

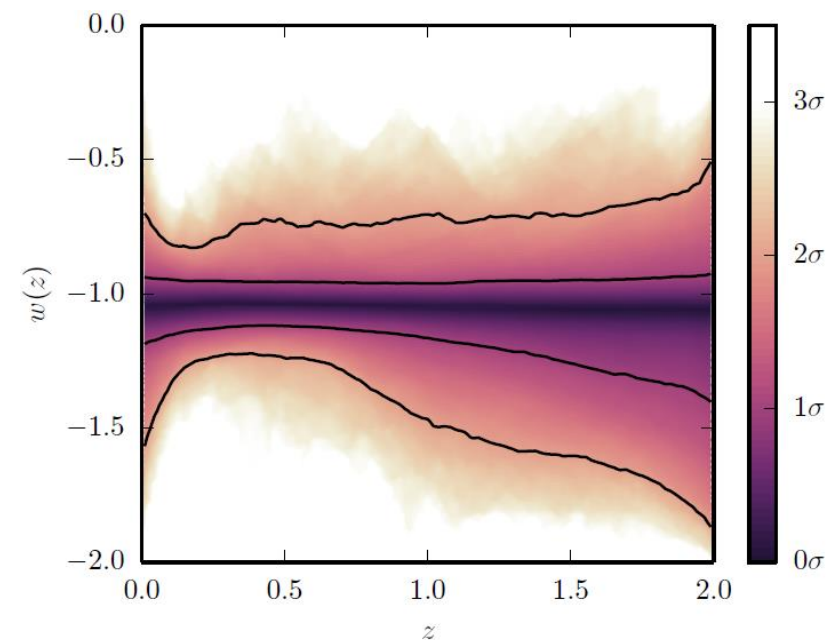
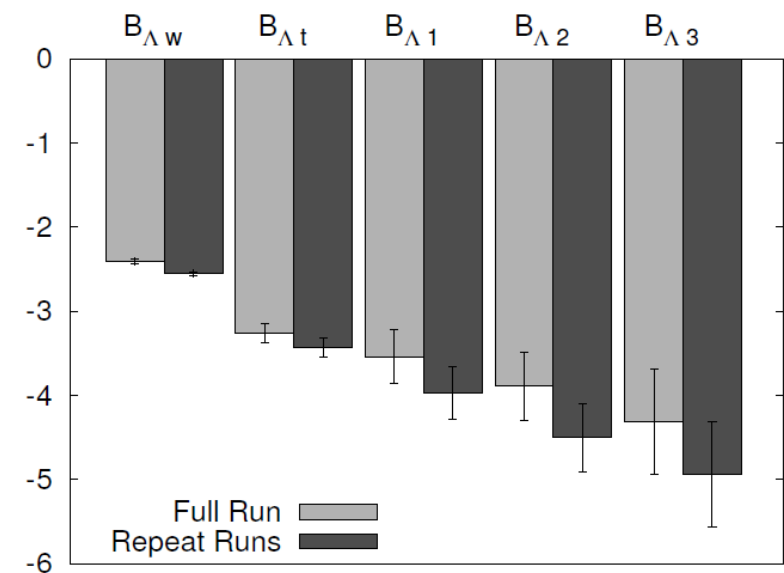
A Novel Technique for Bayesian Model Selection

Dark Energy

-- From WMAP to Planck --

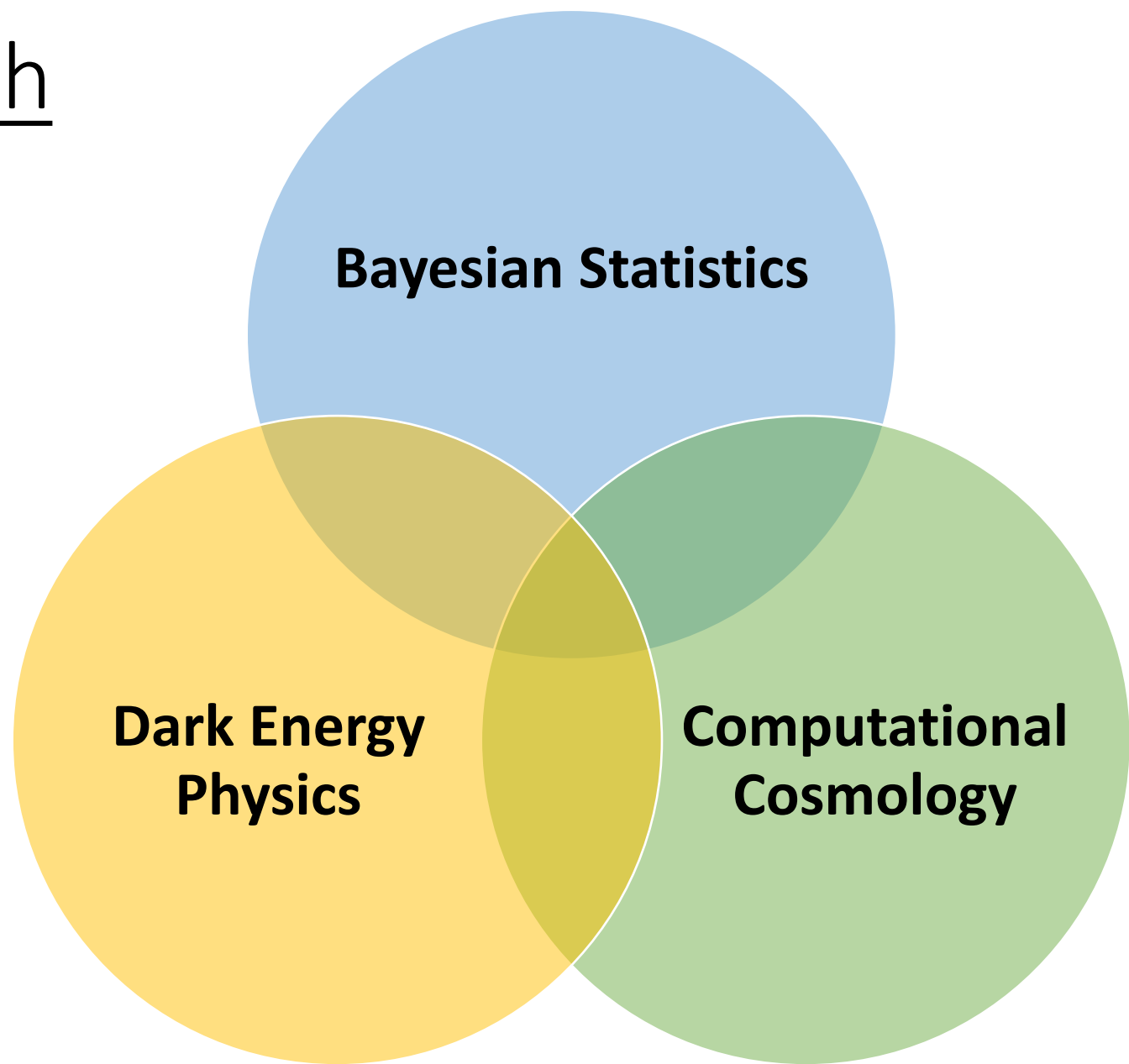
Sonke Hee

Cavendish Astrophysics Group, Cambridge
Kavli Institute for Cosmology Cambridge
Churchill College, Cambridge



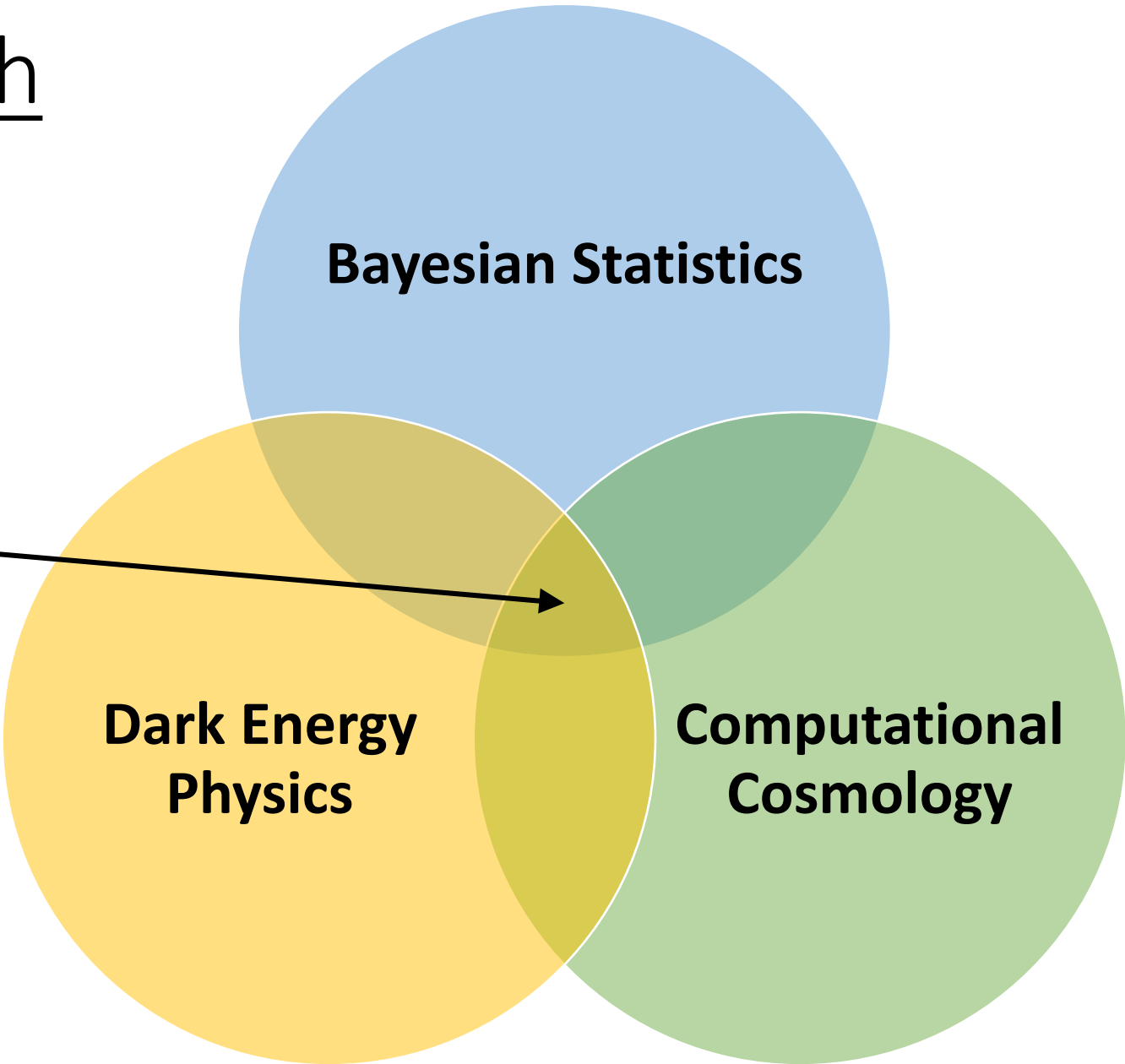


Where is this research taking place?

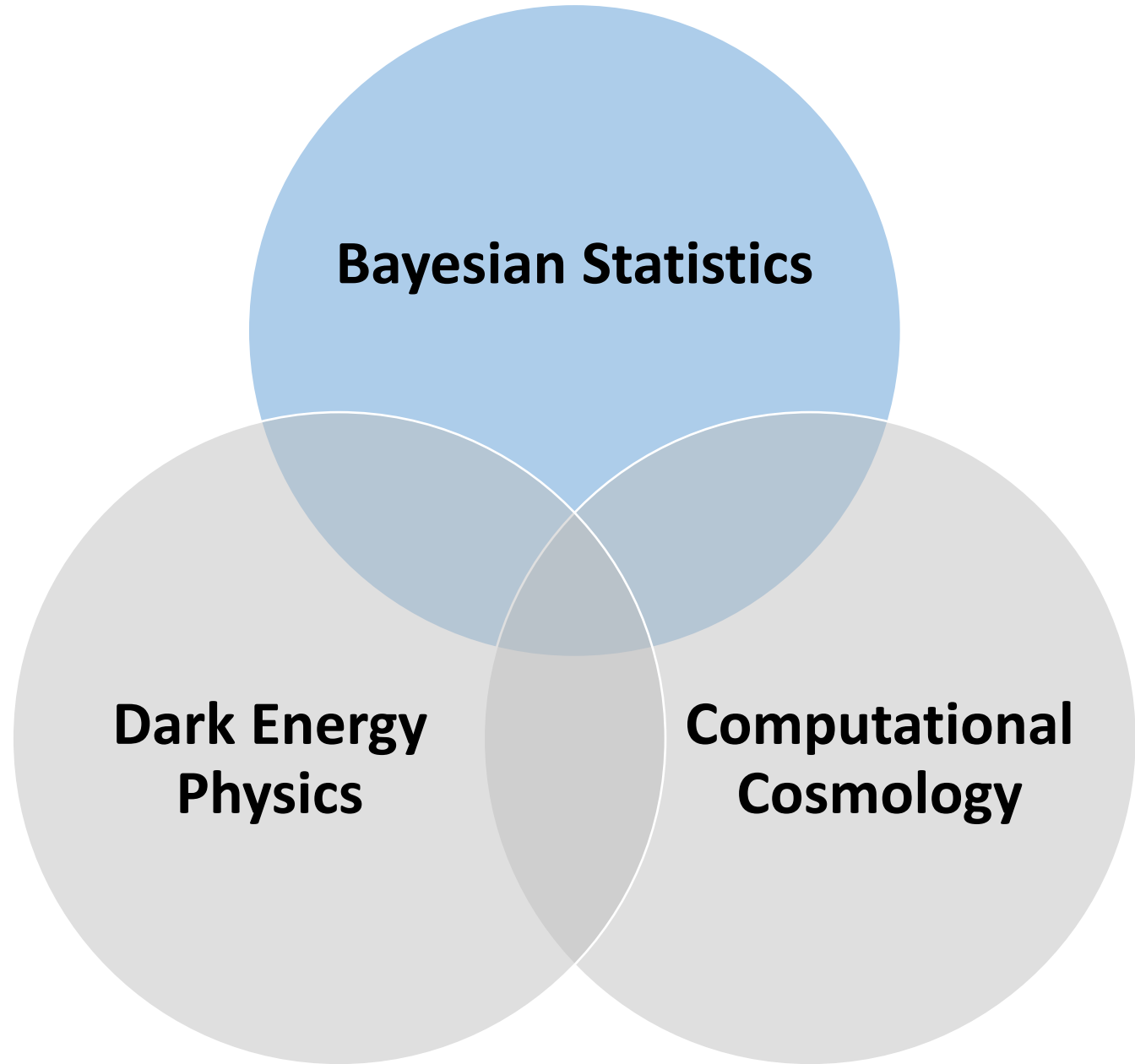


Where is this research taking place?

Here!



Bayesian Statistics

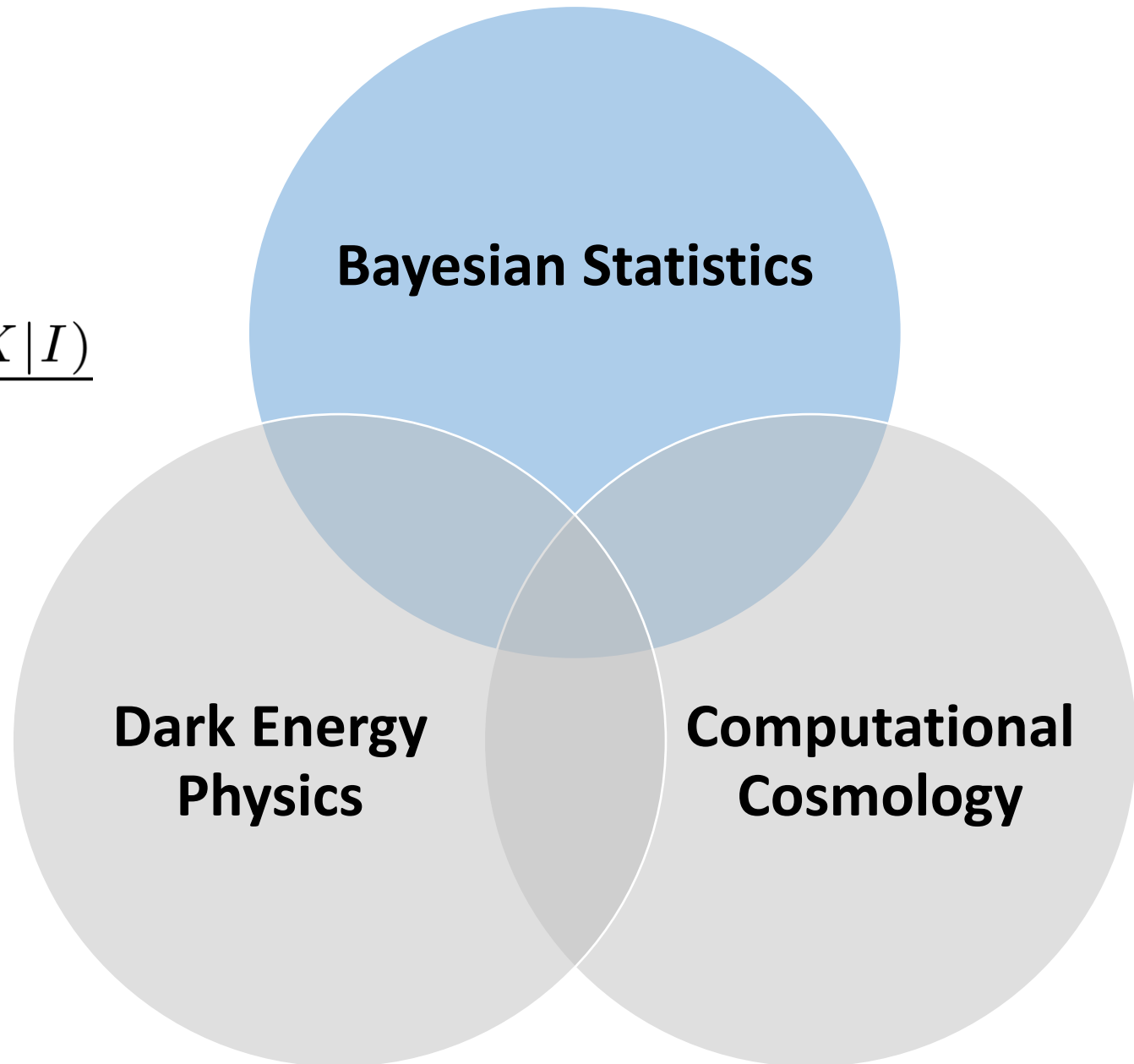




Bayesian Statistics

Bayes Theorem:

