

Quantifying uncertainties in nuclear reactions

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Michigan State University occupies the ancestral, traditional, and contemporary Lands of the Anishinaabeg–Three Fires Confederacy of Ojibwe, Odawa, and Potawatomi peoples. The University resides on Land ceded in the 1819 Treaty of Saginaw.

Outline

- \Diamond The role of optical potentials in reactions
- \Diamond Bayesian analyses of optical potentials
- \Leftrightarrow Propagation to other observables
	- \Diamond Transfer
	- \Diamond Charge-exchange
	- Knockout
- \Diamond Emulators
	- \Leftrightarrow Application to model for breakup
- \Diamond Opportunities for the future

The Optical Potential is an essential ingredient in reaction theory V NN COS^S Language U_{opt}

It's the projection of the many-body scattering problem on the ground state: $P\Psi(\vec{r}, \vec{r}_1, \ldots, \vec{r}_A) = \phi_0(\vec{r}) \Phi_0(\vec{r}_1, \ldots, \vec{r}_A)$

End up with a single-channel scattering equation with potential:

$$
V_{\rm opt} = \mathcal{V}_{00} + \sum_{j,k\neq 0} \mathcal{V}_{0j} \frac{1}{E - H_{jk} + i\eta} \mathcal{V}_{k0}
$$

 U_{opt} = V(R) + iW (R) can be obtained phenomenologically!

Optical potentials are pervasive in reaction models

Inputs necessary for (n,g) ; (p,g) ; (p,n) ; (n,p) ; (d,p) ; (d,n) ; ... Inputs also for breakup, knockout and transfer on heavier probes

Reaction observables are very sensitive to details of the optical potential.

OP is the main source of uncertainty

Need uncertainty quantification!

OP white paper shows current state of the art

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Bayesian statistics [Thomas Bayes \(1701–1761\)](https://en.wikipedia.org/wiki/Thomas_Bayes)

Bayesian analysis

- Calibration of the model M (parameter posterior distributions)
- Estimation of uncertainty in predictions (credible intervals on observables)

- Model assessment (comparison between models and with data)
- Model mixing (admixture between models with different strengths)
- Experimental design (What is the optimum measurement that adds information?)

The model has a set of parameters

 $U_{opt}(R) = V f(R, r, a) + W f(R, r_w, a_w) + W_s f(R, r_s, a_s) + V_{so} + V_c$

We use previous OP parameterizations to set the priors (typically wide priors to allow process to be data driven)

- use MCMC to sample parameter space

Statistical model

Data: elastic scattering angular distributions/polarizations/total xs

- real exp data with evaluated errors
- mock data calculated using KD with 10% errors

Likelihood:

- No correlations and errors normally distributed

$$
p(D|H, M) = \exp[-\chi^2/2]
$$

- Include correlations effectively by dividing by the number of data points N (equivalent to inflating errors)

$$
p(D|H, M) = \exp[-\chi^2/(2N)]
$$

$$
\chi^2 = \sum_{i=1}^{N} \frac{[\sigma_{\exp}(\theta_i) - \sigma_{\text{th}}(\theta_i, x)]^2}{[\Delta \sigma_{\exp}(\theta_i)]^2}.
$$

Bayesian: parameter posterior distributions

Create 95% confidence intervals for observable

Manuel Catacora-Rios

What angular information needed?

⁴⁸Ca(n,n) ⁴⁸Ca at 12 MeV

 $^{48}Ca(p,p)^{48}Ca$ at 21 MeV

Catacora-Rios, King, Lovell and Nunes, PRC (2019)

Single energy versus multiple energy sets? Polarization versus differential cross sections?

95%

credible

intervals

Lovell, Nunes, Catacora-Rios, King, JPG (2020) Catacora-Rios et al. PRC 100, 064615 (2019) King, Lovell, Neufcourt, Nunes PRL (2019) Catacora-Rios et al. PRC 104, 064611 (2021)

What prior to use?

Priors encapsulate our prior knowledge (e.g. a previous global parameterization)

Use gaussian distributions on parameters How wide should these be?

Lovell and Nunes PRC (2018)

FIG. 2: (Color online) Comparison of the posterior distributions (histograms) resulting from various prior distributions (corresponding solid lines) for a wide Gaussian (WG), medium Gaussian (MG), and narrow Gaussian (NG) as defined in Table II for $\frac{90}{2r(n,n)}$ $\frac{90}{2r}$ at 24.0 MeV.

Which likelihood to use?

Complications:

data correlations systematic errors on data underestimated

model correlations

model uncertainties

How to combine sets of angular distributions?

