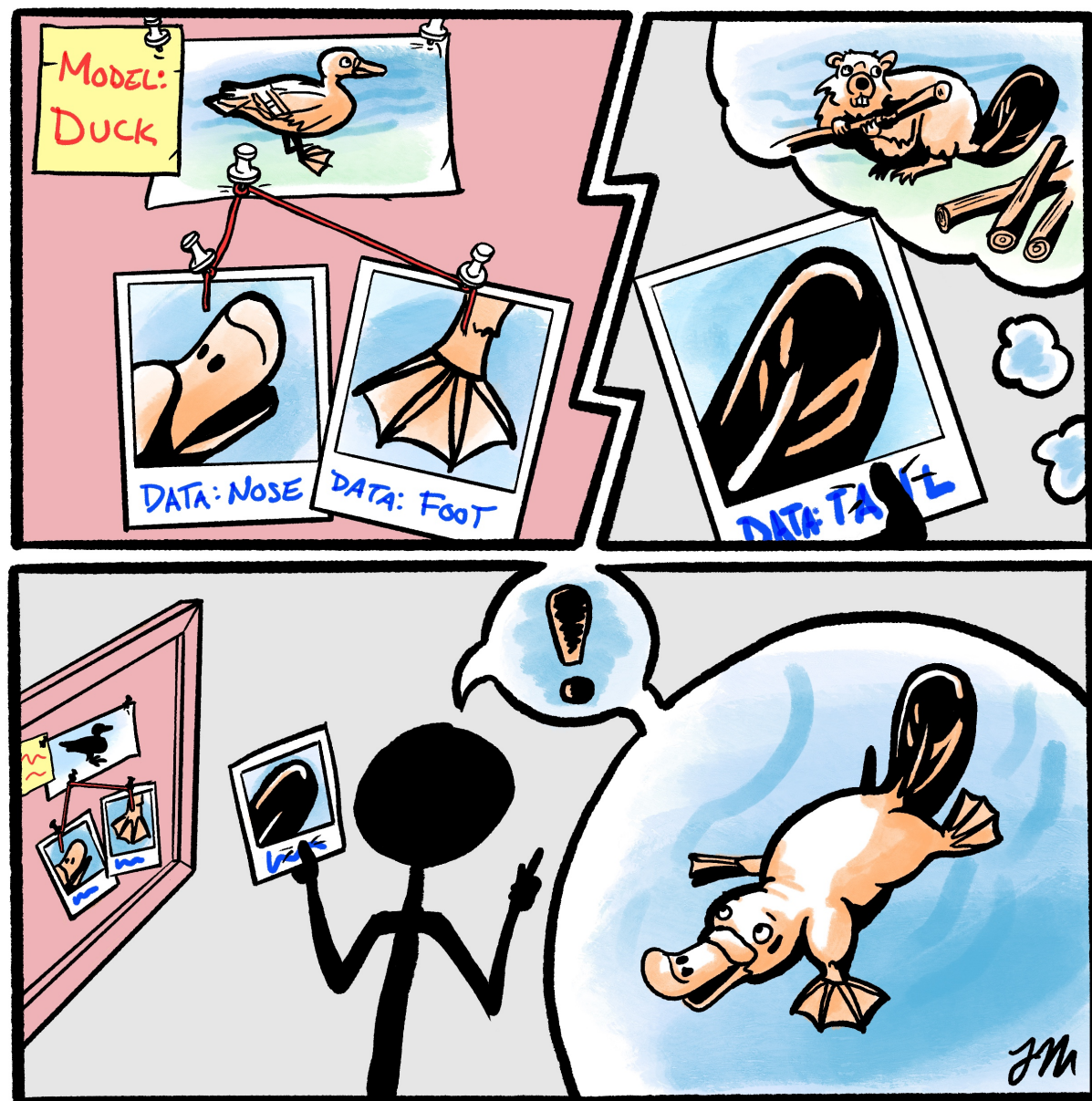




# cosmological Tensions

(and How to Find Them)



# Darkbites

© 2020 Jessie Muir

Pablo Lemos  
University College London -  
University of Sussex  
07-05-2021  
[p.lemos@sussex.ac.uk](mailto:p.lemos@sussex.ac.uk)

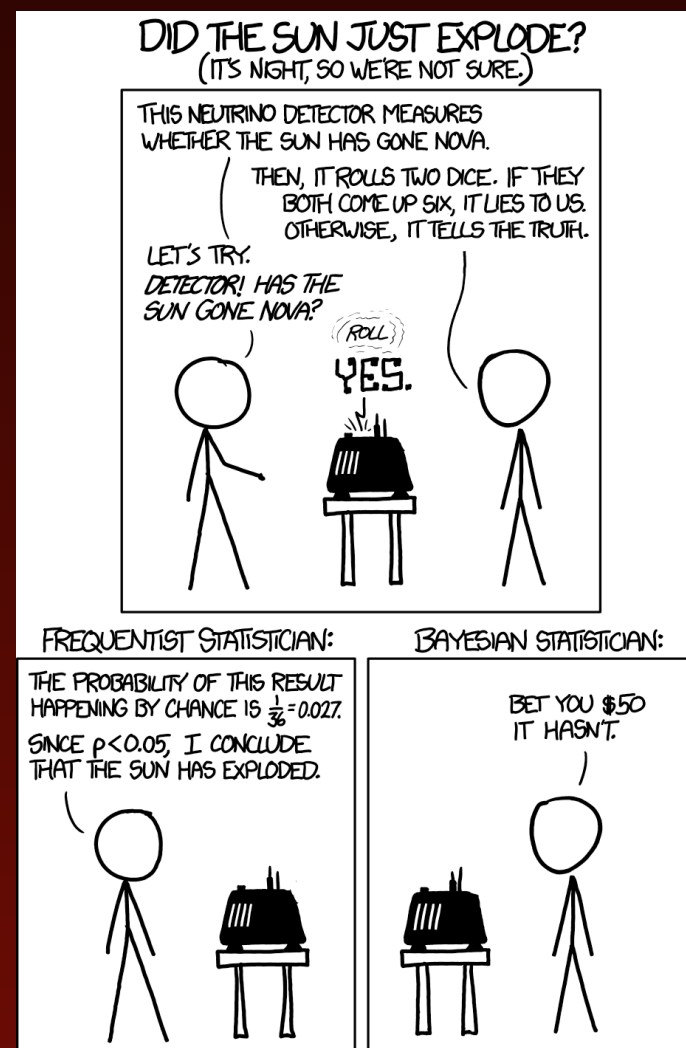
Text: Andresa Campos  
[@AndresaCampos](https://twitter.com/AndresaCampos)

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





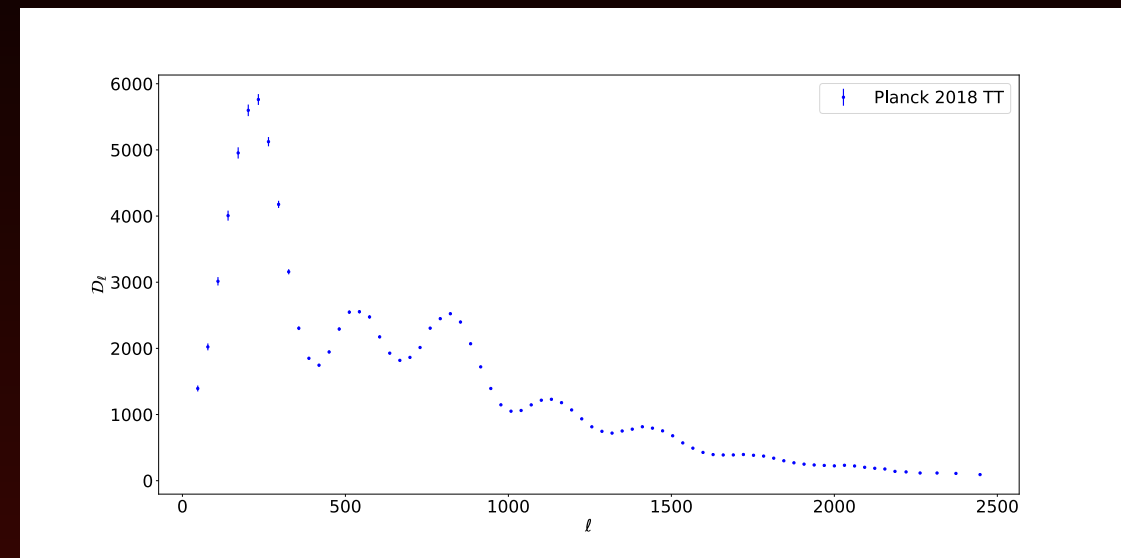
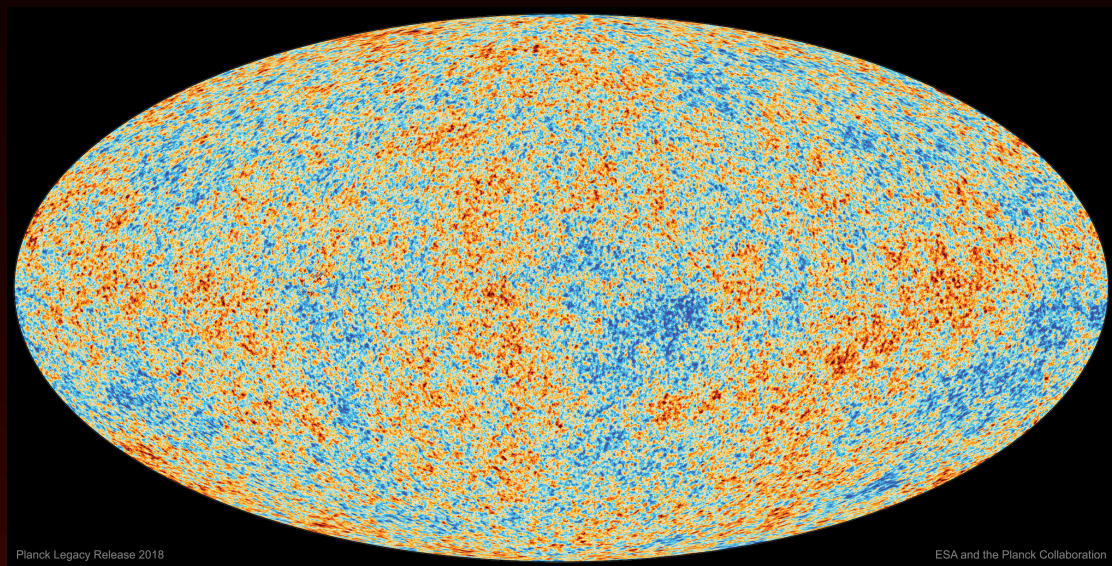
## Bayes' Theorem

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

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

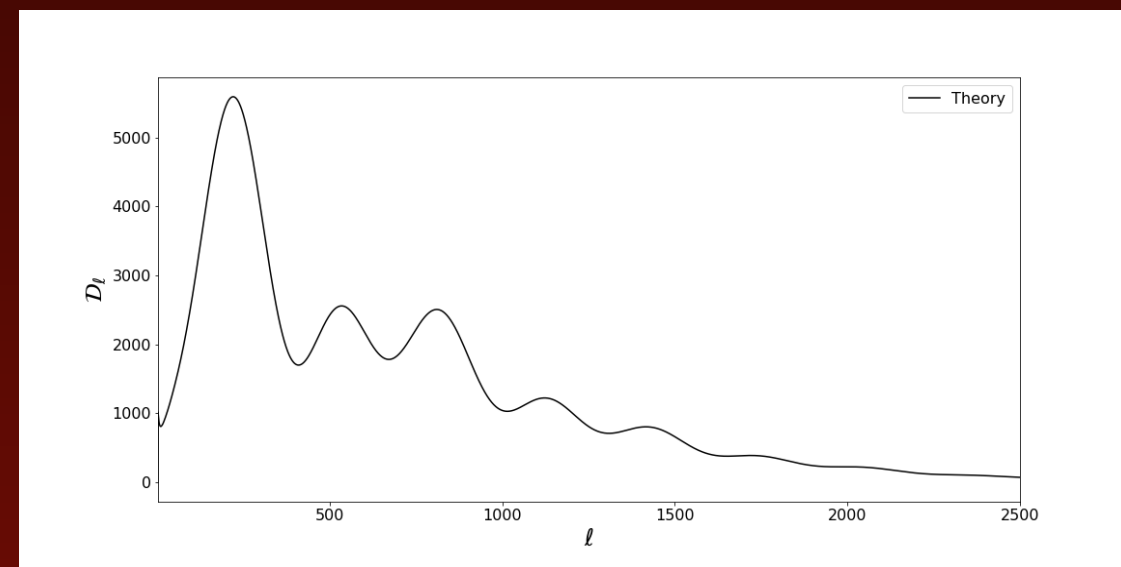
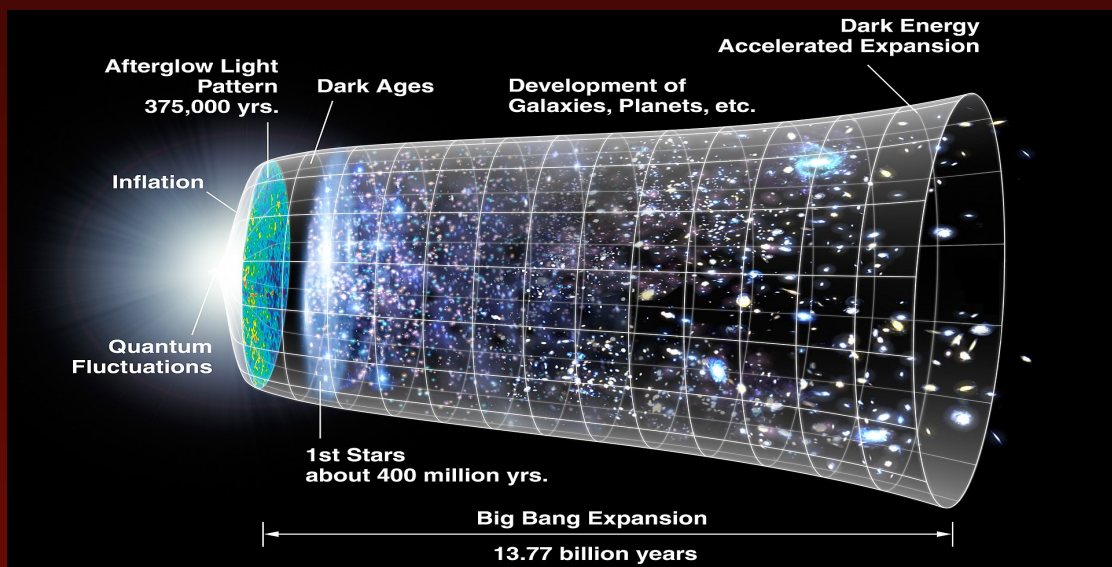
- **$\theta$** : *Parameters*
- **$D$** : *Data*
- **$M$** : *Model*





