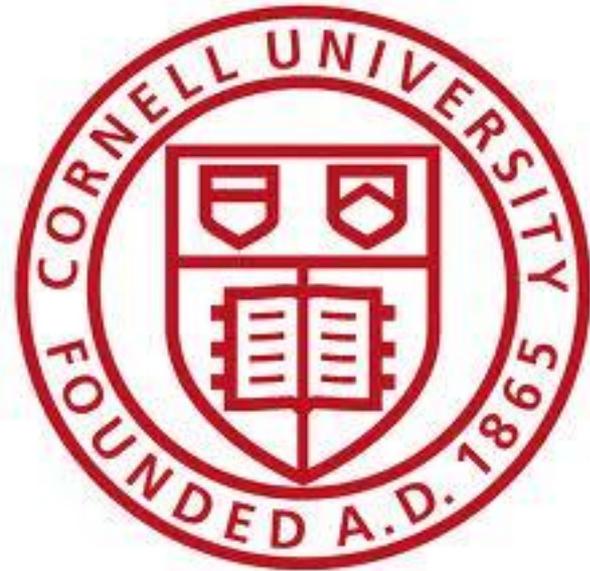


a graph algorithm for a new architecture
and
new algorithms for quantum state evolution

Claire C Ralph

thank you!

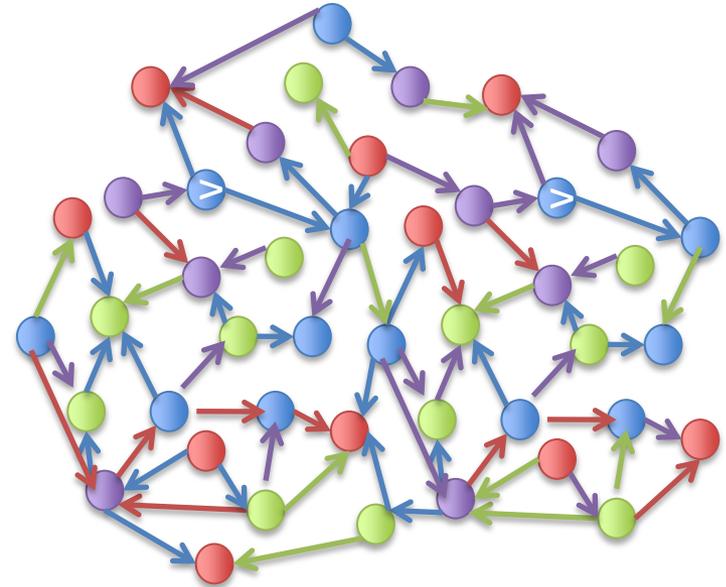


a graph algorithm for a new
architecture

high performance and graphs

graphs

- Data-driven applications
- Unstructured problems
- Poor locality
- High data access to compute ratio



traditional HP not optimal for graphs

New Architecture → New Techniques → New Algorithms

'Tolerate' not 'reduce' latency.

Load balancing not an issue.

