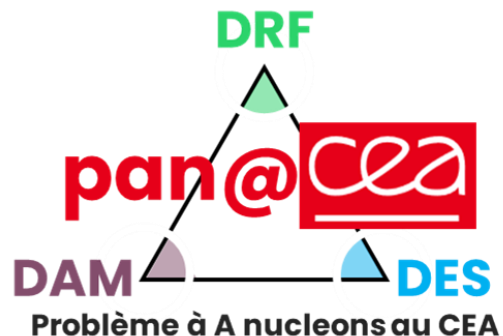




Extending the reach of nuclear ab initio calculations via dimensionality reduction techniques

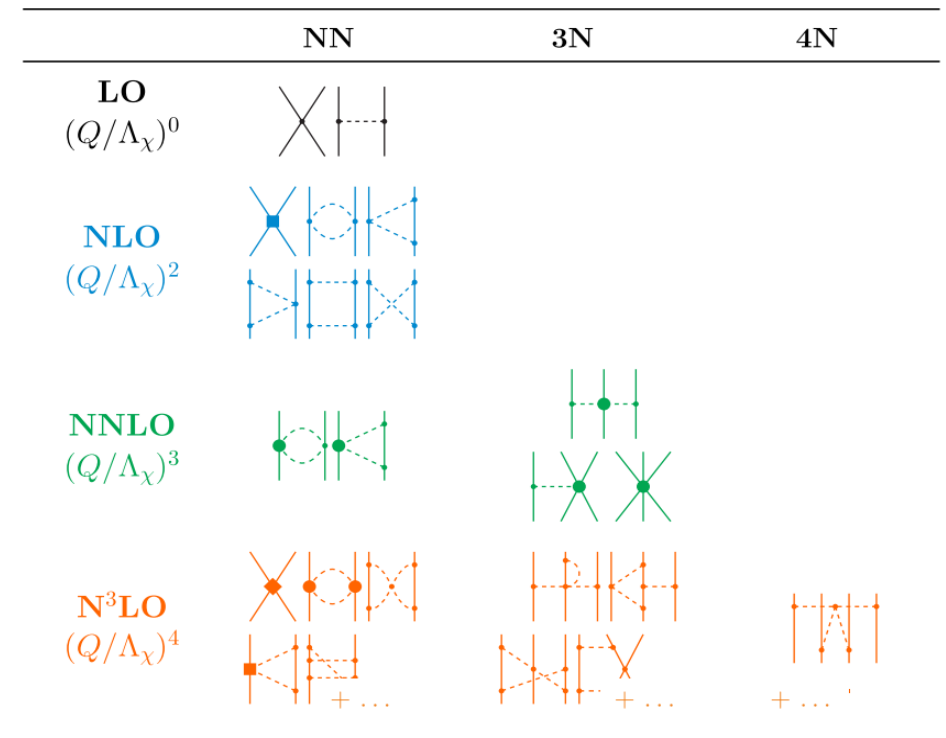
Lars Zurek

with Thomas Duguet, Jean-Paul Ebran, and Mikael Frosini

EuNPC, Caen, September 22, 2025

The nuclear ab initio approach

- Solve many-body Schrödinger equation for nuclear interactions from chiral effective field theory
- Rooted in quantum chromodynamics
- Systematically improvable
- Uncertainty estimates “built-in”



The nuclear ab initio approach

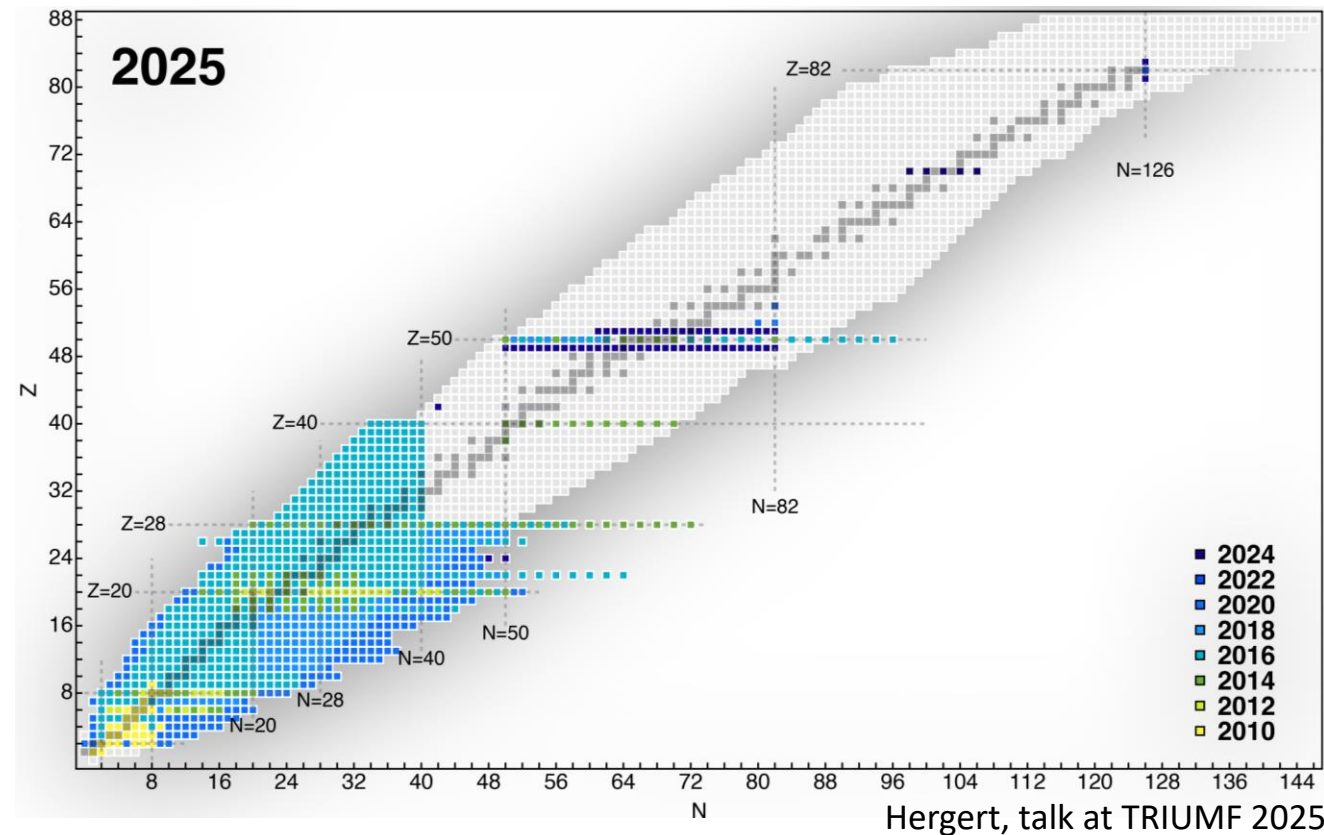
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See talk of V. Somà tomorrow
14:00, Inspire 1