$\theta, M$

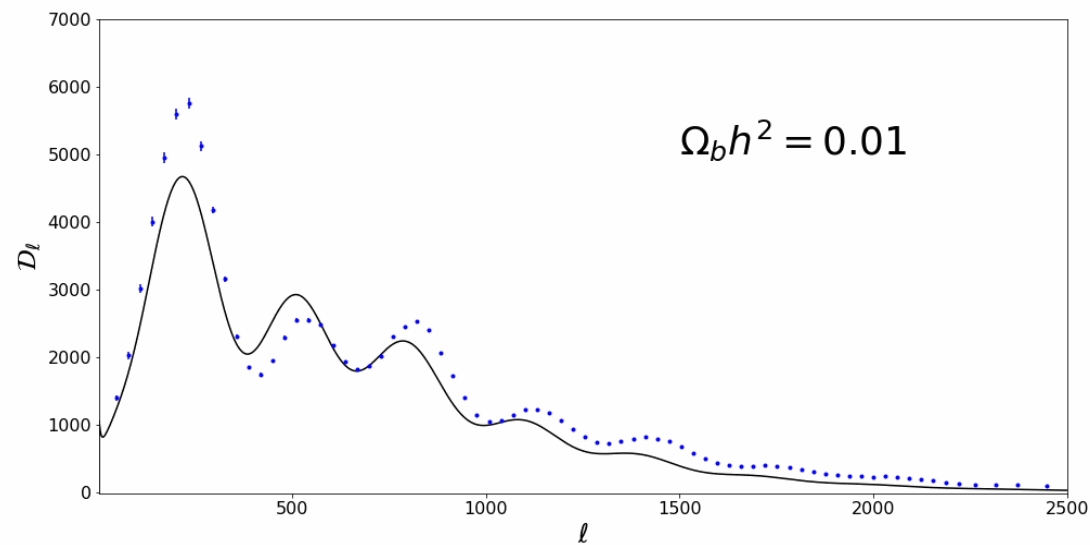
$$\mathcal{L} \equiv P(D | \theta, M)$$



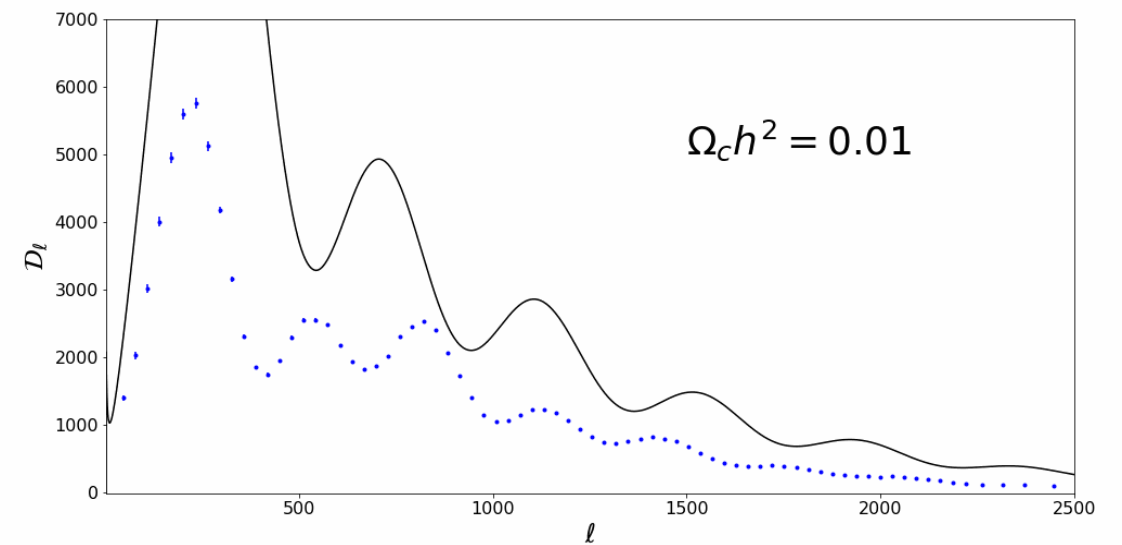




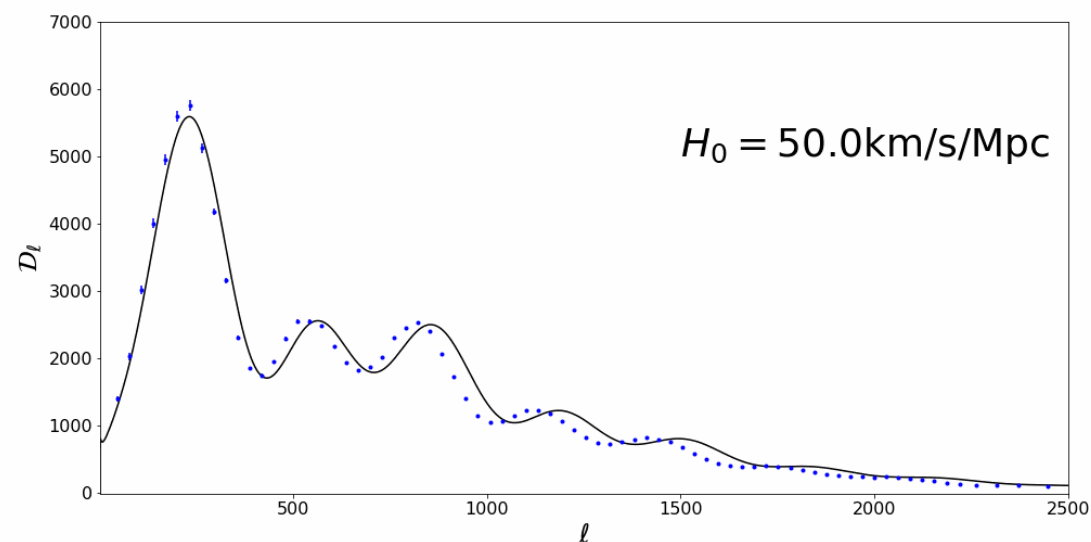
## How much dark energy?



## How much dark matter?



## How fast does it expand?





## Three types of problem

- Parameter estimation
- Model comparison
- Dataset comparison

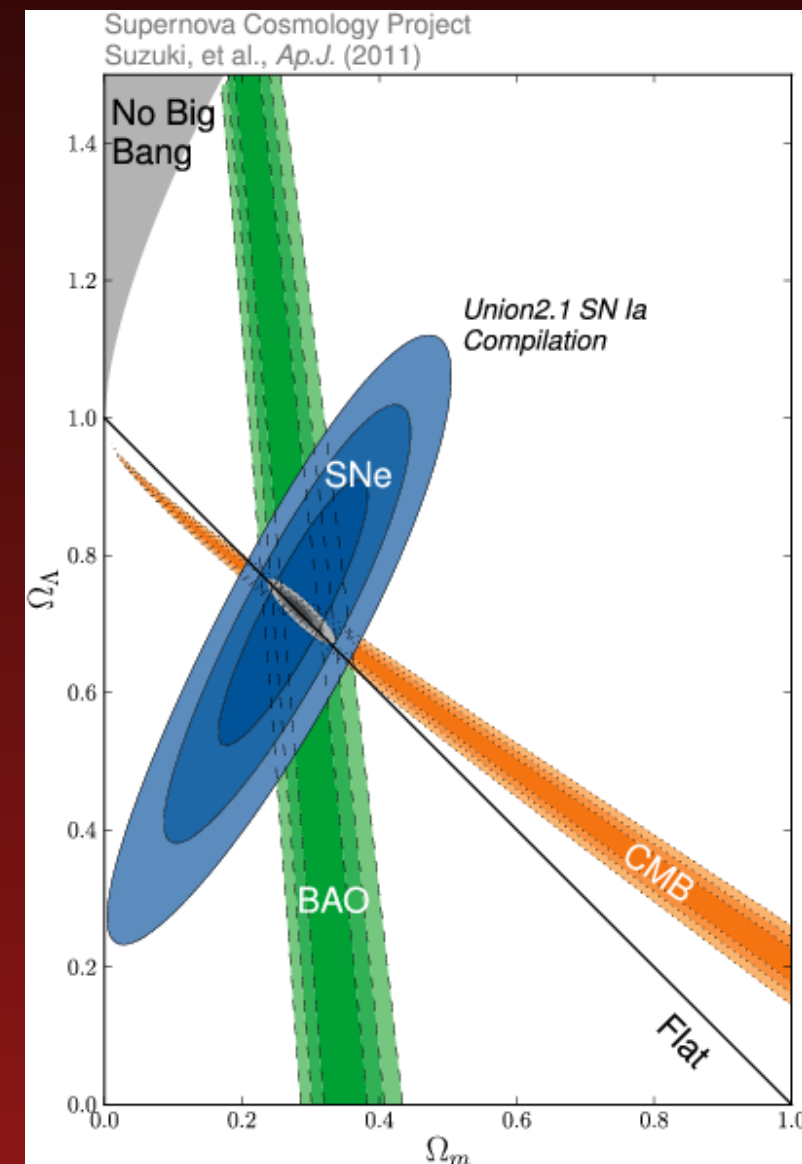
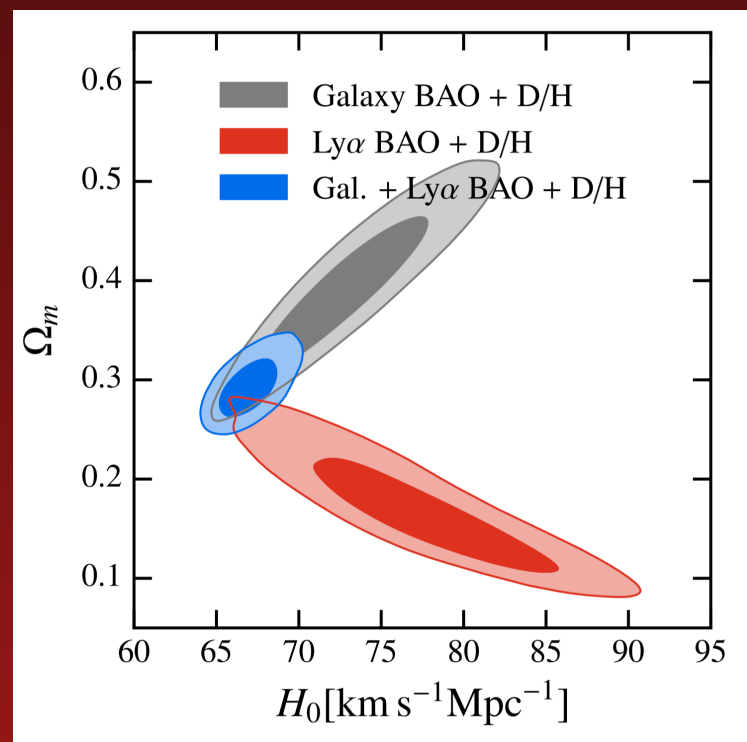
**'Tension'**



