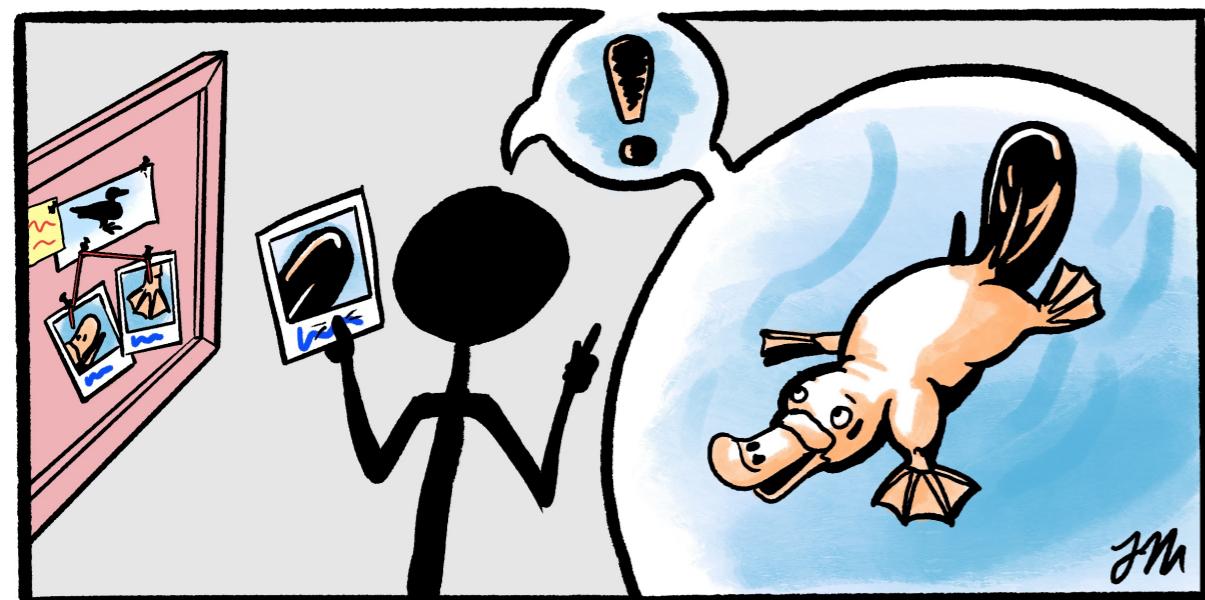
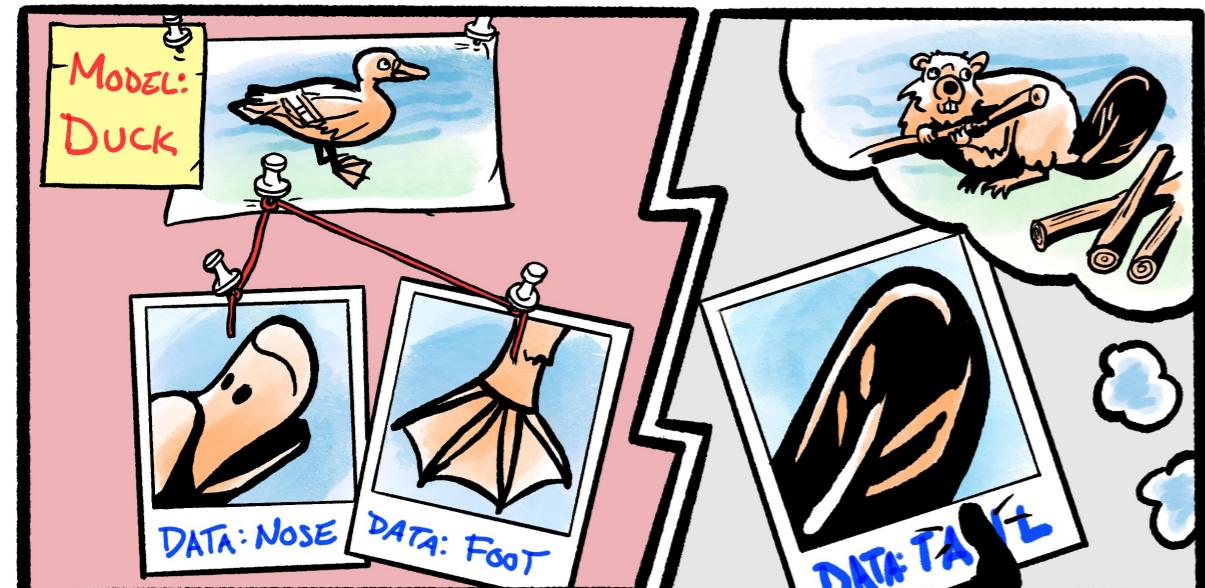




Cosmological Tensions (and How to Find Them)



Darkbites

© 2020 Jessie Muir

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University College London -
University of Sussex
07-05-2021
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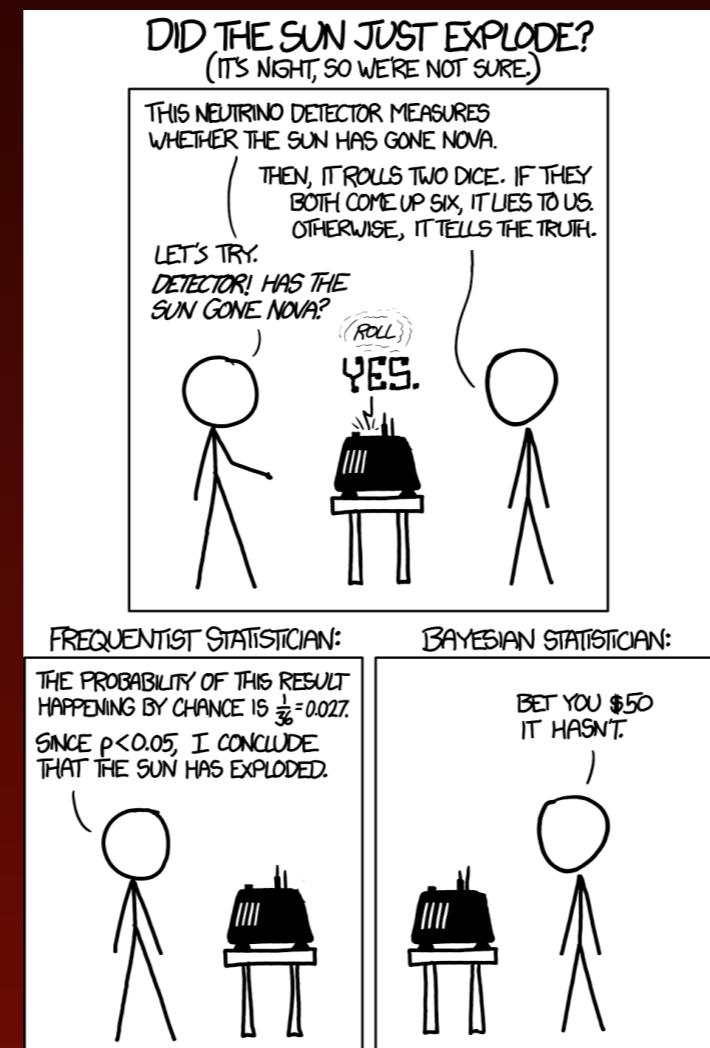
Text: Andresa Campos
[@AndresaCampos](https://twitter.com/AndresaCampos)

Illustration: Jessie Muir
[@jlynnmuir](https://twitter.com/jlynnmuir)

In collaboration with:
George Efstathiou,
Will Handley, Ofer
Lahav, Benjamin
Joachimi, Antony
Lewis and others



Bayesian Statistics



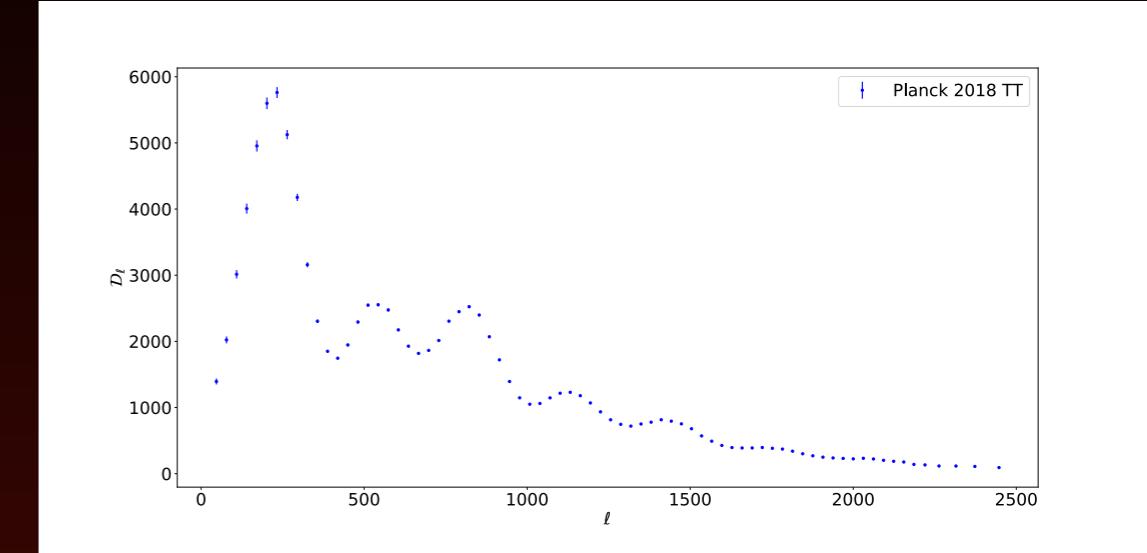
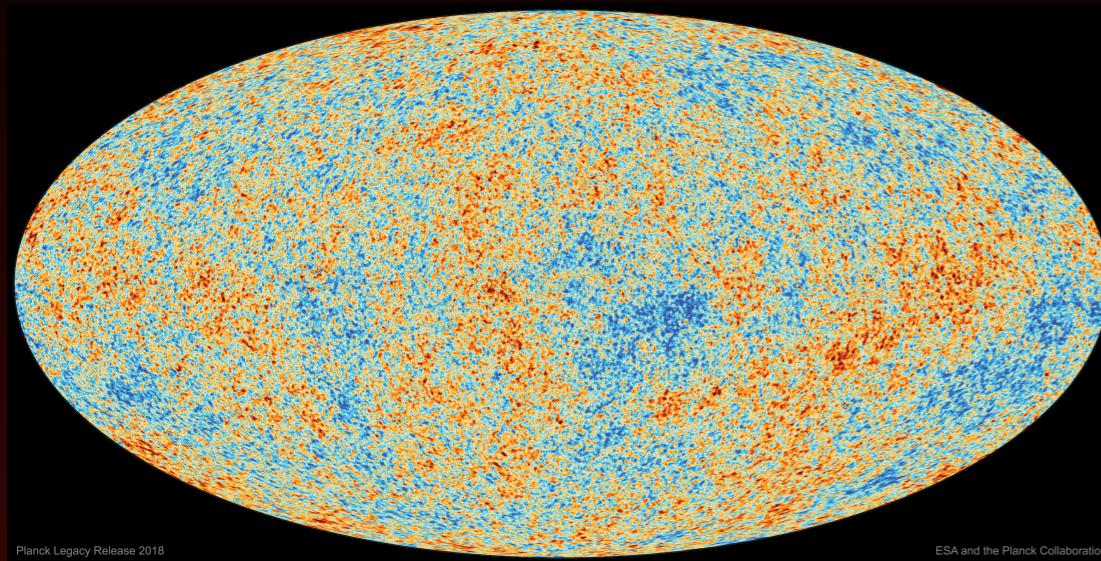


Bayes' Theorem

$$P(\theta|D, M) = \frac{P(D|\theta, M) \cdot P(\theta|M)}{P(D|M)}$$

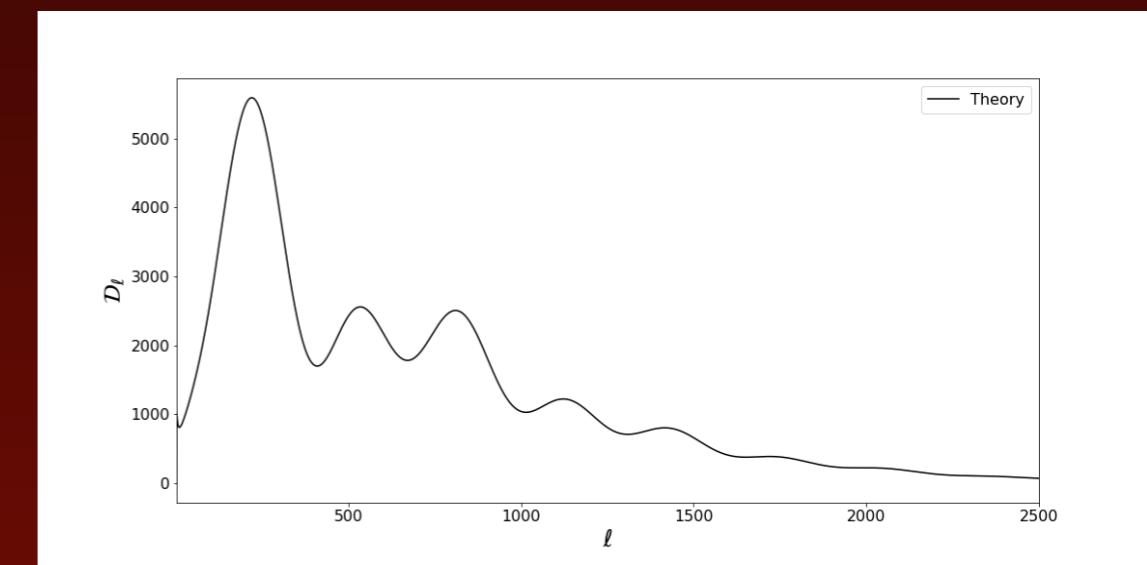
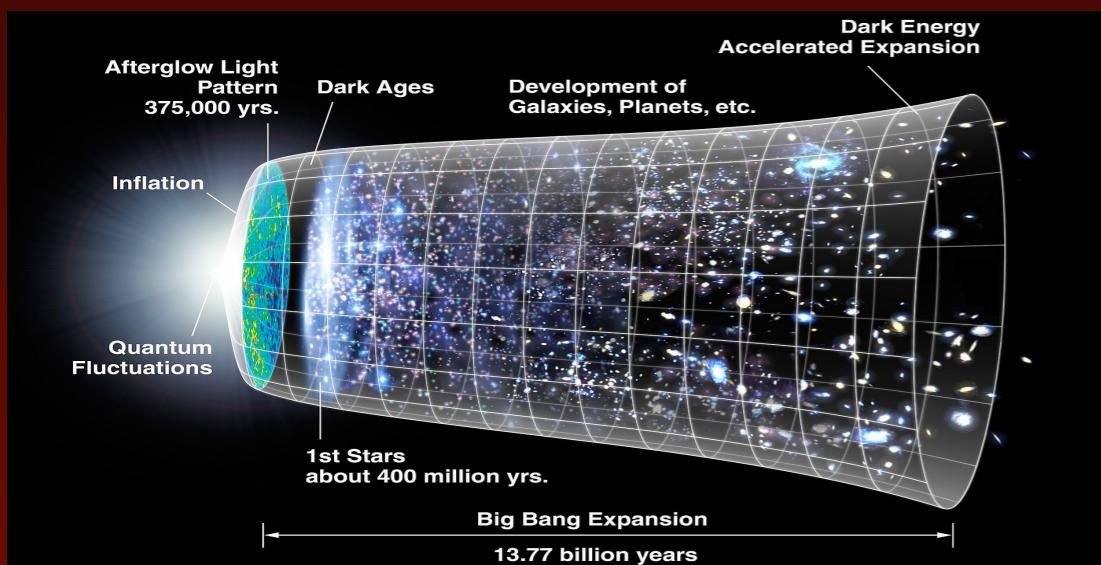
$$\mathcal{P} = \frac{\mathcal{L} \times \Pi}{Z}$$

