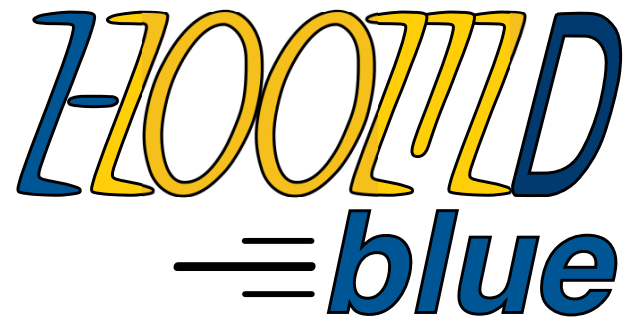
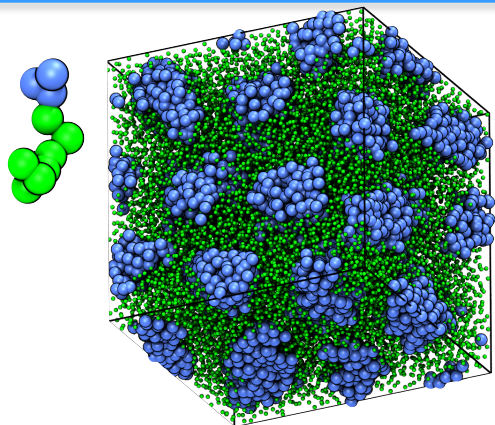




- Fast and flexible many-particle dynamics on the GPU

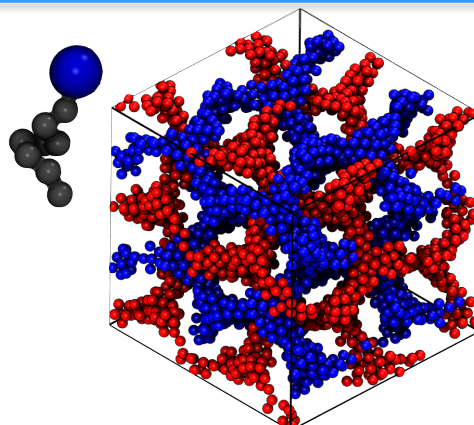
Joshua A. Anderson

Many-particle dynamics - examples



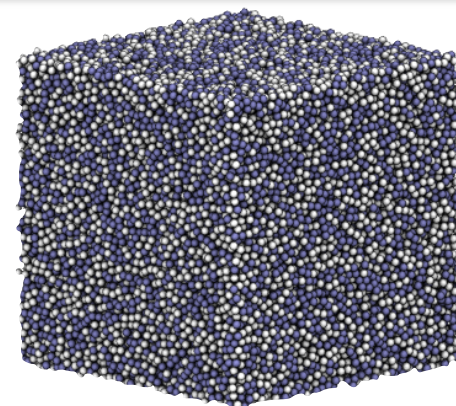
polymer systems

Dissipative Particle Dynamics (DPD)



