



THE DARK ENERGY SURVEY

DES Year 3: Weak lensing and clustering cosmology

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Kavli Fellow, Stanford University

  @astroalexamon





THE DARK ENERGY SURVEY

DES Y3:

Data systematics:

Astrophysical systematics :

The Big picture:

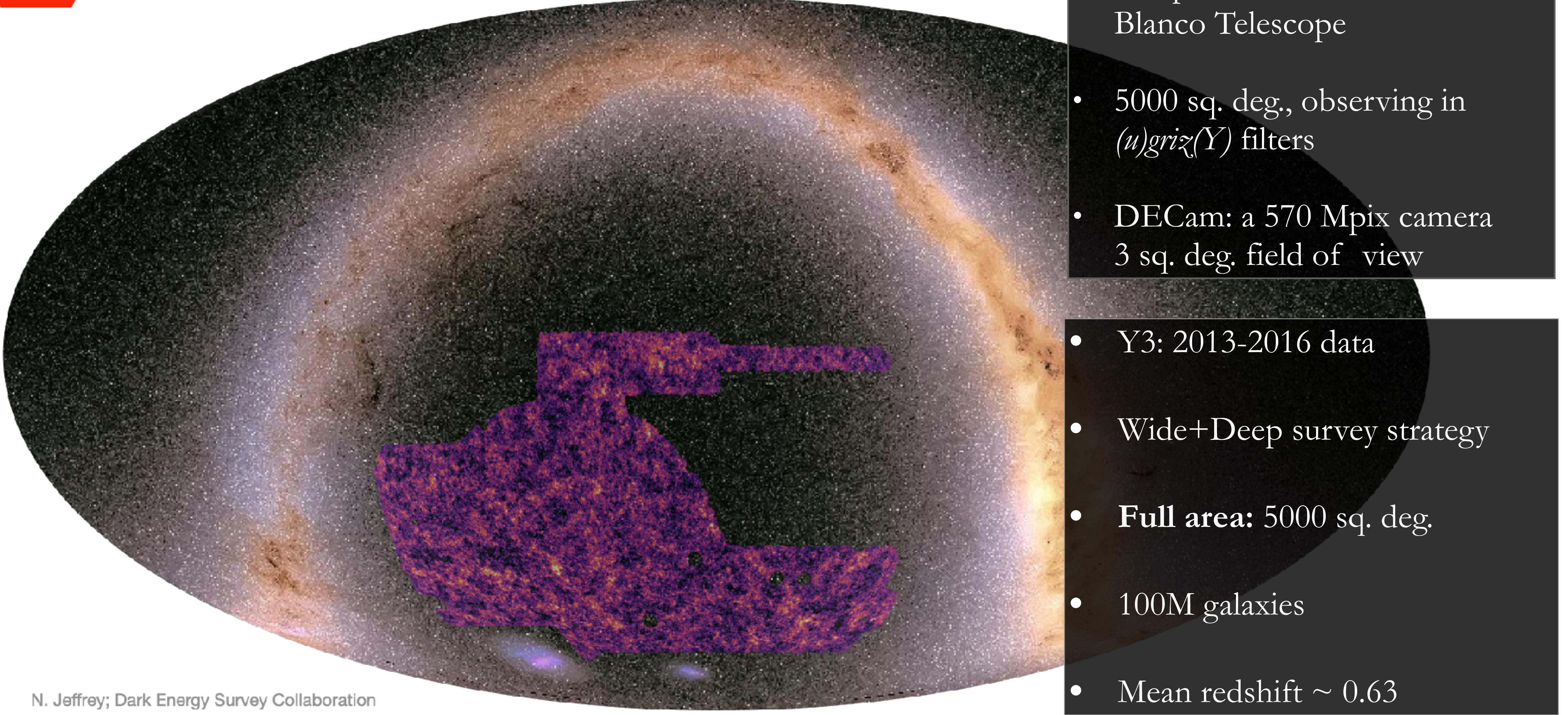
Outlook:

the most powerful weak-lensing catalog & ‘ 3×2 -point’
novel methods
conservative modelling choices
cosmological constraints in context
cosmic shear going forward





Dark Energy Survey: Y3



N. Jeffrey; Dark Energy Survey Collaboration

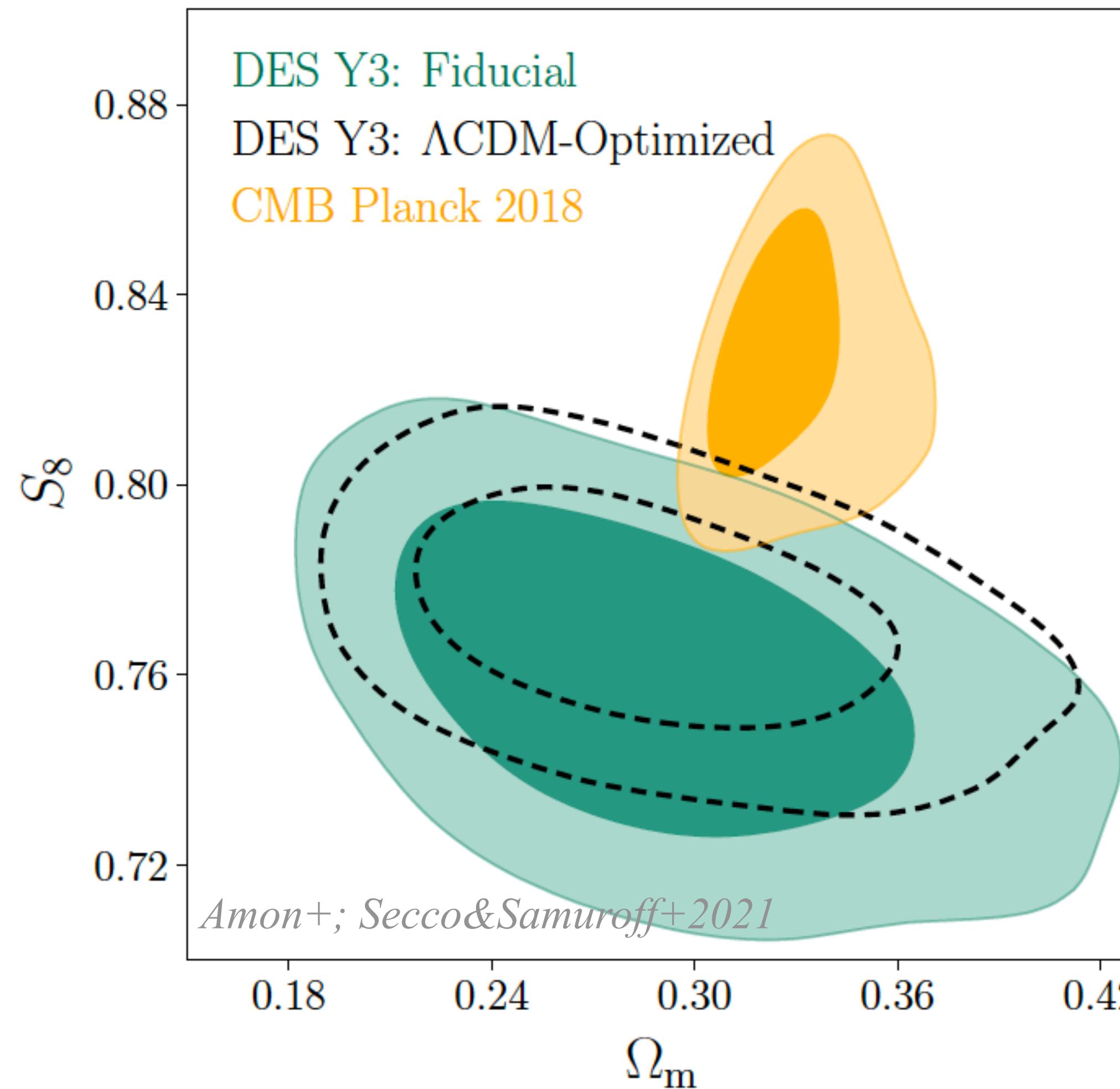
- Survey status: 6 yrs observations complete on the 4m CTIO Blanco Telescope
- 5000 sq. deg., observing in $(u)griz(Y)$ filters
- DECam: a 570 Mpix camera 3 sq. deg. field of view
- Y3: 2013-2016 data
- Wide+Deep survey strategy
- **Full area:** 5000 sq. deg.
- 100M galaxies
- Mean redshift ~ 0.63



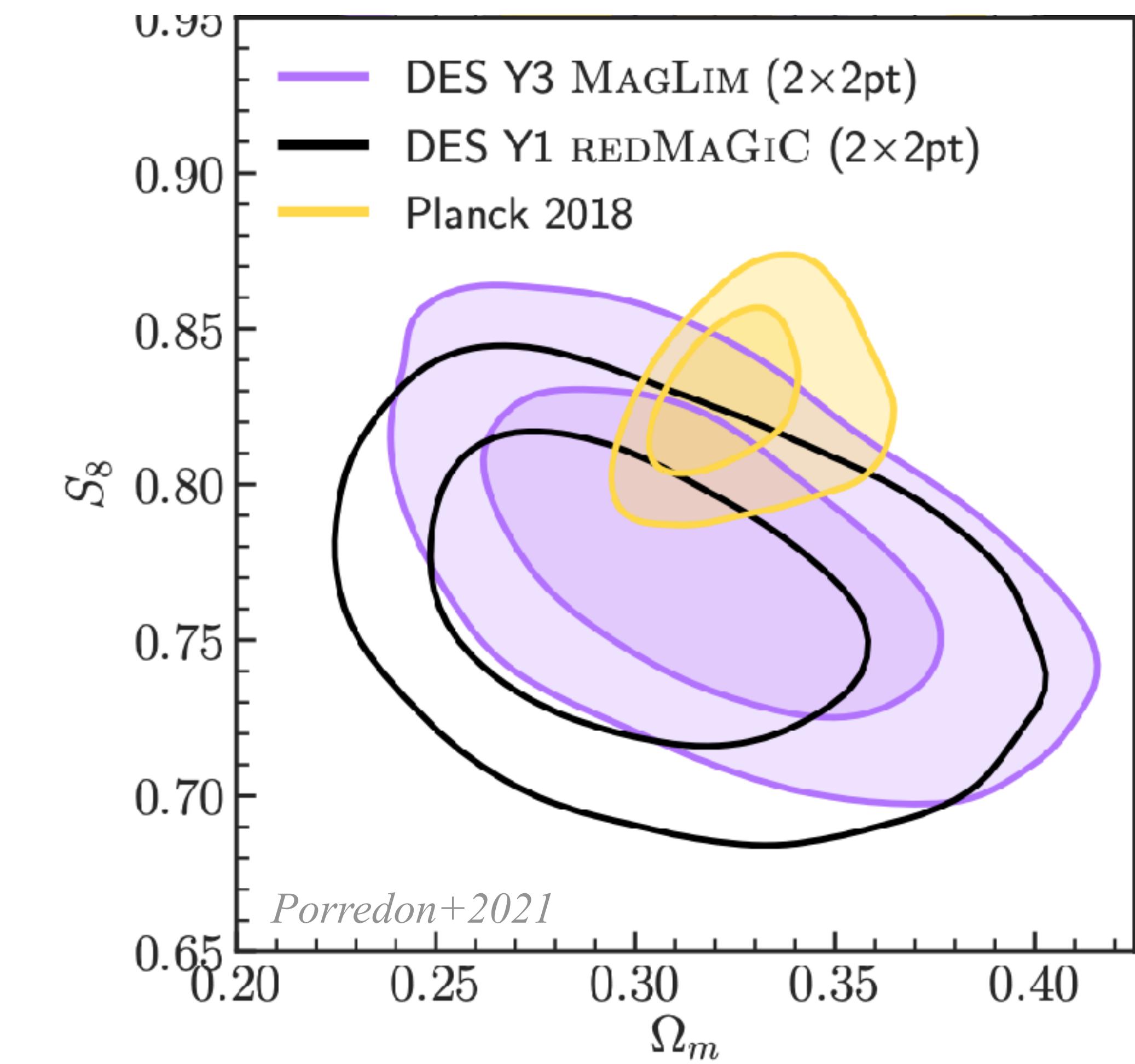


DES Y3 cosmological constraints

cosmic shear



' $2\times 2pt$ '



DES Y3

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Data calibration

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Analysis choices

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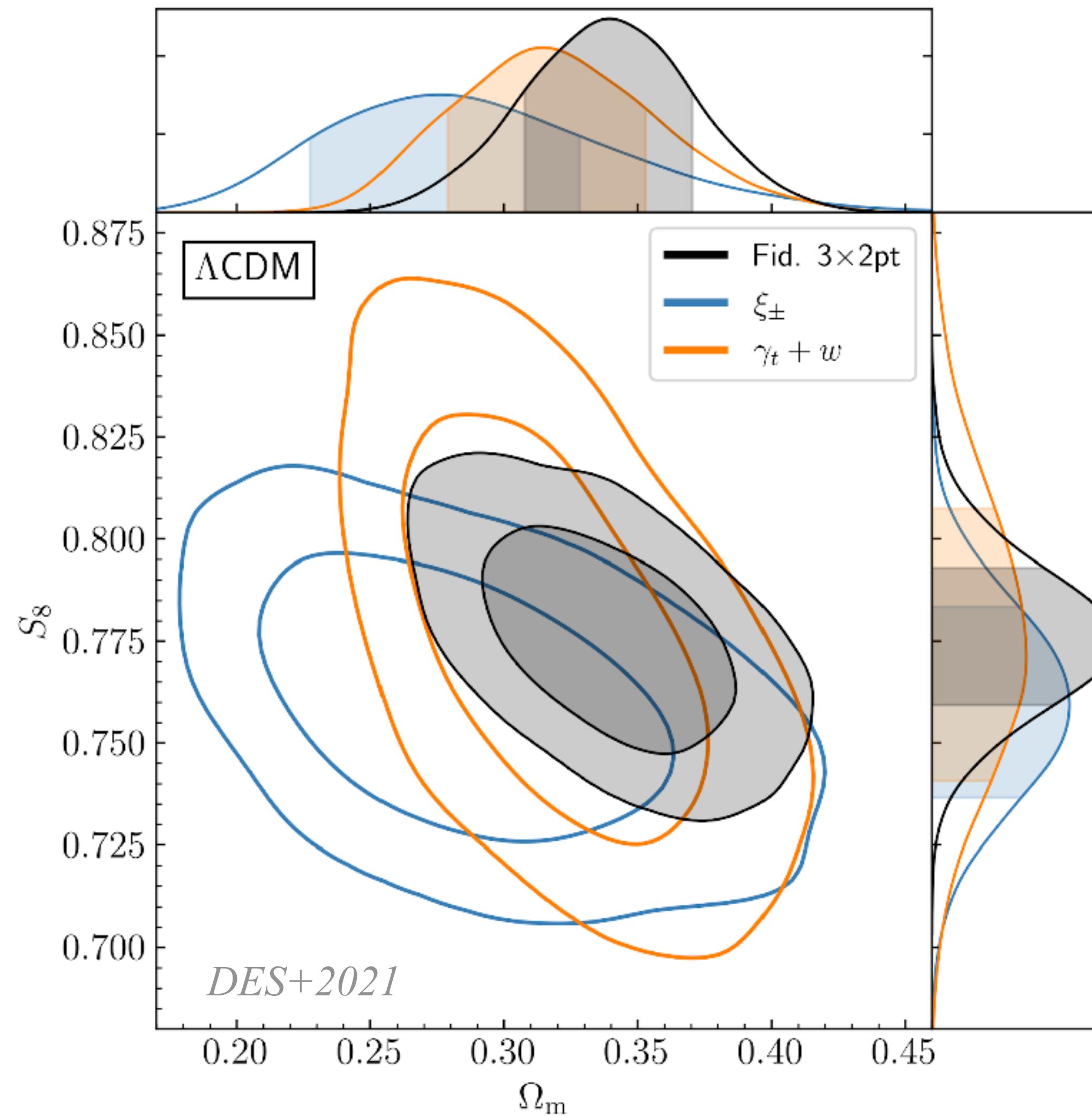
The big picture

—

Looking ahead



DES Y3 $3\times 2pt$



A factor of 2.1 improvement in signal-to-noise from DES Year 1.

No significant evidence of inconsistency between **DES Y3 3x2pt** and *Planck* CMB at $0.7\text{-}1.5\sigma$ or $p=0.13\text{-}0.48$.

$$S_8 = 0.776^{+0.017}_{-0.017} \quad (0.776)$$

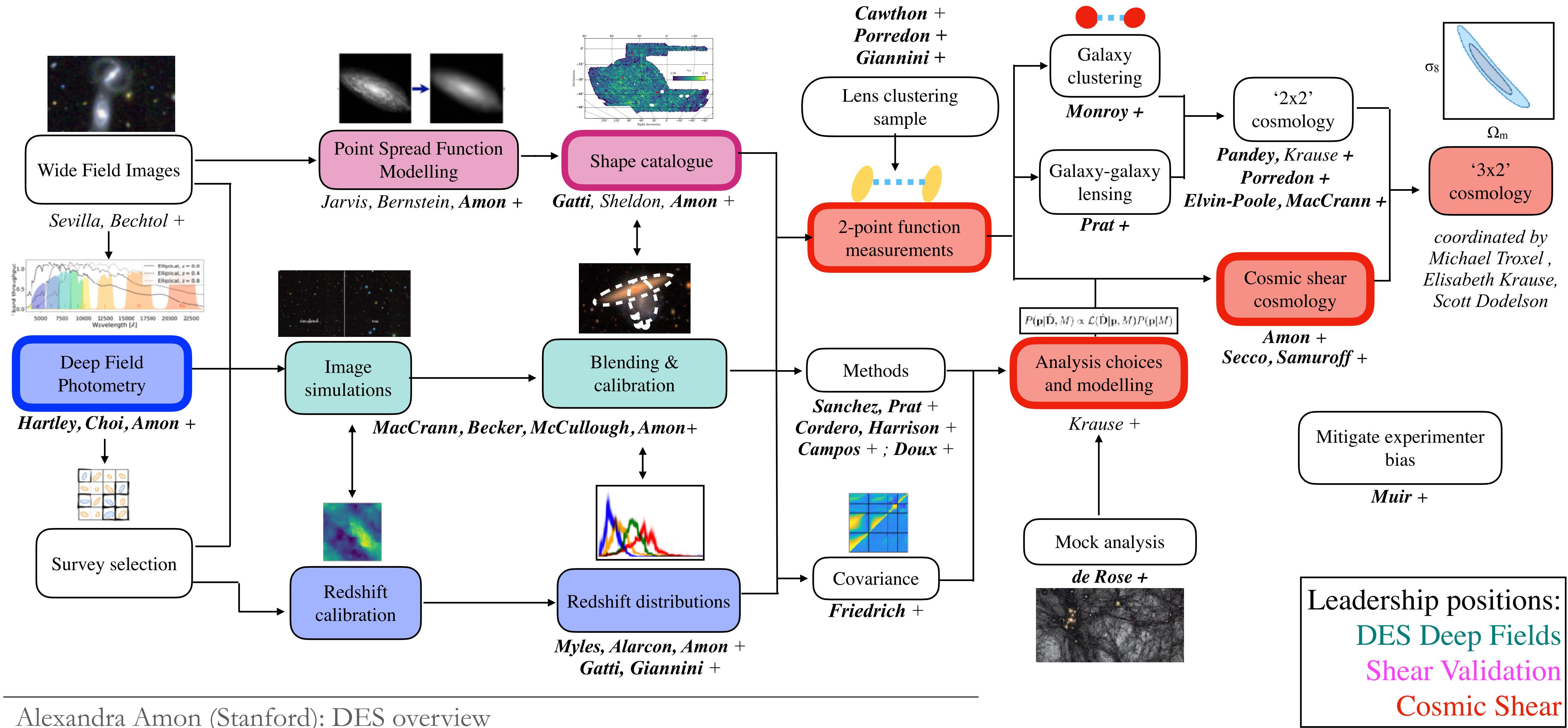
$$\text{In } \Lambda\text{CDM: } \Omega_m = 0.339^{+0.032}_{-0.031} \quad (0.372)$$

$$\sigma_8 = 0.733^{+0.039}_{-0.049} \quad (0.696)$$

$$\text{In } w\text{CDM: } \Omega_m = 0.352^{+0.035}_{-0.041} \quad (0.339)$$
$$w = -0.98^{+0.32}_{-0.20} \quad (-1.03)$$



