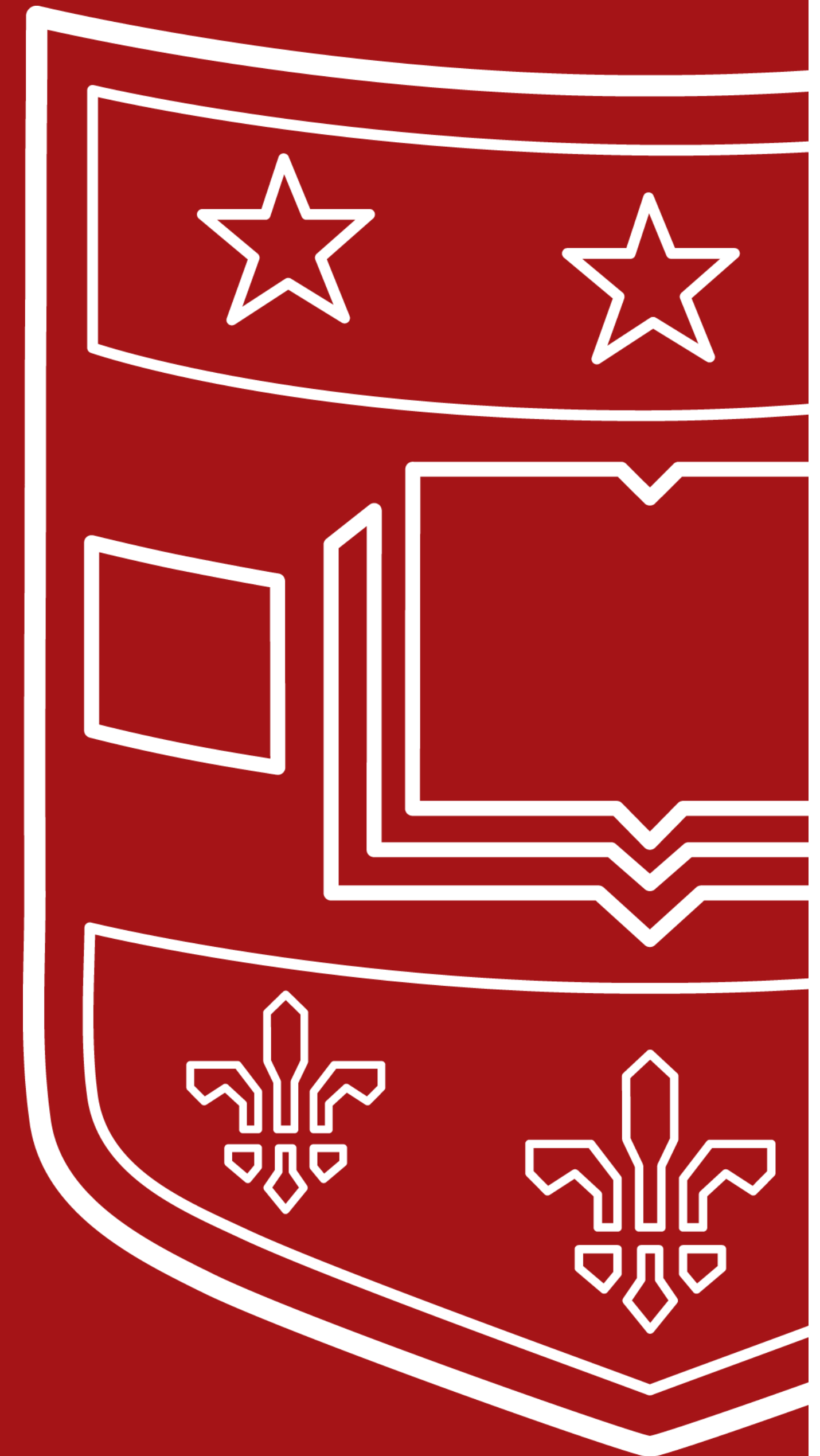


# Certainifying Uncertainty: Bayesian parameter and truncation error estimation in EFTs

Jason Bub

In collaboration with: Maria Piarulli, Daniel Phillips,  
Dick Furnstahl, and Saori Pastore

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# Next-Generation $\chi$ EFT Interactions



We are interested in calibrating the next generation of EFT nucleon-nucleon interactions.

These models should have robust uncertainty quantification:

- Parametric uncertainty
- Truncation uncertainty

This must be accomplished in the model calibration.



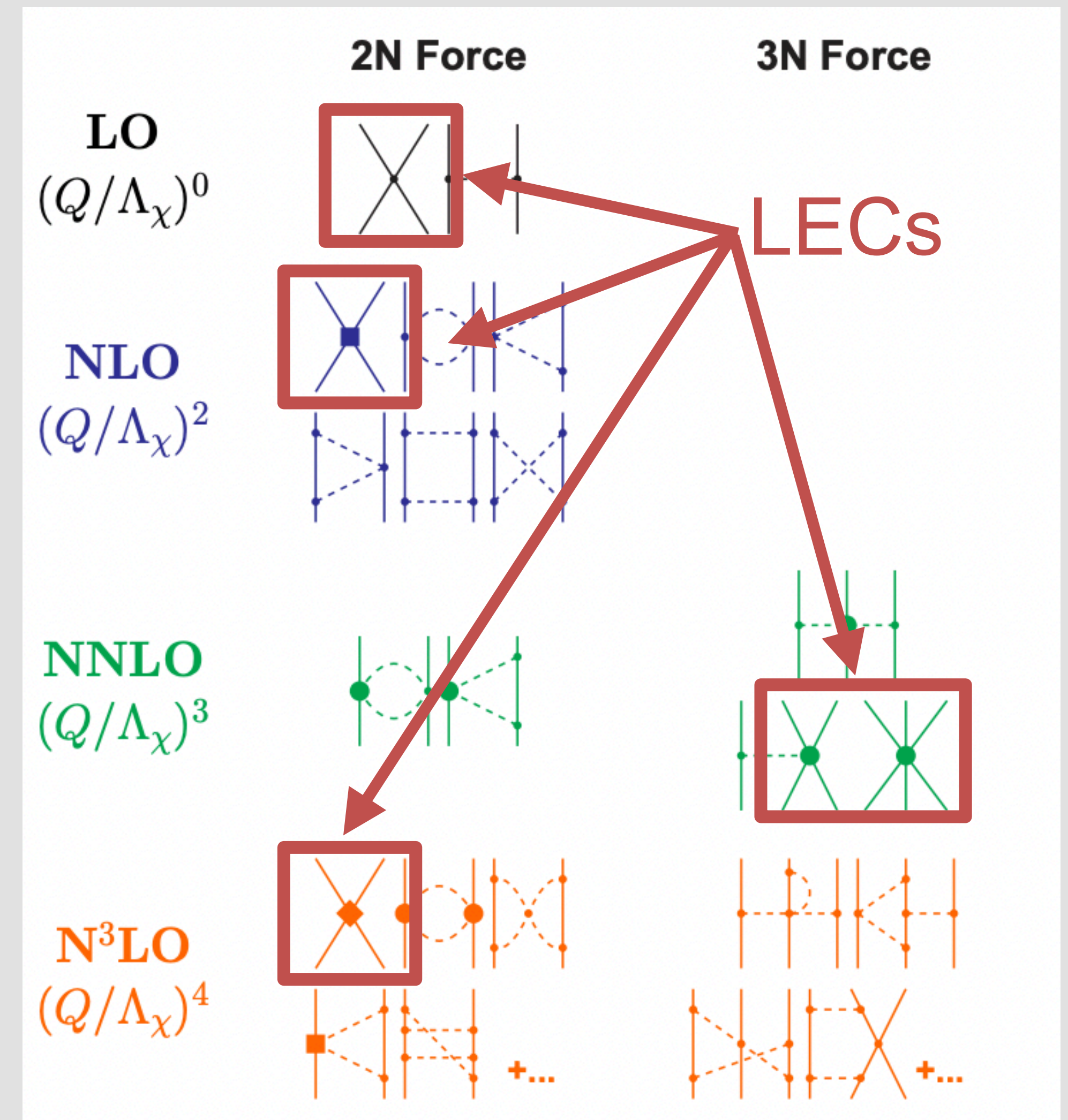
# Effective Field Theory

We take an effective expansion of QCD preserving chiral symmetry with N and  $\pi$  d.o.f.

The interaction can be ordered in terms of powers of  $p/\Lambda_\chi$

- $p$  is a momentum or pion mass
- $\Lambda_\chi$  is the symmetry breaking scale

Gives a systematic ordering to improve the interaction.





# Bayes' Theorem

For a model calibration problem in a Bayesian approach, we have

$$\underbrace{\text{pr}(\vec{a} | \vec{y}, \mathbf{I})}_{\text{Posterior}} \propto \underbrace{\text{pr}(\vec{y} | \vec{a})}_{\text{Likelihood}} \underbrace{\text{pr}(\vec{a} | \mathbf{I})}_{\text{Prior}}$$

The **likelihood** can take the form it uses in normal model fitting

$$\text{pr}(\vec{y} | \vec{a}) \sim e^{-\sum_i \left( y_{\text{exp}}^{(i)} - y_{\text{th}}^{(i)}(\vec{a}) \right)^2 / 2\sigma_i^2} = e^{-\chi^2/2}$$

What the **prior** does for us is **encode any previous information** that we may know.

- Ex: LECs are natural, i.e., order 1  $\rightarrow \text{pr}(\vec{a} | \mathbf{I}) \sim \mathcal{N} \left( \vec{0}, \Sigma_{\text{pr}} \right)$

# Likelihood Improvement



In the simple **likelihood**, we had the  $\chi^2$ ,  $\left(e^{-\chi^2/2}\right)$ , but we can improve this.

We can inform the model calibration with information about the model *itself*.

In what way?

$$\chi^2 = \sum_i \frac{\left(y_{\text{exp}}^{(i)} - y_{\text{th}}^{(i)}(\vec{a})\right)^2}{\sigma_i^2} \longrightarrow \sigma_i^2 \rightarrow \sigma_i^2 + \sigma_{\text{th},i}^2$$



# Theory Uncertainty in Calibration

Why are theory errors necessary in calibration?

