# Bayesian model calibration for nuclear decays with the Skyrme finite-amplitude method

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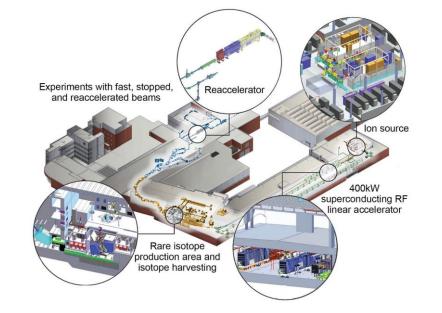


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## Alchemy on the Earth and in the universe





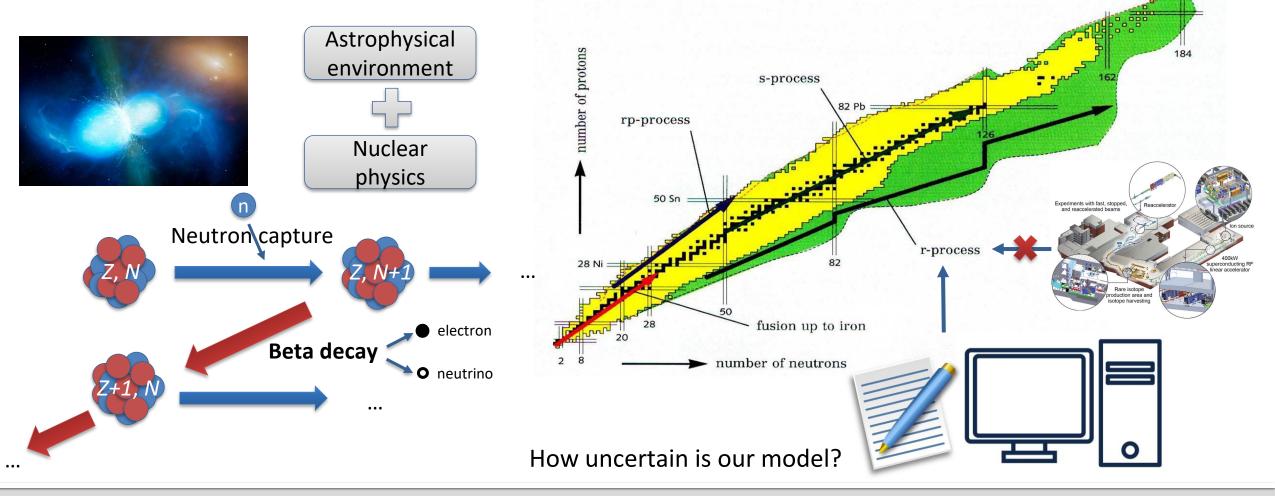


- Medieval alchemists failed to transform one chemical element to another.
- We can now do "modern alchemy" in facilities for nuclear physics!
- Large-scale producers of heavy elements exist in the universe, such as the neutron-star merger.



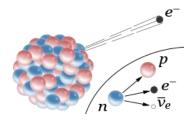
# Rapid neutron-capture process and nuclear inputs for simulation

• The rapid neutron-capture process (*r* process) plays a key role in the synthesis of heavy nuclei.

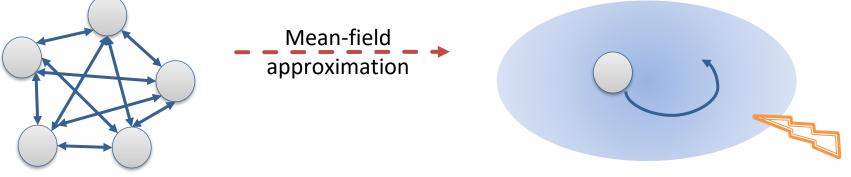




# Finite amplitude method enables large-scale studies



 New algorithms in the framework of nuclear density functional theory make it feasible to perform selfconsistent microscopic calculations of beta decays through the nuclear landscape.



