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Efficient emulators for scattering using eigenvector continuation

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KVP method

$$Du_{\text{exat}}(r) = \left(-\frac{d^2}{dr^2} + \frac{\ell(\ell+1)}{r^2} + \boxed{U(r)} - p^2 \right) u_{\text{exat}}(r) = 0$$

Parameter dependent potential

1. Introduce a functional
2. Put into some ansatz as trial
3. Solve the optimization problem

$$\beta[u_t] = \tau_{\text{trial}} - \int_0^\infty dr u_t(r) Du_t(r) \quad u = \sum_i c_i \psi_i$$

$$u_t(r) \xrightarrow{r \rightarrow \infty} \frac{1}{p} \sin\left(pr - \ell \frac{\pi}{2}\right) + \tau_{\text{trial}} \cos\left(pr - \ell \frac{\pi}{2}\right)$$

We choose a set of basis functions, and minimize the functional by selecting the coefficients

How to construct the functional

Boundary condition of scattering wave function (Definition of the K-matrix)

$$u_{\ell,E}(r) \xrightarrow{r \rightarrow \infty} \frac{1}{p} \sin \left(pr - \ell \frac{\pi}{2} \right) + \frac{\mathcal{K}_{\ell}(E)}{p} \cos \left(pr - \ell \frac{\pi}{2} \right)$$

$$\mathcal{K}_{\ell}(E) = \tan(\delta_{\ell})$$

Take a variation at the exact scattering wave function

$$\delta\beta = \delta\tau - \int_0^{\infty} dr u_{\text{exact}}(r) D\delta u(r) + \mathcal{O}(\delta u^2)$$

Carry out the integration by parts for two times

$$\begin{aligned} \int_0^{\infty} dr u_{\text{exact}}(r) D\delta u(r) & \stackrel{\%}{=} \int_0^{\infty} dr u_{\text{exact}}(r) \left(-\frac{d^2}{dr^2} + \frac{\ell(\ell+1)}{r^2} + U(r) - p^2 \right) u_{\text{exact}} \delta u(r) \\ & = \left(u_{\text{exact}} \delta u' - u'_{\text{exact}} \delta u \right) \Big|_0^{\infty} + \int dr u D u_{\text{exact}} \end{aligned}$$

How to construct the functional

With the boundary condition and some sin/cos algebras

$$\left(u_{\text{exact}} \delta u' - u'_{\text{exact}} \delta u \right) \Big|_0^{\infty} = \delta \tau$$

So we have proved $\delta \beta[u] \Big|_{u_{\text{exact}}} = \delta \tau - \delta \tau = 0$

Remind that $\beta [u_{\text{exact}}] = \frac{1}{p} [\mathcal{K}_{\ell}(E)]_{\text{exact}} = \frac{1}{p} [\tan \delta_{\ell}(E)]_{\text{exact}}$

EC with KVP

EC provides us with a set of basis, and reduce the size of the basis

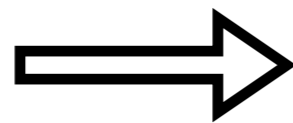
Snapshots

$$D(\theta_i)u_i = 0 \quad i = 1, 2, \dots, N_b$$

Differential equation



Finite difference



Number of the grid points $\gg N_b$



Linear equation



Size of the matrix $\gg N_b$

R-matrix: solve the large linear equations “**exactly**” for 1000 times

EC: solve “**exactly**” N_b times + emulate other situations $1000 - N_b$ times

Where the efficiency comes from

EC with KVP

Insert the trial solution into the functional

$$\begin{aligned}\beta [u_t] &= \sum_j c_j \tau_j(E) - \sum_{jk} c_j c_k \int_0^\infty dr u_j(r; E) \left[-\frac{d^2}{dr^2} + \frac{l(l+1)}{r^2} + 2\mu V(r; \boldsymbol{\theta}) - p^2 \right] u_k(r; E) \\ &= \sum_j c_j \tau_j - \sum_{jk} c_j c_k \int_0^\infty dr u_j(r; E) (2\mu) [V(r; \boldsymbol{\theta}) - V_k(r)] u_k(r; E)\end{aligned}$$

Define the U matrix for convenient

$$\Delta U = \int_0^\infty dr u_j(r; E) (2\mu) [V(r; \boldsymbol{\theta}) - V_k(r)] u_k(r; E)$$