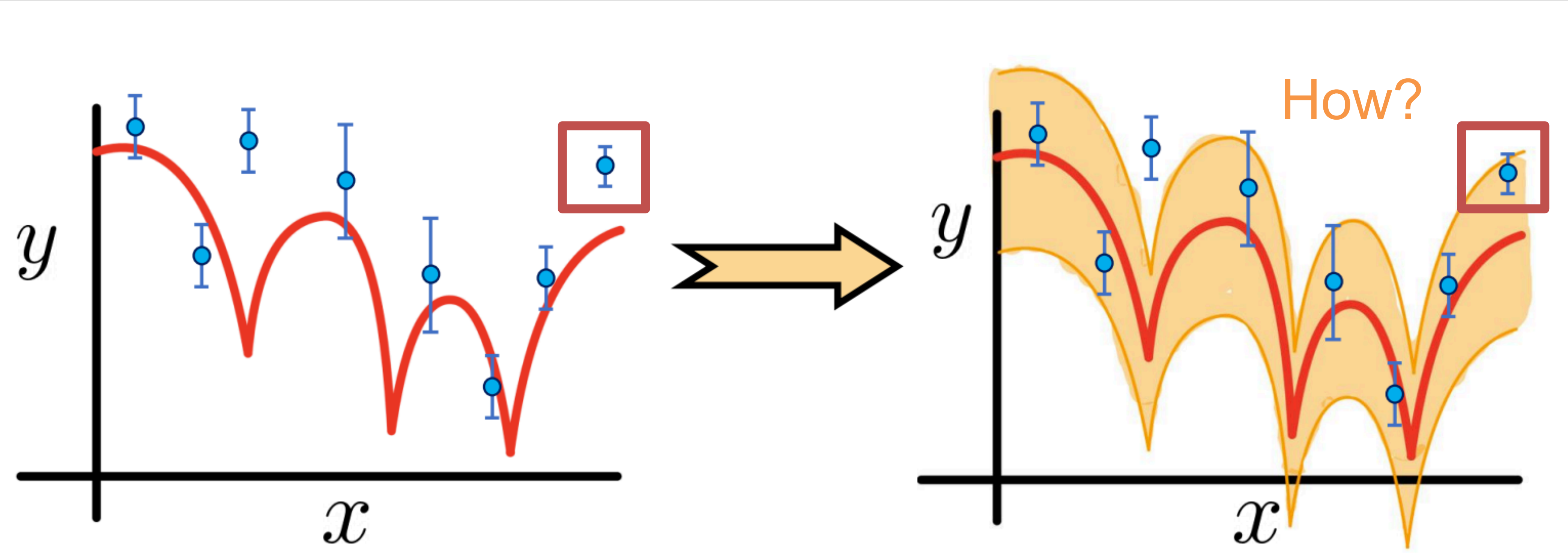


Figure courtesy of Pablo Giuliani



# Modeling the Model

Since our model is a perturbative series, we can model it as such\*:

$$y_{\text{th}}(x) = y_{\text{ref}}(x) \sum_{n=0}^{\infty} c_n(x) Q^n(x), \quad Q \equiv \frac{\max[p_{\text{soft}}, p]}{\Lambda_b},$$

where  $y_{\text{ref}}(x)$  sets a reference scale for the observable  $y_{\text{th}}$  and  $\Lambda_b$  is the EFT breakdown scale.

This series follows the truncation scheme of the EFT:

$$y_{\text{th}}(x) = y_{\text{ref}}(x) \sum_{n=0}^k c_n Q^n + y_{\text{ref}}(x) \sum_{n=k+1}^{\infty} c_n Q^n = y_{\text{th}}^{(k)}(x) + \delta y_{\text{th}}^{(k)}(x).$$

\*R. J. Furnstahl et. al. Phys. Rev. C **92**, 024005



# Truncation Errors

From the neglected terms, we have

$$\delta y_{\text{th}}^{(k)}(x) = y_{\text{ref}}(x) \sum_{n=k+1}^{\infty} c_n(x) Q^n(x).$$

This is a geometric series in  $Q$ , so we can find\*

$$\delta y_{\text{th}}^{(k)}(x) = \frac{y_{\text{ref}} \bar{c} Q^{(k+1)}}{1 - Q},$$

Where we assume that  $c_n | \bar{c} \sim \mathcal{N}(0, \bar{c}^2)$ .





# Theoretical Covariance

From the truncation uncertainty, we can construct a covariance matrix, assuming  $\delta y_{\text{th}}$  is normally distributed,

$$\Sigma_{ij}^{\text{th}} = y_{\text{ref},i} y_{\text{ref},j} \frac{\left( y_{\text{ref},i} \bar{c} Q_i^{(k+1)} \right) \left( y_{\text{ref},j} \bar{c} Q_j^{(k+1)} \right)}{1 - Q_i Q_j} r(x_i, x_j; \vec{l}),$$

where we introduce a kernel  $r(x_i, x_j; \vec{l})$  to smooth and handle correlations.



# Correlated Likelihood

We can build a total covariance,

$$\Sigma_{ij} = \Sigma_{ij}^{\text{exp}} \delta_{ij} + \Sigma_{ij}^{\text{th}}.$$

And our correlated likelihood is now

$$\text{pr}(\vec{y} \mid \vec{a}, \mathbf{I}) \propto e^{-\left(\vec{y}_{\text{exp}} - \vec{y}_{\text{th}}\right)^{\text{T}} \Sigma^{-1} \left(\vec{y}_{\text{exp}} - \vec{y}_{\text{th}}\right)} = e^{-d_M(\vec{a})}$$

where we define the **Mahalanobis distance**

$$d_M(\vec{a}) = \left(\vec{y}_{\text{exp}} - \vec{y}_{\text{th}}\right)^{\text{T}} \Sigma^{-1} \left(\vec{y}_{\text{exp}} - \vec{y}_{\text{th}}\right).$$

Correlated version of  $\chi^2$

# Additional Parameters



In this process, we have introduced two new parameters:  $\bar{c}$  and  $\Lambda_b$ .

This changes the posterior we need to find:

$$\underbrace{\text{pr}(\vec{a}, \bar{c}^2, \Lambda_b | \vec{y}_{\text{exp}}, \mathbf{I})}_{\text{Total posterior}} \propto \underbrace{\text{pr}(\vec{y}_{\text{exp}} | \vec{a}, \Sigma, \mathbf{I})}_{\text{Likelihood for } \vec{a}} \underbrace{\text{pr}(\vec{a} | \mathbf{I})}_{\text{Prior for } \vec{a}} \underbrace{\text{pr}(\bar{c}^2 | \Lambda_b, \vec{a}, \mathbf{I})}_{\text{Posterior for } \bar{c}^2} \underbrace{\text{pr}(\Lambda_b | \vec{a}, \mathbf{I})}_{\text{Posterior for } \Lambda_b} .$$

We can find a closed form of  $\text{pr}(\bar{c}^2 | \Lambda_b, \vec{a}, \mathbf{I})$  and  $\text{pr}(\Lambda_b | \vec{a}, \mathbf{I})$ .



# Posterior for $\bar{c}$

Since we had  $c_n | \bar{c} \sim \mathcal{N}(0, \bar{c}^2)$ , where  $\bar{c}^2$  is a population variance, we make the standard choice of prior for an unknown variance:

$$\bar{c}^2 \sim \chi^{-2}(\nu_0, \tau_0^2)$$

This yields a conjugate posterior

$$\text{pr}(\bar{c}^2 | \mathbf{I}) \sim \chi^{-2}(\nu_0, \tau_0^2) \iff \text{pr}(\bar{c}^2 | \vec{a}, \Lambda_b, \mathbf{I}) \sim \chi^{-2}(\nu, \tau^2(\vec{a}, \Lambda_b)).$$

Where we have hyperparameters:

$$\nu = \nu_0 + N_{\text{obs}} n_{\text{orders}}, \text{ degrees of freedom}$$
$$\tau^2(\vec{a}, \Lambda_b) = \frac{1}{\nu} \left( \nu_0 \tau_0 + \sum_{i,n} c_{n,i}^2(\vec{a}, \Lambda_b) \right), \text{ scale}$$
$$c_{n,i} = \frac{y_i^n - y_i^{(n-1)}}{y_{\text{ref},i} Q_i^n}$$



## Posterior for $\Lambda_b$

Our posterior for the breakdown scale also uses these hyperparameters:

$$\text{pr}(\Lambda_b | \vec{a}, \mathbf{I}) \propto \frac{\text{pr}(\Lambda_b | \mathbf{I})}{\tau^\nu \prod_{n,i} \left( \frac{p_i}{\Lambda_b} \right)^n}$$

This posterior needs to be numerically normalized as the normalization constant is dependent on  $\vec{a}$ .

With all our components, we can estimate our parameters.

# Pionless EFT



We are working in an EFT framework **without pions in Weinberg PC**

Our interaction takes the form:

$$y_{\text{th}}(x) = \frac{y_{\text{ref}}(x)}{y_{\text{exp}}(x)} \sum_{n=0}^{\infty} c_{2n}(x) Q^{2n}(x)$$

$$v_{\text{LO}} = C_S + C_T \sigma_1 \cdot \sigma_2$$

$$v_{\text{NLO}}^{\text{CI}}(\vec{k}, \vec{K}) = C_1 k^2 + C_2 k^2 \sigma_1 \cdot \sigma_2 + C_3 S_{12}(k) + C_4 k^2 \tau_1 \cdot \tau_2$$
$$+ iC_5 \vec{S} \cdot (\vec{K} \times \vec{k}) + C_6 k^2 \tau_1 \cdot \tau_2 \sigma_1 \cdot \sigma_2 + C_7 S_{12}(k) \tau_2 \cdot \tau_2$$
$$v_{\text{NLO}}^{\text{CD}} = C_0^{\text{IT}} T_{12} + C_0^{\text{IV}} (\tau_{1z} + \tau_{2z})$$

# Regularization



To use these interactions, they must be regularized in some fashion and must be local in coordinate space (for QMC).

We employ a Gaussian cutoff in coordinate space, which smears  $\delta$ -functions upon Fourier transformation

$$f(r) = \frac{1}{\pi^{3/2} R_s^3} e^{-\left(\frac{r}{R_s}\right)^2}$$

We choose  $R_s \in [1.5, 2.0, 2.5]$  fm which are  $\sim \frac{400}{R_s}$  MeV in momentum space.