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Propagating uncertainties to transfer

OP constrained with elastic scattering to obtain posterior distributions for parameters

Ekstrom talk: UQ important for decision-making and model assessment

Lovell, Nunes, Catacora-Rios, King, JPG (2020)

Uncertainty quantified **global** optical potential (CHUQ and KDUQ)

Bayesian analysis using the same experimental protocol as in the original CH89 and KD2003 parameterizations

Pruitt et al., Phys. Rev. C 107, 014602 (2023)

Andy Smith

OP uncertainties in charge exchange to IAS

- DWBA formalism
- Using parameter posterior from KDUQ

Dark shade (68% ci) Light shade (95% ci)

Comparing two-body and three-body models for charge exchange

Smith, Hebborn, Nunes, Zegers, PRC **110**, 034602 (2024)

Uncertainty quantified **global** optical potential (East Lansing Model)

ELM uses a much smaller set of data compared to KDUQ

Includes charge-exchange to IAS for key isotopes

Beyer and Nunes, in preparation

Propagating uncertainties to knockout

- Eikonal model
- Using parameter posterior from KDUQ

compare with a consistent ADWA study of transfer 34,26,46Ar(p,d)

dark (light) shade: 68% (95%) credible intervals

Hebborn, Nunes, Lovell, PRL 131, 212503(2023)

Comparing knockout and transfer: linear fit

 $\mathcal{R}(\Delta S) = a\Delta S + b.$

Hebborn, Nunes, Lovell, PRL 131, 212503(2023)

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Emulators for nuclear reactions

An emulator is a fast and efficient replacement for a complex physics model

Extension to coupled channels in development!

Odel et al., Phys. Rev. C 109, 044612 (2024)

Data driven emulator Breakup cross sections needed for astrophysics

Example: 7 Be(p, γ)⁸B reaction relevant for solar fusion

Working horse for modeling these reactions: Continuum Discretized Coupled Channel (CDCC) Large scale (large memory requirements) Long runs (many hours to days)

Impossible to do Bayesian analysis directly with CDCC!

Predictions: Angular distributions and energy distributions of fragments

Emulators for breakup cross sections

 $^7\mathsf{Be}(\mathsf{p},\gamma)^8\mathsf{B}$ reaction relevant for solar fusion

Continuum Discretized Coupled Channel Gaussian-processors emulator for breakup: Angular distribution and energy distribution

uncertainty from ⁷Be+p interaction

mock data generated for set of interactions from G. Goldstein et al., Phys. Rev. C 76, 024608 (2007)

TABLE I: Model parameters and their ranges.

Surer, Nunes, Plumlee, Wild, PRC106, 024607(2022)

Emulators for breakup cross sections

Posterior distributions and correlation plots

Emulators for breakup cross sections

Continuum Discretized Coupled Channel Gaussian-processors emulator for breakup: Angular distribution and energy distribution

uncertainty from ⁷Be+p interaction

Surer, Nunes, Plumlee, Wild, PRC106, 024607(2022)

Opportunities for the future

- Optical potential validated for rare isotopes:
	- nucleon global optical potential with UQ informed by (p,n); ab-initio priors; extension to heavy-ions…
- Bayesian analysis for complex reactions models:

fast and accurate emulators

Opportunities for the future

most important of all are the people!

Filomena Nunes

Chloë Hebborn

Kyle Beyer

Patrick McGlynn

Cate Beckman

Andy Smith Manuel Catacora Rios

Daniel Shiu

Pablo Giuliani

Grigor Sargsyan

few-body reaction group@MSU, summer 2024

Collaborators:

Bayesian Analysis: Amy Lovell (LANL) Chloe Hebborn (MSU) Garrett King (WashU) Manuel Catacora-Rios (MSU) Cole Pruitt (LLNL)

Charge Exchange: Terri Poxon-Pearson (NNSA) Gregory Potel (LLNL) Andy Smith (MSU) Chloe Hebborn Remco Zegers

Knockout: Chloe Hebborn Amy Lovell

Emulators: BAND collaboration

BACKUP

Bird's eye view of nuclear reactions

Nuclear reactions got us from the lightest elements all the way to the wide range of elements found in our solar system!

Bird's eye view of nuclear reactions

Probe of neutron capture: breakup and transfer

Probe of single-particle structure: knockout

Probe of electron capture: charge-exchange

Reactions are the most diverse probes to extract astrophysics and structure information, especially for unstable isotopes…

But reaction theory is key for translation!

Phenomenological potentials fitted to stable nuclei

Reaction theory maps the many-body into a few-body problem

❑ isolating the important degrees of freedom in a reaction ❑ effective nucleon-nucleus interactions (or nucleus-nucleus) usually referred to as **optical potentials**

> main cause of uncertainty

Landscape of global optical potentials

Optical potentials from theory

Microscopic optical potential:

- Non-local, typically not global, no simple general form
- depends on the EFT: cutoffs, regularizations, etc.
- agreement with data is variable...

Landscape of microscopic optical potentials

How do optical models compare?

How do optical models compare?

credible Total cross section as a function of energy intervals **SCGF NSM**

95%

How do optical models compare?

Asymmetry of total cross section

What model encapsulates more information?

Bayesian evidence: provides information contained in a data set. Integral of the likelihood times the prior over full parameters space

$$
p(d|\mathcal{M}) = \int_{\Omega_{\mathcal{M}}} p(d|\alpha, \mathcal{M}) p(\alpha|\mathcal{M}) d\alpha_{\mathcal{M}}
$$

$$
\text{Bayesian factor:} \quad \frac{\bar{p}(d|\mathcal{M})_{(d\sigma/d\Omega)}}{\bar{p}(d|\mathcal{M})_{(iT_{11})}} \quad < 3
$$

Kass and Raftery, J. Amer. Stat. Assoc 9 (430) 791

TABLE I. Bayesian evidence (multiplied by 10^{-3}) for the surface model (second row) and the volume model (third row) for both beam energies considered (first row). The ratio between the Bayesian evidence of the volume model over that with the surface model is in the fourth row (the Bayes' factor).

Catacora-Rios et al. PRC 104, 064611 (2021)