DES Y3 $3\times 2pt$: pixels to cosmology



➡ Respond at PollEv.com/alexandraamon538

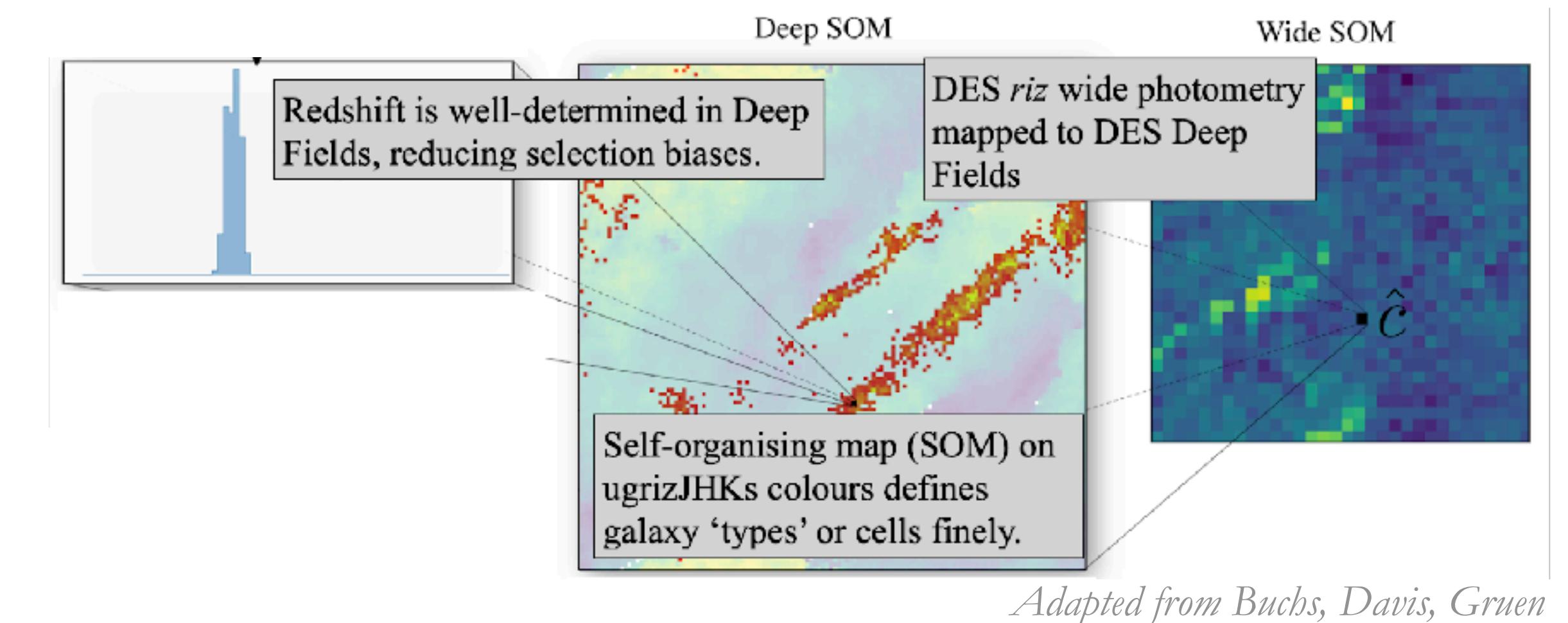
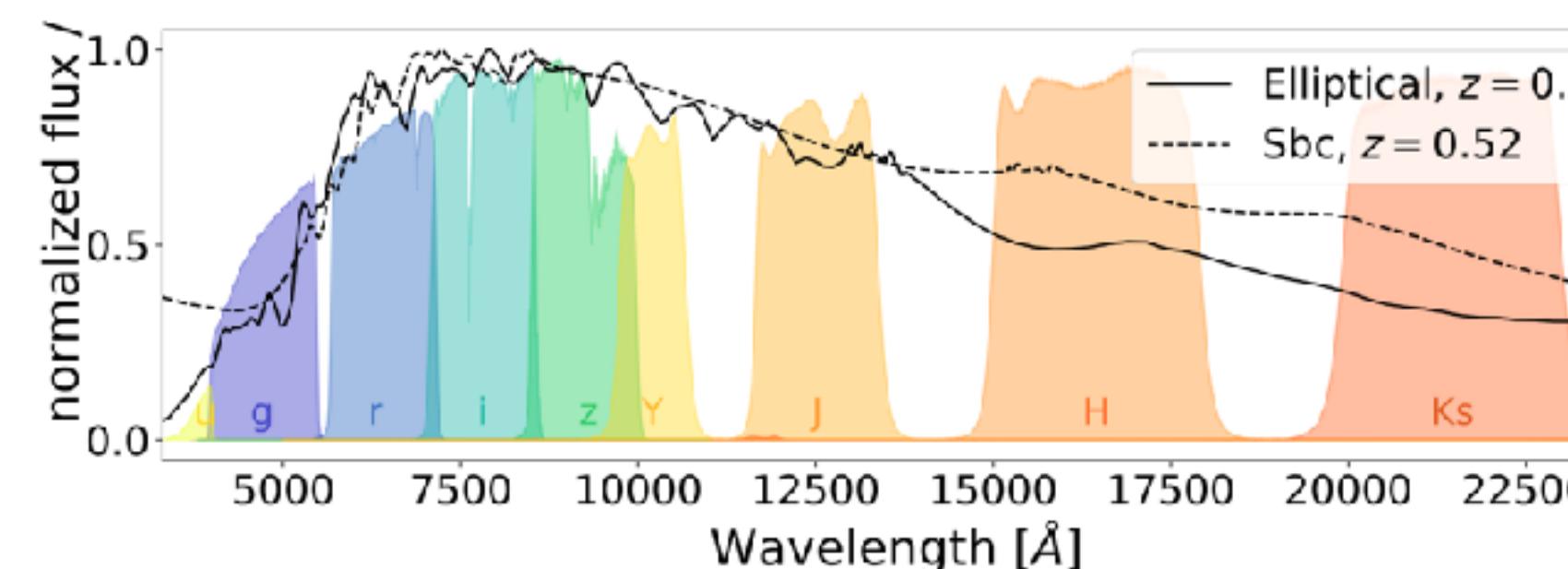
What part of the DES Y3 analysis are you worried about?

- Redshift calibration
- Shape calibration
- Intrinsic alignments
- Baryonic effects
- 2x2 Maglim Results
- Summer hols: DES Y3 is great!

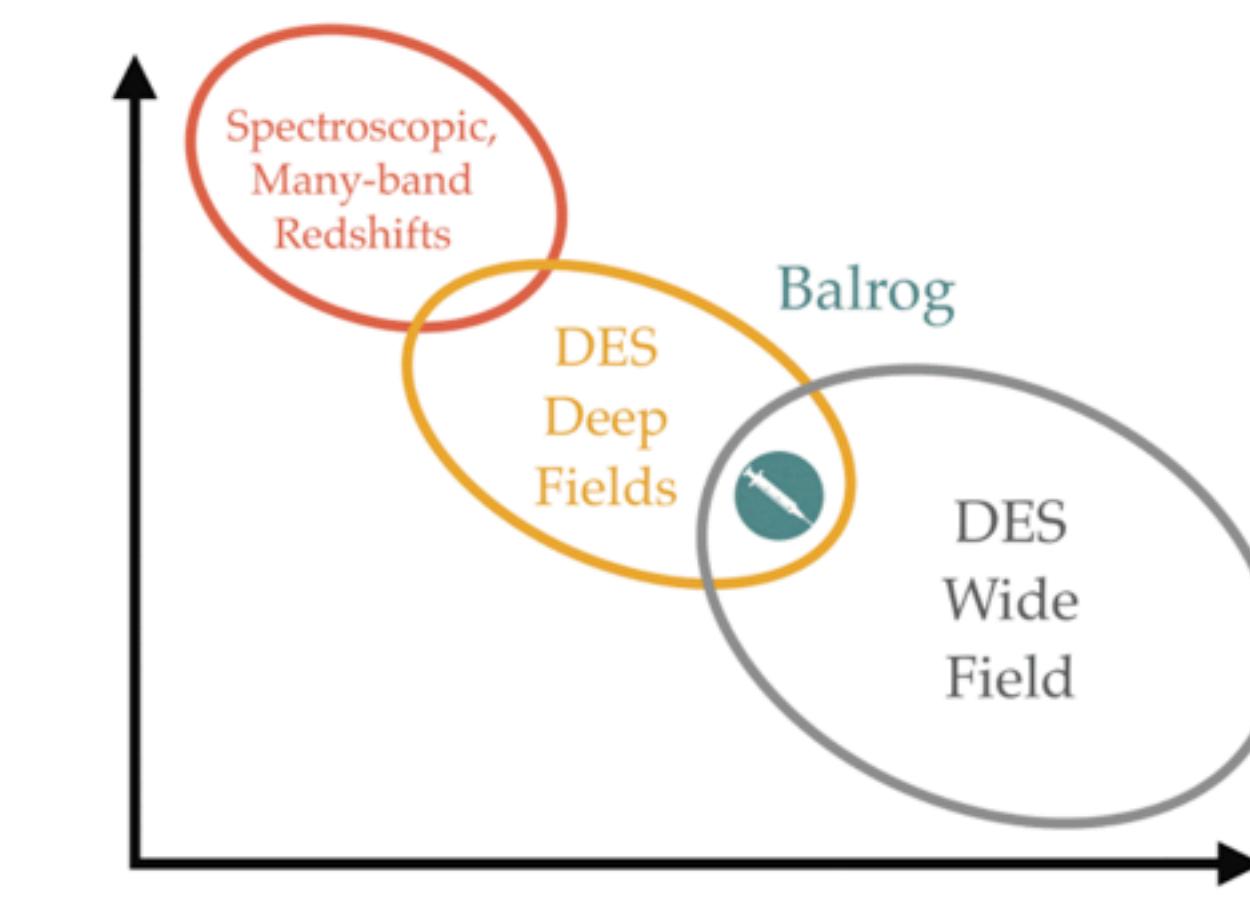


DES Y3: novel framework for redshift calibration

1. Breaks colour-redshift degeneracies with a **Deep-Wide survey strategy** (*Hartley, Choi, Amon+*) and a machine-learning approach



Adapted from Buchs, Davis, Gruen



(Myles, Alarcon, Amon +2020)

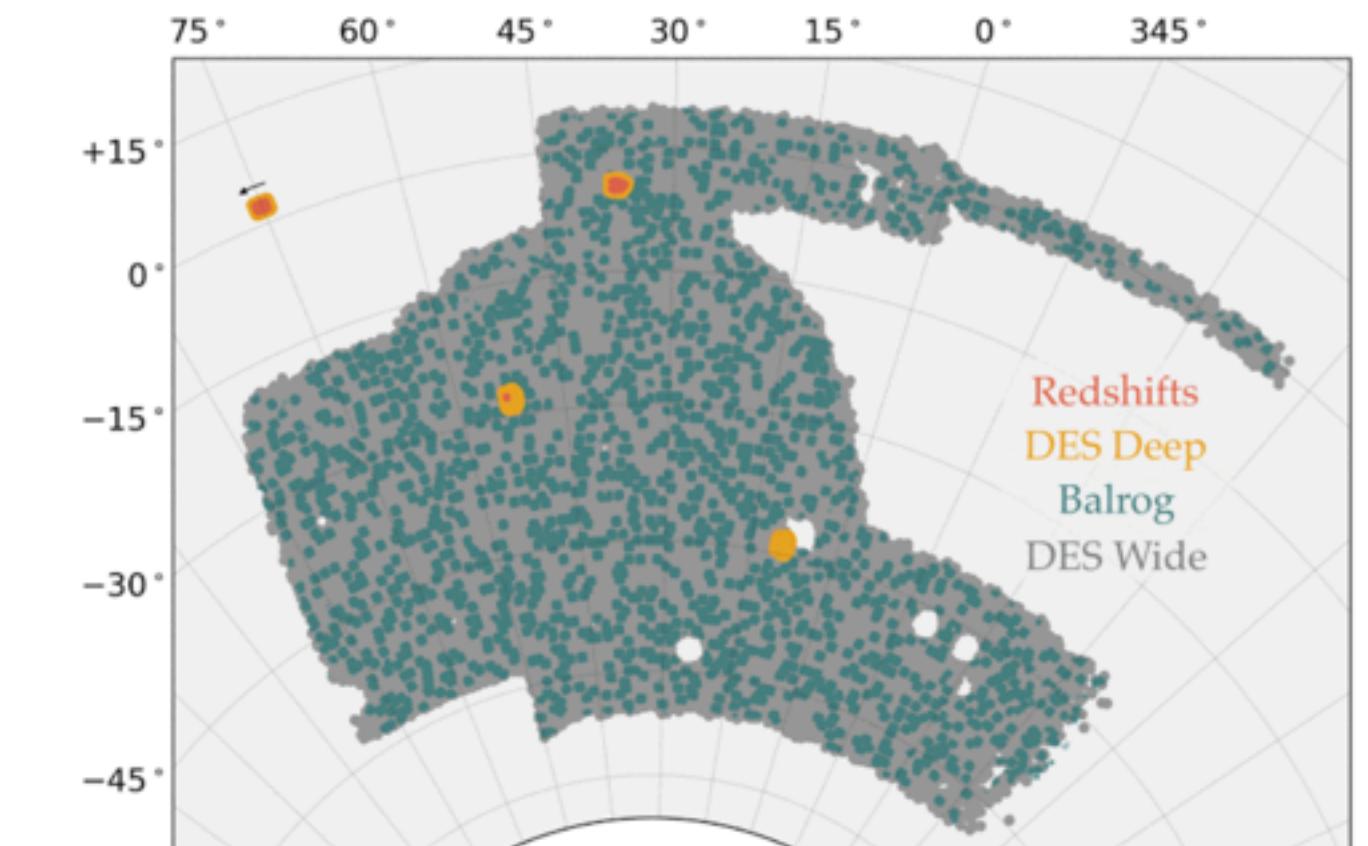
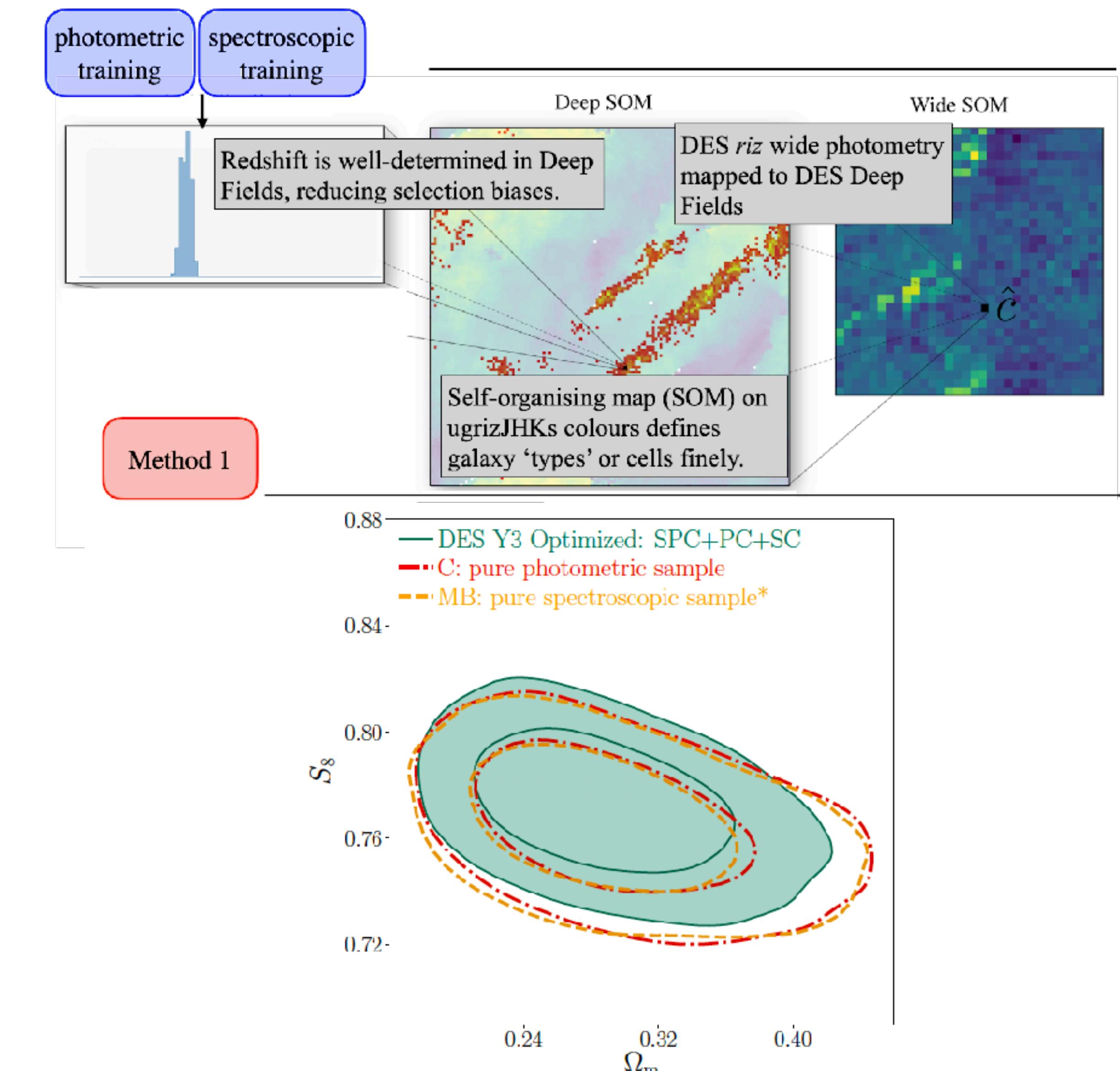