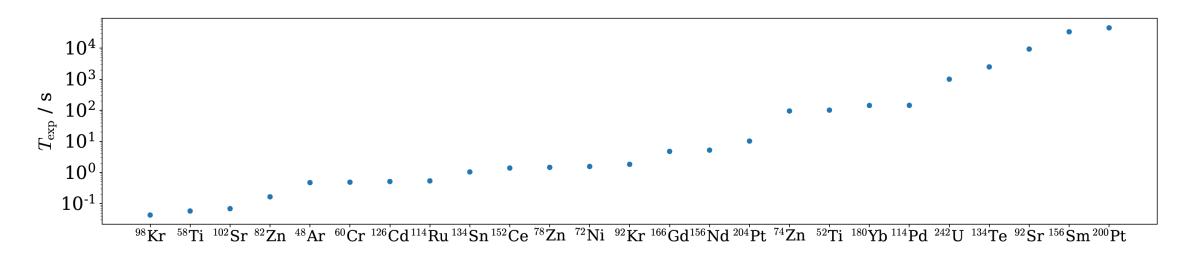
- Each nucleon moves in the **mean field** generated by other nucleons. The mean field is specified by the Skyrme energy density functional (EDF).
- The ground state is described by the Hartree-Fock-Bogoliubov (HFB) approach.
- The beta-decay transition is calculated through the finite-amplitude method (FAM), where an external charge-changing field is applied as perturbation to trigger the transition.
- Some EDF parameters need to be constrained by experimental data on the beta decay.



#### Parameters to calibrate and fit targets

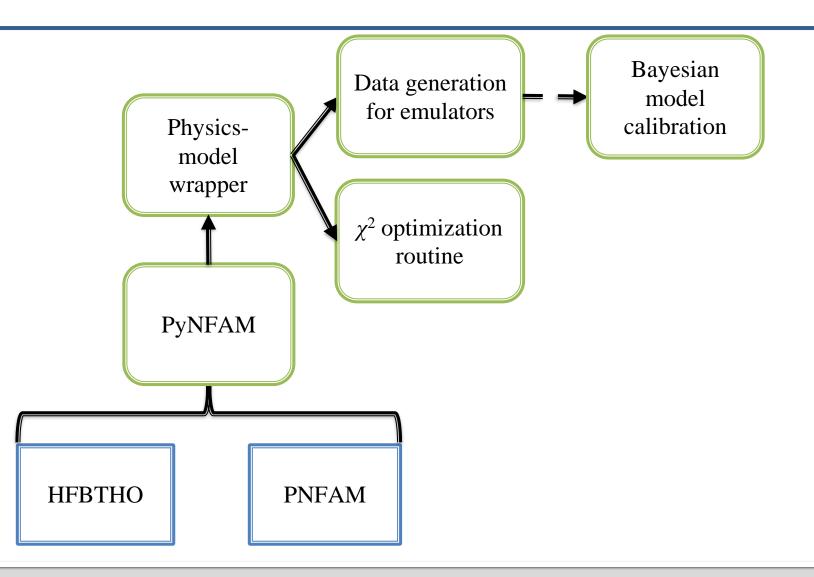
- Skyrme parameterization UNDEF1-HFB is employed.
- Parameters to calibrate include:
  - Landau-Migdal parameter  $g_0'$ ,
  - Normalized isoscalar pairing strength  $v_0$ ,
  - Axial-vector coupling  $g_A$  (quenching effect).

- Fit targets include two types of data:
  - Gamow-Teller-resonance energies of 4 selected doubly and semi-magic systems,
  - $-\beta^{-}$ -decay half lives of 25 even-even nuclei.
- The data selection is based on <u>Phys. Rev. C 93</u>, <u>014304 (2016).</u>





#### **Numerical Framework**



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# $\chi^2$ optimization is the first step for model calibration

• We minimize the weighted sum of squared errors (residuals) with POUNDERS:

$$\chi^{2}(\mathbf{x}) = \frac{1}{n_{d} - n_{x}} \sum_{i=1}^{n_{d}} \left( \frac{s_{i}(\mathbf{x}) - d_{i}}{w_{i}} \right)^{2}.$$

• Parameter values obtained from the  $\chi^2$  optimization are shown in the table below. These parameters are weakly correlated.