# Parameter Estimation Algorithm

To estimate all of these parameters, we need data to calibrate to:

Our choice of data is the pp and np Granada database (differential cross sections, total cross sections) up to 5 MeV + deuteron binding energy + nn scattering length.

We then use Markov Chain Monte Carlo (MCMC) to sample the posteriors at LO ( $Q^0$ ), NLO ( $Q^2$ ), and N3LO ( $Q^4$ ), allowing for the order-by-order convergence analysis for LO  $\rightarrow$  NLO and NLO  $\rightarrow$  N3LO to estimate  $\bar{c}$  and  $\Lambda_b$ .

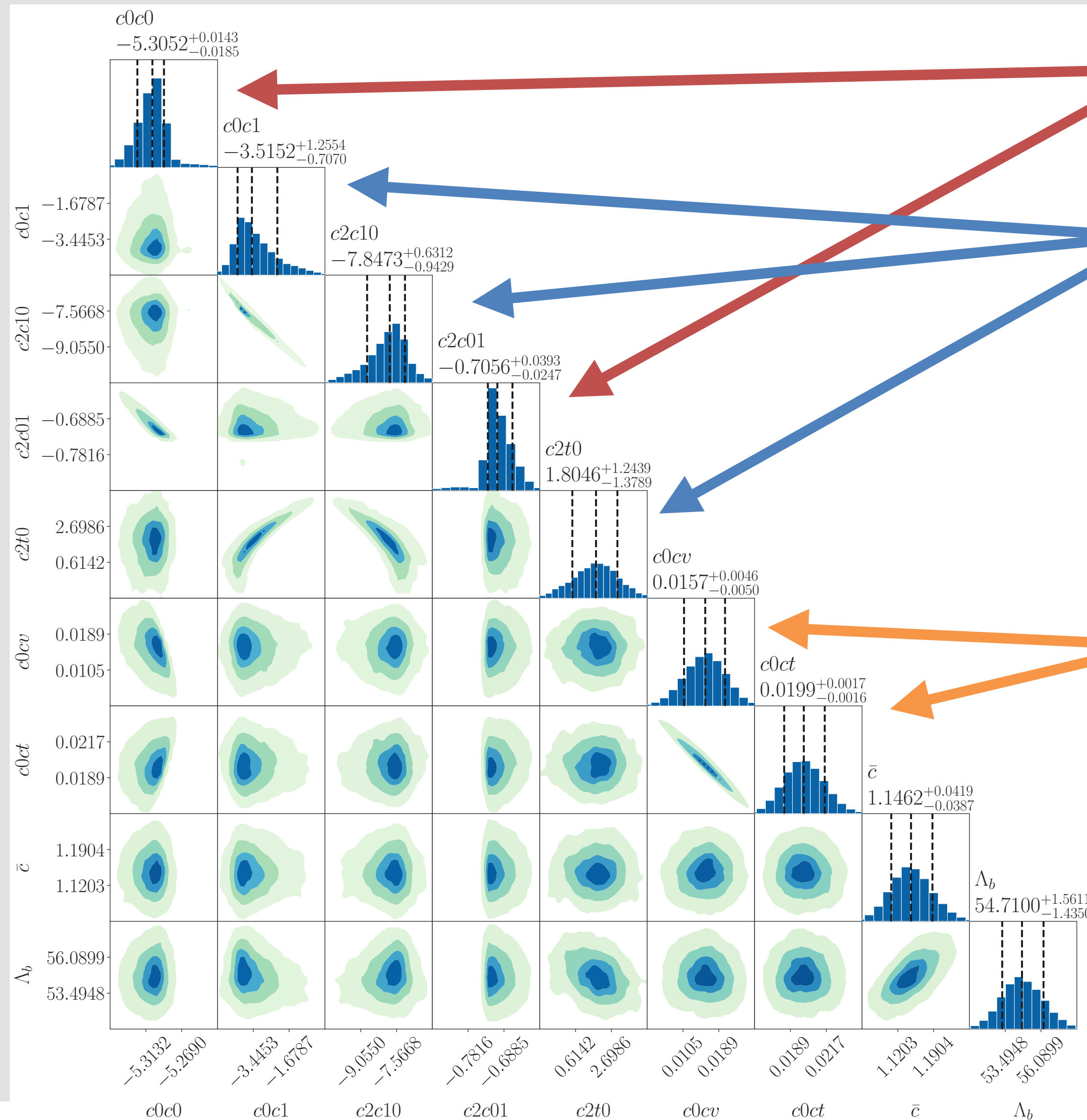




# Prior Choices

- $\text{pr}(\vec{a} | \text{I}) \sim \mathcal{N} \left( \vec{a}_{\text{p.s}}^{\text{MAP}}, \overline{10^2} \right)$
- $\text{pr}(\Lambda_b | \text{I}) \sim \mathcal{N} (500 \text{ MeV}, 1000^2 \text{ MeV})$
- $\text{pr}(\bar{c}^2 | \text{I}) \sim \chi^{-2}(\nu_0 = 1.5, \tau_0^2 = 1.5^2)$
- $r(x_i, x_j; \vec{l}) = e^{|p_i - p_j|/2l_p} e^{|\theta_i - \theta_j|/2l_\theta} \delta_{\text{type}_i, \text{type}_j}, \quad l_p = 0.3 \text{ MeV}, l_\theta = 20^\circ$
- $P_{\text{soft}} = \begin{cases} p_d \sim 45 \text{ MeV}/c, & \text{for } np \text{ and } nn \text{ scattering} \\ 1/{}^1a_{pp} \sim 25 \text{ MeV}, & \text{for } pp \text{ scattering.} \end{cases}$