The functional now turns into a matrix form

$$\beta = \sum_j c_j \tau_j - \sum_{jk} c_j c_k \Delta U_{jk}$$

EC with KVP

Take the partial derivatives and add an Lagrange multiplier

$$\frac{\partial}{\partial c_i} \left[\sum_j c_j \tau_j - \sum_{jk} c_j c_k \Delta U_{jk} - \lambda \left(\sum_j c_j - 1 \right) \right] = 0$$
$$\frac{\partial}{\partial \lambda} \left[\sum_j c_j \tau_j - \sum_{jk} c_j c_k \Delta U_{jk} - \lambda \left(\sum_j c_j - 1 \right) \right] = 0.$$

Finally we have the expression for the coefficients and multiplier

$$c_j = \sum_i (\Delta \widetilde{U})_{ji}^{-1} (\tau_i - \lambda) \quad \lambda = \frac{\sum_{ij} (\Delta \widetilde{U})_{ji}^{-1} \tau_i}{\sum_{ij} (\Delta \widetilde{U})_{ji}^{-1}}$$

Kohn anomalous issues

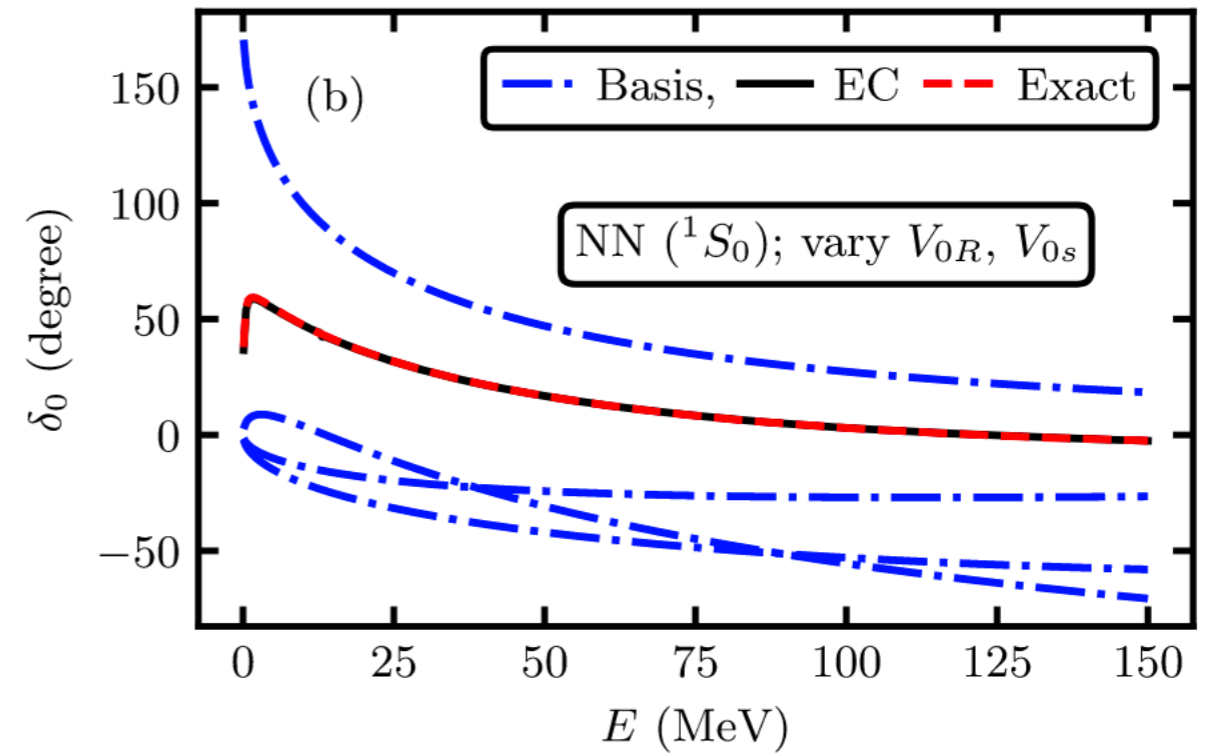
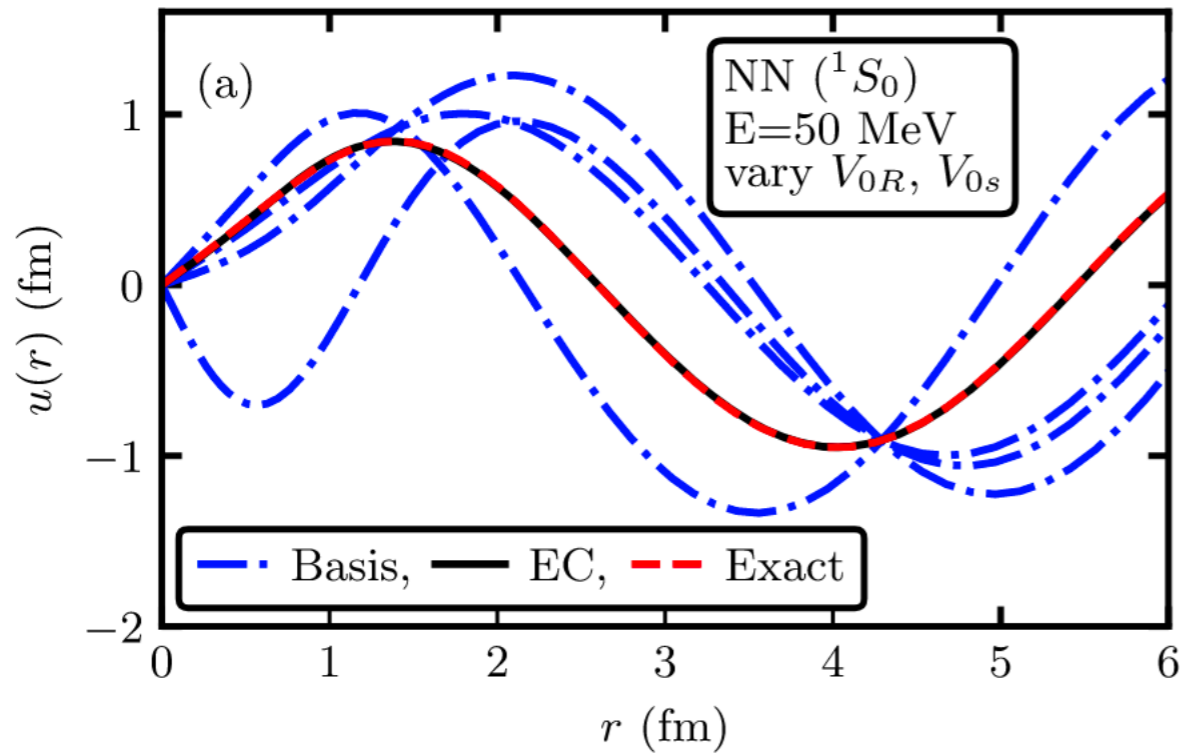
The matrix to be inverted may be expected to be increasingly **ill-conditioned** as the basis size increases.