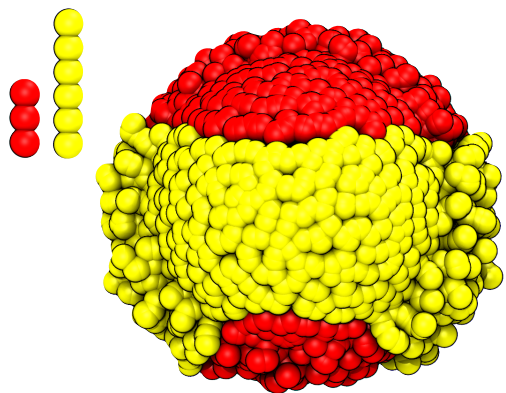
tethered nanospheres

Brownian Dynamics



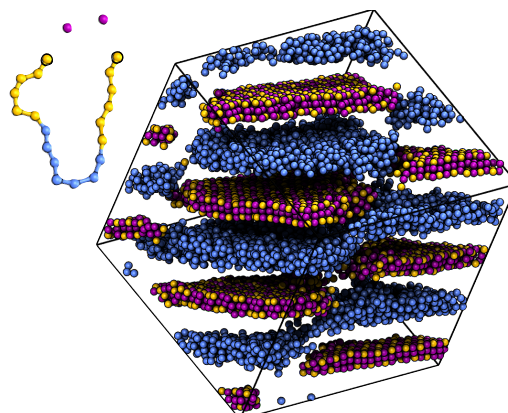
supercooled liquids

Molecular Dynamics



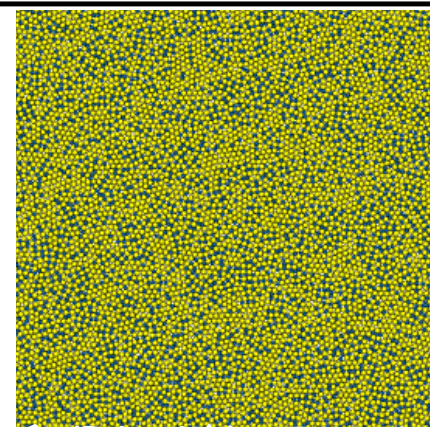
surfactant coated surfaces

DPD with constraints



polymer nanocomposites

coarse-grained Molecular Dynamics



supercooled liquids

2D Molecular Dynamics

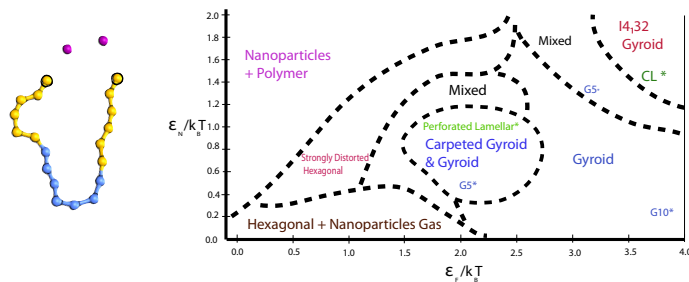
Live demo

Demo of HOOMD-blue outside of presentation

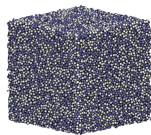
Benefits of GPU computing

Papers with CPU jobs...

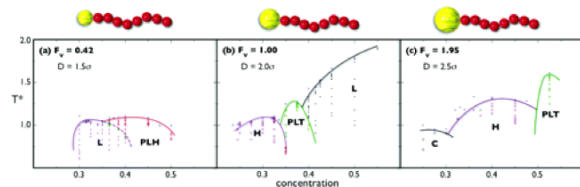
run thousands of serial jobs - often **one month** of CPU time for each



compute a phase diagram for one polymer architecture



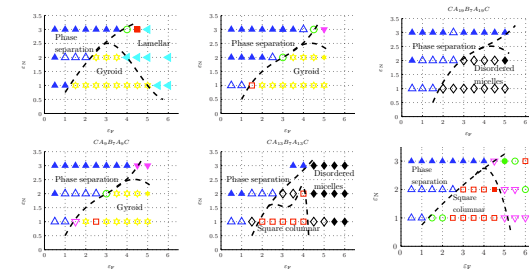
study one supercooled liquid model



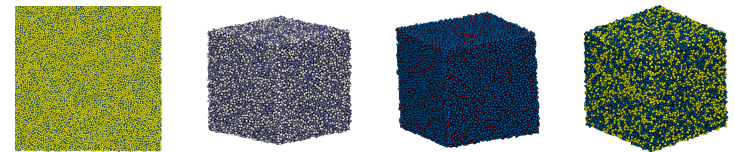
study three monodisperse tethered nanospheres

Papers with GPU jobs...

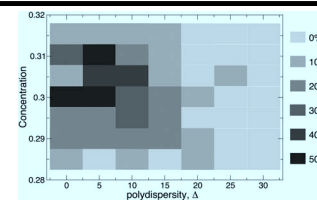
run thousands of single GPU jobs - **one day** of GPU time for each