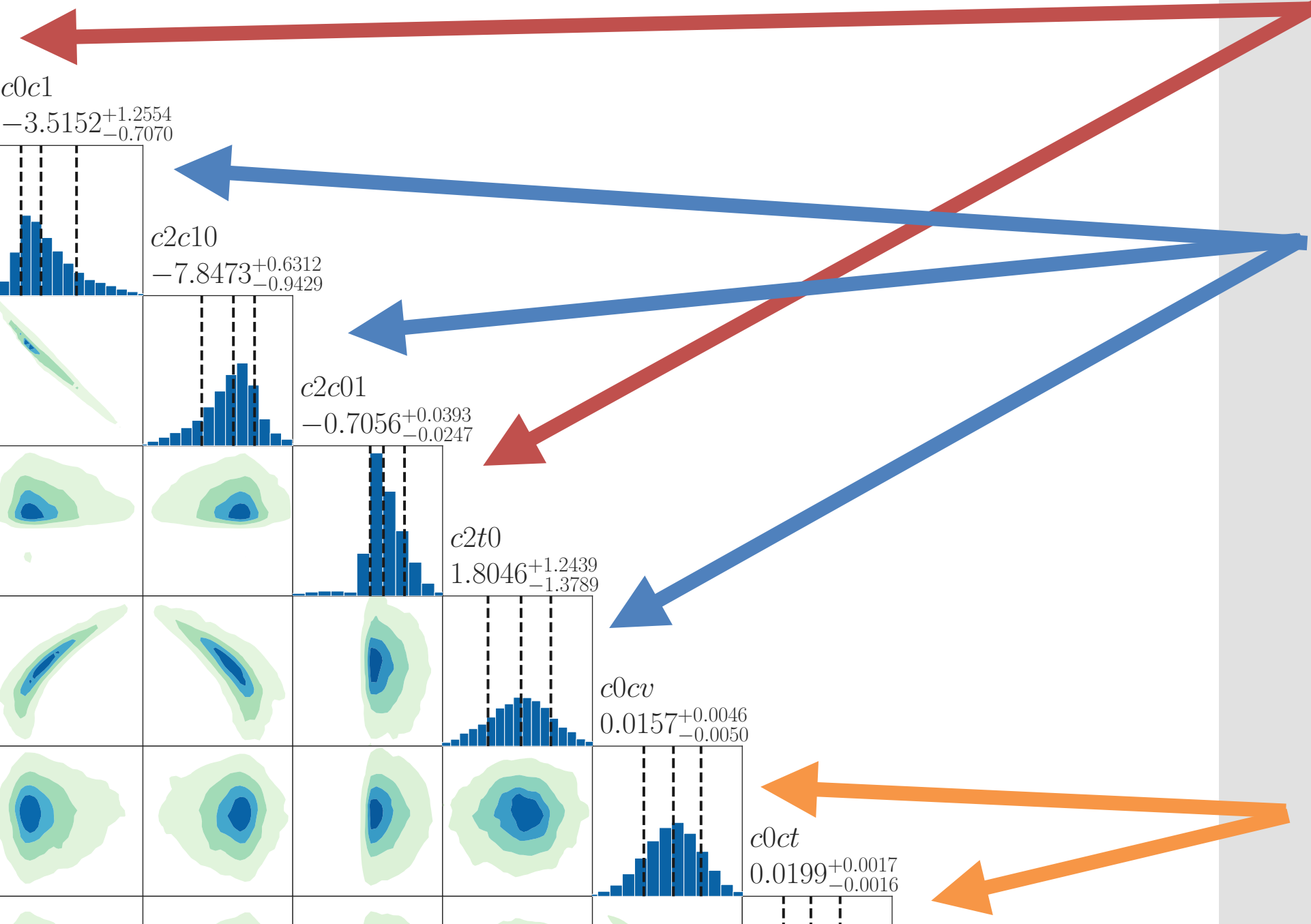
# NLO Posterior



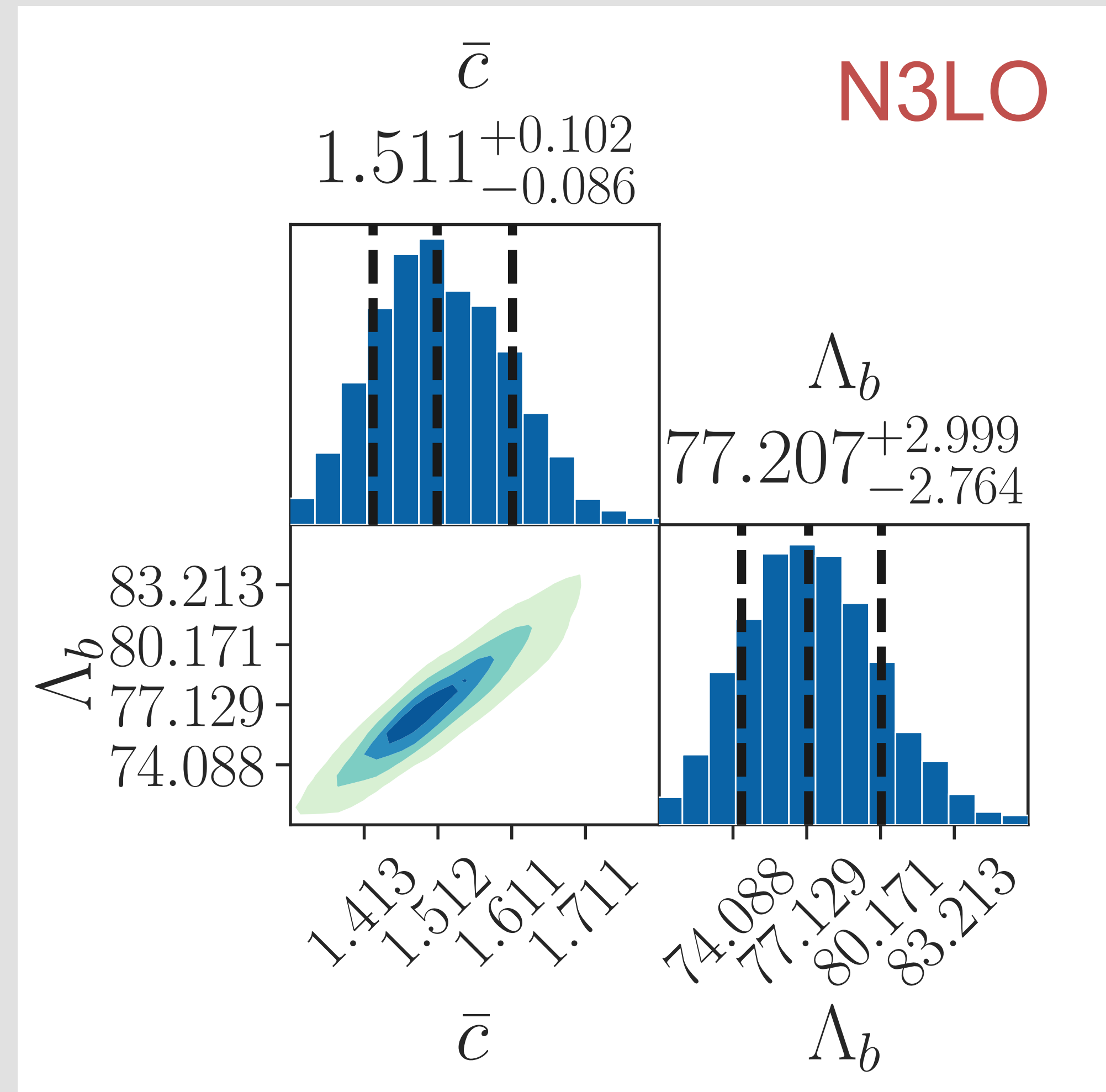
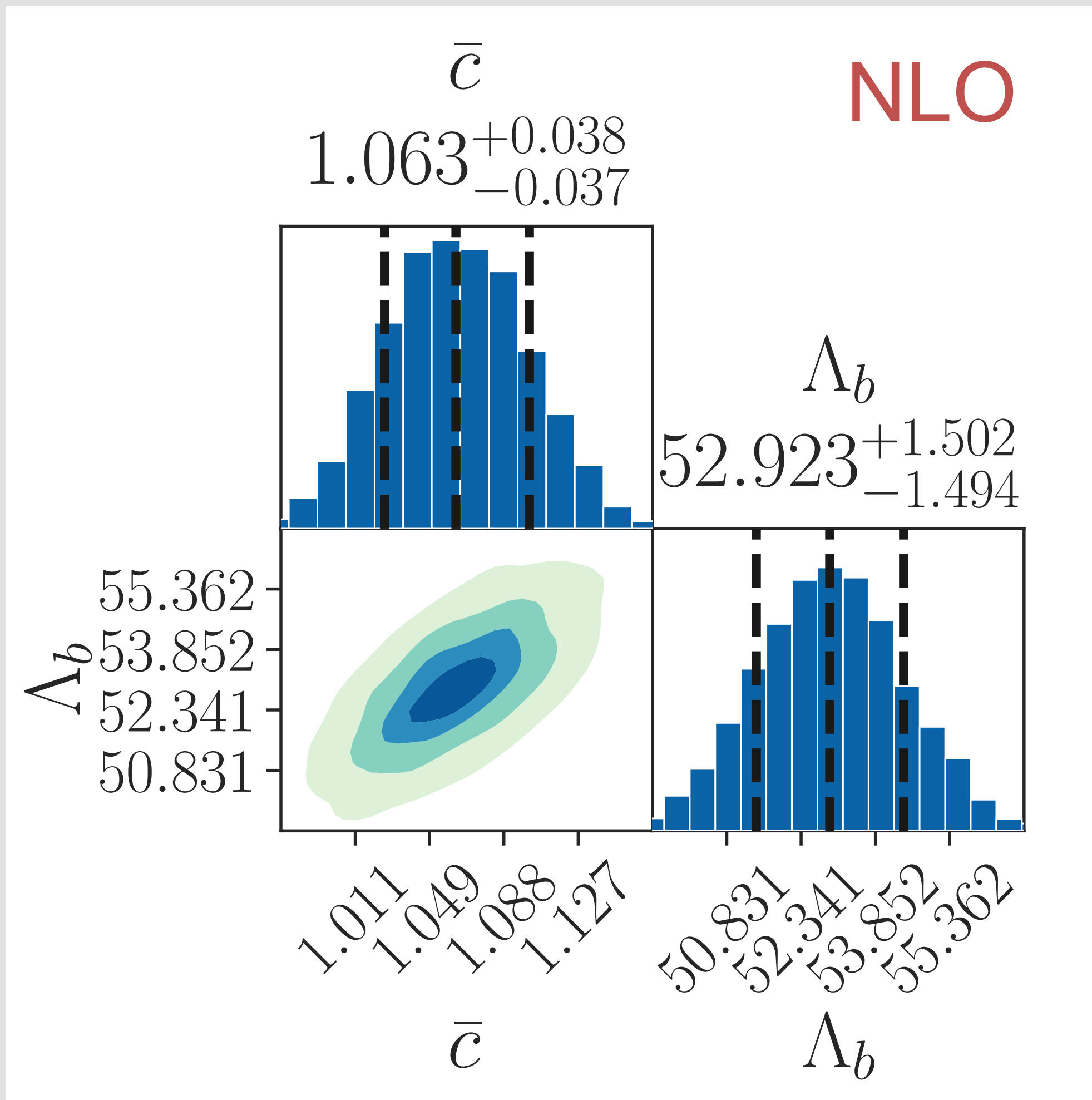
$(S, T) = (0, 1)$