1. Complex formulation of S - matrix
2. Pseudo inverse method
3. **Regularization: adding a small value to the diagonal**

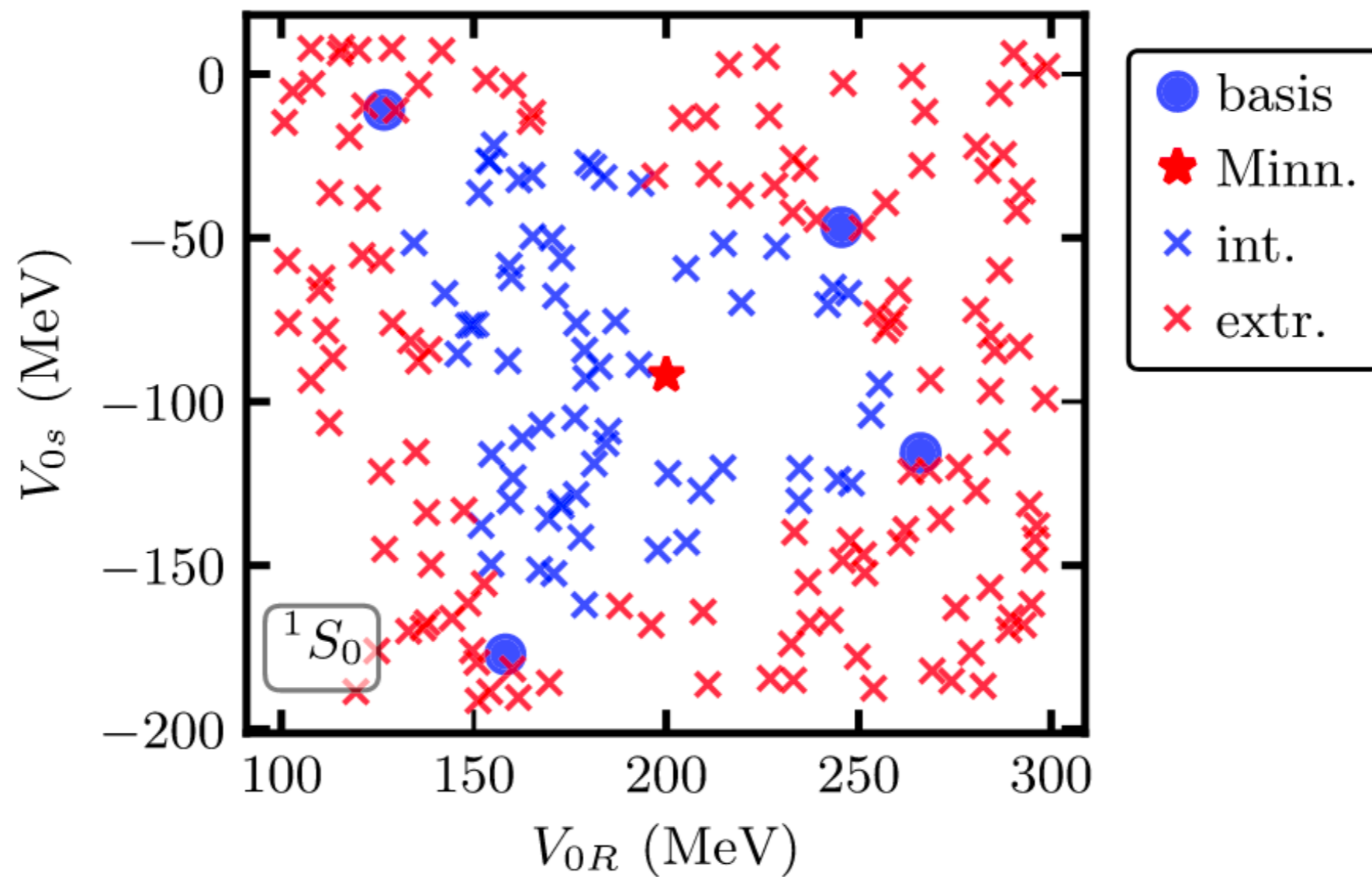
Nugget: $10^{-8} \sim 10^{-10}$

N-N scattering

$$V_{1S_0}(r) \equiv V_{0R}e^{-\kappa_R r^2} + V_{0s}e^{-\kappa_s r^2}$$
$$V_{3S_1}(r) \equiv V_{0R}e^{-\kappa_R r^2} + V_{0t}e^{-\kappa_t r^2}$$