compute phase diagrams for six polymer architectures



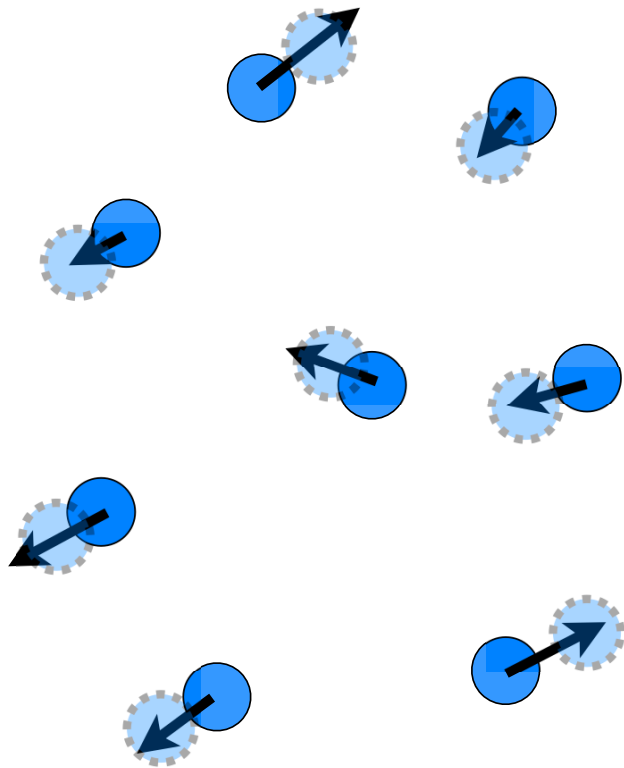
provide an in-depth comparison of four different models



study the effects of of varying polydispersity

490 jobs just for one figure!

Method overview



$$\vec{r}_i(t)$$

$$\vec{v}_i(t)$$



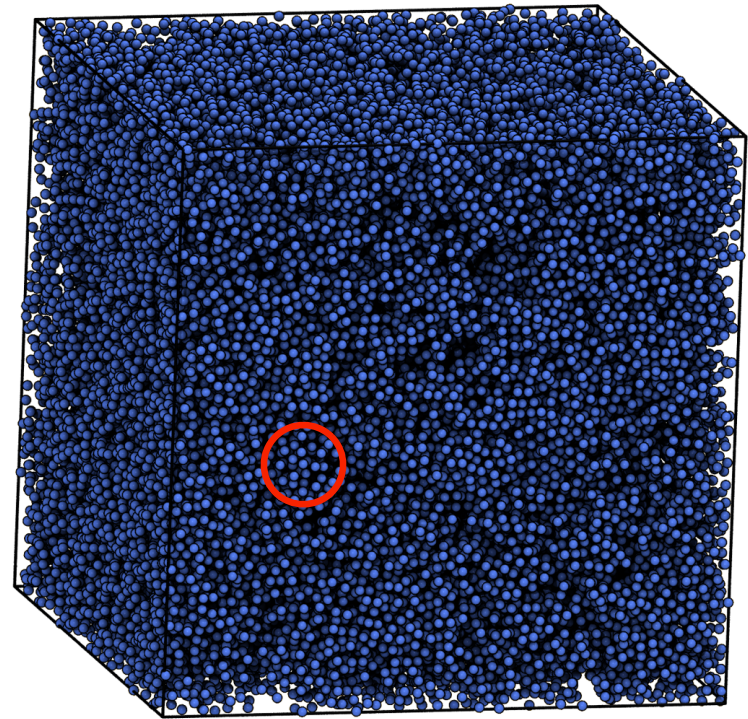
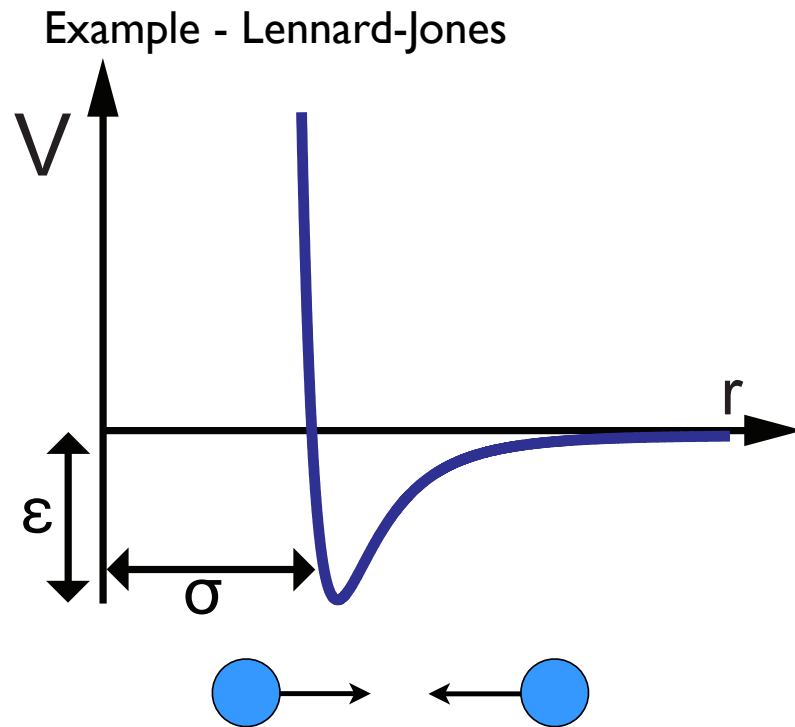
Calculate accelerations