$$Pr(X|Y, I) = \frac{Pr(Y|X, I) \times Pr(X|I)}{Pr(Y|I)}$$

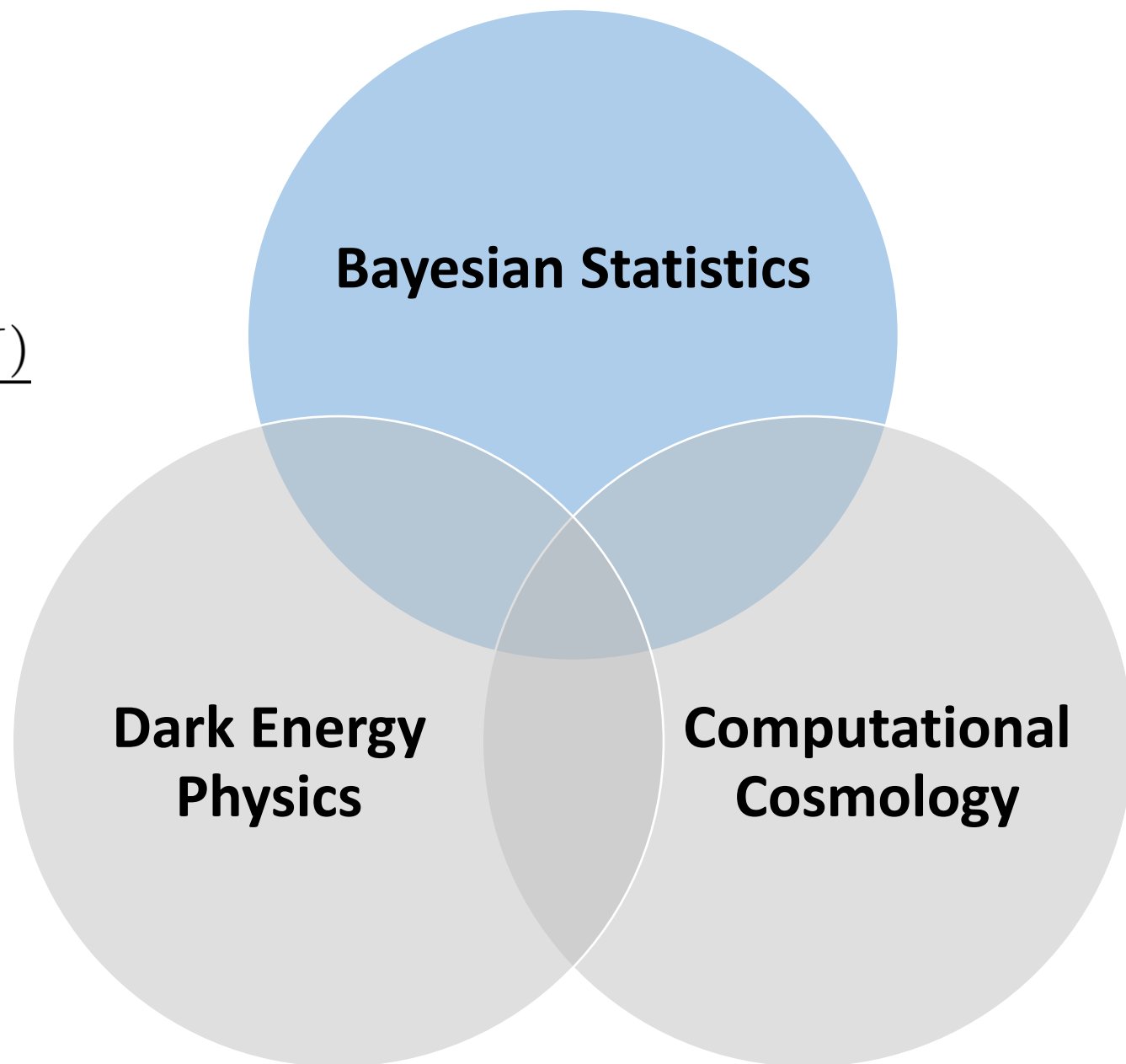




Bayesian Statistics

(1) Parameter Estimation:

$$Pr(\theta|\mathcal{D}, I) = \frac{Pr(\mathcal{D}|\theta, I) \times Pr(\theta|I)}{Pr(\mathcal{D}|I)}$$





Bayesian Statistics

(1) Parameter Estimation:

$$Pr(\theta|\mathcal{D}, I) = \frac{Pr(\mathcal{D}|\theta, I) \times Pr(\theta|I)}{Pr(\mathcal{D}|I)}$$

(2) Model Selection

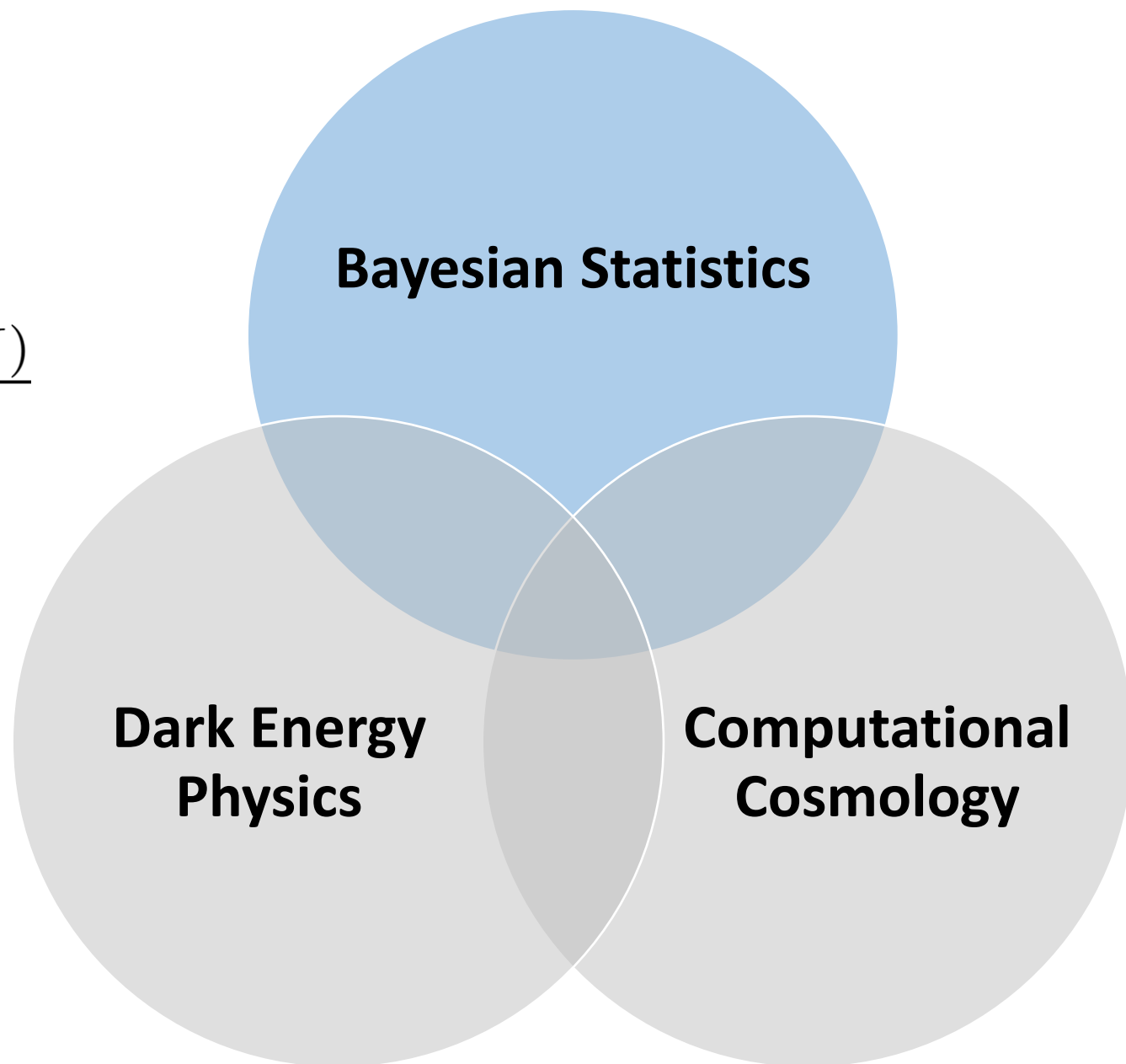
$$\mathcal{B}_{ij} = \ln\left(\frac{Pr(\mathcal{M}_j|\mathcal{D})}{Pr(\mathcal{M}_i|\mathcal{D})}\right) = \ln\left(\frac{\mathcal{Z}_i}{\mathcal{Z}_j}\right)$$

\mathcal{B}_{ij} = Bayes Factor

\mathcal{M}_i = Model i

\mathcal{D} = Data

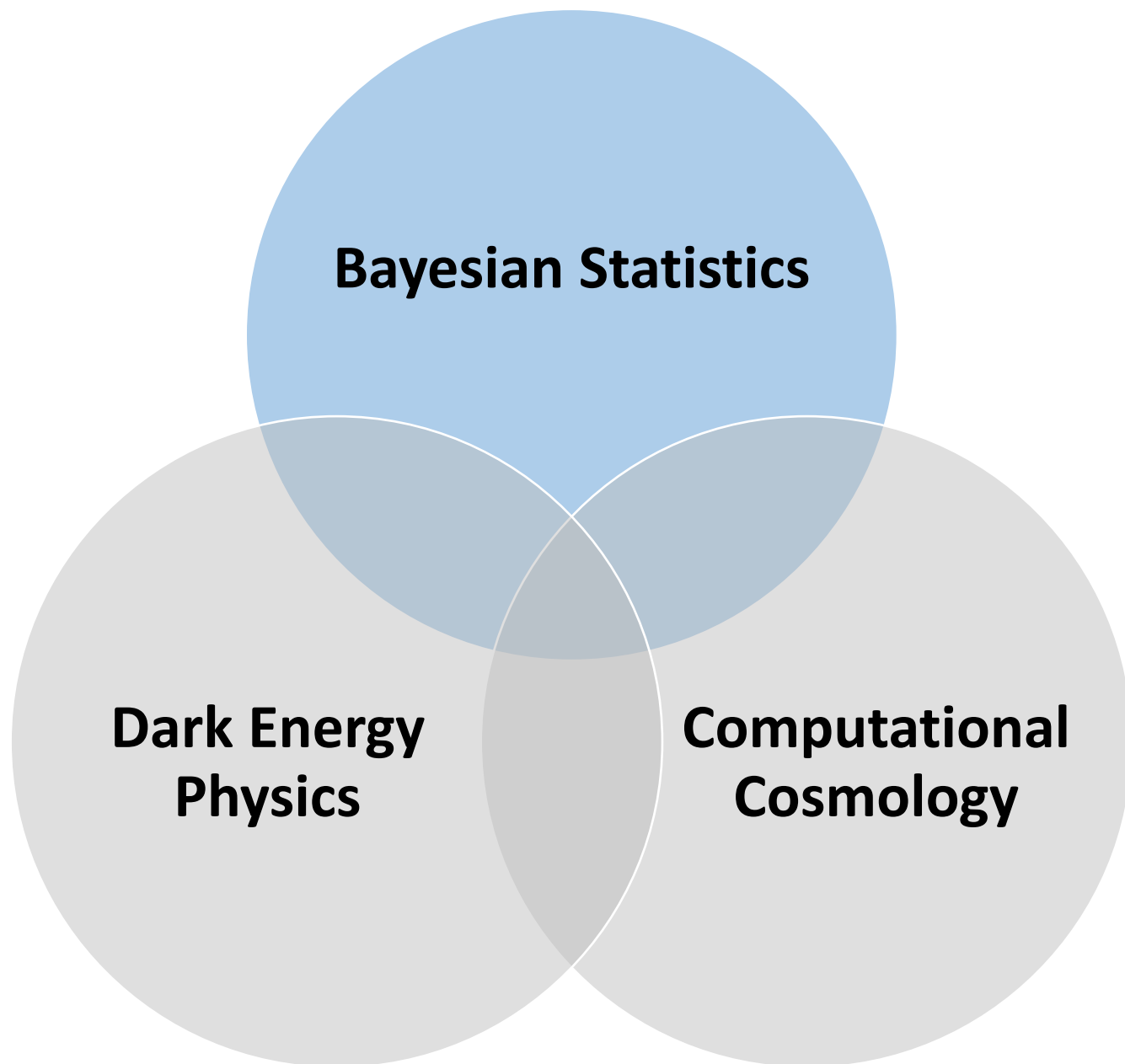
$\mathcal{Z} = Pr(\mathcal{D}|I)$ = ‘Evidence’





Bayesian Statistics

$$\mathcal{Z} = \int_{all \theta} \mathcal{L}(\theta) \pi(\theta) d\theta$$

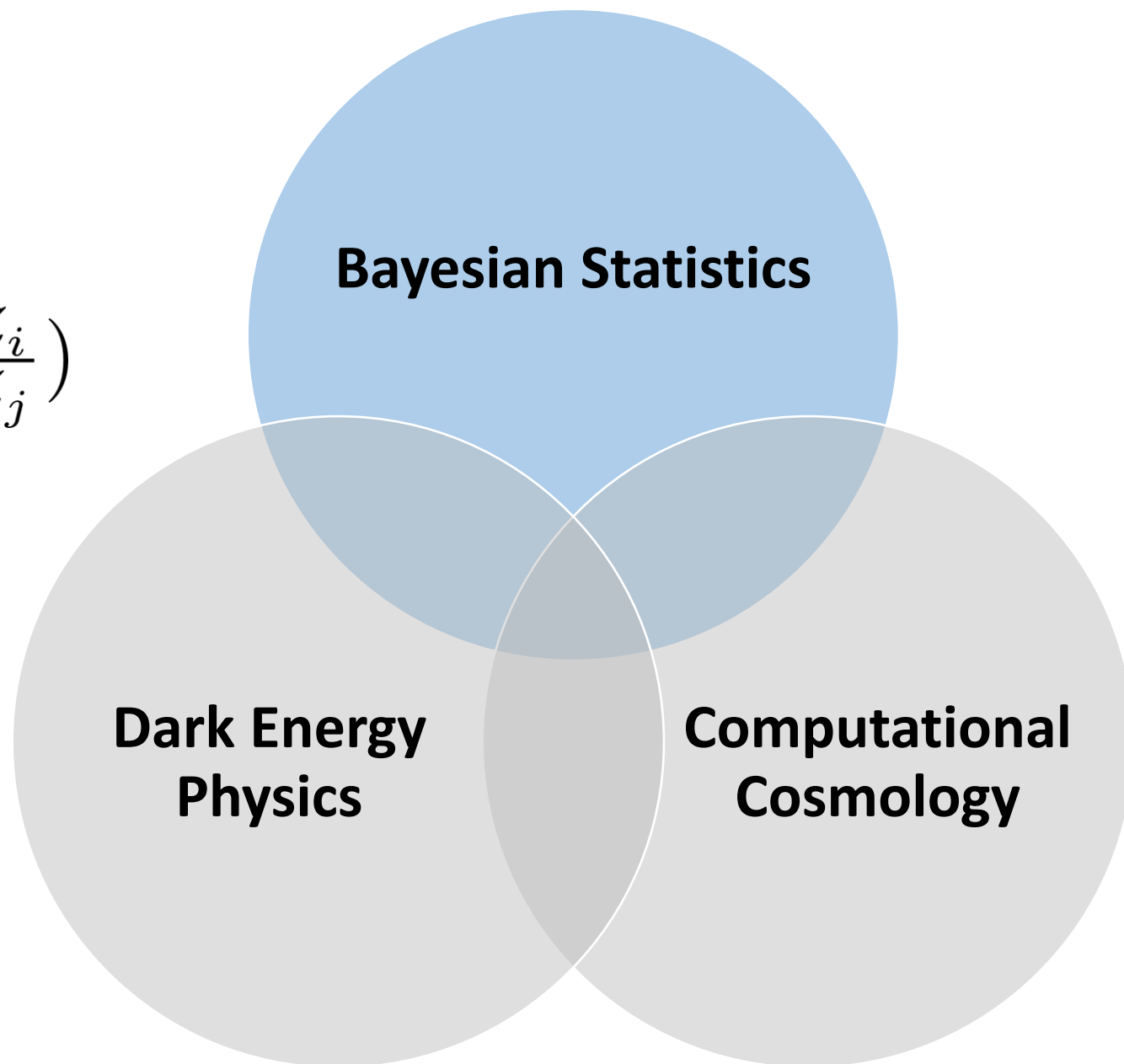




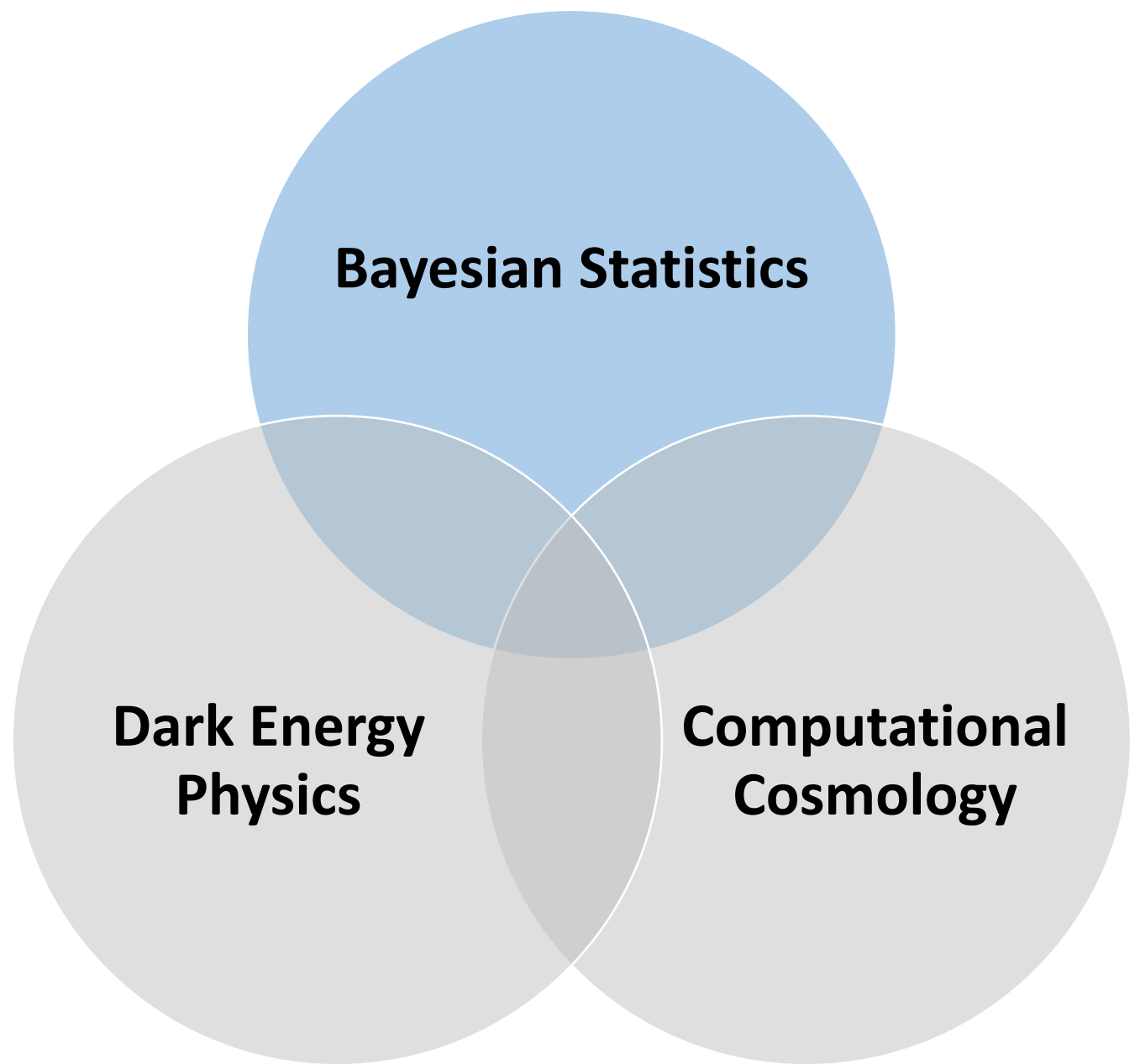
Bayesian Statistics

$$\mathcal{B}_{ij} = \ln\left(\frac{\text{Pr}(\mathcal{M}_j|\mathcal{D})}{\text{Pr}(\mathcal{M}_i|\mathcal{D})}\right) = \ln\left(\frac{\mathcal{Z}_i}{\mathcal{Z}_j}\right)$$