# Bayesian Statistics

## 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!



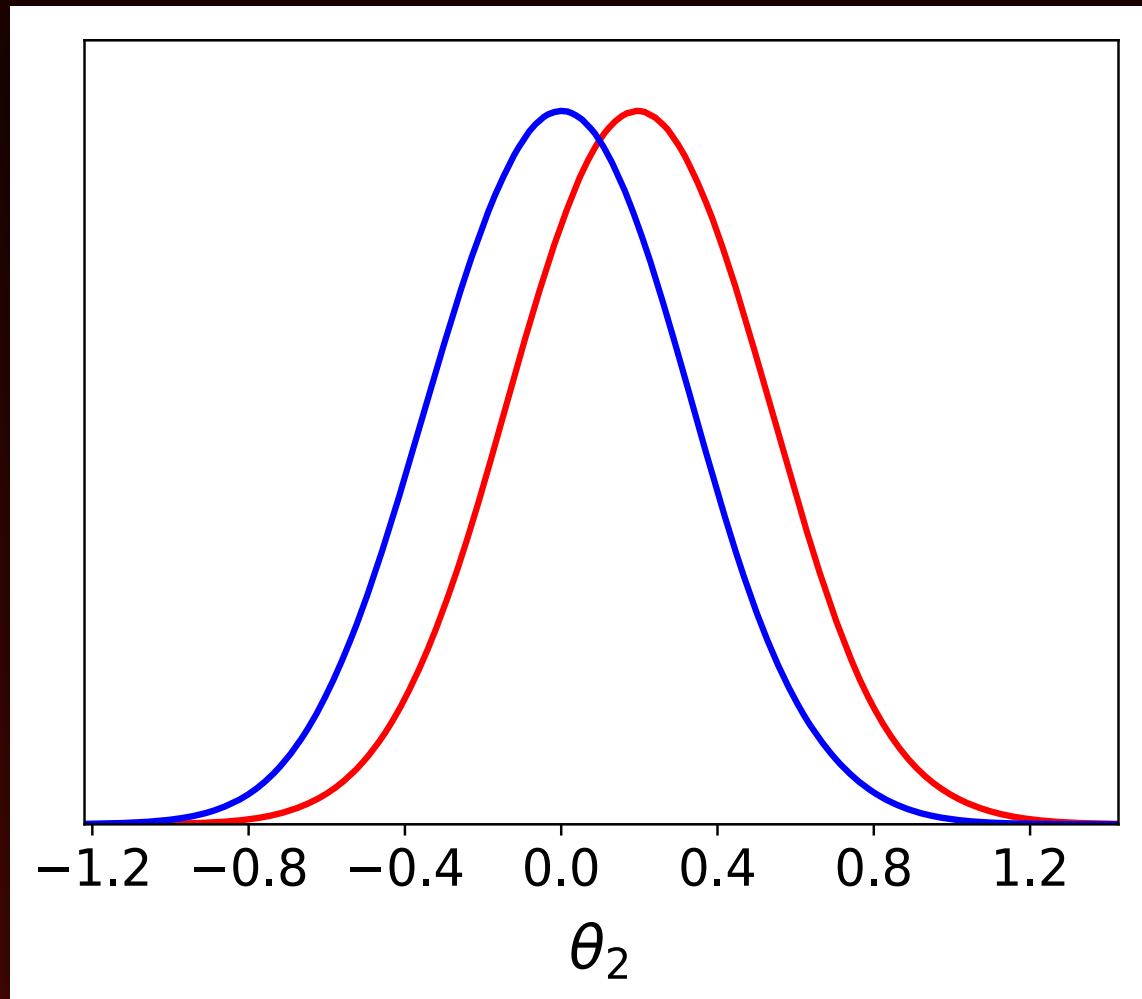




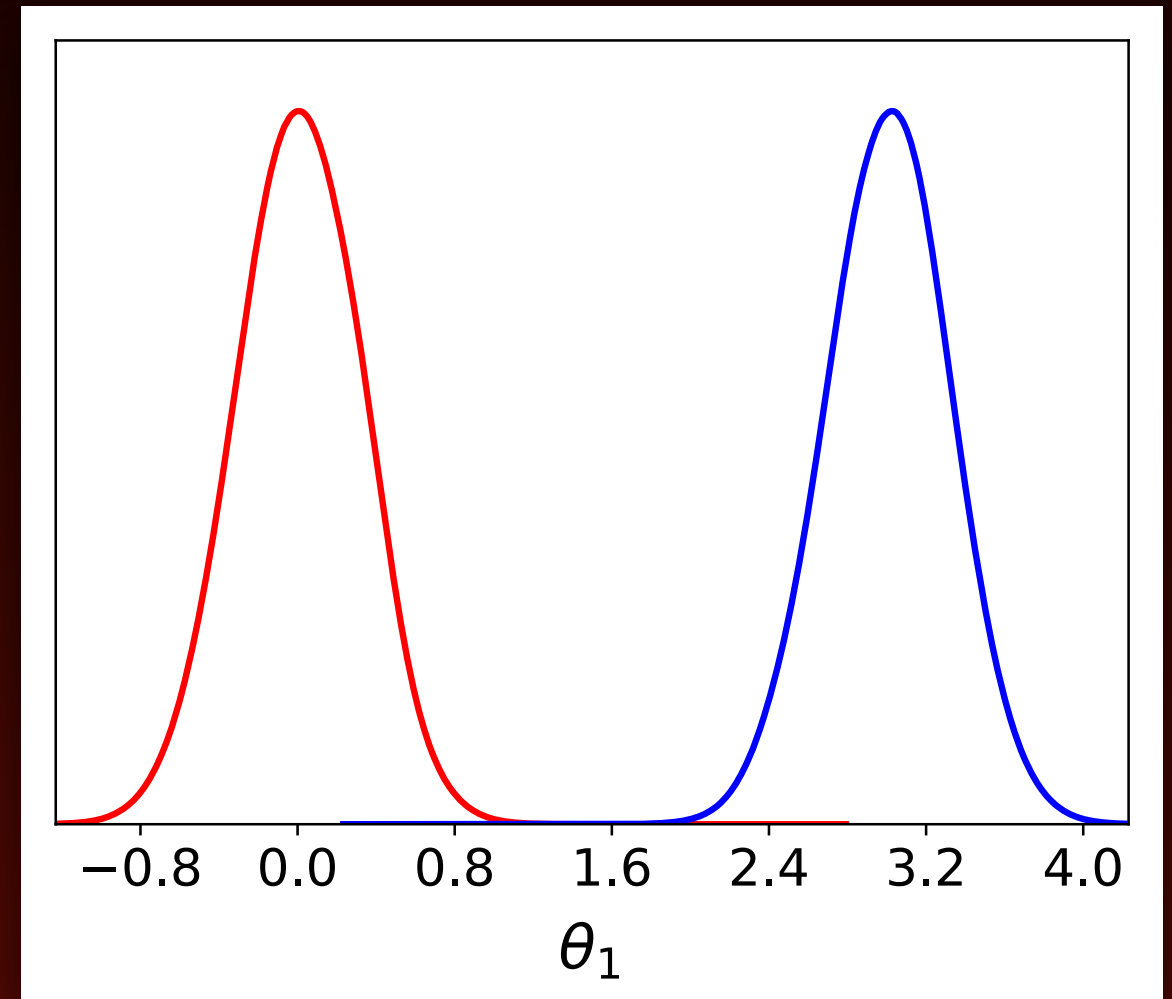
# Why is Data Set Comparison Non-Trivial?



# Trivial?



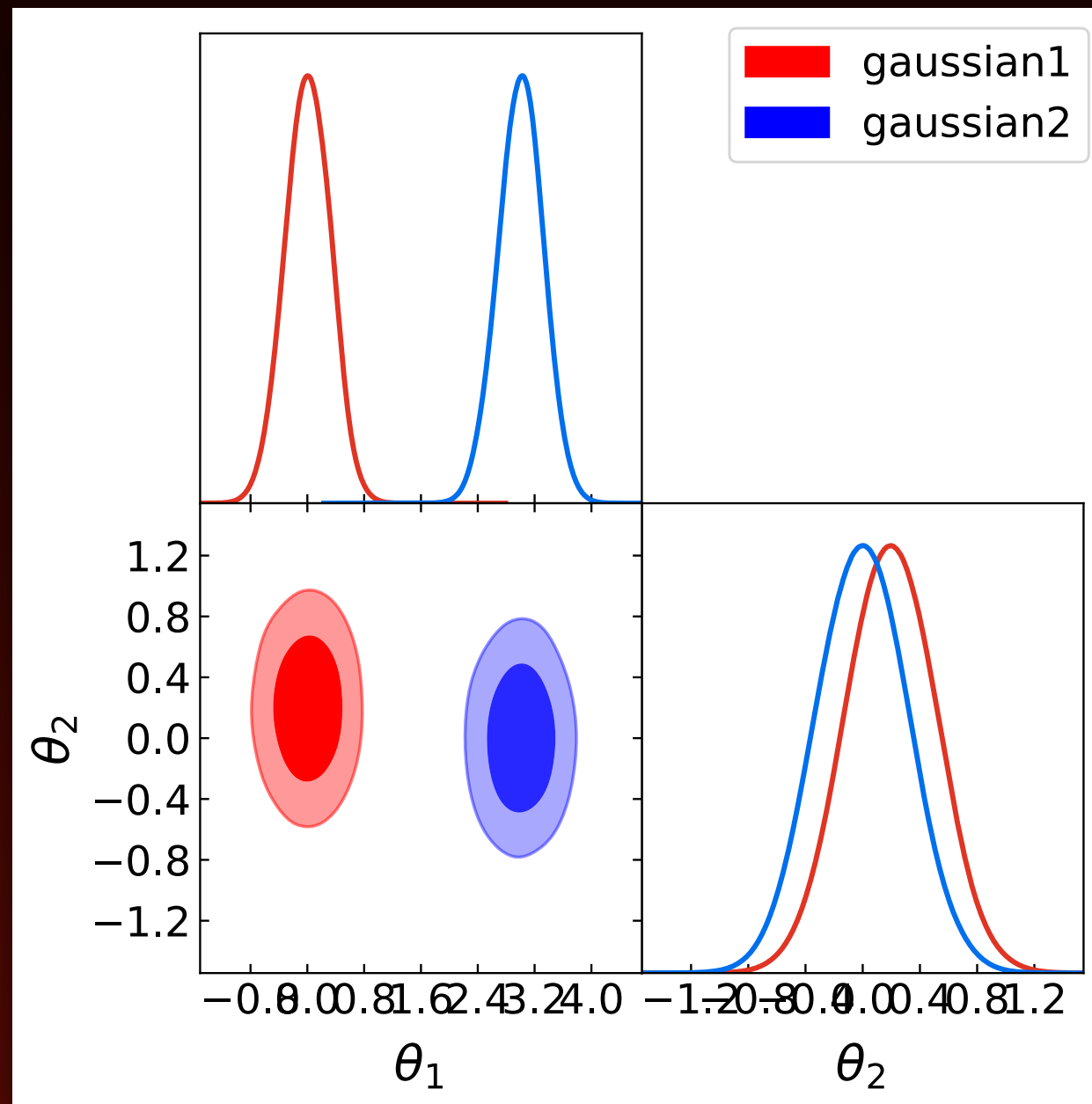
Consistent



Inconsistent



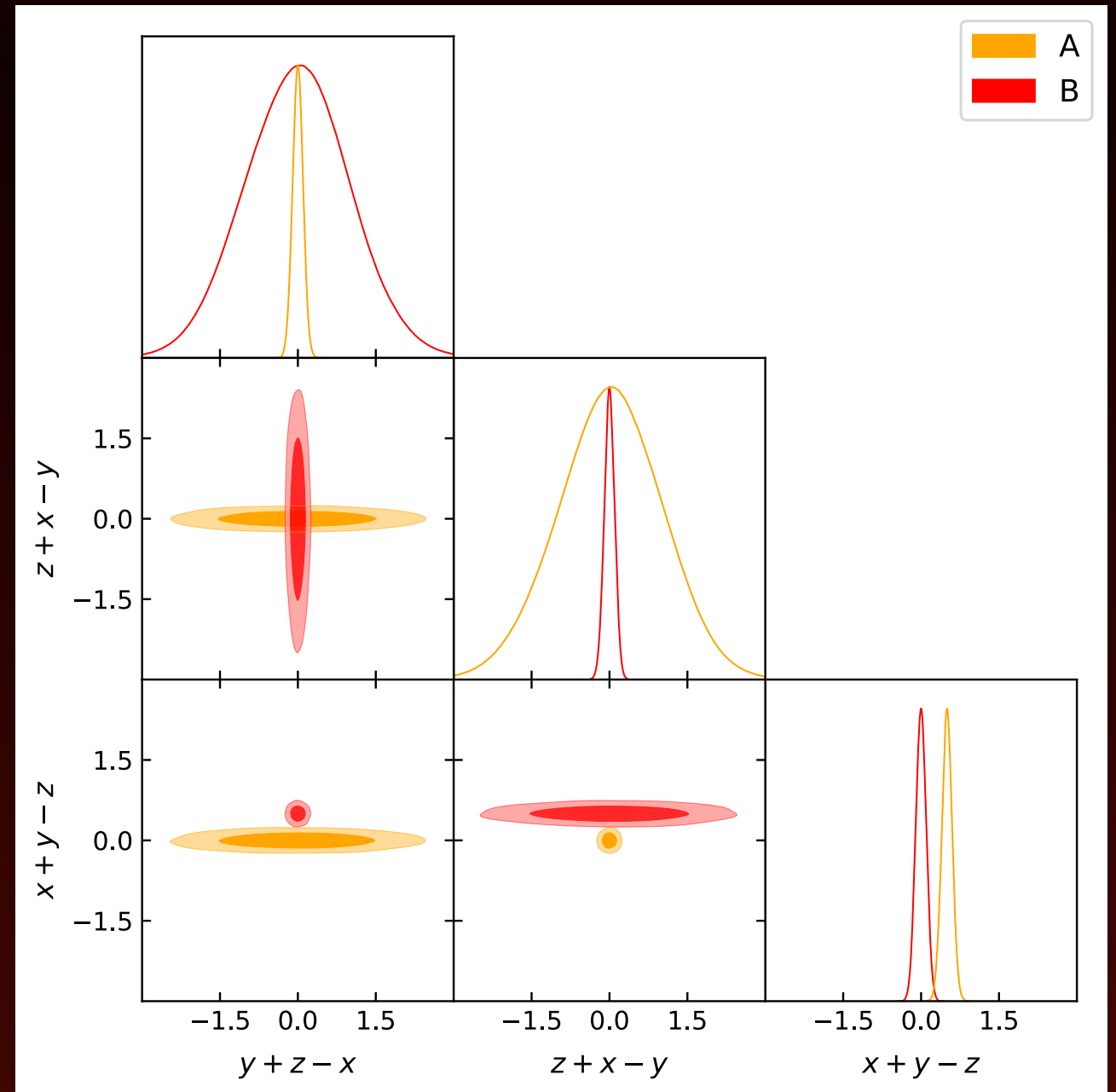
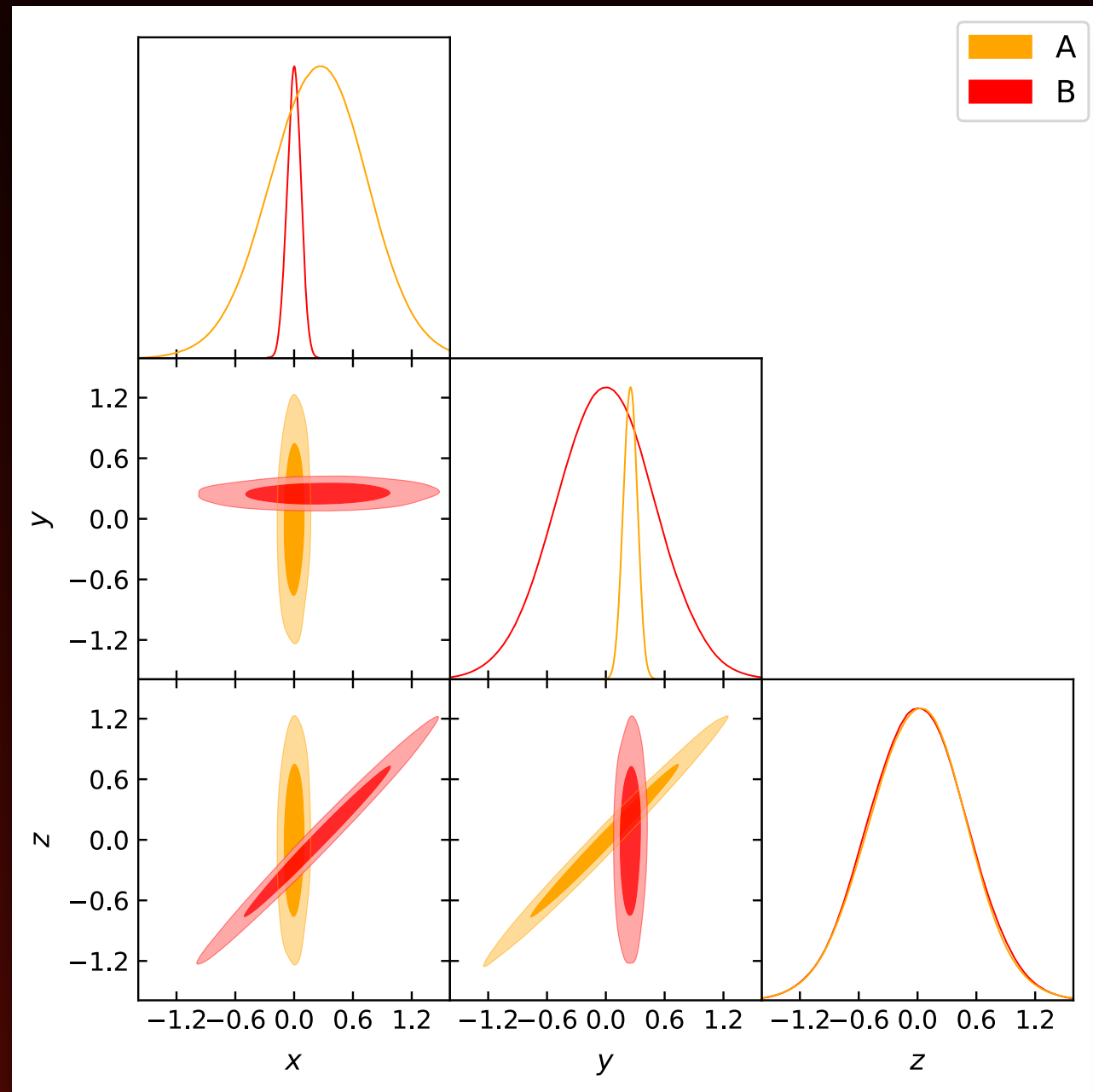
# Trivial?







