## Scanning for cosmological tensions across a DiRAC-enabled grid of models, datasets and samplers (with AI coda)

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21<sup>st</sup> May 2025







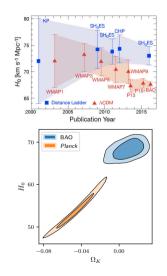




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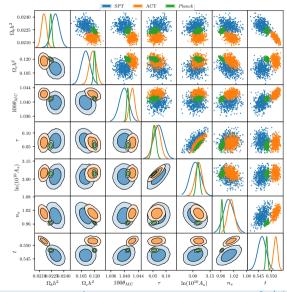
# The Challenge: Cosmological Tensions & Analysis Robustness

- Precision cosmology has revealed potential discrepancies in key parameters:
  - ► *H*<sub>0</sub> (Hubble constant) [1907.10625]
  - Ω<sub>K</sub> (spatial curvature) [1908.09139], [1911.02087]
  - $\sigma_8/S_8$  (matter clustering) [1610.04606]?
- This underscores the critical importance of robust cosmological analysis.
- Key goals:
  - Accurately quantifying tensions between datasets and models.
  - Identifying and addressing biases, e.g., ACDM bias from fiducial assumptions in likelihoods.
- We need systematic exploration across a wide range of models and datasets to distinguish new physics from systematics.



## The importance of global tension metrics

- If you have a well-defined parameter like H<sub>0</sub>, this can be a good proxy for assessing the consistency of two fits
- However, in e.g. Planck vs DES we're asking: "In this 6 dimensional space, how well does this 6d grape fit with this 3d banana/blanket"?
- It's easy to make things look worse than they are when there are 6 degrees of freedom
- Inappropriate to pick a single parameter and quote a "sigma"
- Many methods have been developed to quantify "global" tensions [1902.04029].



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# **Bayesian Inference Pillars & The Problem of Scale**

**Parameter Estimation** 

What do data tell us about model parameters?

 $\mathcal{P}(\theta|D, M) = \frac{\mathcal{L}(D|\theta, M)\pi(\theta|M)}{\mathcal{Z}(D|M)}$ 

#### **Model Comparison**

How much do data support a model? (Occam's Razor [2102.11511])

$$P(M|D) = \frac{\mathcal{Z}(D|M)P(M)}{P(D)}$$

**Tension Quantification** 

Are datasets consistent under a model? [1902.04029]

$$\mathcal{R} = \frac{\mathcal{Z}_{AB}}{\mathcal{Z}_{A}\mathcal{Z}_{B}}, \quad \log \mathcal{S} = \dots$$

Also evidence-based.

 Model comparison and tension quantification are computationally hard, especially across many models/datasets.

Relies on the Bavesian

Evidence  $\mathcal{Z}$ .

This necessitates efficient tools and pre-computed resources.

# The Planck legacy archive

- Planck collaboration science products
- Distributed cosmology inference results as MCMC chains
- Across a grid of:
  - subsets/combinations of *Planck* data
    - TT, lowl, lowE, lensing
  - ACDM extensions
    - base, mnu, nrun, omegak, r
- importance sampling across some other likelihoods (BAO, JLA,...)
- Cannot compute evidences in high dimensions from MCMC chains
  - Only parameter estimation
  - no model comparison



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# **DiRAC Allocations: Powering the Systematic Search**

#### DiRAC 13: Next Gen Cosmo Analysis

- Goal: Create a Planck Legacy Archive (PLA) equivalent, but using nested sampling.
  - Enabling model comparison & tension quantification.
- Systematic scan over models and modern datasets (Planck, DES, SNe).
- A public grid of nested sampling & MCMC chains.

### DiRAC 17: New Horizons in Cosmology

- Extending DiRAC 13 success.
- Incorporating next-gen data: DESI, DES Y5 SNe, Pantheon+, KiDS-1000, HSC, Euclid forecasts.