Fit	${oldsymbol{g}}_{oldsymbol{0}}^{\prime}$	$oldsymbol{g'_1}$	$v_0$	$g_{\scriptscriptstyle A}$
А	1.59560 (0.039)	0 (fixed)	-0.99993 (0.178)	1 (fixed)
В	1.59184 (0.034)	0 (fixed)	-1.19745 (0.179)	0.50345 (0.143)

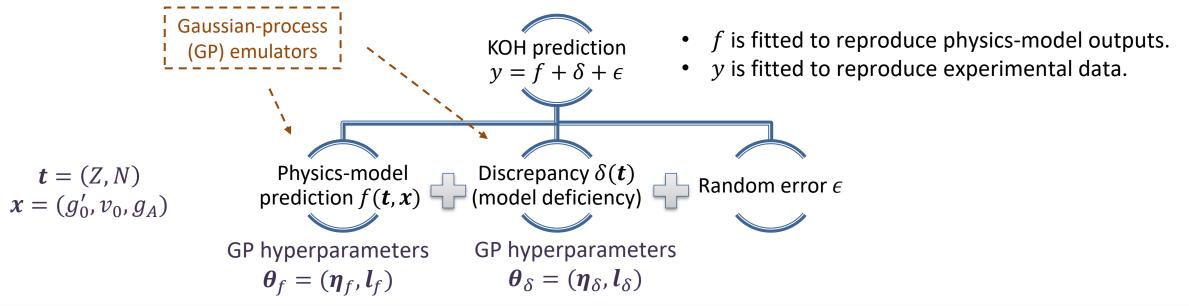


# **Bayesian model calibration with model deficiency**

- Compared with the  $\chi^2$  optimization, Bayesian inference provides a reliable approach to obtain the distributions of parameter values, which is useful for uncertainty quantification and propagation.
- The foundation of the Bayesian model calibration is Bayes' theorem:

 $P(\text{ params } | \text{ obs }) \propto P(\text{ obs } | \text{ params })P(\text{ params }).$ 

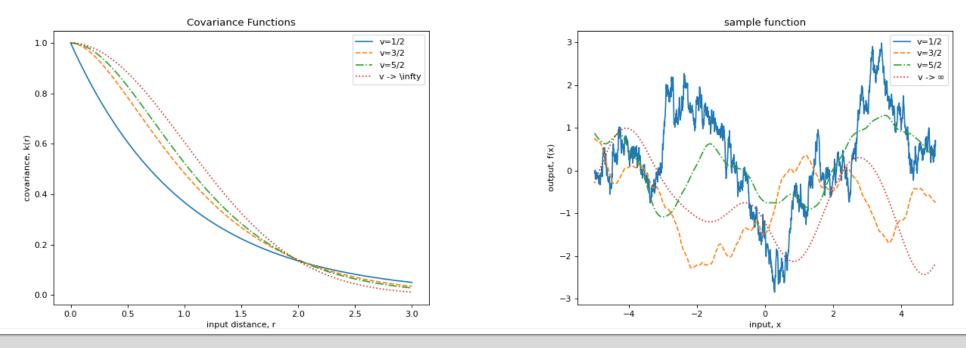
The likelihood is specified by the Kennedy-O'Hagan (KOH) model.





#### **Unsmooth covariance function is better for GP emulators**

- We use the Matérn kernel as the covariance function:  $k(\mathbf{x}, \mathbf{x}') = \eta \frac{2^{1-v}}{\Gamma(v)} \left(\frac{\sqrt{2v}|\mathbf{x}-\mathbf{x}'|}{l}\right)^v K_v \left(\frac{\sqrt{2v}|\mathbf{x}-\mathbf{x}'|}{l}\right)$ , where  $K_v$  is the modified Bessel function.
- Compared with the radial basis function, samples generated by the GP with a Matérn covariance function are **not smooth**, making it easier to emulate unsmooth outputs of the physics model.