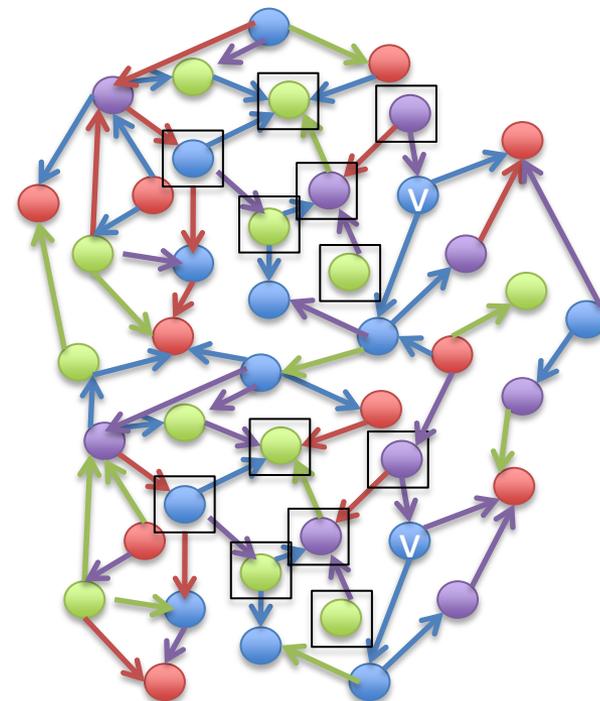
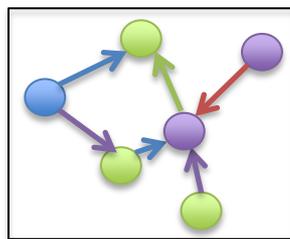
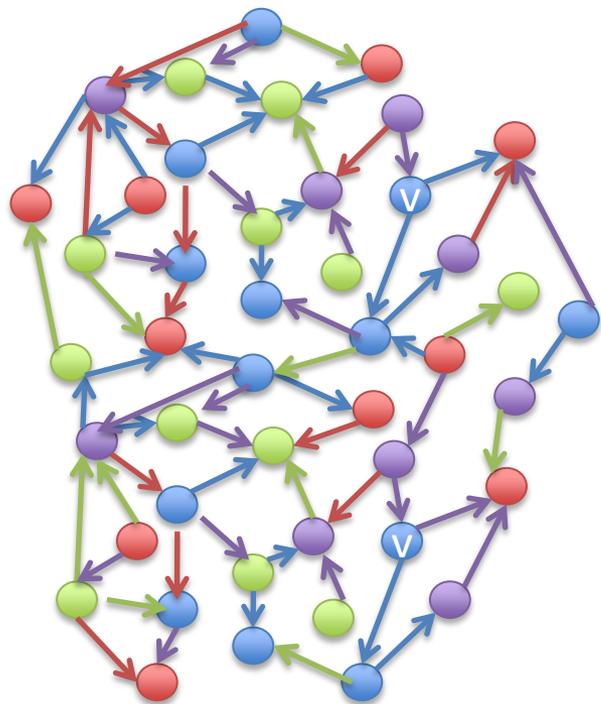
Hotspotting still a problem!

The Cray XMT

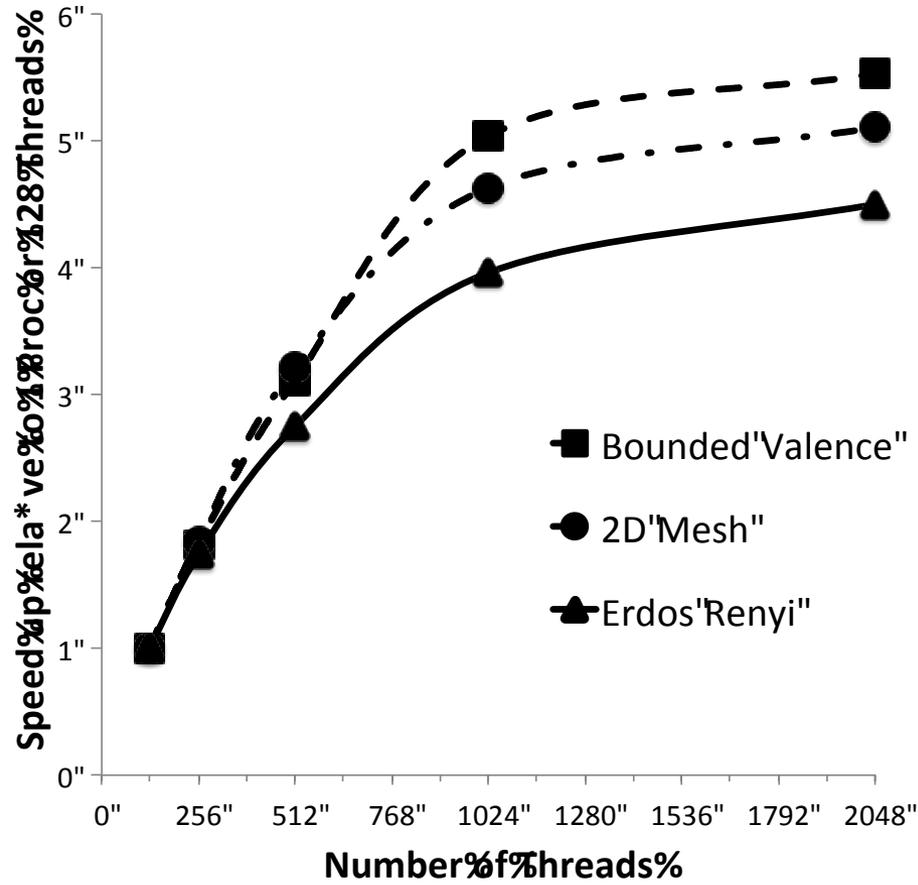
- 500 MHz single 64 bit Cray Threadstorm Processor
- 128 threads per processor
 - Single cycle switching
- 8GB shared memory/processor
- Automated dynamic scheduling
- Largest is 512 processors
 - 65,536 threads
- Largest SMP has 2560 cores
 - 4096 threads



Subgraph Isomorphism



Speed up of Final Algorithm



Scales to
1000
threads.

