# Extreme quantum mechanics in MATLAB 

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## What is quantum mechanics?

- A mathematical description of reality at small scales
- All that is knowable about a system is contained in one function
- Squared amplitude of that function is probability density


Erwin
Schrödinger


Paul
Dirac

antimatter


## What is spin?

- A relativistic symmetry that gives charged particles a magnetic moment
- No classical analogue, mathematically similar to angular momentum
- Responsible for most magnetic properties of matter



## Magnetic resonance industry


"By 2020, nuclear magnetic resonance (NMR) spectrometry is likely to lead the spectroscopy market in terms of annual revenue, which is expected to reach approximately $\$ 1.7$ billion worldwide."

Transparency Market Research Report, Aug 2014

"Estimates suggest that the global MRI market was worth about $£ 4.3$ billion in 2010 and is expected to grow to around $£ 6.2$ billion by 2015, equivalent to an annual growth of $7.7 \%$ a year."

Oxford Economics Report, Nov 2012


Birds use a spin-selective chemical reaction to sense the direction of the Earth's magnetic field.

DOI: 10.1021/bi060330h


Spin is at the core of every magnetic interaction and every technology that uses magnetism.

DOI: 10.1063/1.2010287


molecular structure

## Magnetic resonance simulations

quantum degrees of freedom
$\hat{\rho}$
$\hat{H}$
matrix of probabilities
matrix of energies
sizeable ( $\mathrm{dim}>10^{3}$ ) and very
sparse complex matrices

## equation of motion

$\frac{\partial}{\partial t} \hat{\rho}(t)=-i[\hat{H}(t), \hat{\rho}(t)]$
Liouville - von Neumann equation

> (basically Schrödinger equation for ensembles)
classical degrees of freedom

| $\{x, y, z\}$ | coordinates |
| :---: | :---: |
| $\{\alpha, \beta, \gamma\}$ | orientations |
| $\{a, \varphi\}$ | radio waves |

$\left\{\varphi_{1}, \varphi_{2}, \ldots\right\} \quad$ sample spinning

+ conformations, concentrations, etc.


## equation of motion

$\frac{\partial}{\partial t} p(\ldots)=-\operatorname{div}[\operatorname{flux}[p(\ldots)]]$
Fokker-Planck equation
(probability balance with a great number of special cases)
spatial derivative operators are turned into matrices using finite difference approximations


Principal problem: sum-of-direct-products-times-a-vector (can just about store the vector).

## Polyadic object in MATLAB

## A sum of matrix krons acting on a vector:

## $(\alpha[\mathbf{A} \otimes \mathbf{B} \otimes \ldots]+\beta[\mathbf{C} \otimes \mathbf{D} \otimes \ldots]+\ldots) \mathbf{x}$ <br> 

A short sum of krons of small matrices! Times a vector...

$$
\begin{aligned}
& \operatorname{dim}[\mathbf{A}]=1000 \\
& \operatorname{dim}[\mathbf{B}]=1000 \\
& \operatorname{dim}[\mathbf{A}(\mathrm{x}) \mathbf{B}]=10^{6} \\
& \text { numel }(\mathbf{v})=10^{6} \\
& \operatorname{dim}[\mathbf{V}]=1000
\end{aligned}
$$