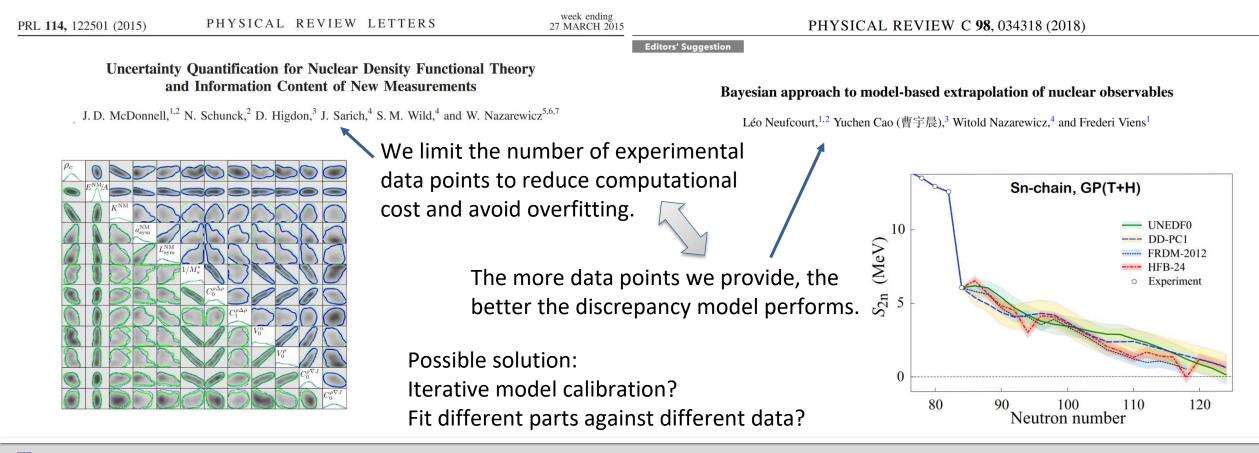






# Can the KOH model kill two birds with one stone?

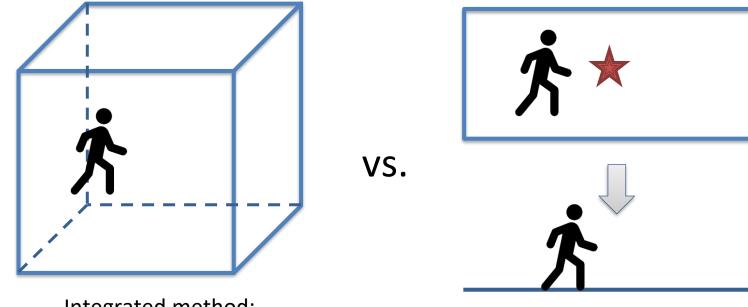
 Fitting the KOH model does not only calibrate parameters x used in the physics model, but also build a statistical model δ that can correct the physics model.





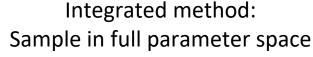
# **Integrated vs. modular approaches**

- With all the priors of parameters (including GP hyperparameters) and the likelihood specified, we can directly perform MCMC sampling in full parameter space.
- A faster but more approximate method is to sample different sets of parameters step-by-step.



- 1. Build the physics-model emulator f: Sample  $\theta_f$  to fit physics-model outputs.
- 2. Fix  $\boldsymbol{\theta}_f$  at its mean value.
- 3. Sample other parameters to fit experimental data.

