Eigenvector continuation in nuclear physics

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SK, A. Ekström, K. Hebeler, A. Sarkar, D. Lee, A. Schwenk, in preparation



Outline

Genesis

Evolution

Exodus

Emulation

Motivation

Many physics problems are tremendously difficult...

- huge matrices, possibly too large to store
 - ever more so given the evolution of typical HPC clusters
- most exact methods suffer from exponential scaling
- interest only in a few (lowest) eigenvalues





Motivation

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Martin Grandjean, via Wikimedia Commons (CC-AS 3.0)

Introducing eigenvector continuation

Frame et al., PRL 121 032501 (2018)



KDE Oxygen Theme

- novel numerical technique
- can solve otherwise untractable problems
- amazingly simple in practice
- broadly applicable
- this talk: nuclear nails

Hubbard model

- three-dimensional Bose-Hubbard model (4 bosons on $4 \times 4 \times 4$ lattice)
- hopping parameter t, on-site interaction $U \rightsquigarrow H = H(c = U/t)$



- Bose gas for c > 0, weak binding for -3.8 < c < 0, tight cluster for c < -3.8
- eigenvector continuation can extrapolate across regimes



General idea

Scenario

Frame et al., PRL **121** 032501 (2018)

- consider physical state (eigenvector) in a large space
- parametric dependence of Hamiltonian H(c) traces only small subspace

Procedure

- calculate $|\psi(c_i)
 angle$, $i=1,\ldots N_{
 m EC}$ in "easy" regime
- solve generalized eigenvalue problem $H|\psi
 angle=\lambda N|\psi
 angle$ with
 - $H_{ij} = \langle \psi_i | H(c_{ ext{target}}) | \psi_j
 angle$
 - $N_{ij}=\langle \psi_i|\psi_j
 angle$

Prerequisite

- smooth dependence of H(c) on c
- enables analytic continuation of $|\psi(c)
 angle$ from c_{easy} to c_{target}

Part I

SRG evolution

SRG evolution

- unitary transformation of Hamiltonian: $H o H_\lambda = U_\lambda H U_\lambda^\dagger \rightsquigarrow V_\lambda$
- decouple low and high momenta at scale λ



R. Furnstahl, HUGS 2014 lecture slides

- interaction becomes more amenable to numerical methods...
- ...at the cost of induced many-body forces!





SRG evolution = ODE solving

$$rac{\mathrm{d} H_s}{\mathrm{d} s} = rac{\mathrm{d} V_s}{\mathrm{d} s} = [[G,H_s],H_s]$$
 , $\lambda = 1/s^{1/4}$

ordinary differential equation ensures smooth parametric dependence

\hookrightarrow SRG evolution satisfies EC prerequisites!

Reverse SRG

Consider A = 3,4 test cases

• EMN N3LO(500) interaction, Jacobi NCSM calculation

Entem et al., PRC 96 024004 (2017); A. Ekström implementation of Navratil et al., PRC 61 044001 (2000)

Reverse SRG

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- possible to extrapolate back from small λ to bare interaction
- information about missing many-body forces in wavefunctions
 - ▶ not in any single wavefunction, but in how they change

Mind the gap

Still no free lunch, however...

- EC is a variational method
- cannot go beyond what bare interaction gives in same model space!



So now what?

Part II

Escaping the model space

Model-space perturbation theory

• consider a Hamiltonian diagonalized in a (small) subspace

$$H=egin{pmatrix} H_{\phi\phi} & H_{\phi\psi}\ H_{\psi\phi} & H_{\psi\psi} \end{pmatrix}$$

 $N_0 = \dim H_{\phi\phi} \, \ll \, \dim H = N_1$

$$H_{\phi\phi}= ext{diag}(\{\lambda_i\}_{i=1,\cdots N_0})$$



- factor out large number X from diagonal entries of $H_{\psi\psi}$
- perturbative expansion for lowest eigenvalue and vector

$$|\psi_1
angle = \sum_{n=0}^\infty X^{-n} \left(\sum_{i=1}^{N_0} x_i^{(n)} |\phi_i
angle + \sum_{j=N_0+1}^{N_1} x_j^{(n)} |\psi_j
angle
ight) \;,\; \lambda_1^{\mathrm{full}} = \sum_{n=0}^\infty X^{-n} \lambda_1^{(n)}$$

▶ matching powers gives coupled recursive expressions for $x_i^{(n)}$ and $\lambda_1^{(n)}$

Model-space perturbation theory (cont'd)

Diagonalizing a small space can still be too expensive...

Model-space perturbation theory (cont'd)

Diagonalizing a small space can still be too expensive...

- actually, a partial diagonalization per se is ok (ightarrow Lanczos)
- but transforming the Hamiltonian is problematic...



- cost for adjusting off-diagonal elements is significant
 - ► scales with size of the full (large) space

Way out

Start from one-dimensional space ($N_{\rm max} = 0$)...