	NN	3N	4N
LO $(Q/\Lambda_\chi)^0$			
NLO $(Q/\Lambda_\chi)^2$			
NNLO $(Q/\Lambda_\chi)^3$			
N³LO $(Q/\Lambda_\chi)^4$			

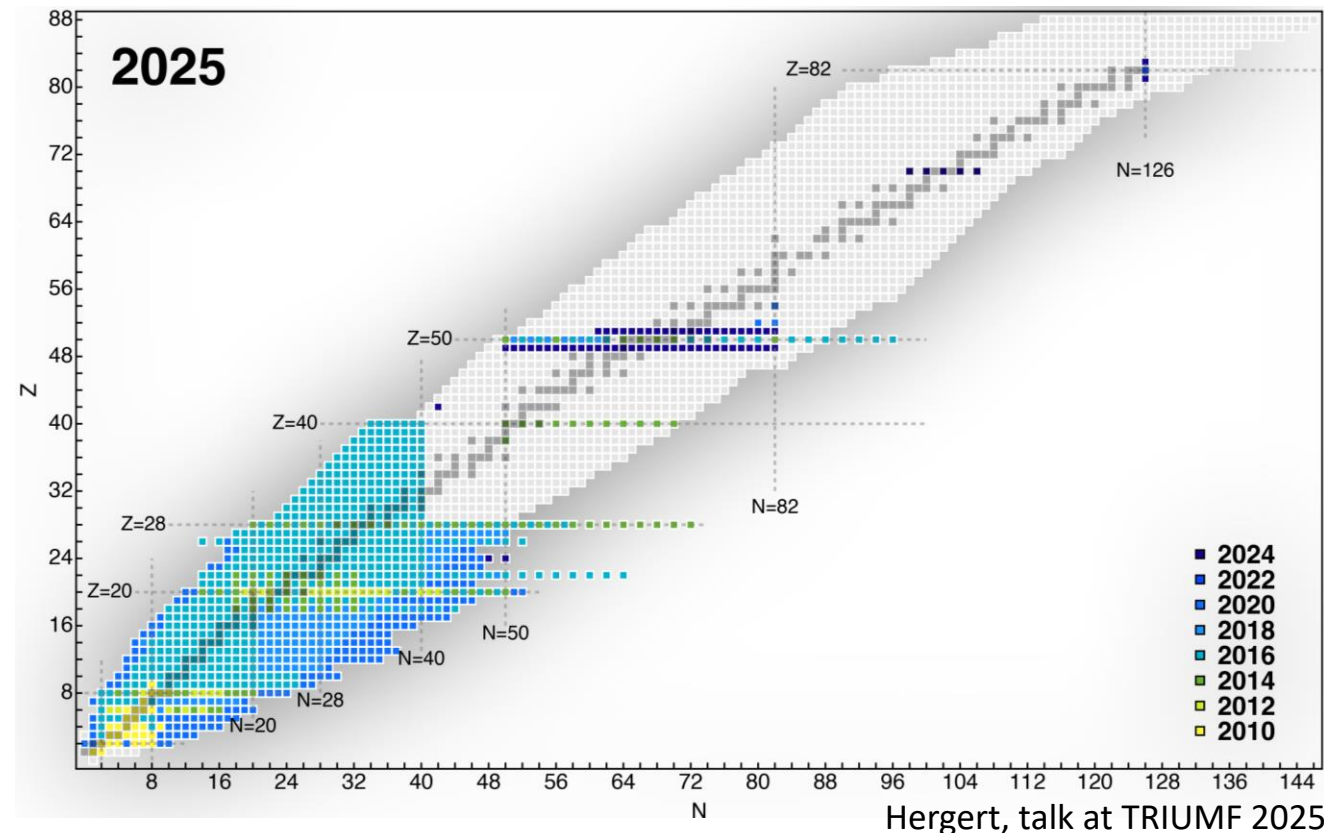
The nuclear ab initio approach

- “Exact” solution of many-body Schrödinger equation (by diagonalization)
limited to lightest nuclei
- Instead: correlation expansion methods
 - Heavy nuclei
 - Open-shell nuclei
 - Precision calculations



Ab initio challenges

- “Exact” solution of many-body Schrödinger equation (by diagonalization) limited to lightest nuclei
- Instead: correlation expansion methods
 - Heavy nuclei
 - Open-shell nuclei
 - Precision calculations
- But: current ab initio calculations cannot do all three at the same time



Dimensionality reduction

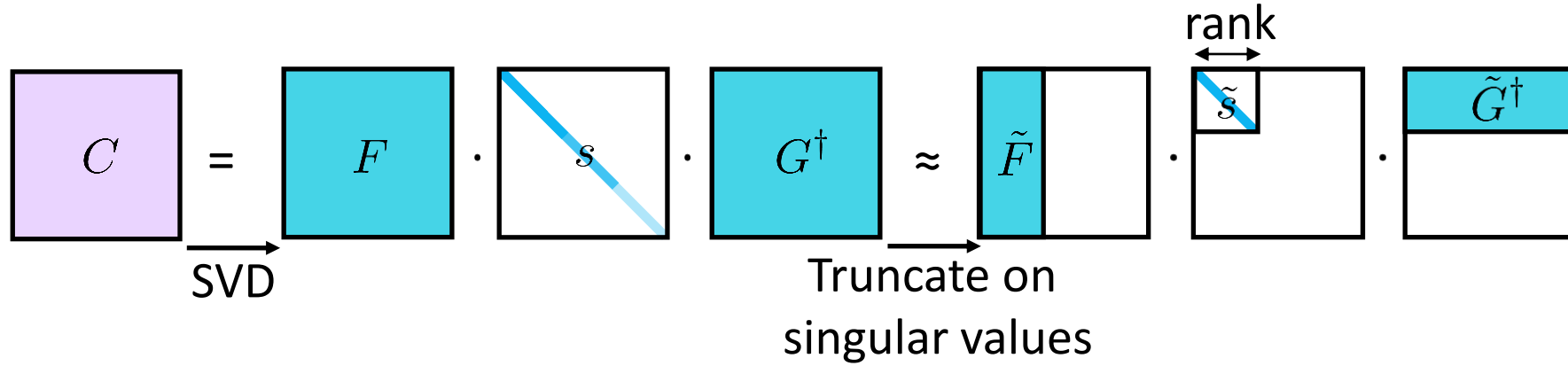
- Different strategies to reduce computational burden
- Using a better single-particle basis: approximate natural orbitals
Tichai et al., PRC **99** (2019)
Scalesi et al., EPJA **61** (2025)
- Eliminate irrelevant matrix elements by importance truncation
Roth et al., PRL **99** (2007)
Porro et al., EPJA **57** (2021)
- Replace tensors by multiple smaller ones via tensor factorization
Frosini et al., EPJA **60** (2024)
Tichai et al., PRR **6** (2024)