Image credit: Carles Sanchez
Deep Fields: Hartley, Choi, Amon +2020
Balrog: Everett +2020



DES Y3: novel framework for redshift calibration

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2. Primary method **mitigates biases in the colour-redshift relation** due to selection effects or photometric outliers using combined samples



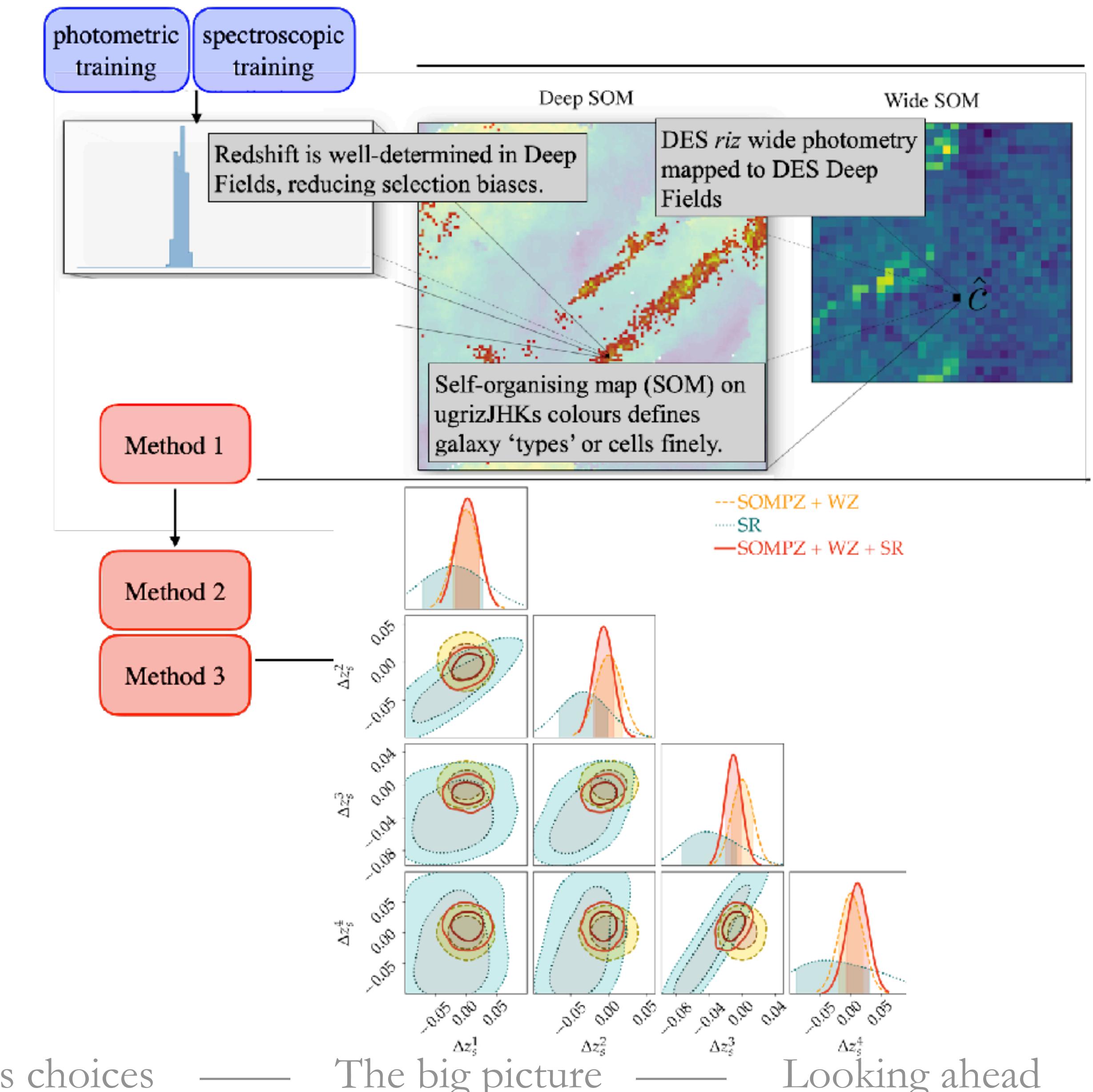
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3. Cross-checks **and combines** with independent sources of information: clustering redshifts (*Gatti, Giannini+*) and shear ratios (*Sanchez, Prat+*)

(*Myles, Alarcon, Amon +2020*)

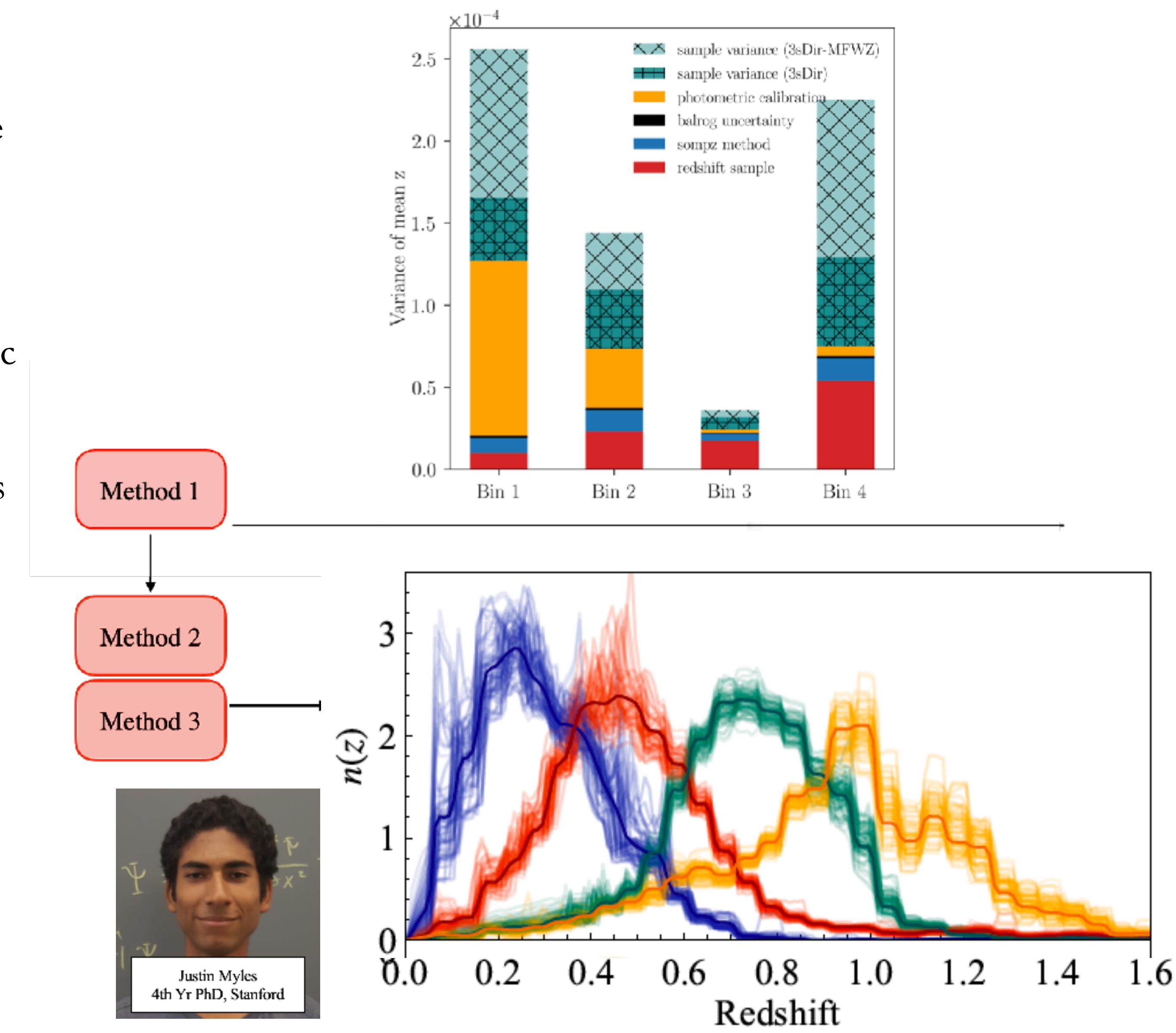




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3. Cross-checks **and combines** with independent sources of information: clustering redshifts (*Gatti, Giannini+*) and shear ratios (*Sanchez, Prat+*)
4. **Characterises the full uncertainty** on the shape and mean of the redshift distributions, including any flux calibration errors, sample variance and the uncertainty on the method as determined by simulations.

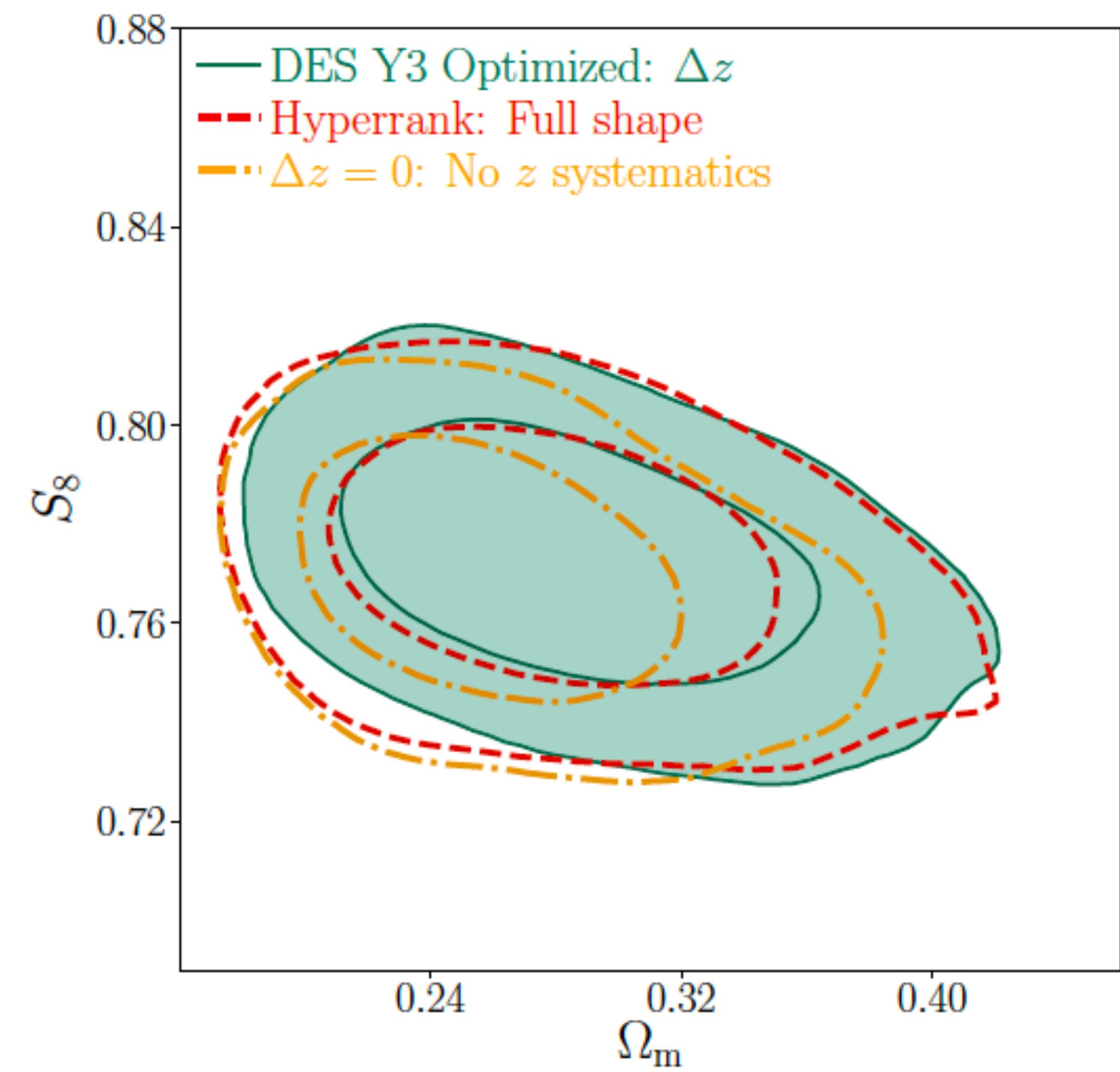
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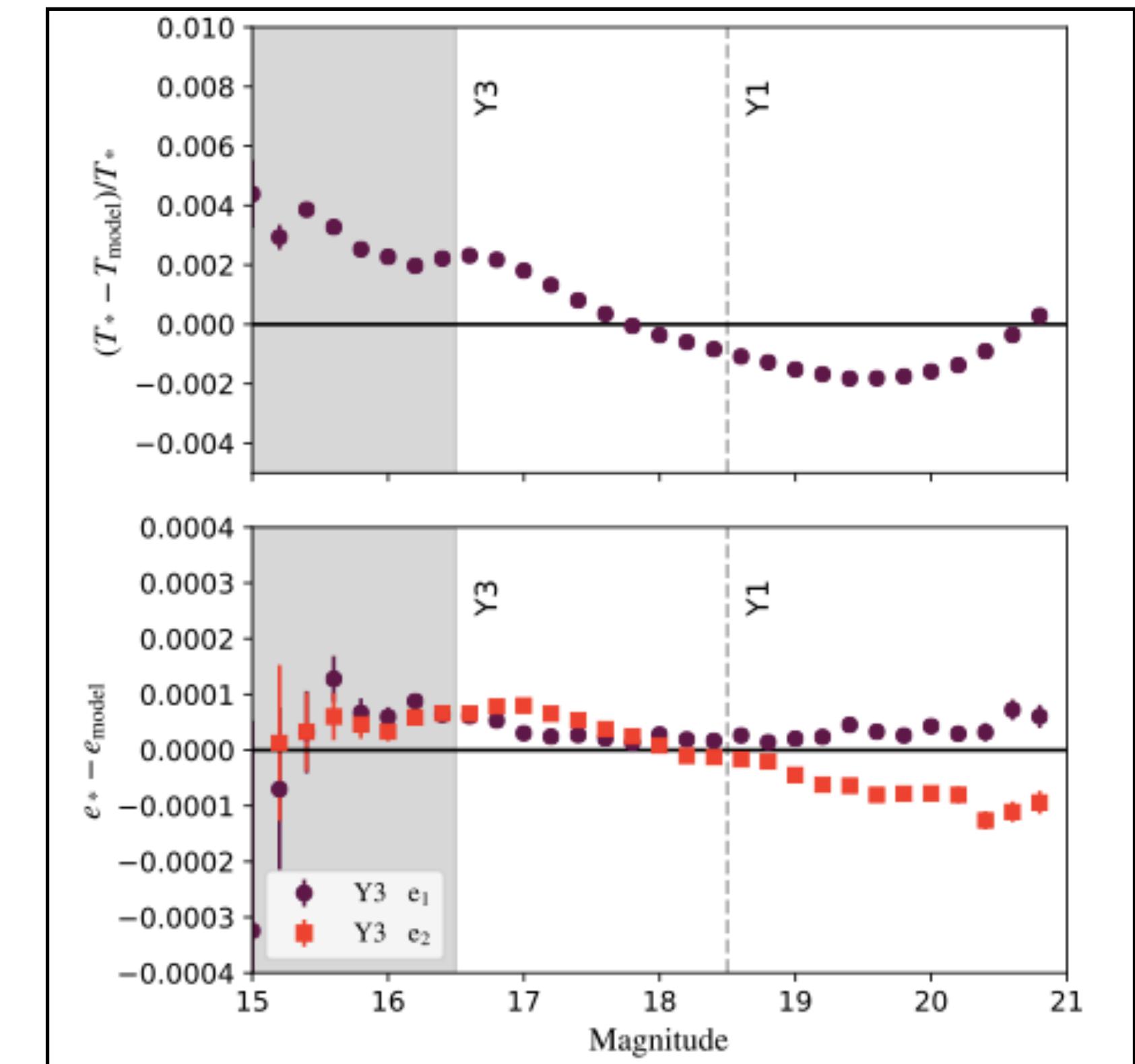
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4. **Characterises the full uncertainty** on the shape and mean of the redshift distributions, including any flux calibration errors, sample variance and the uncertainty on the method as determined by simulations.
5. Can marginalise over ensemble with Hyperrank (*Cordero+*)
(*Myles, Alarcon, Amon +2020*)





DES Y3: calibration to account for ‘blending’

1. New Point Spread Function modelling (*Jarvis, Bernstein, Amon+*) :



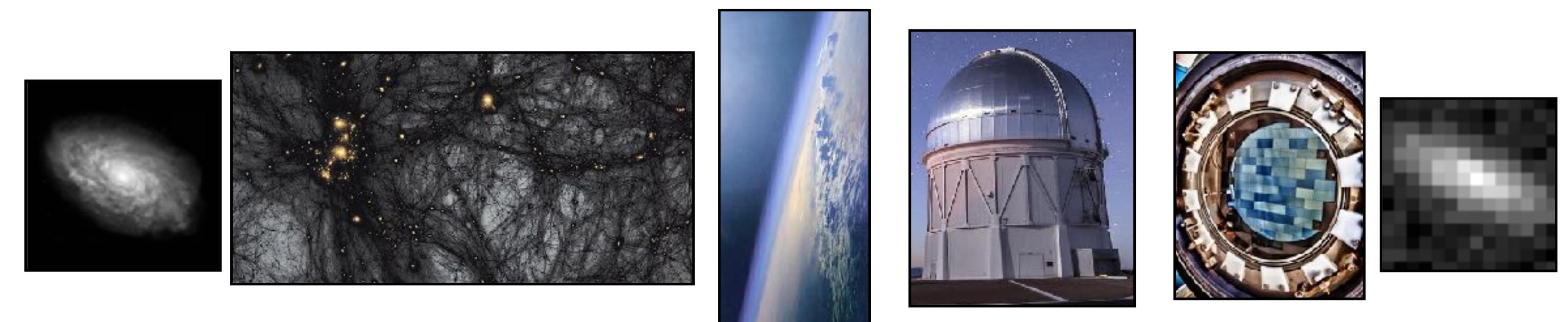


DES Y3: calibration to account for ‘blending’

1. New Point Spread Function modelling (*Jarvis, Bernstein, Amon+*) and extended suite of tests on the metacalibration catalogue (*Gatti, Sheldon, Amon+*).

| Bin | no. objects | n_{eff} | σ_e | $\langle R_\gamma \rangle$ | $\langle R_s \rangle$ |
|------|-------------|------------------|------------|----------------------------|-----------------------|
| Full | 100 204 026 | 5.590 | 0.268 | | |
| 0 | 24 940 465 | 1.476 | 0.243 | 0.7636 | 0.0046 |
| 1 | 25 280 405 | 1.479 | 0.262 | 0.7182 | 0.0083 |
| 2 | 24 891 859 | 1.484 | 0.259 | 0.6887 | 0.0126 |
| 3 | 25 091 297 | 1.461 | 0.301 | 0.6154 | 0.0145 |