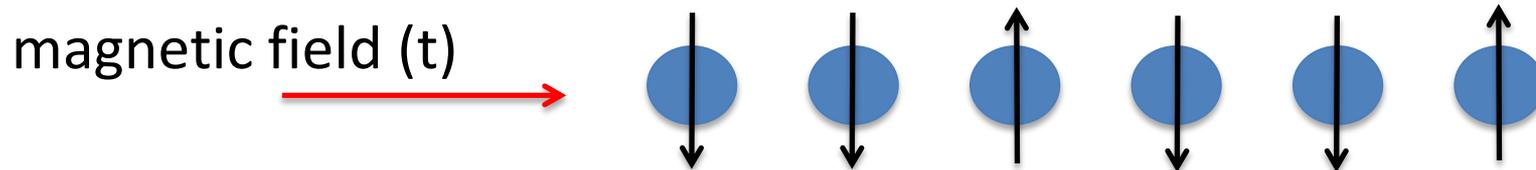
in summary

- developed a new graph algorithm
- got to work with a cool new architecture
- pretended to be a computer scientist for a couple of months

new algorithms for quantum state
evolution

time dependent Schrodinger Equation for a system of n spins

$$i \frac{\partial}{\partial t} \Psi(\sigma_1, \dots, \sigma_n) = \hat{H} \Psi(\sigma_1, \dots, \sigma_n)$$



$$\Psi_t = e^{-iH\Delta t} \Psi_o$$

a technical detail

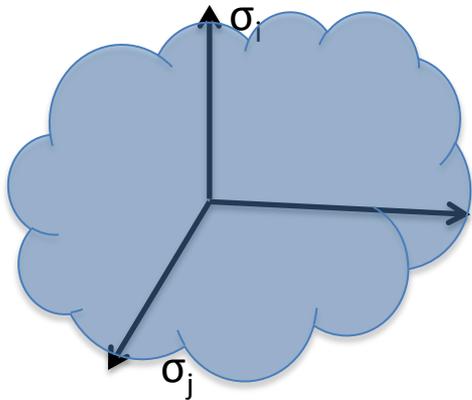
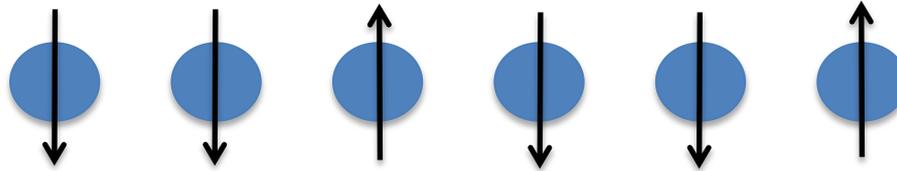
For a closed system:

$$\Psi_t = e^{-iH\Delta t} \Psi_o$$

For an open system:

$$\Gamma_t = e^{-iL\Delta t} \Gamma_o$$

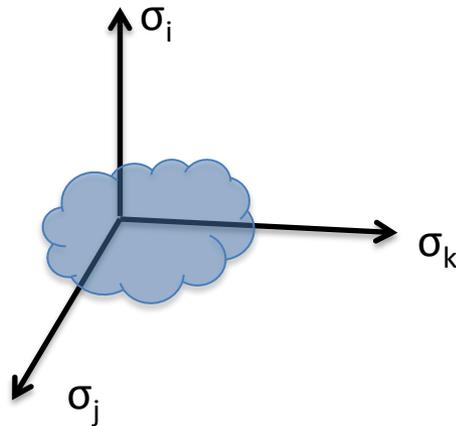
why so complicated?