Singular value decomposition

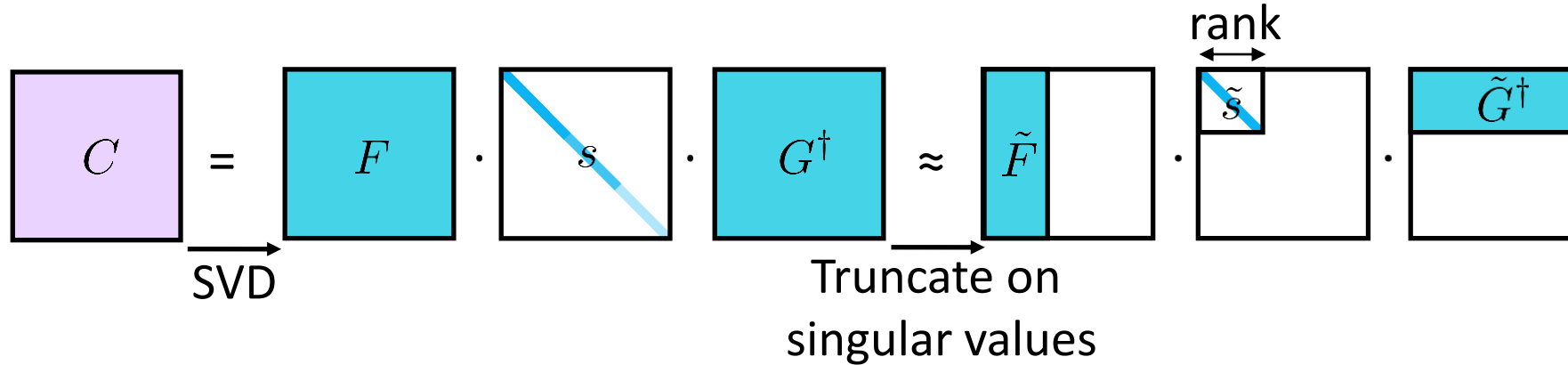
The diagram illustrates the Singular Value Decomposition (SVD) of a matrix C . On the left, a purple square labeled C is shown. An arrow labeled "SVD" points from C to the right. To the right of the arrow is an equals sign, followed by three matrices multiplied together. The first matrix is a cyan square labeled F . This is followed by a dot, then a white square with a blue diagonal line from the top-left to the bottom-right, labeled s . This is followed by another dot, then a cyan square labeled $G^\dagger$.

$$C \xrightarrow{\text{SVD}} F \cdot s \cdot G^\dagger$$

Singular value decomposition

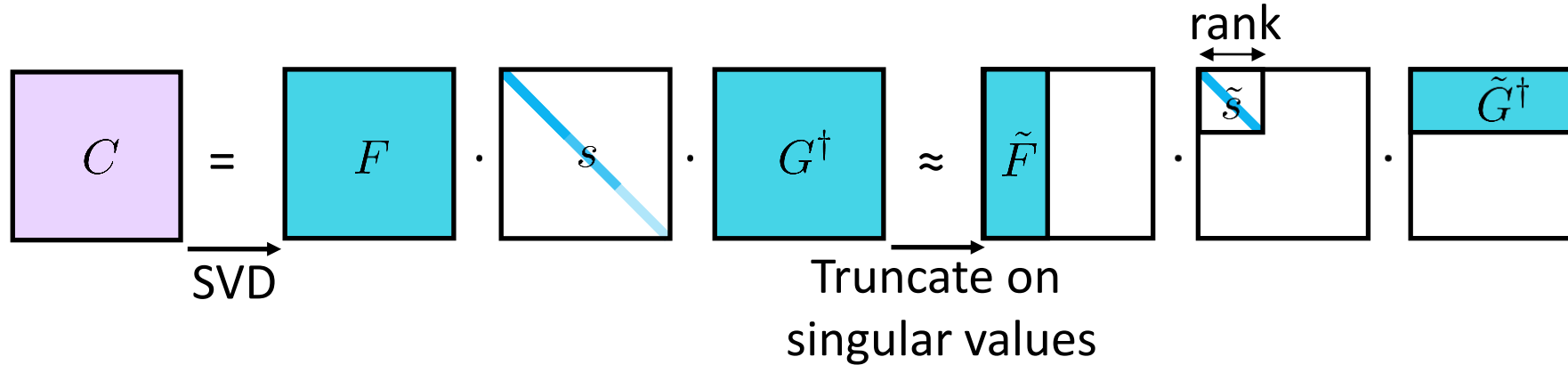


Singular value decomposition



- Eckart-Young theorem: truncated SVD gives best possible rank- r approximation of C (measured in terms of Frobenius norm)

Singular value decomposition



- Eckart-Young theorem: truncated SVD gives best possible rank- r approximation of C (measured in terms of Frobenius norm)
- Large computational benefits possible if small ranks can be used
- Maximal gain only when full matrices are never reconstructed

Bogoliubov many-body perturbation theory

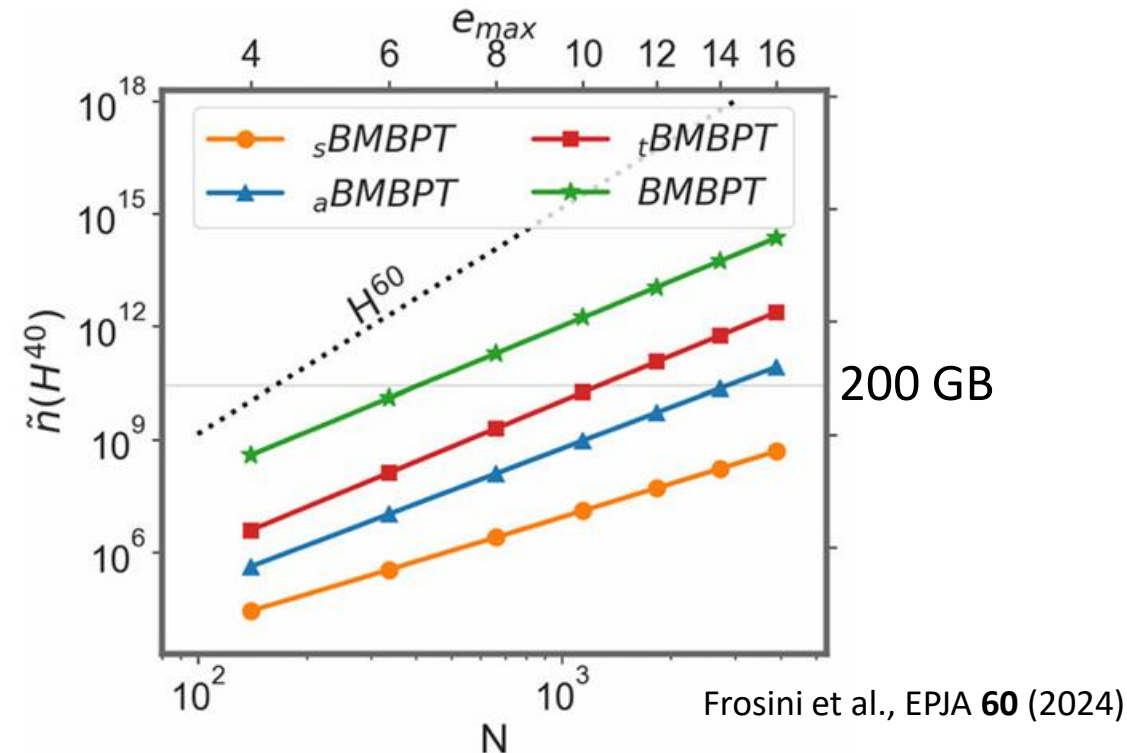
Duguet, Signoracci, JPG **44** (2016) Tichai et al., PLB **786** (2018) Arthuis et al., CPC **240** (2019)

