p(x_{64,extra}|x_{64}, y) \\
 & \quad + \log p(x_{128,extra}|x_{128}, y) + C
 \end{aligned}$$

Reliable Uncertainty Quantification with NF posteriors

- Consistent posteriors from different scales



Posterior is proportional to likelihood $p(x|y)$ times prior $p(y)$ (which is assumed to be flat here)

We worked hard to train NF to give reliable uncertainty quantification in terms of 68% and 95% c.i. coverage probability

Method	$n_g = 10\text{arcmin}^{-2}$	$n_g = 20\text{arcmin}^{-2}$	$n_g = 50\text{arcmin}^{-2}$	$n_g = 100\text{arcmin}^{-2}$
Multiscale Flow $p(x_{512} y)$	72.8%, 96.8%	74.4%, 95.2%	73.6%, 97.6%	66.4%, 97.6%
Multiscale Flow $p(x_{256} y)$	70.4%, 96.0%	76.8%, 95.2%	74.4%, 97.6%	68.8%, 96.8%
Multiscale Flow $p(x_{128} y)$	76.0%, 96.0%	74.4%, 97.6%	76.0%, 97.6%	73.6%, 96.8%
Multiscale Flow $p(x_{64} y)$	80.8%, 95.2%	70.4%, 94.4%	72.0%, 95.2%	74.4%, 95.2%

Robustness — unknown systematic effects on WL maps: baryonic effects

Likelihood, ie density estimation, can be used for anomaly detection (generalization of reduced χ^2 test, ie PPD): if the data are from Out of Distribution then the density will be low

This can be done as a function of scale

Scale dependence of posteriors is an independent test

Both are comparable in ROC and powerful!

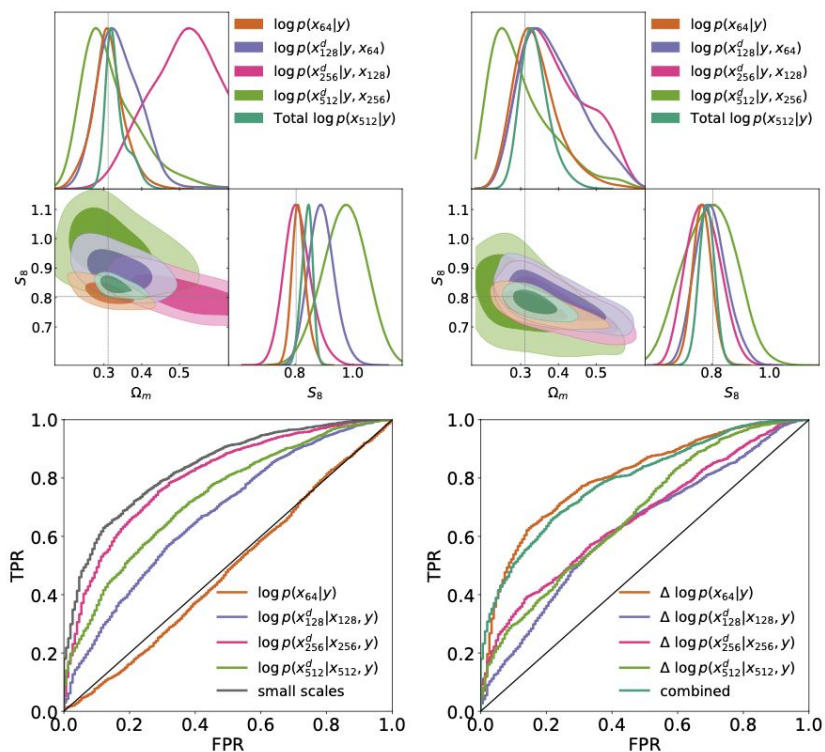


Fig. 5. Top panel: scale-dependent posterior analysis of a baryon-corrected convergence map using Multiscale Flow trained on dark-matter-only maps (left), and Multiscale Flow trained on BCM maps (right). Bottom panel: ROC curve of identifying distribution shift with $\log p$ (left) and $\Delta \log p$ (right). The "small scales" in the lower left panel represent combining the three small scale terms. In these experiments, we consider 30arcmin^{-2} galaxy shape noise.

Two independent ways to identify unknown unknowns!