$(S, T) = (1, 0)$

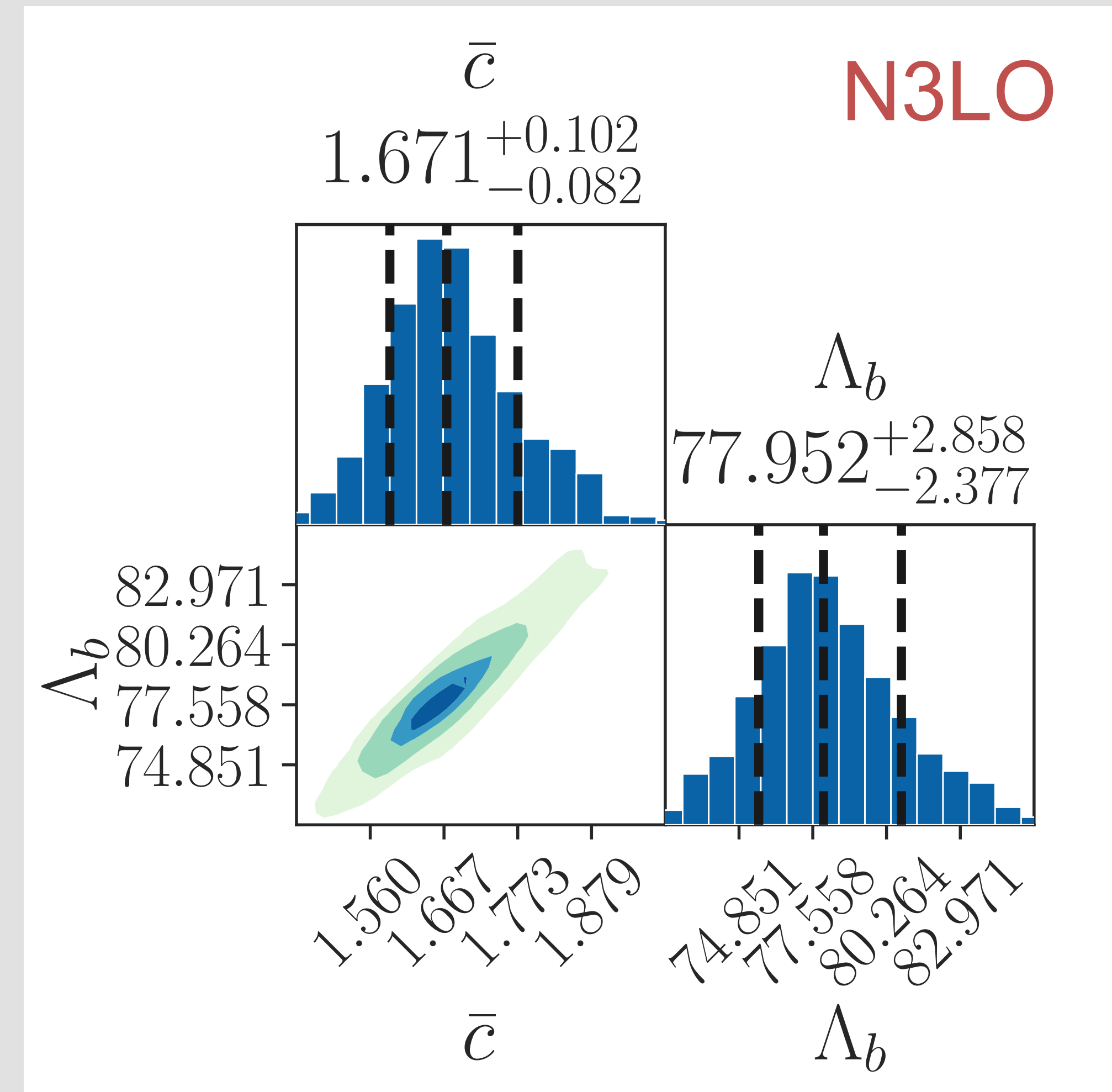
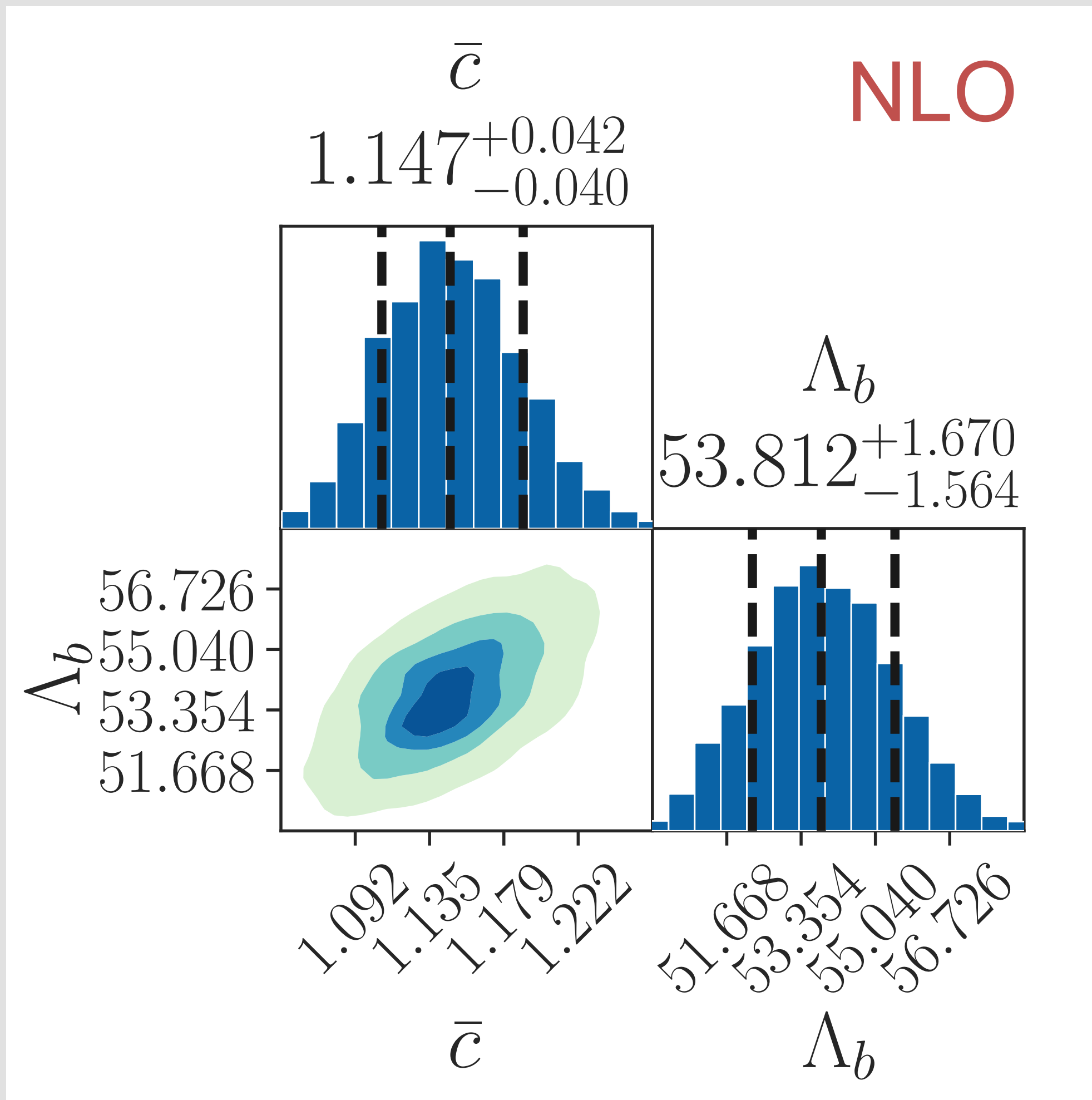
$C^{CD}$



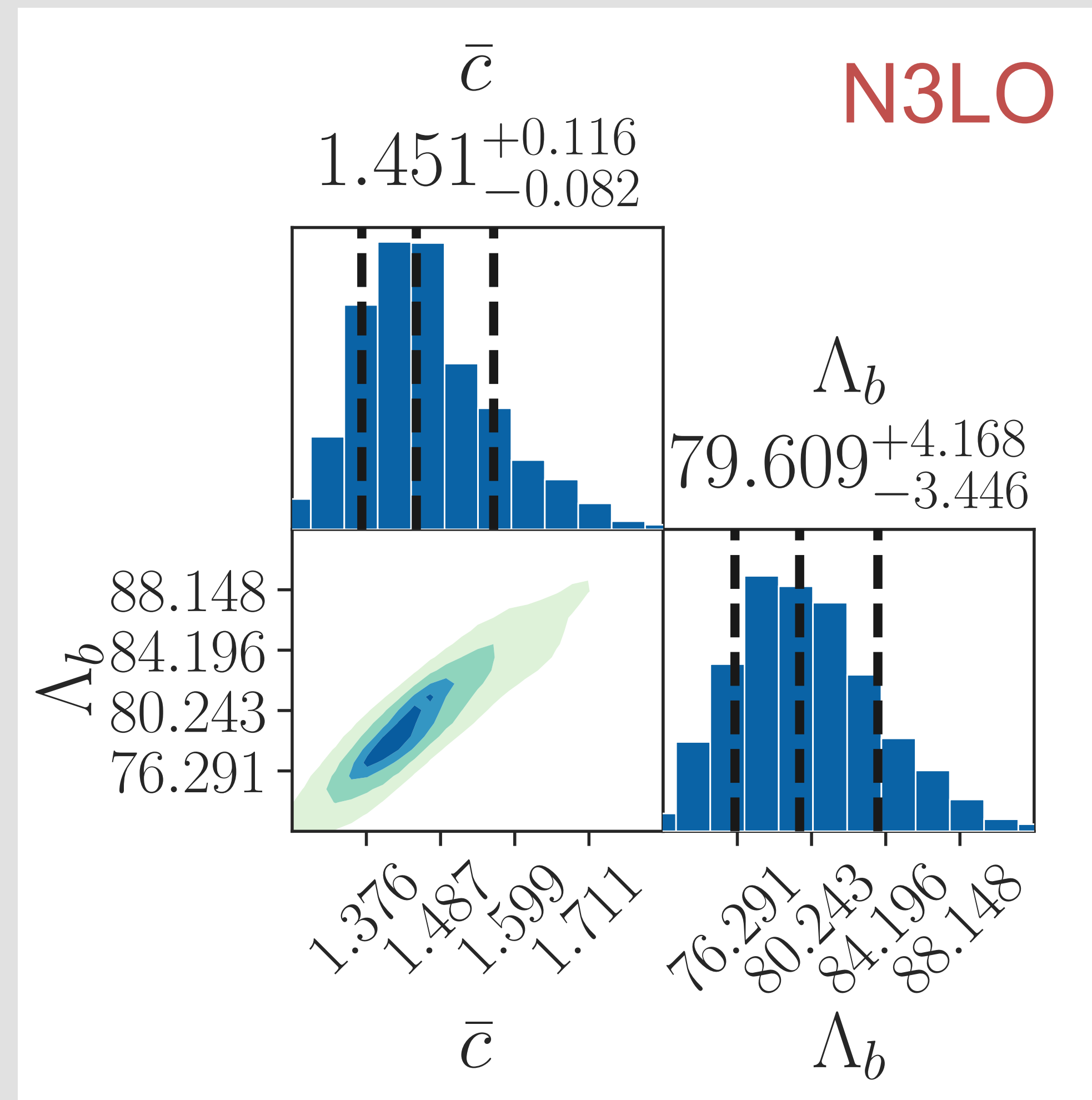
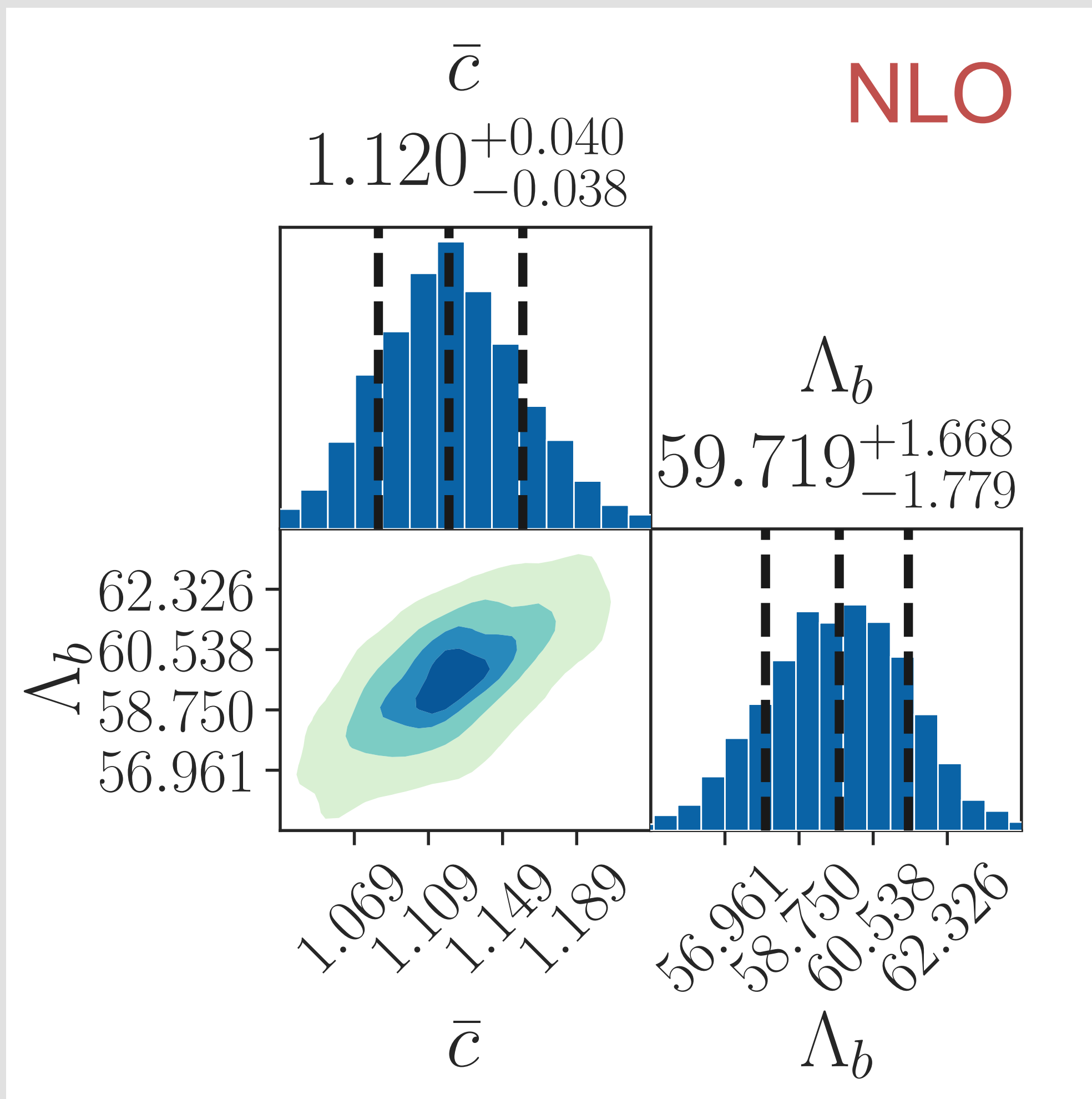
# 2.5 fm $\bar{c}$ and $\Lambda_b$ Posteriors