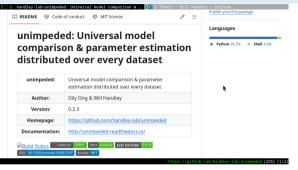
# Introducing unimpeded



Universal Model comparison and Parameter Estimation Distributed over Every Dataset

- Core Idea: A re-usable library of MCMC chains, Nested Sampling runs, and ML emulators. [Abstract]
- Built from DiRAC allocations (DP192 & DP264).
- Systematic Coverage:
  - Current: ~10 cosmological models (e.g., ΛCDM, kΛCDM, wΛCDM).
  - $\blacktriangleright$  ~60 datasets & pairwise combinations (CMB, BAO, SNe, WL).
- Python tool for seamless download, upload, and caching of analysis products.
- Data stored on Zenodo; fast HDF5 local caching.

```
<wh260@cam.ac.uk> willhandley.co.uk/talks
```



github.com/handley-lab/unimpeded

## unimpeded in Action: Easy Access to Complex Results

### Key Features & Benefits:

- Easily access pre-computed results with a few lines of Python.
- Facilitates:
  - Parameter estimation.
  - Model comparison (via nested sampling evidences).
  - Tension quantification.
  - Pairwise dataset comparisons.
- Removes computational barriers for many common analyses.
- Enables robust science by allowing checks over many models and data combinations.
- Foundation for building more complex, multi-probe analyses.
   <wh260@cam.ac.uk>

```
1 from unimpeded.database import
      DatabaseExplorer
_2 dbe = DatabaseExplorer()
_{3} planck = dbe.download_samples(
      method='ns', model='walcdm'.
      dataset='planck_2018_CamSpec')
6 sdss = dbe.download_samples(
      method='ns', model='walcdm'.
      dataset='bao.sdss_dr16')
8
  planck_sdss = dbe.download_samples(
9
      method='ns', model='walcdm',
10
11
      dataset='bao.sdss_dr16+
      planck_2018_CamSpec')
```

## unimpeded preliminary results

- Models on y-axis
- Dataset combinations on x axis
- numbers and colours refer to σ-values for their tension
- Lots to unpack here (a very expensive plot)!