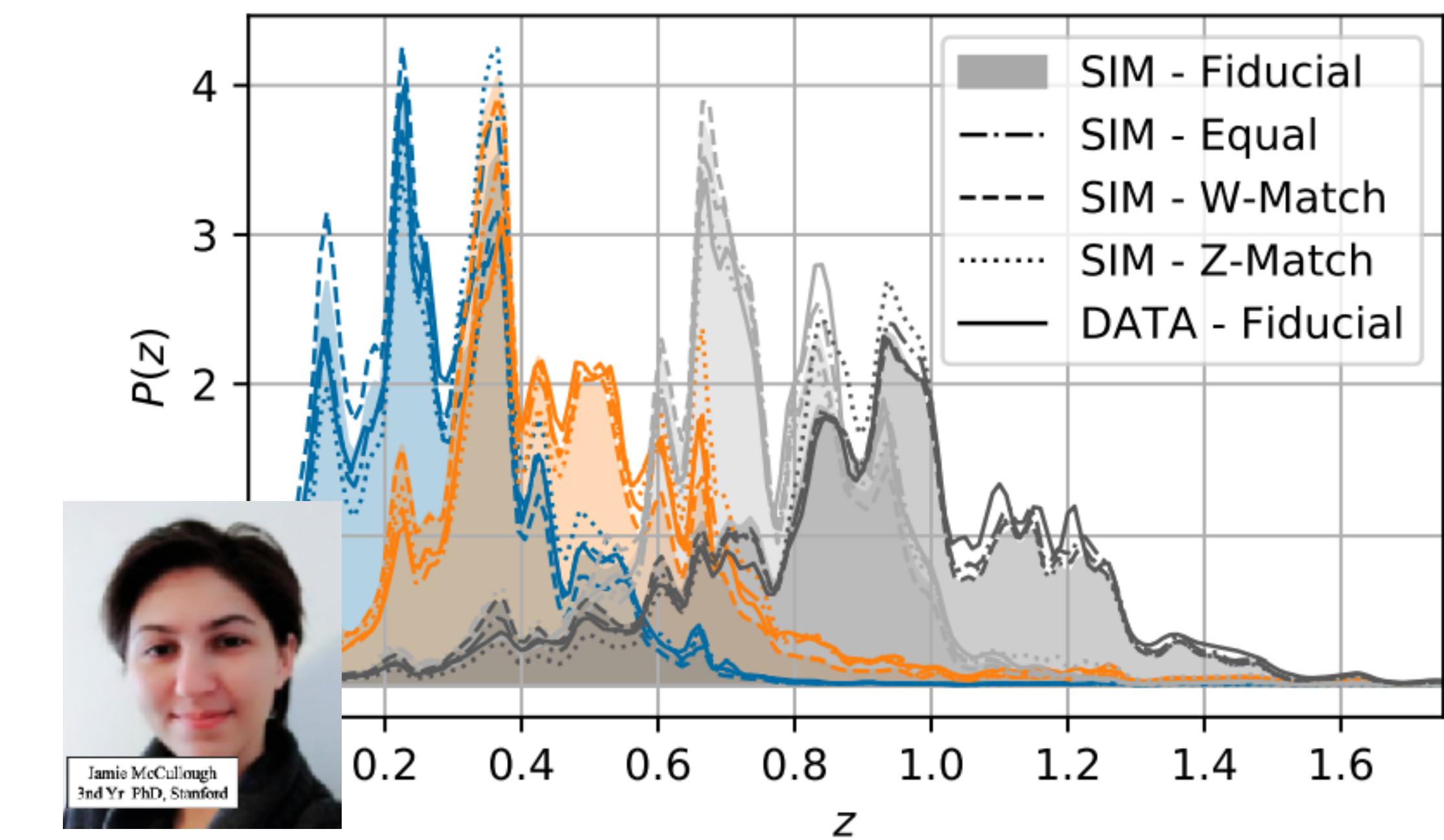
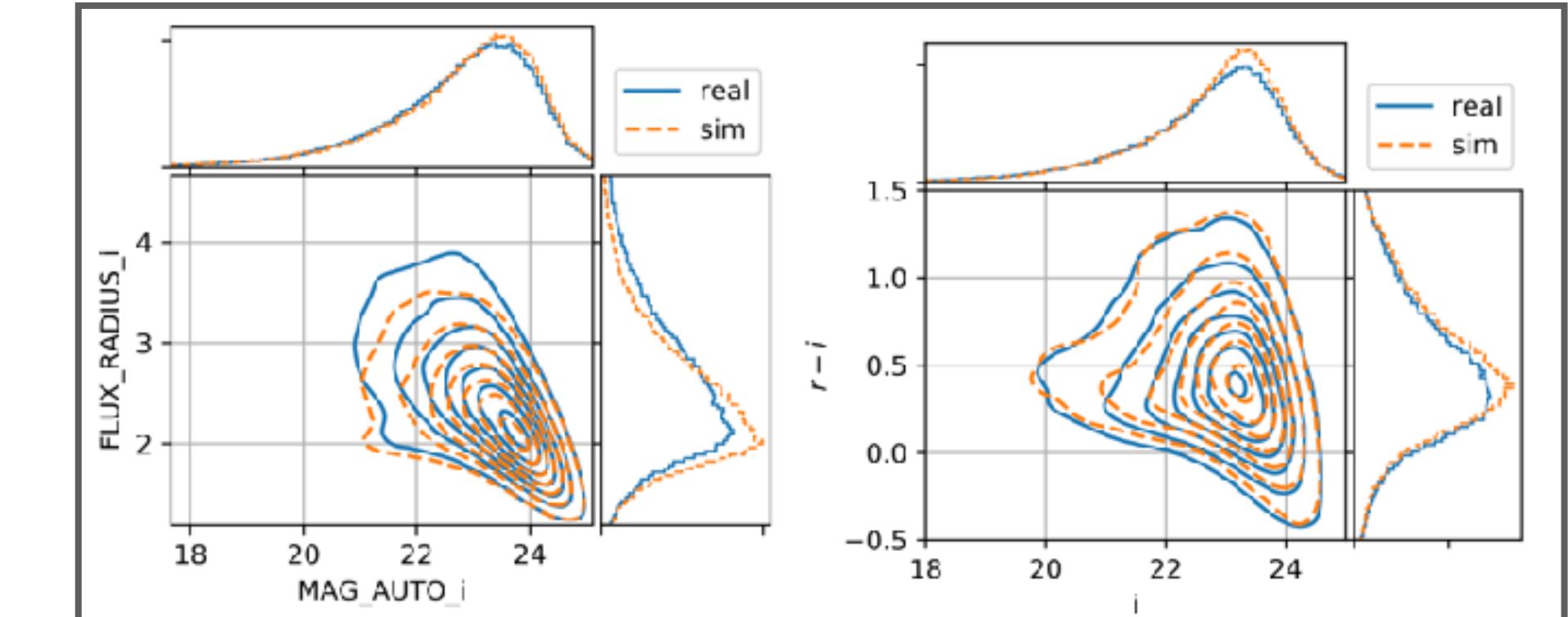
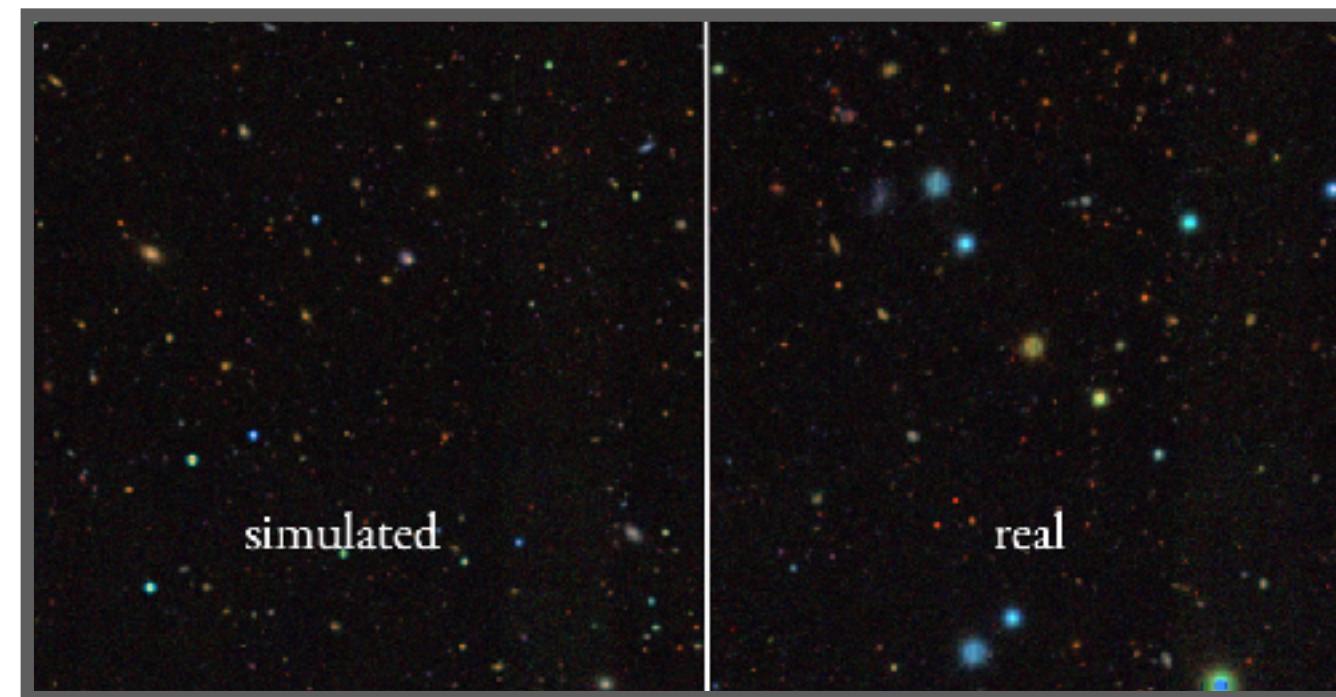
$$\langle R_s \rangle_{i,j} = \frac{\langle e_i \rangle^{s_+} - \langle e_i \rangle^{s_-}}{\Delta \gamma_j}$$





DES Y3: calibration to account for ‘blending’

1. New Point Spread Function modelling (*Jarvis, Bernstein, Amon+*) and extended suite of shape catalogue tests (*Gatti, Sheldon, Amon+*).
2. Developed **image simulations** that are well-matched to data used as a testing bed
3. Perform the full **redshift analysis** on simulations to understand shear redshift-dependent effect of crowded galaxy fields

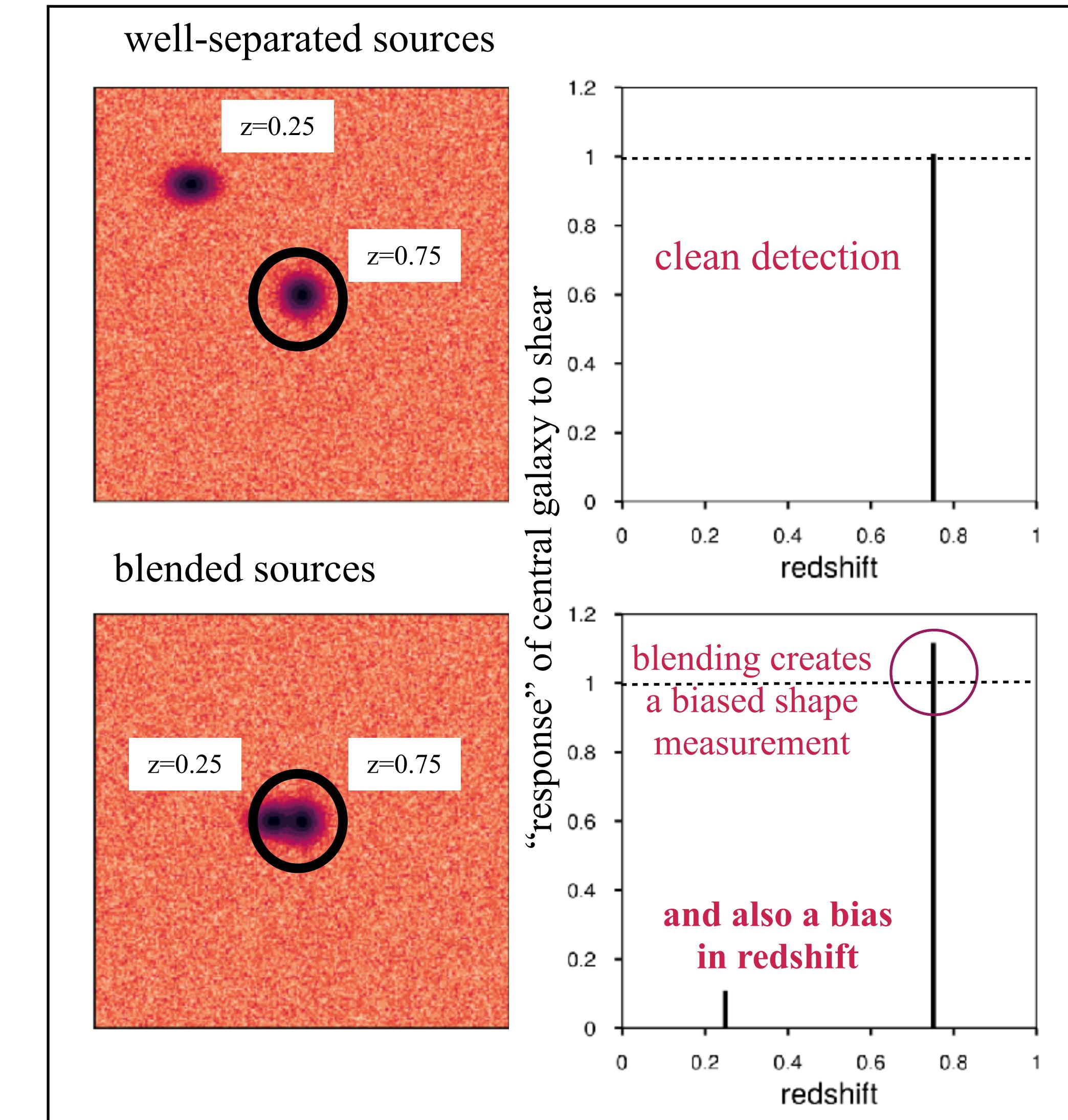




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4. **Detected that measured shapes ‘respond’ to the shear of galaxies at other redshifts.**
5. Modelled and accounted for the impact of blending as a redshift-mixing effect

(*MacCrann, Becker, McCullough, Amon+2020*)

