

The Bayesian Analysis of Nuclear Dynamics CI Framework



<https://bandframework.github.io/>

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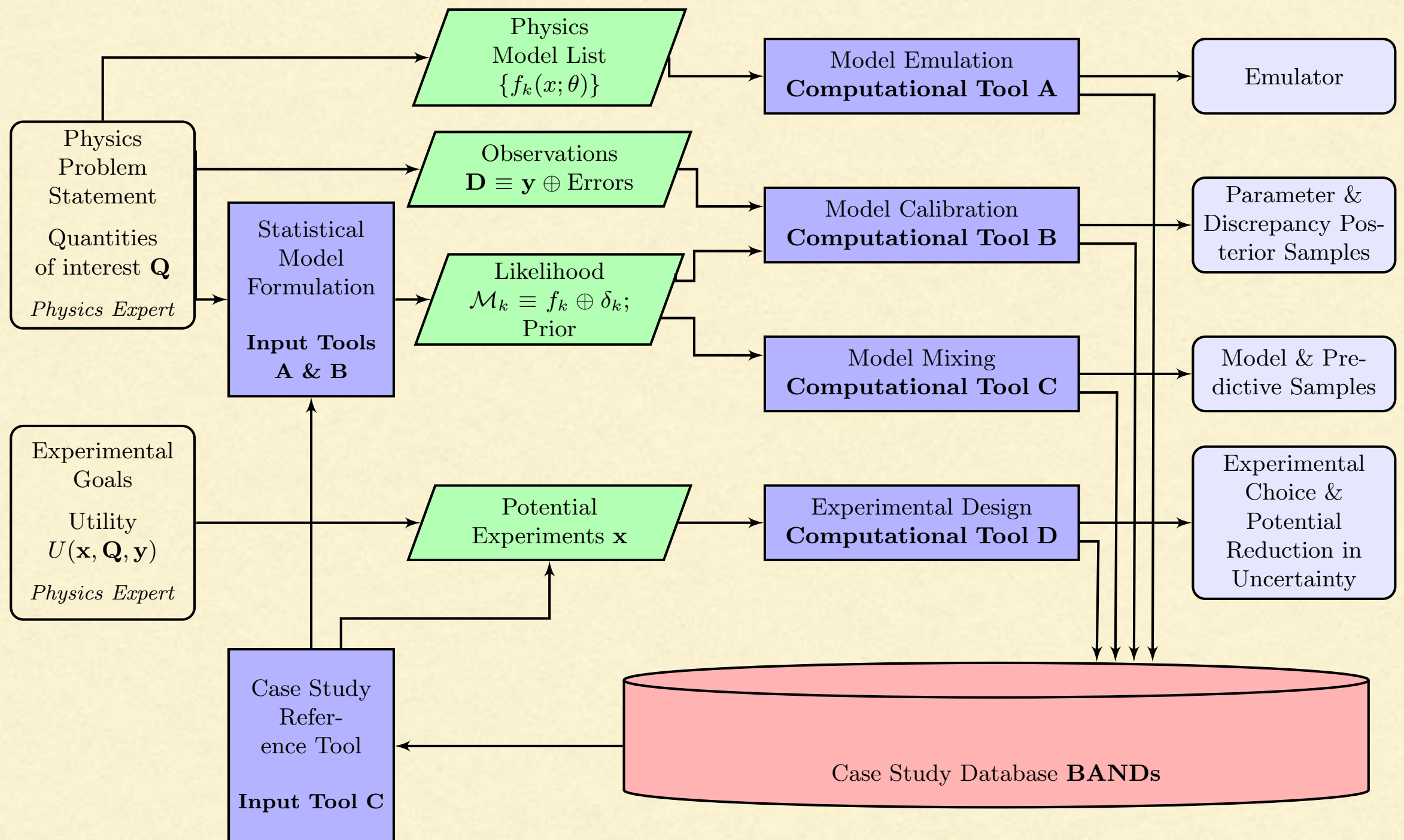
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Bayesian Analysis of Nuclear Dynamics