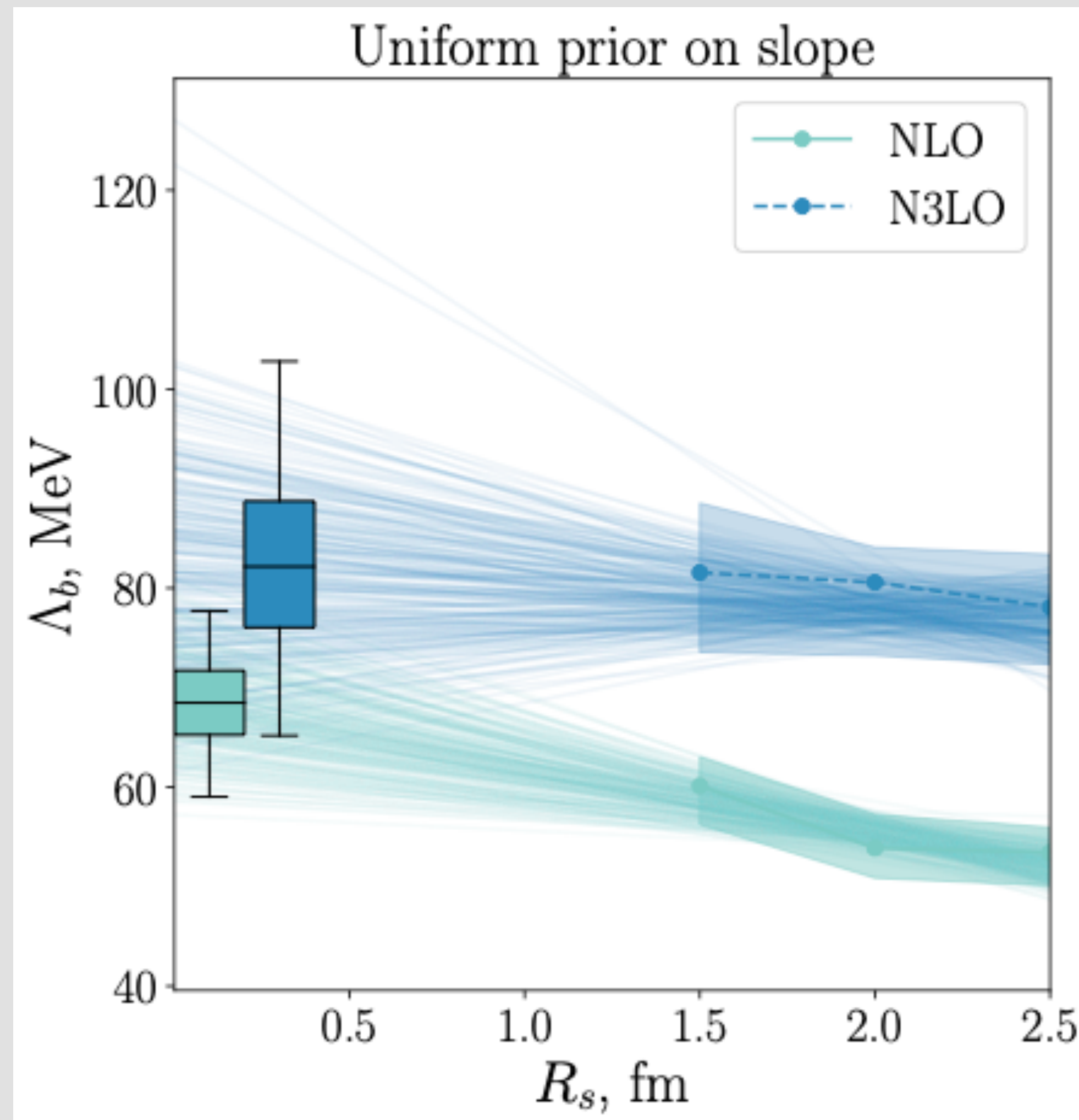
# 2.0 fm $\bar{c}$ and $\Lambda_b$ Posteriors



# 1.5 fm $\bar{c}$ and $\Lambda_b$ Posteriors



# Extrapolation to Remove Artifacts



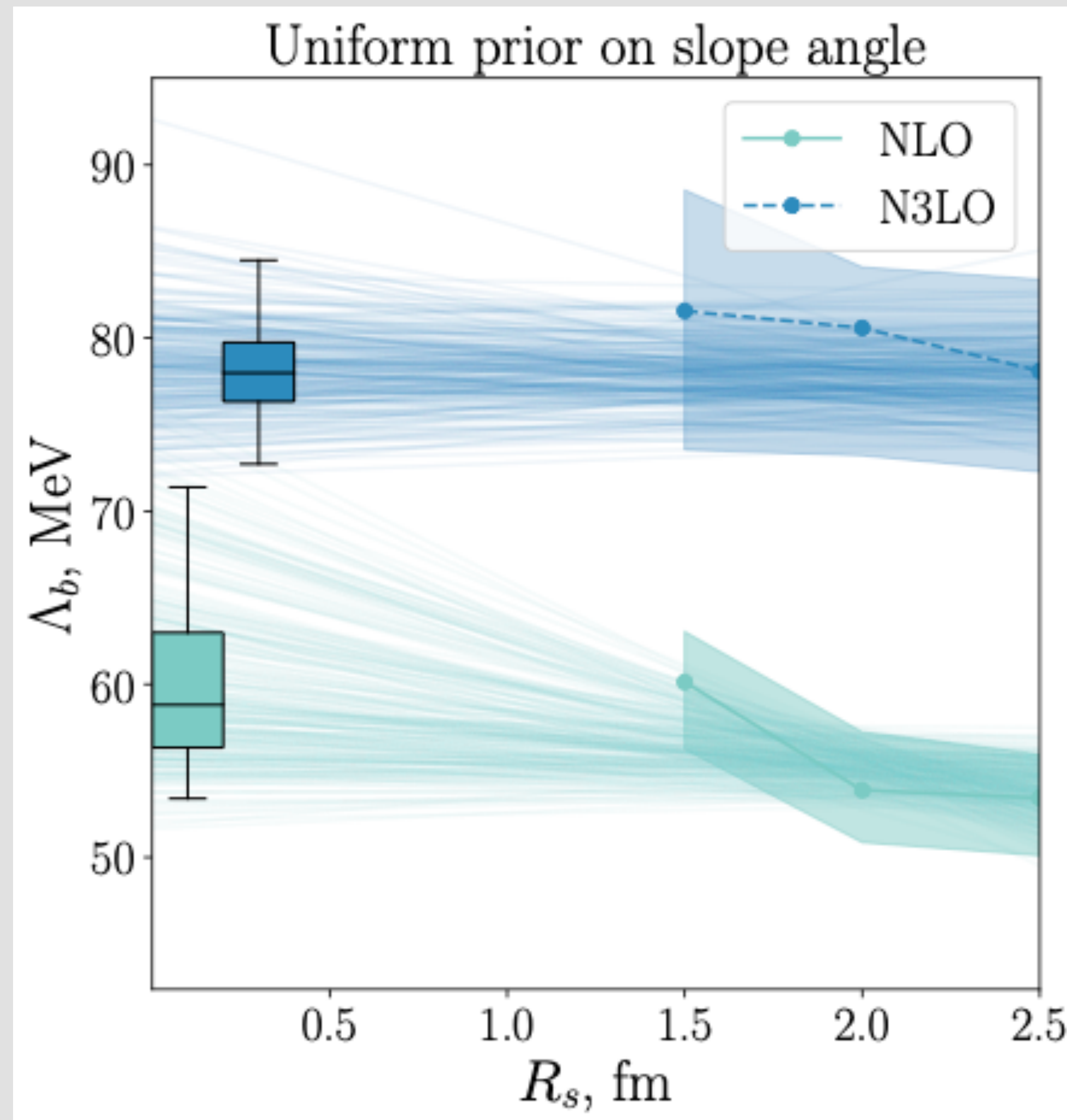
Why is there dependence on the order?