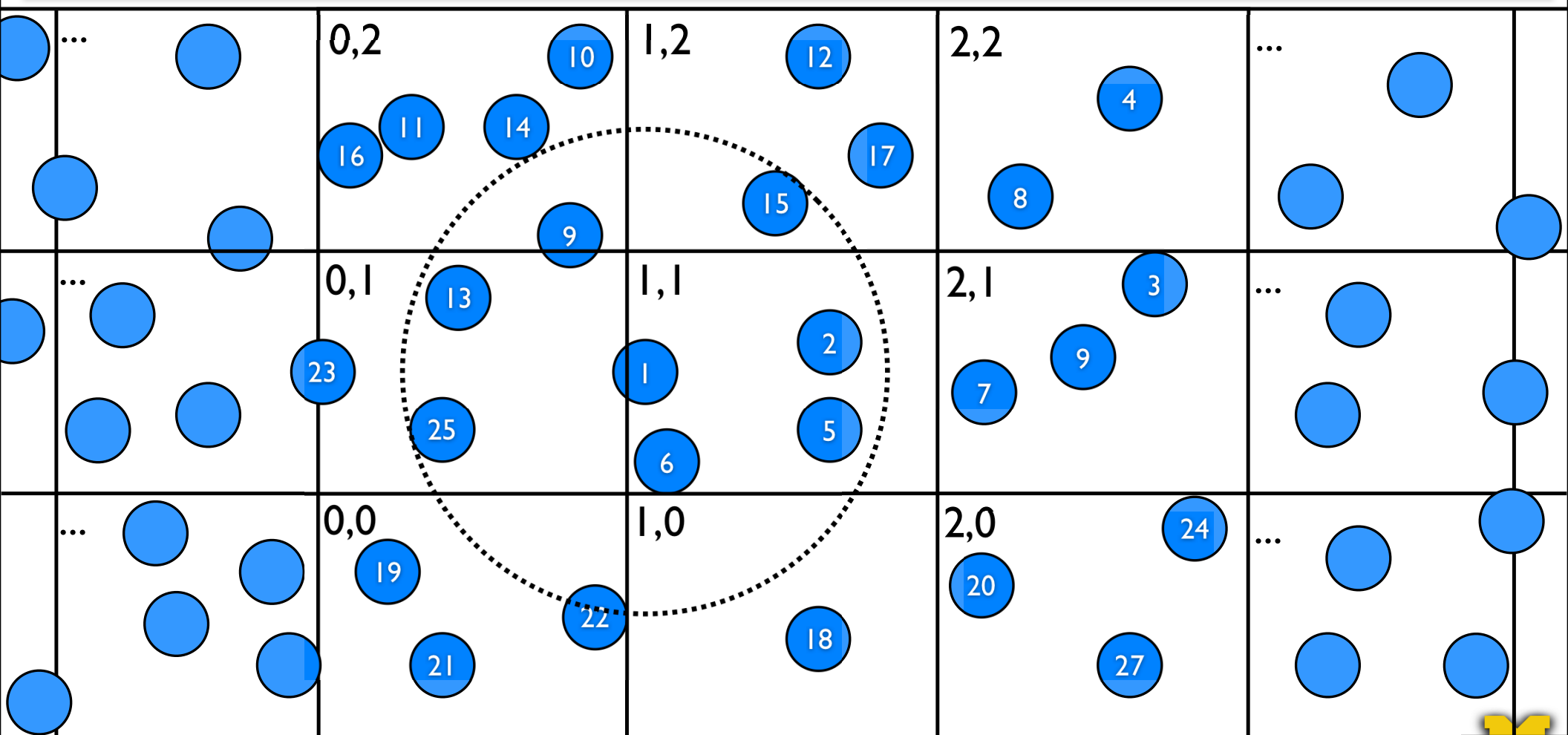
$$\vec{r}_i(t + \delta t)$$

$$\vec{v}_i(t + \delta t)$$

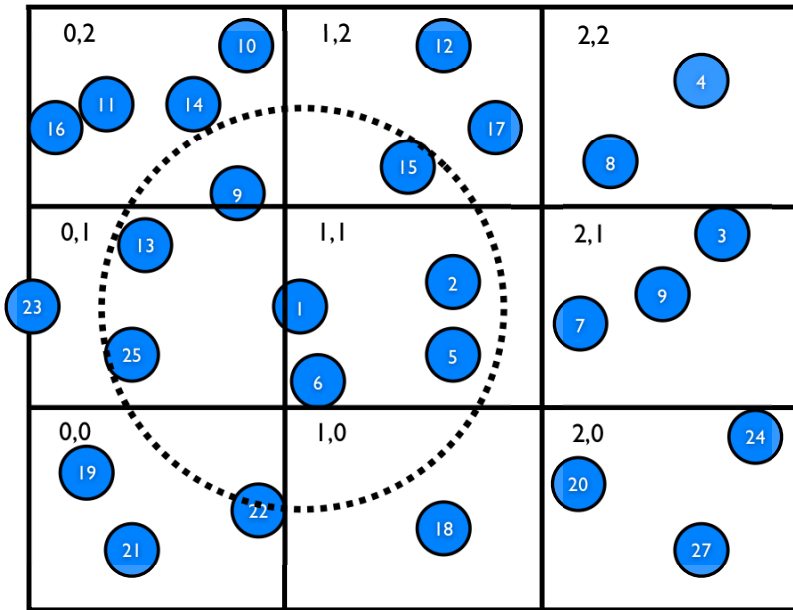
Pair potential



Cell list



Generating the cell list



```