ACDM	0.31 ±0.11	0.08 ±0.26	1.29 ±0.08	2.78 ±0.12	2.58 ±0.10	2.78 ±0.17	2.97 ±0.18	0.24 ±0.49	0.93 ±0.06	0.92 ±0.07	0.98 ±0.07	1.01 ±0.08	1.15 ±0.08	0.04 ±0.08	0.67 ±0.10	0.37 ±0.39	0.62 ±0.39	0.09 ±0.20	1.11 ±0.40	0.00 ±0.00	2.61 ±0.13	2.39 ±0.11	2.57 ±0.14	2.44 ±0.11	$^{1.41}_{\pm 0.08}$	0.50 ±0.24	0.73 ±0.10	0.18 ±0.29	0.91 ±0.25	- 3.5
KACDM	0.12 ±0.09	0.00 ±0.00	1.38 ±0.19	2.47 ±0.11	2.84 ±0.13	2.90 ±0.22	3.27 ±0.10	1.31 ±0.24	0.43 ±0.06					0.19 ±0.06		0.37 ±0.11	0.56 ±0.20	0.22 ±0.41	0.82 ±0.08	0.00 ±0.00	2.93 ±0.12	3.19 ±0.12	3.19 ±0.15	3.57 ±0.10	1.05 ±0.32		2.33 ±0.21	1.61 ±0.31	2.44 ±0.12	- 3.0
NACDM	0.01 ±0.04	0.00 ±0.00	1.16 ±0.08		1.79 ±0.14	2.69 ±0.31		0.01 ±0.05	0.59 ±0.06	0.65 ±0.07	0.51 ±0.07	0.59 ±0.06	0.60 ±0.07	0.23 ±0.12	0.74 ±0.12	0.21 ±0.31	0.32 ±0.13	0.56 ±0.12	0.22 ±0.24	0.00 ±0.00	2.55 ±0.14			2.54 ±0.12	0.98 ±0.09	0.19 ±0.24	0.15 ±0.22	0.35 ±0.25	0.58 ±0.18	- 2.5
mΛCDM	0.05 ±0.10	0.24 ±0.60	1.15 ±0.07				2.94 ±0.23	1.88 ±0.53	0.82 ±0.05	0.81 ±0.06	0.75 ±0.07			0.24 ±0.13	0.63 ±0.07	0.68 ±0.22	0.33 ±0.30	0.68 ±0.20	0.16 ±0.33	0.01 ±0.09					1.25 ±0.08	0.62 ±0.17	0.59 ±0.33	0.95 ±0.22	0.69 ±0.50	£
nrun∧CDM	0.00 ±0.03	0.01 2±0.06	1.17 ±0.24		2.94 ±0.17	2.94 ±0.19	2.98 ±0.23	0.19 ±0.46	0.32 ±0.09	0.88 ±0.06	1.04 ±0.08			0.08 ±0.11	0.32 ±0.28	0.34 ±0.17	0.67 ±0.29	0.69 ±0.20	0.79 ±0.30	0.00 ±0.00			2.84 ±0.23	3.10 ±0.32	1.10 ±0.19	0.56 ±0.27	0.08 ±0.20	0.56 ±0.12	0.88 ±0.11	- 2.0
rACDM	±0.10	0.00 ±0.03		±0.12	±0.15	±0.20	±0.20	±0.06	±0.06	±0.07	±0.06	±0.20	±0.09	±0.09	±0.14	±0.06	±0.03	±0.00	±0.06	±0.00	±0.22		±0.21	±0.16		±0.22	±0.17	±0.11	±0.30	- 1.5
wACDM	±0.05		±0.06			±0.14	±0.24	±0.04	±0.04	±0.08	±0.08	±0.12	±0.15	±0.04	±0.11	±0.28	±0.18	±0.15	±0.23	±0.07			±0.10		±0.07	±0.11		±0.28	±0.30	- 1.0
nΛCDM	±0.11	0.27 ±0.58	±0.09					±0.02	±0.04	±0.05	±0.05	±0.08	±0.05	±0.09	±0.10	±0.26	±0.22	±0.35	±0.24	±0.00		±0.14			±0.07	±0.19	±0.32	±0.12	±0.19	- 1.0
w <sub>0</sub> w <sub>a</sub> ACDM	±0.01	0.00 ±0.00	±0.16			±0.15		±0.00	±0.05					±0.05	±0.19	±0.36	±0.09		±0.14	±0.09						±0.30	±0.38	±0.29	±0.32	- 0.5
AACDM	0.17 ±0.14	0.11 ±0.39	1.50 ±0.12	2.25 ±0.11	2.24 ±0.12	2.45 ±0.24	2.29 ±0.24	0,00 ±0.00	1.09 ±0.07	1.00 ±0.07	$\frac{1.03}{\pm 0.09}$	0.94 ±0.08	0.95 ±0.08	0.03 ±0.08	1,17 ±0.28	0.40 ±0.27	0.00 ±0.03	0.20 ±0.25	0.25 ±0.33	0.00 ±0.00	2.06 ±0.10	1.90 ±0.09	1,86 ±0.09	$1.90 \pm 0.11$	1.59 ±0.13	0.46 ±0.14	0.04 ±0.11	0.09 ±0.17	0.37 ±0.18	- 0.0
	H0.riess2020Mb+bao.sdss_dr16 -	H0.riess2020Mb+bicep_keck_2018	H0.riess2020Mb+des_y1.joint	H0.riess2020Mb+planck_2018_CamSpec -	H0.riess2020Mb+planck_2018_CamSpec_nolens	H0.riess2020Mb+planck_2018_plik	H0.riess2020Mb+planck_2018_plik_nolens -	bao.sdss_dr16+bicep_keck_2018 -	bao.sdss_dr16+des_y1.joint	bao.sdss_dr16+planck_2018_CamSpec -	bao.sdss_dr16+planck_2018_CamSpec_nolens -	bao.sdss_dr16+planck_2018_plik	bao.sdss_dr16+planck_2018_plik_nolens	bao.sdss_dr16+sn.pantheon	bicep_keck_2018+des_y1.joint -	bicep_keck_2018+planck_2018_CamSpec -	bicep_keck_2018+planck_2018_CamSpec_nolens -	bicep_keck_2018+planck_2018_plik	bicep_keck_2018+planck_2018_plik_nolens	bicep_keck_2018+sn.pantheon	des_y1.joint+planck_2018_CamSpec -	des_y1.joint+planck_2018_CamSpec_nolens -	des_y1.joint+planck_2018_plik	des_y1.joint+planck_2018_plik_nolens	des_y1.joint+sn.pantheon	planck_2018_CamSpec+sn.pantheon -	planck_2018_CamSpec_nolens+sn.pantheon -	planck_2018_plik+sn.pantheon	planck_2018_plik_nolens+sn.pantheon -	