- Power counting?
- Flawed assumption of geometric series?
- Fierz transformation breaking?



# Extrapolation to Remove Artifacts

Proper choice of prior!!!



Why is there dependence on the order?

- Power counting?
- Flawed assumption of geometric series?
- ~~Fierz transformation breaking?~~



# Unconstrained $p$ -waves

With  $\Lambda_b \sim 50$  MeV, the max lab energy is given by

$$\frac{p_{\text{c.m.}}^{(max)}}{\Lambda_b} = \frac{\sqrt{E_{\text{lab}}^{(max)} \mu}}{\Lambda_b} = 1 \Rightarrow E_{\text{lab}}^{(max)} = \frac{\Lambda_b^2}{\mu} \sim 5 \text{ MeV}$$

The Granada database has 4 data (polarized cross sections) up to 5 MeV that constrains  $^1P_1$  and  $^3P_0$  channels.

We can explore removing  $p$ -wave interactions from our models.





# Posterior Predictive Density

We can now easily and **rigorously** propagate uncertainty to observable calculations.

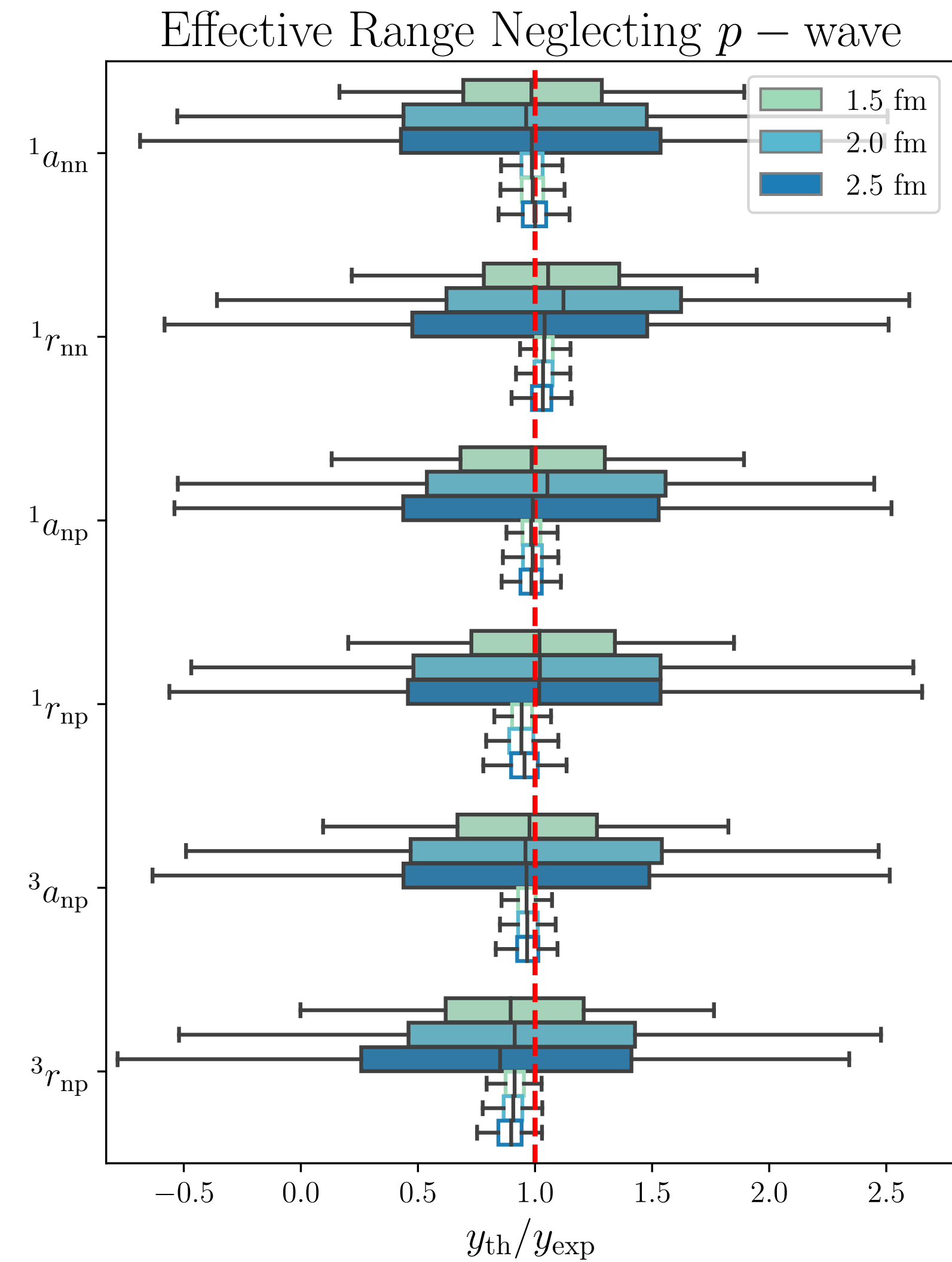
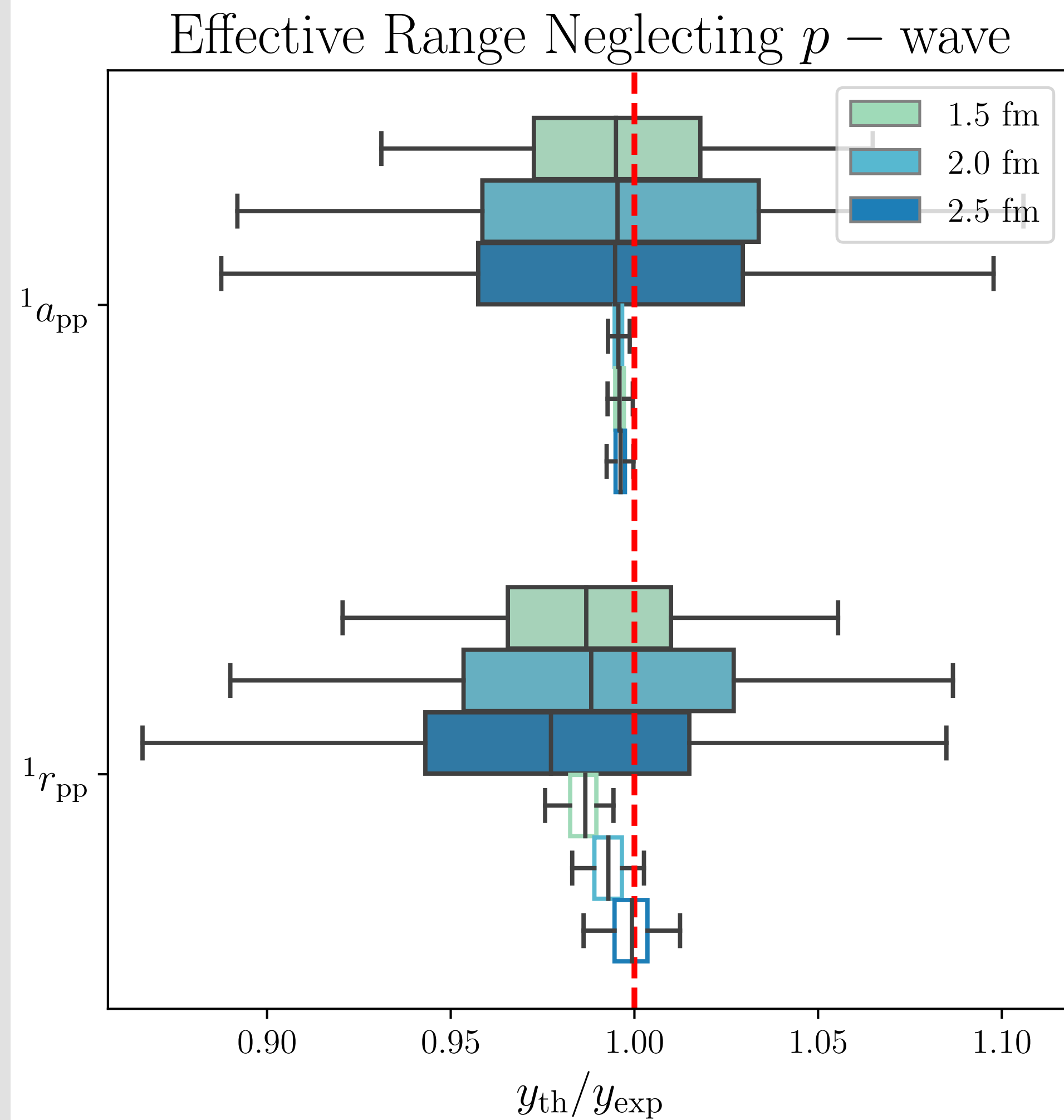
We calculate a **posterior predictive distribution (p.p.d.)** for the observables

$$\text{pr}(\vec{y}_{\text{th}} | \vec{y}, \vec{x}, I) = \int d\vec{a} d\bar{c}^2 d\Lambda_b \mathcal{N}(\vec{y}_{\text{th}}, \Sigma_{\text{th}}) \text{pr}(\vec{a}, \bar{c}^2, \Lambda_b | \vec{y}_{\text{exp}}, I)$$