Bayes factor	Interpretation
$0.0 \leq \mathcal{B}_{ij} < 1.0$	Not significant
$1.0 \leq \mathcal{B}_{ij} < 2.5$	Slight
$2.5 \leq \mathcal{B}_{ij} < 5.0$	Strong
$5.0 \leq \mathcal{B}_{ij}$	Decisive

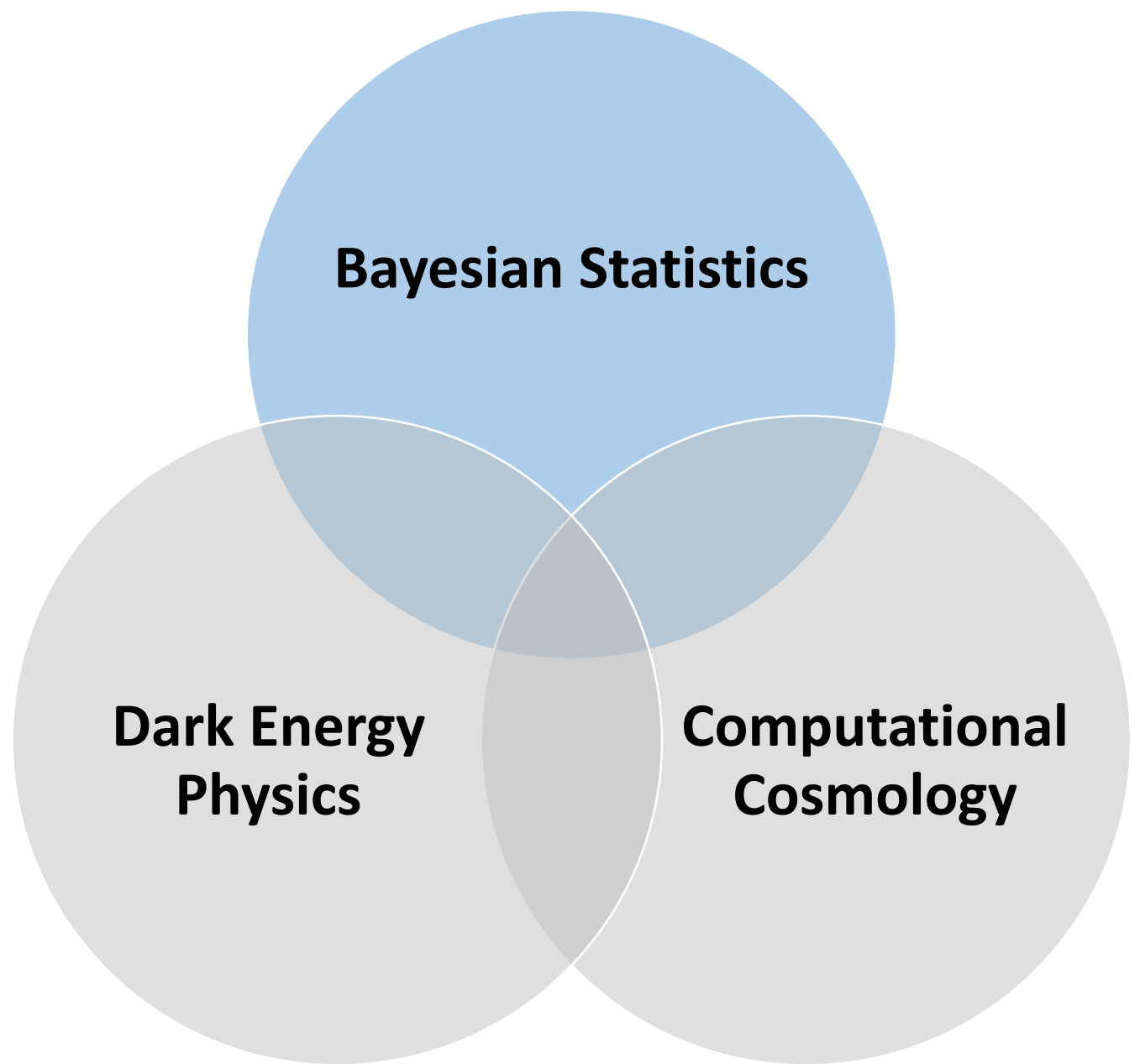
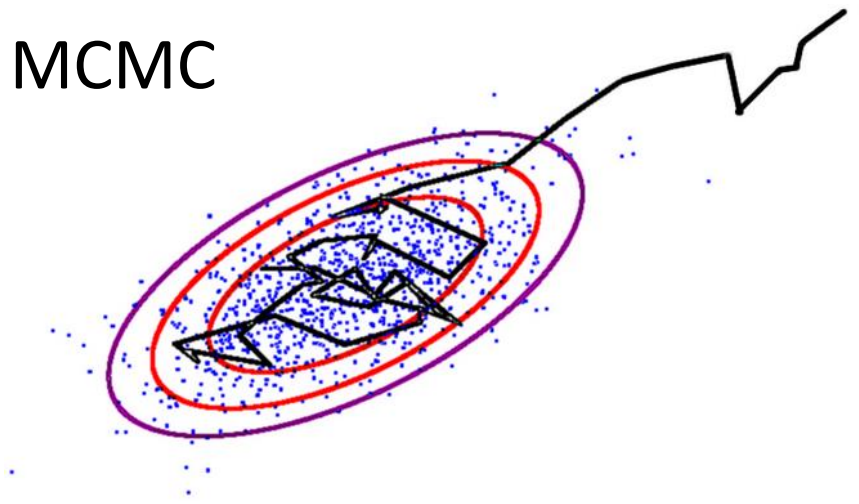


Bayesian Samplers



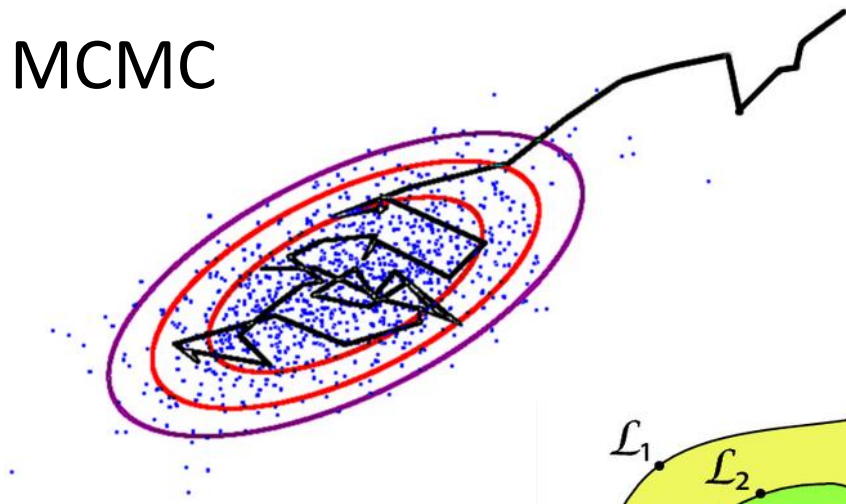
Bayesian Samplers

MCMC

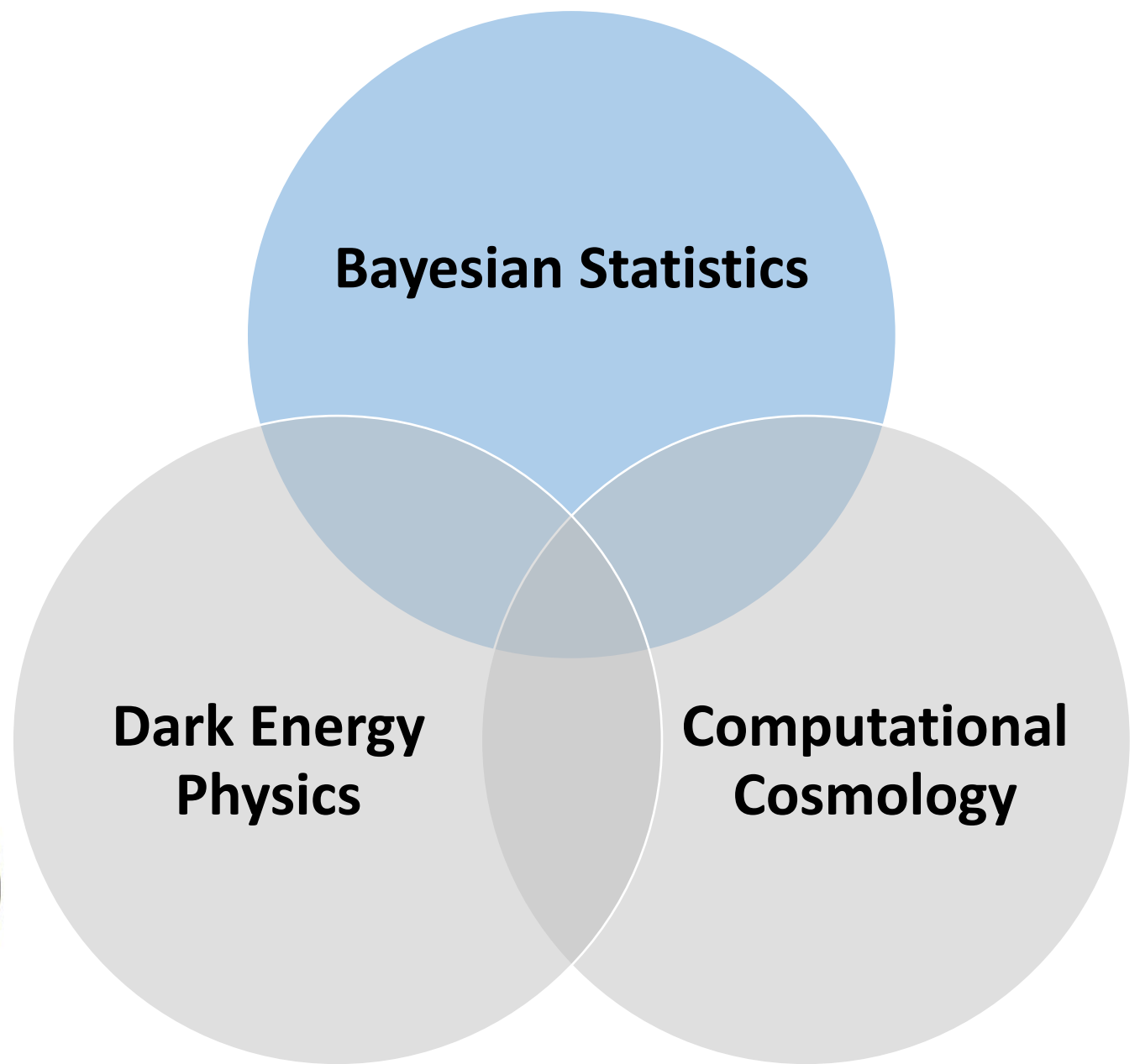
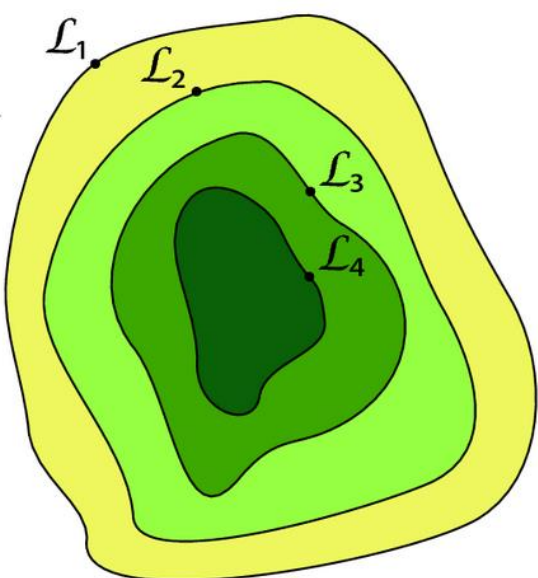