Figure of Merit: inverse error area on matter density and amplitude

Method	$n_g = 10\text{arcmin}^{-2}$	$n_g = 30\text{arcmin}^{-2}$	$n_g = 100\text{arcmin}^{-2}$
Multiscale Flow $p(x_{512} y)$	95	246	733
Multiscale Flow $p(x_{256} y)$	91	229	644
Multiscale Flow $p(x_{128} y)$	74	180	435
Multiscale Flow $p(x_{64} y)$	52	123	272
power spectrum	30 (30)	52 (51)	81 (79)
peak count	(30)	(89)	(162)
CNN	(44)	(121)	(292)
scattering transform $s_0 + s_1 + s_2$	(\lesssim 50)	(\lesssim 140)	(\lesssim 329)

Factor of 5 better than power spectrum

Factor of 3 better than other recent methods (CNN, scattering transform)

Equivalent to 3-5 larger survey area of the sky!

Table 2. Similar to Table 1, with baryonic effects.

Method		$n_g = 10\text{arcmin}^{-2}$	$n_g = 20\text{arcmin}^{-2}$	$n_g = 50\text{arcmin}^{-2}$	$n_g = 100\text{arcmin}^{-2}$
Fix baryon parameters at fiducial values	Multiscale Flow $p(x_{512} y)$	166	310	617	1072
	Multiscale Flow $p(x_{256} y)$	164	297	558	947
	Multiscale Flow $p(x_{128} y)$	124	214	415	704
	Multiscale Flow $p(x_{64} y)$	81	136	247	387
	power spectrum	41(41)	61 (58)	95(87)	127(111)
	CNN	-	(\sim 93)	(\sim 146)	(\sim 194)
Marginalize over baryon parameters	Multiscale Flow $p(x_{512} y)$	149	220	362	521
	Multiscale Flow $p(x_{256} y)$	147	213	341	494
	Multiscale Flow $p(x_{128} y)$	112	166	269	398
	Multiscale Flow $p(x_{64} y)$	75	113	183	259
	power spectrum	34(33)	48(48)	68(65)	84 (78)
	CNN	-	(\sim 77)	(\sim 109)	(\sim 136)

Next Step: include systematics (photoz calibration, shear calibration, intrinsic alignments...)

Next step: apply to the real data.

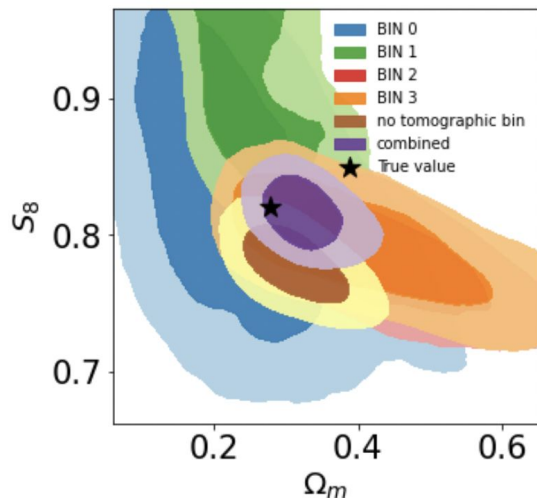
First HSC on Subaru Later Rubin etc.

1. When fixing the baryon parameters at fiducial values, the FoM of CNN are estimated from Lu et al. (51). Lu et al. (51) estimated the 1 σ area of a 1500h^{-2} survey and reports their results with the error ratio from direct comparison with our experiments.

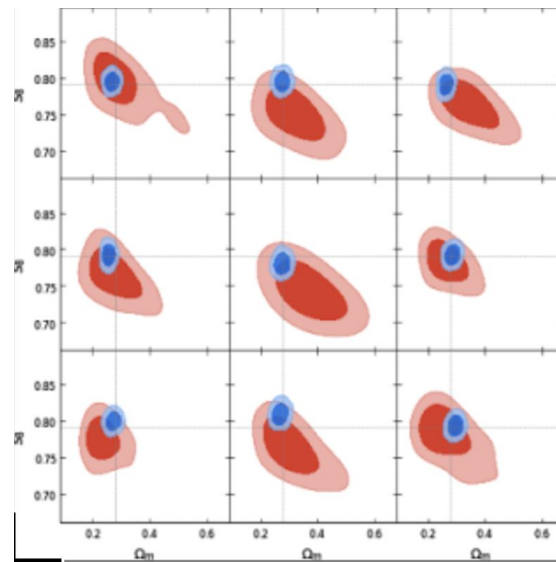
Tomographic bins: conditional likelihoods

