Stress testing the Oslo Method for heavy nuclei Fabio Zeiser, University of Oslo & collaborators

Current Work

Motivation: The Oslo Method has been widely used to find the basic research: Models nuclear level density (NLD) and γ -ray strength function (γ SF) of the atomic nucleus below the neutron separation threshold S_n.

What happens if some of its assumptions are violated?

Specific problem: There is a mismatch between the intrinsic spinparity distribution of the levels in the nucleus, and the spinsparities populated by the (d,p) reaction. This is expected to be most significant for heavy nuclei.

applications (nuclear & astrophysics)

Optional: $\rho_{\rm red}$ Correction



Approach: Produce synthetic data of the nuclear decay of following the 239 Pu(d,p) 240 Pu reaction, where input NLD, γ SF and spin-parity distributions are known. Then we can infer systematic deviations by applying the Oslo Method to this data.





Fig: Producing synthetic data to infer systematic errors in the Oslo Method

Results

- No systematic deviations when assumptions are fulfilled, i.e. populating levels proportionally with intrinsic spin-parity
- For the calculated population from the ²³⁹Pu(d,p)²⁴⁰Pu reaction, we infer significant distortions in the extracted nuclear level density and γ-ray strength
- In progress: Using the presented approach to correct for the

Fig: NLD and γSF extracted with the Oslo Method, (with an optional correction r) from the synthetic dataset. The populated spin distribution g_{pop} was either chosen equal to the intrinsic distribution g_{int} or according to the calculations for ²³⁹Pu(d,p)²⁴⁰Pu

deviations

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Inverse Problems

need to parametrize?

Another, more systematic approach:

Can we find the *true* NLD and ySF (parameters) using a nuclear decay code and Markov Chain Monte Carlo methods? The goodness of fit would evaluated comparing the calculated decay data to the experimental data

ottow about degeneracies (zmultiple modes); y=a.b 2



Another project: Multidimensional unfolding via Machine Learning?

Challenge: Find the true multiplicity distribution (and energy spectrum) from the measured multiplicities and energies

Approaches:

i) Matrix inversion

ii) *Bayesian* multidimensional unfolding: Want about 100 energy bins, and multiplicities up to at least 4, but the response matrix for 25 bins is already 500 GB large.

iii) Can we use Machine Learning & image recognition?







Fig: Example of the detector response for multiplicity M = 2