Module method: Sample step-by-step





# Hierarchical priors ensure that we calibrate the physics model

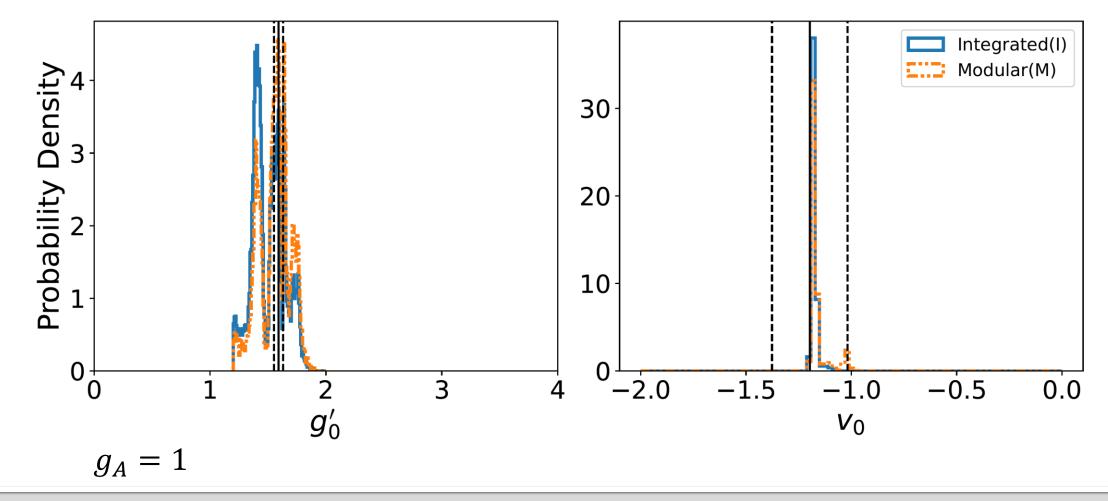
- If priors of GP hyperparameters are not chosen carefully, *f* can perform badly (far from experimental data) but δ can well reproduce experimental observations.
- A hierarchical prior structure for the magnitudes of variances ( $\sigma_{\epsilon}^2 < \sigma_{\delta}^2 < \sigma_{f}^2$ ) is thus adopted:

$$\begin{array}{l} 1/\eta_f ~~ \sim \Gamma(a=10,b=10), \\ 1/\eta_\delta ~~ \sim \Gamma(a=10,b=0.3), \\ 1/\sigma_\epsilon^2 ~~ \sim \Gamma(a=10,b=0.012). \end{array}$$

- The mean value of the Gamma distribution is a/b.
- This prior structure ensures that the variance in experimental data is most explained by *f*, less by δ, and least by ε.
- Is the choice of  $\eta_{\delta}$  for different data types equivalent to the choice of weights in the  $\chi^2$  optimization?



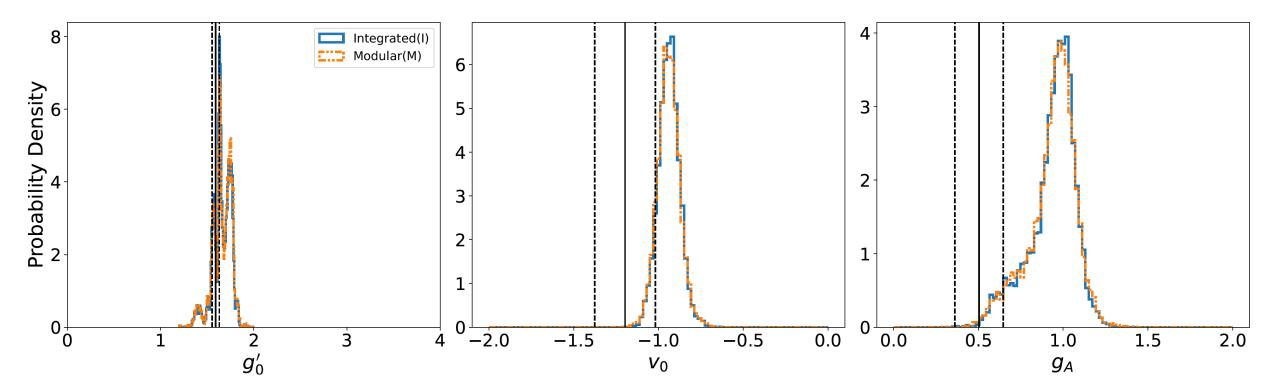
#### **Preliminary results: Posterior distributions**