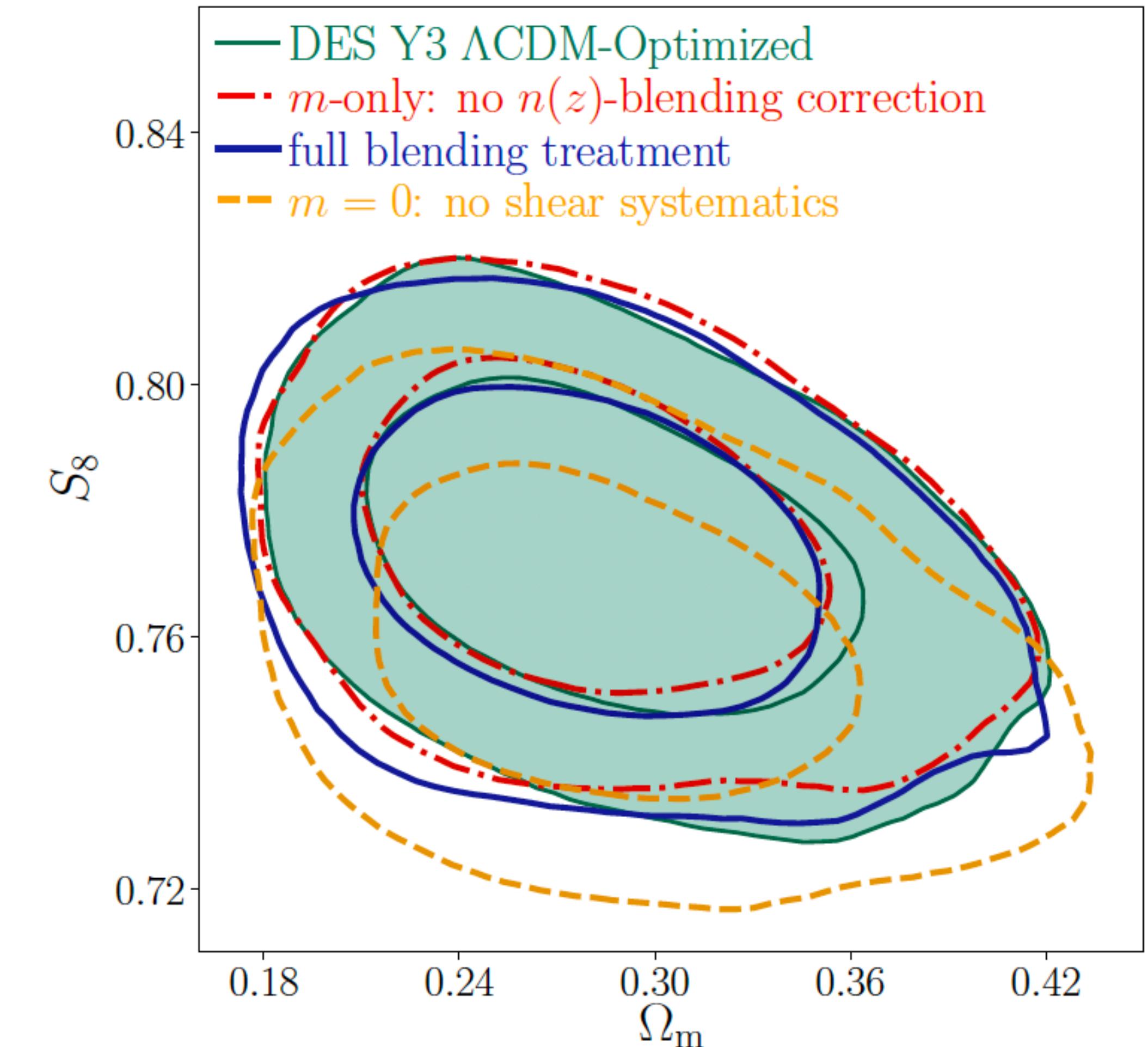




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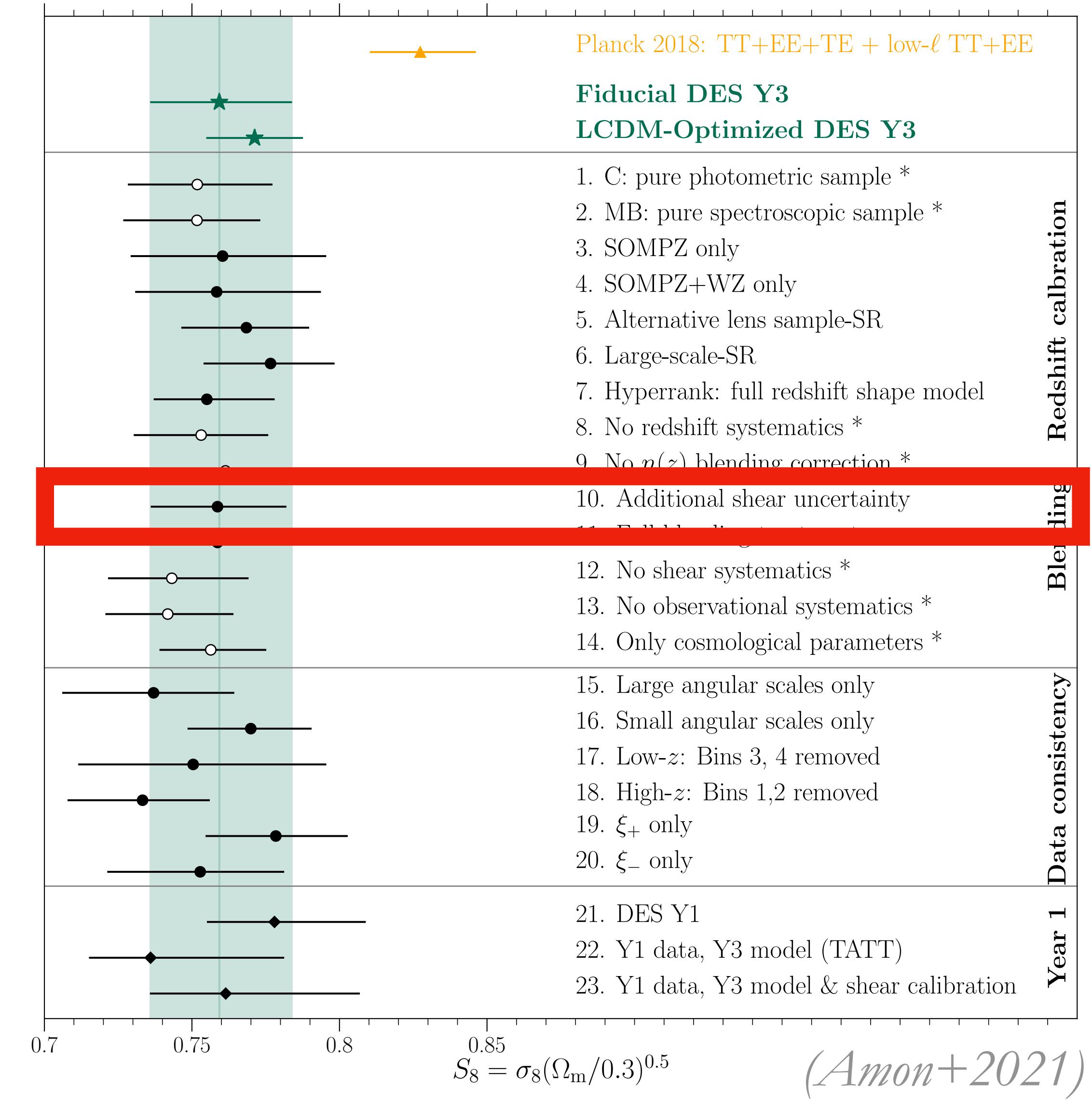
(*Amon, Gruen, Troxel, MacCrann, Dodelson+2021*)



DES Y3: calibration to account for ‘blending’

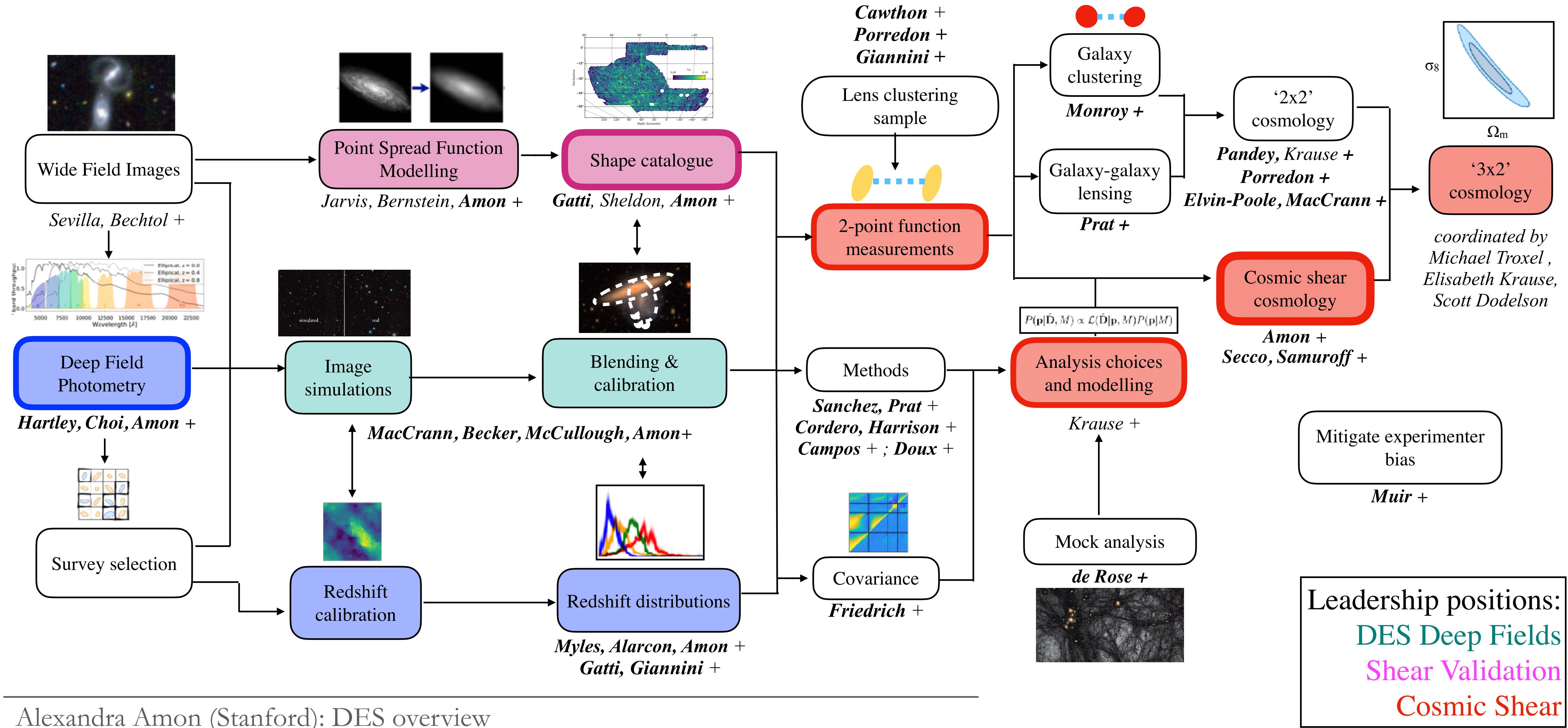
Understanding blending in the future:

- clustering in the simulations, estimated in Martinet+2019 as an upper bound of 1% effect
- We tested that the cosmic shear results were stable to this additional uncertainty
- With Jamie McCullough, Daniel Gruen and others, we are doing more detailed investigations of blending effects using the DES Y3 image simulations.





DES Y3 $3\times 2pt$: pixels to cosmology



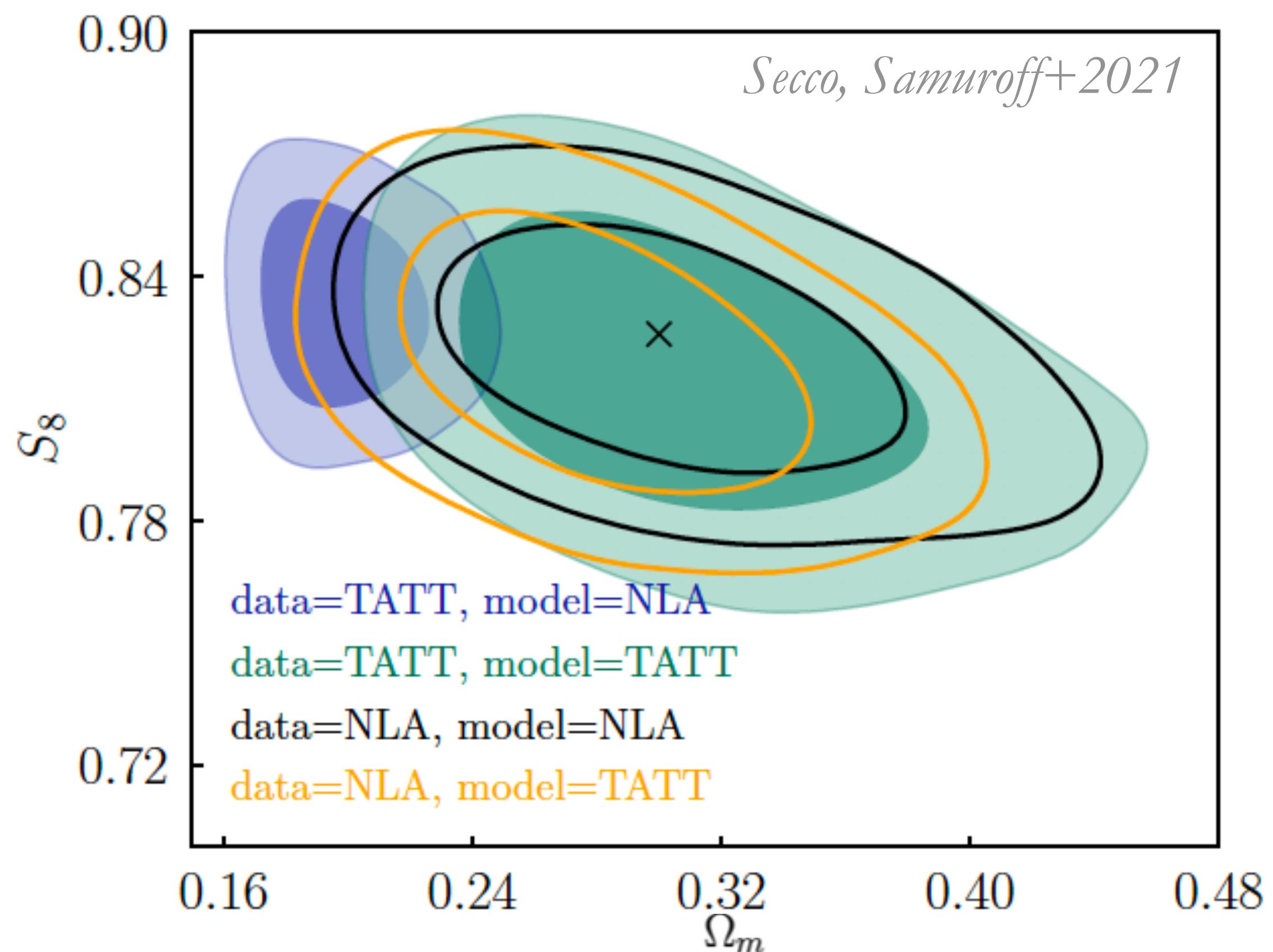


DES Y3: model choices

Intrinsic alignments: Tidal Alignment & Tidal Torquing model, TATT

- TATT is a superspace of the NLA model
- physically motivated to account for late-types
- simulated tests found NLA model can bias cosmology*

Baryonic effects: eliminate small-scale measurements

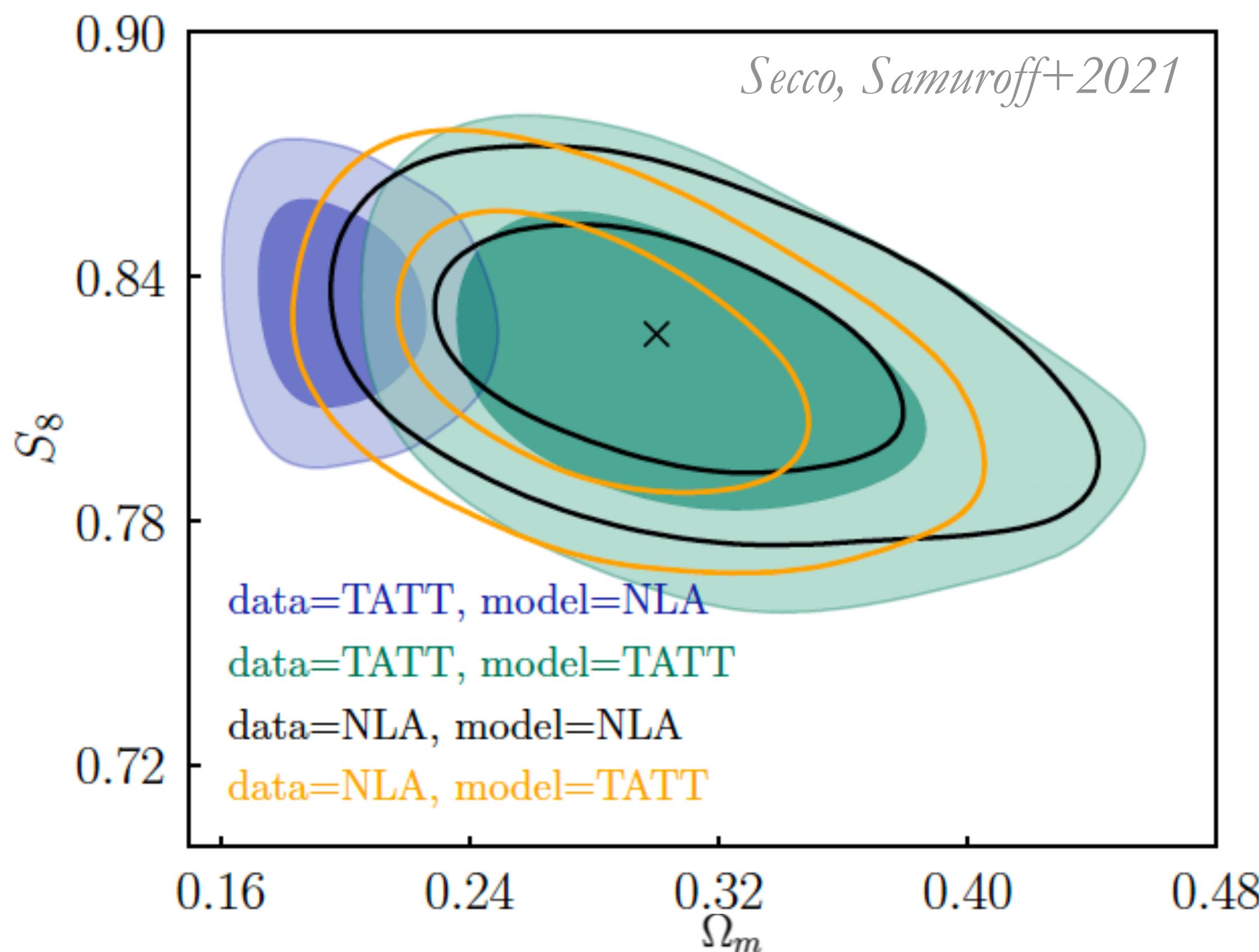




DES Y3: model choices

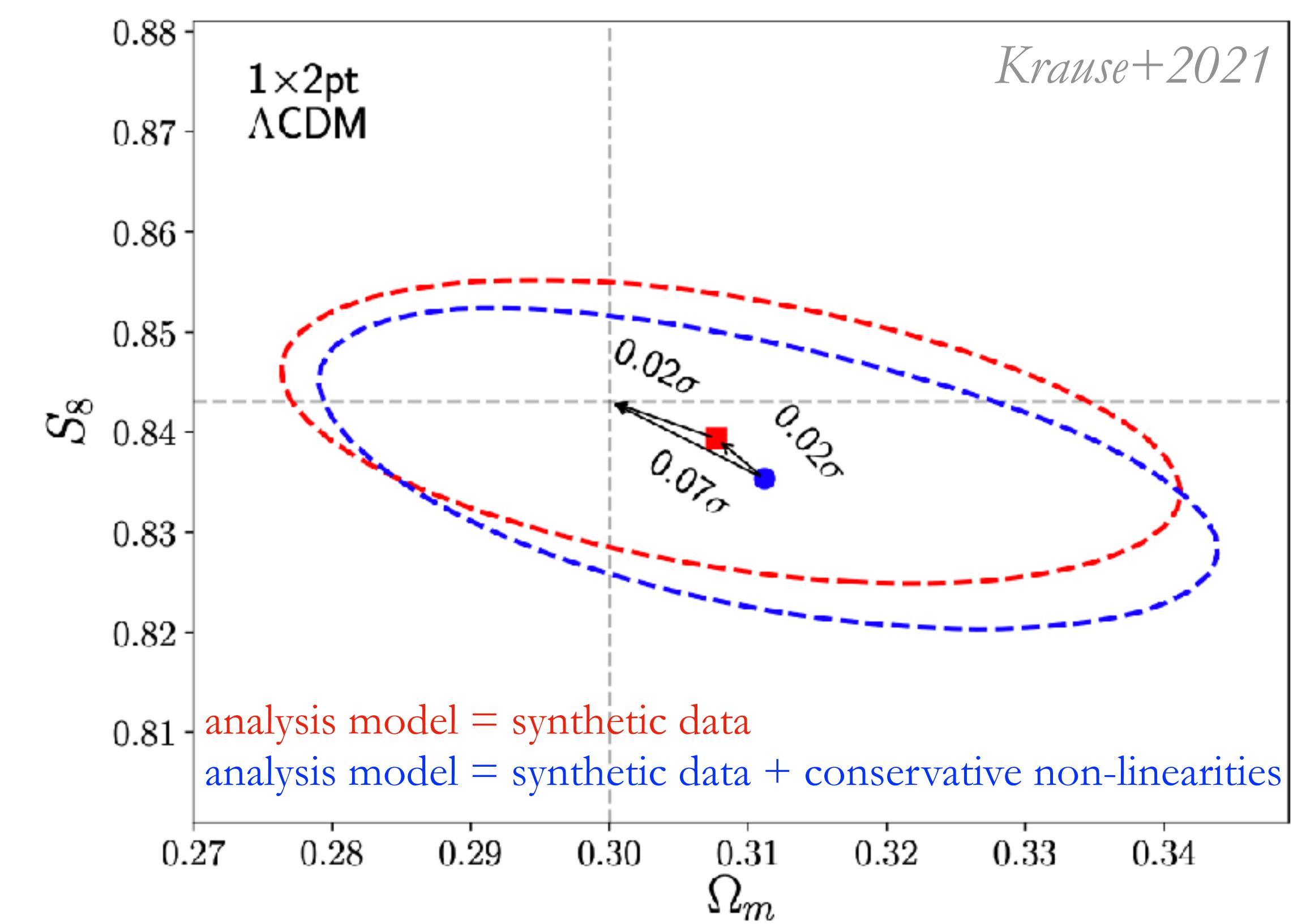
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Baryonic effects: eliminate small-scale measurements

- Bracket impact of effect with OWLS-AGN hydro-simulation
- Threshold of bias in 3x2pt cosmology constraints is 0.3σ
- Simulated tests find a maximum bias of 0.1σ for cosmic shear



DES Y3

— Data calibration

— Analysis choices

— The big picture

— Looking ahead



DES Y3: model choices

Cosmology (7 parameters):