- Much progress on Uncertainty Quantification in Nuclear Physics in last few years
 - But still some inhibitions regarding use of Bayesian methods:
 - What prior should I choose?
 - Isn't MC sampling too computationally expensive a way to estimate the parameters I care about?
 - How do I use Bayesian methods to assess model uncertainty?
 - BAND will provide solutions!
 - Guidance from experienced & expert statisticians and physicists on prior selection
 - Fast emulators for expensive physics models
 - Use “Bayesian Model Mixing” to provide error bars that reflect full error bar for a nuclear-physics prediction, based on best available Nuclear Physics knowledge
 - Consistently calibrated and mixed nuclear-physics models can then be used for optimal design of experiments
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The Framework



The Team

Senior Investigators

- Ohio U., Daniel Phillips: Nuclear Physics (PI)
- Michigan State U., Taps Maiti, Frederi Viens: Statistics;
Witek Nazarewicz, Filomena Nunes, Scott Pratt: Physics
- Northwestern U., Matthew Plumlee: Engineering/Statistics,
Stefan Wild: Applied Math/Computer Science
- Ohio State U., Dick Furnstahl, Uli Heinz: Physics,
Matthew Pratola, Statistics



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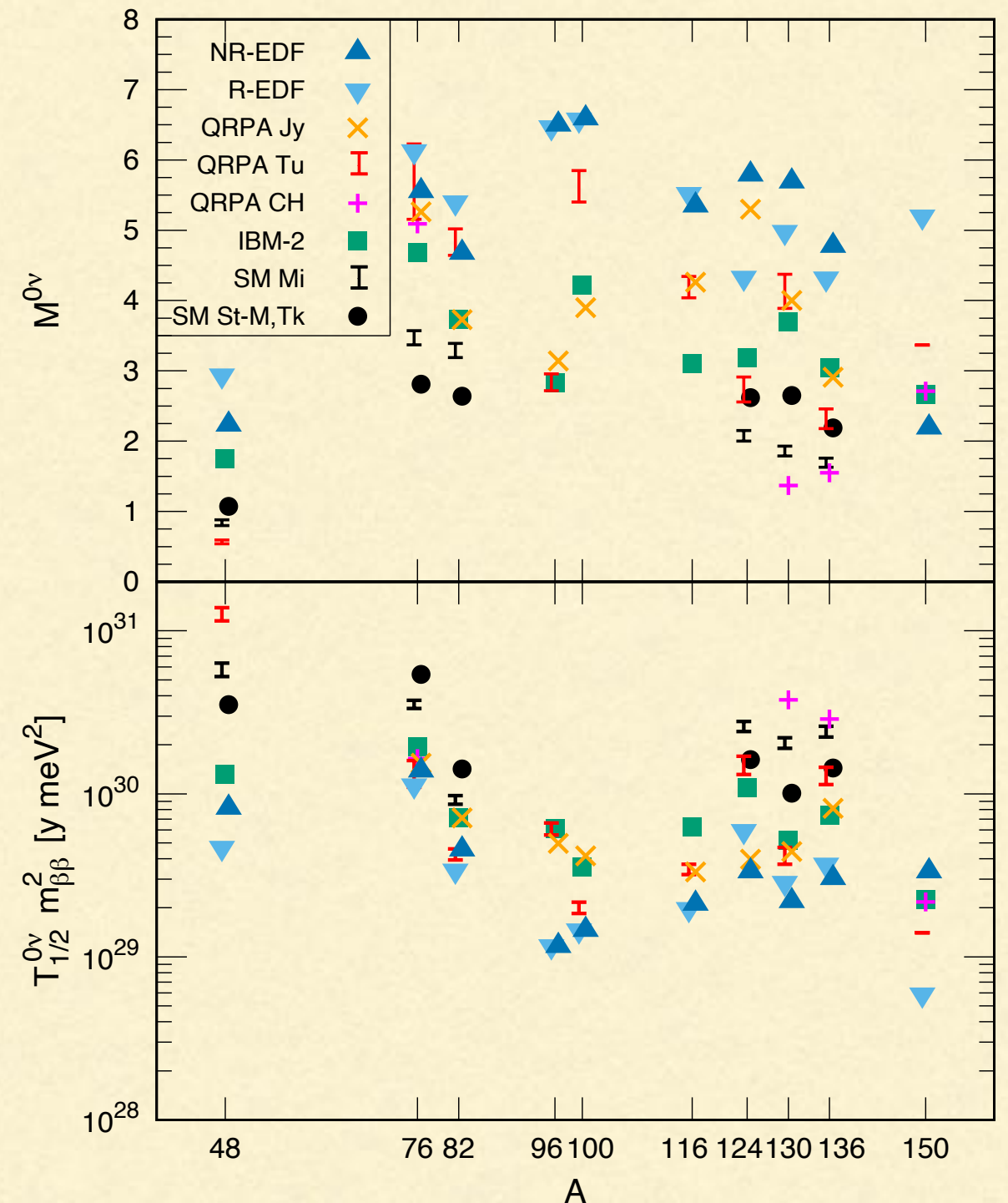
Nuclear Science motivation