- Split nuclear Hamiltonian: $H = \underbrace{H^{00} + H^{11}}_{\text{Exact (HFB)}} + \underbrace{H^{22} + H^{31} + H^{13} + H^{40} + H^{04}}_{\text{via perturbative corrections}}$
- 3N interaction treated through rank reduction Frosini et al., EPJA **57** (2021)

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- 3N interaction treated through rank reduction Frosini et al., EPJA **57** (2021)
- First correction to HFB ground state energy requires $H_{k_1 k_2 k_3 k_4}^{40}$



Randomized SVD

Halko et al., SIREV **53** (2011)
Martinsson, Tropp, ActaNum **403** (2020)
Tropp, Webber, 2306.12418 (2023)

- Decompose H^{40} despite it being too big for memory
- Lanczos-type algorithm to find largest singular values
- Number of necessary singular values determined on the fly using stochastic estimator of decomposition quality
Tropp, Webber, 2306.12418 (2023)
- Based on matrix-vector products

Implicit product

Frosini et al., EPJA **60** (2024)

- To circumvent construction and storage of $H_{k_1 k_2 k_3 k_4}^{40}$ consider instead only “matrix-vector” products

$$\sum_{k_3 k_4} H_{\underline{k_1 k_2} \underline{k_3 k_4}}^{40} \underline{X_{k_3 k_4}^{02}}$$

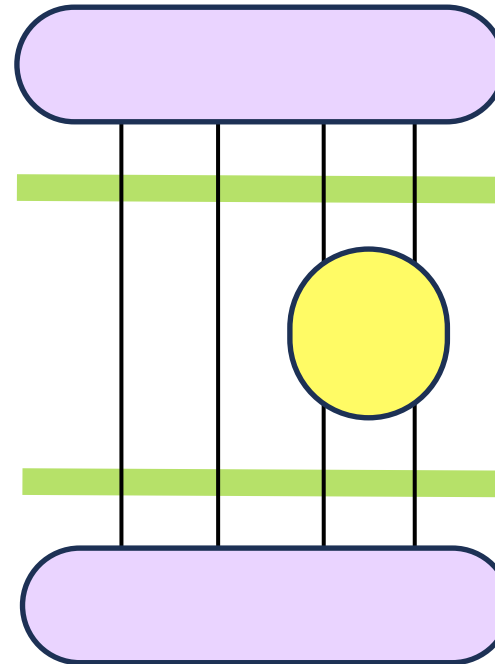
- Efficient implicit FAM-like products Carlsson et al., PRC **86** (2012)
 1. transform X^{02} to underlying spherical HO basis
 2. calculate product in that basis
 3. transform back

BMBPT(3)

- Next correction to HFB ground state energy:

$$e_{\text{BMBPT}}^{(3)} = \frac{1}{8} \sum_{k_1 \dots k_6} \frac{H_{k_1 k_2 k_3 k_4}^{04} H_{k_3 k_4 k_5 k_6}^{22} H_{k_5 k_6 k_1 k_2}^{40}}{(E_{k_1} + E_{k_2} + E_{k_3} + E_{k_4})(E_{k_5} + E_{k_6} + E_{k_1} + E_{k_2})}$$

- Evaluation of energy scales as N^6



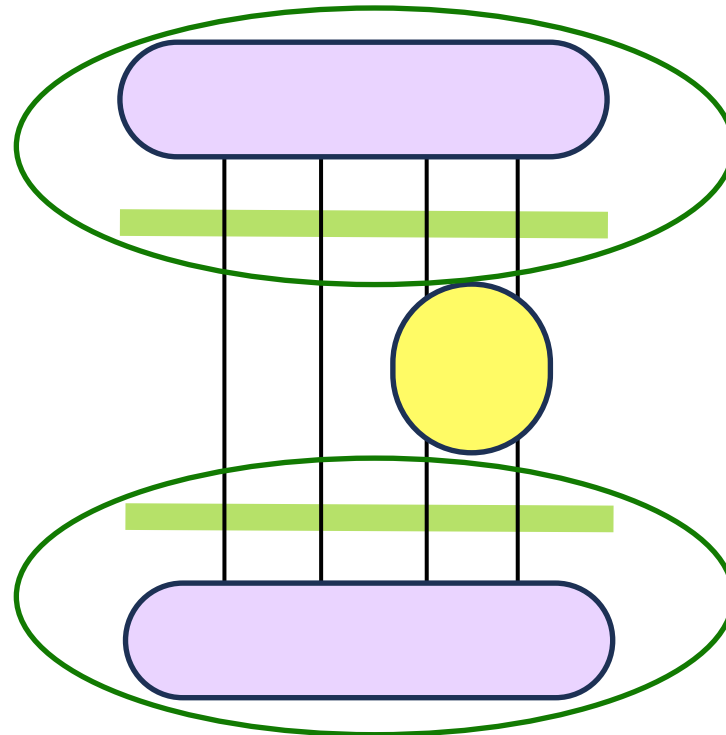
BMBPT(3)