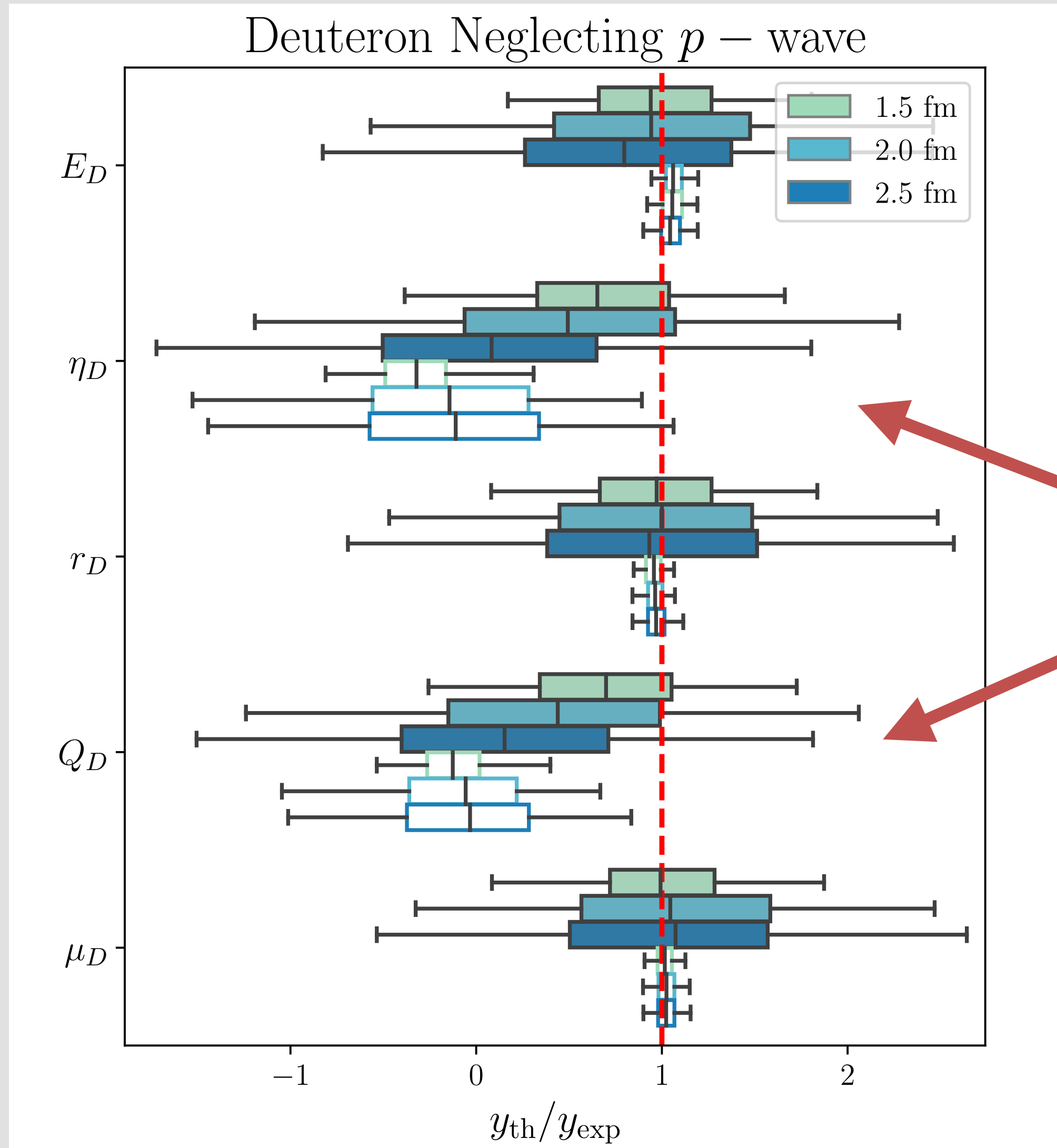
which is done via sampling the posterior.

This can be done for **any** calculation of nuclear observables.

# Propagation of Errors for ERPs



# Propagation of Errors for Deuteron



- Poorly constrained  $d$ -waves
- 2b corrections at  $O(Q^5)$

# Long-term Goals



- For pion- and  $\Delta$ -full interactions, we must look at higher energy data ( $\sim 200$  MeV)
- Emulation for calculation of scattering observables



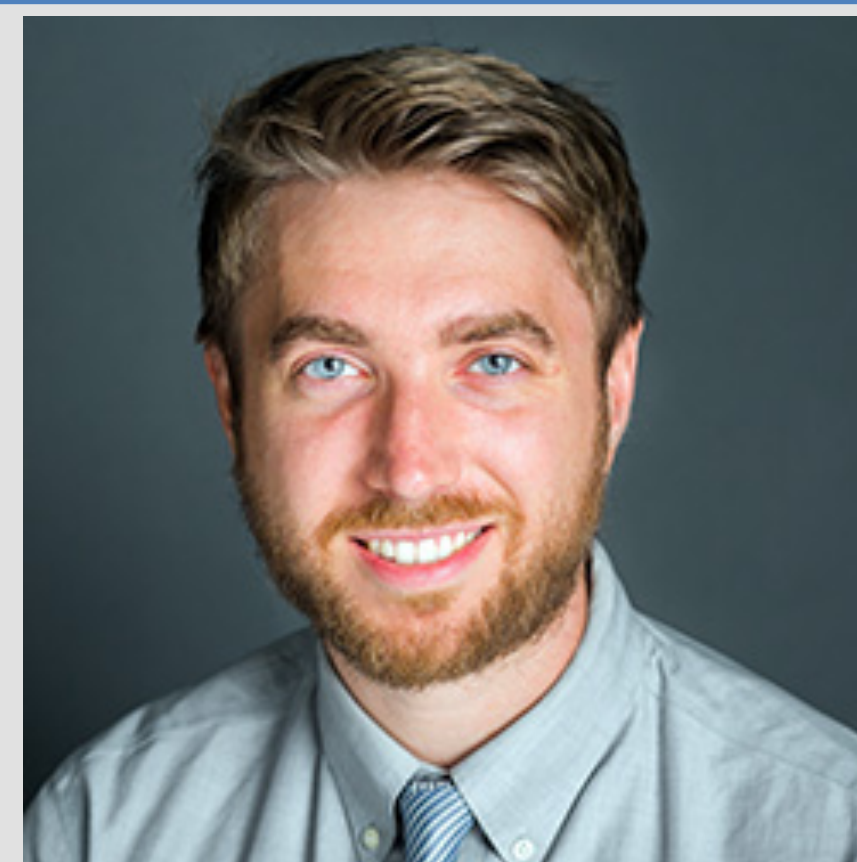
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Stefan Wild

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Matt Plumlee

Amazon



Pablo Giuliani

MSU/FRIB



Daniel Odell

SRNL

Gaussian Process Emulation

Reduced Basis Methods via  
Galerkin Projection

# Open Questions

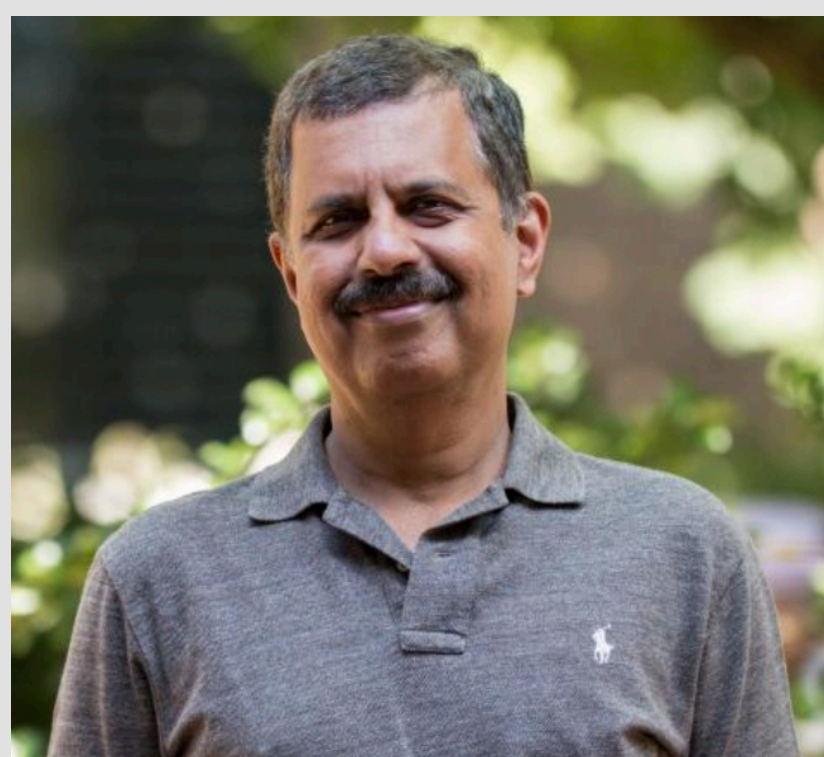


- Application to few- and many-body observables
  - How do we generate ppds using expensive many-body methods?
  - Estimation of the momentum to treat model discrepancy?
- Model mixing EFT model
  - Across degrees-of-freedom
  - Cutoffs
  - Regulators

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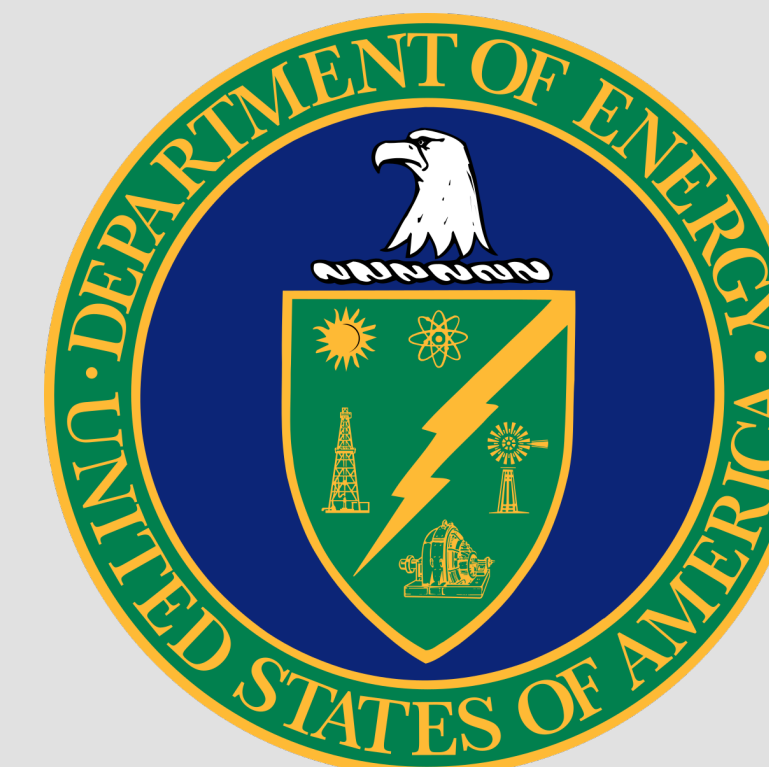


Sai Iyer  
WashU

## Computational Resources



## Funding



## Fellowship/Travel