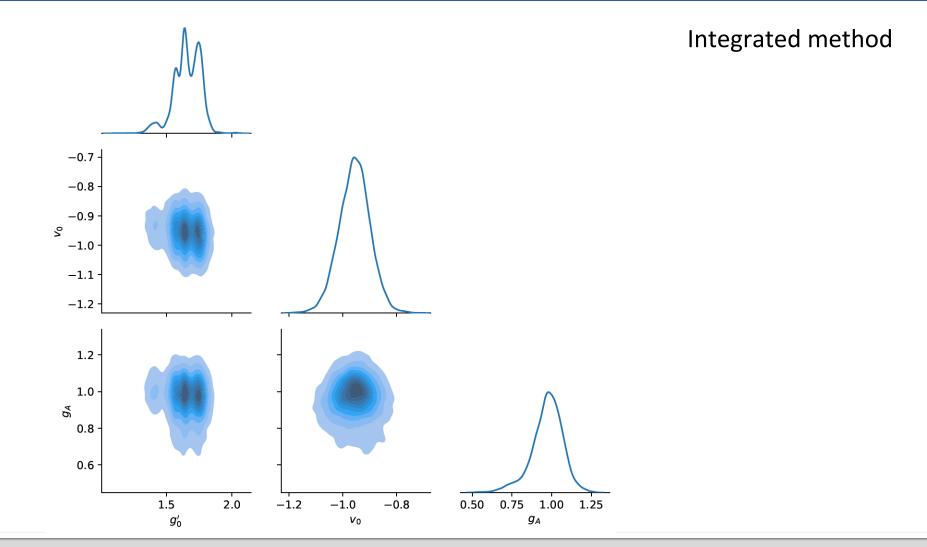


#### **Preliminary results: Posterior distributions**





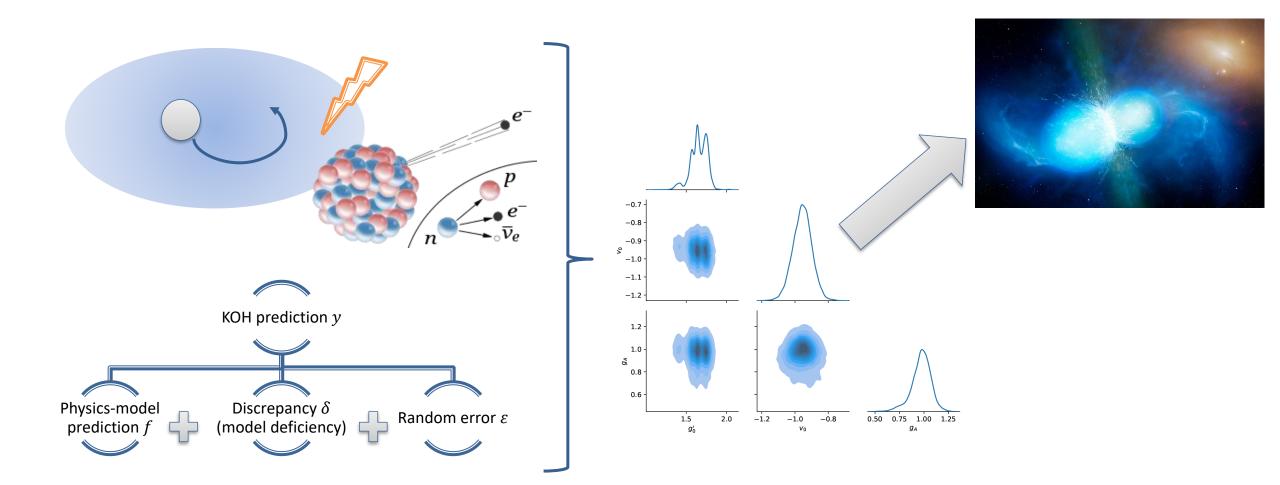
#### **Preliminary results: Correlation**



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#### **Conclusions and outlook**

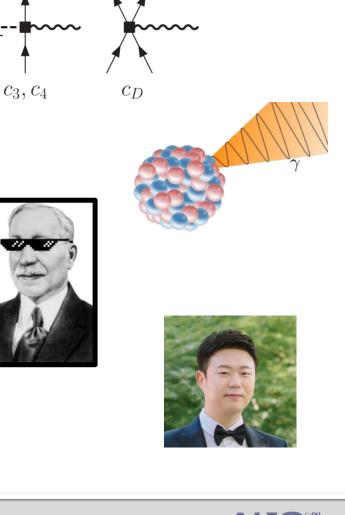




# **Conclusions and outlook**

- Astrophysical simulations may help us on the selection of fit targets.
- The FAM with two-body currents is ready to use, which eliminates the necessity to fit g<sub>A</sub> but introduces new unknown parameters.
  We can use the current framework to calibrate these new parameters.
- We will also extend our framework for other nuclear observables, such as the gamma decay.
- We are going to explore the application of the reduced basis method for better emulators of FAM codes.
- Statisticians are exploring the variational Bayesian method to totally remove the MCMC sampling in our framework.
- Bayesian model mixing can help us build models that vary with Z and N and provide best results in various regions.