- Next correction to HFB ground state energy:

$$-C_{k_5 k_6 k_1 k_2}^{40}(2)$$

$\equiv$

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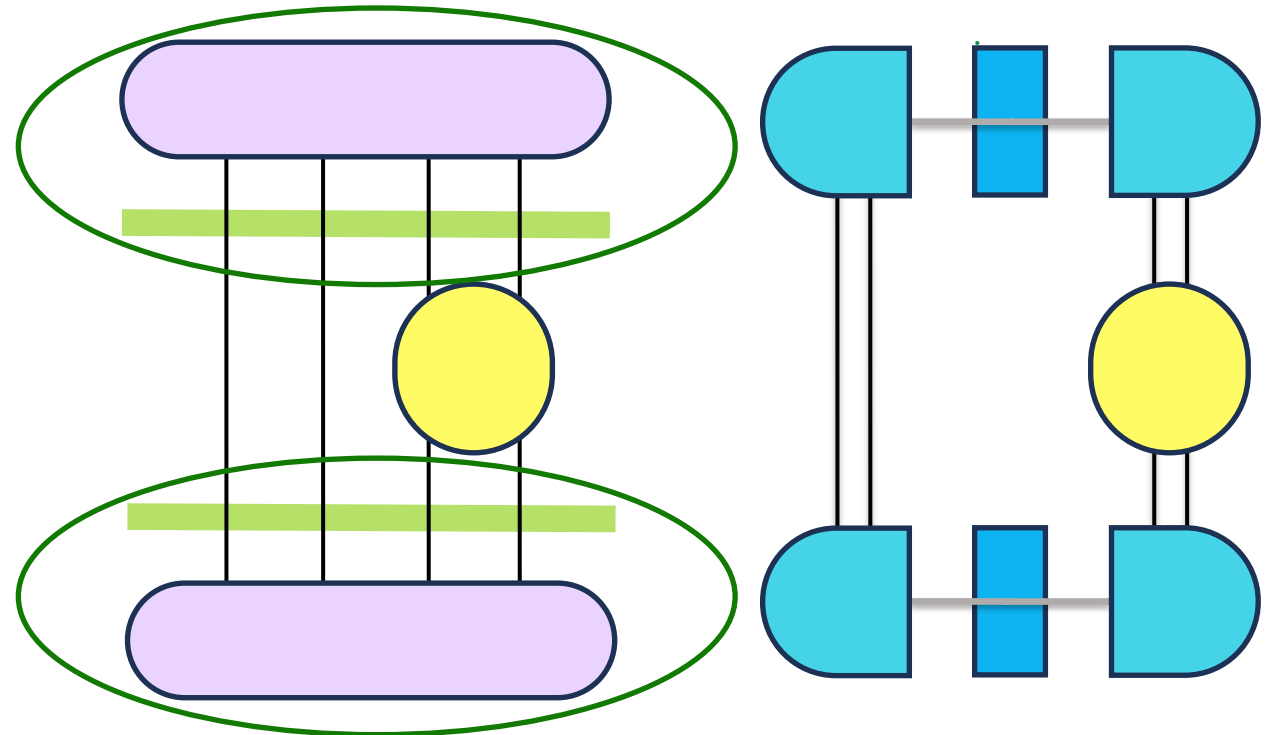
BMBPT(3)

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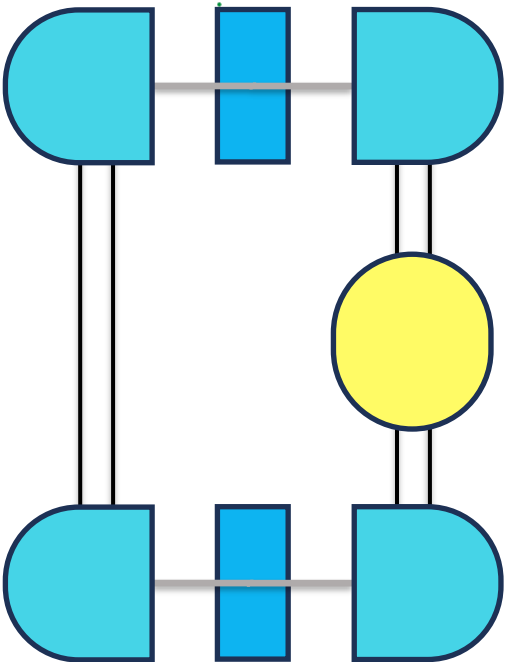
$$C_{k_1 k_2 k_3 k_4}^{40}(2) \approx \sum_{\mu=1}^r F_{k_1 k_2}^{\mu} s_{\mu} G_{k_3 k_4}^{\mu}$$



SVD-BMBPT(3)

- Project H^{22} to both sides on subspace spanned $C^{40}(2)$ by singular vectors

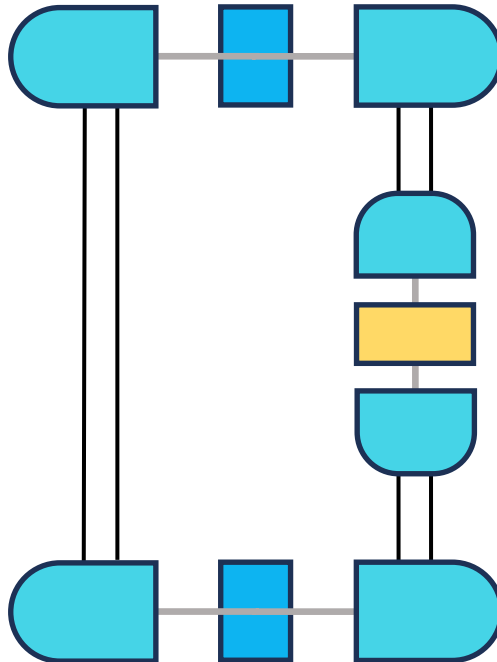
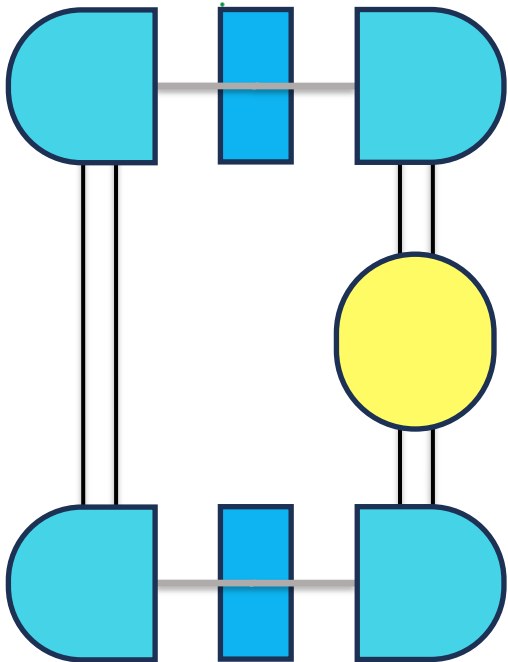
$$s'_{\mu\nu} = \sum_{k_1 \dots k_4} F_{k_1 k_2}^{\mu*} H_{k_1 k_2 k_3 k_4}^{22} F_{k_3 k_4}^{\nu}$$