...i.e., directly use the given Hamiltonian

Failure

³H NCSM calculation, $N_{\rm max} = 12$ model space

• EMN N3LO 500 interaction

Entem et al., PRC 96 024004 (2017)



- perturbation theory does not converge!
 - however, interaction clearly "more perturbative" for small SRG λ
 - convergence perhaps for very small λ

Saved by EC

- span space by the wavefunction corrections $|\psi_1^{(n)}
 angle o x_j^{(n)}$, $n=0,\cdots$ $ext{order}$
- evaluate Hamiltonian between these states
- interpretation: $H = H_{
 m diag} + c \, H_{
 m off\text{-}diag}$, EC-extrapolate to c=1



• same input as PT, but now things converge (to the correct result!)

Part III

EC as efficient emulator

Hamiltonian parameter spaces

• consider now a Hamiltonian depending on several parameters:

$$H = H_0 + V = H_0 + \sum_{k=1}^d c_k V_k$$
 (1)

- in particular, V can be a chiral potential with LECs c_k
- Hamiltonian is element of *d*-dimensional parameter space
- typical for $\mathcal{O}(Q^3)$ calculation: 14 two-body LECs + 2 three-body LECs
- convenient notation: $\vec{c} = \{c_k\}_{k=1}^d$

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Generalized EC

- EC construction is straightforward to generalize to this case:
- simple replace $c_i
 ightarrow ec{c}_i$ in construction

•
$$|\psi_i
angle=|\psi(ec{c}_i)
angle$$
 for $i=1,\cdots N_{
m EC}$

• $H_{ij} = \langle \psi_i | H(ec{c}_{ ext{target}}) | \psi_j
angle$, $N_{ij} = \langle \psi_i | \psi_j
angle$

Note: sum in Eq. (1) can be carried out in small (dimension = $N_{\rm EC}$) space!

Need for emulators

1. Fitting of LECs to few- and many-body observables

- common practice now to use A>3 to constrain nuclear forces, e.g.:
 - JISP16, NNLO_{sat}, α-α scattering
 Shirokov et al., PLB 644 33 (2007); Ekström et al., PRC 91 051301 (2015); Elhatisari et al., PRL 117 132501 (2016)
- fitting needs many calculations with different parameters

2. Propagation of uncertainties

- statistical fitting gives posteriors for LECs
- LEC posteriors propagate to observables Wesolowski et al., JPG **46** 045102 (2019)
- need to sample a large number of calculations
- cf. talks yesterday by Sarah and Gautam



Emulators

Exact calculations can be prohibitively expensive!

Options

- multi-dimensional polynomial interpolation
 - simplest possible choice
 - typically too simple, no way to assess uncertainty
- Gaussian process



► statistical modeling, iteratively improvable

Ekström et al., arXiv:1902.00941

interpolation with inherent uncertainty estimate

Recall

Eigenvector continuation can extrapolate!

Interpolation and extrapolation

Hypercubic sampling

- want to cover parameter space $S = \{ ec{c}_i \}$ efficiently
- Latin Hypercube Sampling can generate near random sample
- for examples that follow:
 - ullet sample each component $c_k \in [-2,2]$
 - vary d LECs, fix the rest at NNLO_{sat} point

Ekström et al., PRC **91** 051301 (2015);

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Convex combinations

- distinguish interpolation and extrapolation target points
- interpolation region is convex hull of the $\{ec{c}_i\}$
 - $\operatorname{conv}(S) = \sum_i lpha_i ec{c}_i$ with $lpha_i \geq 0$ and $\sum_i lpha_i = 1$
- extrapolation for $ec{c}_{ ext{target}}
 ot \in \operatorname{conv}(S)$
- EC can handle both!



Pbroks13, Wikimedia Commons

Performance comparison: energy

Cross validation

- compare emulation prediction agains exact result for set $\{ec{c}_{ ext{target},j}\}_{j=1}^N$
- underlying calculation: Jacobi NCSM (again)
- transparent symbols indicate extrapolation targets



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Performance comparison: radius

Operator evaluation

- generalized eigenvalue problem
- EC gives not only energy, but also a continued wavefunction
- straightforward (and inexpensive) to evaluate arbitrary operators



EC uncertainty estimate

- EC is a variational method
 - projection of Hamiltonian onto a subspace
 - dimension of this subspace determines the accuracy
 - ► rate of convergence currently being analyzed

D. Lee + A. Sarkar, work in progress

Bootstrap approach

• leave out basis vectors one at a time, take mean and standard deviation



Summary and outlook

This talk

- eigenvector continuation can be used to reverse SRG
 - conceptually interesting: implicit information about induced forces
- convergent perturbative model-space extension
 - effectively tame divergent expansion coefficients
 - interesting as computational method
- eigenvector continuation efficient emulator
 - highly competitive, accurate and efficient
 - can both interpolate and extrapolate from training set

Future directions

- larger systems, other methods (in particular: m-scheme NCSM)
- combined model-space and SRG EC
- application for large-scale uncertainty quantification

Thanks...

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