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# **Backup Slides**



# Additional remarks on the $\chi^2$ optimization

No.	Nucleus	$E_{\rm GTR}$	Error	No.	Nucleus	$E_{\rm GTR}$	Error
1	<sup>208</sup> Pb	15.6	0.2	3	$^{90}$ Zr	8.7	_
2	$^{132}\mathrm{Sn}$	16.3	0.6	4	$^{112}$ Sn	8.94	0.25

- POUNDERS (Practical Optimization Using No Derivatives for sums of Squares) in PETSc/TAO is the numerical tool for the optimization.
- It was first used in the fits of UNEDF functionals (<u>Phys. Rev. C 82</u>, 024313 (2010)).

- The  $\chi^2$  optimization is based on the assumption that all the normalized residuals are independently Gaussian distributed:  $\varepsilon \sim N(0, \sigma^2 I_{n_d})$ .
- The probability distribution of  $\mathbf{x}$  is  $P(\mathbf{x}) \propto \exp\left[-\frac{\chi^2(\mathbf{x})}{2\sigma^2}\right] \approx C \exp\left[-\frac{1}{4\sigma^2}(\mathbf{x}-\widehat{\mathbf{x}})^T H(\mathbf{x}-\widehat{\mathbf{x}})\right]$ , where  $\widehat{\mathbf{x}}$  is the minimum point of  $\chi^2$ , H is the Hessian matrix (proportional to the covariance matrix of  $\mathbf{x}$ ).
- Approximate formula for the covariance matrix  $Cov(\hat{x}) \approx \chi^2(\hat{x})[G^T(\hat{x})G(\hat{x})]^{-1}$ , where Jacobian matrix G is defined as  $G_{ij} = \frac{\partial \varepsilon_i}{\partial x_j}$ .



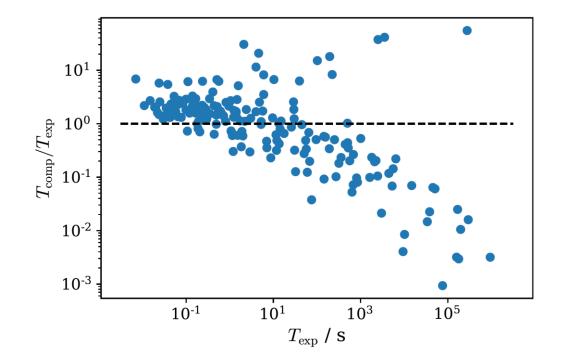
# Results with the tensor term included ( $\chi^2$ optimization)

Fit	$g_0'$	$oldsymbol{g'_1}$	$v_0$	${oldsymbol{g}}_A$
А	1.59560 (0.039)	0 (fixed)	-0.99993 (0.178)	1 (fixed)
В	1.59184 (0.034)	0 (fixed)	-1.19745 (0.179)	0.50345 (0.143)
С	1.73245 ( <mark>0.820</mark> )	-0.37034 ( <mark>2.143</mark> )	-0.99920 (0.183)	1 (fixed)
D	2.72206 (0.422)	-2.54125 (0.781)	-1.23511 (0.179)	0.41168 (0.132)

Strong correlation between  $g'_0$  and  $g'_1$  is seen in Fits C and D.