Bayesian Samplers

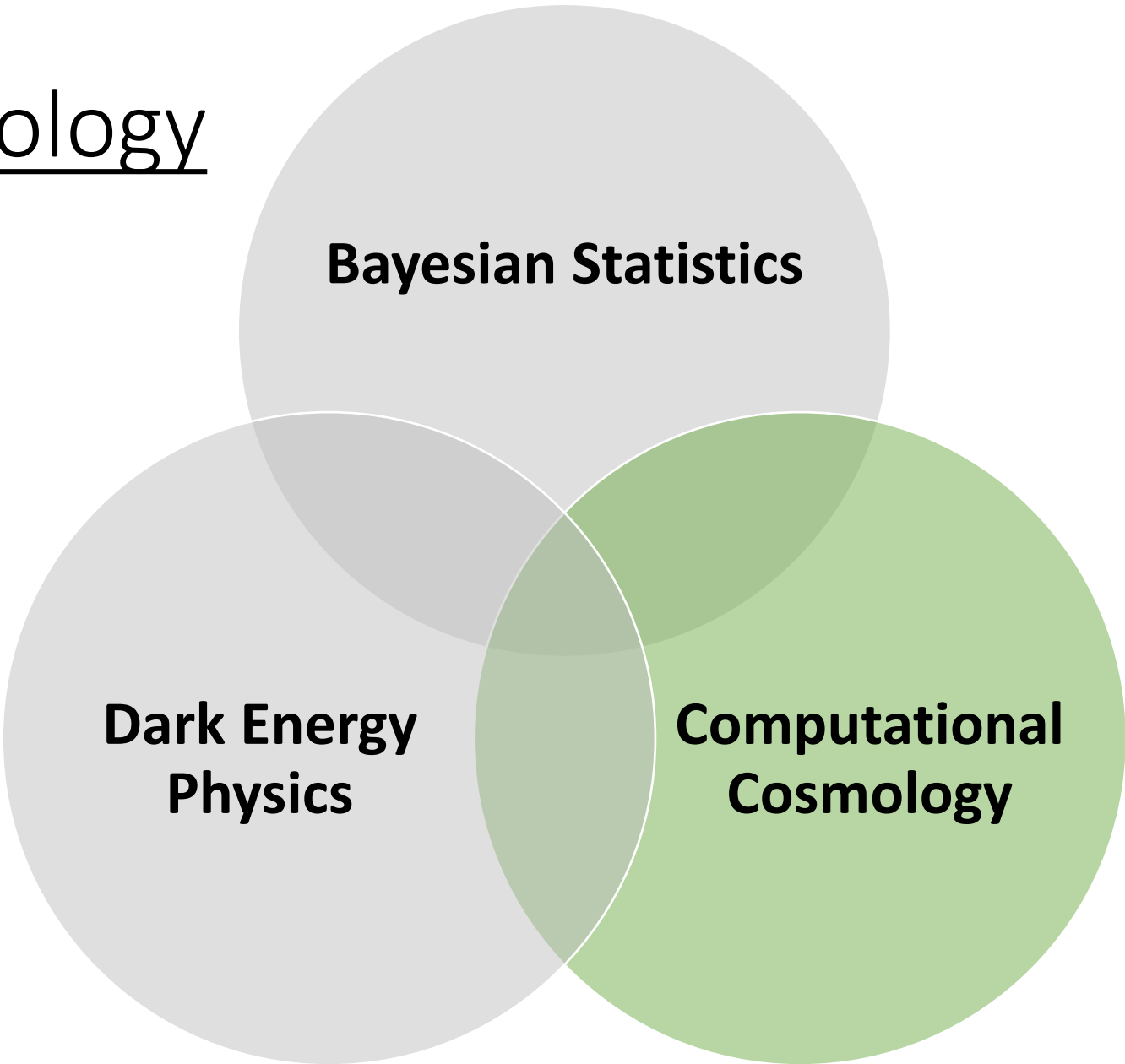
MCMC



Nested Sampling

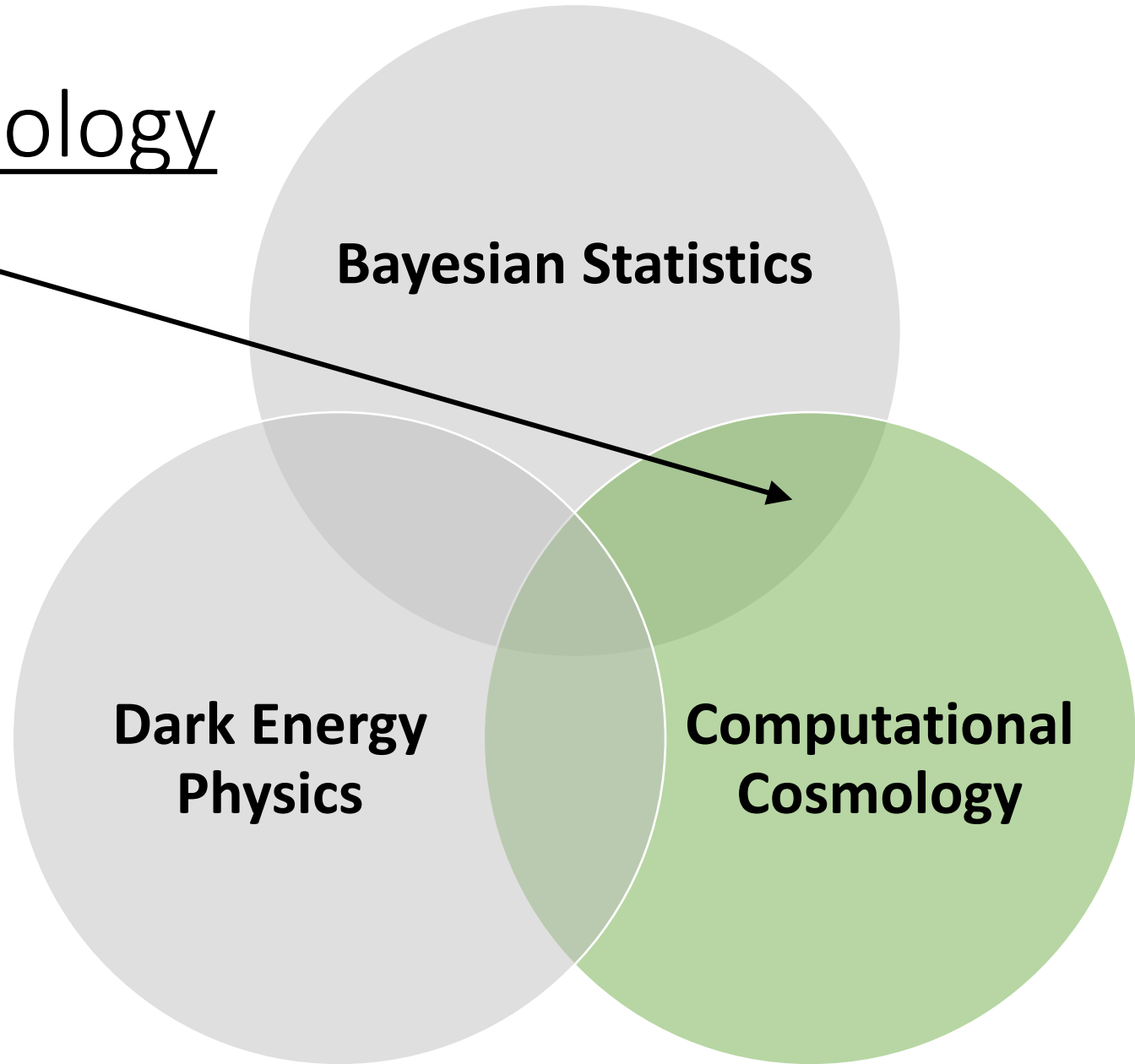


Computational Cosmology



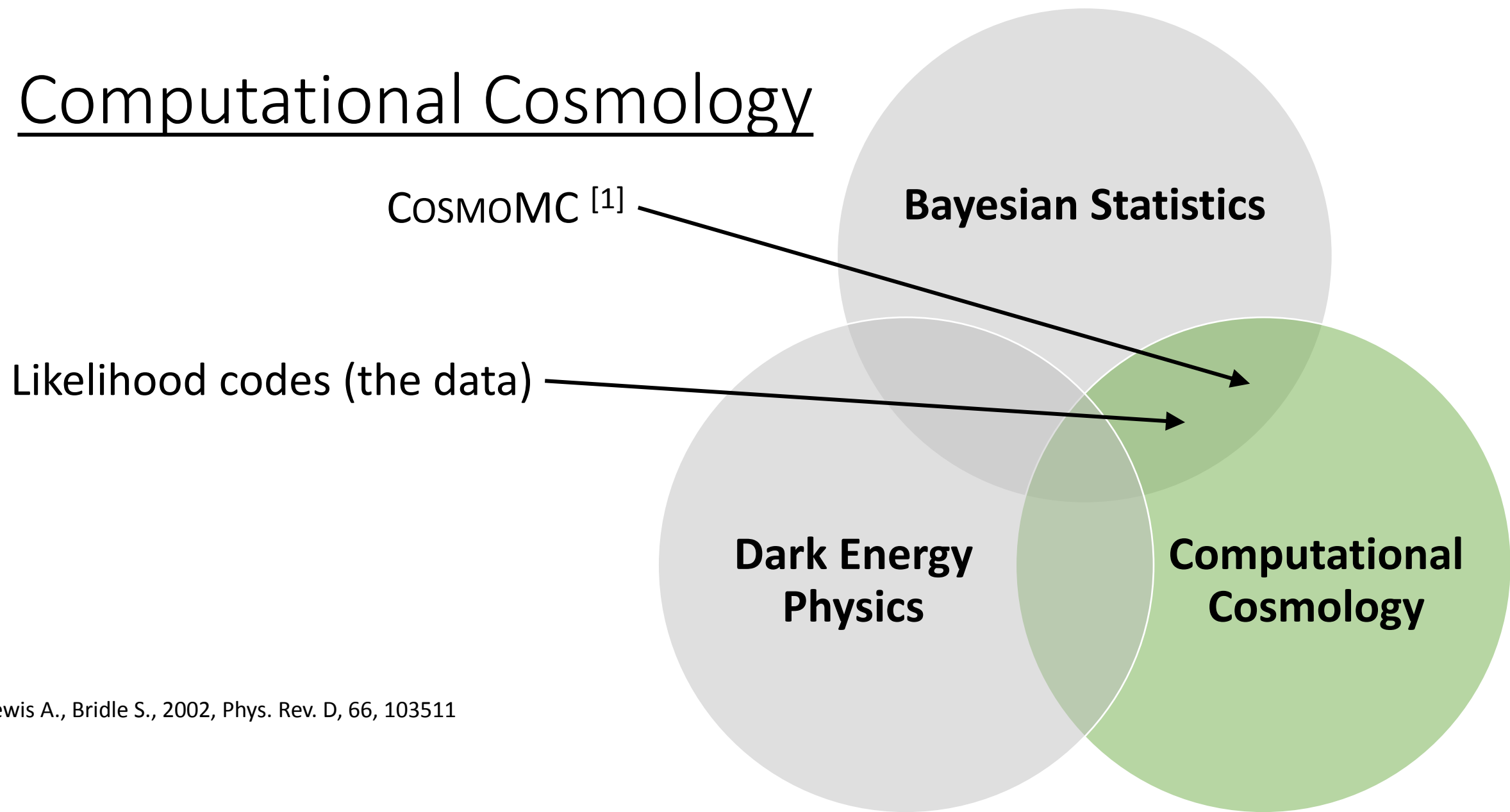
Computational Cosmology

COSMOMC [1]



[1] Lewis A., Bridle S., 2002, Phys. Rev. D, 66, 103511

Computational Cosmology



COSMOMC [1]

Likelihood codes (the data)

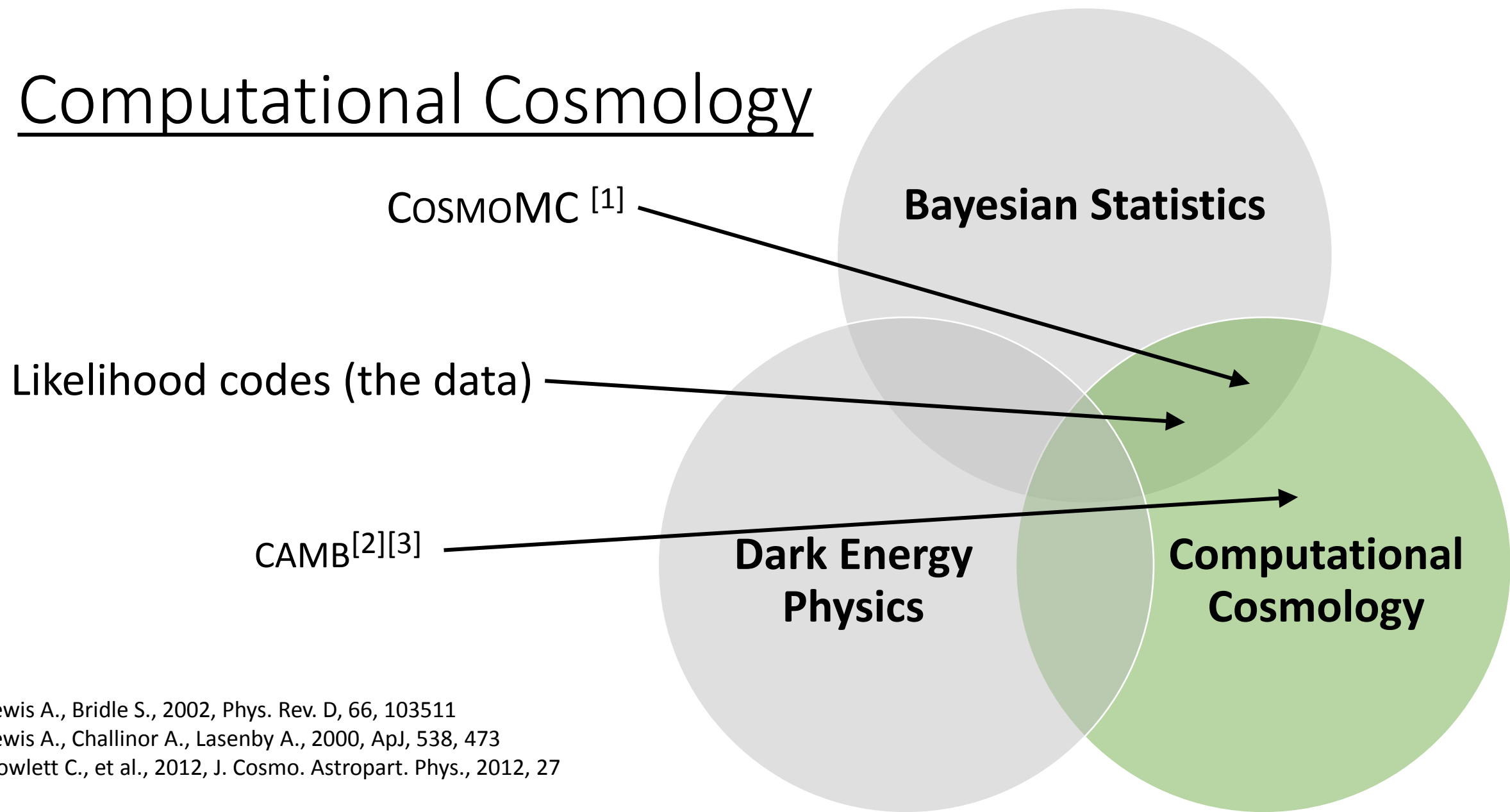
Bayesian Statistics

Dark Energy Physics

Computational Cosmology

[1] Lewis A., Bridle S., 2002, Phys. Rev. D, 66, 103511

Computational Cosmology



COSMOMC [1]

Bayesian Statistics

Likelihood codes (the data)

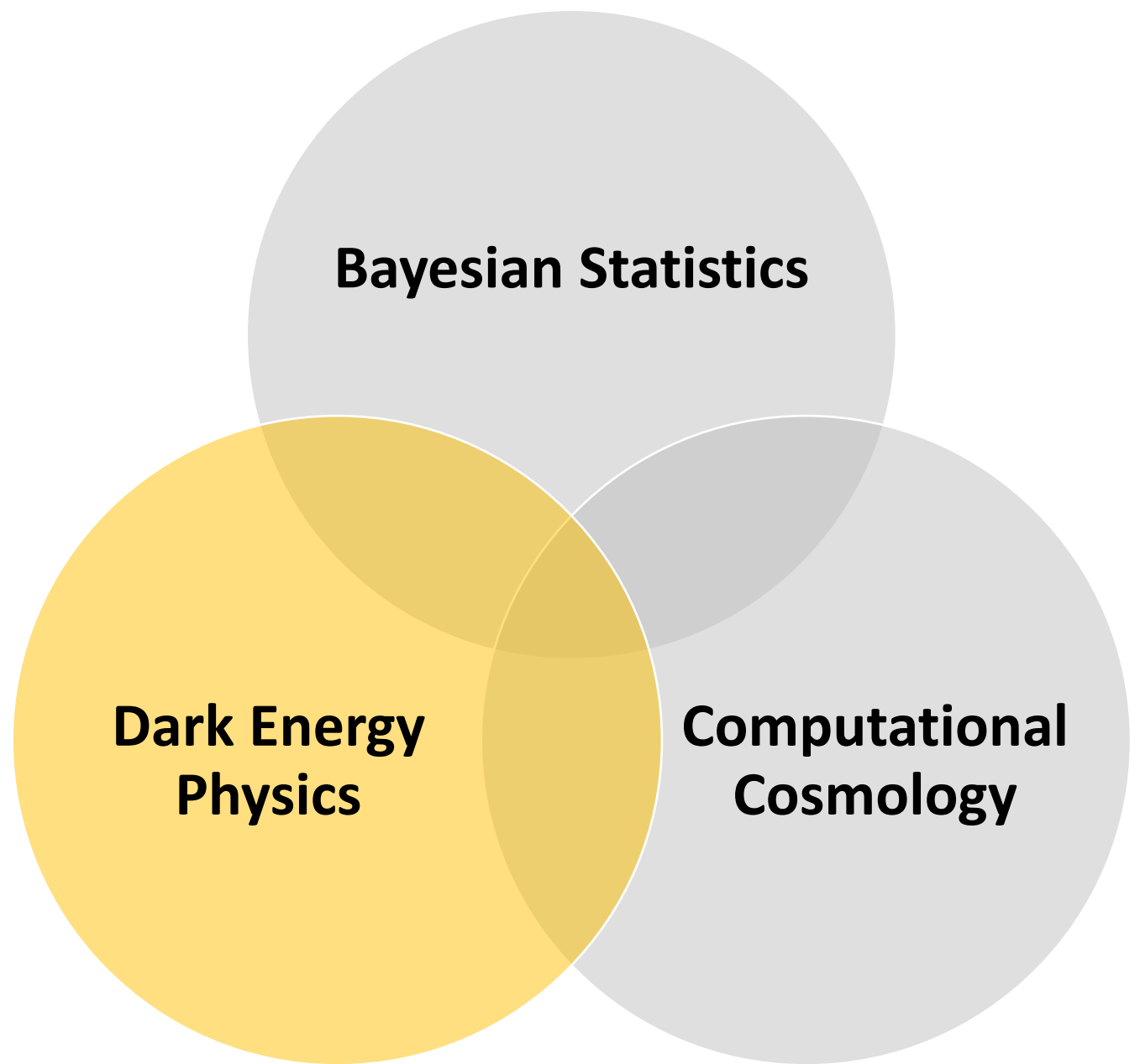
CAMB[2][3]

Dark Energy Physics

Computational Cosmology

[1] Lewis A., Bridle S., 2002, Phys. Rev. D, 66, 103511
[2] Lewis A., Challinor A., Lasenby A., 2000, ApJ, 538, 473
[3] Howlett C., et al., 2012, J. Cosmo. Astropart. Phys., 2012, 27

Dark Energy Physics



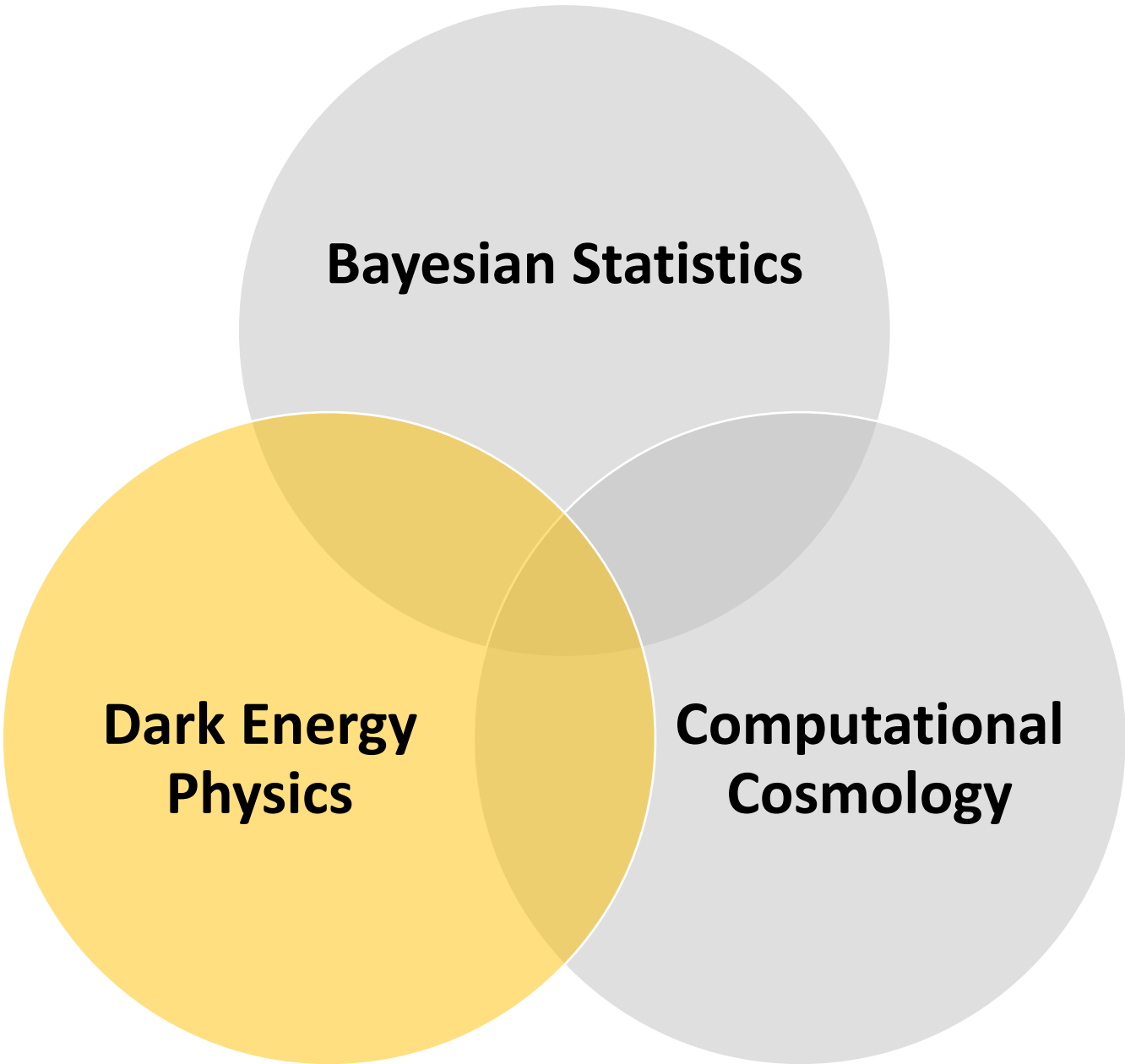


Dark Energy Physics

LCDM?

Quintessence?

Phantom?





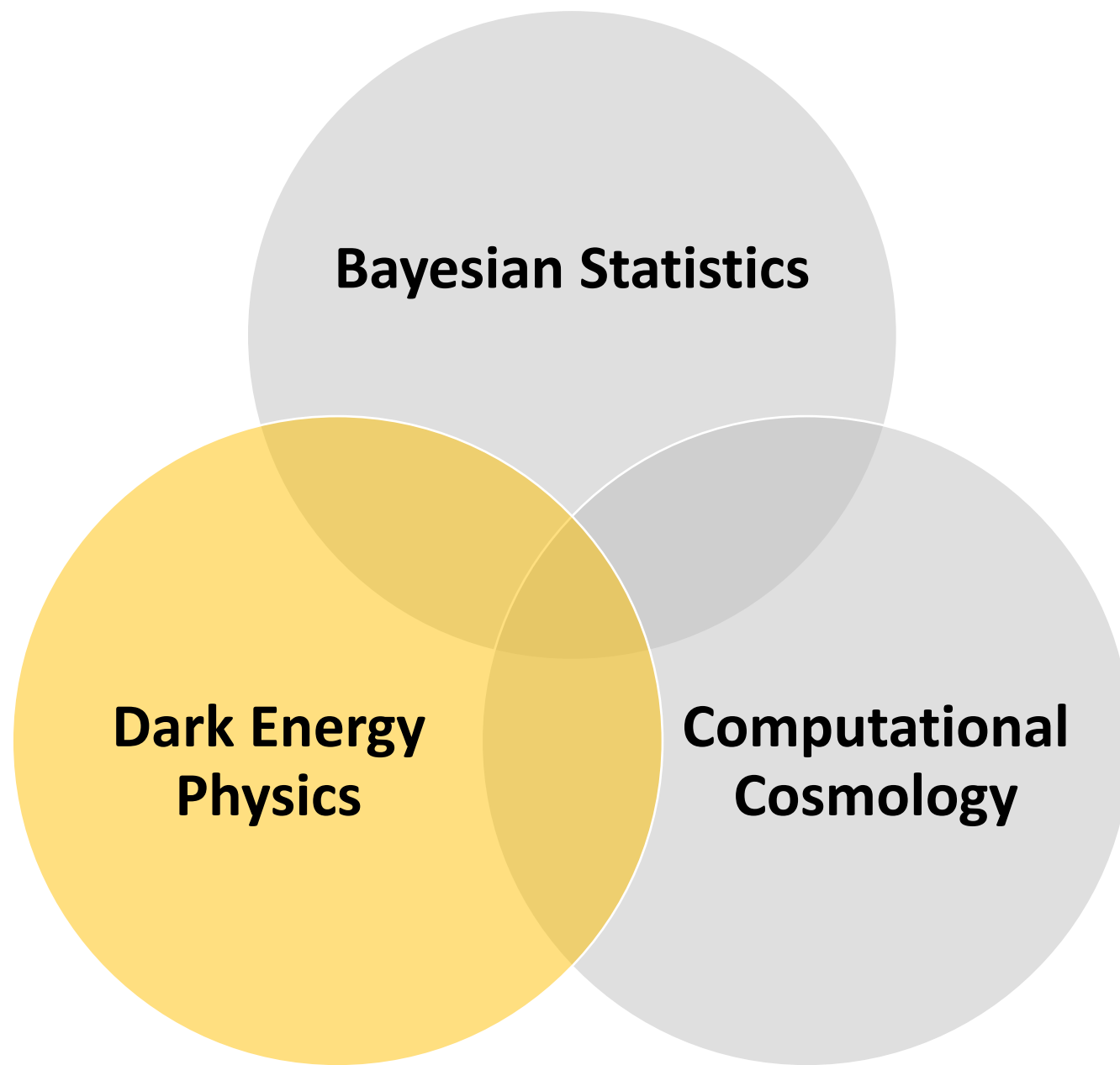
Dark Energy Physics

LCDM?