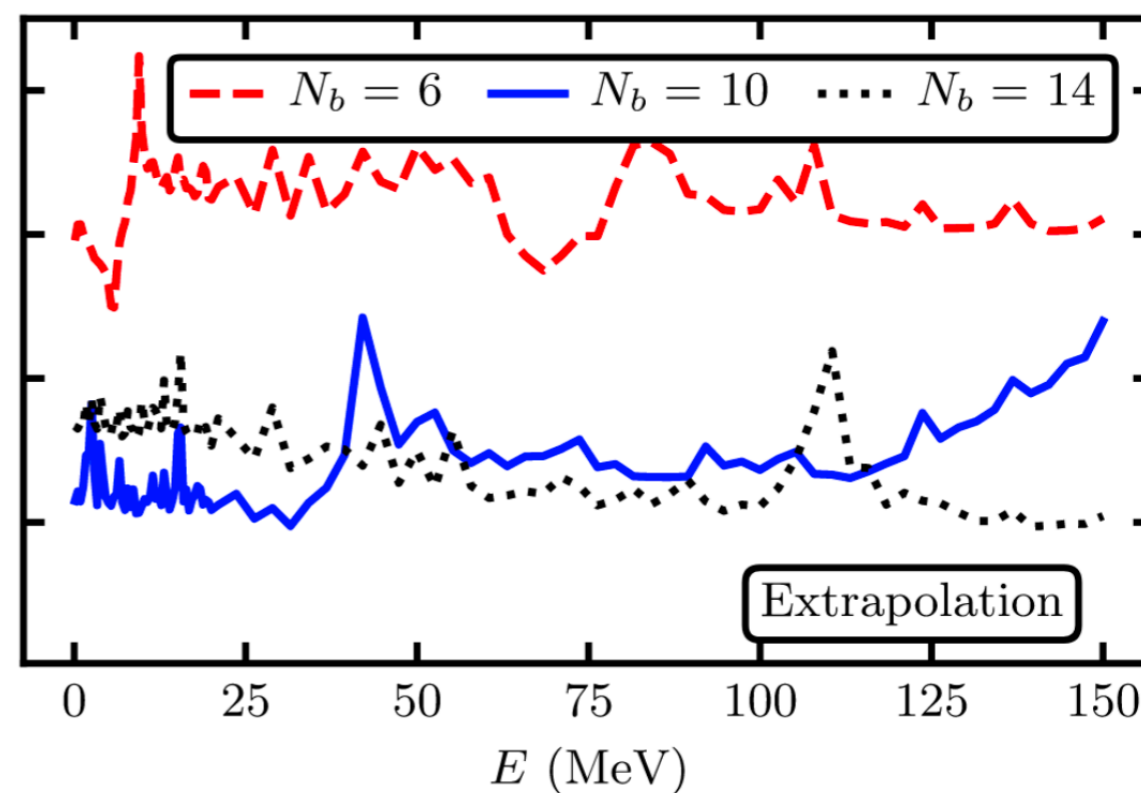