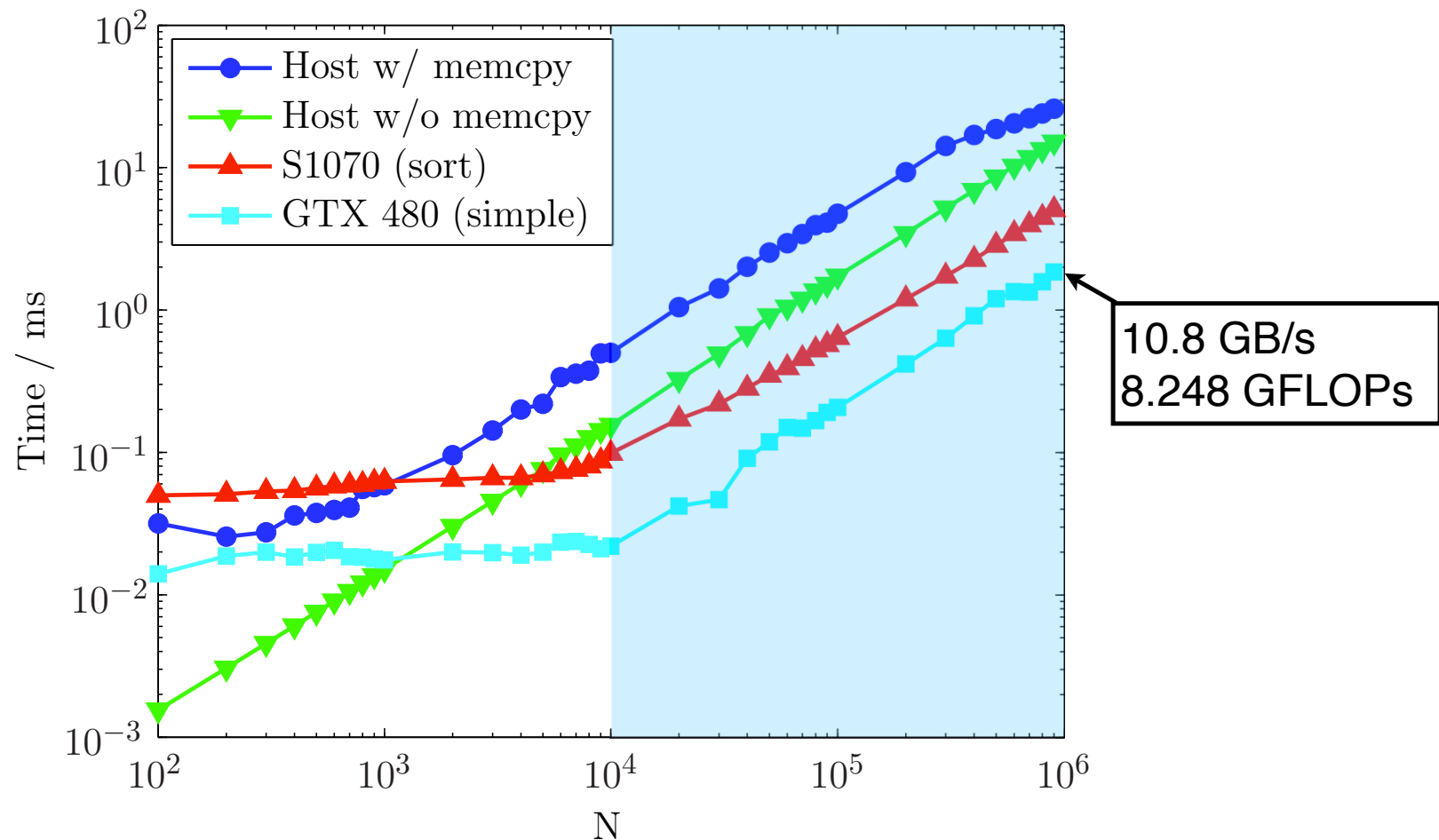
for each particle i in parallel
  load position pos[i]
  compute cell index ci
  cur_size = atomicInc length[ci]
  write (ci, pos[i]) to
    cell_list[ci][cur_size]
    
```