- $\log p(\mathbf{x}_{512}) = \log p(\mathbf{x}_{32}) + \log p(\mathbf{x}_{32}^d | \mathbf{x}_{32}) + \log p(\mathbf{x}_{64}^d | \mathbf{x}_{64}) + \log p(\mathbf{x}_{128}^d | \mathbf{x}_{128}) + \log p(\mathbf{x}_{256}^d | \mathbf{x}_{256})$
- $\log p(\mathbf{x}) = \log p(\mathbf{x}_{\text{BIN1}}) + \log p(\mathbf{x}_{\text{BIN2}} | \mathbf{x}_{\text{BIN1}}) + \log p(\mathbf{x}_{\text{BIN3}} | \mathbf{x}_{\text{BIN2}}) + \log p(\mathbf{x}_{\text{BIN4}} | \mathbf{x}_{\text{BIN3}})$

$p(\mathbf{x}_{128} | \mathbf{x}_{64}, y)$ scale ~ 7.5 - 15 arcmin



Up to 10x higher
Figure of Merit for
realistic HSC data!



$$p(\text{bin1}, \text{bin2}, \text{bin3}, \text{bin4}) = p(\text{bin1}) * p(\text{bin2} | \text{bin1}) * p(\text{bin3} | \text{bin1}, \text{bin2}) * p(\text{bin4} | \text{bin1}, \text{bin2}, \text{bin3})$$

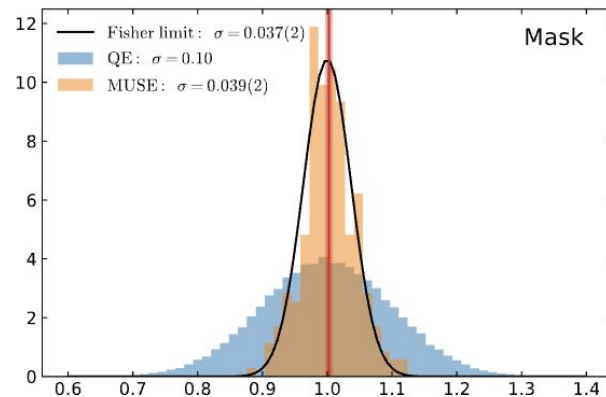
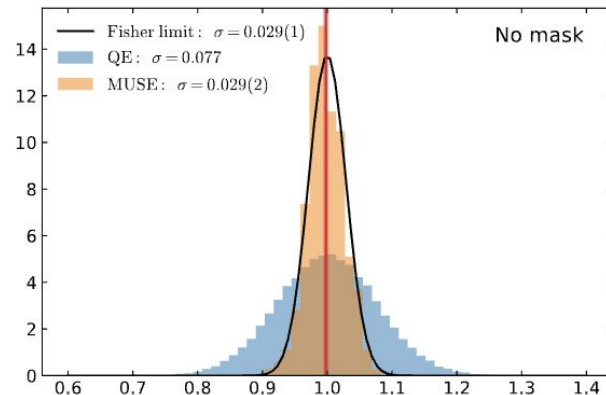
Explicit Likelihood Lensing of CMB: the ultimate LSS probe?

CMB lensing is almost systematics free

Current state of the art (e.g. ACT analysis):
Quadratic Estimators (QE)

With low noise (eg SPT, CMB S4) explicit likelihood analysis (Hirata & Seljak 2003, Millea & Seljak 2022) gets much better results

High dimensional sampling at field level is slow,
but new methods of sampling such as
MicroCanonical Langevin Monte Carlo make it
feasible



Summary

- Lambda CDM is the current standard model of our universe and currently explains all the data
- Recent analyses do not support claims of tensions in the data (e.g. Hubble and amplitude tensions)
- Next generation surveys are coming online soon (DESI, Euclid, Rubin, SO etc.)
- Next generation analyses are being developed and in some cases will be equivalent to an order of magnitude increase in data volume