# 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 data (WMAP, ACBAR, CBI, VSA), SDSS and concordance is favoured and the tightening justified.

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

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

$$\mathcal{Z} =$$

Probability that both datasets come **DIFFERENT** Universes

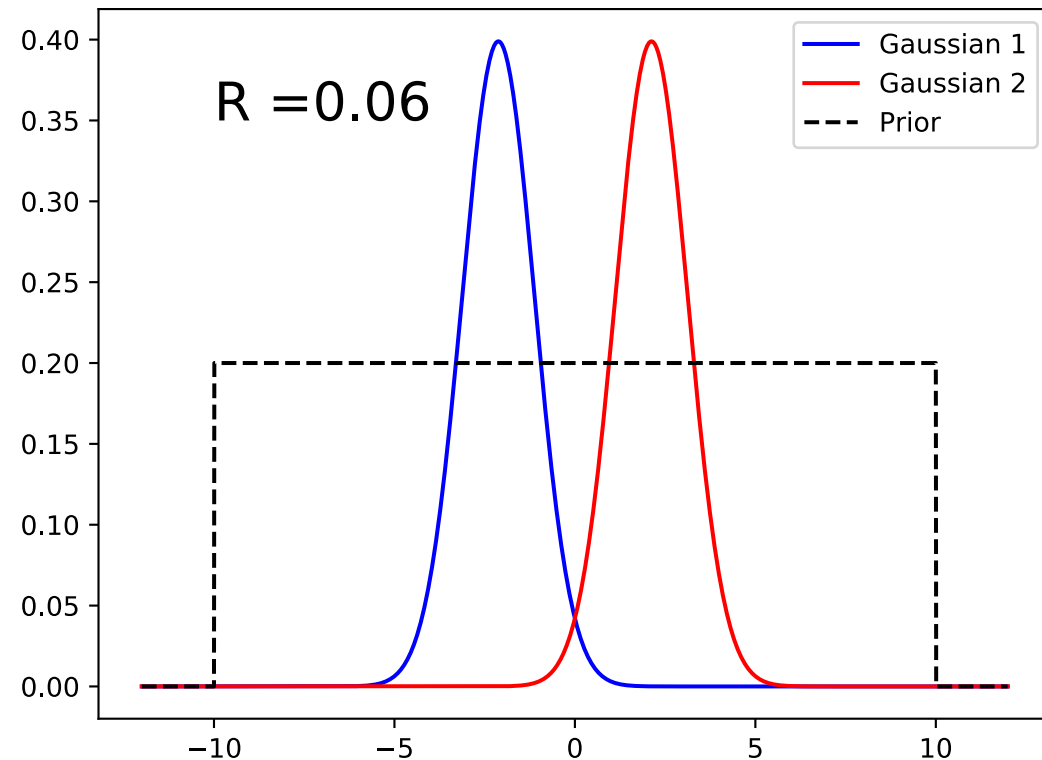




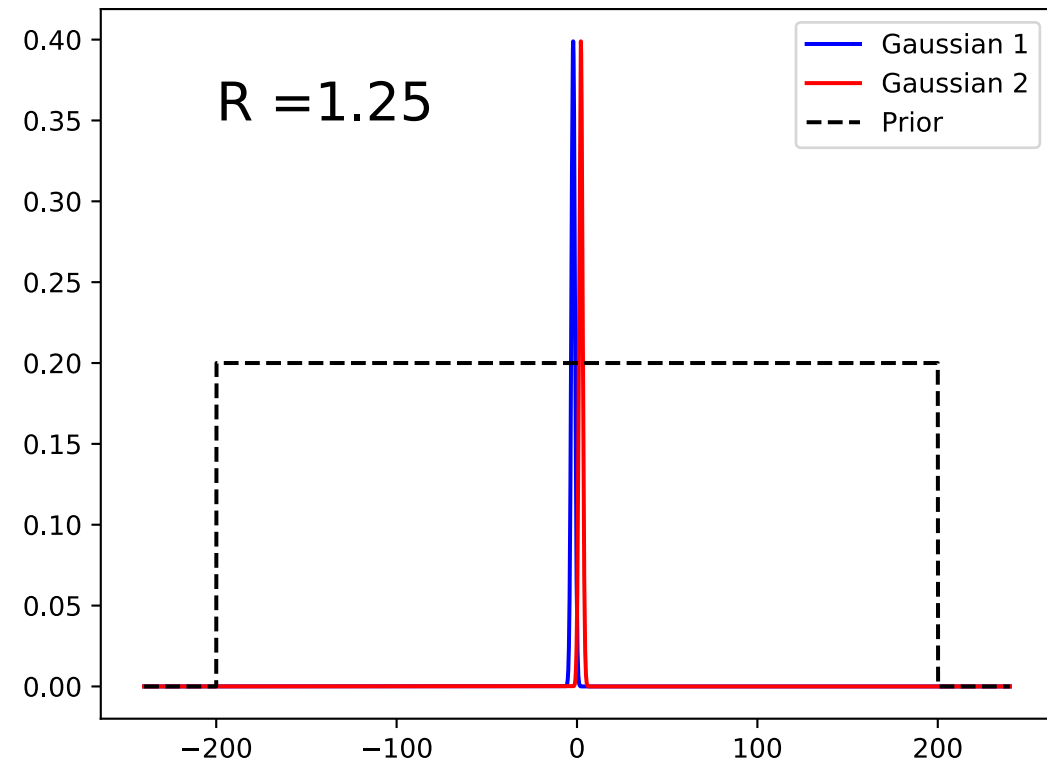
# Bayes Ratio

Toy example: 1D Gaussians:

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



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

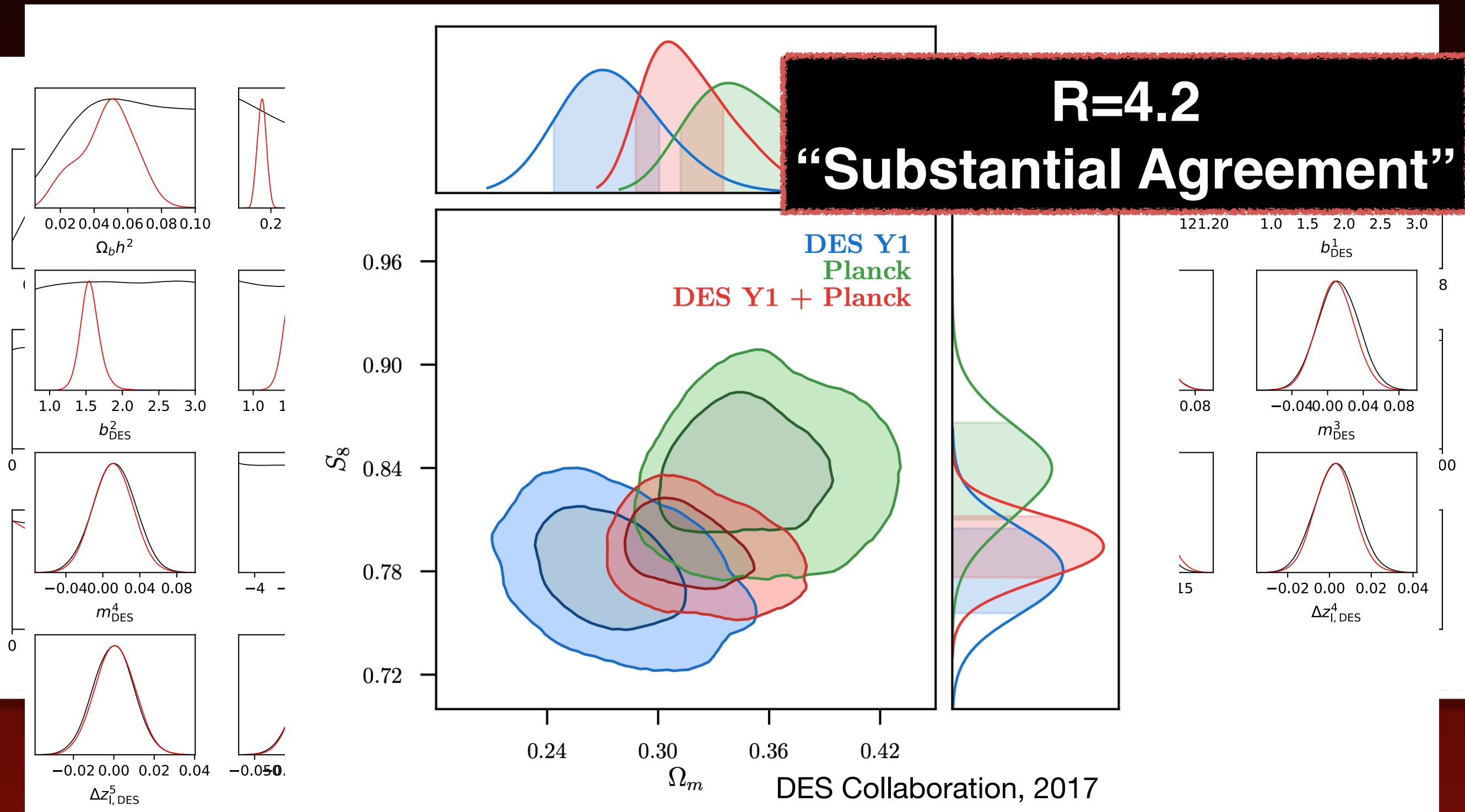


Prior: -200, 200 -> **Concordant**



# 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*



# 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

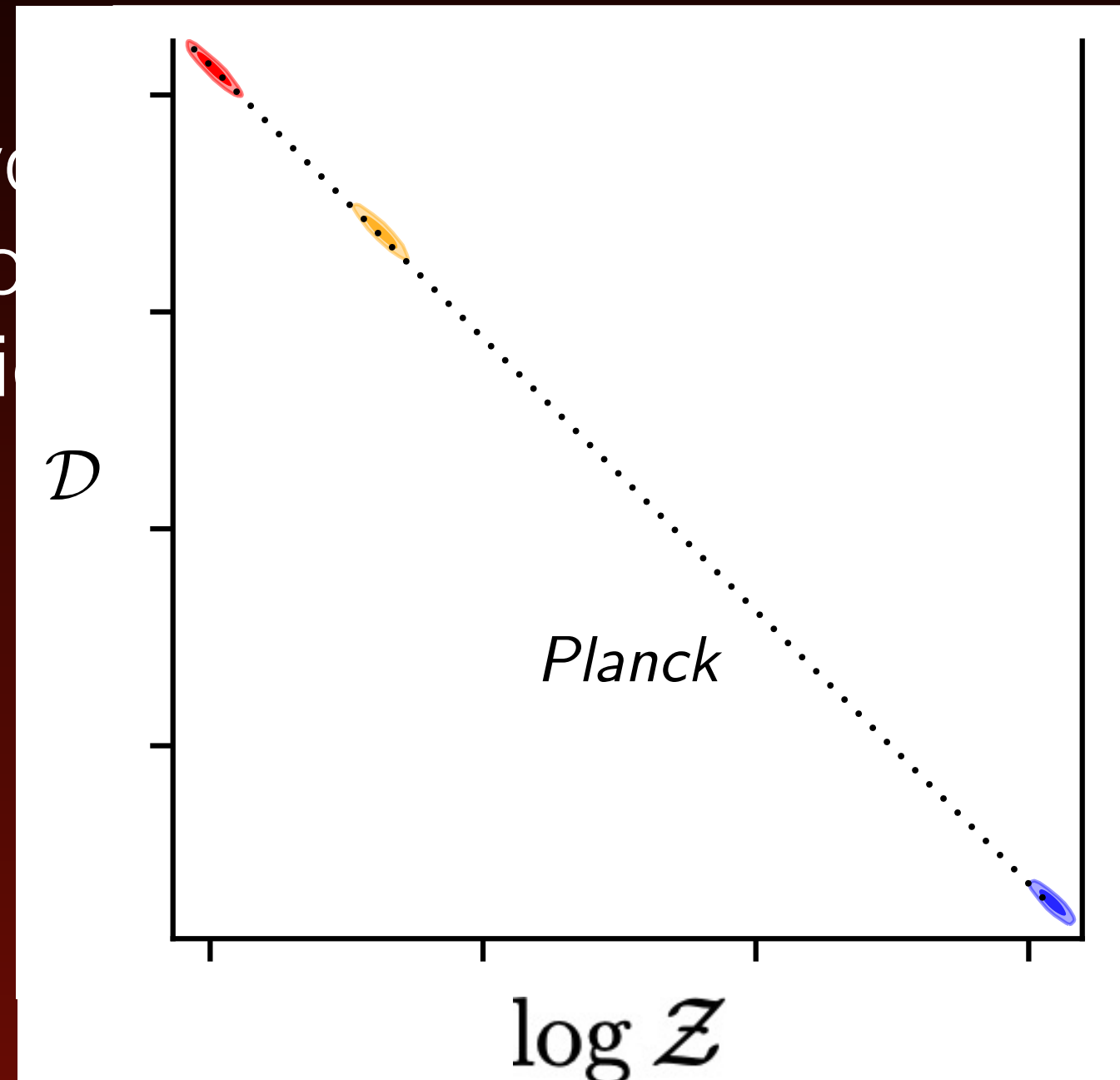
What part of the Bayes Ratio carries the ‘prior volume dependence’?