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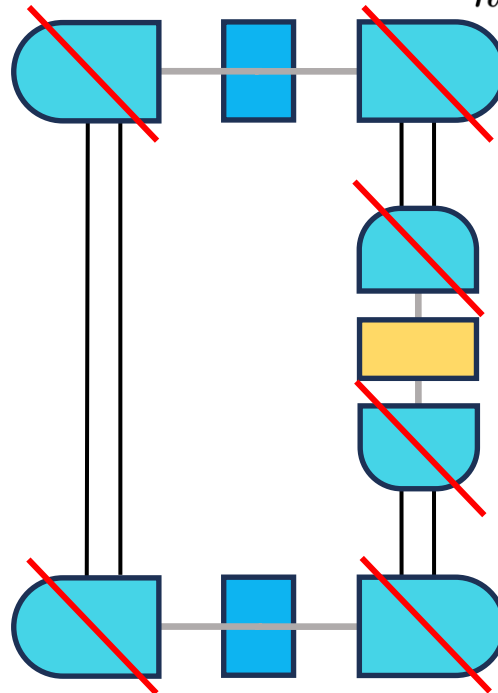
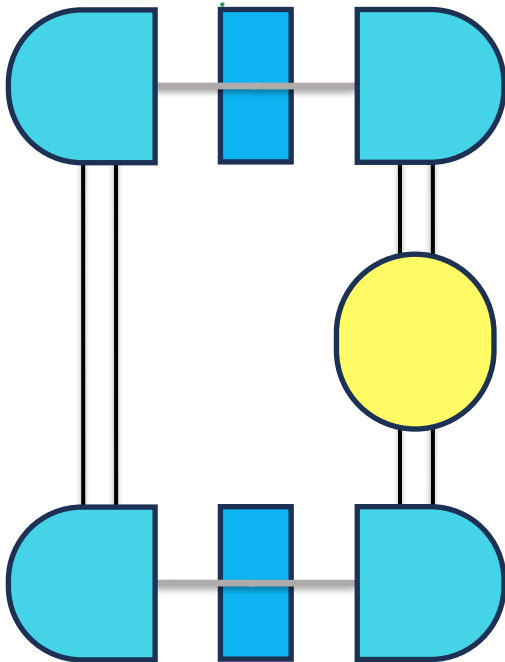


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- Singular vectors form unitary matrices $\sum_{k_1 k_2} F_{k_1 k_2}^{\mu*} F_{k_1 k_2}^{\nu} = \delta_{\mu\nu}$

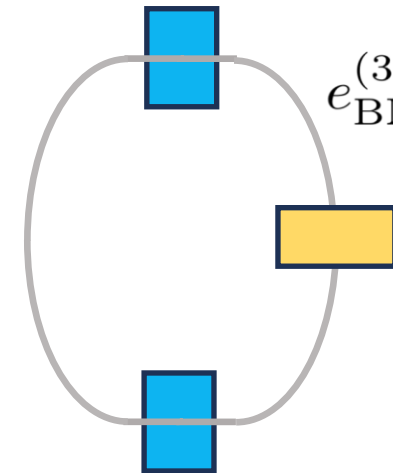
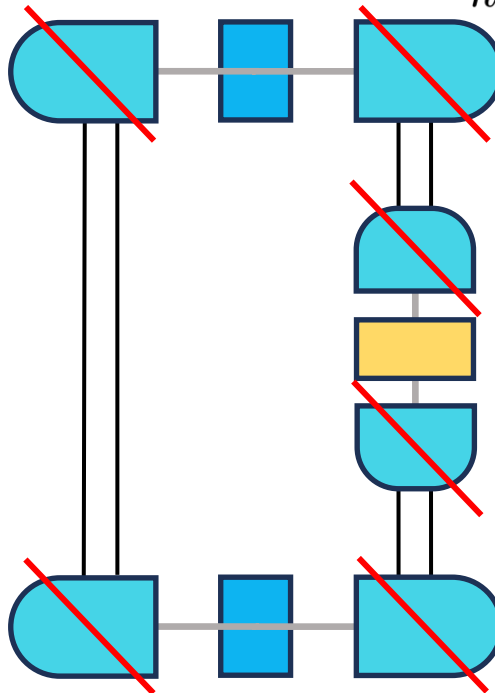
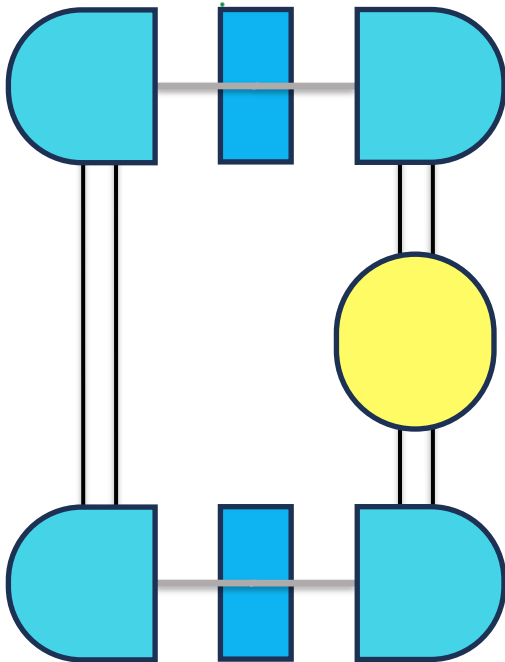


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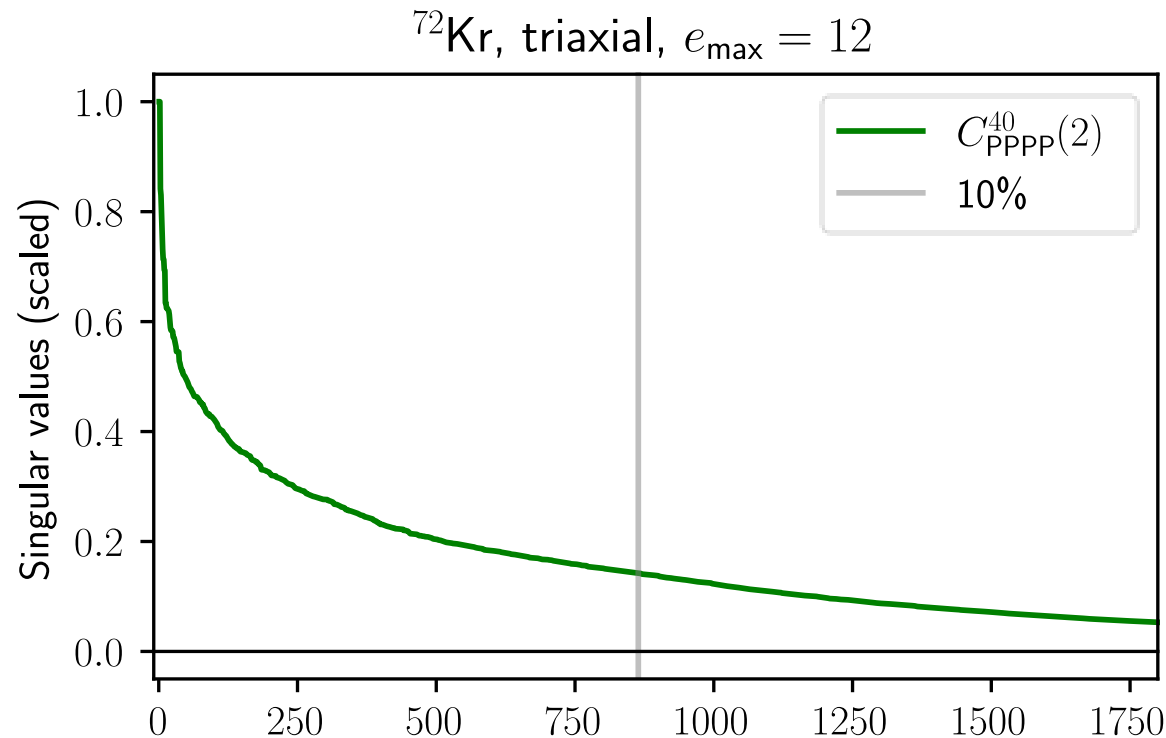
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SVD-BMBPT(3)