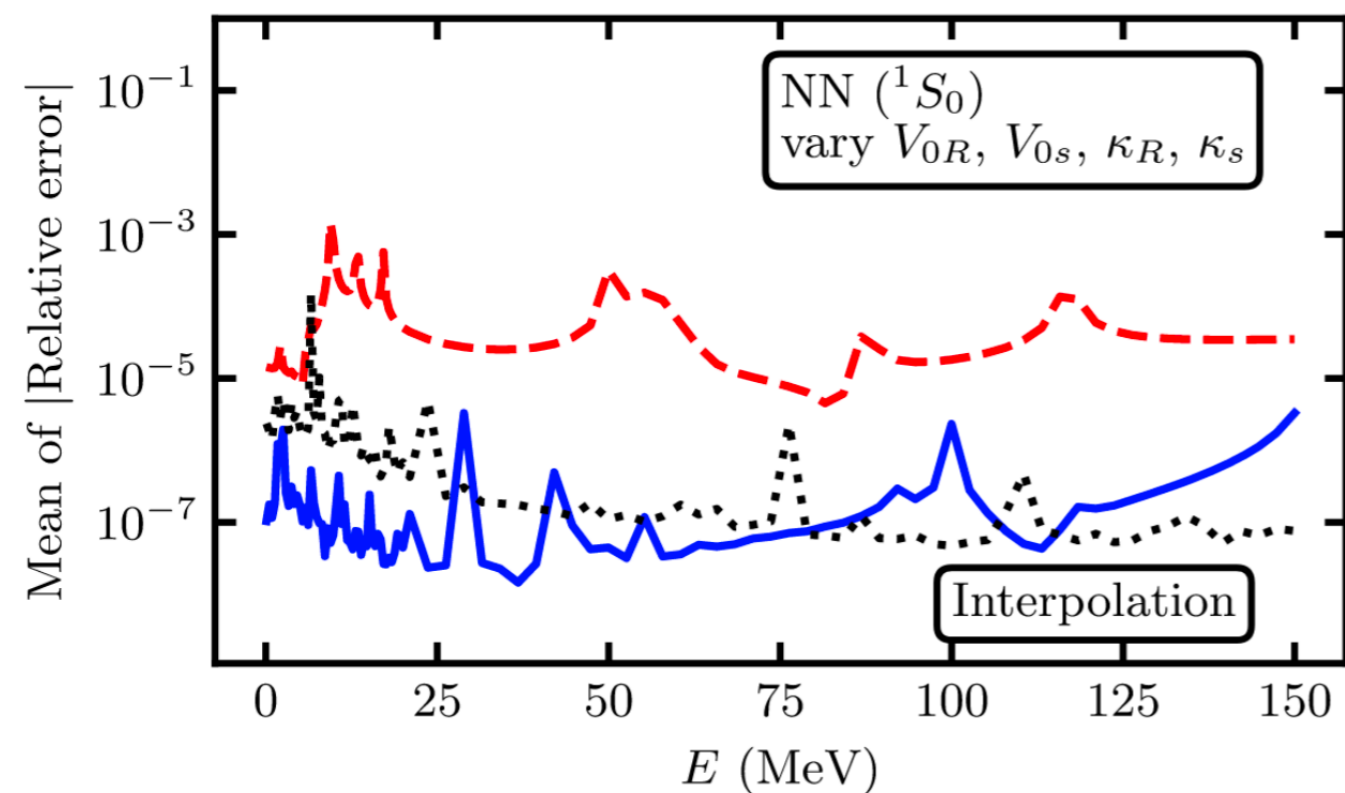
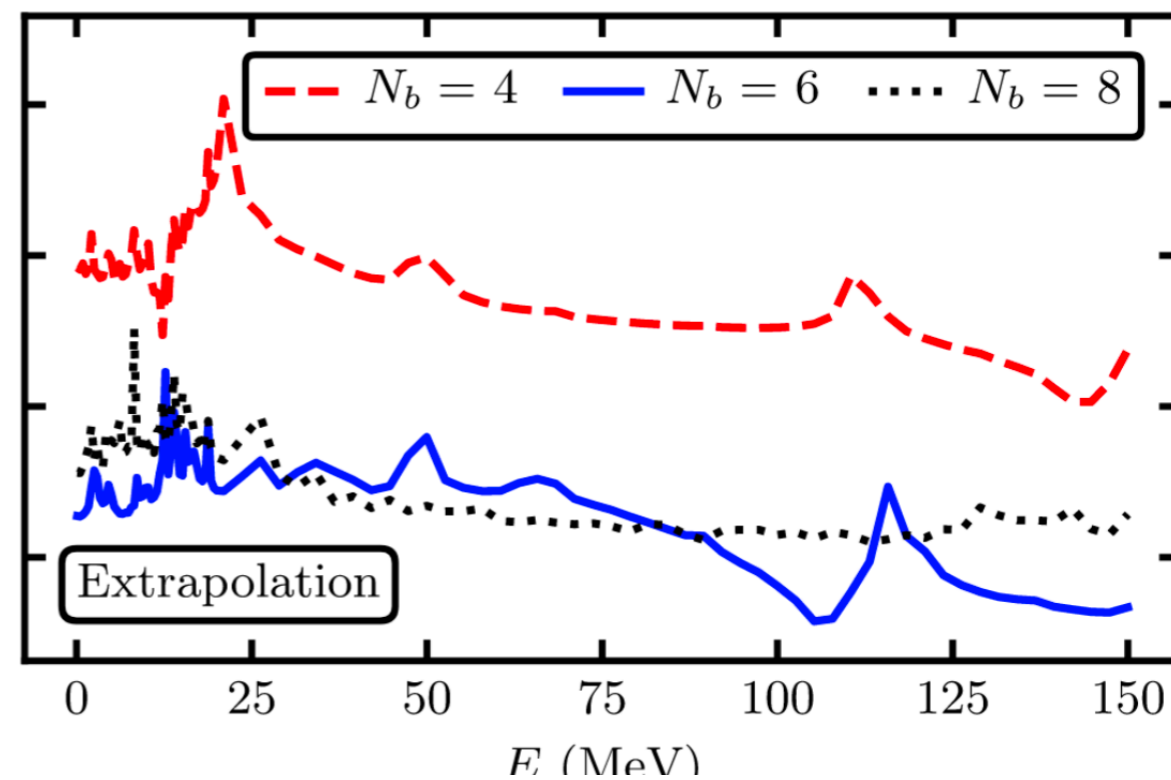