- **θ** : *Parameters*
- **D** : *Data*
- **M** : *Model*



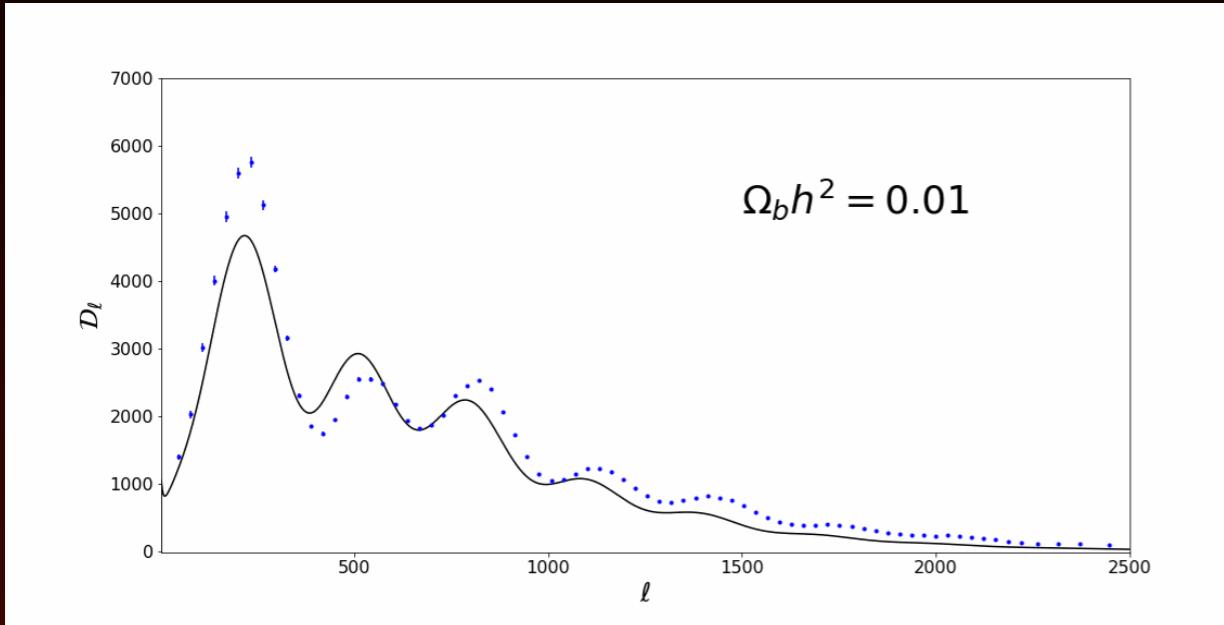
θ, M

$$\mathcal{L} \equiv P(D \mid \theta, M)$$

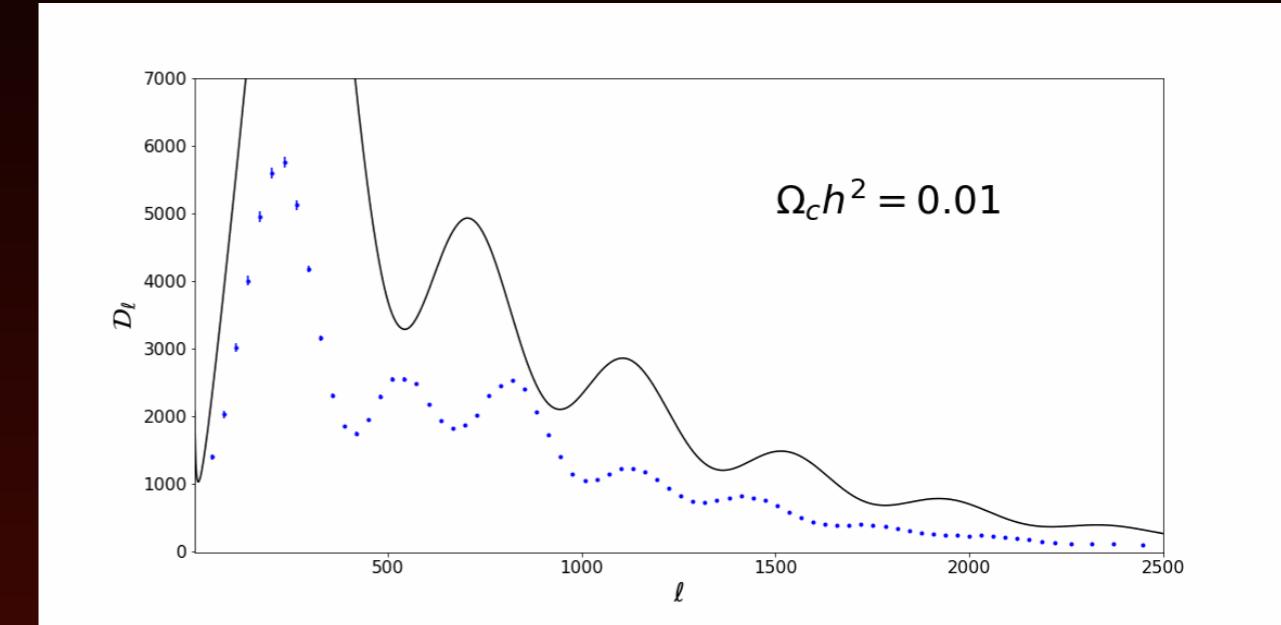




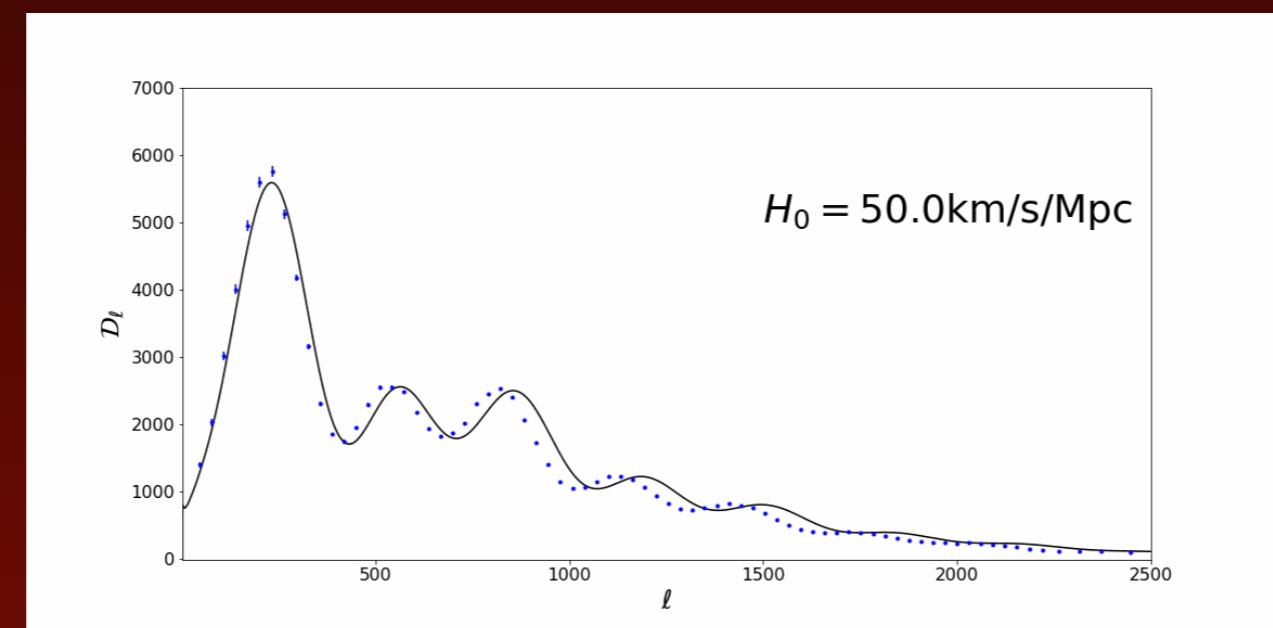
How much dark energy?



How much dark matter?



How fast does it expand?





Three types of problem

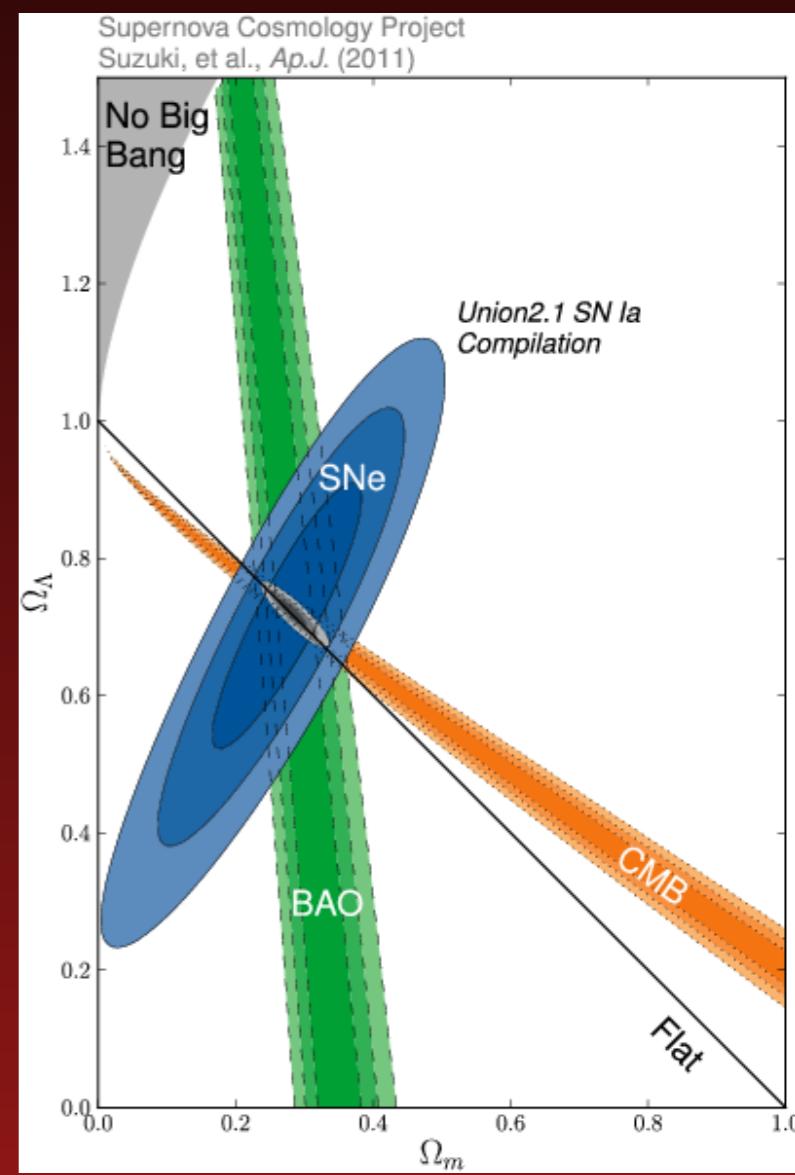
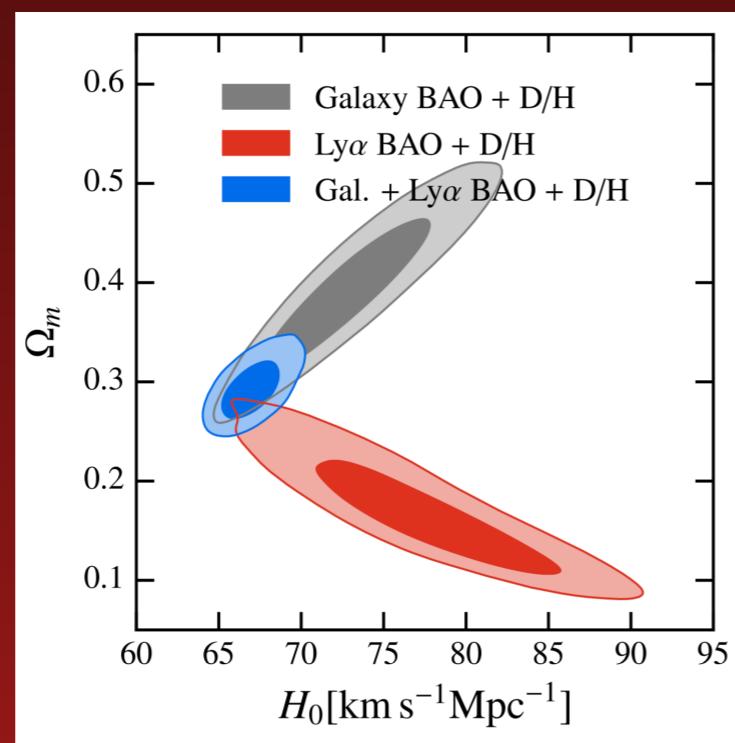
- Parameter estimation
- Model comparison
- Dataset comparison

‘Tension’



Why is tension important

- We can only combine data Sets that are **CONSISTENT**. Data set combinations are crucial to break degeneracies.
- If two data sets are in tension, there are two explanations: One (or both) data sets are wrong, or the underlying model is wrong.
- We need a method to accurately quantify tension!