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

## Kullback-Leibler Divergence

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

*Kullback, Leibler, 1951*  
*doi: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)**:

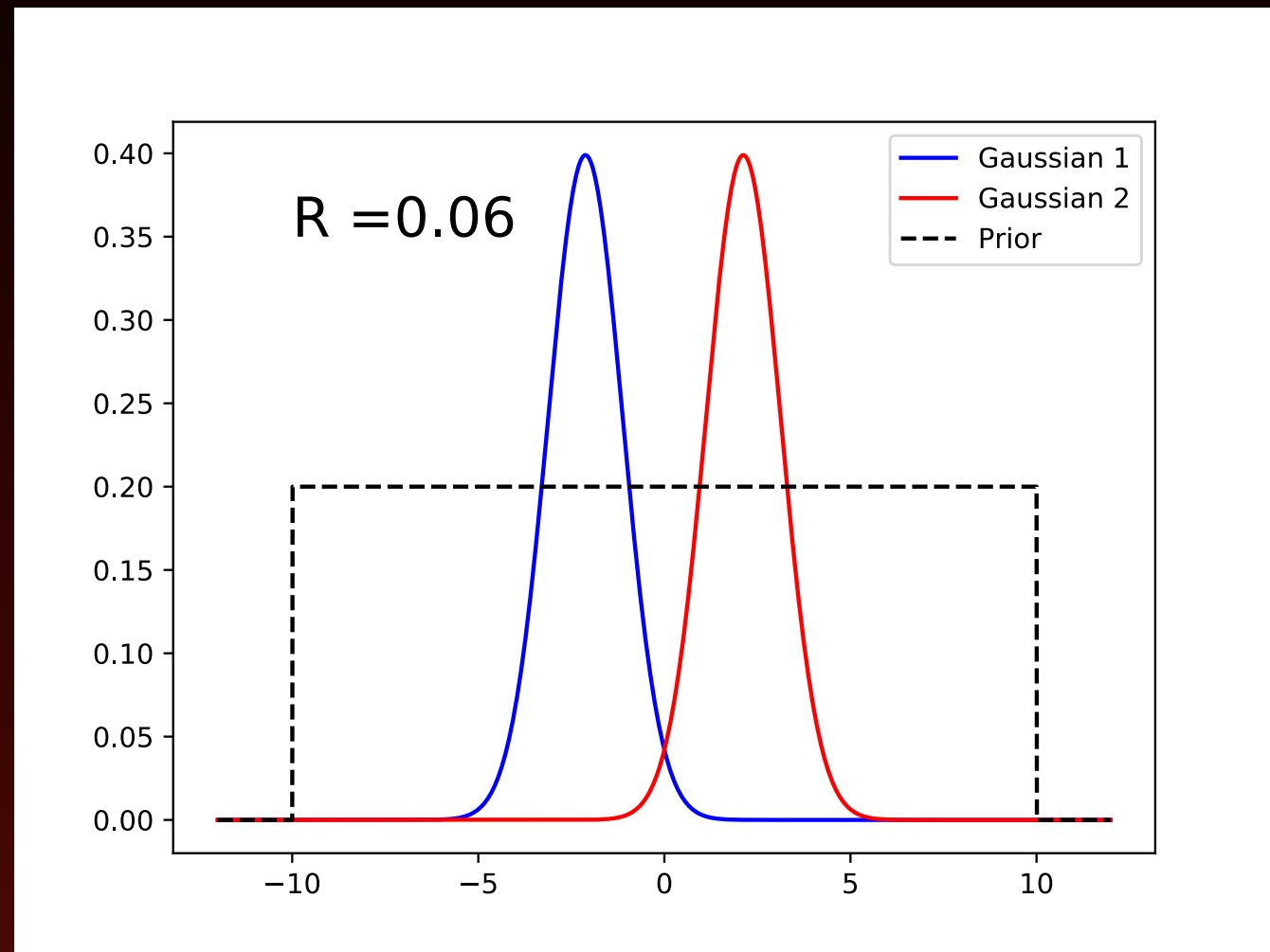
$$\log R = \log I + \log S$$



# 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.



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

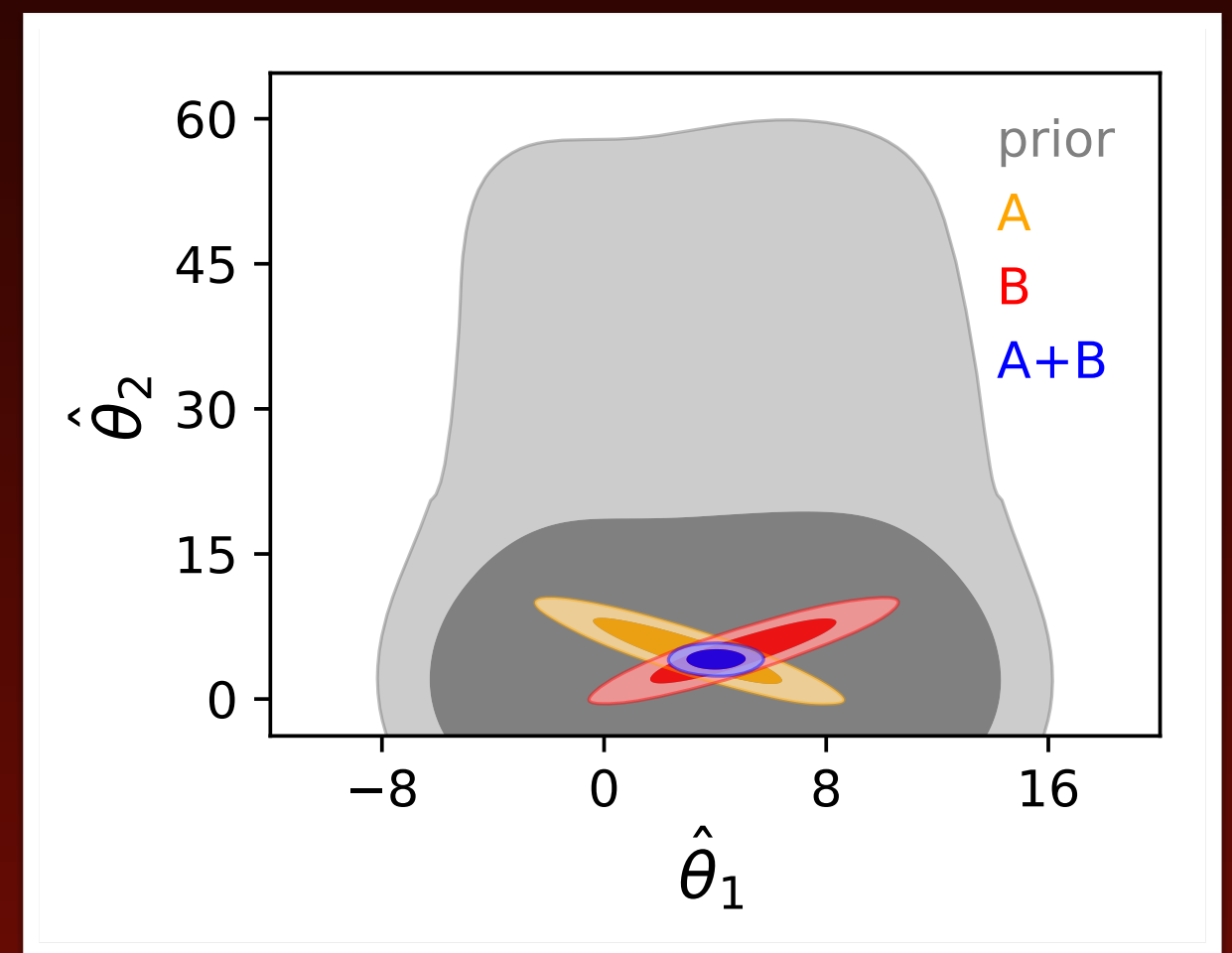
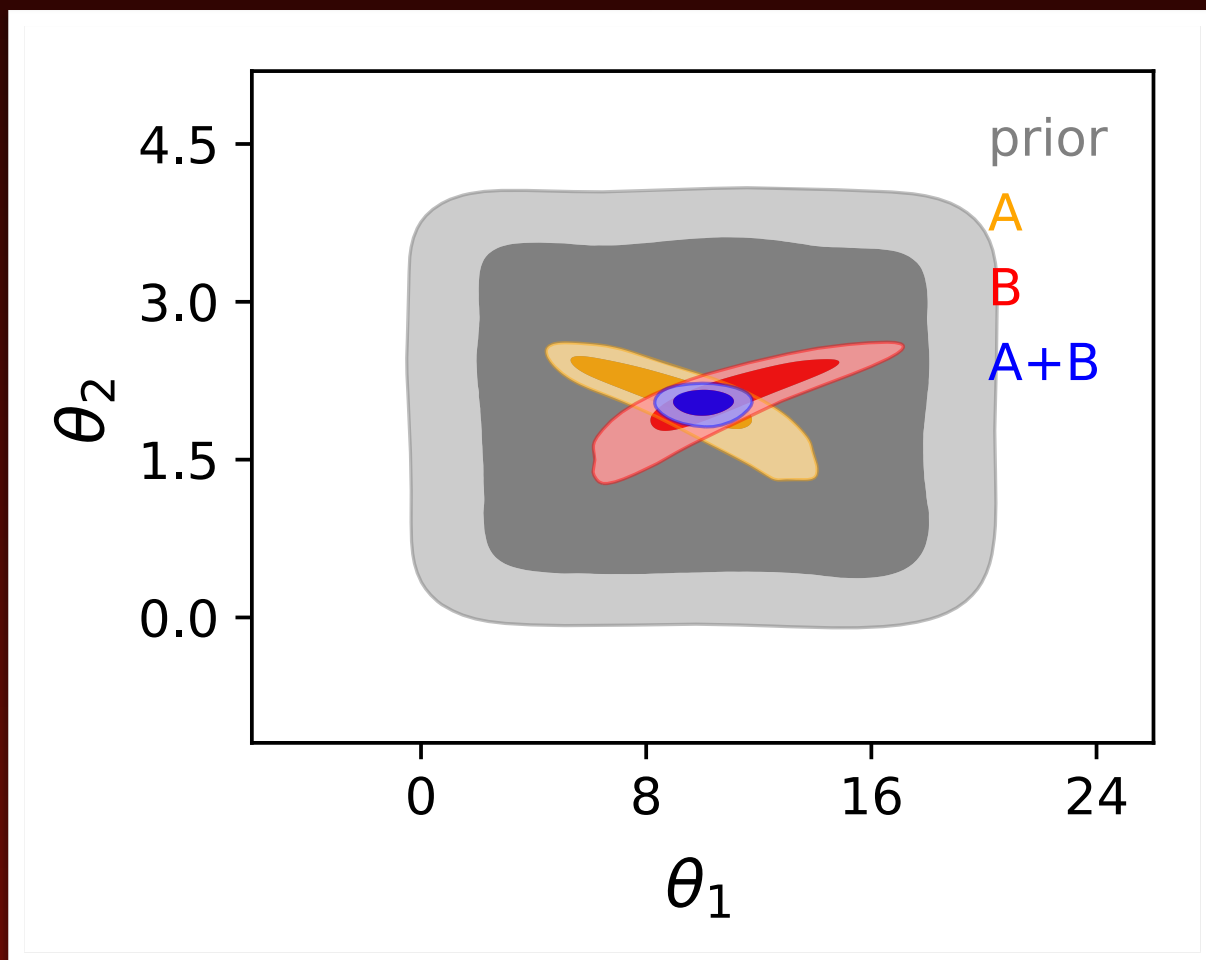