- r-process: extrapolation to the dripline and beyond; ties in to other nuclear-structure issues
- Heavy-ion collisions: energy deposition; pre-hydrodynamic stage; conversion of hydrodynamic output to final-state particles
- Mixing different approaches to reaction dynamics → nuclear data evaluation with fully quantified uncertainties
- Neutrinoless double beta decay

Ultimate goal is to build framework that is *generally useful*

Some specific $0\nu\beta\beta$ context

- Wide range of predictions for $M^{0\nu}$ in experimentally relevant nuclei
- Should we just take an average?
- Or weight them somehow according to performance on other relevant observables?
- What if some approaches are better for small A and others for large A ?
- Bayesian Model Mixing provides a way forward

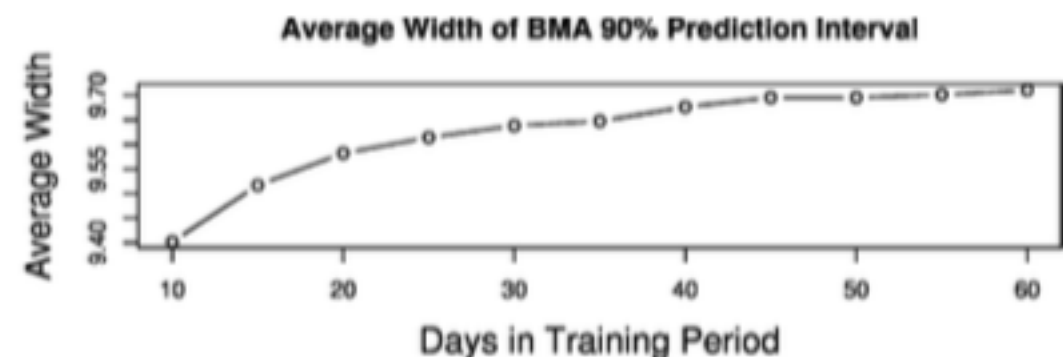
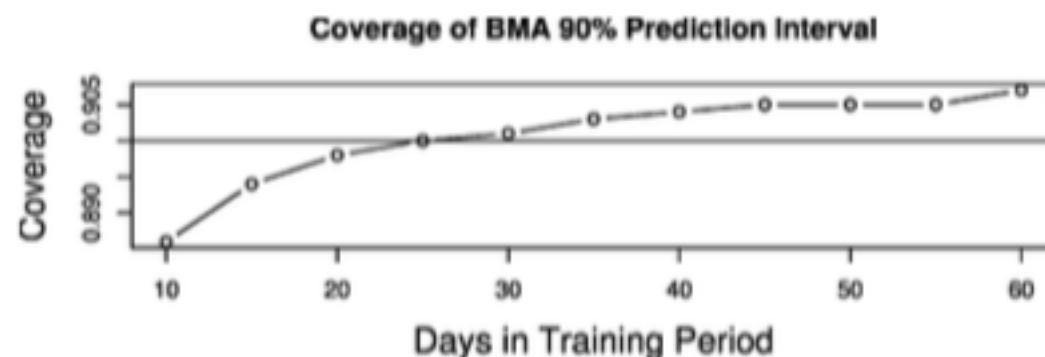
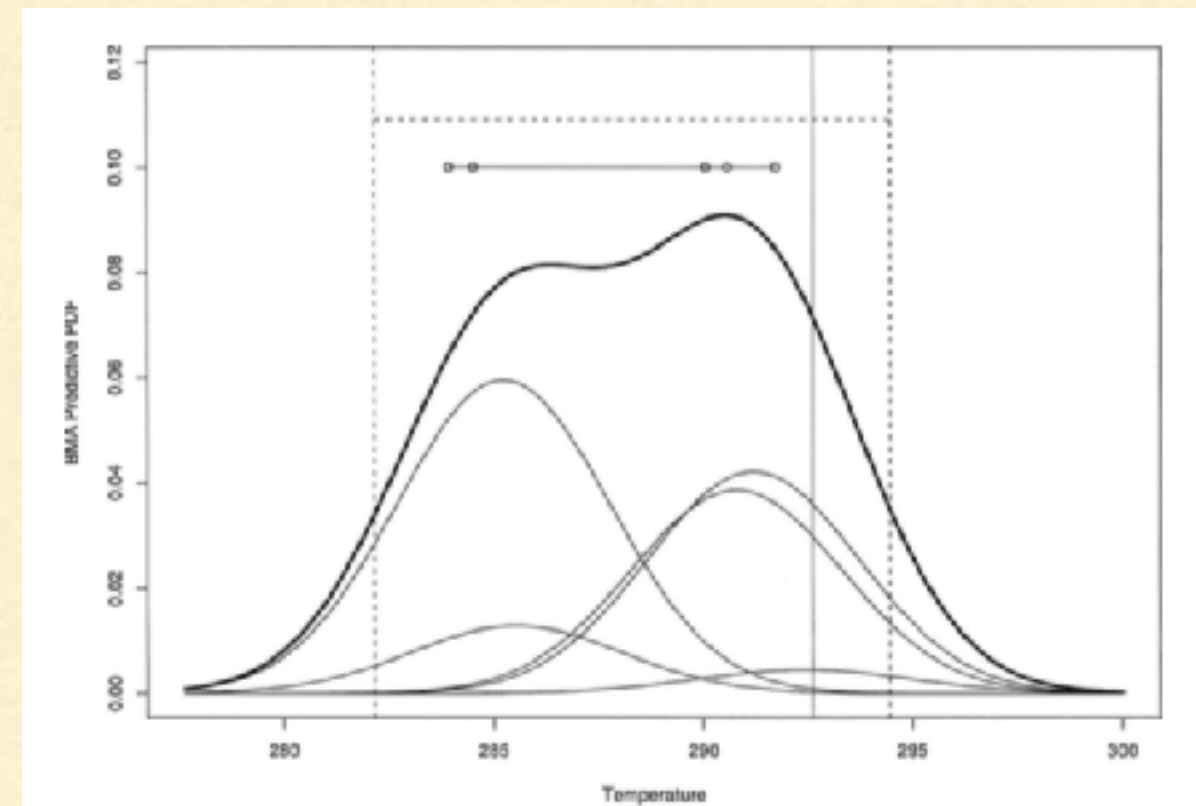


Bayesian Model Mixing

Simplest version is “Bayesian Model Averaging” (BMA)

$$\text{pr}(Q | D, I) = \sum_M \text{pr}(Q | M, D, I) \text{pr}(M | D, I)$$

- Used in several other fields
- Improves predictive performance in weather forecasting Raftery et al. (2005)
- Application to EDFs, proton-emission, etc. Neufcourt et al. (2019-2022)
- Applied to EFT expansion in toy context Connell, Billig, Phillips (2021)



Timeline



<https://bandframework.github.io/>

- Year 1: Release of BAND Manifesto; Nuclear-physics codes in repo
 - Year 2: Version 1 demo released
 - Year 3: Version 2 framework released
 - Years 4 & 5: Mature version of BAND Framework released with database; POC demos for experimental planning and forefront nuclear theory; workshop for other disciplines
 - Throughout: Roundtables with community, BAND camps, tutorials
 - Collaboration & input welcomed
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