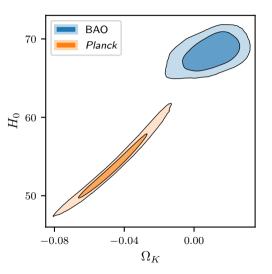
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## Addressing Biases: Beyond **\CDM** Fiducial Assumptions

- Many cosmological likelihoods (e.g., CMB lensing, BAO reconstruction) rely on fiducial ACDM assumptions for:
  - Simulation calibrations.
  - Theoretical templates.
  - Data correction algorithms.
- This can introduce an inherent bias towards ACDM or affect parameter inferences when testing extended models.
- Example: Curvature constraints from CMB lensing or BAO are derived assuming an underlying flat fiducial cosmology. This can mask or create artificial tensions.
- ▶ Robustly testing beyond-ACDM physics requires assessing the impact of these assumptions.
- unimpeded's grid, with varied models and datasets, helps explore these effects without being locked into one fiducial for all analyses.

# **Curvature tension** $\Omega_K$

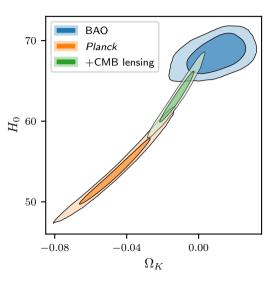
- ACDM assumes the universe is flat
- If you allow  $\Omega_K \neq 0$ , *Planck* (plikTTTEEE) has a moderate preference for closed universes (50:1)
- Planck+CMB lensing +BAO strongly prefer a flat universe
- But, *Planck* vs lensing is  $2.5\sigma$  in tension, and Planck vs BAO is  $3\sigma$ .
- $\blacktriangleright$  This is reduced if plik  $\rightarrow$  camspec
  - Di Valentino et al [1911.02087]
  - Handley [1908.09139]
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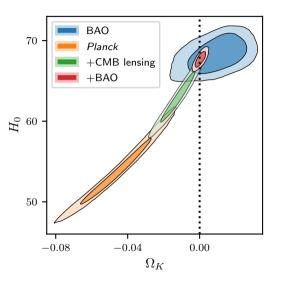
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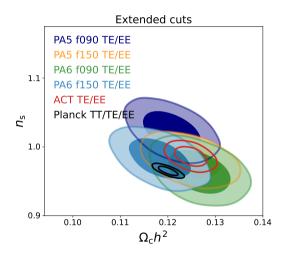


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## A Note on Blinding: Aspirations vs. Reality

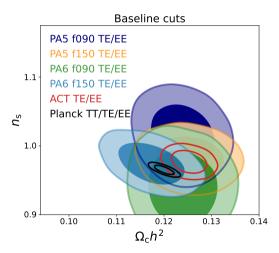
- Blinding is a common strategy to mitigate confirmation bias in cosmological analyses.
- However, the ideal of a pure "one-shot" unblinding is often challenged in practice.
- Examples from recent large analyses (e.g., DES Y3, ACT DR4/DR6) show that post-unblinding adjustments or re-evaluations do occur.
  - These might be necessary due to unforeseen complexities or subtle effects.
  - But they invalidate the intended benefits of the blinding protocol.
- Openly available chains and analysis tools (like unimpeded) can aid in community scrutiny and reproducibility, whether analyses are blinded or not.