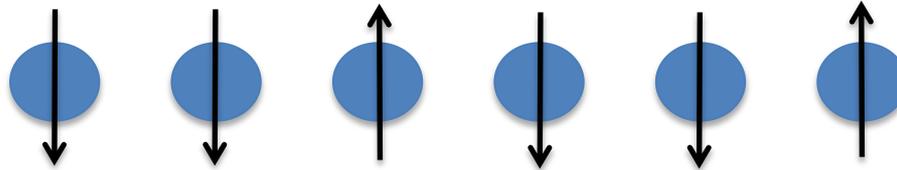
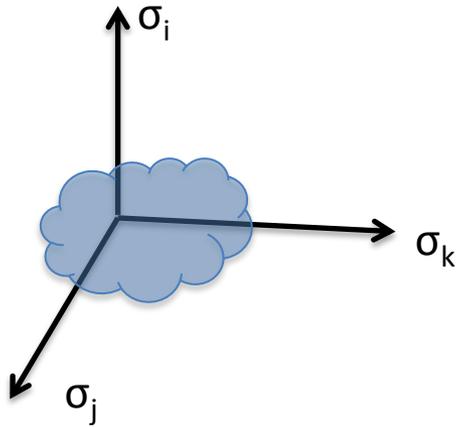
$$|\Psi\rangle = \sum_{\sigma} c_{\sigma_1 \dots \sigma_n} |\sigma_1 \dots \sigma_n\rangle$$

n spins \rightarrow tensor of dim 2^n

$$c^{\sigma_1 \dots \sigma_n}$$



strategies for approximation



1. state space restriction.

$$C^{\sigma_1 \dots \sigma_n}$$

tensor of dim 2^n is mostly zero

2. quantum monte carlo

not promising for this problem

3. low dimensional factorization

$$C^{\sigma_1 \dots \sigma_n} \rightarrow A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_n}$$

the matrix product state

$$A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_n}$$

We have a form, so:

yes for time independence

?? for dynamics

1. how do we find the elements that make up the 'best' product?
2. can we calculate properties (take expectation values) directly from this form?

a new algorithm: aMPS

1. how do we find the elements that make up the 'best' product?

derived EofM for individual matrices

Fast and
Accurate

linearize these using special structure

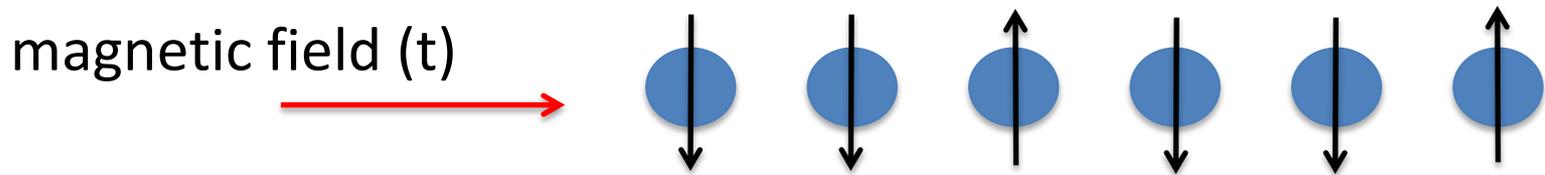
1. can we calculate useful properties from this form?

response functions and excitations



Adv. Chem. Phys. (2012)
arXiv:1103.2155v1 [cond-mat.str-el]

so we can use the MPS, but is it useful?



NMR

ESR

high temp limit



SSR



MPS

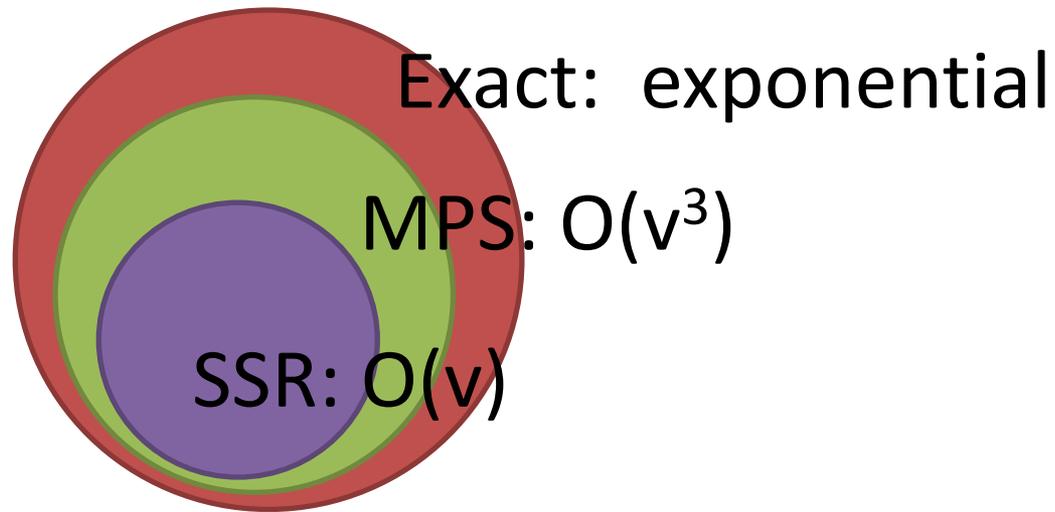
?