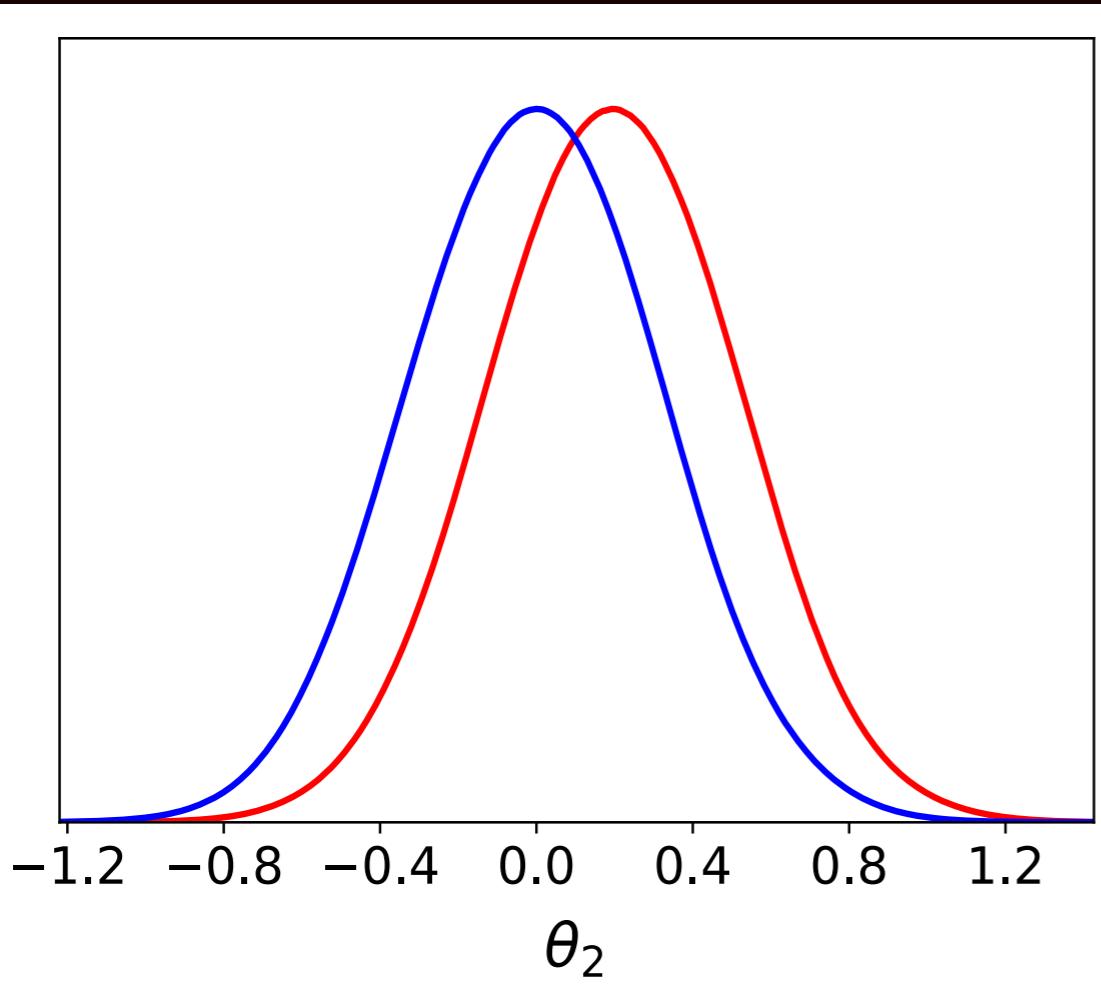




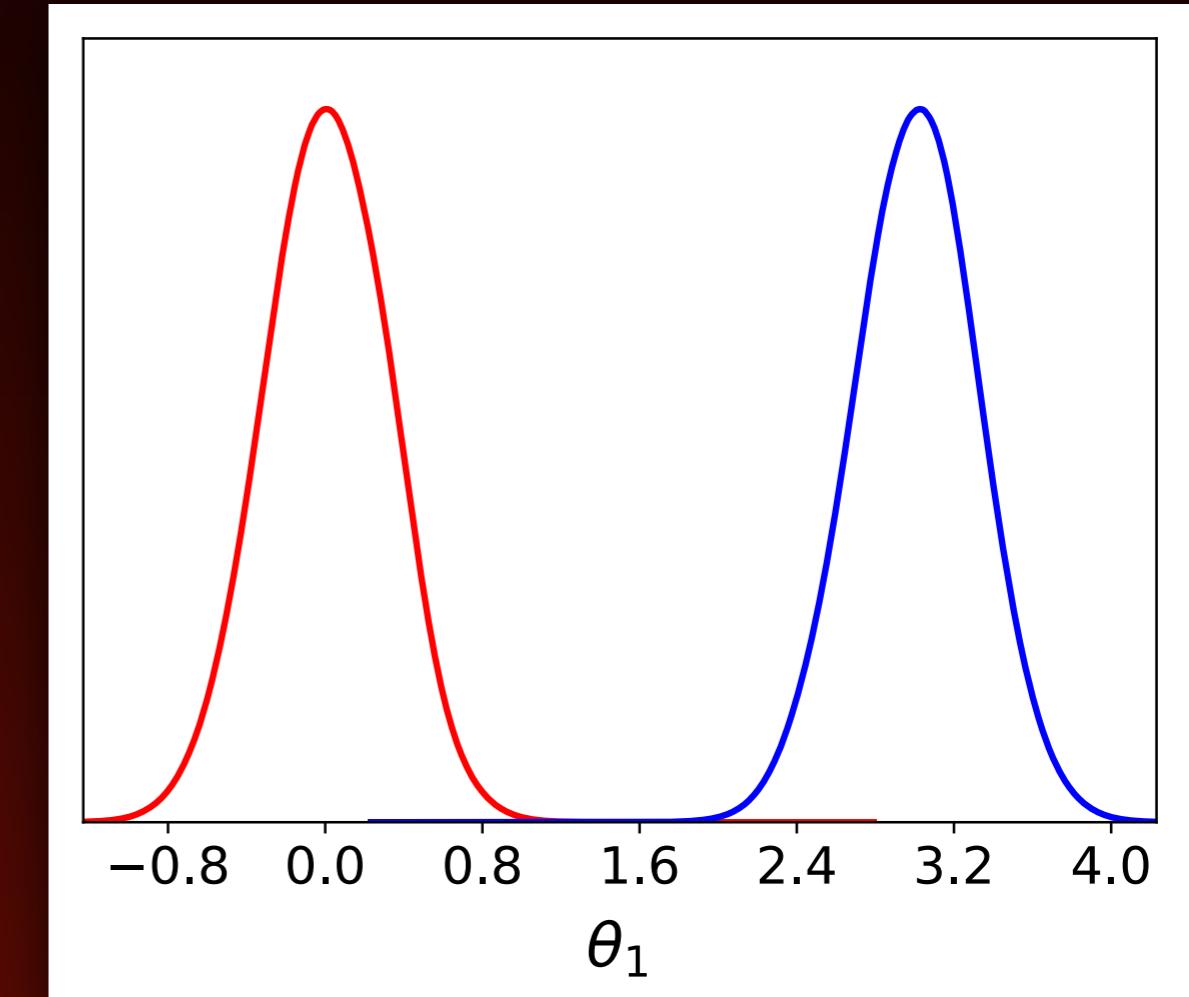
Why is Data Set Comparison Non-Trivial?



Trivial?



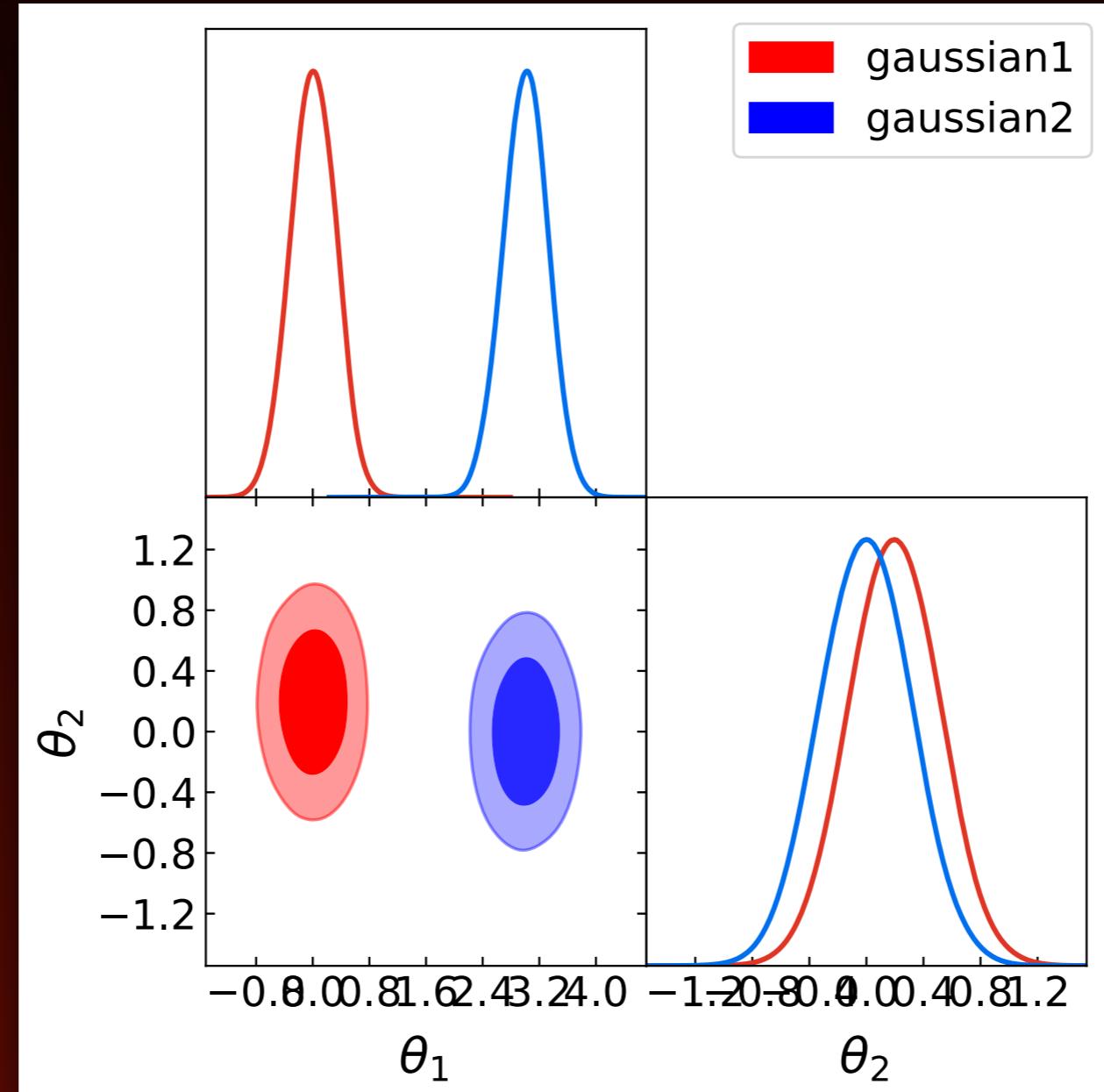
Consistent



Inconsistent

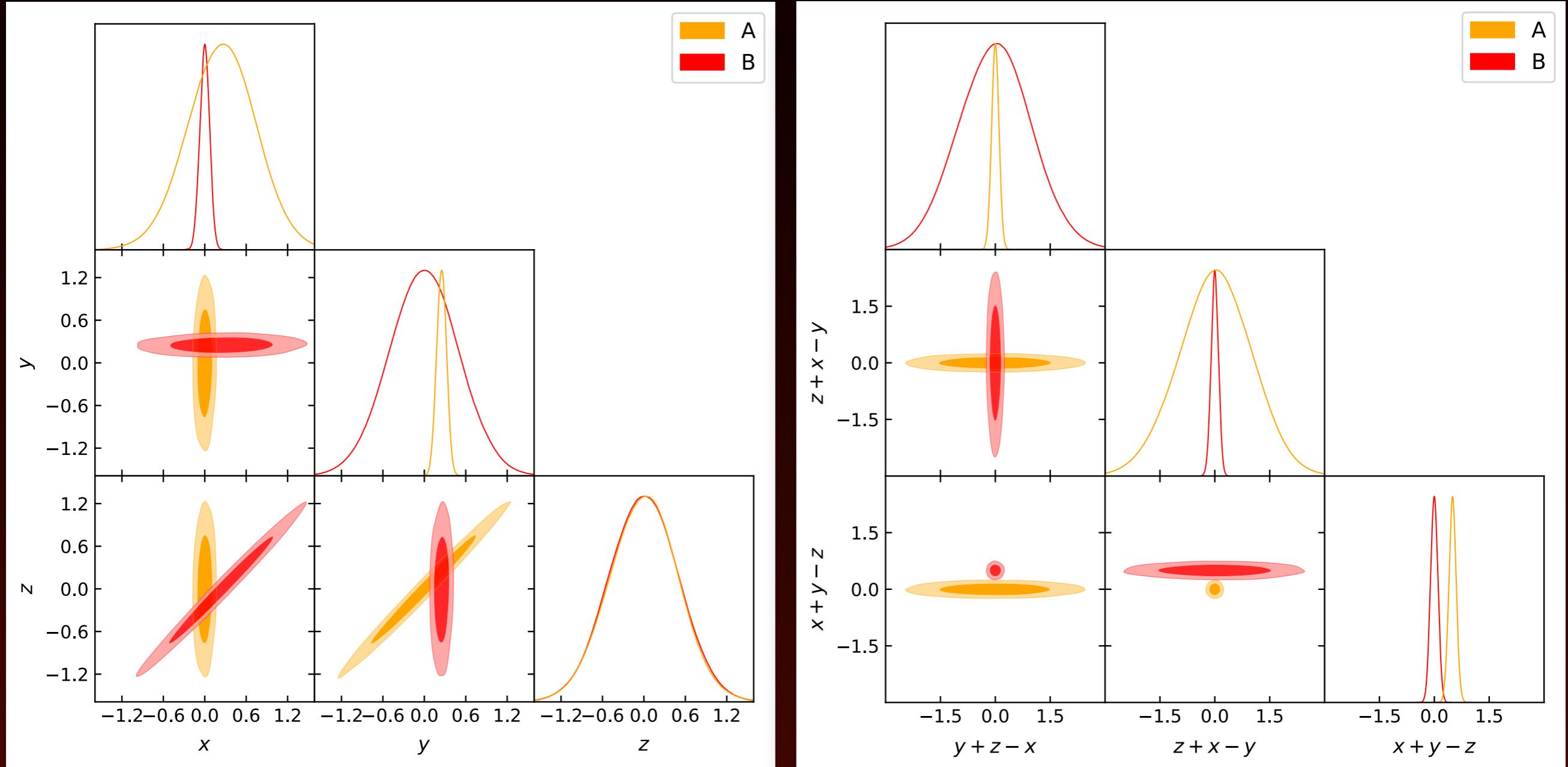


Trivial?





Trivial?





The Bayes Ratio



Bayes Ratio

Bayesian evidence as a tool for comparing datasets

Phil Marshall

Kavli Institute for Particle Astrophysics and Cosmology, Stanford University, USA

Nutan Rajguru

Astrophysics Group, Cavendish Laboratory, Madingley Road, Cambridge, UK

Anže Slosar

Faculty of Mathematics and Physics, University of Ljubljana, Slovenia

(Dated: February 2, 2008)

We introduce a new conservative test for quantifying the consistency of two or more datasets. The test is based on the Bayesian answer to the question, “How much more probable is it that all my data were generated from the same model system than if each dataset were generated from an independent set of model parameters?”. We make explicit the connection between evidence ratios and the differences in peak chi-squared more cheaply calculated. Calculating evidence for data (WMAP, ACBAR, CBI, VSA), SDSS and concordance is favoured and the tightening of constraints justified.