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# Machine Learning Enhancement: Emulators

Harry Bevins

- ▶ The unimpeded framework includes providing machine learning emulators. [Abstract]
- These emulate marginalised likelihoods or posteriors.
- Based on techniques like Masked Autoregressive Flows (MAFs) (e.g., as in margarine [2205.12841], [2207.11457]).
- Benefits:
  - Dramatically speeds up inference for re-use in new analyses.
  - Provides a fast and flexible alternative to full MCMC/NS runs for many applications.
  - ▶ Enables easy application of, e.g., a "Planck prior" without running Planck likelihoods.
- ▶ Part of the DiRAC 17 vision: leveraging advanced SBI and ML.

# The Future: Next-Gen Tools AI in Cosmological Robustness

### Simulation-Based Inference (SBI)

- Shifts focus from explicit likelihoods to forward simulations. [ERC Grant, DiRAC 17 CfS]
- Powerful for complex, non-linear probes and intractable likelihoods.

### BlackJAX Nested Sampling

- New JAX-based nested sampling algorithm.
- Designed for modern hardware (GPUs).
- Aims for high performance and accessibility (open source). github.com/williamjameshandley/talks/blob/cosmoverse\_2025/blackjax\_ns.py

## Large Language Models (LLMs)

- LLMs (like the one assisting with this talk) show increasing potential:
  - ▶ Rapid code generation for analysis pipelines (e.g., custom Boltzmann solvers).
  - Synthesis of complex information, literature review.
  - Structuring analyses and drafting papers/talks.
- ▶ A capable prompt engineer could significantly accelerate research workflows.





github.com/handley-lab/group

- Cosmological tensions (e.g.,  $H_0$ ,  $S_8$ ,  $\Omega_K$ ) demand systematic and robust analysis frameworks.
- unimpeded, powered by DiRAC allocations (DP192 & DP264), provides a growing, publicly accessible library of:
  - MCMC chains & Nested Sampling runs.
  - ML-emulators for fast likelihood/posterior evaluations.
- Covers  $\sim 10$  models and  $\sim 60$  dataset combinations (and growing).
- ▶ Facilitates parameter estimation, model comparison, and critical tension quantification.
- Key for addressing fiducial biases and rigorously exploring beyond-ΛCDM physics.
- Future:
  - Expansion with DiRAC 17: Next-gen datasets (DESI, Euclid, etc.).
  - Integration of advanced SBI and AI/LLM techniques.
  - Continued community-focused development.
- We are seeking α-testers and collaborators!

## Bonus Slide: The Role of LLMs in This Talk

This presentation was drafted with the assistance of a Large Language Model. LLMs can be powerful tools for:

- Synthesizing complex information from multiple lengthy documents.
- Structuring scientific narratives and identifying key talking points.
- Generating initial drafts for talks, papers, and code.
- Assisting with literature reviews.
- Brainstorming slide content and visual aids.

The future of scientific research, communication, and analysis will likely involve closer and more sophisticated human-AI collaboration.

 
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 will@maxwell 21 May 07:45

 Objective:
 Generate a LaTeX Beamer presentation (slides only not a full document) for a 30minute [mild] (target < 15 slides) to be delivered on 2025-05-20.</td>

#### Core Topic & Narrative:

The initial island focus on binarcasing the Burkgeded apftare and its application to results fr on DIMA cillocations (specifically 1986.23' completed, 1986.21' compring). A key there is the impertance of robust cossolgital analysis, quantifying tensions (like the curvature tension) and addressing binase (tes, f. Col Mais from fideling languages), and the impledies the second binary of the second se

Target Audience: Cosmoverse workshop (see below)

#### Stylistic Guidelines:

- Minimal use of overlays (\pause, \onslide).
- Use of columns for layout.
  - Bullet points typically in one column.
- Figures often placed in the other column.
   Cite arXiv papers using \arXiv(<arXiv\_id>).

#### Input Materials & How to Use Them:

1. Cosmoverse Workshop Details:

\* Purpose: context on the **audience** and other speakers. ```text

[INSERT cosmoverse.txt]

#### Talk abstract:

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#### DiRAC Cases for Support:

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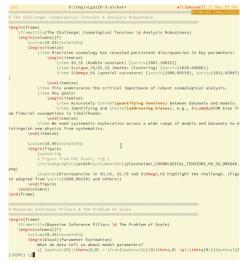
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