WOUTER RYSSENS



Nuclear Astrophysics

from the microscopic description of nuclei to the macroscopic explosion of stars

W. Ryssens and S. Goriely

April 21st 2023



wryssens@ulb.be

wryssens.com

(id)

Andy Sproles, ORNL





Graphic created by Jennifer Johnson http://www.astronomy.ohio-state.edu/~jaj/nucleo/ Astronomical Image Credits: ESA/NASA/AASNova

The nuclear chart...



The nuclear chart...



The nuclear chart and the processes traversing it



Different <u>isotopes</u>



Masses, reaction and decay rates



Temperature, densities,



Simulating **one** nucleus



The nuclear many-body problem

- quantum N-body problem
- ... for A~10 300 particles
- ... with Coulomb interaction
- ... with strong nuclear interaction



Simulating **one** nucleus





The nuclear many-body problem

- quantum N-body problem
- ... for A~10 300 particles
- ... with Coulomb interaction
- ... with strong nuclear interaction

Density Functional Theory with MOCCa

• <u>non-linear</u> optimisation problem

Simulating **one** nucleus



The nuclear many-body problem

- quantum N-body problem
- ... for A~10 300 particles
- ... with Coulomb interaction
- ... with strong nuclear interaction

Typical job: 0.3-1h, < 2GB, 1 CPU Extreme job: 6 - 7h , ~ 4GB, 1 CPU On LUMI: ?



Nuclear Density Functional Theory

- <u>non-linear</u> optimisation problem
- 3D coordinate space representation
- need to explore different shapes

Simulating **7000** nuclei: building models

Fit to experimental data:

- 2457 nuclear masses
- 884 charge radii
- 45 fission barriers

Optimisation problem:

- high-dimensional
- function evaluation ex. costly
- highly non-linear



Simulating **7000** nuclei: building models

Fit to experimental data:

- 2457 nuclear masses
- 884 charge radii
- 45 fission barriers

Optimisation problem:

- high-dimensional
- function evaluation ex. costly
- highly non-linear

Machine learning:

- committee of neural networks
- each emulates MOCCa
- efficient exploration of parameter space



Simulating **7000** nuclei: building models

Fit to experimental data:

- 2457 nuclear masses
- 884 charge radii
- 45 fission barriers

Optimisation problem:

- high-dimensional
- function evaluation ex. costly
- highly non-linear

Machine learning:

- committee of neural networks
- each emulates MOCCa
- efficient exploration of parameter space

BSkG3: 1.5 years of development, ~2x10⁶ CPU hours! 25 model parameters

properties of 2457 nuclei

Reaction networks

Typical job: 4 days, ~ 1GB, 1 CPU



One "trajectory"

- ~ 5000 stiff, coupled, first-order ODEs
- one set of astrophysical conditions
- variable runtime

Reaction networks

Typical job: 4 days, ~ 1GB, 1 CPU



One "trajectory"

- ~ 5000 stiff, coupled, first-order ODEs
- one set of astrophysical conditions
- variable runtime

Realistic simulations ~ 1k-2k trajectories

- different colliding stars
- ejecta with different temperatures/densities
- uncertainty study of nuclear models

Conclusion

From the **microscopic** description of **nuclei** and their **many** properties

- non-linear optimisation problem with 3D geometries
- large-scale parameter adjustments with neural networks
- large-scale calculations for the extraction of reaction rates



Conclusion

From the **microscopic** description of **nuclei** and their **m**

and their **many** properties

- non-linear optimisation problem with 3D geometries
- large-scale parameter adjustments with neural networks
- large-scale calculations for the extraction of reaction rates

.... to the **macroscopic** explosion of **stars**

- large reaction-network calculations
- for different thermodynamic conditions
- for different nuclear models



Conclusion

From the **microscopic** description of **nuclei** and their **many** properties

- non-linear optimisation problem with 3D geometries
- large-scale parameter adjustments with neural networks
- large-scale calculations for the extraction of reaction rates to the **macroscopic** explosion of **stars**
 - large reaction-network calculations
 - for different thermodynamic conditions
 - for different nuclear models

... to find out where the **elements** come from!



Thank you

..... to all our Belgian collaborators!



<u>S. Goriely</u>, <u>N. Chamel</u>, <u>M. Godefroid</u>, <u>S. Van Eck</u>, G. Grams, L. Batail, N. Shchechilin, L. Perot, V. Allard, A. Choplin, S. Martinet, R. Giribaldi, J. De Prince ... for the computing time!

...and the patient user support!



KU LEUVEN	<u>P. Van Duppen, T. Cocolios, R. Raabe, G.</u> <u>Nevens, R. De Groote</u>
------------------	--

UMONS <u>P. Quinet</u>, <u>P. Palmeri</u>

..... to our international partners!

..... for the funding!



finis fivo

..... for your attention!