Probability that both datasets come
from **THE SAME** Universe

$$R \equiv \frac{z_{AB}}{z_A \times z_B}$$

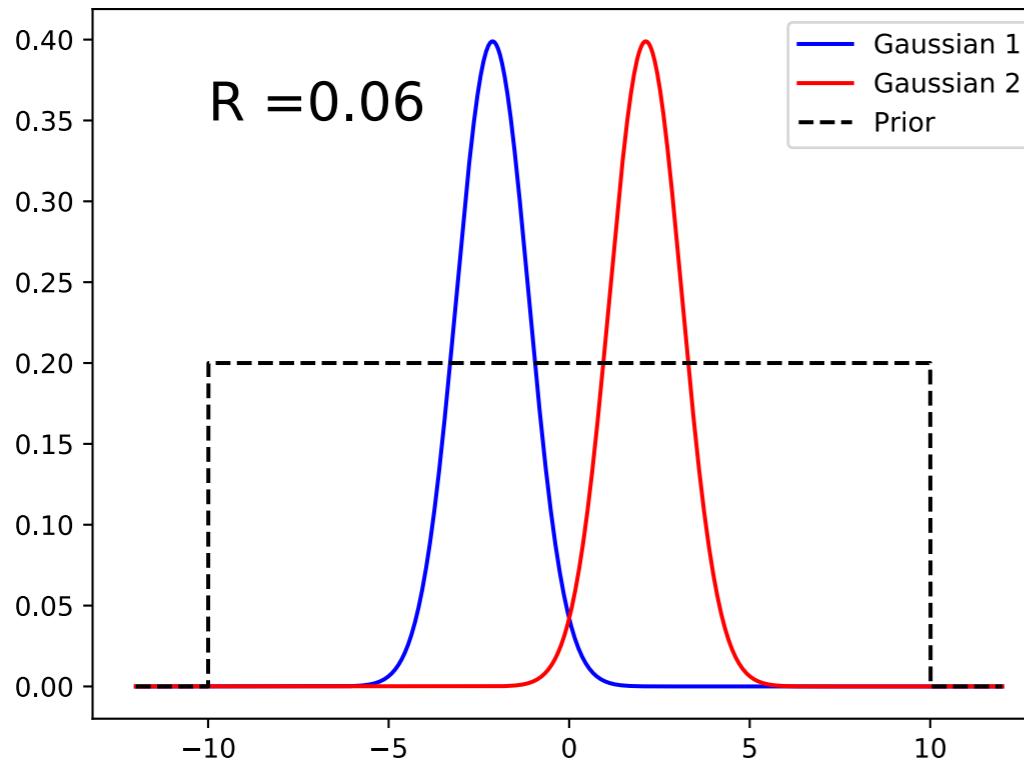
r
Probability that both
datasets come
DIFFERENT Universes



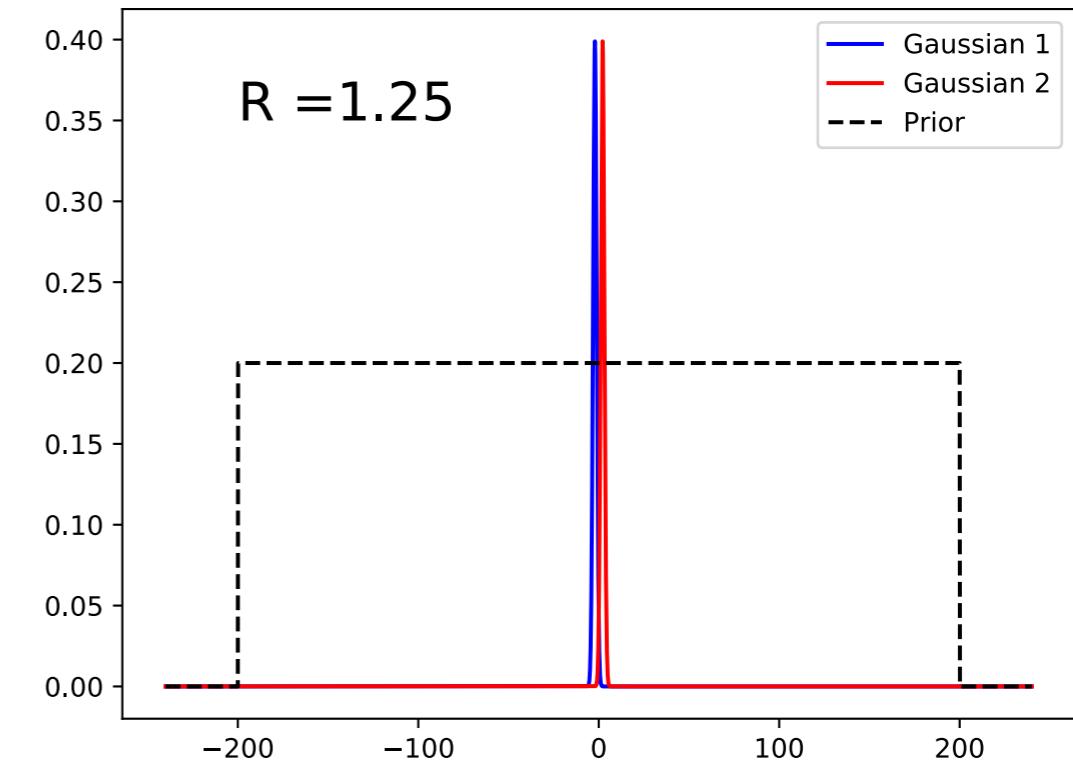
Bayes Ratio

Toy example: 1D Gaussians:

$$R \equiv \frac{\mathcal{Z}_{AB}}{\mathcal{Z}_A \times \mathcal{Z}_B} \propto V_\pi$$



Prior: -10, 10 -> **Discordant**

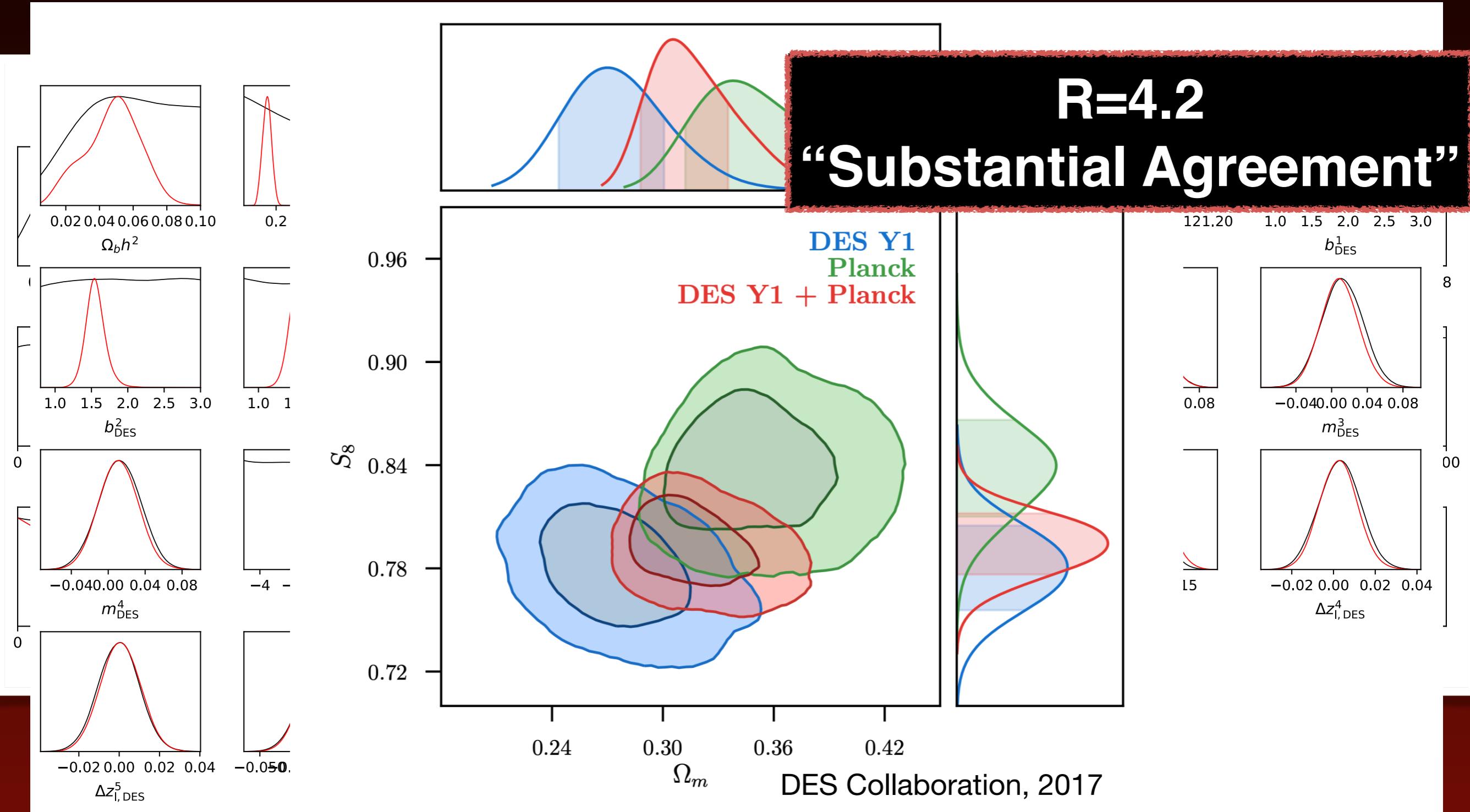


Prior: -200, 200 -> **Concordant**



Bayes Ratio

Is this a problem in Cosmology?





The ‘Suspiciousness’

In collaboration with:
Will Handley



Proposition 1

Proposition 1:

*If there are **any** physically reasonable priors which render R significantly less than 1, then as Bayesians we should consider these datasets **in tension**.*

Handley & PL, 2019, arXiv: 1902.04029
[10.1103/PhysRevD.100.043504](https://doi.org/10.1103/PhysRevD.100.043504)



Suspiciousness

We want a method that

- Is formed of fully Bayesian quantities.
- Is independent of choice of parameterisation.
- **Has an intuitive interpretation**
- **Does not depend on prior volume**



Suspiciousness

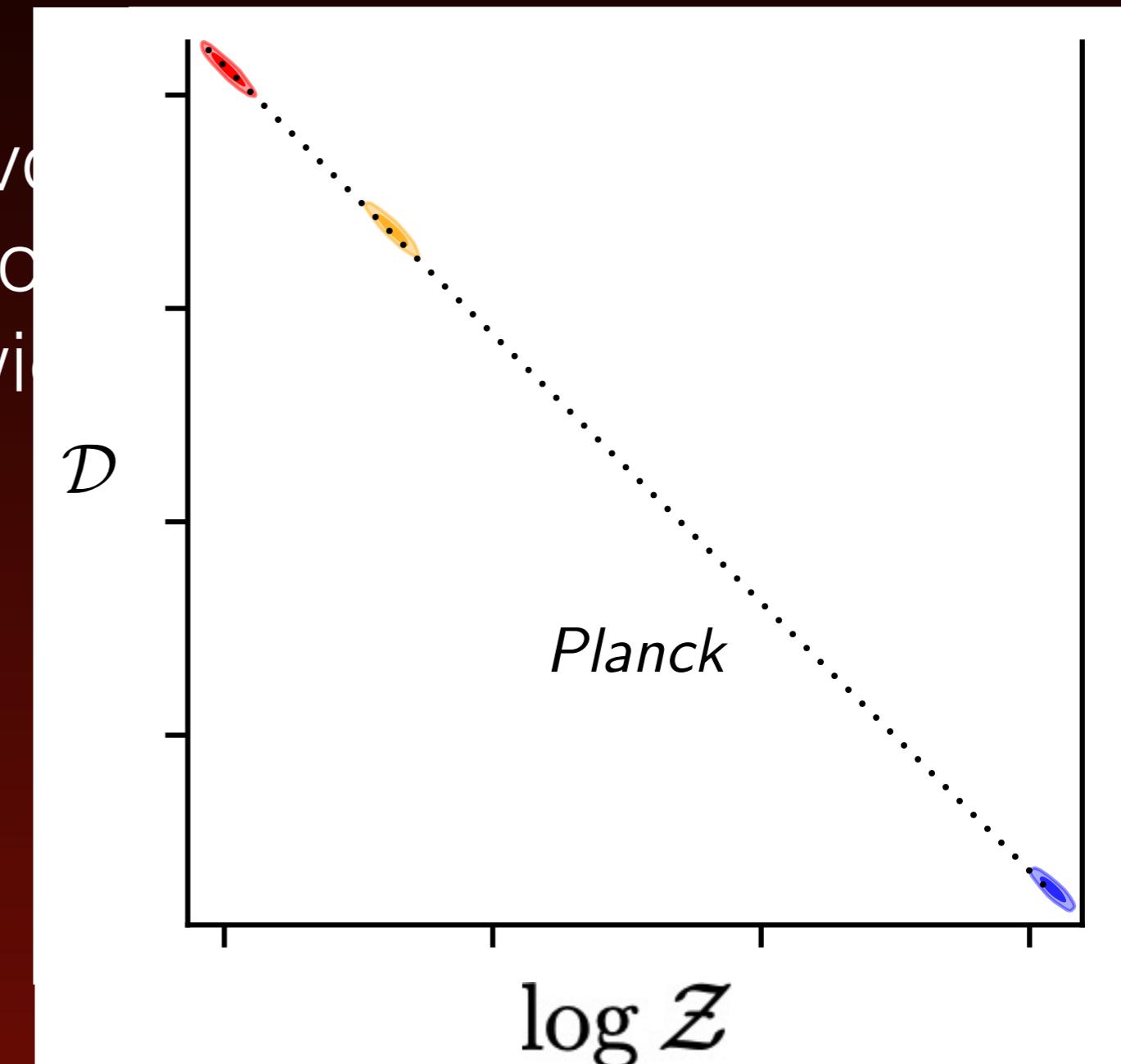
What part of the Bayes Ratio carries the ‘prior volume dependence’?

i.e. if I double the prior volume of the space of many possible states, so the prior becomes twice as large

Kullback-Leibler Divergence

$$\mathcal{D} \equiv \int d\theta \mathcal{P} \log \left(\frac{\mathcal{P}}{\Pi} \right)$$

Kullback, Leibler, 1951
[doi:10.1214/aoms/1177729694](https://doi.org/10.1214/aoms/1177729694)





Suspiciousness

So a part of the **BAYES RATIO (R)**:

$$\log R = \log \mathcal{Z}_{AB} - \log \mathcal{Z}_A - \log \mathcal{Z}_B$$

Encloses its dependence on the prior volume. We call this part the **INFORMATION (I)**:

$$\log I = \mathcal{D}_A + \mathcal{D}_B - \mathcal{D}_{AB}$$

The part of R that is left, is what we call the **SUSPICIOUSNESS (S)**:

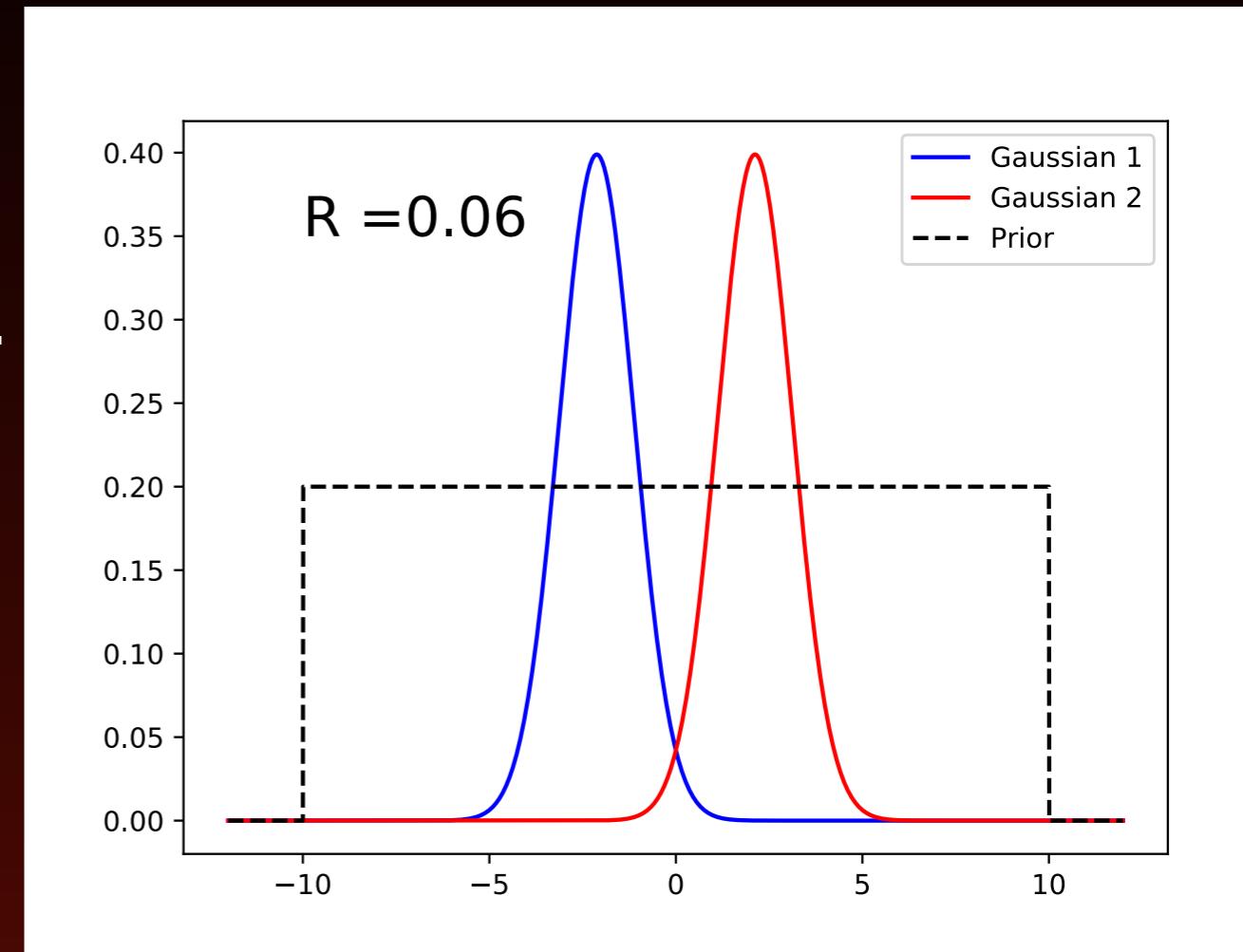
$$\log R = \log I + \log S$$



Suspiciousness

For **Gaussian Likelihoods**,
the Suspiciousness follows
a **chi-squared distribution**.

Therefore we can assign a
tension probability, and
interpret the result with a
'number of sigma' tension.



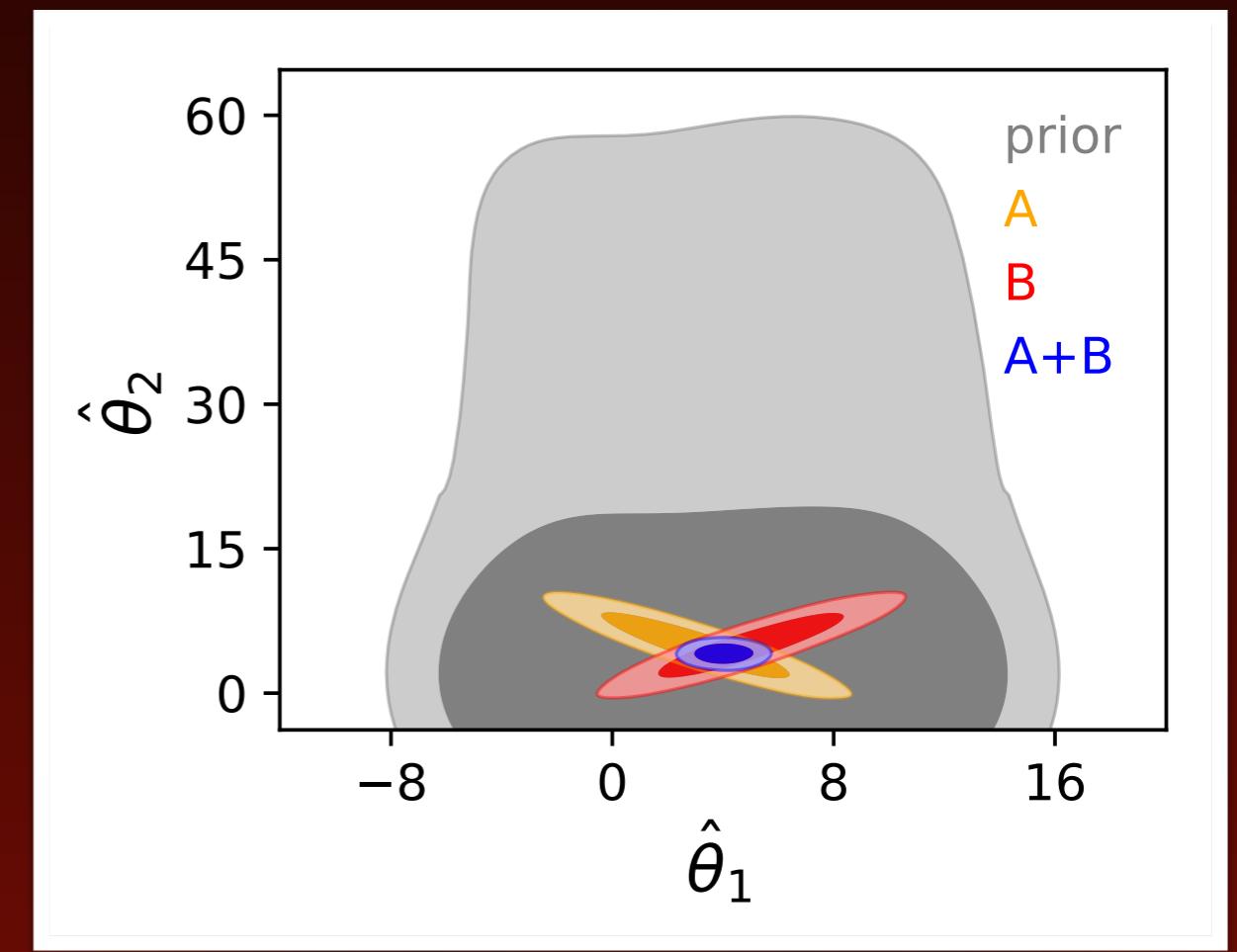
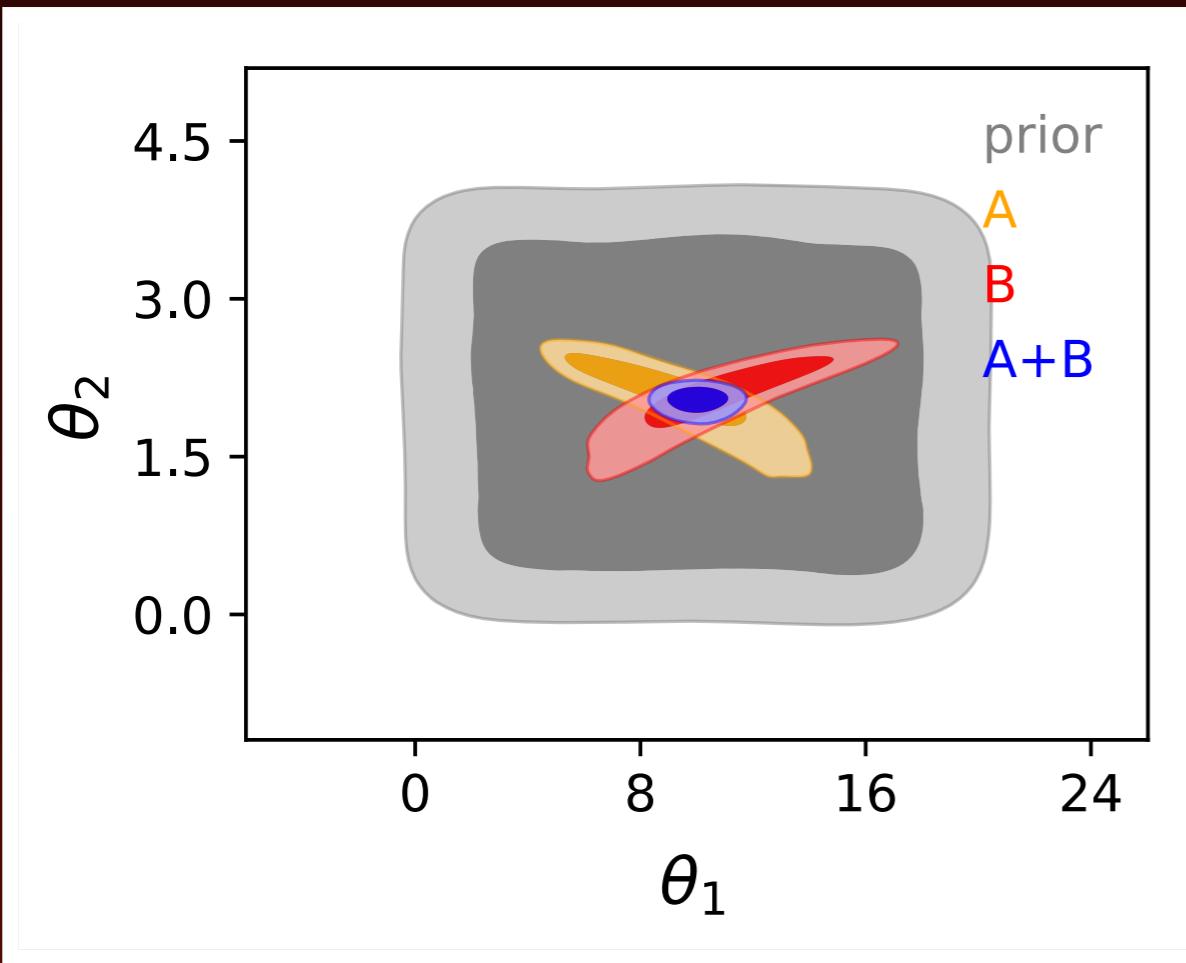
$$p = \int_{d-2 \log S}^{\infty} \chi_d^2(x) dx = \int_{d-2 \log S}^{\infty} \frac{x^{d/2-1} e^{-x/2}}{2^{d/2} \Gamma(d/2)} dx$$



Suspiciousness

What about non-Gaussian posteriors

Box Cox transformations can ‘Gaussianize’ the posterior, and they preserve the Suspiciousness





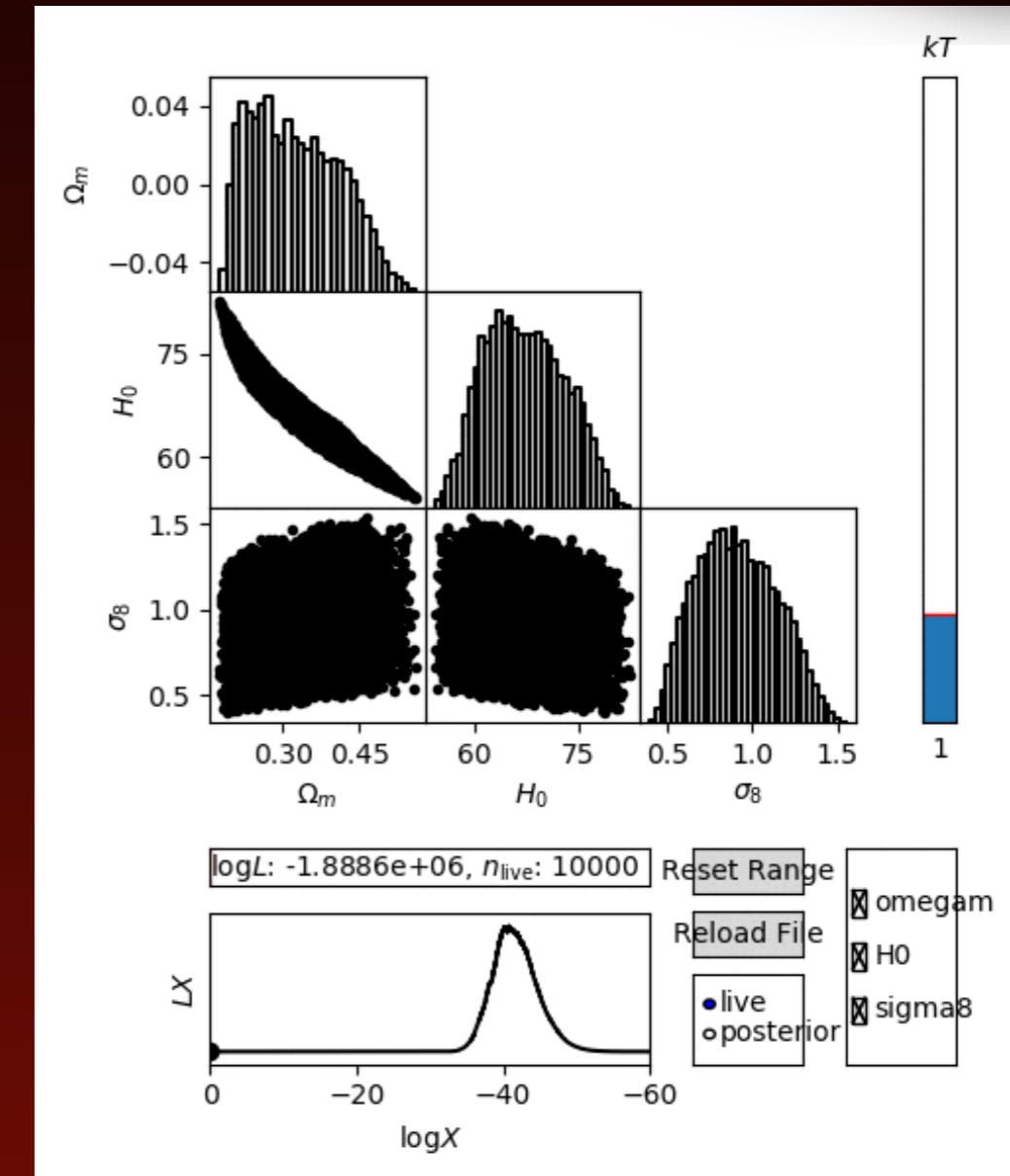
How to calculate this



Anesthetic

<https://github.com/williamjameshandley/anesthetic>

- Public python code
- Computation of Evidences, KL divergences, Bayesian model dimensionalities...
- Marginalised 1D and 2D plots
- Dynamic replaying of nested sampling





How does this work in practice?