- Memory usage: $\mathcal{O}(rN^2) \leq \mathcal{O}(N^4)$
- CPU time: $\mathcal{O}(rN^4) \leq \mathcal{O}(N^6)$

SVD-BMBPT(3)

- Memory usage: $\mathcal{O}(rN^2) \ll \mathcal{O}(N^4)$
- CPU time: $\mathcal{O}(rN^4) \ll \mathcal{O}(N^6)$
- Need only $r \sim 1000$ singular values to reach desired accuracy

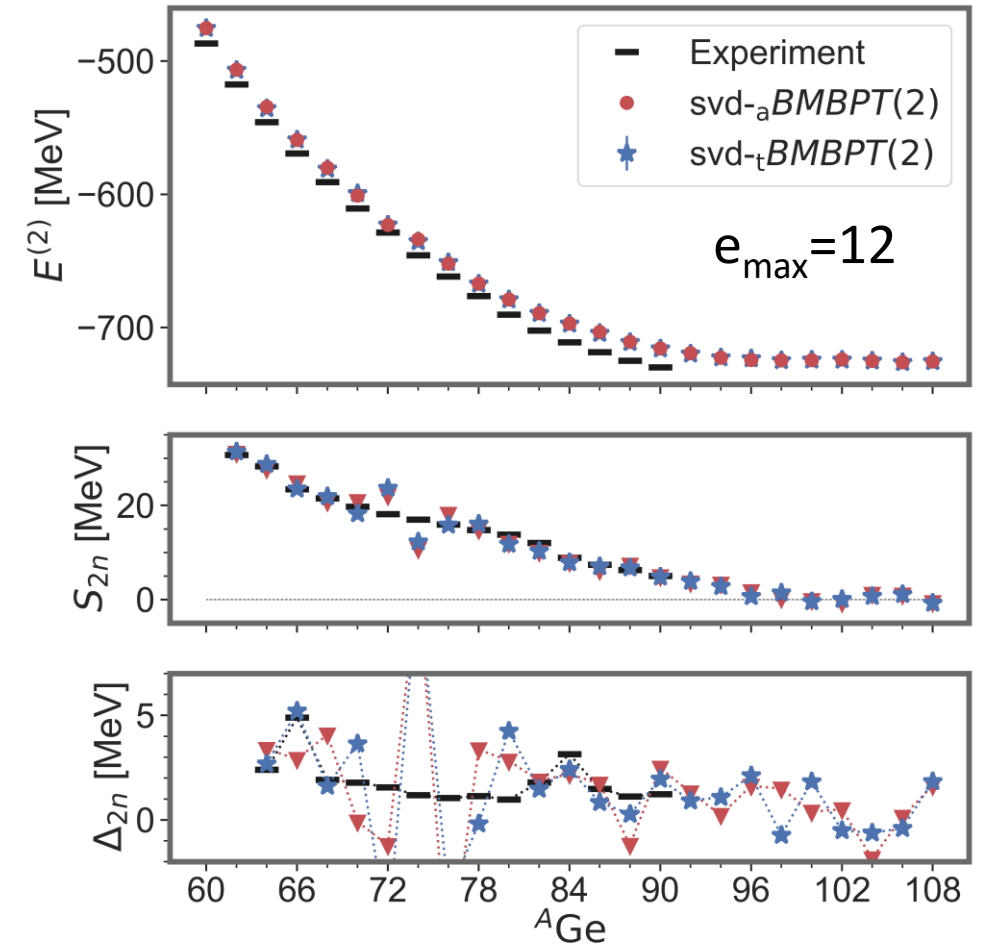


Full rank: $8 \cdot 10^5$

Summary

- SVD selects efficient subspace to compute BMBPT corrections
- Dimensionality reduction through tensor factorization makes previously unfeasible ab initio calculations possible

Frosini, ..., Zurek, EPJ Conf **302** (2024)



Outlook

- Particle number constraint at BMBPT(3) level
- Ideas can be applied to other many-body methods

Demol et al., AOP **424** (2021)

Demol et al., EPJA **61** (2025)

Parrish et al., JChemPhys **150** (2019)

Lesiuk, JCTC **18** (2022)

	BMBPT(2)	BMBPT(3)	BMBPT(4)
Traditional	$O(N^5)$	$O(N^6)$	$O(N^8)$
Tensor factorized	$O(r N^4)$	$O(r N^4)$	?