Λ CDM/wCDM with massive neutrinos

Astrophysical model (9 parameters):

- Galaxy bias
- Magnification (fixed)
- Intrinsic Alignments (TATT)

Control non-linear modeling uncertainties through scale cuts
(Krause+, DeRose+)

Calibration systematics (16 parameters)

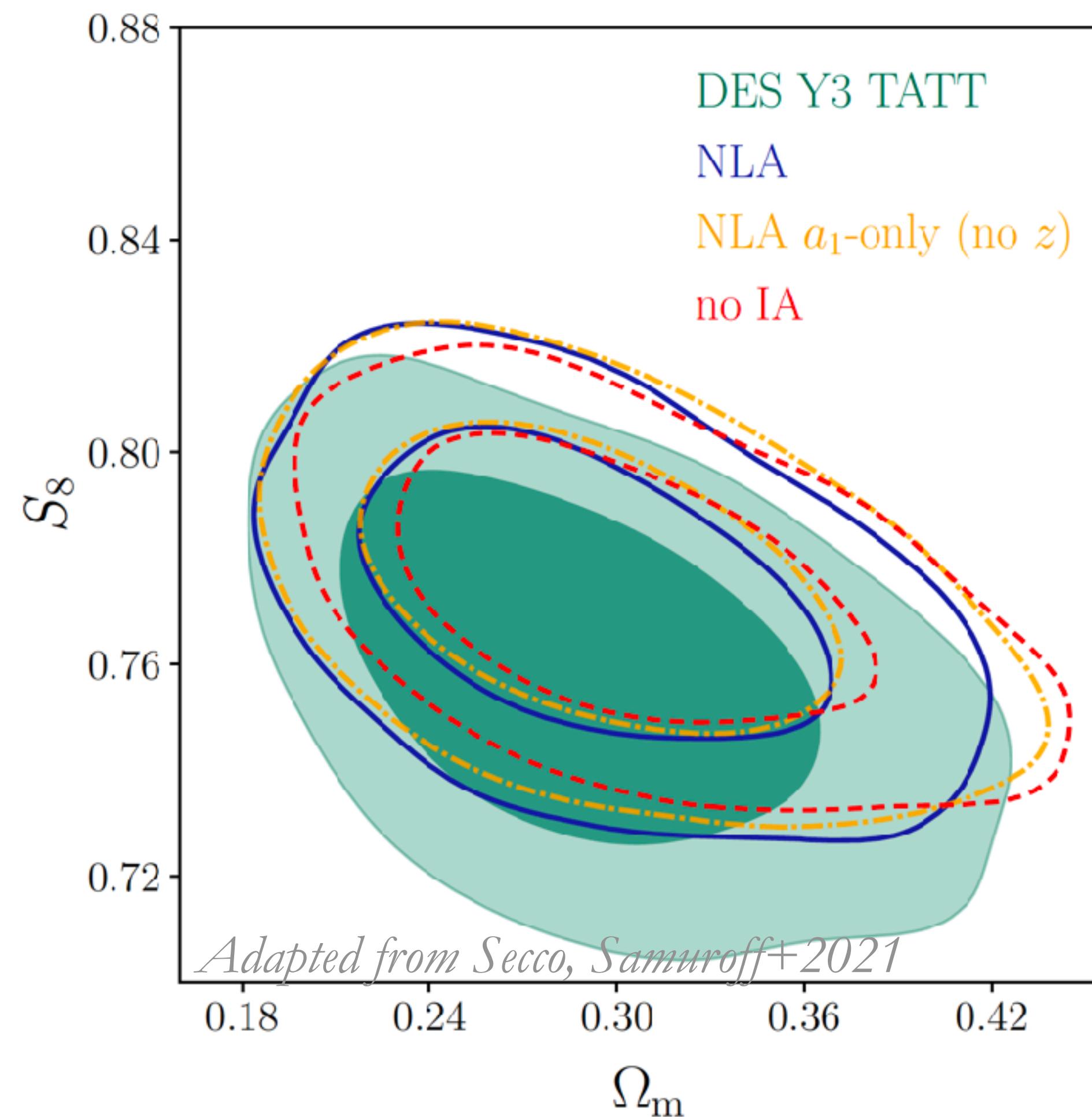
| Parameter | Prior |
|----------------------------|-----------------------|
| Cosmology | |
| Ω_m | Flat (0.1, 0.9) |
| $10^9 A_s$ | Flat (0.5, 5.0) |
| n_s | Flat (0.87, 1.07) |
| Ω_b | Flat (0.03, 0.07) |
| h | Flat (0.55, 0.91) |
| $10^3 \Omega_\nu h^2$ | Flat (0.60, 6.44) |
| w | Flat (-2.0, -0.33) |
| Lens Galaxy Bias | |
| $b_i (i \in [1, 4])$ | Flat (0.8, 3.0) |
| Lens magnification | |
| C_1^1 | Fixed 1.21 |
| C_1^2 | Fixed 1.15 |
| C_1^3 | Fixed 1.88 |
| C_1^4 | Fixed 1.97 |
| Lens photo-z | |
| $\Delta z_1^1 \times 10^2$ | Gaussian (-0.9, 0.7) |
| $\Delta z_1^2 \times 10^2$ | Gaussian (-3.5, 1.1) |
| $\Delta z_1^3 \times 10^2$ | Gaussian (-0.5, 0.6) |
| $\Delta z_1^4 \times 10^2$ | Gaussian (-0.7, 0.6) |
| $\sigma_{z,1}^1$ | Gaussian (0.98, 0.06) |
| $\sigma_{z,1}^2$ | Gaussian (1.31, 0.09) |
| $\sigma_{z,1}^3$ | Gaussian (0.87, 0.05) |
| $\sigma_{z,1}^4$ | Gaussian (0.92, 0.05) |
| Intrinsic Alignment | |
| $a_i (i \in [1, 2])$ | Flat (-5, 5) |
| $\eta_i (i \in [1, 2])$ | Flat (-5, 5) |
| b_{TA} | Flat (0, 2) |
| z_0 | Fixed 0.62 |
| Source photo-z | |
| $\Delta z_s^1 \times 10^2$ | Gaussian (0.0, 1.8) |
| $\Delta z_s^2 \times 10^2$ | Gaussian (0.0, 1.5) |
| $\Delta z_s^3 \times 10^2$ | Gaussian (0.0, 1.1) |
| $\Delta z_s^4 \times 10^2$ | Gaussian (0.0, 1.7) |
| Shear calibration | |
| $m^1 \times 10^2$ | Gaussian (-0.6, 0.9) |
| $m^2 \times 10^2$ | Gaussian (-2.0, 0.8) |
| $m^3 \times 10^2$ | Gaussian (-2.4, 0.8) |
| $m^4 \times 10^2$ | Gaussian (-3.7, 0.8) |



DES Y3: model choices

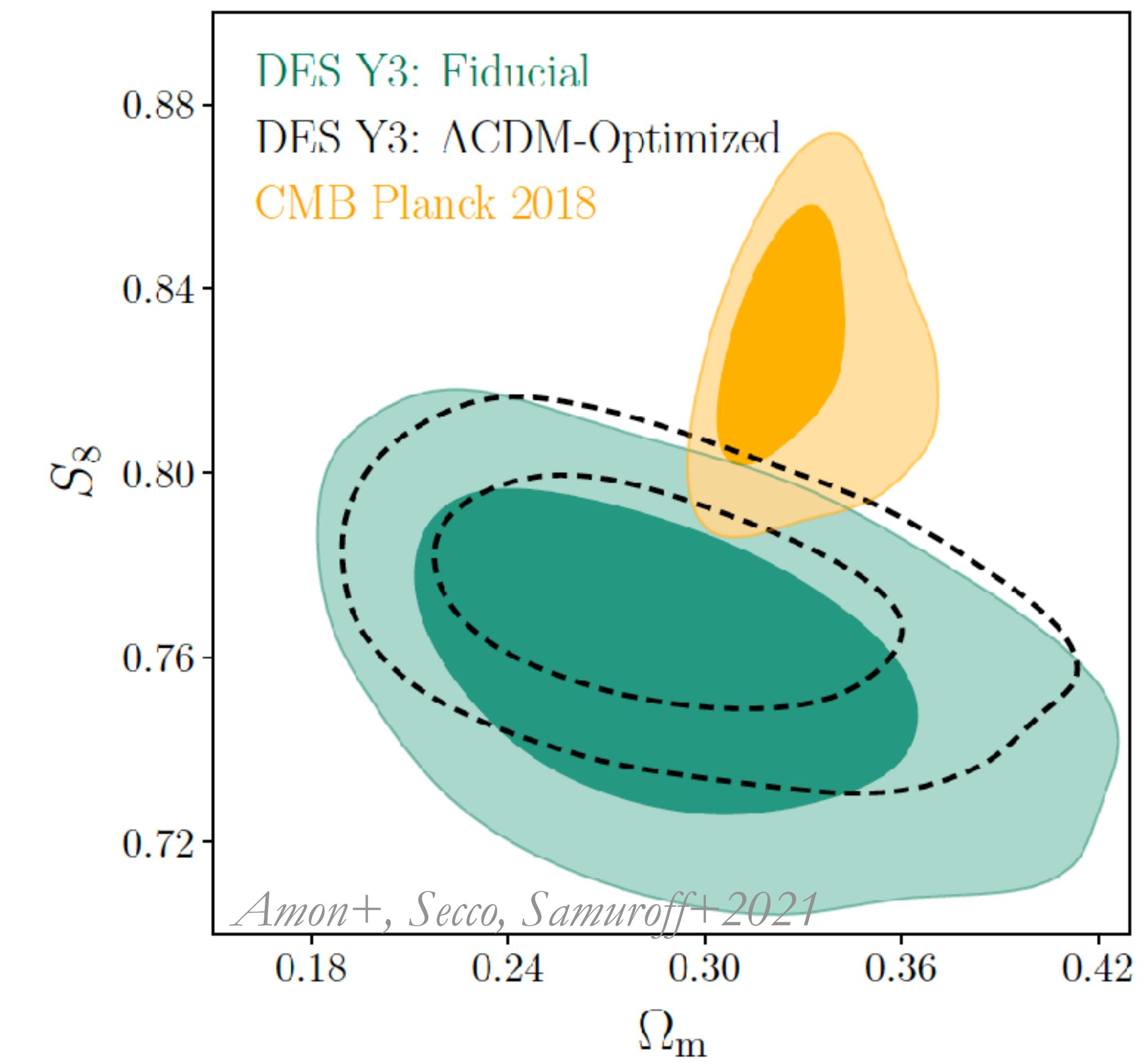
Intrinsic alignments:

Cosmology stable with the simpler NLA and NLA a_1 -only model.



Baryonic effects:

LCDM-Optimized analysis that uses more small-scale information gives consistent results with the Fiducial



DES Y3

— Data calibration

— Analysis choices

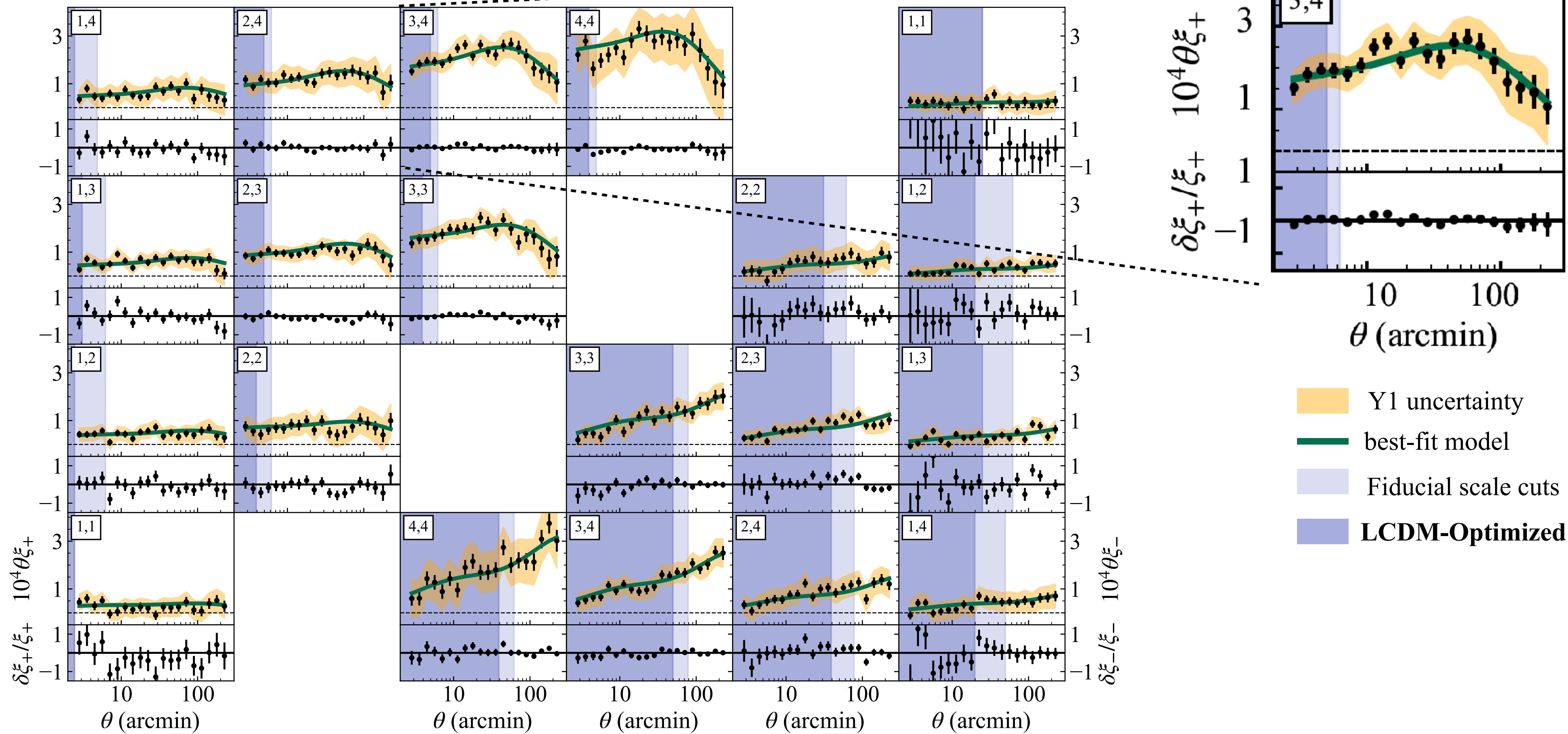
— The big picture

— Looking ahead



DES Y3: model choices

Amon+2021



DES Y3

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Data calibration

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Analysis choices

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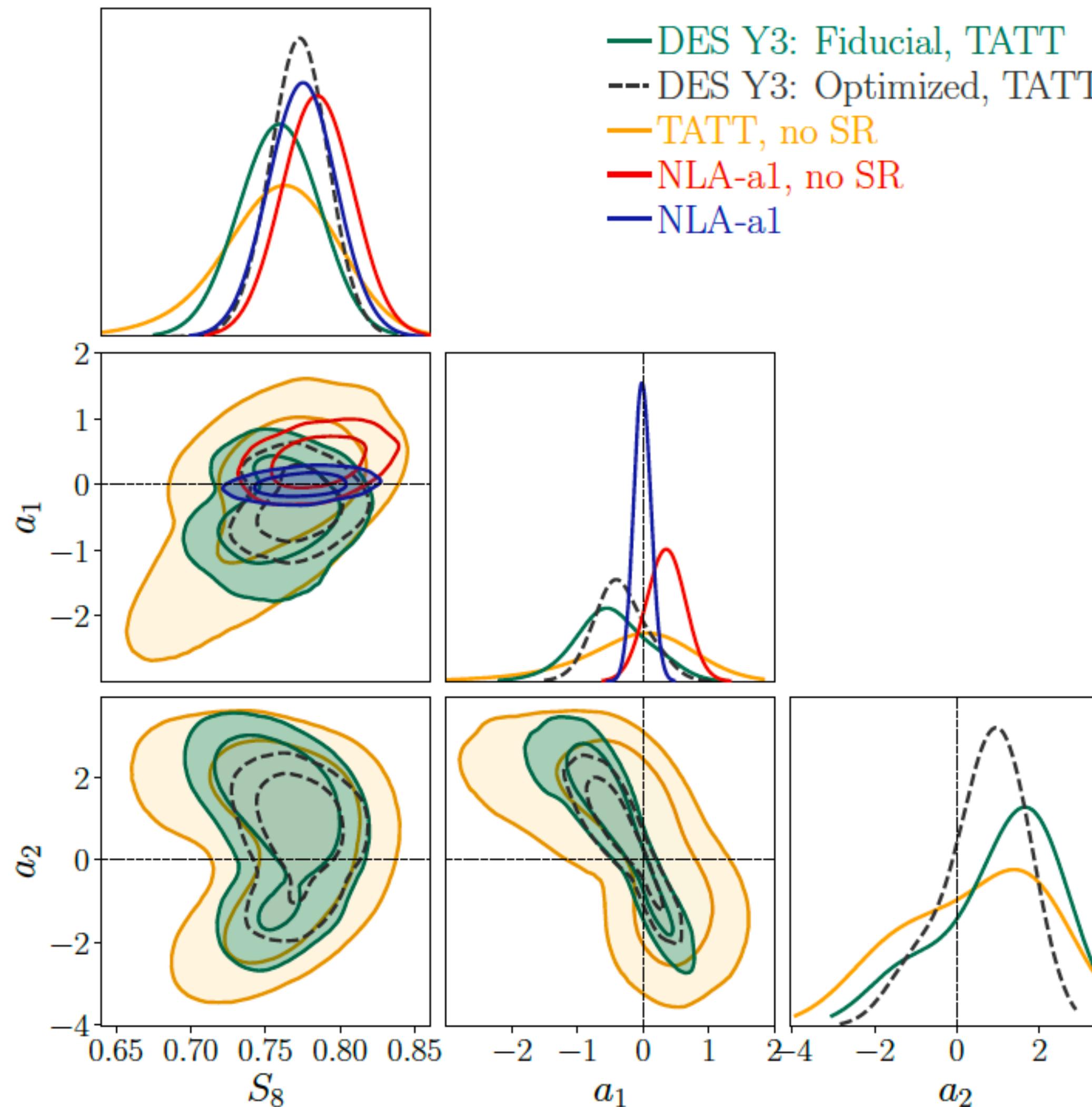
The big picture