Quintessence?

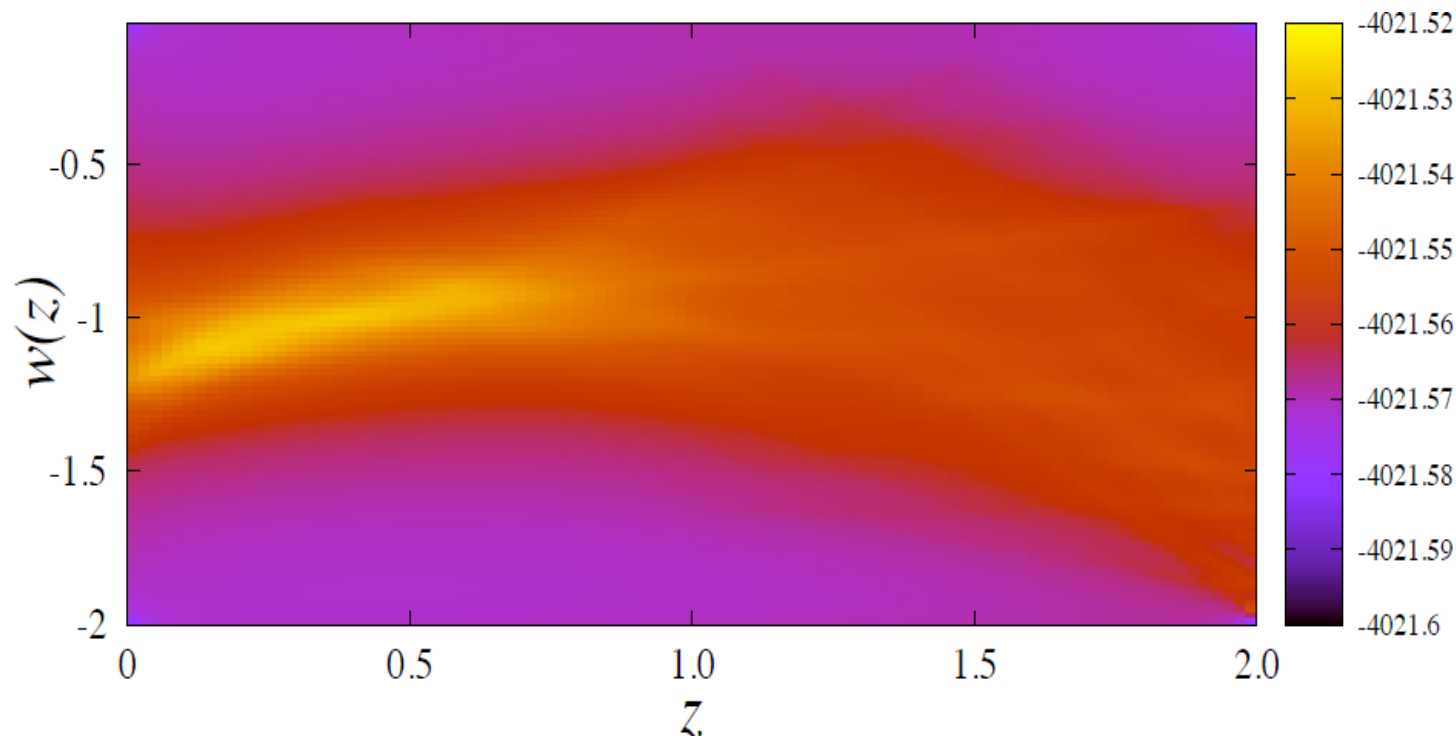
Phantom?

We investigate $w(z)$





The Problem we are Investigating



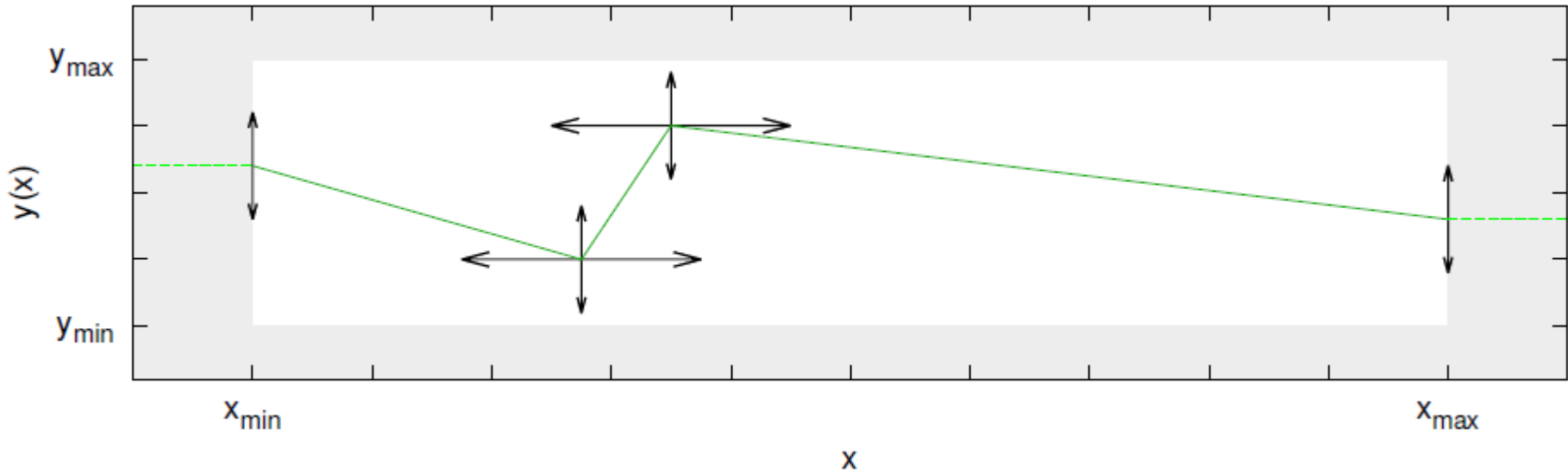
J.A. Vázquez, M. Bridges, M.P. Hobson and A.N. Lasenby (2012)

“Reconstruction of the dark energy equation of state”. JCAP, 09, 20.

(arxiv 1205.0847)

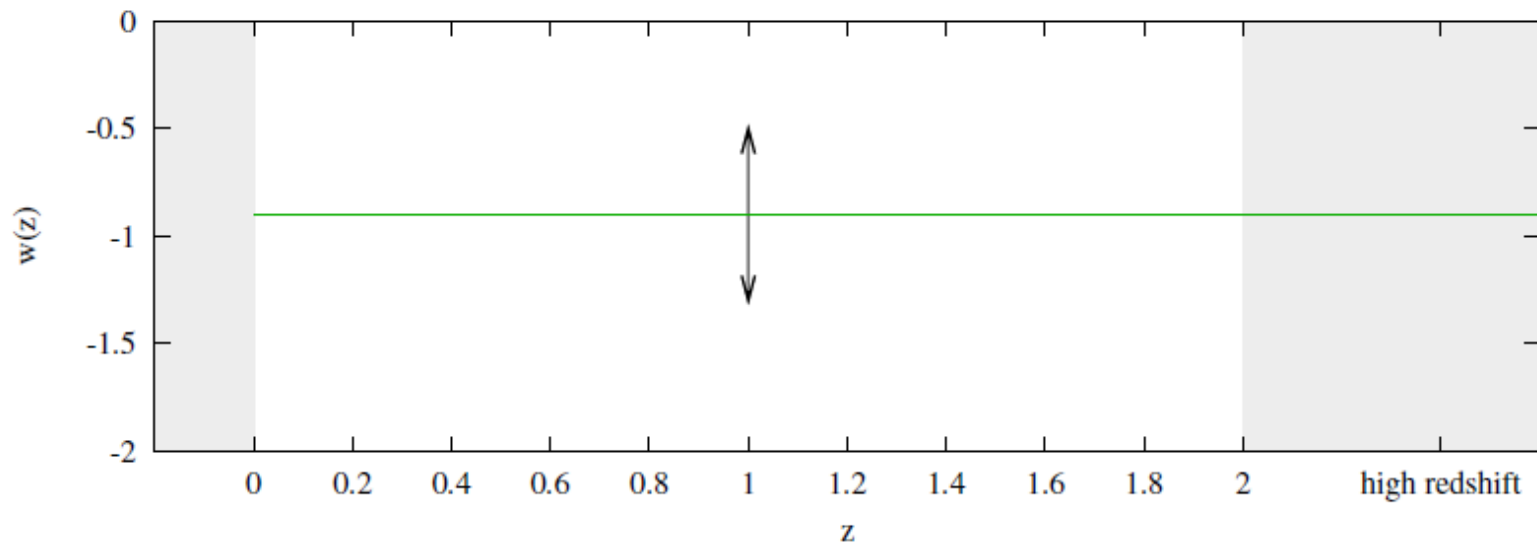


Previous Results - WMAP





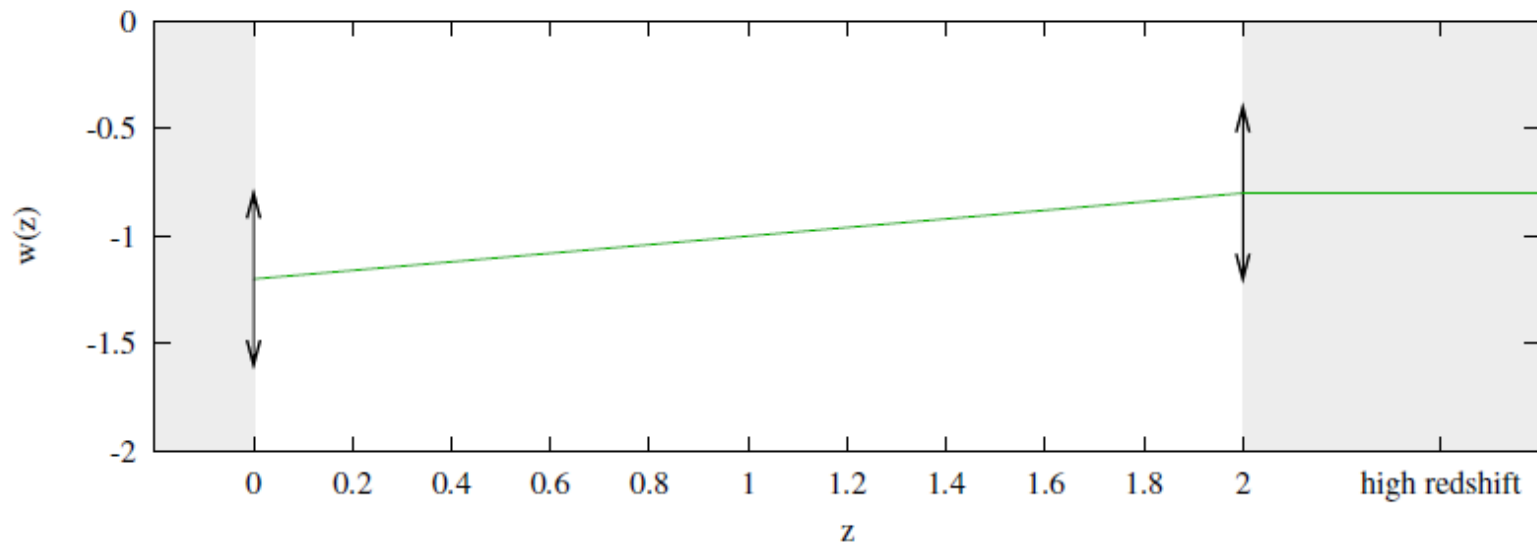
Previous Results - WMAP



Model (number of nodes)	Description
wCDM (1)	Constant $w(z)$
tCDM (2)	Tilted $w(z)$
1CDM (3)	Sudden change
2CDM (4)	Kinked model
3CDM (5)	Spikes/Kinks, etc

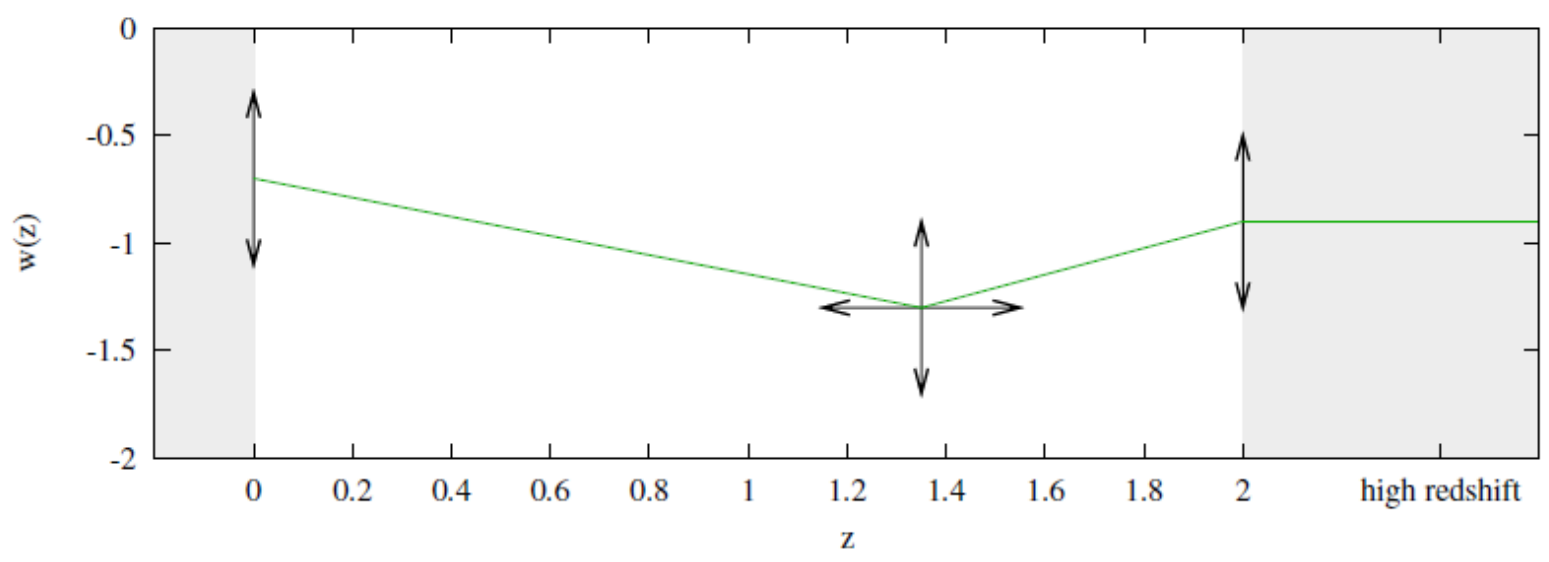


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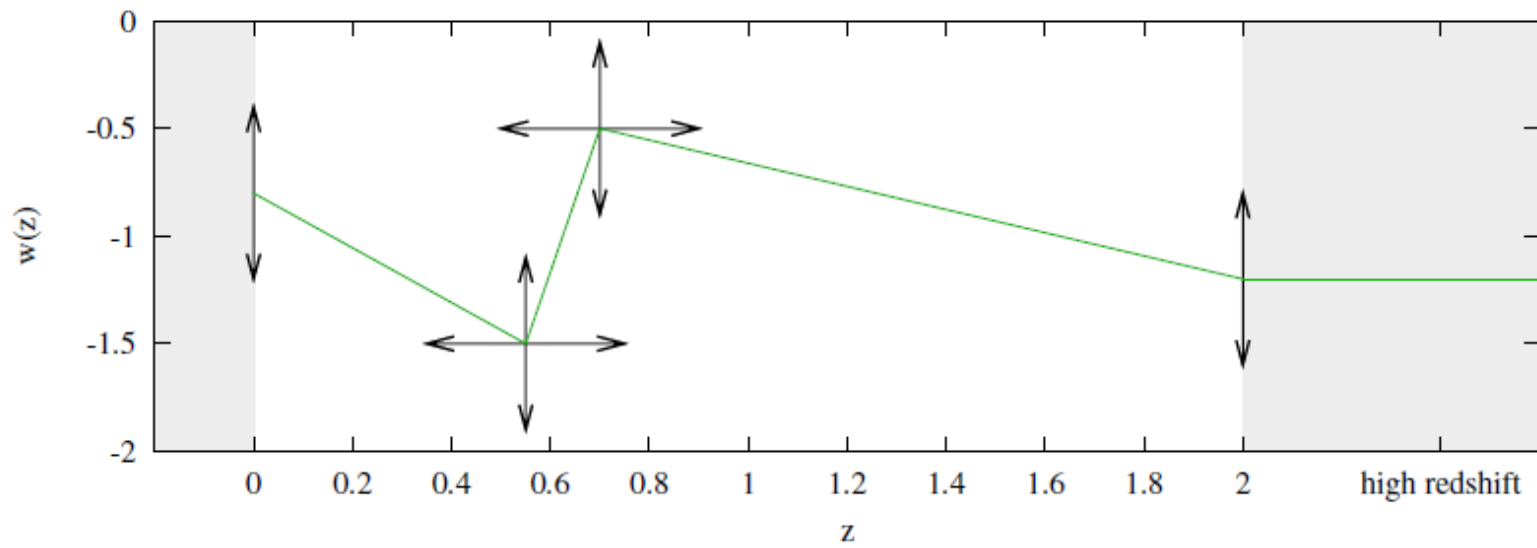
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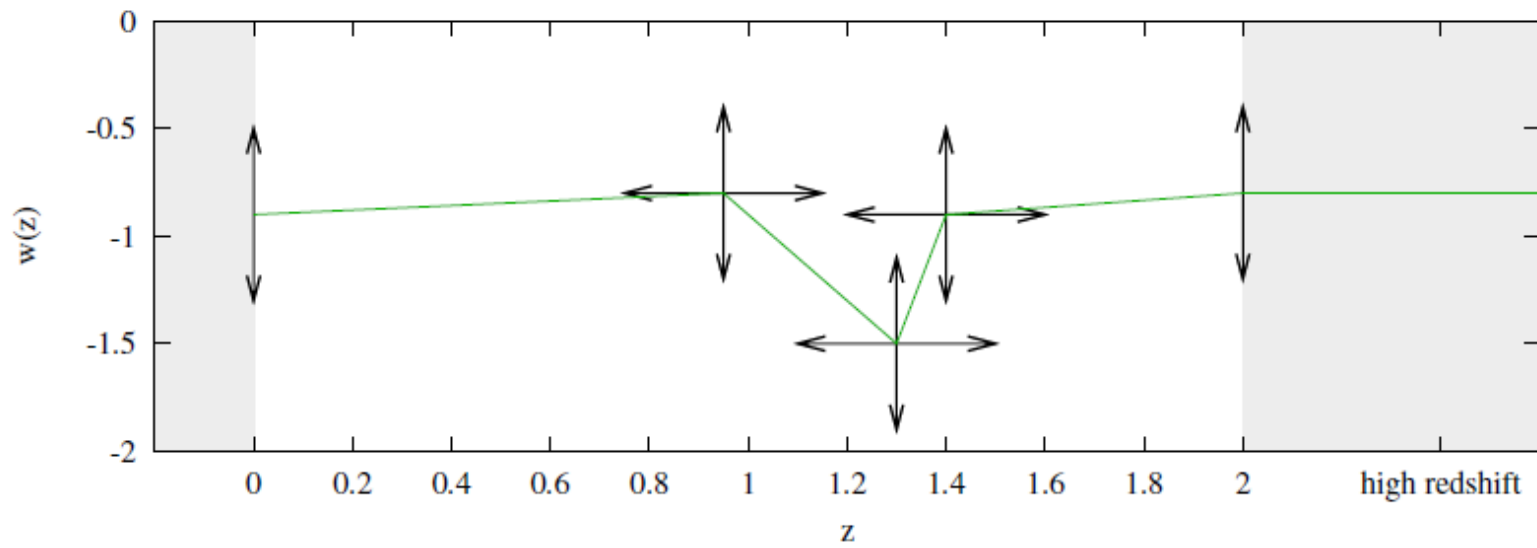
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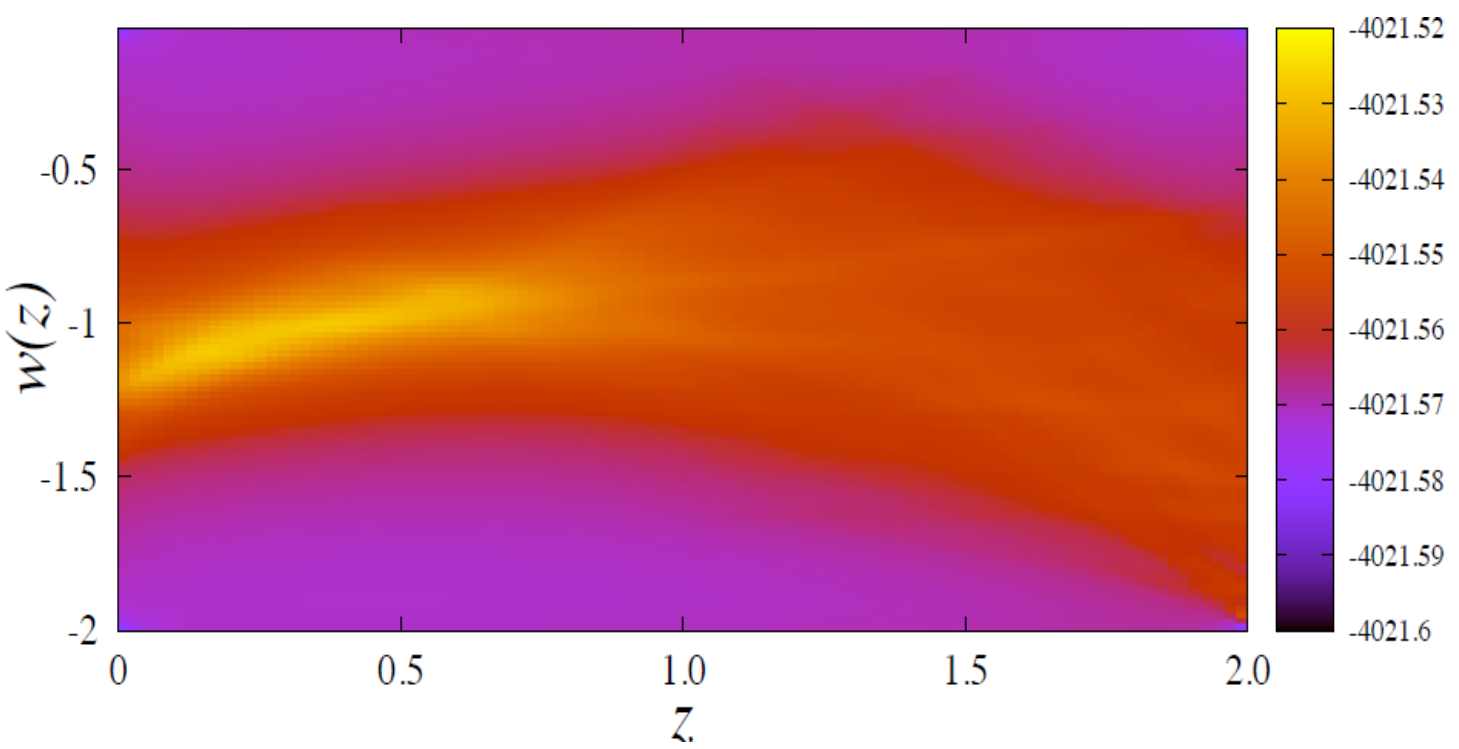