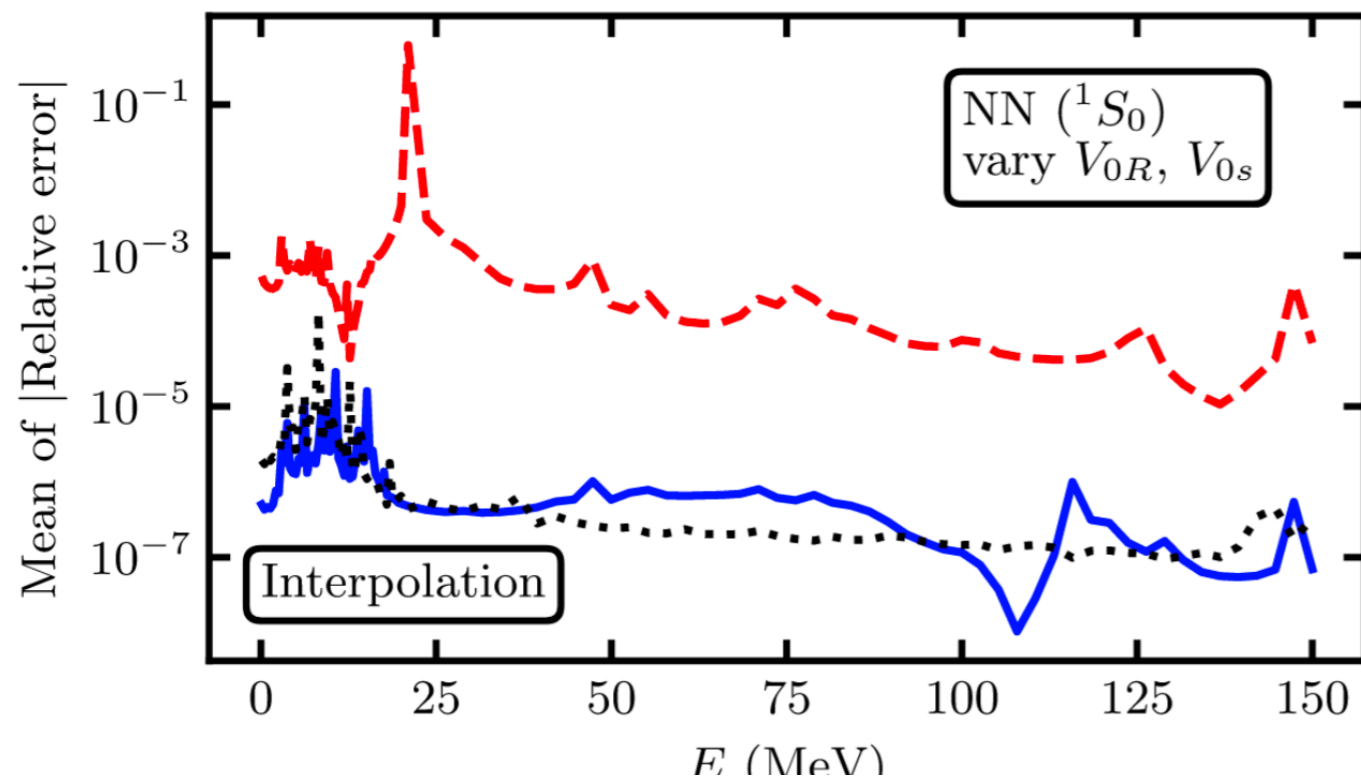