?

aMPS 'automatically' gives optimal basis

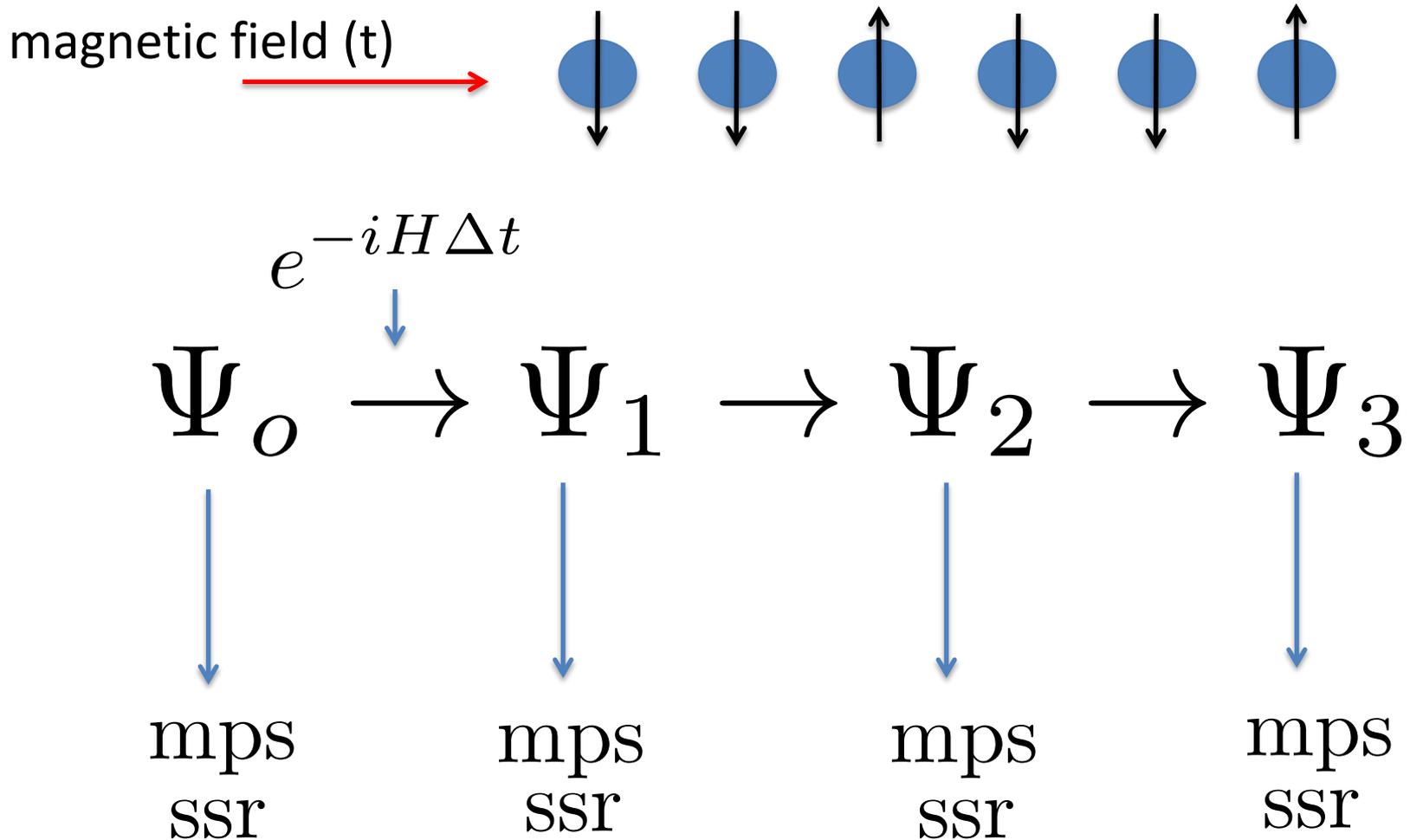
relationship between SSR and MPS: MPS covers the space of the SSR basis with $O(n^3)$ storage.