Previous Results - WMAP



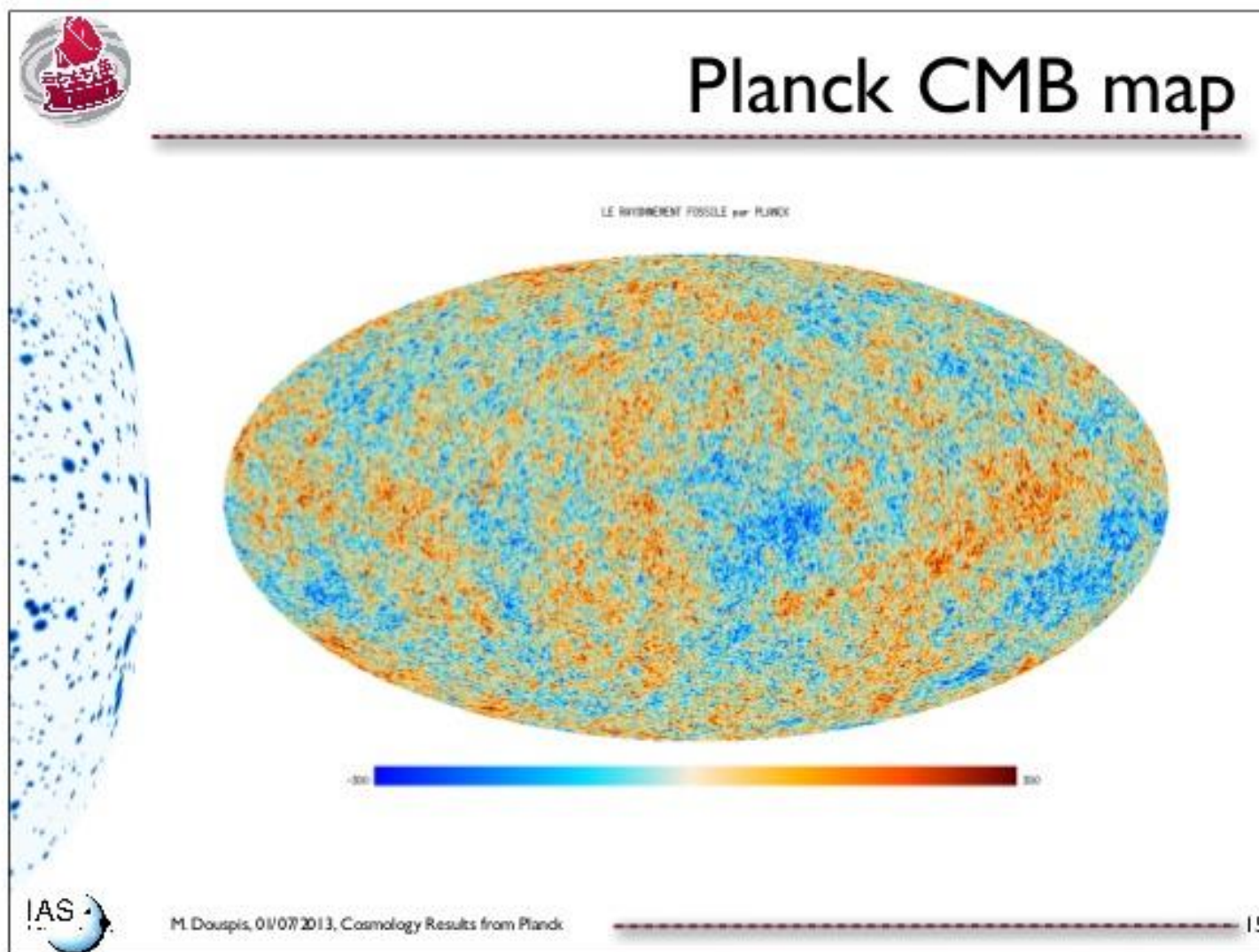
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1CDM (3)	Sudden change
2CDM (4)	Kinked model
3CDM (5)	Spikes/Kinks, etc

Previous Results - WMAP



Model (number of nodes)	Bayes Factor compared to LCDM
wCDM (1)	-2.19 +/- 0.35
tCDM (2)	-2.34 +/- 0.35
1CDM (3)	-1.27 +/- 0.35
2CDM (4)	-0.81 +/- 0.35
3CDM (5)	-0.95 +/- 0.35

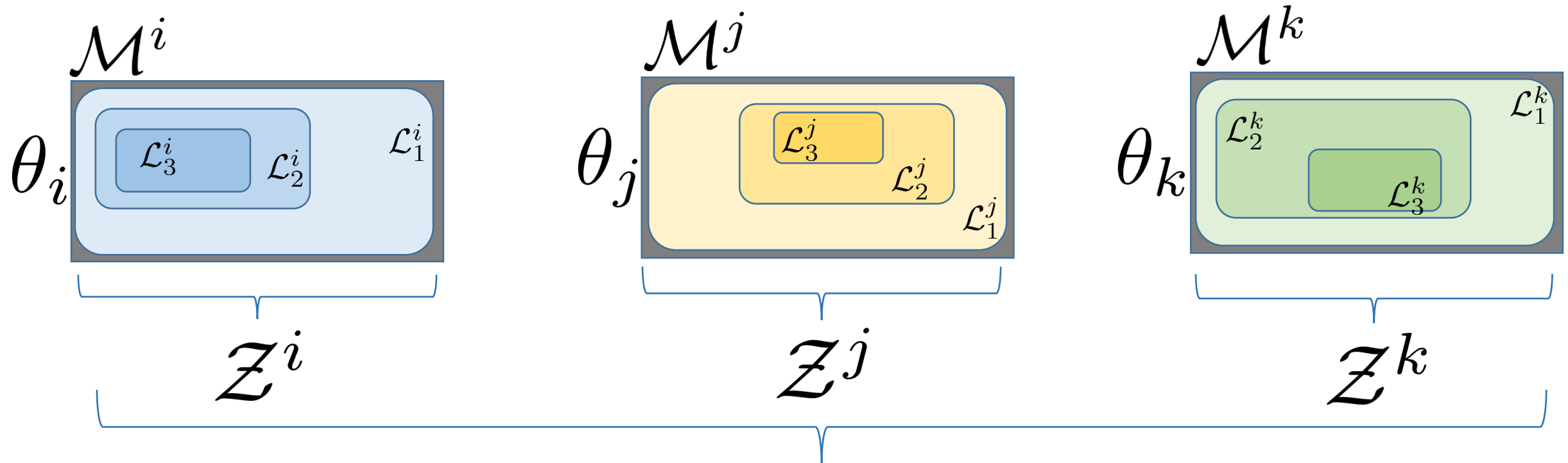
The Problem to Solve



Parameter	Prior Range	Prior Type
$\Omega_b h^2$	[0.019, 0.025]	uniform
$\Omega_c h^2$	[0.095, 0.145]	uniform
$100\theta_{MC}$	[1.03, 1.05]	uniform
τ	[0.01, 0.4]	uniform
n_s	[0.885, 1.04]	uniform
$\ln(10^{10} A_s)$	[2.5, 3.7]	uniform
A_{100}^{PS}	[0, 360]	uniform
A_{143}^{PS}	[0, 270]	uniform
A_{217}^{PS}	[0, 450]	uniform
A_{143}^{CIB}	[0, 20]	uniform
A_{217}^{CIB}	[0, 80]	uniform
A_{143}^{iSZ}	[0, 10]	uniform
$r_{143 \times 217}^{PS}$	[0, 1]	uniform
$r_{143 \times 217}^{CIB}$	[0, 1]	uniform
γ^{CIB}	[-2, 2]	uniform
c_{100}	[0.98, 1.02]	uniform
c_{217}	[0.95, 1.05]	uniform
$\xi^{iSZ-CIB}$	[0, 1]	uniform
A^{kSZ}	[0, 10]	uniform
β_1^l	[-20, 20]	uniform
$w(z_1)$	[-2, -0.01]	uniform
$w(z_2)$	[-2, -0.01]	uniform
$w(z_3)$	[-2, -0.01]	uniform
$w(z_4)$	[-2, -0.01]	uniform
$w(z_5)$	[-2, -0.01]	uniform
z_2	[0.01, 2.0]	sorted-uniform
z_3	[0.01, 2.0]	sorted-uniform
z_4	[0.01, 2.0]	sorted-uniform
n	[$\Lambda, w, r, 1, 2, 3$]	uniform

‘Vanilla’ Method

- Calculate Bayes factors with Evidences

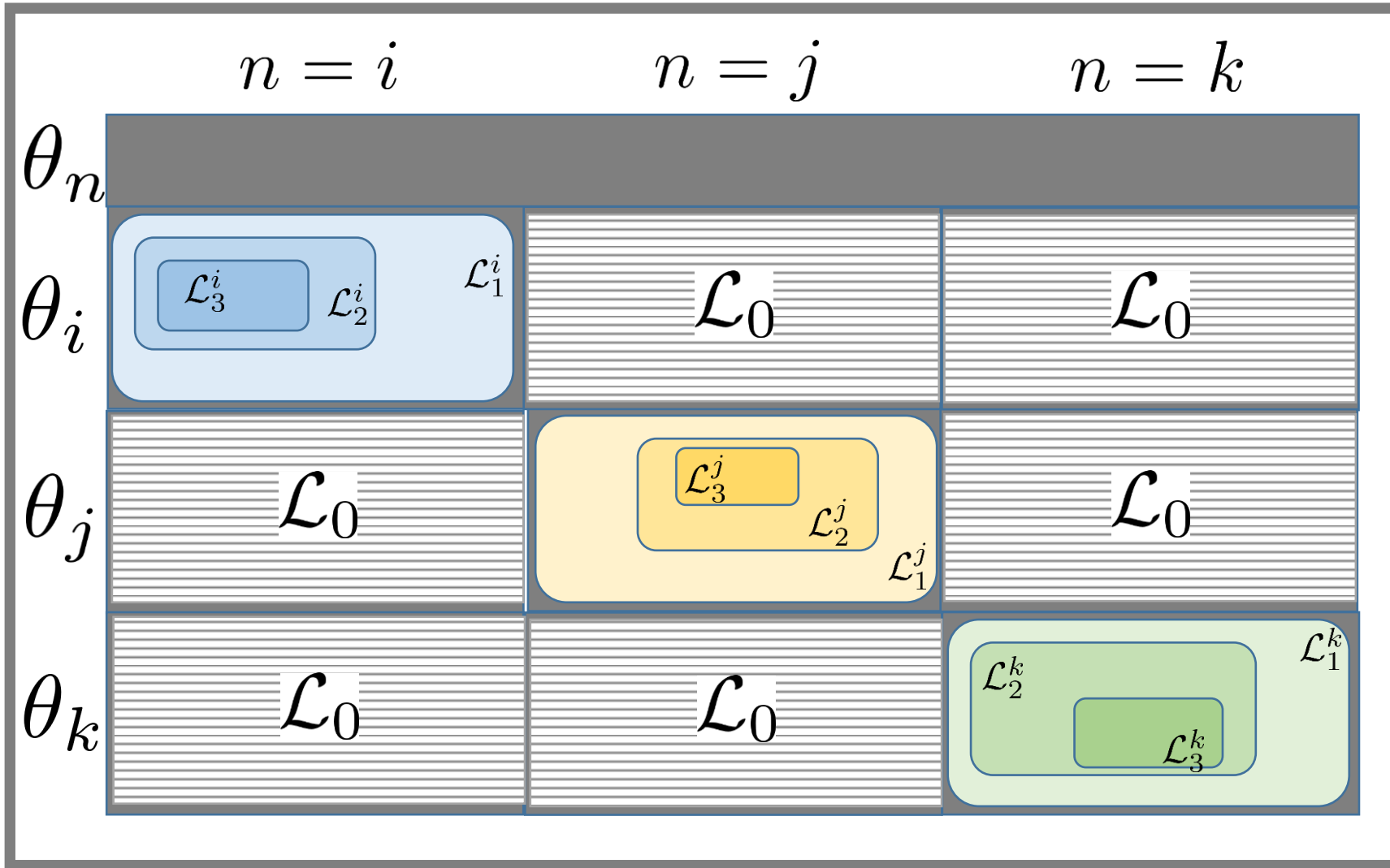


$$\mathcal{B}_{ij} = \ln\left(\frac{\mathcal{Z}^j}{\mathcal{Z}^i}\right)$$

$$\mathcal{B}_{ik} = \ln\left(\frac{\mathcal{Z}^k}{\mathcal{Z}^i}\right)$$

Our Solution

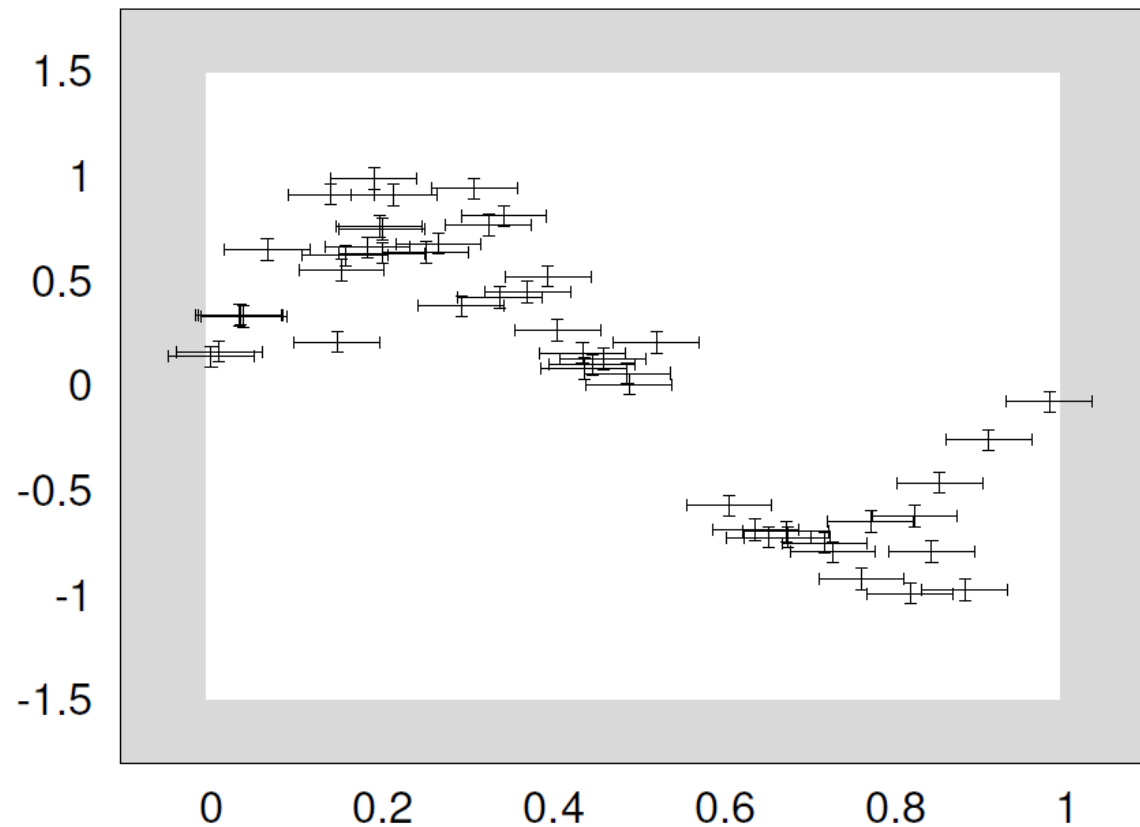
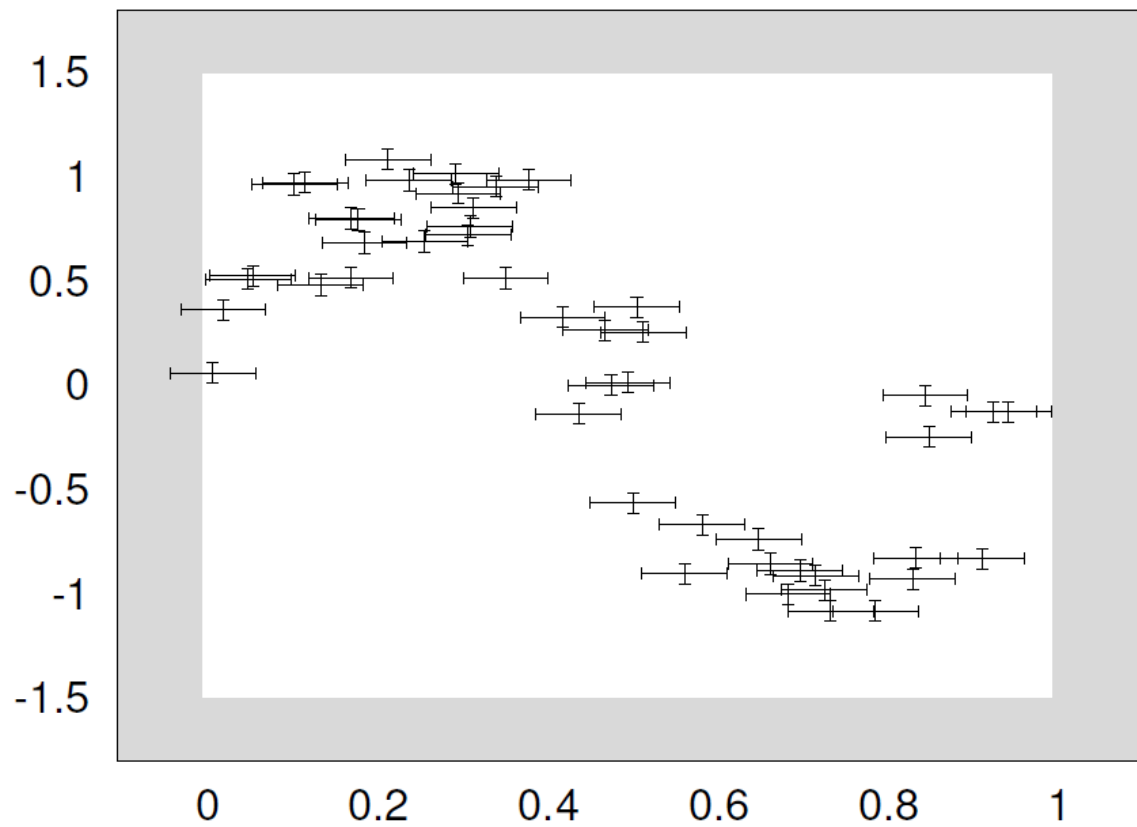
Hypermodel \mathcal{M}^n