[^0]```
% Bring forward n-th dimension
dims=1:numel(x_dims); dims(n)=[];
dims=[n,dims]; x=permute(%,dims);
% Unroll other dimensions
=reshape(%),[col_dims(n), numel(x)/col_dims(n)]);
% Run multiplication and update dimension map
x=Q{nmats-n+1}*x; x_dims(n)=row_dims(n);
% Roll other dimensions back up
=reshape(full(x),[row_dims(n),x_dims(dims(2:end))]);
% Put the current dimension back
=ipermute x, dims);
>> A=randn (1000); B=randn(200); C=randn (500);
>> H=polyadic ({{A,B,C}})
H =
    100000000\times100000000 polyadic array with properties:
    cores: {{1\times3 cell}}
    prefix: {}
    suffix: {
>> cheap_norm(H)
ans =
    5.4739e+07
>> tic; (H+H')*rand(1e8,1); toc
Elapsed time is 9.662600 seconds
```


## Polyadic object in MATLAB

Addition is implemented as buffering of terms:

$$
\mathbf{A} \otimes \mathbf{B}+\mathbf{C} \otimes \mathbf{D} \otimes \mathbf{E}+\ldots \quad \Leftrightarrow \quad\{\{\mathbf{A}, \mathbf{B}\},\{\mathbf{C}, \mathbf{D}, \mathbf{E}\}, \ldots\}
$$

The buffer is replayed every time an action on a vector is needed:

$$
(\alpha[\mathbf{A} \otimes \mathbf{B} \otimes \ldots]+\beta[\mathbf{C} \otimes \mathbf{D} \otimes \ldots]+\ldots) \mathbf{x}=\alpha[\mathbf{A} \otimes \mathbf{B} \otimes \ldots] \mathbf{x}+\beta[\mathbf{C} \otimes \mathbf{D} \otimes \ldots] \mathbf{x}+\ldots
$$

The same applies to pre- and post-multiplication:

\[

\]

| Wall clock time, <br> polyadic rep | Wall clock time, <br> explicit rep |
| :---: | :---: |
| $0.37 \pm 0.01 \mathrm{~ms}$ | $0.88 \pm 0.12 \mathrm{~ms}$ |
| $1.8 \pm 0.3 \mathrm{~ms}$ | Out of RAM |
| $97 \pm 14 \mathrm{~ms}$ | Out of RAM |
| $0.21 \pm 0.01 \mathrm{~ms}$ | $0.05 \pm 0.01 \mathrm{~ms}$ |
| $2.1 \pm 0.3 \mathrm{~ms}$ | $11.4 \pm 1.6 \mathrm{~ms}$ |
| $105 \pm 16 \mathrm{~ms}$ | Out of RAM |



## MRI simulation using polyadics

Result: arbitrary spatial dynamics with quantum mechanical description of spin.

A. Allami, M.G. Concilio, P. Lally, I. Kuprov, Science Advances, 2019, 5(7), eeaw8962.

## Spinach toolbox

- Magnetic resonance theory library for large-scale timedomain simulation work
- All types of magnetic resonance (NMR, EPR, MRI, DNP, PHIP, SQUID, etc.)
- Over 600 pages of docs and tutorials, over 100 real-life simulation examples
- Well-annotated open-source code, clear variable names, informative error messages
- Parallel processing, GPU support, tensor structured object support
- Over 50 developers and contributors, 12 years of full-time programming




## Parallelisation strategies

- Some strategies are trivial (parfor over ensembles, etc.)
- SPMD permits more sophisticated techniques


[^1]
## Nested switchable parallelisation

## Extended parfor syntax passes parallelisation opportunities down:

\% Powder averaging loop parfor ( $\mathrm{n}=1$ : numel (weights), nworkers)
\% Localise the parameter array
localpar=parameters;
\% Get the full Hamiltonian at the current orientation H=I+orientation (Q,[alphas(n) betas(n) gammas(n)]); H=(H+H')/2;


Table 1. Scaling behaviour of the parallel propagation algorithms

| Number of CPU <br> cores | Time steps per wall clock second |  |  |
| :---: | :---: | :---: | :---: |
|  | Algorithm A <br> (observable) | Algorithm B1 <br> (final state) | Algorithm B2 <br> (final state) |
| 1 | 1.2 | 3.1 | 1.9 |
| 2 | 2.5 | 6.2 | 3.7 |
| 4 | 4.9 | 12.5 | 7.4 |
| 8 | 9.9 | 25.1 | 14.8 |
| 16 | 18.9 | 49.7 | 29.8 |
| 32 | 29.4 | 72.7 | 48.1 |
| 64 | 48.4 | 112.8 | 78.6 |
| 128 | 68.0 | 151.7 | 110.9 |

## Writing a MATLAB package


H.J. Hogben, M. Krzystyniak, G.T.P. Charnock, P.J. Hore, I. Kuprov, Journal of Magnetic Resonance, 2011, 208(2), 179-194.

## Phosphine ligand chemistry

A di-tert-butylphosphine... 22 spins, meaning matrix dimension $2^{22}$ (very sparse).
Confused chemists - the pattern of NMR peaks is emphatically not textbook.



SPMD across 512 cores: 15 minutes, perfect match to the experiment.
Industrial gold extraction, etc.

## Complex sparse GPU arithmetic



Metal locations in metalloproteins using Tikhonov regularisation:

$$
\text { (least squares gradient) }=[2 \times F F T, 2 \times D G E M M]
$$

Gradient explicit, but Hessian implicit (hello, Optimisation Toolbox):
(Hessian-times-vector) $=[4 \times F F T, 4 \times$ DGEMM $]$

Quadruple 3D FFT of a $512 \times 512 \times 512$ dataset - over 2 GB of data!



## Complex sparse GPU arithmetic



A module for Spinach library:
> run time: minutes
$>$ good match to DEER data
> sensible match to MD data
> new use for old PCS data
> some structural insight


Relevant operations:
> matrix exponentiation
$>$ time propagation
> very large Fourier transforms
> element-wise operations

[^2]
## Molecular distance measurement

Signal modulation by dipolar coupling:



Straightforward theory for two point spin $1 / 2$ particles:

$$
\gamma(r, t) \sim \frac{1}{\sqrt{D t}} \cdot \operatorname{trig}(D t) \cdot \operatorname{fresnel}(\sqrt{D t}) \quad D \sim \frac{1}{r^{3}}
$$

Can extract distance distributions:

$$
\operatorname{DEER}(t)=\int_{0}^{\infty} p(r) \gamma(r, t) d r
$$



## Inverse QM with neural nets

Neural networks are surprisingly good at getting distance distributions:


ensemble statistics


10 hours of unattended training vs. 10 years of programming!

## Philosophical matters...



We managed to find out how a two-layer DEERNet works...


A black box neural network is not compatible with Descartes/Popper framework of science! issues: interpretability and trust


The net spontaneously evolved:

1. A bandpass filter $v s$ the noise
2. A notch filter $v s$ the baseline
3. A frequency axis rectifier from cubic to linear: $\omega \sim 1 / r^{3}$ in DEER, but the plot is linear.

All packed into a single linear (!) transform in one (!!) layer.


## Summary

- Simulations in spin physics are stupidly hard!
- They can be done with some tensor algebra tricks...
- ...but you need a language that understands tensor structures!
- ...which chemists know how to use
- ...that is compatible with version control
- ... and of which there is only one dialect.


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Ben Tordoff


[^0]:    + various technicalities: sums, products by scalars, etc.

[^1]:    L.J. Edwards, I. Kuprov, Journal of Chemical Physics, 2012, 136(4), 044108.

[^2]:    E.A. Suturina, D. Häussinger, K. Zimmermann, L. Garbuio, M. Yulikov, G. Jeschke, I. Kuprov, Chemical Science, 2017, 8(4), $2751-2757$.

