# Bayesian methods in nuclear effective field theories

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## The BUQEYE collaboration

**Bayesian uncertainty quantification: errors for your EFT** 



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#### Visit us online at [buqeye.github.io]

## **BUQEYE** priorities

Full uncertainty quantification for low-energy nuclear theory (structure/reactions)

Using statistics and machine learning techniques to gain physics insight

Collaboration and coordination with experts in statistics/machine learning

Pushing through computational bottlenecks (mostly from theory calculations)

- emulation/surrogates (gaussian processes, eigenvector continuation, etc.)
- simplifications that ease computational burden

Informing experimental design from the theory side



### Uncertainty Quantification in Nuclear Physics

To produce meaningful experimental measurements and theoretical predictions, it is essential to quantify uncertainties!

 $y_{\rm th} = y_{\rm exp} + \delta y_{\rm exp} + \delta y_{\rm th}$ 

Theory discrepancy:

Experimental discrepancy:

 $\delta y_{
m th}$ 

Made up of the following:

- missing physics
- numerical/ method errors
- fitting to uncertain data

#### Notes

- likely to be "systematic"
- not usually fully quantified
- often assumed to be normal

Made up of the following:

- counting statistics
- background and selection effects

 $\delta y_{\rm exp}$ 

• systematic uncertainties

#### Notes

- systematic errors may not be well understood or inflated
- often assumed to be normal

## Effective field theories are special

EFT convergence properties allow us to model theory discrepancies

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Also the possibility of regulator artifacts, alternate power-counting, etc.

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A healthy EFT improves systematically with order

Ignore other uncertainty sources other than EFT truncation uncertainty  $\delta y_k$ 

$$y = y_{\text{ref}} \sum_{n=0}^{k} c_n Q^n + y_{\text{ref}} \sum_{n=k+1}^{\infty} c_n Q^n$$
  
theory calculation

where  $Q \sim \frac{p}{\Lambda_b}$  and  $c_n$  s extracted from order-by-order calculations.

EFTs have unconstrained parameters, the low-energy constants (LECs)

### Strategy for nuclear interactions: constrain LECs using data

For an EFT at order k, let LECs be given by  $\vec{a}_k$ 

Interaction( $\vec{a}_k$ )  $\rightarrow$  Observable calculation  $\rightarrow$  Prediction( $\vec{a}_k$ )

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Interaction( $\vec{a}_k$ )  $\rightarrow$  Observable calculation  $\rightarrow$  Prediction( $\vec{a}_k$ )

Then compare the value with data



Adjust parameters  $\vec{a}_k$  according to favorite optimization procedure OR

Use a Bayesian approach and compute a parameter pdf for  $\vec{a}_k$  6 / 17

### nucleon-nucleon sector for chiral effective field theories

e.g.  ${}^{1}S_{0}$  channel at N ${}^{3}$ LO (4 parameters)





 $\operatorname{pr}(\vec{a}_k | D, I) \propto e^{-\chi^2/2}$ 

$$\chi^2(\vec{a}_k) = \sum_{i=1}^N \left( \frac{d_i - y_k(p_i; \vec{a}_k)}{\sigma_i} \right)^2$$

Bottleneck = theory calculation

This does not include  $\delta y_{
m th}$ 

### Including theory discrepancy in the parameter estimation procedure

SW et al., JPG 46, 045102 (2019)

Use model of EFT truncation error to include  $\delta y_{\text{th}}$ 

$$y = y_{\text{ref}} \sum_{n=0}^{k} c_n Q^n + y_{\text{ref}} \sum_{n=k+1}^{\infty} c_n Q^n$$
  
theory calculation

Assume a prior pdf on  $c_n$ s and marginalize to get

$$\operatorname{pr}(\vec{a}_k | D, I) \propto e^{-\chi^2(\vec{a}_k)/2} = e^{-\frac{1}{2}(\mathbf{y}_{\exp} - \mathbf{y}_{th}(\vec{a}_k))^T (\Sigma_{\exp} + \Sigma_{th})^{-1} (\mathbf{y}_{\exp} - \mathbf{y}_{th}(\vec{a}_k))}$$

Simple modified  $\chi^2$  form under certain assumptions

### Including theory discrepancy in the parameter estimation procedure

#### SW et al., JPG 46, 045102 (2019)

Accounting for truncation error absorbs higher-order effects



Purple: not accounting for  $\delta y_{
m th}$ 

Green: stabilizes when including  $\delta y_{
m th}$ 

### The three-nucleon sector (fitting $c_D$ and $c_E$ ) [preliminary]

SW, Andreas Ekstrom, et al. 2020 [in preparation]



Data:

- energy/radius of <sup>4</sup>He
- energy of  ${}^{3}$  H
- <sup>3</sup> H  $\beta$ -decay half-life

Fixed NN potential: NNLO<sub>sat</sub>

As before, including estimated  $\delta y_{
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Not including uncertainty from NN sector

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Made possible by eigenvector continuation Frame et al., PRL 121 032501 (2018)

## Status report 1: parameter estimation The three-nucleon sector (fitting $c_D$ and $c_E$ ) [preliminary]

SW, Andreas Ekstrom, et al. 2020 [in preparation]





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### Further improve model of truncation error: include correlations

Jordan Melendez, SW, et al. PRC 100 044001 (2019)

### Gaussian process model to study EFT truncation errors

For observables that vary over a continuous domain x

E.g., NN scattering observables in energy and angle  $x = \{E_{lab}, \theta\}$ 

$$y(x) = y_{\text{ref}}(x) \sum_{n=0}^{k} c_n(x)Q(x)^n + y_{\text{ref}}(x) \sum_{n=k+1}^{\infty} c(x)_n Q(x)^n$$

Still using lower-order convergence to inform missing corrections

E.g., truncation error at  $E_{\text{lab}} = 50 \text{ MeV}$  won't be too different from 51 MeV

Good evidence in NN sector that truncation errors are well-modeled by GPs



Differential cross-section of np scattering,  $E_{\text{lab}} = 150 \text{ MeV}$ 

From order-by-order predictions from LO to N<sup>4</sup>LO (5 orders)

GPs are random curves with some mean function and covariance kernel

J. Melendez, SW, et al. PRC 96 024003 (2017)

A correlated vs. pointwise truncation error model J. Melendez, SW, et al. PRC 100 044001 (2019)



Major feature: can use statistical validation tools to gain physics insight!

In preparation: Melendez, SW, et al. (2020)

Use GP model with modern NN potentials to diagnose convergence



This figure: toy problem

Estimate correlation length and EFT breakdown scale!

Currently applying diagnostics to modern NN potentials

Package gsum freely available to use: buqeye.github.io/software

## More ongoing BUQEYE projects

- Convergence analysis of infinite matter calculations Drischler, Melendez, Furnstahl, Phillips
- Bayesian experimental design for proton Compton scattering experiments Melendez, Furnstahl, Griesshammer, McGovern, Phillips
- Ongoing questions about Bayesian model selection in EFTs

### See everything at buqeye.github.io

## Summary and status

- Bayesian statistics is the ideal tool for effective field theories
- Build and include physics assumptions explicitly through Bayesian priors
- A data-driven study of theory expectations

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### Gain physics insight from statistical tools

- Continuing need for expertise in statistics and machine learning
- Model selection for EFTs? Power-counting comparison and diagnosis?
  - major bottleneck is observable computation
- Fully quantify uncertainties on EFT predictions
  - understand full correlation/interplay of all errors
  - easily usable approaches for nuclear structure practitioners