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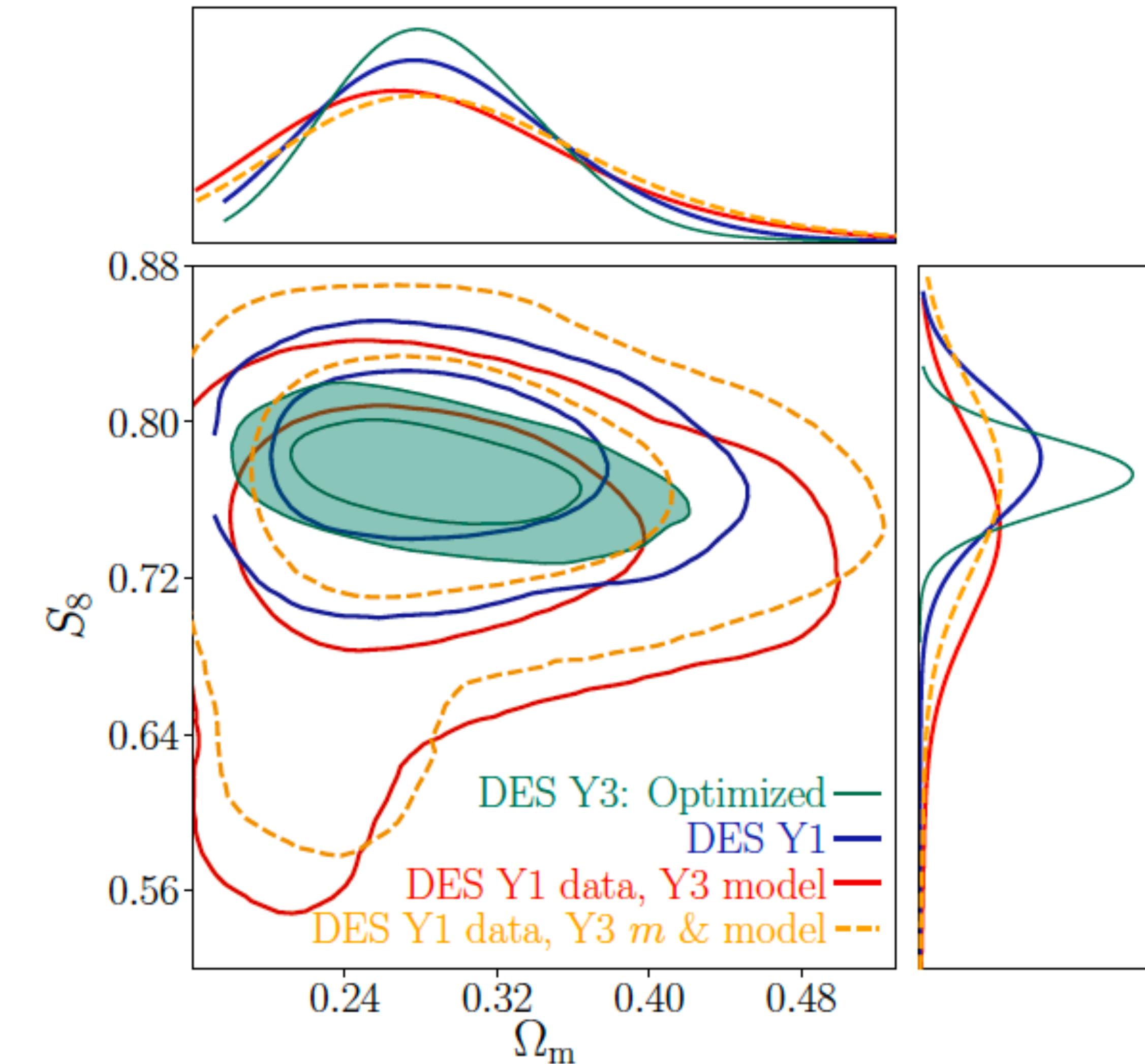
Looking ahead



DES Y3: conservative model choices



Amon, Gruen, Troxel, MacCrann, Dodelson+2021



DES Y3

— Data calibration

— Analysis choices

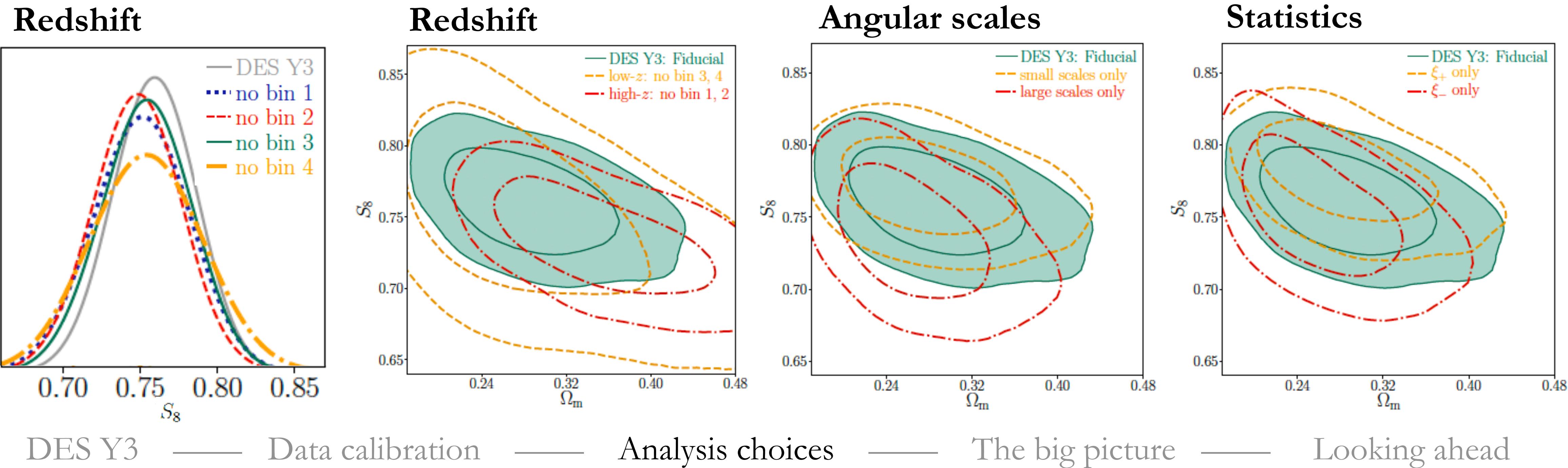
— The big picture

— Looking ahead



DES Y3: internally consistent

- Internal consistency as a model check
- Assess at the parameter level and with Posterior Predictive Distribution (PPD), with a threshold of model fit $p>0.01$ (methodology in *Doux, Baxter+2020*)
- DES Y3 cosmic shear consistent across redshift, angular scales and correlation function statistic
Amon, Gruen, Troxel, MacCrann, Dodelson+2021

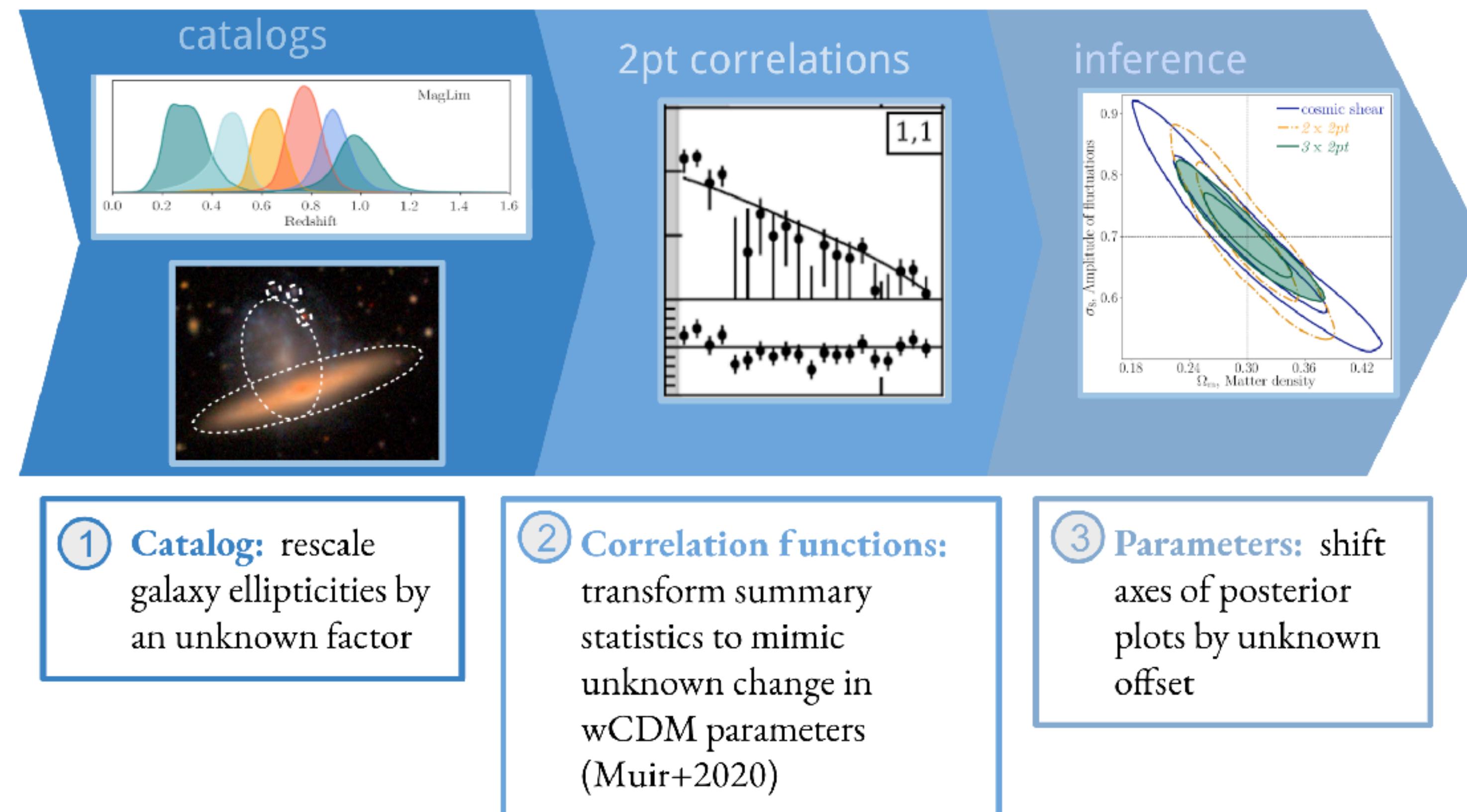




DES Y3: mitigating experimenter bias

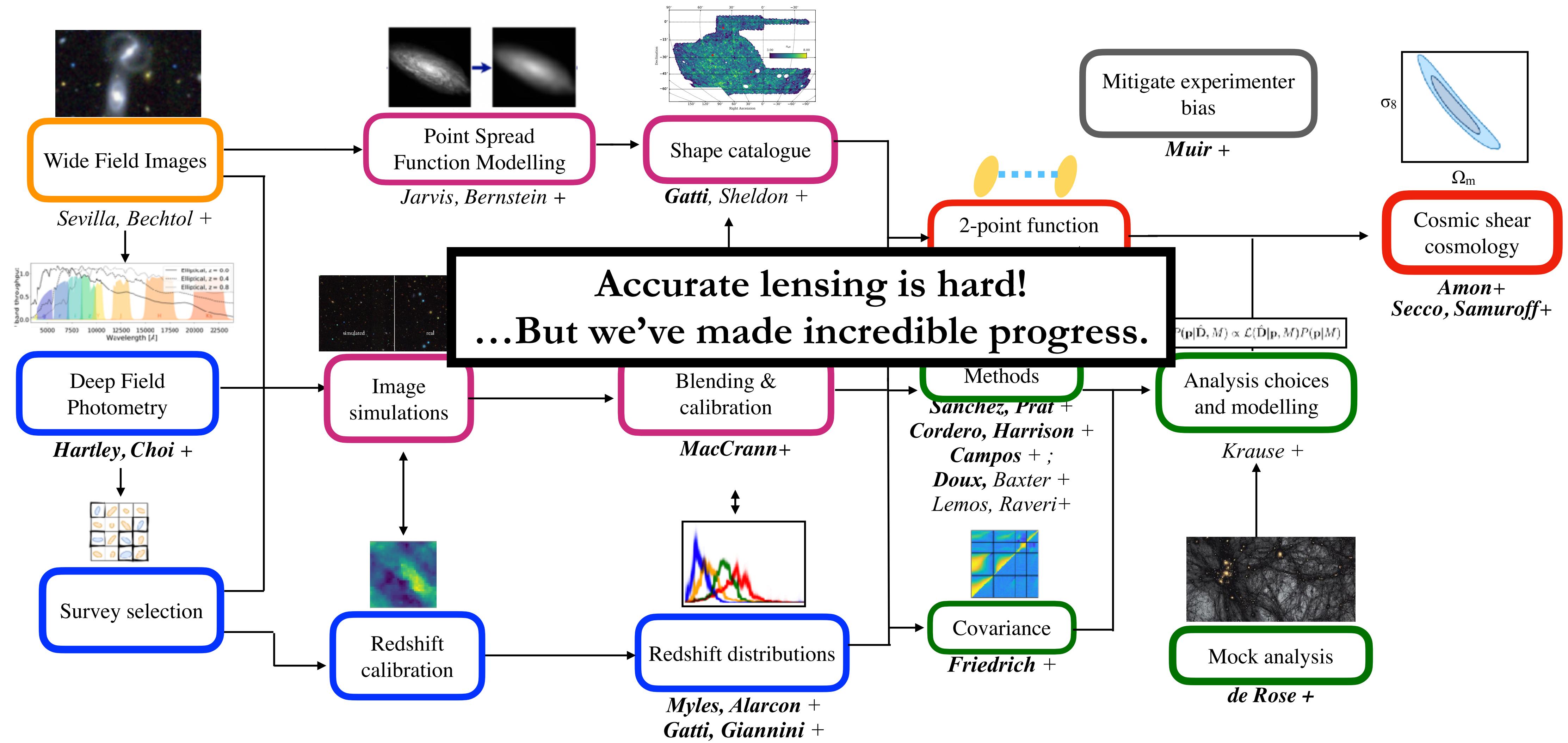
We minimize observer bias using a 3-stage blinding strategy. Before ‘unblinding’ we froze:

- Modeling choices & calibration priors
- Data vector measurements
- Planned list of robustness tests & combinations with external data





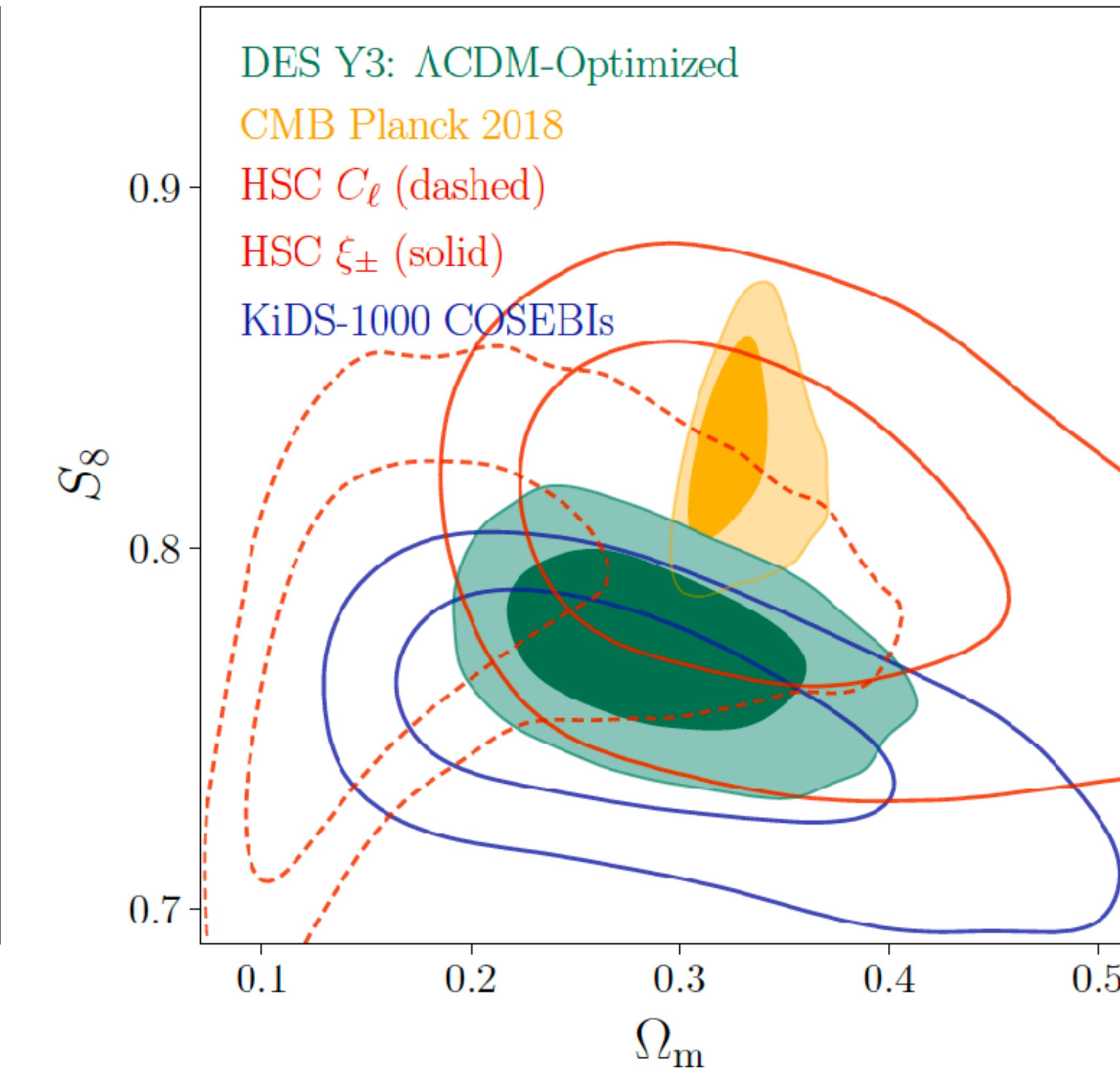
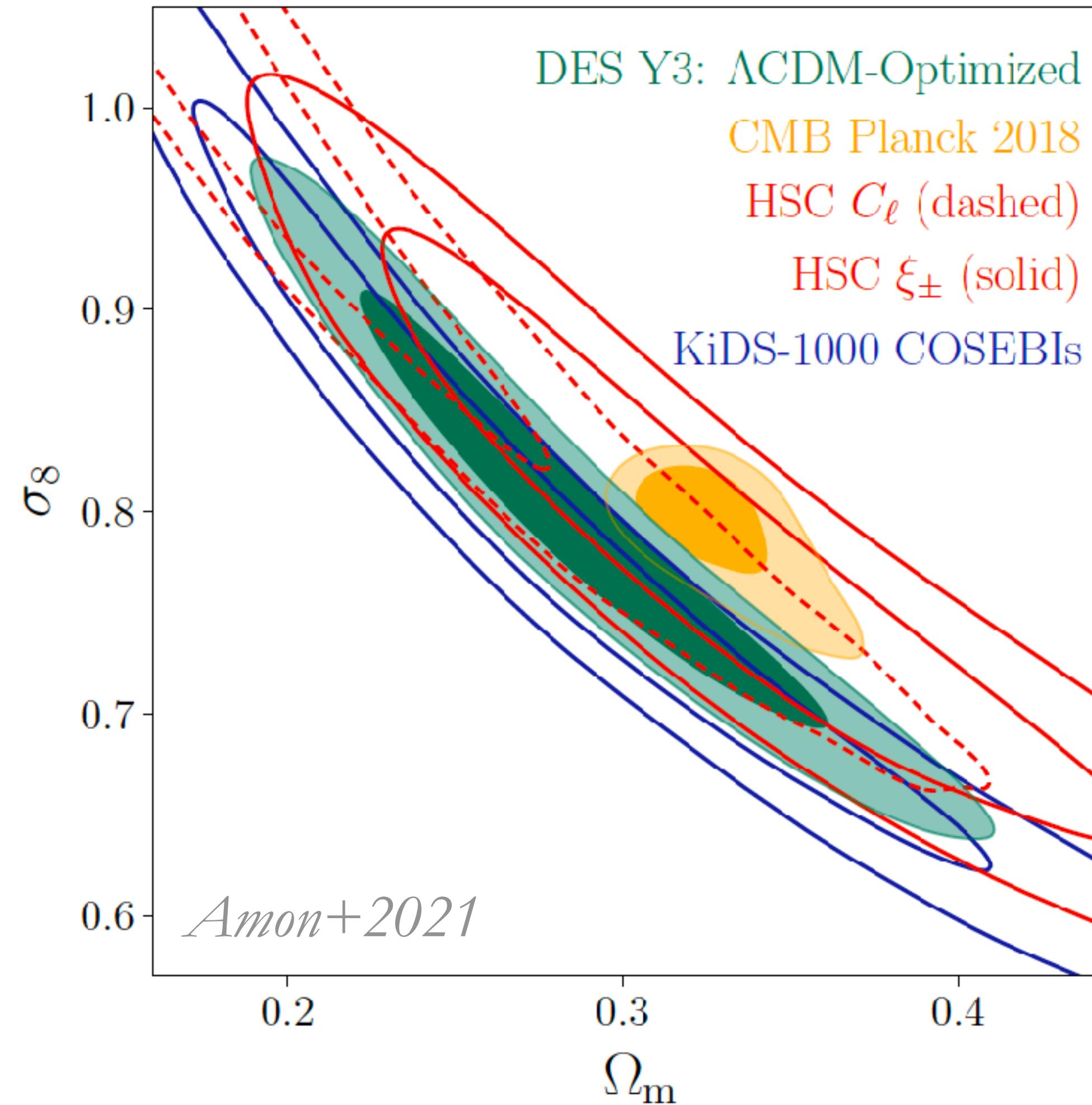
DES Y3 : pixels to cosmology





The lensing perspective on S_8 consistency

We find no significant evidence of inconsistency between **DES Y3 shear** and **Planck CMB** at $\sim 2\sigma$ or $p=0.05 > 0.01$.