# Suspiciousness

What about non-Gaussian posteriors

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





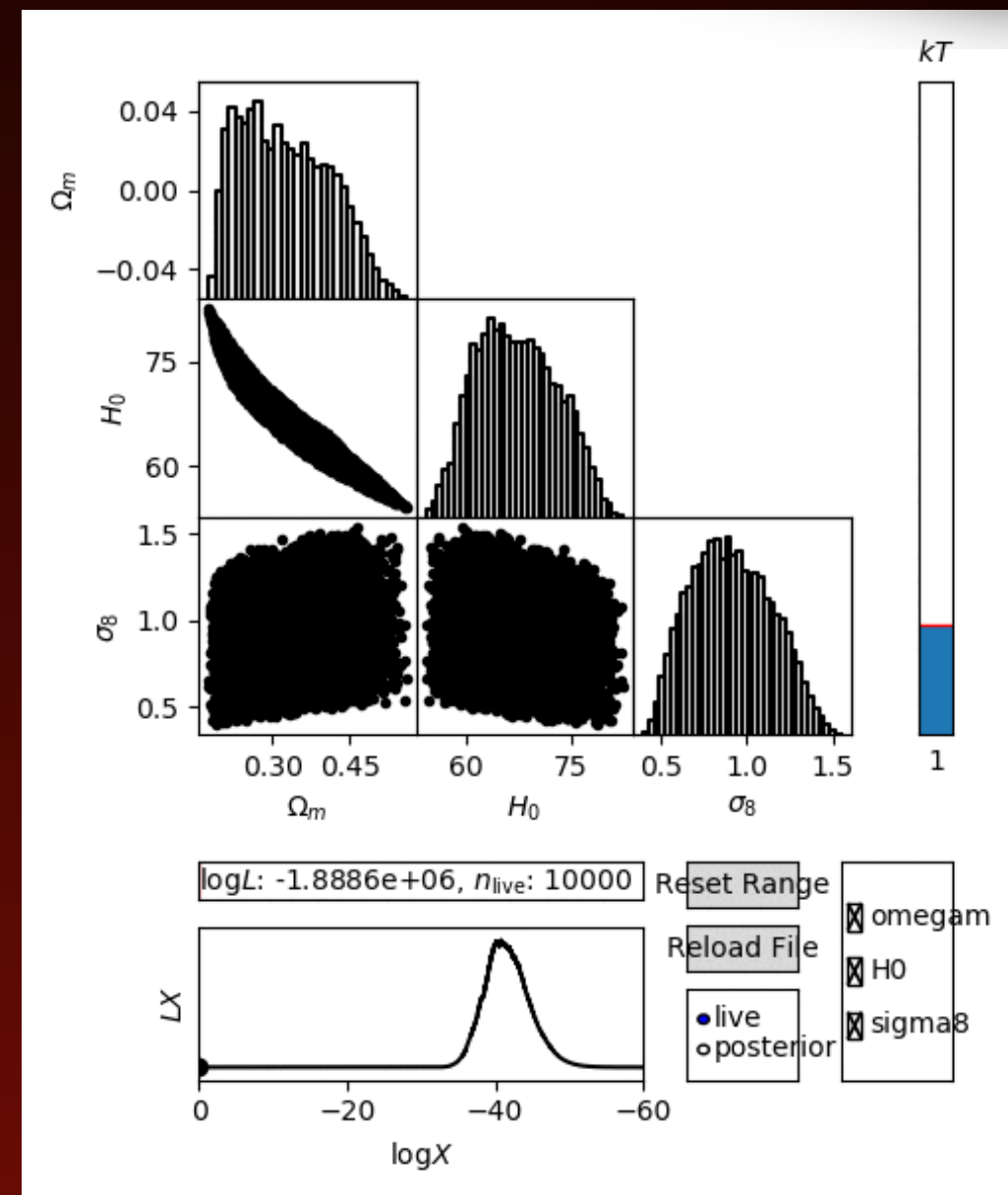
# How to calculate this



# 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





# Application to Cosmology

# How does this work in practice?

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

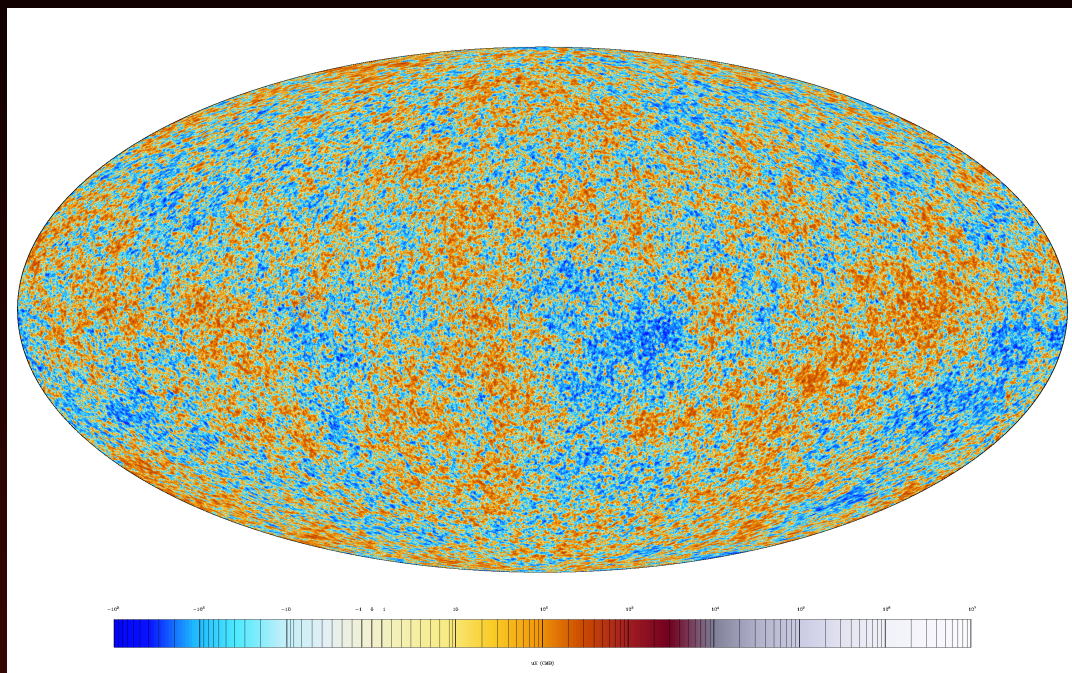


# Application to Cosmology

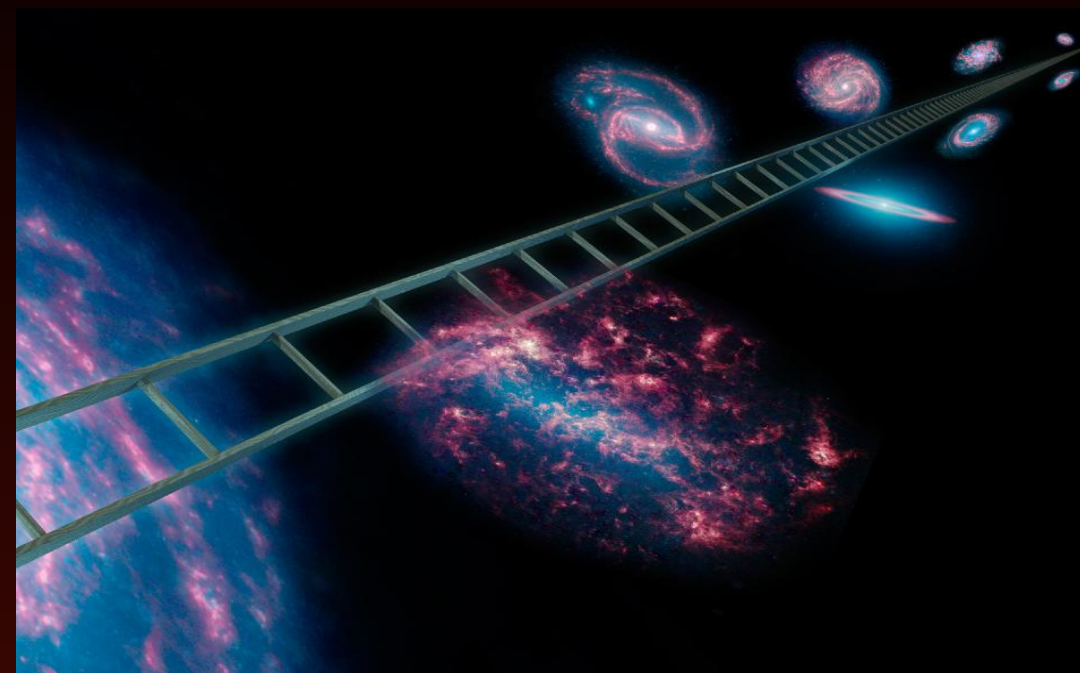




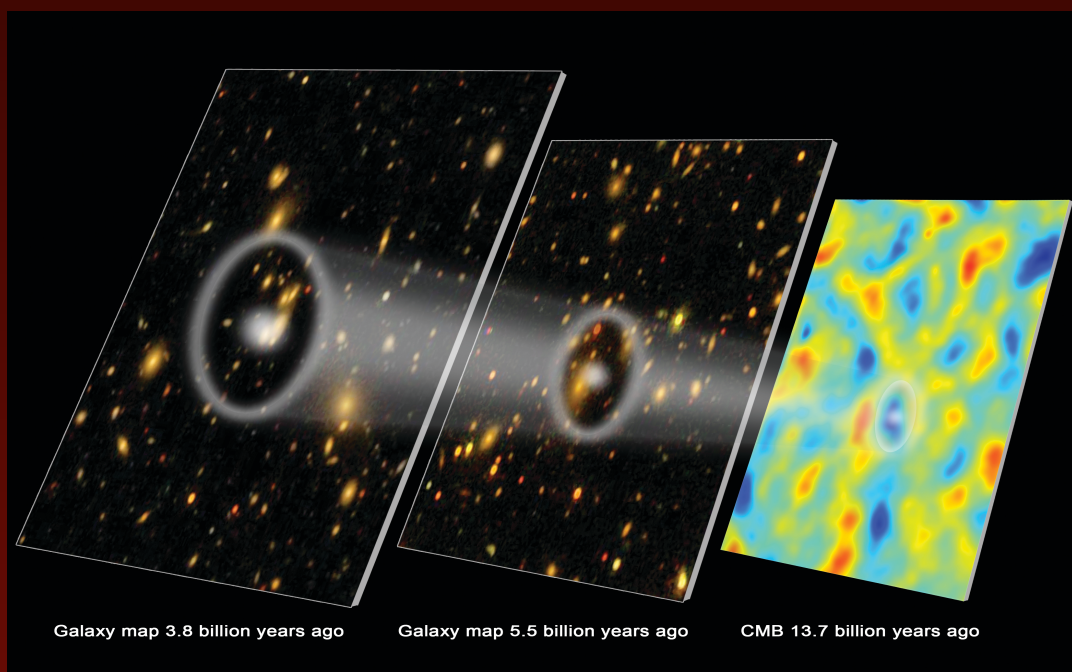
# Application to Cosmology



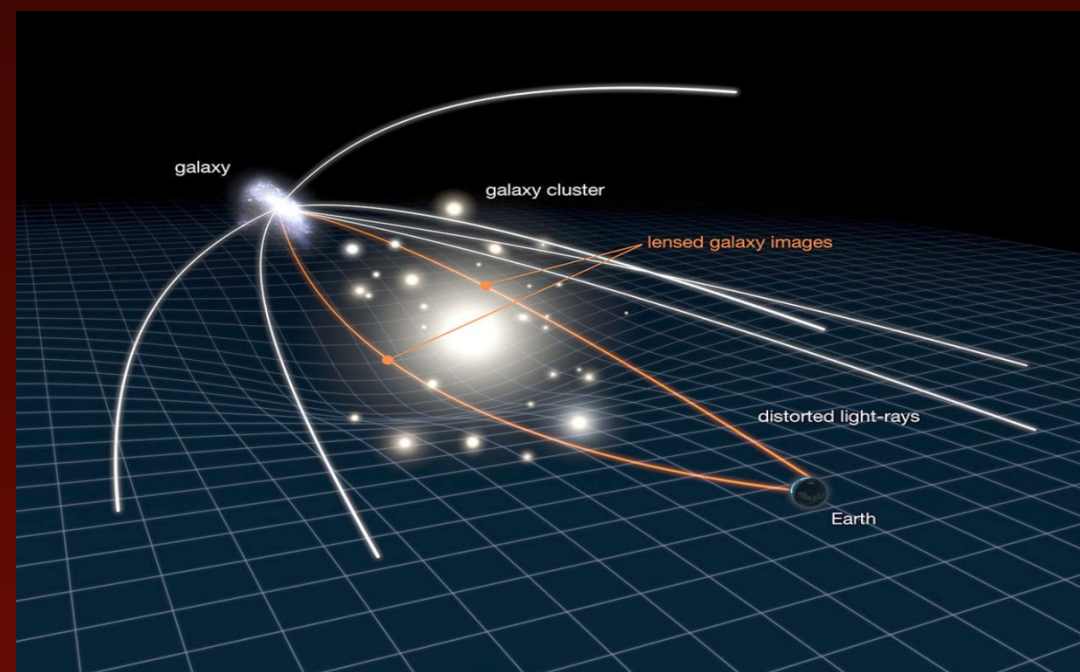
*Cosmic Microwave Background (CMB) - **PLANCK***



*Cosmic Distance Ladder - **SH0ES***



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



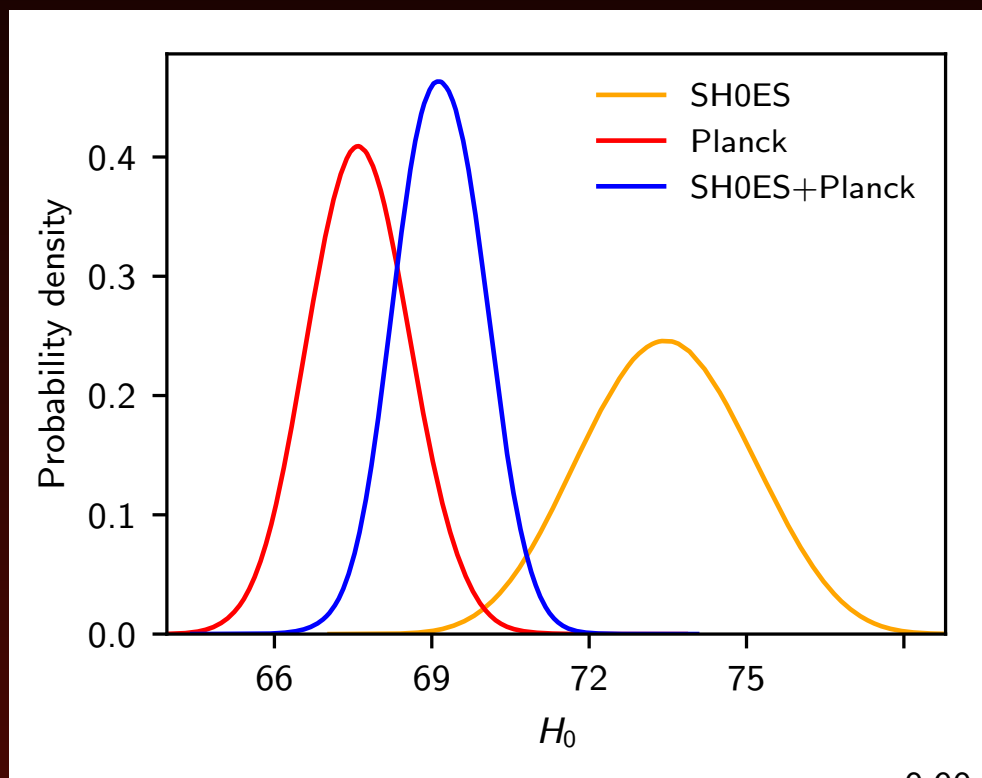
*Weak Galaxy Lensing - **DES***



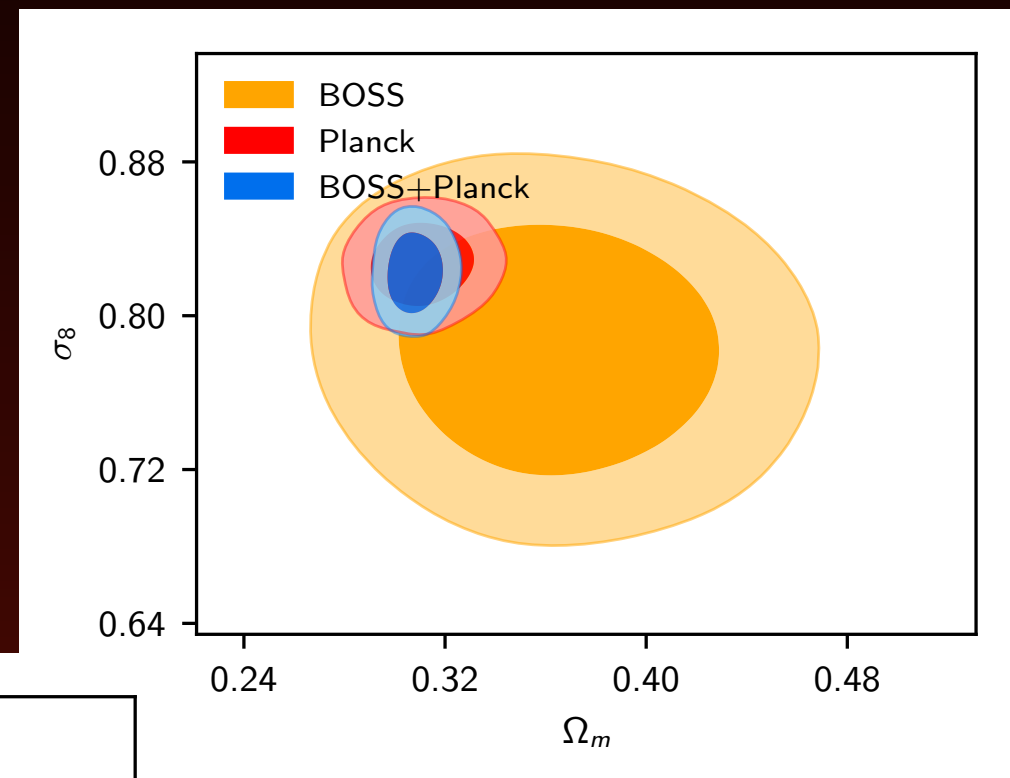
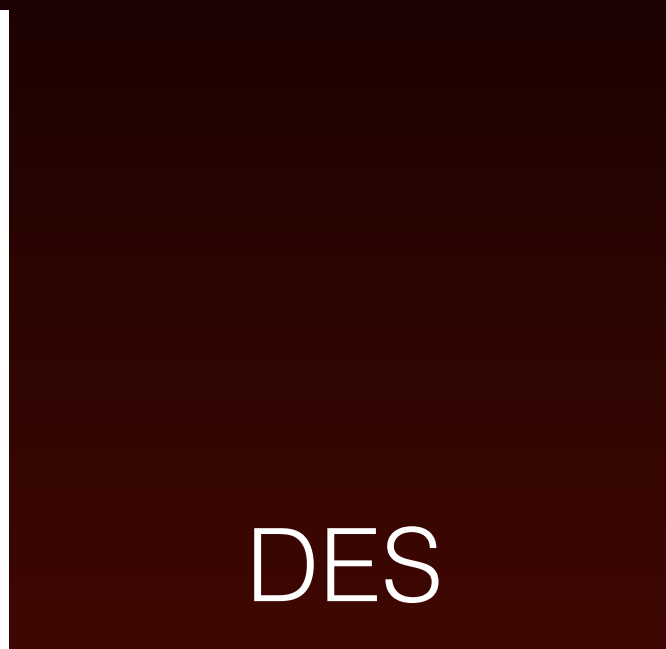