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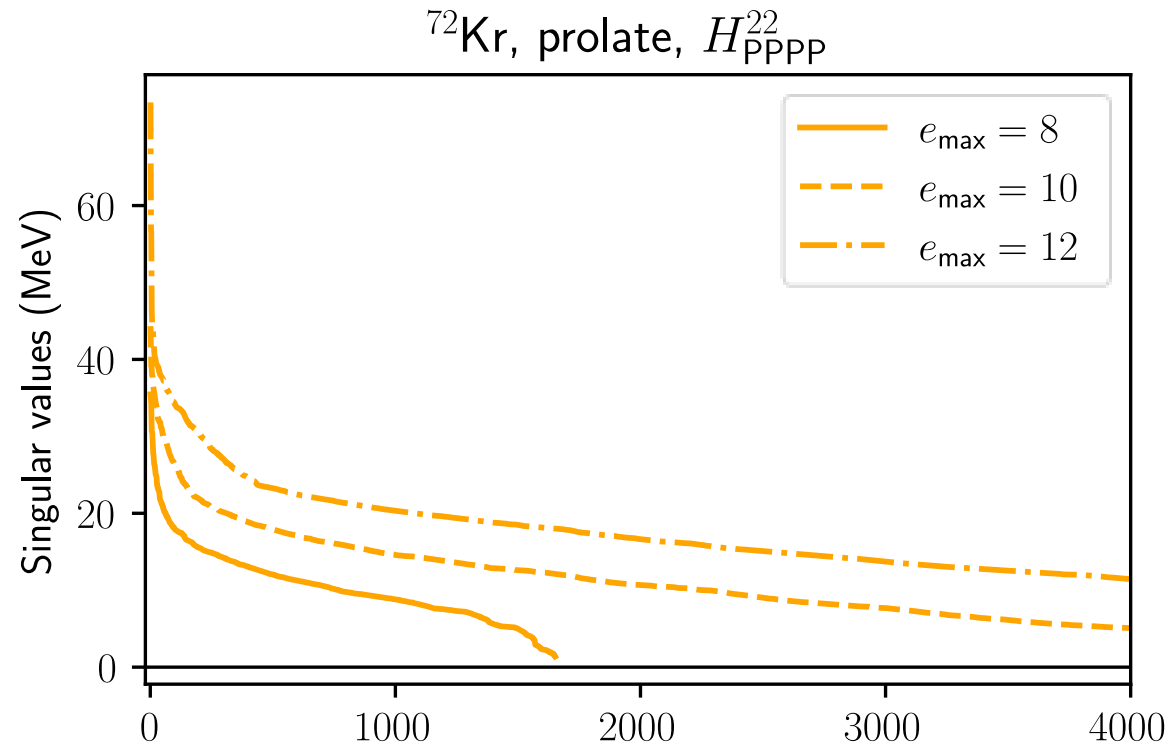
Thanks for your attention

and to

Thomas Duguet, Jean-Paul Ebran, and Mikael Frosini

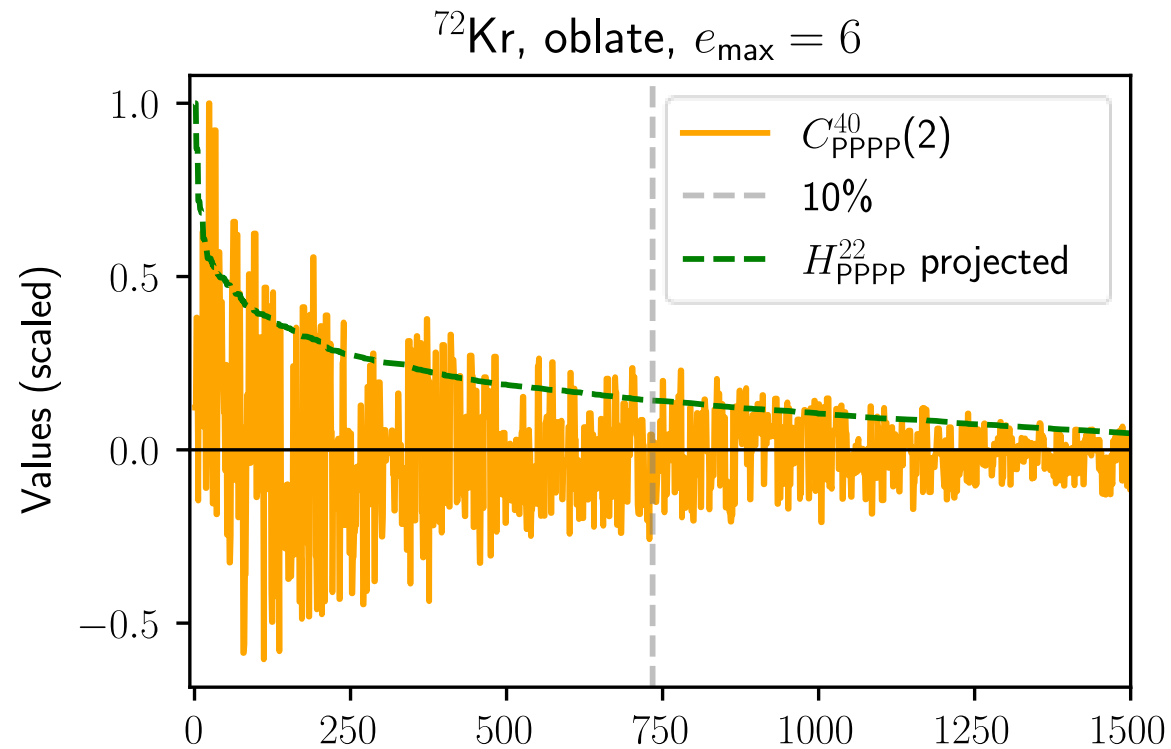
Singular spectra

- The spectrum of H^{22} does not converge with increasing $e_{\max}$



Singular spectra

- H^{22} projected on subspace spanned $C^{40}(2)$ by singular vectors falls off



Required accuracy

- BMBPT as formal power series

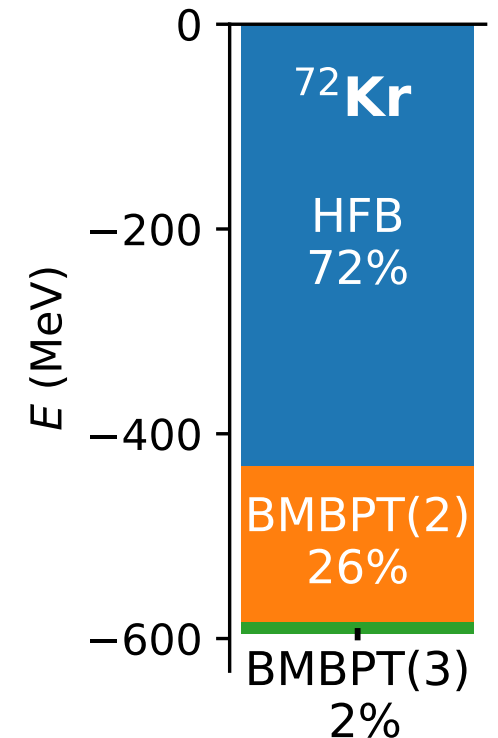
$$E^{[P]} = \langle \Phi | H \sum_{p=0}^{P-1} (R_0 \Omega_1^{(P)})^p | \Phi \rangle_c$$

Demol et al., EPJA **61** (2025)
Svensson et al., 2507.09079

- Conservative estimate of missing next-order contribution

$$\left| e_{\text{BMBPT}}^{(4)} \right| \sim \frac{1}{10} \left| e_{\text{BMBPT}}^{(3)} \right| \sim \frac{1}{100} \left| e_{\text{BMBPT}}^{(2)} \right|$$

→ Defines calculation accuracy goal at BMBPT(3) level



Required accuracy: needed basis size

- Need $e_{\max} \sim 14$ for BMBPT(2), ~ 10 for BMBPT(3)