N-N scattering



Latin-hypercube sampling method

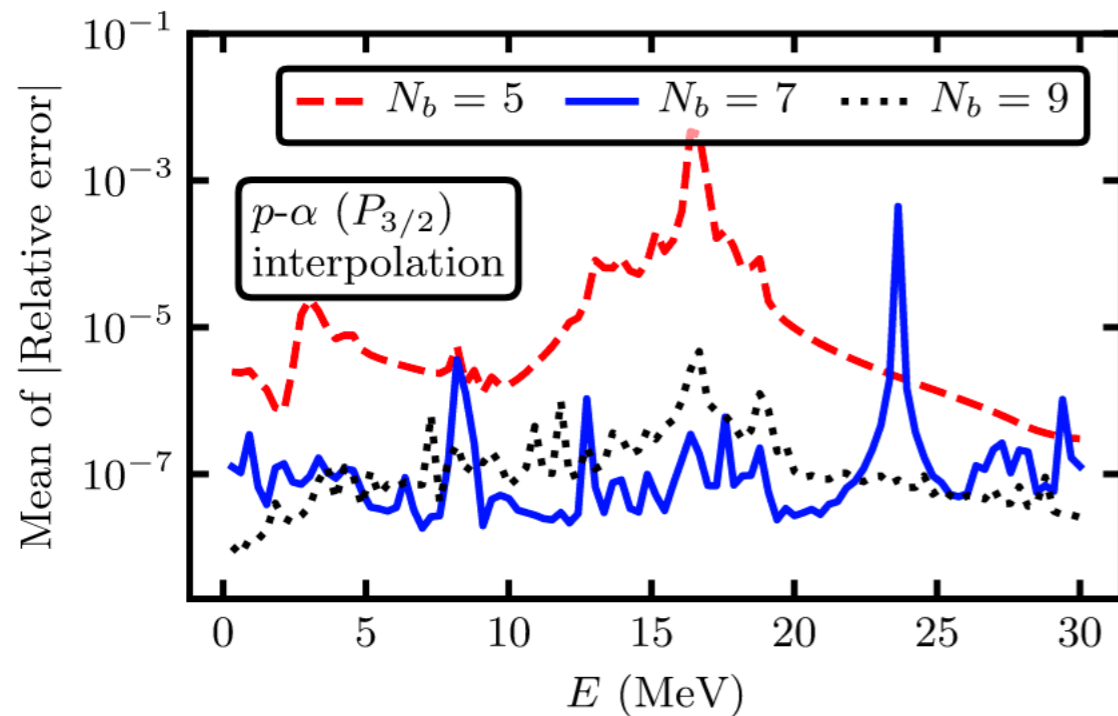
N-N scattering



Other applications

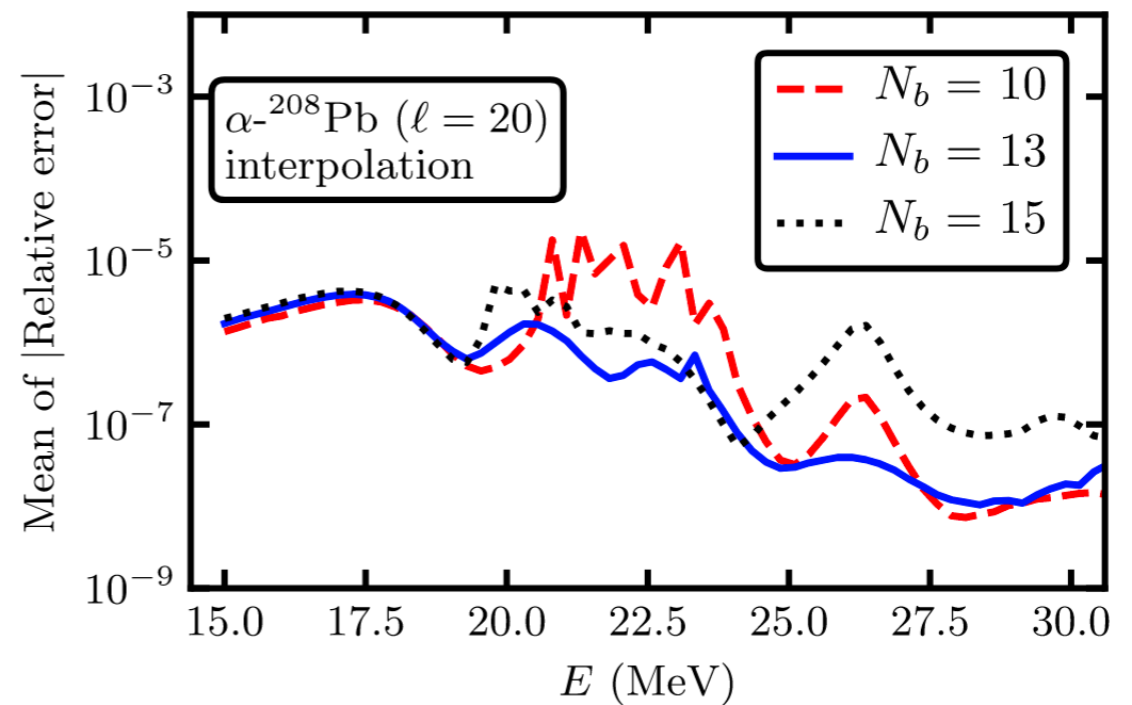
$p - \alpha$ scattering

$$V_\ell(r', r) = V_{p\alpha, \ell}^{(0)} r'^\ell r^\ell e^{-\beta_\ell (r' + r)}$$



$\alpha - {}^{208}\text{Pb}$ scattering

Woods-Saxon form



Summary

EC basis for KVP is accurate and efficient:

1. It can produce good results compared with the exact R-matrix method
2. Instead of carrying out the diagonalization, only the inverse operations and some multiplication are needed.

But there are some limitations:

1. Can only deal with a specific energy
2. Can only deal with a specific partial wave
3. Cannot emulate the observable