# Application to Cosmology

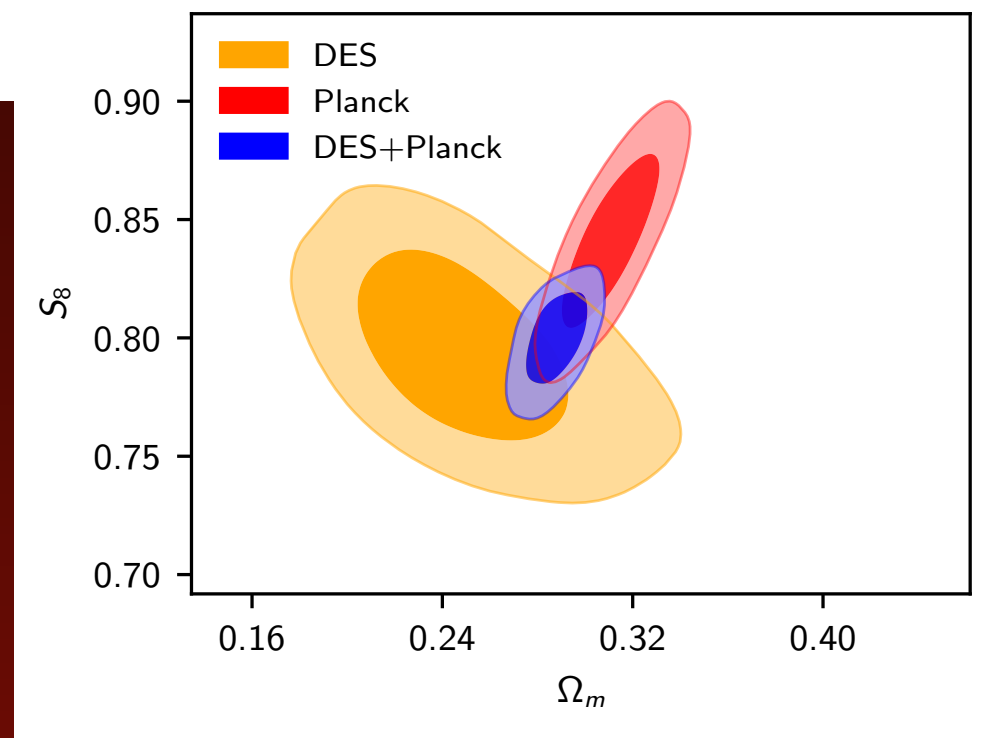
## Planck vs...



SH0ES



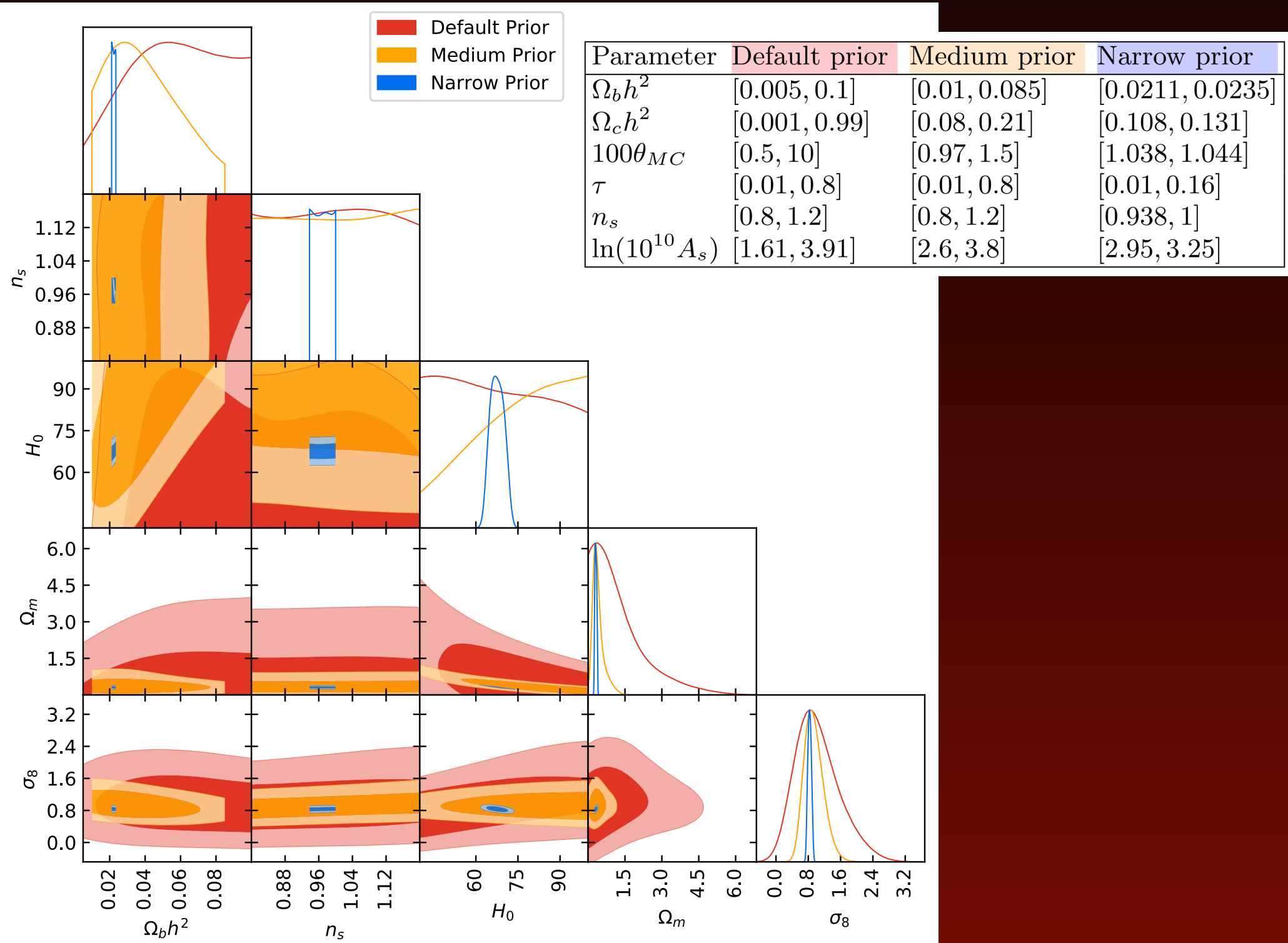
BOSS





# Application to Cosmology

## Three different priors





# Application to Cosmology

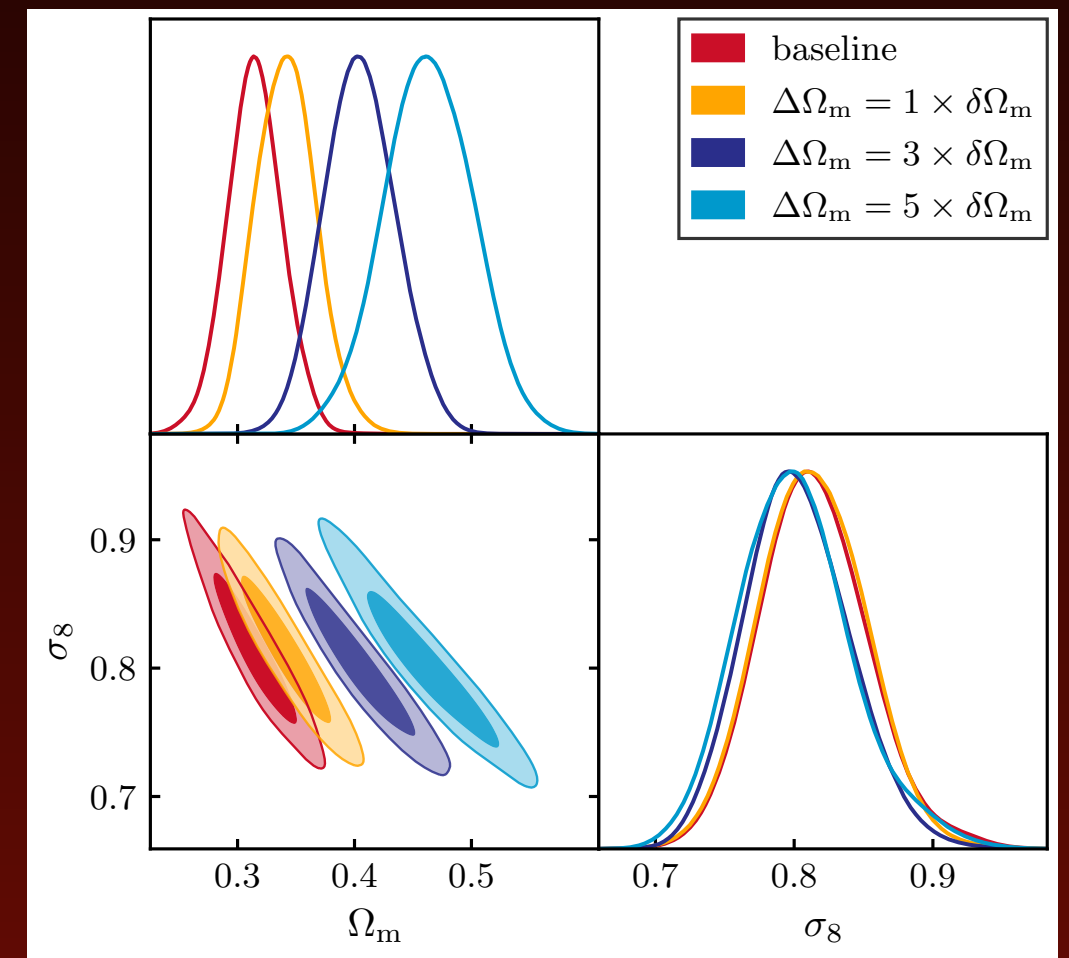
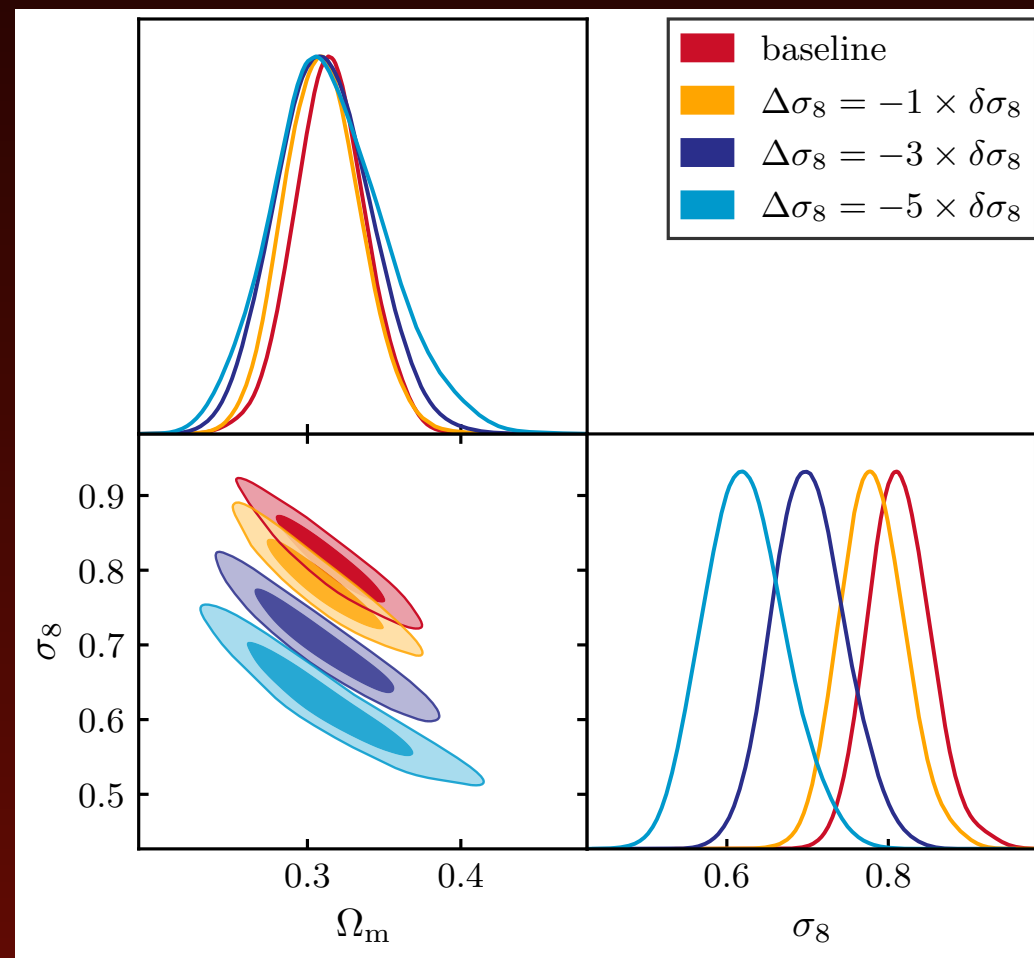
## Results