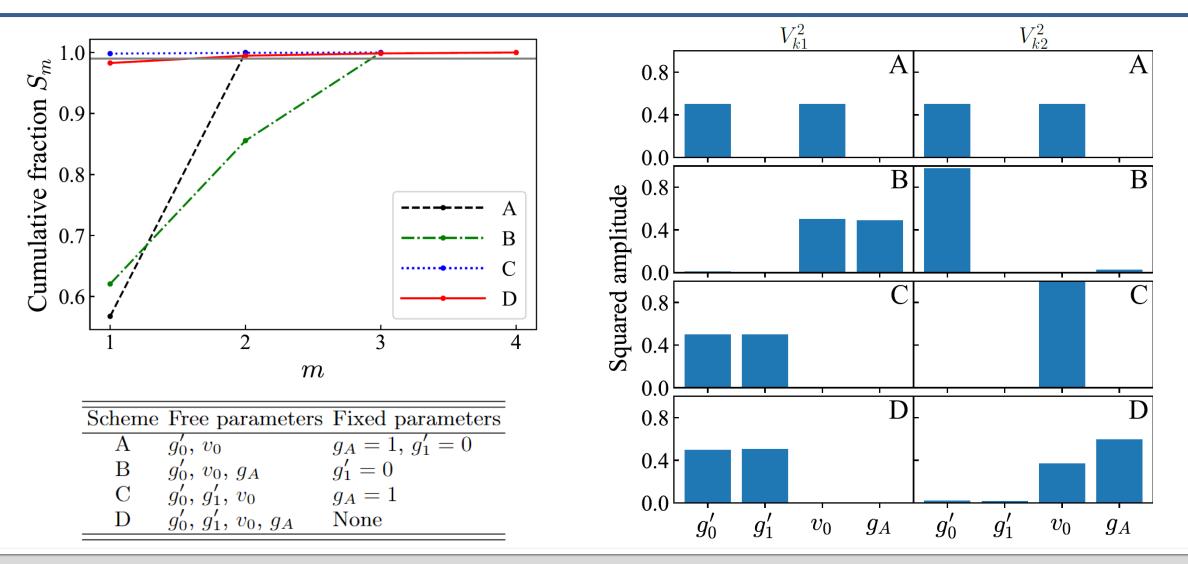
# **Predictions of Fit B**



- For even-even nuclei with known β<sup>-</sup>-decay half lives, the figure shows ratios between calculated (with optimal parameter values) and experimental half lives.
- The error is typical for beta-decay calculations.
- Errors are larger for nuclei with longer half lives, primarily due to the leptonic phase-space factor.

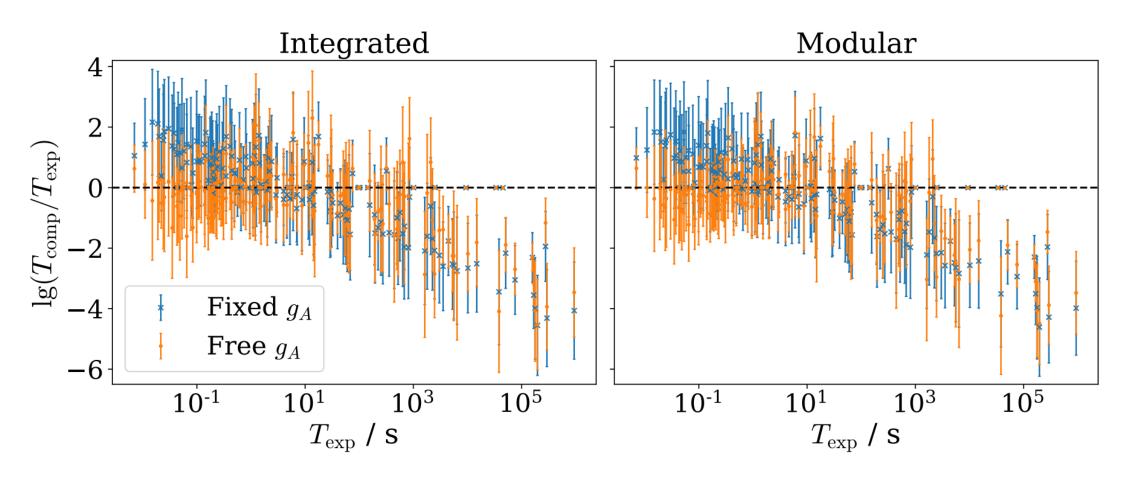


# Number of effective parameters (PCA, $\chi^2$ optimization)





# Predictions by the KOH model after calibration







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