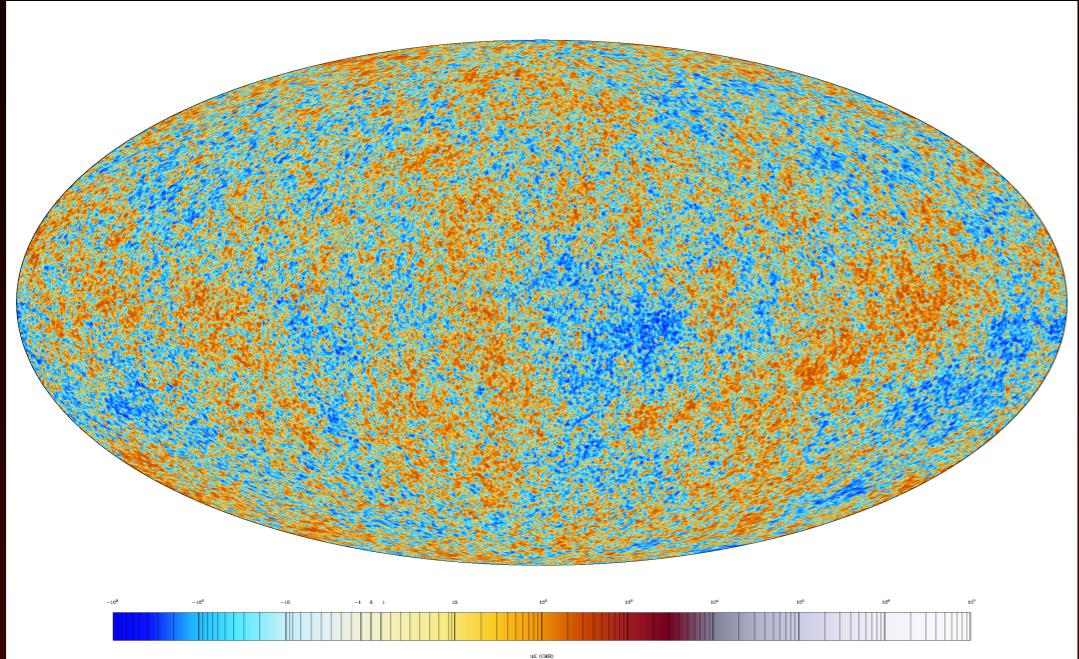
<https://github.com/Pablo-Lemos/Suspiciousness-CosmoSIS.git>



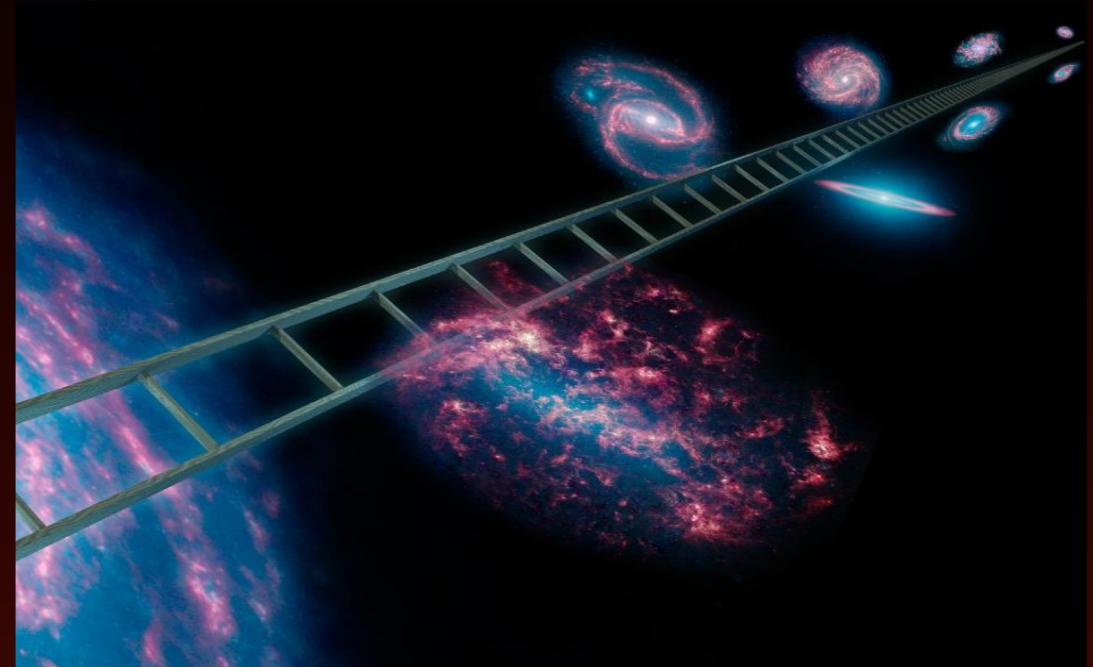
Application to Cosmology



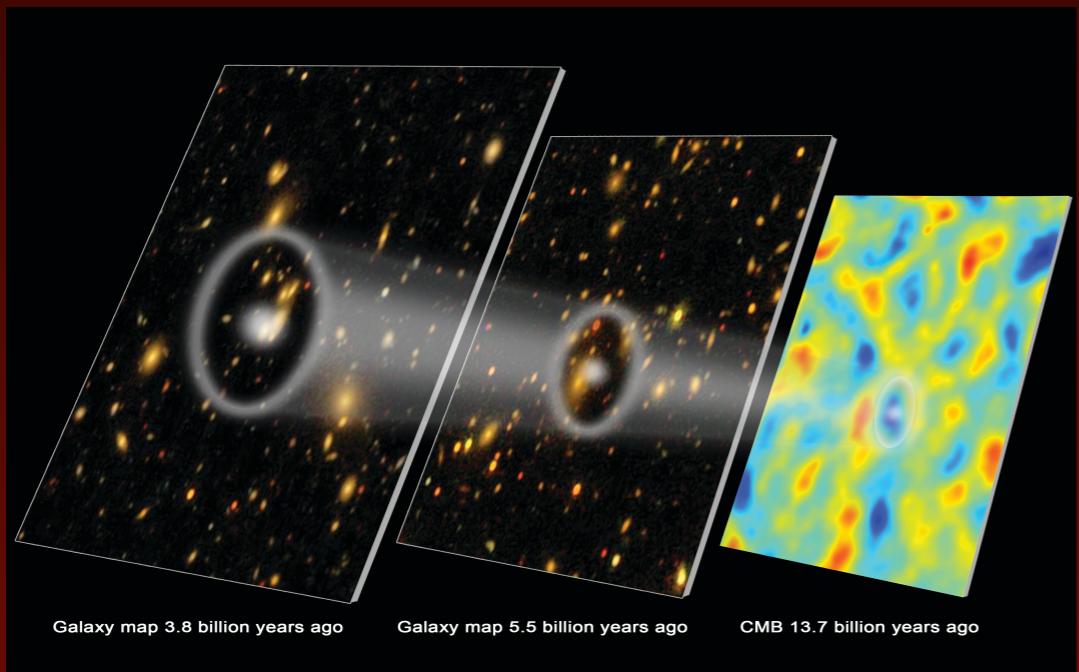
Application to Cosmology



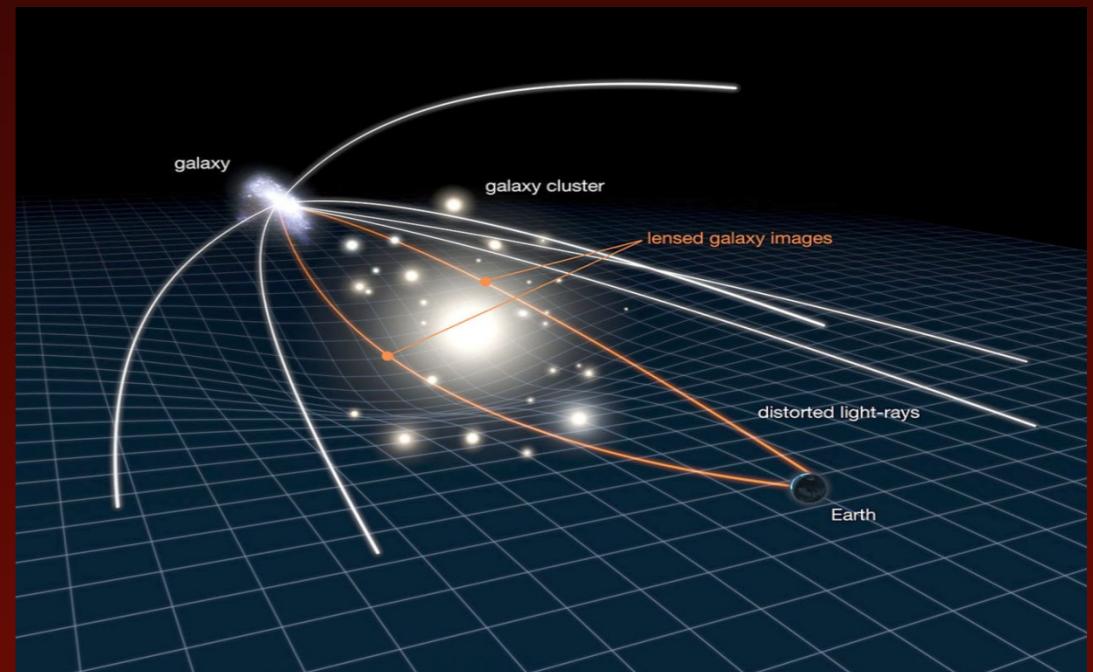
*Cosmic Microwave Background (CMB) - **PLANCK***