Gives us:
 $Post(n) = Pr(n|\mathcal{D}, \theta)$

From which we get
 $\mathcal{B}_{ij} = \ln\left(\frac{Post(n=j)}{Post(n=i)}\right)$

Our Solution – Testing





Introduction



Background



The Problem



The Solution

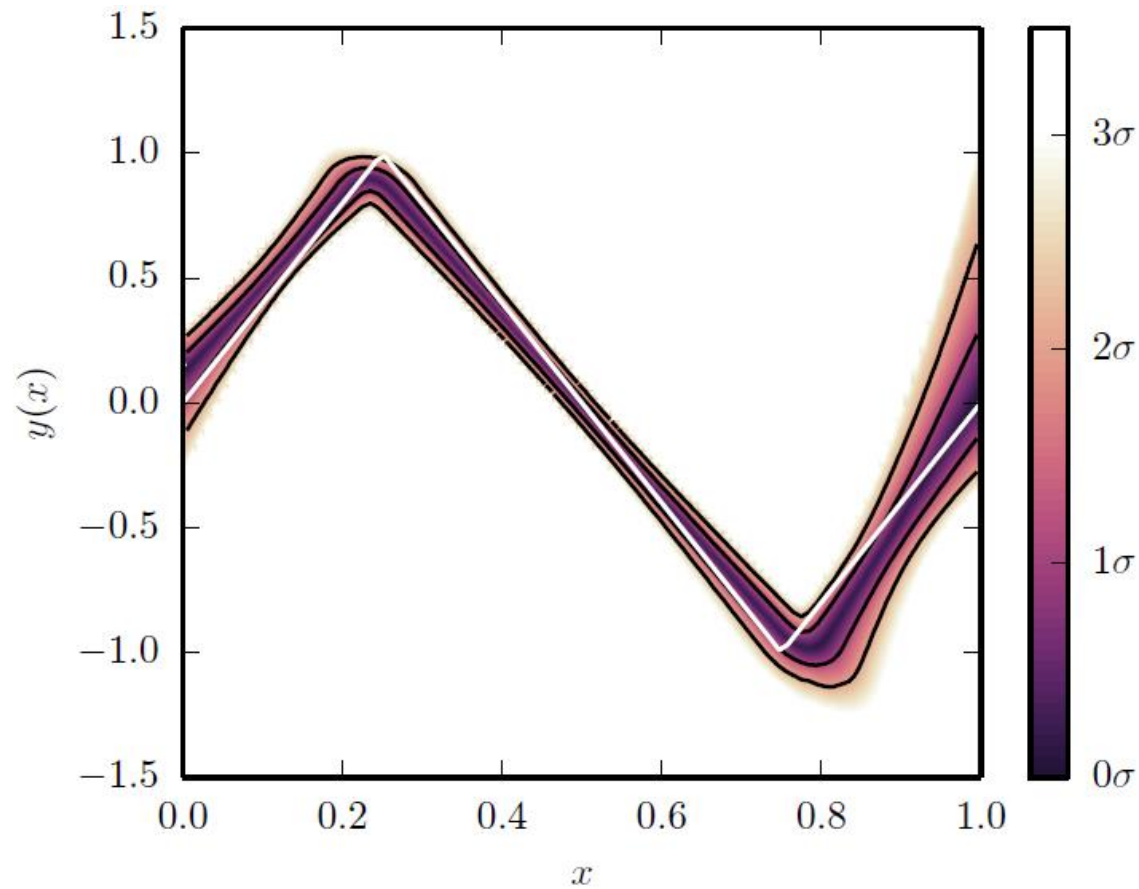
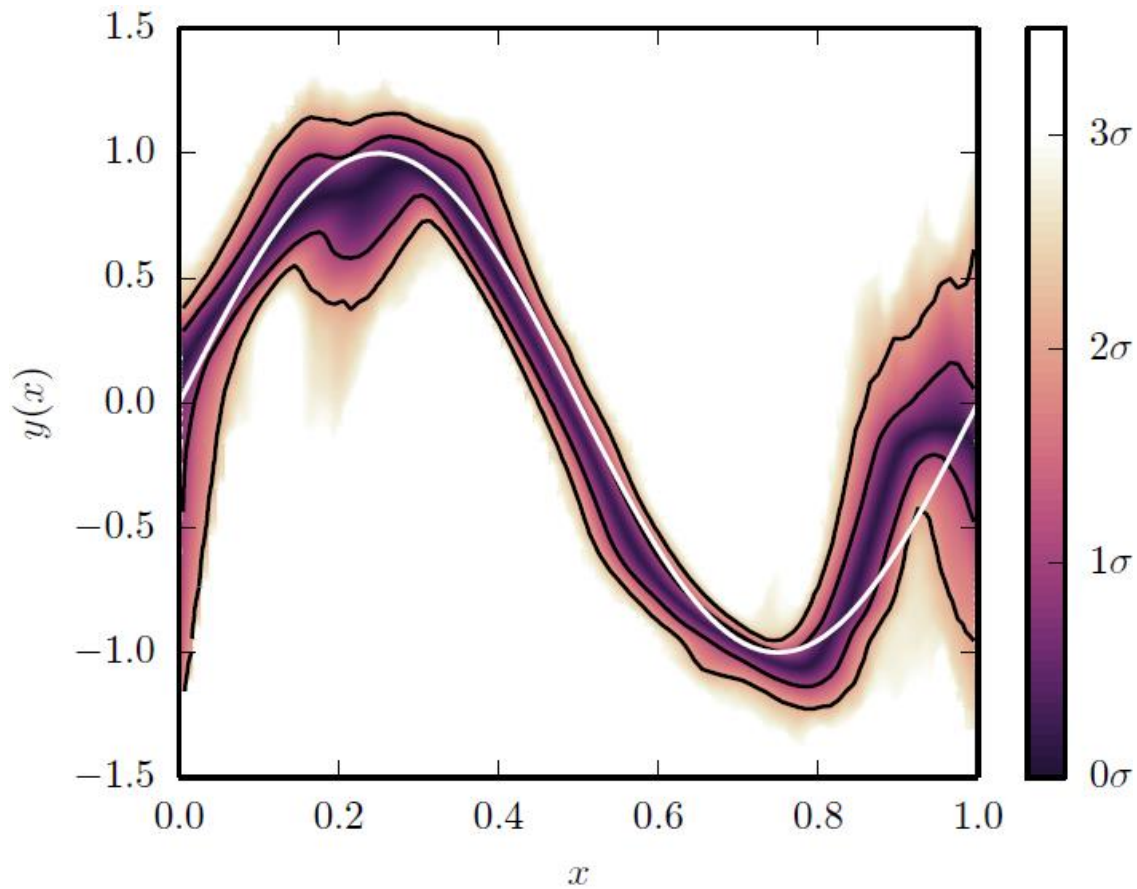


Physics – finally!

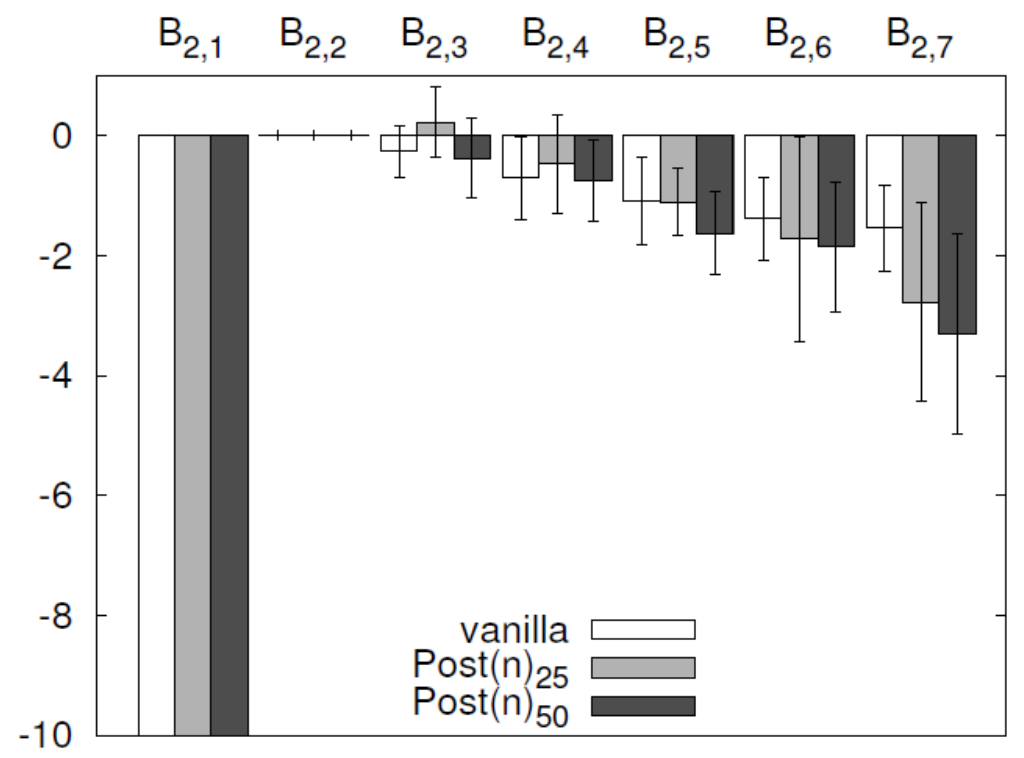
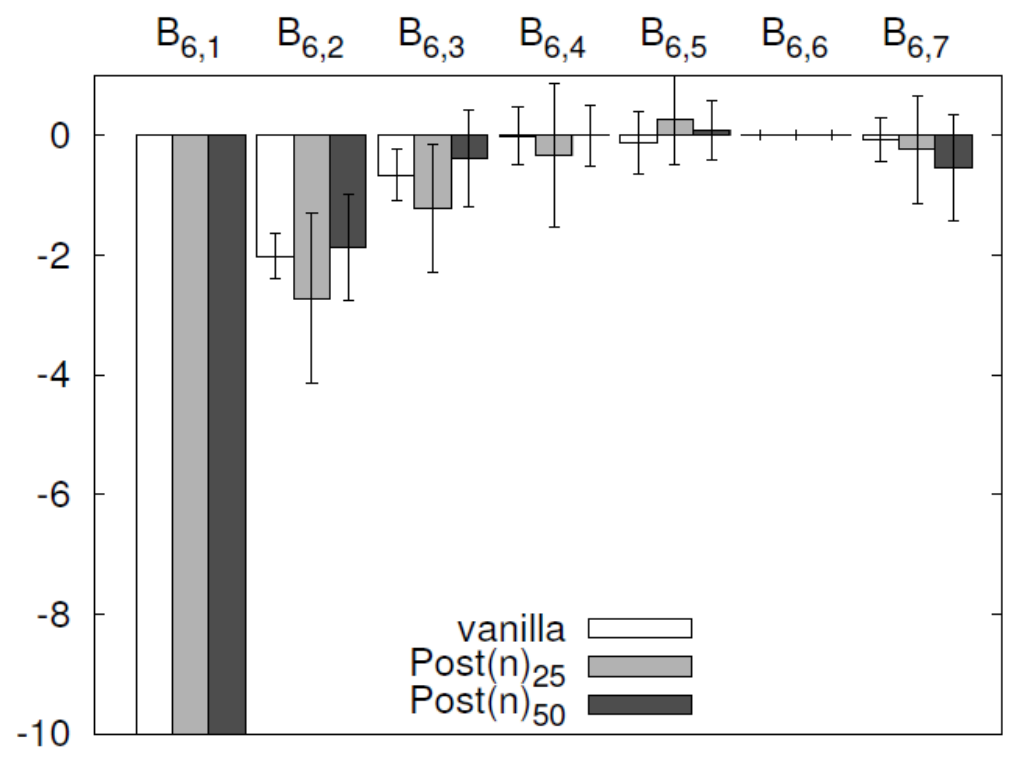


Thanks!

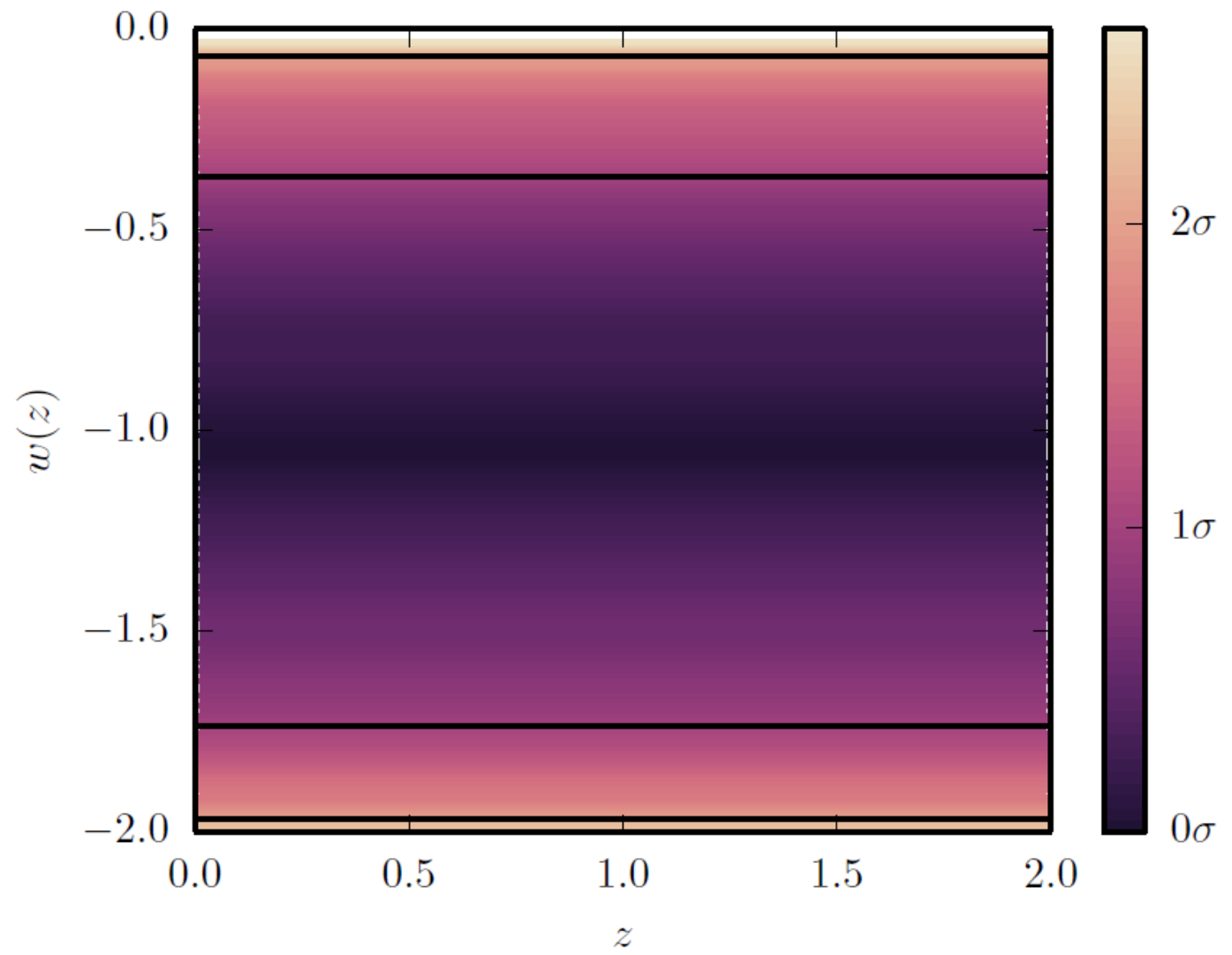
Our Solution – Works!



Our Solution – Works!