storage of SSR to MPS

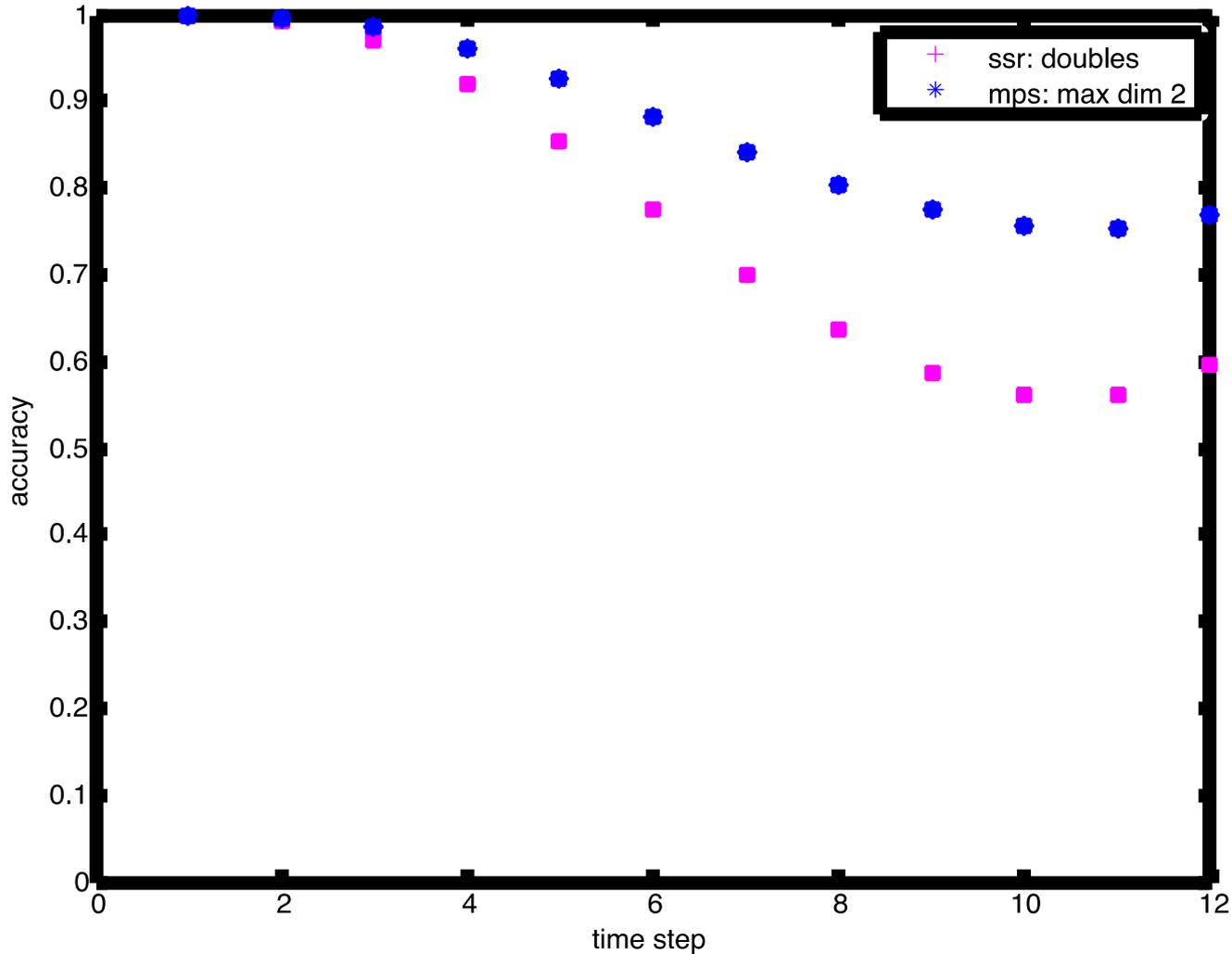


MPS guaranteed to do better than SSR
with polynomial increase in storage cost.