*Cosmic Distance Ladder - **SH0ES***



*Baryon Acoustic Oscillations (BAO) - **BOSS***

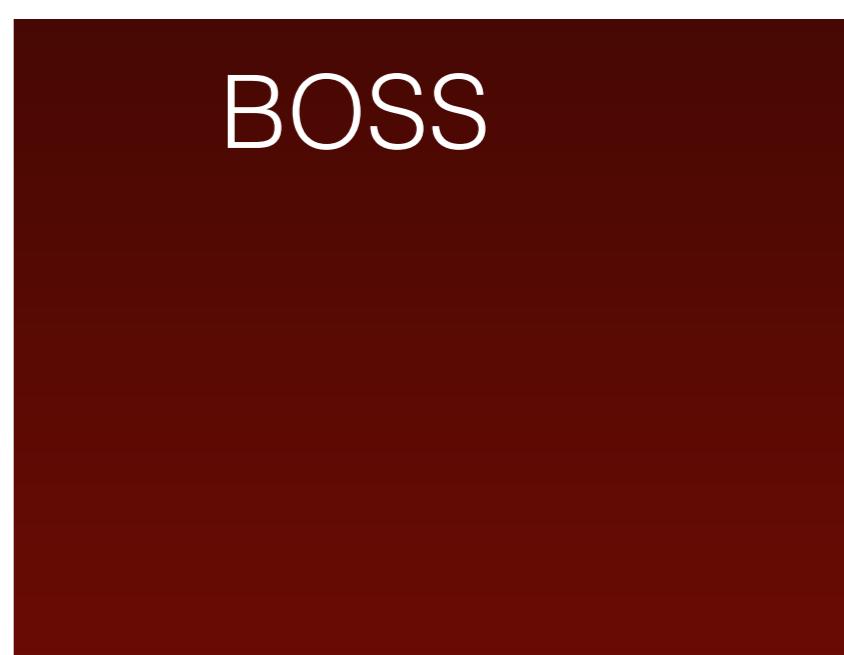
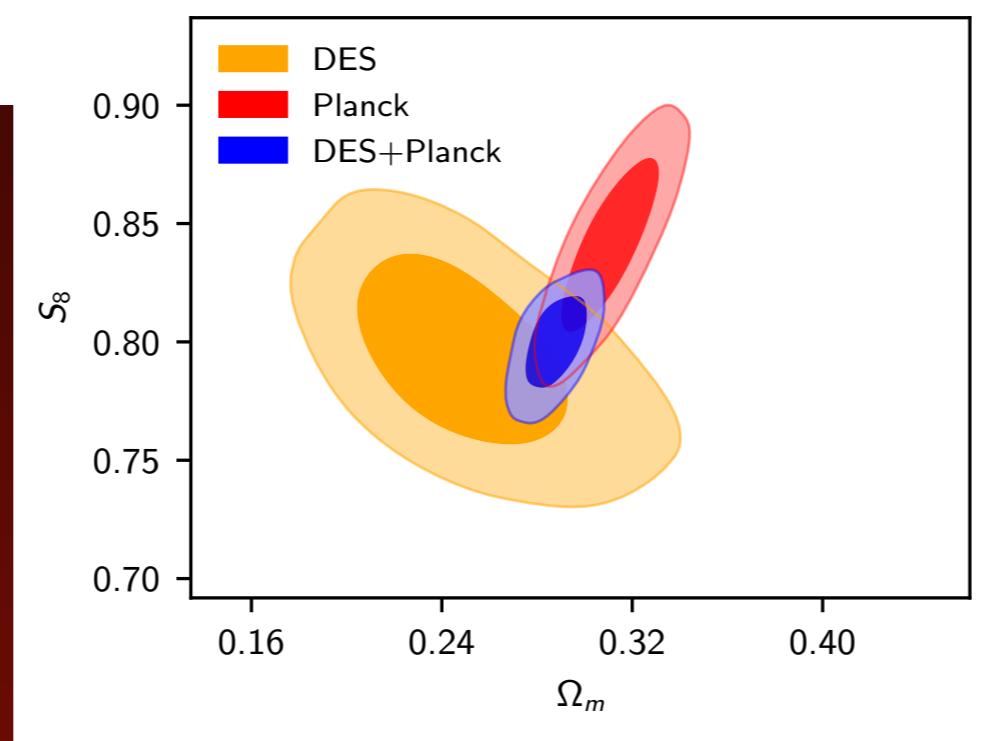
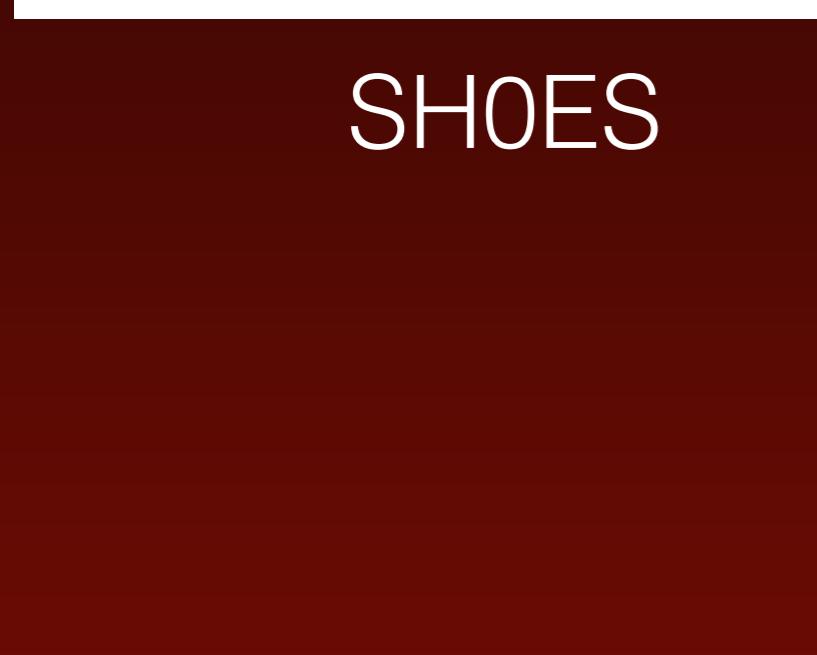
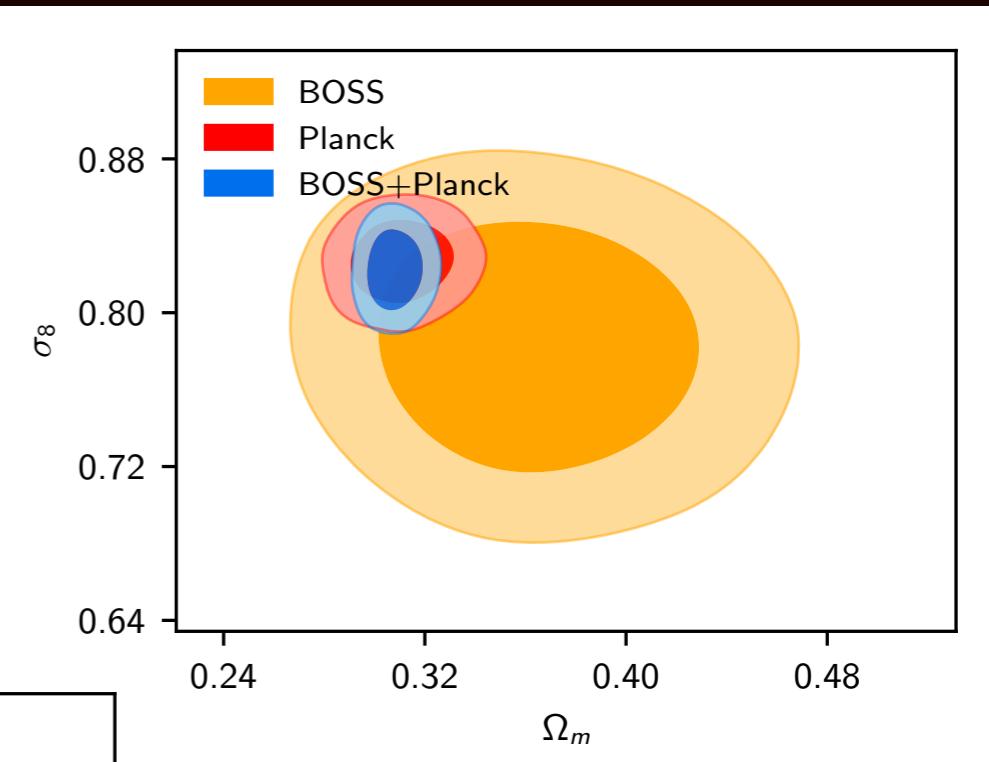
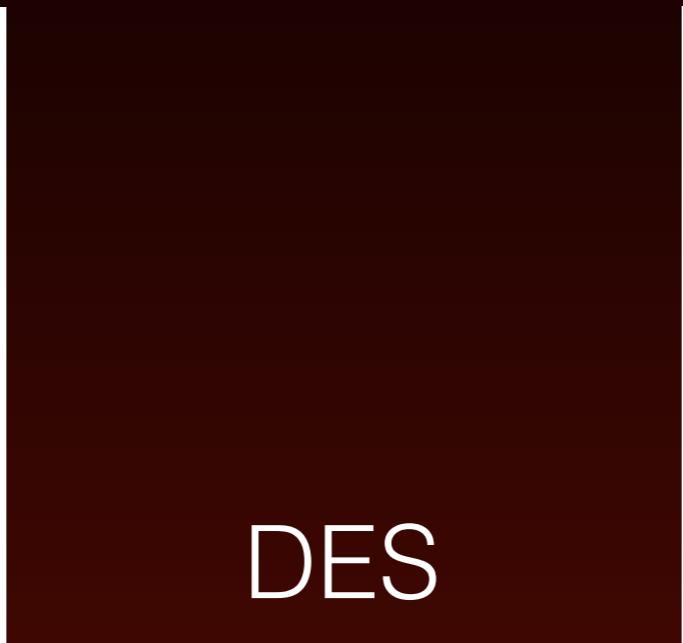
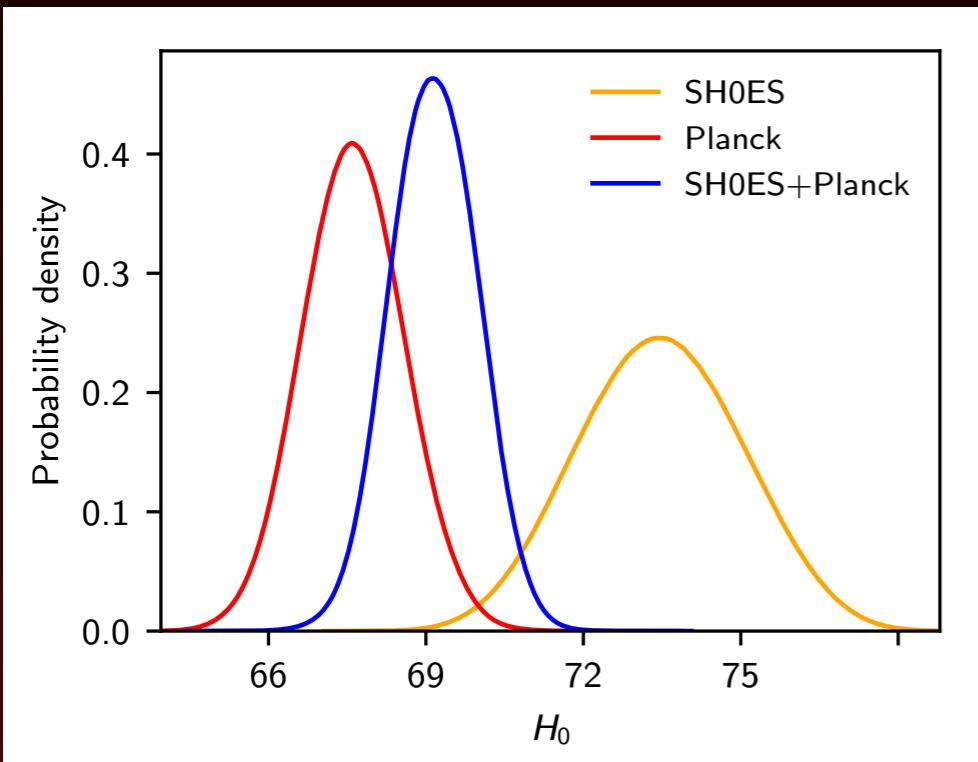


*Weak Galaxy Lensing - **DES***



Application to Cosmology

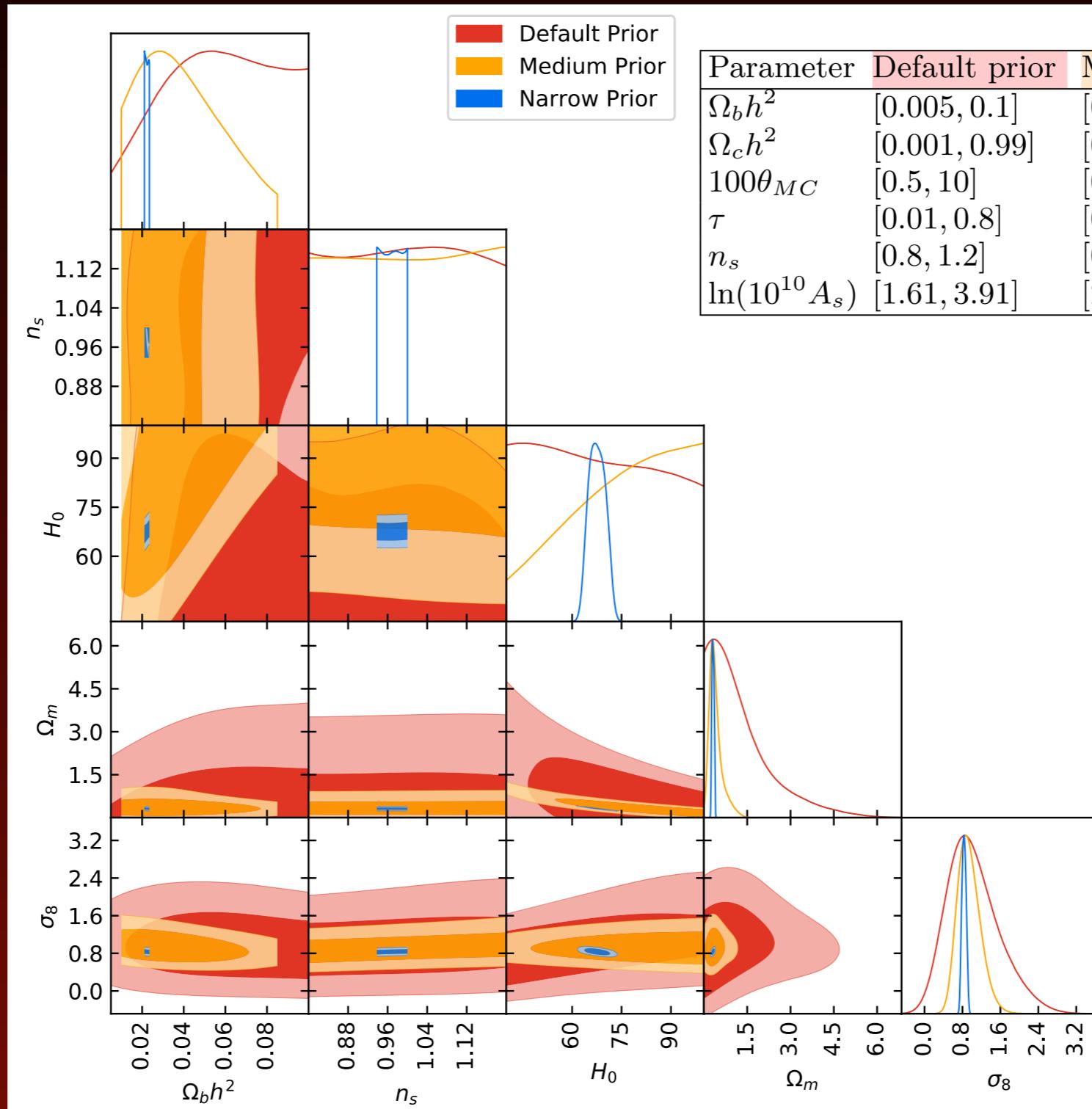
Planck vs...





Application to Cosmology

Three different priors



Parameter	Default prior	Medium prior	Narrow prior
$\Omega_b h^2$	[0.005, 0.1]	[0.01, 0.085]	[0.0211, 0.0235]
$\Omega_c h^2$	[0.001, 0.99]	[0.08, 0.21]	[0.108, 0.131]
$100\theta_{MC}$	[0.5, 10]	[0.97, 1.5]	[1.038, 1.044]
τ	[0.01, 0.8]	[0.01, 0.8]	[0.01, 0.16]
n_s	[0.8, 1.2]	[0.8, 1.2]	[0.938, 1]
$\ln(10^{10} A_s)$	[1.61, 3.91]	[2.6, 3.8]	[2.95, 3.25]

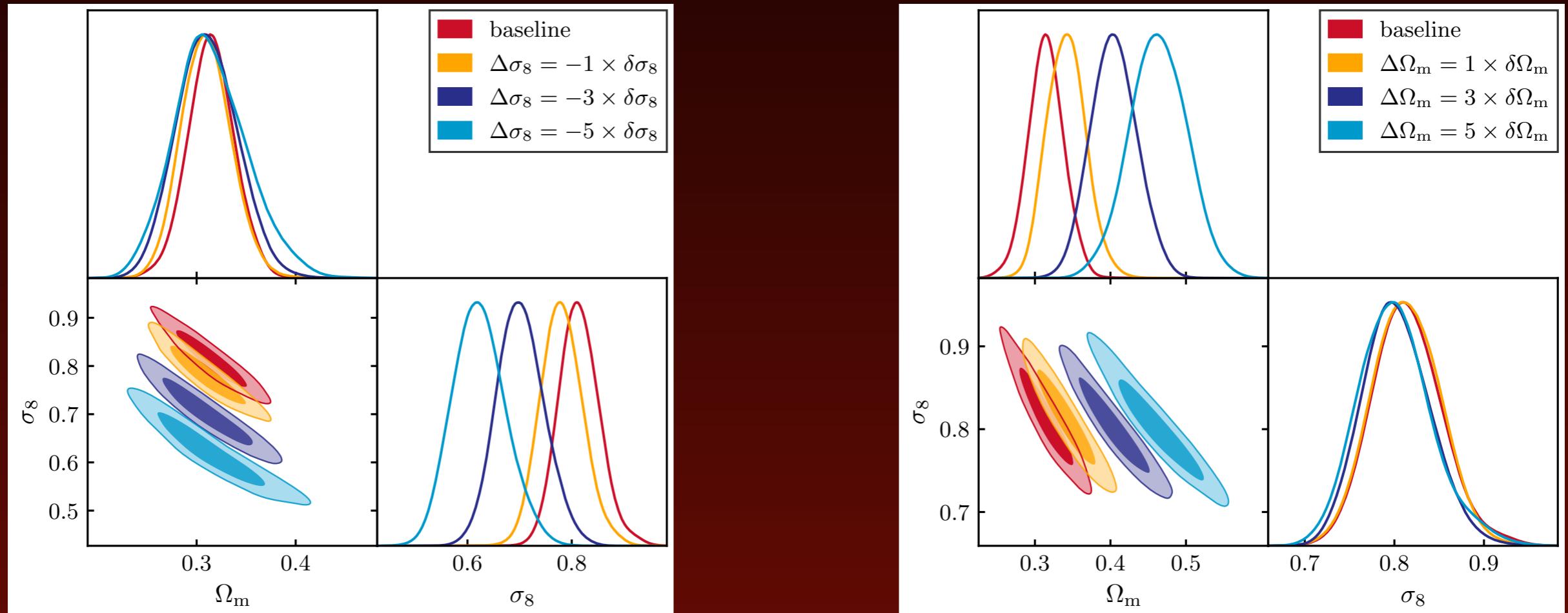


Results

Dataset	Prior	$\log R$	$\log I$	$\log S$	d	$p(\%)$
BOSS- <i>Planck</i>	default	6.30 ± 0.29	6.18 ± 0.29	0.11 ± 0.29	2.91 ± 0.51	42.66 ± 4.28
	medium	4.51 ± 0.28	4.06 ± 0.28	0.46 ± 0.28	3.30 ± 0.55	55.12 ± 4.47
	narrow	1.30 ± 0.23	0.69 ± 0.22	0.61 ± 0.22	1.67 ± 0.54	77.12 ± 14.10
DES- <i>Planck</i>	default	2.88 ± 0.35	6.15 ± 0.34	-3.28 ± 0.34	3.97 ± 0.82	3.23 ± 1.00
	medium	0.51 ± 0.34	4.00 ± 0.34	-3.49 ± 0.34	3.13 ± 0.81	2.04 ± 0.79
	narrow	-1.88 ± 0.29	0.90 ± 0.29	-2.78 ± 0.29	1.15 ± 0.77	1.44 ± 0.91
SH ₀ ES- <i>Planck</i>	default	-2.03 ± 0.29	1.96 ± 0.28	-3.99 ± 0.28	0.78 ± 0.52	0.25 ± 0.17
	medium	-2.50 ± 0.28	1.56 ± 0.28	-4.06 ± 0.28	1.77 ± 0.51	0.56 ± 0.24
	narrow	-2.00 ± 0.23	1.43 ± 0.23	-3.43 ± 0.23	1.92 ± 0.52	1.17 ± 0.45