DES Y3

— Data calibration

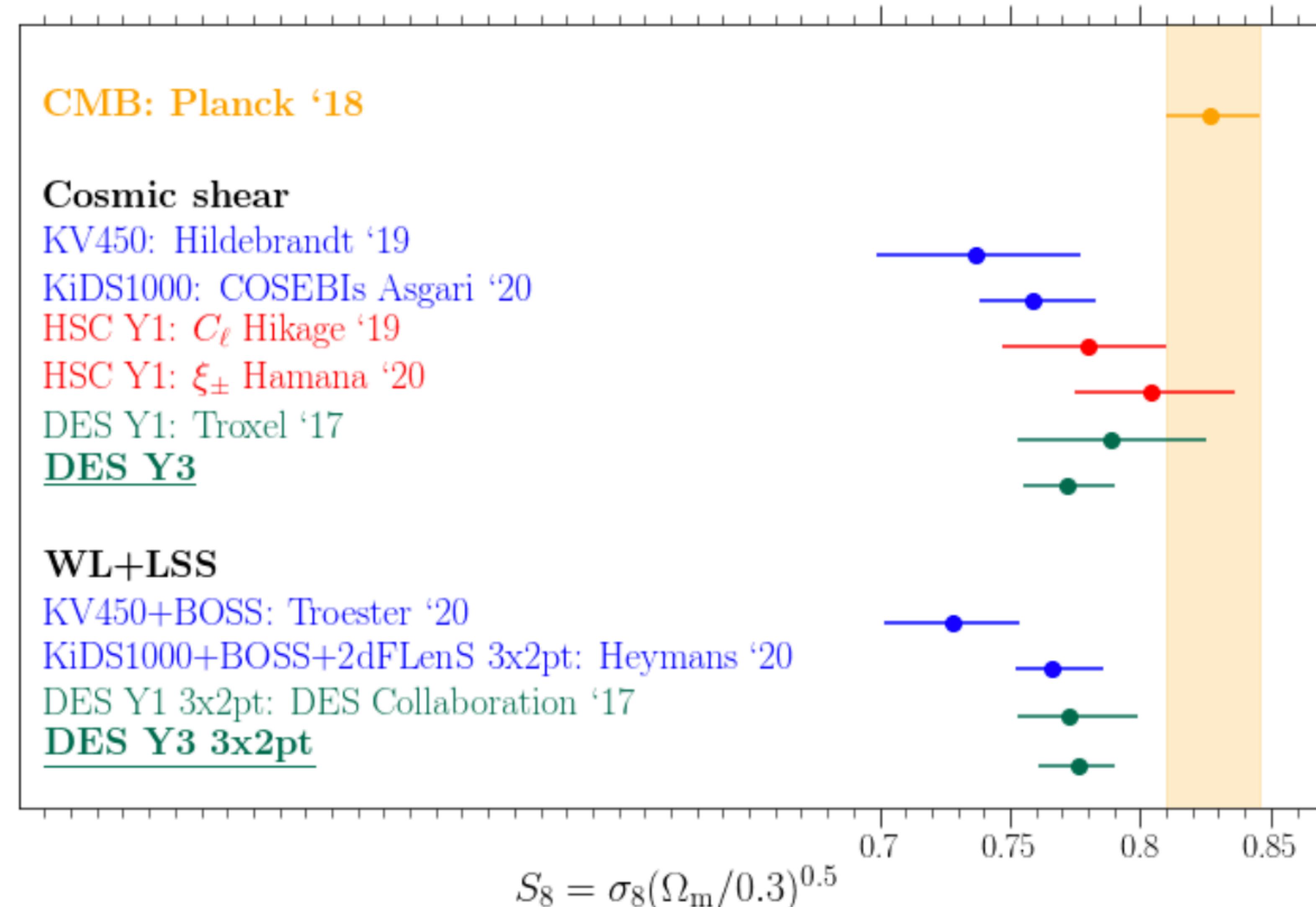
— Analysis choices

— The big picture

— Looking ahead



The lensing perspective on S_8 consistency



DES Y3

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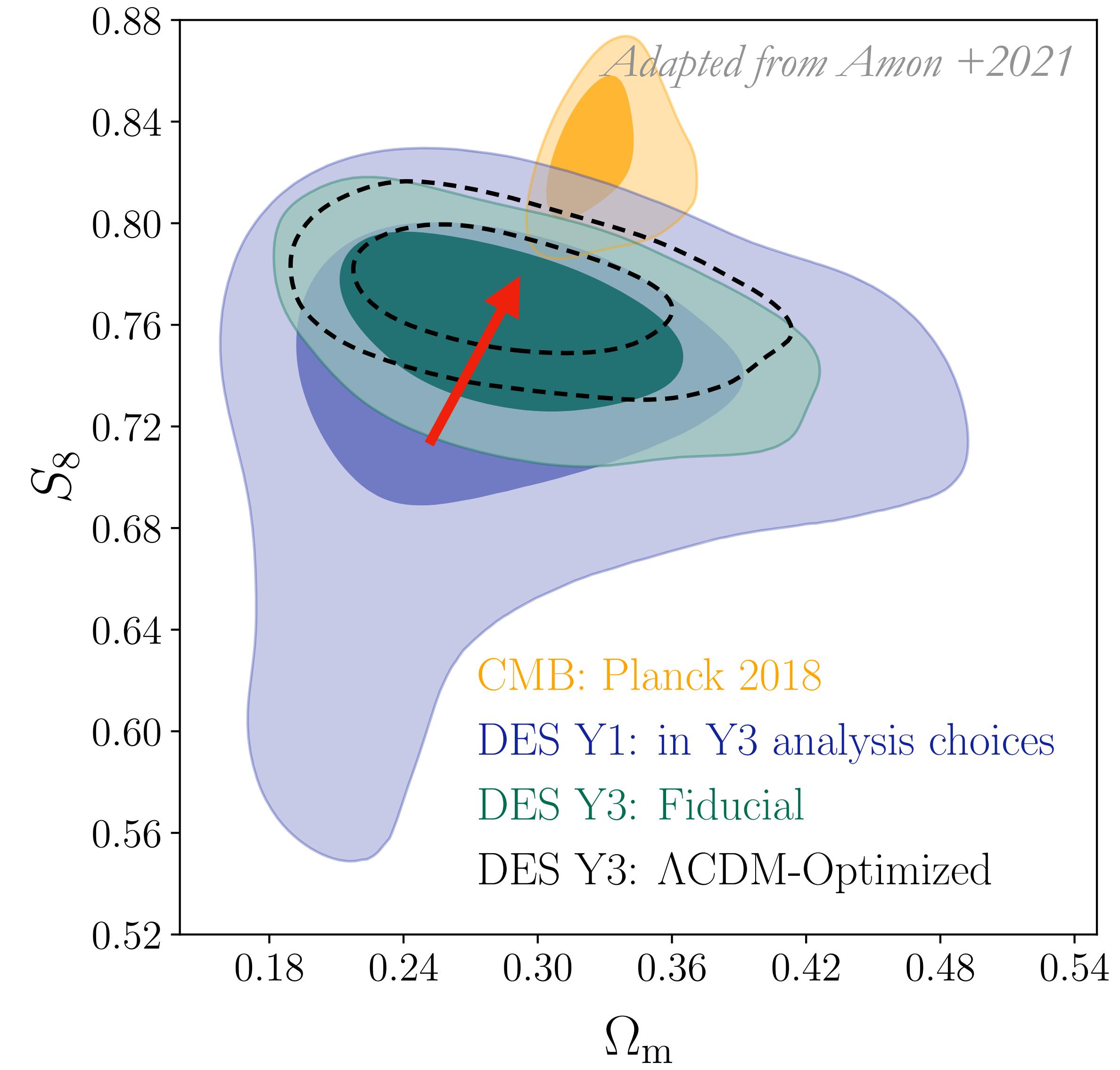


The lensing perspective on S_8 consistency

When comparing on equal footing with consistent analysis choices,

DES LCDM-Optimised represents a x1.5 improvement over the Fiducial Y3 and a x2 improvement over DES Y1.

—> Gains in statistical power have, in both cases, shown no increase in S_8 tension with planck.



DES Y3

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Data calibration

—

Analysis choices

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The big picture

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Looking ahead



Looking ahead for cosmic shear

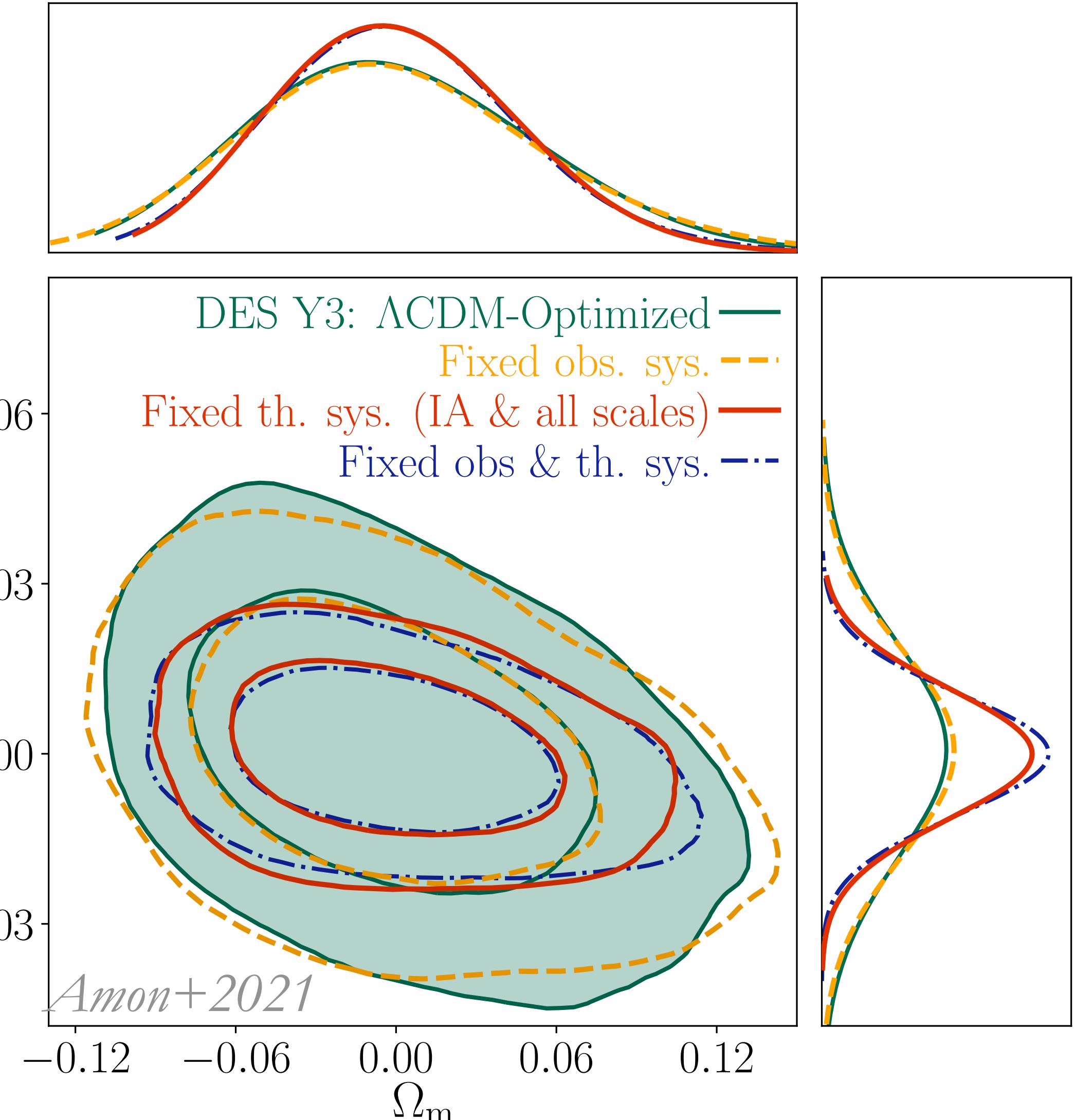
Tested the impact of fixing uncertainties in the analysis to their best fit values:

Redshift and shear calibration uncertainties negligible in the Y3 analysis.*

Fixing theoretical systematics, both intrinsic alignment modelling and small-scale baryonic effects (by using all scales) improves the S8 constraining power by ~ 2 .**

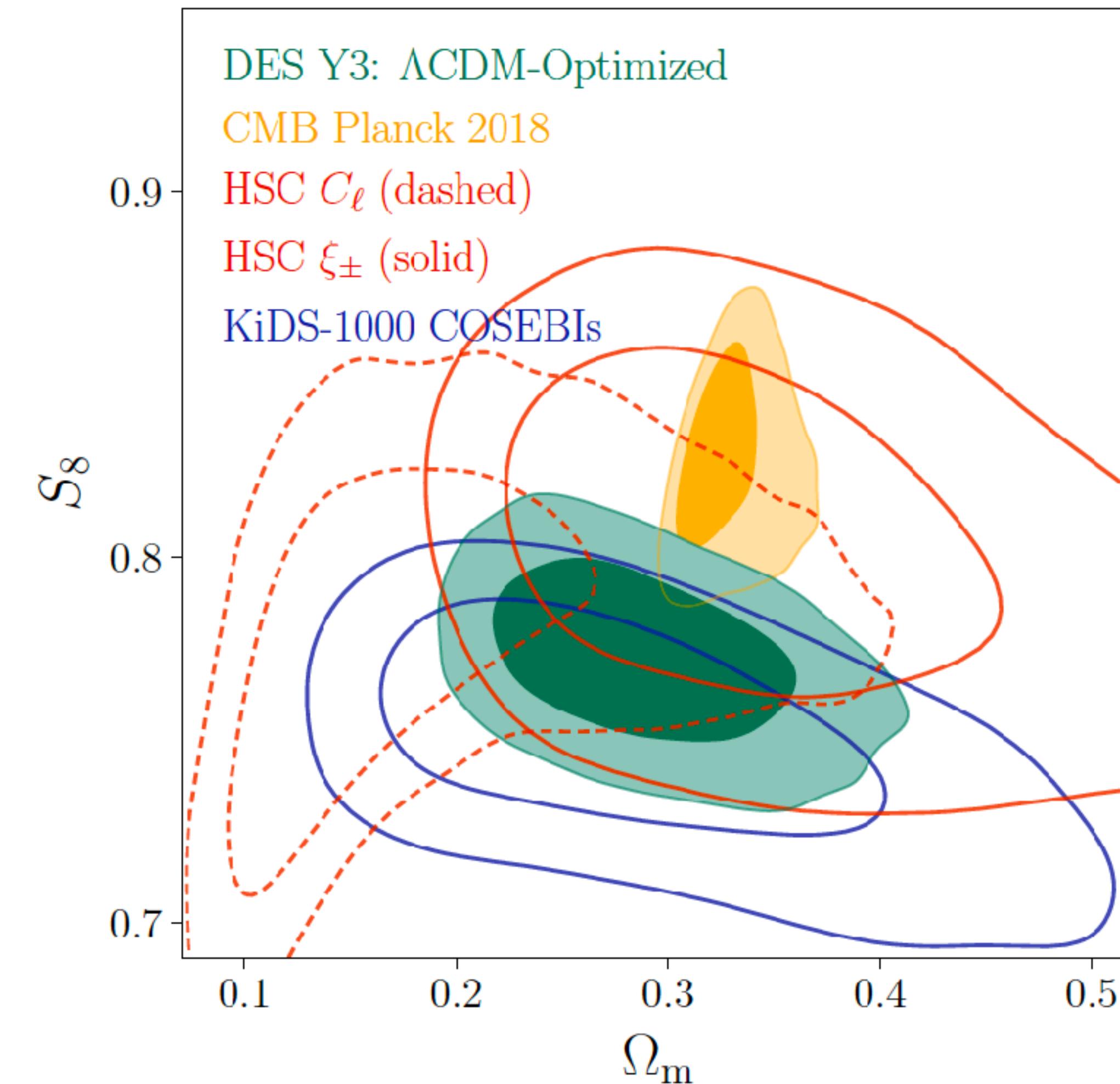
* Maintaining their accuracy is still crucial and challenging

** Future decreases in measurement noise may not lead to concomitant decreases in cosmological uncertainties.





Looking ahead for cosmic shear : KiDS + DES !





Novel advancements built to utilize the statistical power of the DES Y3 data

DES Y3 cosmology stable to data calibration and modelling choices

Y3 finds a slightly higher clustering amplitude and matter density than in DES Y1, and a factor of ~ 2 improvement in power

and no significant evidence for inconsistency in Λ CDM between DES and *Planck*



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#Darkbites