# Graph Partitioning Methods for Fast Parallel Quantum Molecular Dynamics 

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## Talk outline

- Background and motivation of partitioning approach
- Quantum MD background
- Recursive polynomial expansion of Hamiltonian matrices
- Partitioned evaluation of matrix polynomials
- Formulation of the GP problem and its application
- CH-partitioning definition
- Application to matrix polynomial evaluation
- Correctness of approach
- Development of CH-partitioning algorithms
- Experimental analysis
- Conclusion

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## Quantum MD background

- Classical MD simulations
- Atoms as bodies that move based on Newton's laws of motion
- Forces between atoms calculated using interatomic potentials
- Positions of atoms updated in small time steps
- Interaction models use a priori knowledge of the system
- Cannot explain events on atomic and subatomic level
- Quantum MD simulations
- Based on laws of quantum mechanics
- Density functional theory (DFT) most used model
- Second-order spectral projection (SP2) approach
- Density matrix as a function $f$ of the Hamiltonian
- Representing $f$ as a recursive polynomial expansion

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## Recursive polynomial matrix expansion

- Given Hamiltonian $H$, compute density matrix $D$

$$
\begin{aligned}
& D=\lim _{n \rightarrow \infty} f_{n}\left(f_{n-1}\left(\ldots f_{0}(H) \ldots\right)\right) \\
& f_{0}(X)=\alpha I-\beta X \\
& f_{i}(X)= \begin{cases}X^{2}, & \text { if } \operatorname{Tr}[X]>N_{i} \\
2 X-X^{2}, & \text { otherwise }\end{cases}
\end{aligned}
$$

- The degree grows at an exponential rate, hence 20-30 iterations suffice
- Thresholding used to reduce MM complexity

$$
D=\lim _{n \rightarrow \infty} f_{n} t_{n}\left(\ldots f_{0} t_{0}(H) \ldots\right)
$$

## Parallel evaluation of matrix polynomial for $D$

- Large number of time steps $\left(10^{4}-10^{6}\right)$ - need parallelism
- Bottleneck operation $Y=X^{2}$ for a sparse matrix $X$
- Sparse matrix algebra
- Works well in sequential and shared-memory environment
- Speedup of distributed implementation goes down with the \# nodes due to communication overhead
- Partitioning based approach
- Computational overhead (total number of operations higher)
- Reduced communication overhead
- Scalable parallelism


## Partitioned evaluation

- Model the sparsity structure of $H$ by a graph $G=G(H)$
- Partition $G$ into (overlapping) graphs $G_{i}$
- core vertices of $G_{1}, \ldots, G_{p}$ form a partition of $V(G)$
- halo vertices are neighbors of core vertices \& not in the core
- CH-partitioning (core-halo)

- Send submatrix $H_{i}$ of $H$ defined by $G_{i}$ to node $i$
- Compute polynomial $P\left(H_{i}\right)$ by node $i$
- Copy core elements of $P\left(H_{i}\right)$ to $D:=P(H)$


## The CH -partitioning problem

- The partitioned algorithm correctly computes during the $i$-th iteration $D\left(H_{i}\right)$ assuming
- Time step is small enough so that density matrix does not change a lot in one iteration
- Graph used for partitioning is based on $\left(D_{i-1}+H_{i}\right)^{2}$
- Thresholding is used after each matrix computation
- CH-partitioning problem formulation:

Given an undirected graph $G$ and $q \geq 2$, find a partition $C_{1}, \ldots, C_{q}$ of $V(G)$ with corr. halos $H_{1}, \ldots, H_{q}$ that minimizes

$$
\sum_{i}\left(\left|C_{i}\right|+\left|H_{i}\right|\right)^{3} \text { (or, alternatively, } \max _{i}\left\{\left|C_{i}\right|+\left|H_{i}\right|\right\} .
$$

## Partitioning algorithms

- Standard graph partitioning
- Related, but different than CH-graph partitioning


Standard graph partitioning


- Solvers Metis, hMetis, KaHIP
- New algorithms
- Kernighan-Lin based
- Simulated annealing
- Metis+SA

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## Experimental setup

## - Test cases motivated by physical systems

| No. | Name | $n$ | $m$ | $m / n$ | Description |
| :--- | :--- | ---: | ---: | ---: | :--- |
| 1 | polyethylene dense crystal | 18432 | 4112189 | 223.1 | crystal molecule in water low threshold |
| 2 | polyethylene sparse crystal | 18432 | 812343 | 44.1 | crystal molecule in water high threshold |
| 3 | phenyl dendrimer | 730 | 31147 | 42.7 | polyphenylene branched molecule |
| 4 | polyalanine 189 | 31941 | 1879751 | 58.9 | poly-alanine protein solvated in water |
| 5 | peptide 1aft | 385 | 1833 | 4.76 | ribonucleoside-diphosphate reductase protein |
| 6 | polyethylene chain 1024 | 12288 | 290816 | 23.7 | chain of polymer molecule, almost 1-d |
| 7 | polyalanine 289 | 41185 | 1827256 | 44.4 | large protein in water solvent |
| 8 | peptide trp cage | 16863 | 176300 | 10.5 | small protein dissolved in $\mathrm{H}_{2} \mathrm{O}$ molecules |
| 9 | urea crystal | 3584 | 109067 | 30.4 | organic compound |

## Test matrices



Phenyl dendrimer system with its molecular representation (left)

2D plot representation of the Hamiltonian (middle)
Thresholded density matrix (right)

## Comparison of accuracies



## Comaprison of running times



## QMD running time comparison



## Conclusion

- New graph partitioning problem with applications in materials science and sparse matrix polynomials
- Parts overlap
- Objective function not directly related to edge cut
- Several implementations
- Classical GP algorithms + SA postprocessing
- KaHIP+SA gives best quality
- Metis+SA best running time and best overall
- Parallel QMD implementation based on CHP runs about 10 times faster than SM based version