Simulated DES vs Planck

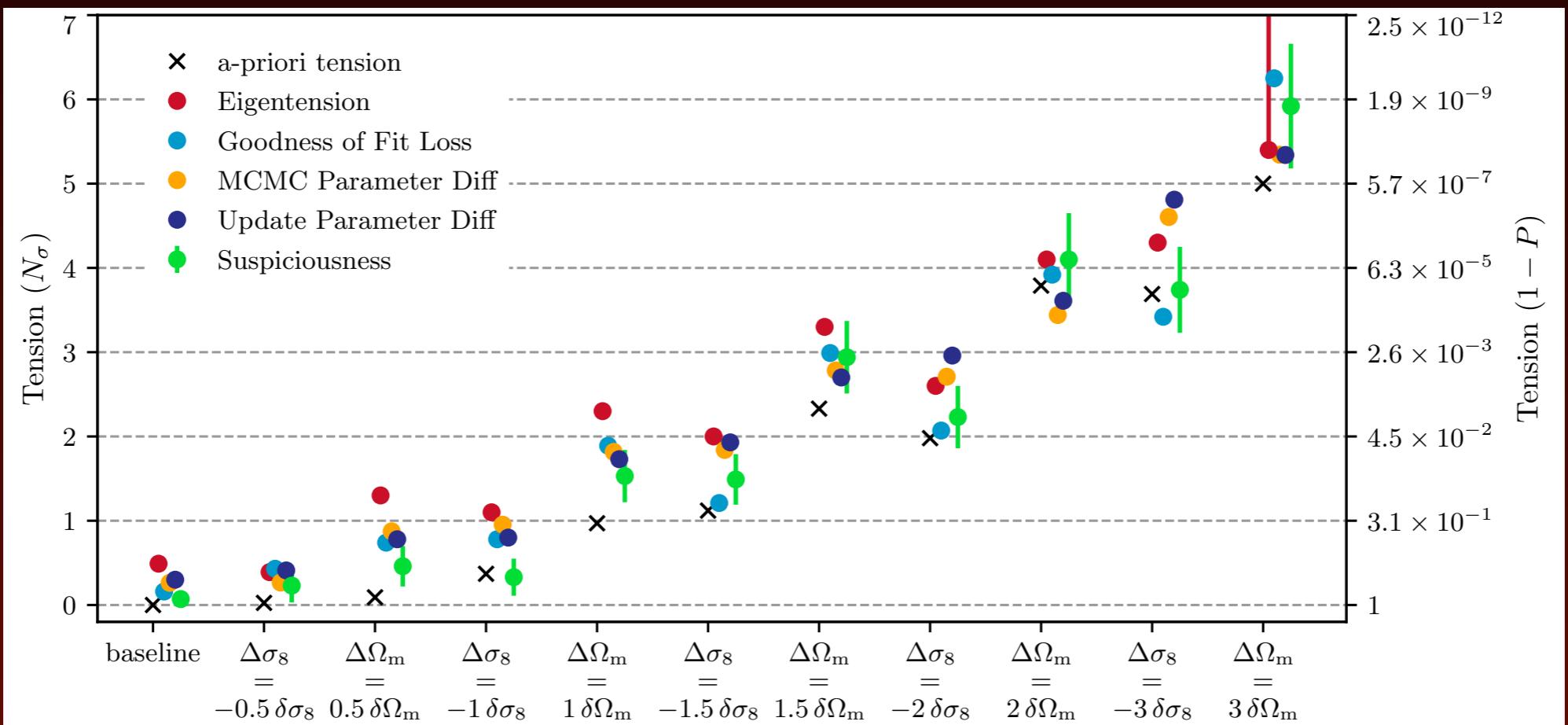
We generated simulated DES data vectors, at cosmologies with a given ‘a priori’ tension with *Planck*.



PL, Raveri & DES Collaboration;
[arXiv: 2012.09554](https://arxiv.org/abs/2012.09554)

Simulated DES vs Planck

We then used the Suspiciousness, Bayes Ratio, and other statistics to quantify the tension between these simulations & *Planck*.



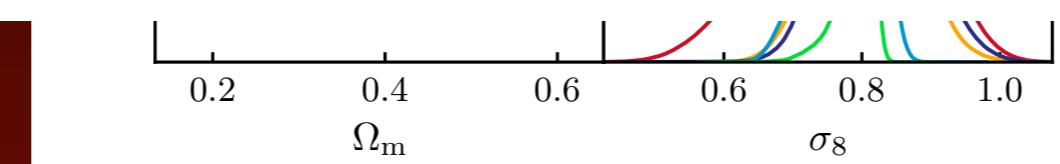
PL, Raveri & DES Collaboration;
[arXiv: 2012.09554](https://arxiv.org/abs/2012.09554)

DES Y1 vs Planck

Finally, we used these metrics to recalibrate the tension between DES Y1 & *Planck*.



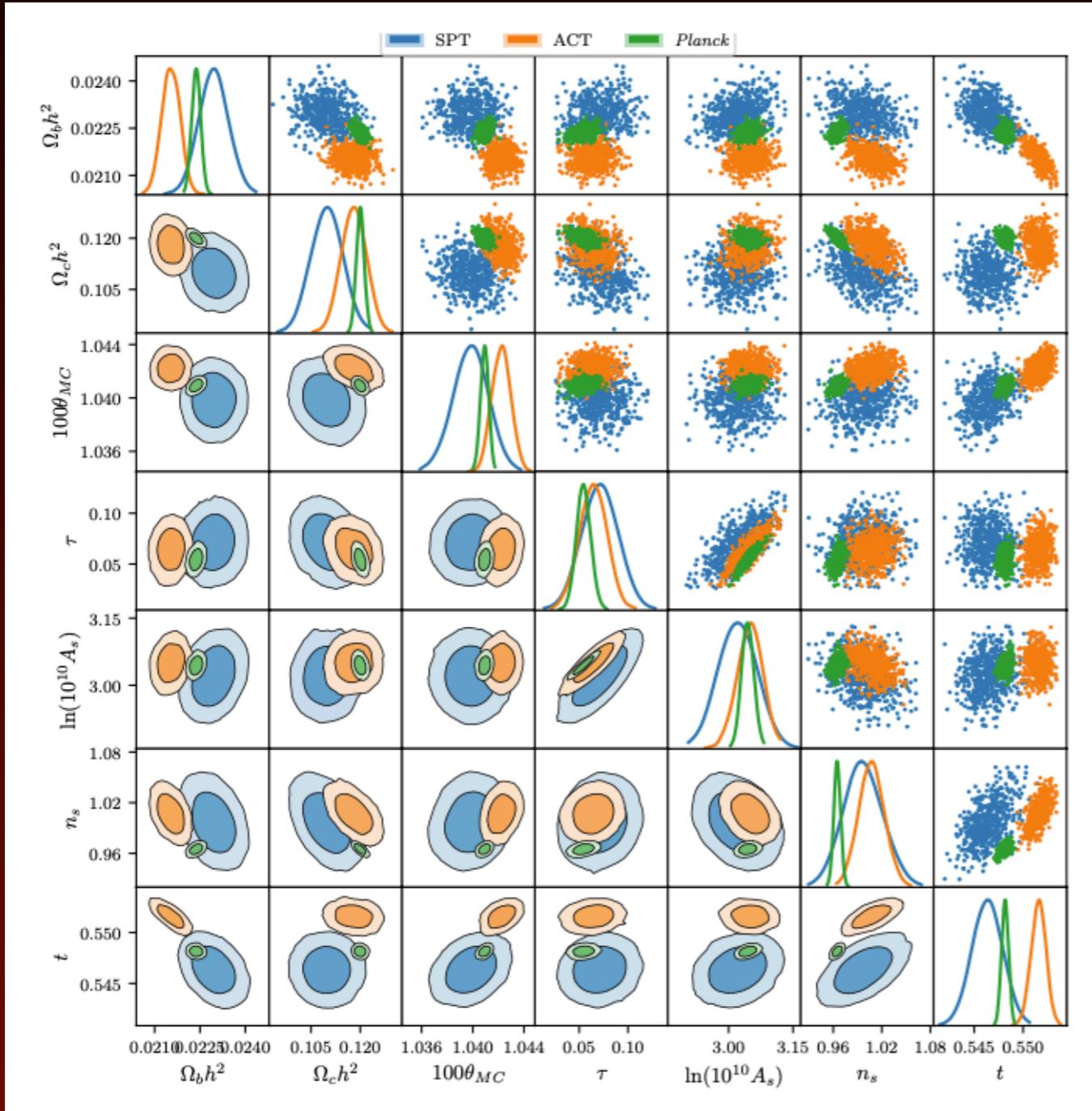
data set	$\log R$	Bayes ratio Interpretation	Eigentension	GoF Loss	MCMC/Update Param Shifts	Suspiciousness
DES cosmic shear vs. <i>Planck</i> 15	2.2 ± 0.5	Substantial Agreement	1.8σ	1.3σ	$1.3/1.2 \sigma$	$(0.7 \pm 0.4) \sigma$
DES 3 \times 2pt vs. <i>Planck</i> 15	1.0 ± 0.5	No Evidence	2.4σ	2.7σ	$2.2/2.2 \sigma$	$(2.4 \pm 0.2) \sigma$
DES 5 \times 2pt vs. <i>Planck</i> 15	1.1 ± 0.5	Substantial Agreement	2.4σ	2.8σ	$2.1/2.3 \sigma$	$(2.2 \pm 0.3) \sigma$
DES 5 \times 2pt vs. <i>Planck</i> 15 + lensing	1.0 ± 0.6	No Evidence	2.4σ	2.5σ	$2.1/2.3 \sigma$	$(2.2 \pm 0.4) \sigma$
DES 5 \times 2pt + <i>Planck</i> lensing vs. <i>Planck</i> 15	6.1 ± 0.6	Strong Agreement	1.6σ	2.4σ	$1.9/2.2 \sigma$	$(1.8 \pm 0.2) \sigma$
DES cosmic shear vs. <i>Planck</i> 18	3.3 ± 0.4	Strong Agreement	1.5σ	1.0σ	$1.0/1.1 \sigma$	$(0.5 \pm 0.3) \sigma$
DES 3 \times 2pt vs. <i>Planck</i> 18	2.2 ± 0.6	Substantial Agreement	2.2σ	1.6σ	$2.0/2.3 \sigma$	$(2.4 \pm 0.2) \sigma$



PL, Raveri & DES Collaboration;
[arXiv: 2012.09554](https://arxiv.org/abs/2012.09554)



Application to the CMB



Dataset combination	χ^2	p	tension	$\log S$
ACT vs <i>Planck</i>	17.2	0.86%	2.63σ	-5.60
ACT vs SPT	15.4	1.77%	2.37σ	-4.68
<i>Planck</i> vs SPT	9.1	16.82%	1.38σ	-1.55
ACT vs <i>Planck</i> +SPT	18.4	0.52%	2.79σ	-6.22
ACT+SPT vs <i>Planck</i>	12.2	5.81%	1.90σ	-3.09
ACT+ <i>Planck</i> vs SPT	10.3	11.09%	1.59σ	-2.17

Handley & PL;
arXiv: 2007.08496



Application to KiDS

KiDS-1000 Cosmology: Multi-probe weak gravitational lensing and spectroscopic galaxy clustering constraints

Catherine Heymans^{1,2*}, Tilman Tröster^{1**}, Marika Asgari¹, Chris Blake³, Hendrik Hildebrandt², Benjamin Joachimi⁴, Konrad Kuijken⁵, Chieh-An Lin¹, Ariel G. Sánchez⁶, Jan Luca van den Busch², Angus H. Wright², Alexandra Amon⁷, Maciej Bilicki⁸, Jelte de Jong⁹, Martin Crocce^{10,11}, Andrej Dvornik², Thomas Erben¹², Maria Cristina Fortuna⁵, Fedor Getman¹³, Benjamin Giblin¹, Karl Glazebrook³, Henk Hoekstra⁵, Shahab Joudaki¹⁴, Arun Kannawadi^{15,5}, Fabian Köhlinger², Chris Lidman¹⁶, Lance Miller¹⁴, Nicola R. Napolitano¹⁷, David Parkinson¹⁸, Peter Schneider¹², Huan Yuan Shan^{19,20}, Edwin A. Valentijn⁹, Gijs Verdoes Kleijn⁹, and Christian Wolf¹⁶

Handley & Lemos (2019) propose the ‘suspiciousness’ statistic S that is based on the Bayes factor, R , but hardened against prior dependences. We find that the probability of observing our measured suspiciousness statistic is 0.08 ± 0.02 , which corresponds to a *KiDS-Planck* tension at the level of $1.8 \pm 0.1 \sigma$ (see Appendix G.3 for details).

The second equality follows from Bayes theorem: $P = \mathcal{L}\pi / Z$. Using this definition of \mathcal{D} allows us to rephrase the suspiciousness solely in terms of the expectation values of the log-likelihoods:

$$\begin{aligned} \ln S = & \langle \ln \mathcal{L}_{3 \times 2\text{pt} + \text{Planck}} \rangle_{P_{3 \times 2\text{pt} + \text{Planck}}} - \langle \ln \mathcal{L}_{3 \times 2\text{pt}} \rangle_{P_{3 \times 2\text{pt}}} \\ & - \langle \ln \mathcal{L}_{\text{Planck}} \rangle_{P_{\text{Planck}}}. \end{aligned} \quad (\text{G.9})$$



Thanks for listening!

- Cosmological ‘**Tensions**’ could be a hint of new physics, and must therefore be understood.
- Quantifying tension is therefore crucial. We propose the ‘**Suspiciousness**’ as the optimal metric of tension in Cosmology.
- The method can be extended to any other problem of assessing consistency between data sets, in astrophysics or otherwise.