a model system

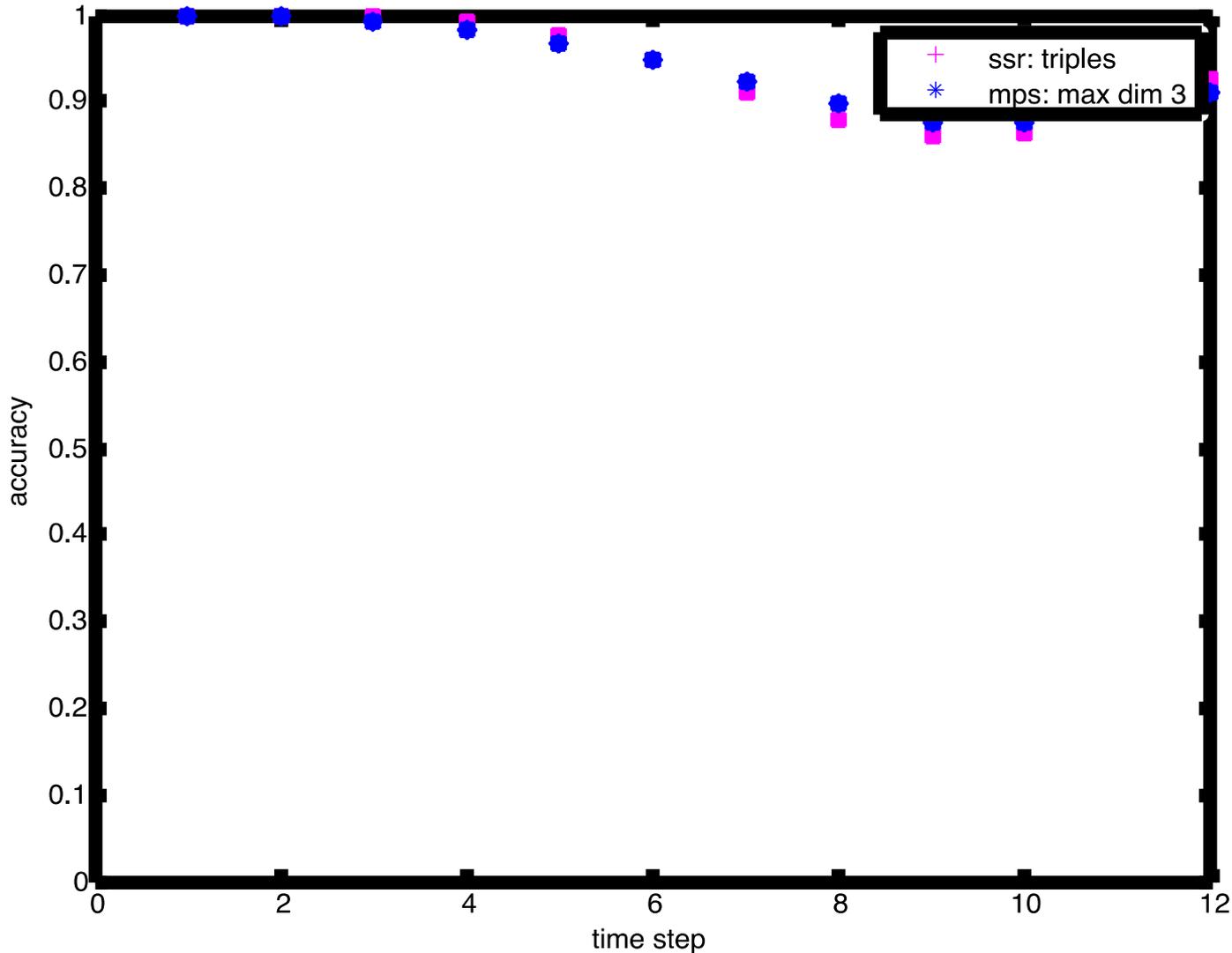


numerical comparison of SSR vs MPS



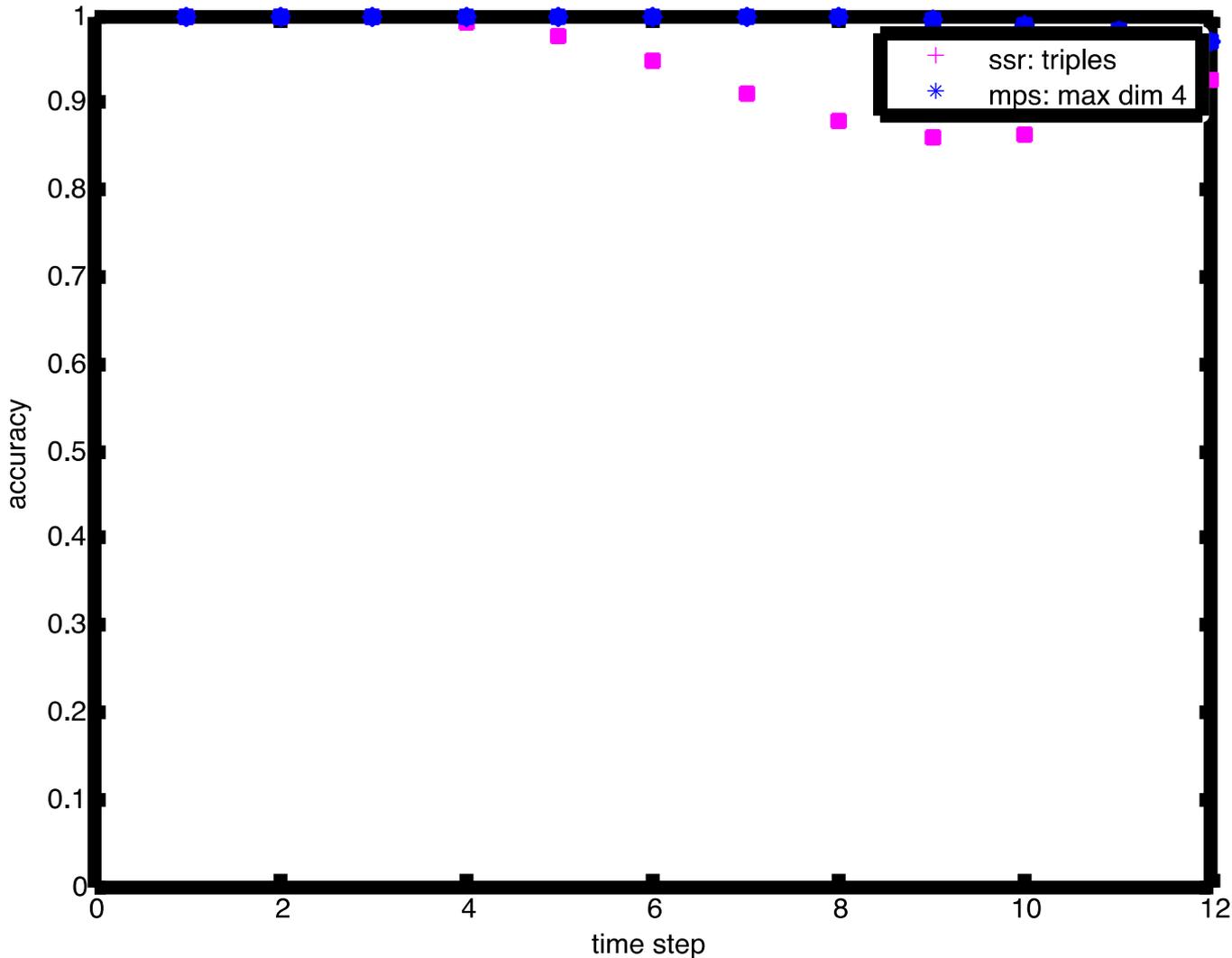
Ext	nelem	256
SSR	nelem	67
	nnz	16
MPS	nelem	48
	nnz	45

numerical comparison of SSR vs MPS



Ext	nelem	256
SSR	nelem	175
	nnz	40
MPS	nelem	96
	nnz	92

numerical comparison of SSR vs MPS



Ext	nelem	256
SSR	nelem	175
	nnz	40
MPS	nelem	160
	nnz	148

in summary

- derived a new time propagation scheme for a beloved wavefunction ansatz, MPS
- derived methods for computing important physical properties such as response functions
- analyzed relationship between MPS and SSR
- initial numerical results are promising

next step:

implementation of time propagation algorithm

thank you!