Dataset	Prior	$\log R$	$\log I$	$\log S$	$d$	$p(\%)$
BOSS- <i>Planck</i>	default	$6.30 \pm 0.29$	$6.18 \pm 0.29$	$0.11 \pm 0.29$	$2.91 \pm 0.51$	$42.66 \pm 4.28$
	medium	$4.51 \pm 0.28$	$4.06 \pm 0.28$	$0.46 \pm 0.28$	$3.30 \pm 0.55$	$55.12 \pm 4.47$
	narrow	$1.30 \pm 0.23$	$0.69 \pm 0.22$	$0.61 \pm 0.22$	$1.67 \pm 0.54$	$77.12 \pm 14.10$
DES- <i>Planck</i>	default	$2.88 \pm 0.35$	$6.15 \pm 0.34$	$-3.28 \pm 0.34$	$3.97 \pm 0.82$	$3.23 \pm 1.00$
	medium	$0.51 \pm 0.34$	$4.00 \pm 0.34$	$-3.49 \pm 0.34$	$3.13 \pm 0.81$	$2.04 \pm 0.79$
	narrow	$-1.88 \pm 0.29$	$0.90 \pm 0.29$	$-2.78 \pm 0.29$	$1.15 \pm 0.77$	$1.44 \pm 0.91$
$SH_0$ ES- <i>Planck</i>	default	$-2.03 \pm 0.29$	$1.96 \pm 0.28$	$-3.99 \pm 0.28$	$0.78 \pm 0.52$	$0.25 \pm 0.17$
	medium	$-2.50 \pm 0.28$	$1.56 \pm 0.28$	$-4.06 \pm 0.28$	$1.77 \pm 0.51$	$0.56 \pm 0.24$
	narrow	$-2.00 \pm 0.23$	$1.43 \pm 0.23$	$-3.43 \pm 0.23$	$1.92 \pm 0.52$	$1.17 \pm 0.45$



# Simulated DES vs Planck

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

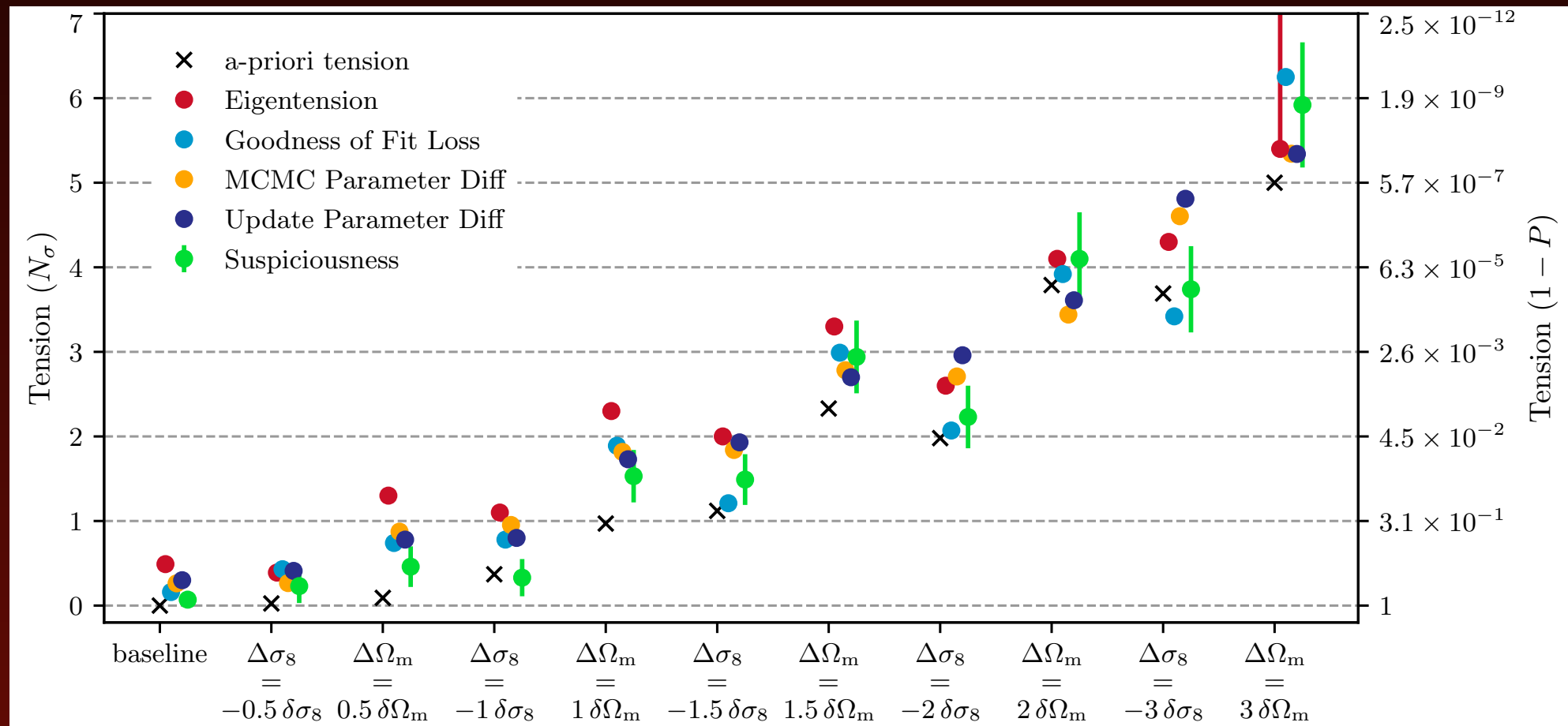


*PL, Raveri & DES Collaboration;*  
[arXiv: 2012.09554](https://arxiv.org/abs/2012.09554)



# 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

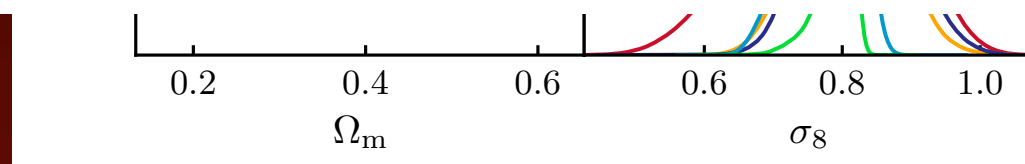


# DES Y1 vs Planck

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	$2.2 \pm 0.5$	Substantial Agreement	$1.8 \sigma$	$1.3 \sigma$	$1.3/1.2 \sigma$	$(0.7 \pm 0.4) \sigma$
<b>DES <math>3 \times 2</math>pt vs. <i>Planck</i> 15</b>	$1.0 \pm 0.5$	No Evidence	$2.4 \sigma$	$2.7 \sigma$	$2.2/2.2 \sigma$	$(2.4 \pm 0.2) \sigma$
DES $5 \times 2$ pt vs. <i>Planck</i> 15	$1.1 \pm 0.5$	Substantial Agreement	$2.4 \sigma$	$2.8 \sigma$	$2.1/2.3 \sigma$	$(2.2 \pm 0.3) \sigma$
DES $5 \times 2$ pt vs. <i>Planck</i> 15 + lensing	$1.0 \pm 0.6$	No Evidence	$2.4 \sigma$	$2.5 \sigma$	$2.1/2.3 \sigma$	$(2.2 \pm 0.4) \sigma$
DES $5 \times 2$ pt + <i>Planck</i> lensing vs. <i>Planck</i> 15	$6.1 \pm 0.6$	Strong Agreement	$1.6 \sigma$	$2.4 \sigma$	$1.9/2.2 \sigma$	$(1.8 \pm 0.2) \sigma$
DES cosmic shear vs. <i>Planck</i> 18	$3.3 \pm 0.4$	Strong Agreement	$1.5 \sigma$	$1.0 \sigma$	$1.0/1.1 \sigma$	$(0.5 \pm 0.3) \sigma$
<b>DES <math>3 \times 2</math>pt vs. <i>Planck</i> 18</b>	$2.2 \pm 0.6$	Substantial Agreement	$2.2 \sigma$	$1.6 \sigma$	$2.0/2.3 \sigma$	$(2.4 \pm 0.2) \sigma$



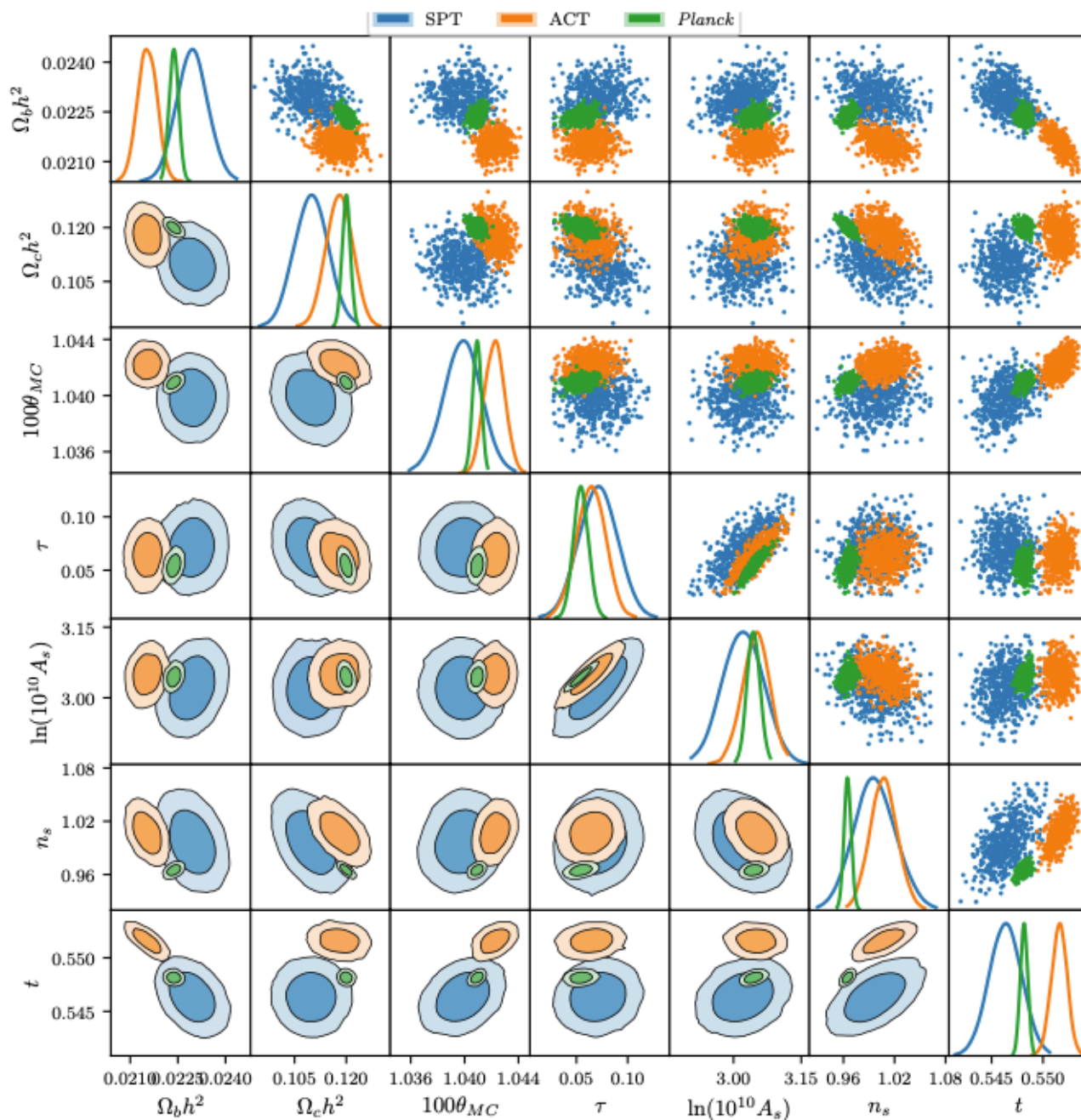
*PL, Raveri & DES Collaboration;*  
arXiv: 2012.09554





# Application to Cosmology

# Application to the CMB



Dataset combination	$\chi^2$	$p$	tension	$\log S$
ACT vs <i>Planck</i>	17.2	0.86%	$2.63\sigma$	-5.60
ACT vs SPT	15.4	1.77%	$2.37\sigma$	-4.68
<i>Planck</i> vs SPT	9.1	16.82%	$1.38\sigma$	-1.55
ACT vs <i>Planck</i> +SPT	18.4	0.52%	$2.79\sigma$	-6.22
ACT+SPT vs <i>Planck</i>	12.2	5.81%	$1.90\sigma$	-3.09
ACT+ <i>Planck</i> vs SPT	10.3	11.09%	$1.59\sigma$	-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<sup>1,2\*</sup>, Tilman Tröster<sup>1\*\*</sup>, Marika Asgari<sup>1</sup>, Chris Blake<sup>3</sup>, Hendrik Hildebrandt<sup>2</sup>, Benjamin Joachimi<sup>4</sup>, Konrad Kuijken<sup>5</sup>, Chieh-An Lin<sup>1</sup>, Ariel G. Sánchez<sup>6</sup>, Jan Luca van den Busch<sup>2</sup>, Angus H. Wright<sup>2</sup>, Alexandra Amon<sup>7</sup>, Maciej Bilicki<sup>8</sup>, Jelte de Jong<sup>9</sup>, Martin Crocce<sup>10,11</sup>, Andrej Dvornik<sup>2</sup>, Thomas Erben<sup>12</sup>, Maria Cristina Fortuna<sup>5</sup>, Fedor Getman<sup>13</sup>, Benjamin Giblin<sup>1</sup>, Karl Glazebrook<sup>3</sup>, Henk Hoekstra<sup>5</sup>, Shahab Joudaki<sup>14</sup>, Arun Kannawadi<sup>15,5</sup>, Fabian Köhlinger<sup>2</sup>, Chris Lidman<sup>16</sup>, Lance Miller<sup>14</sup>, Nicola R. Napolitano<sup>17</sup>, David Parkinson<sup>18</sup>, Peter Schneider<sup>12</sup>, HuanYuan Shan<sup>19,20</sup>, Edwin A. Valentijn<sup>9</sup>, Gijs Verdoes Kleijn<sup>9</sup>, and Christian Wolf<sup>16</sup>

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  $0.08 \pm 0.02$ , 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:

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



# 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.