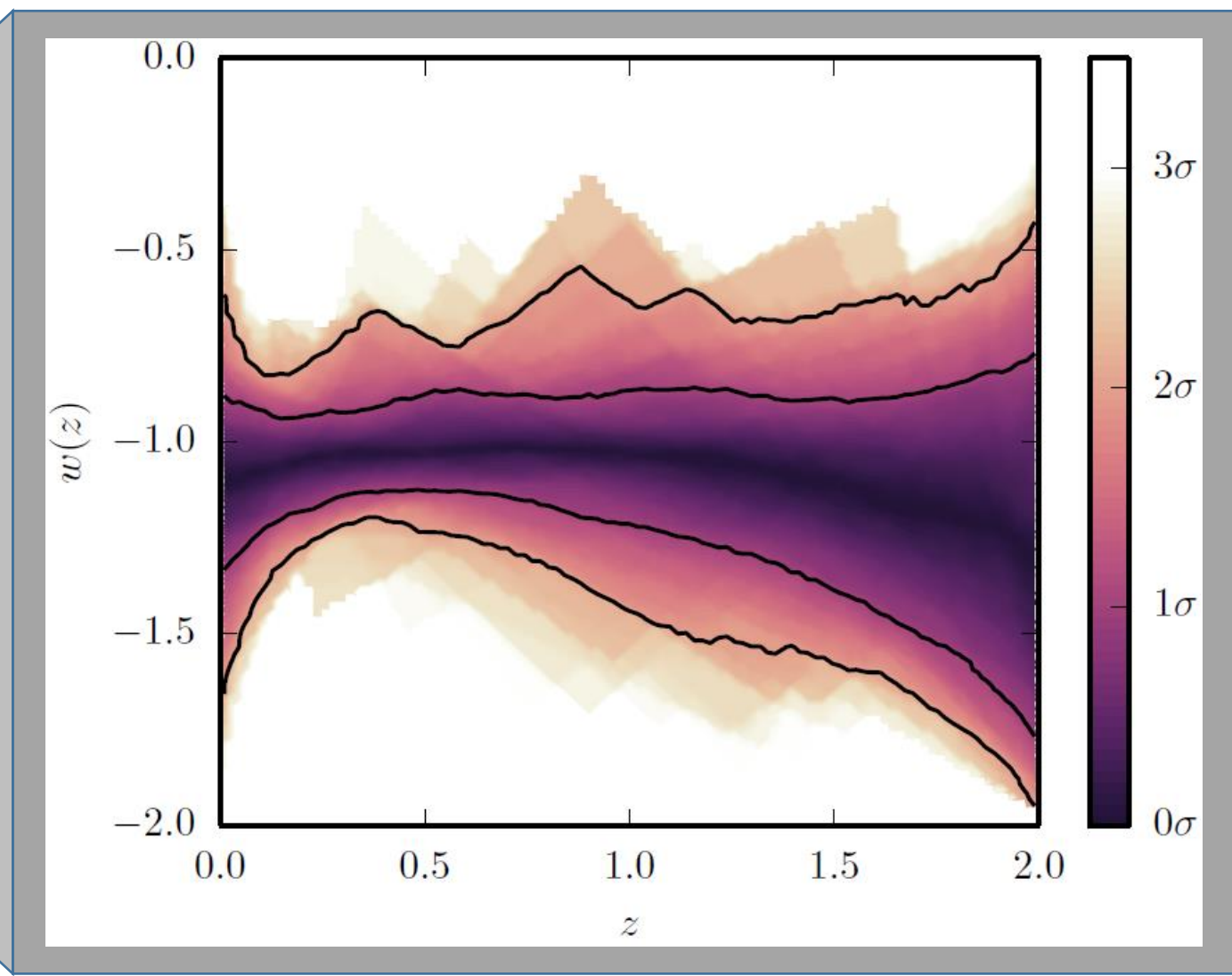
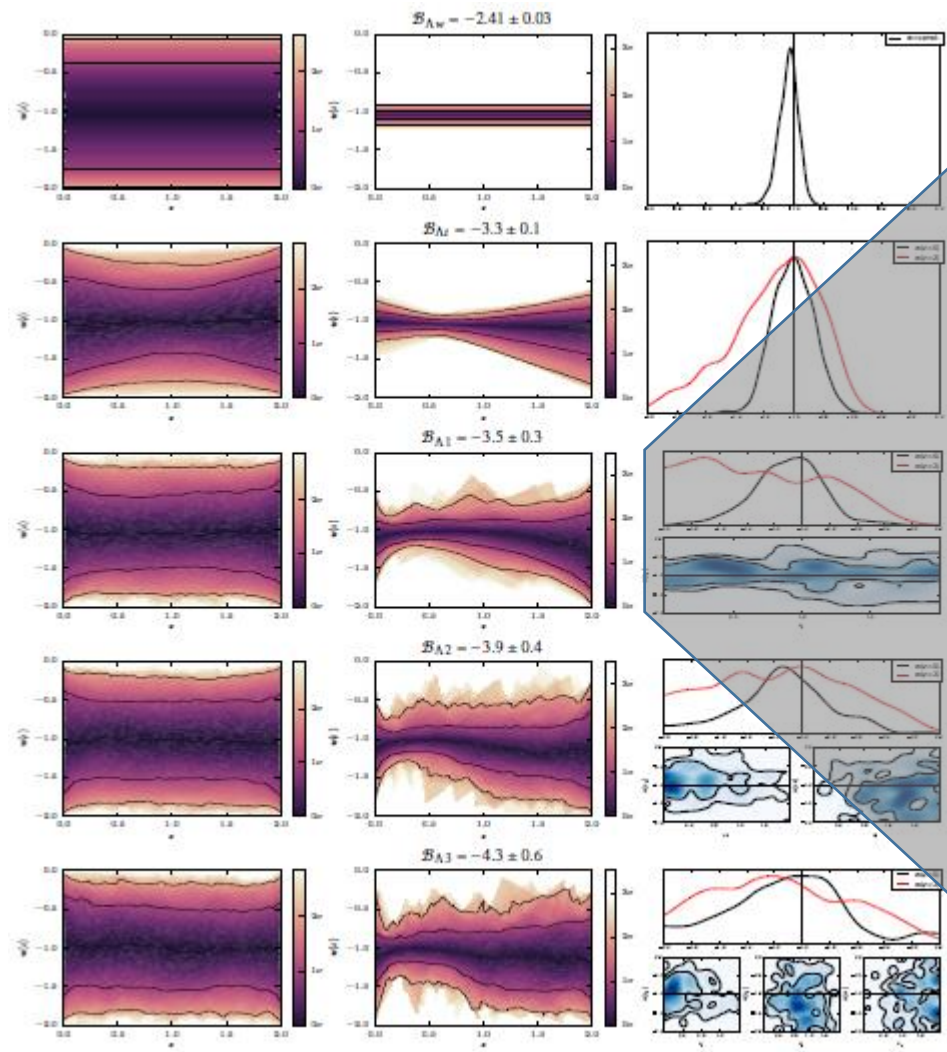
So let's do some Physics – finally!



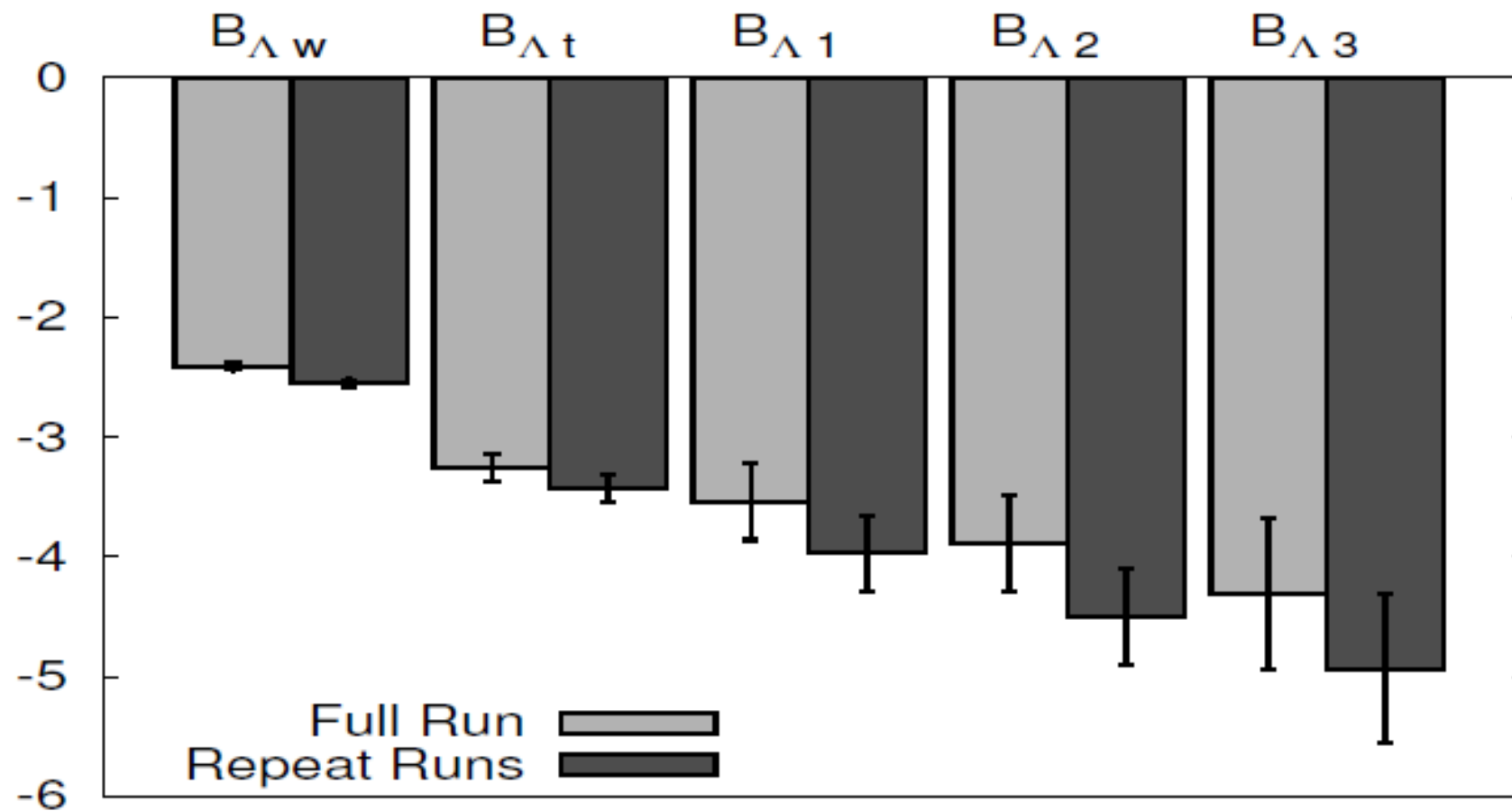
+

Planck 2013 CMB data
SN Ia Union 2.0
BAO (BOSS DR11)

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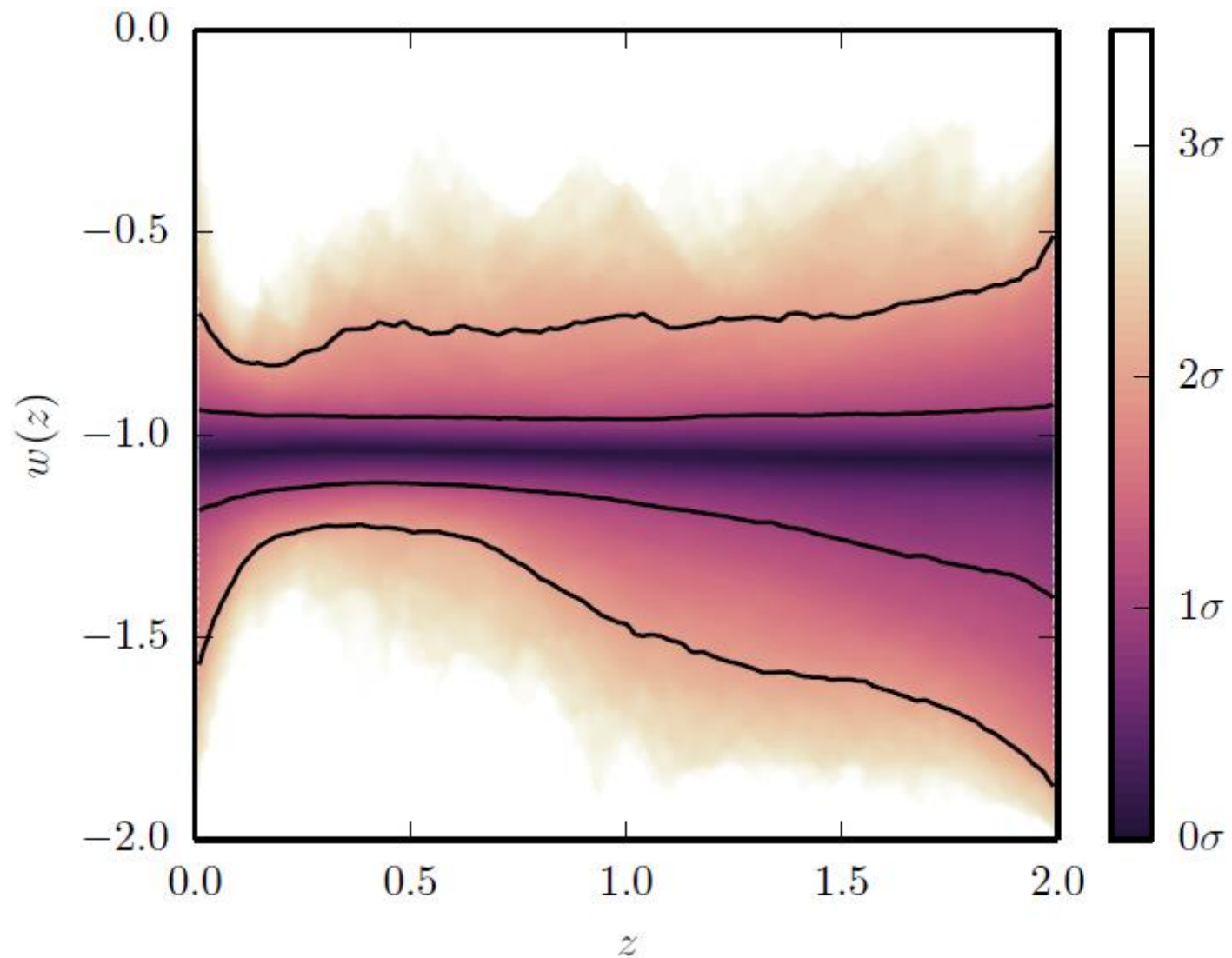


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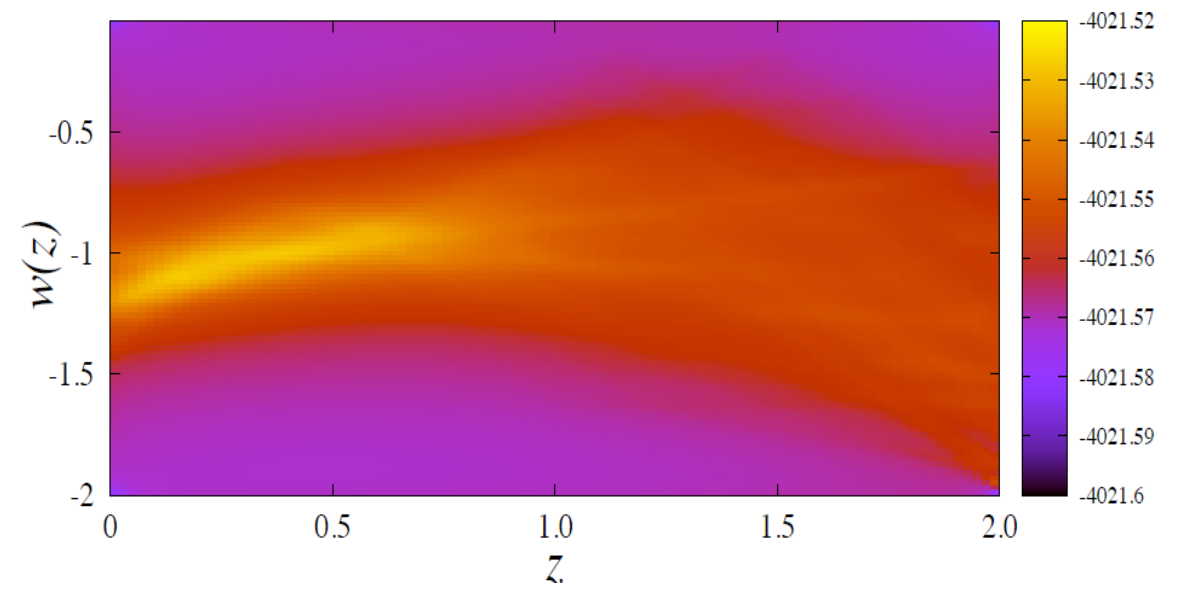


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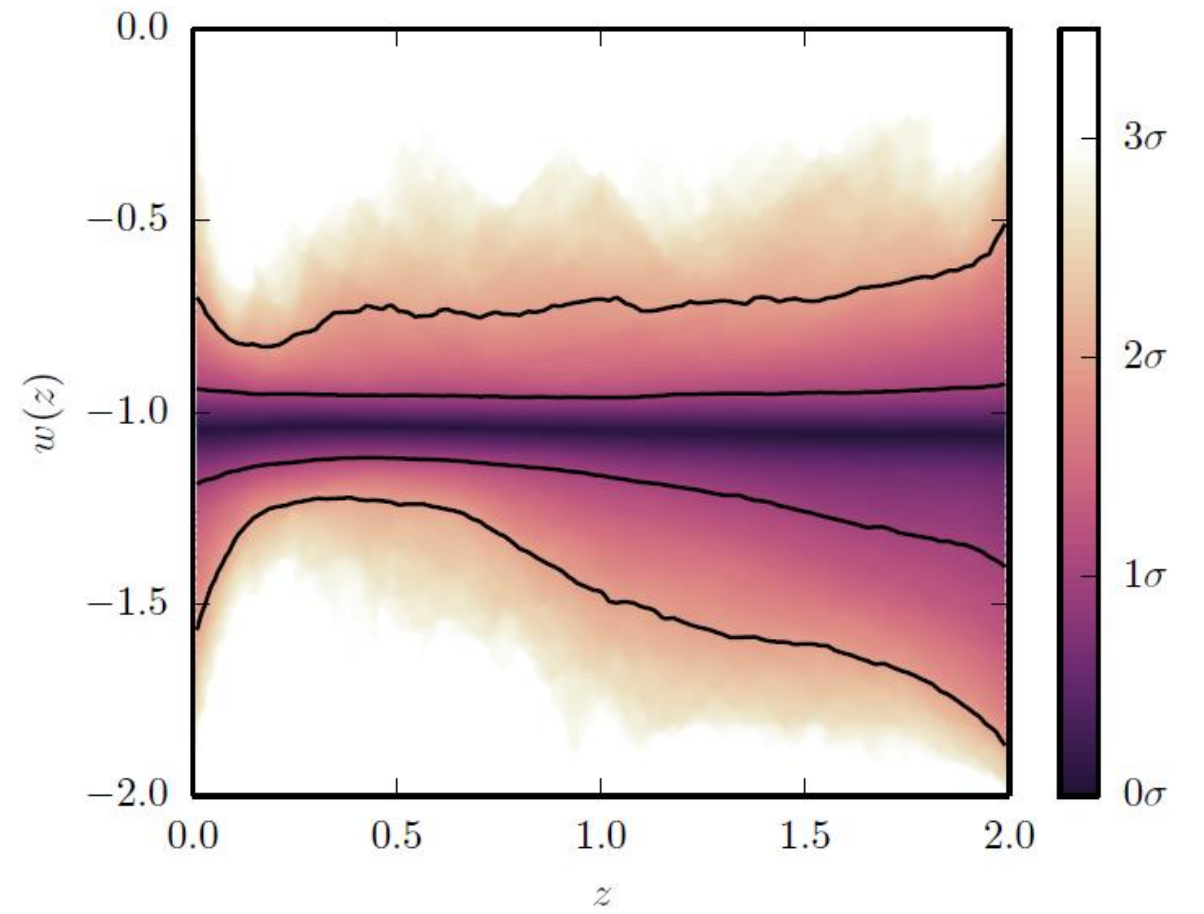


Final Conclusions

WMAP era Datasets



Planck 2013 era Datasets





Thanks

S. Hee, W. Handley, M. Hobson and A. Lasenby (2015)
“Bayesian model selection without evidences: application to the dark energy equation-of-state”. Arxiv: 1506.09024, submitted to MNRAS.

To my supervisors!

To Will! (POLYCHORD)

To Rich and Do Young!

Cambridge University,
and STFC for funding!