	1	2	3	4	5	6	7	8	9	10
0,0	3					19	21	22		
1,0	1					18				
2,0	3					20	24	27		
0,1	3					23	25	13		
1,1	4					1	6	2	5	
2,1	3					7	9	3		
0,2	5					16	11	14	10	9
1,2	3					12	15	17		
2,2	2					8	4			

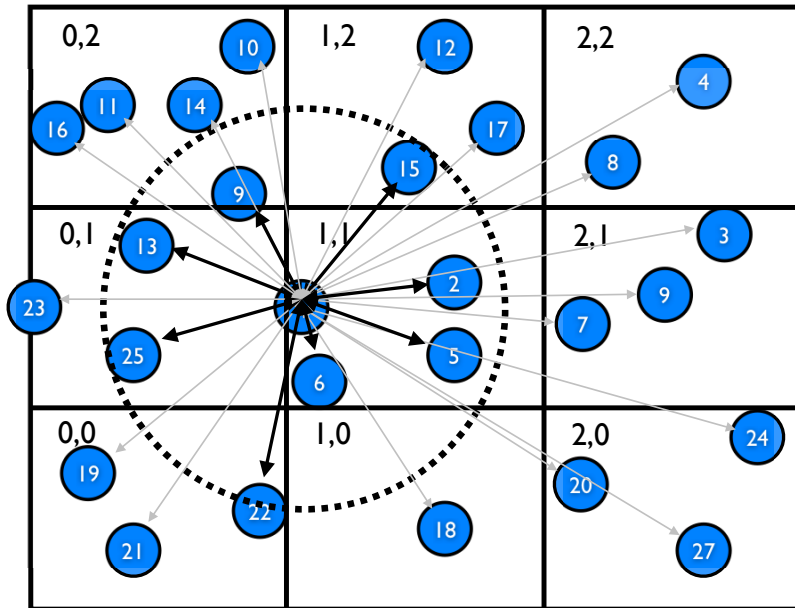
• • •



Cell list performance



Creating the neighbor list



1	2	3	4	5
---	---	---	---	---

25	1	7	3	18
13	6	9	8	20
6	5	8	...	1
2	7	4		2
5	15	...		6
9	17			7
15				15

Neighbor list

7	6	...	...	7
---	---	-----	-----	---

Num Neighbors

```

for each particle i in parallel
  load position pos[i]
  compute cell index ci
  for each nearby cell cn
    for each particle p in cn
      load cell_list[cn][p]
      if distance < rcut
        append to n_list[[i]
    
```



Neighbor list on GF100 and G200

GF100

```
for each particle i in parallel
  load position pos[i]
  compute cell index ci
  for each nid from 0 to 27
    cn=ld.global adj_list[ci][nid]
    for each particle p in cn
      ld.global cell_list[cn][p]
      if distance < rcut
        append to n_list[][i]
```

Notes

- Semi-random memory reads performed from L1
- Activate 48k L1 for best perf.
- Spatial sorting (later) increases cache hit ratio
- Bottleneck becomes the incoherent *n_list* append (only 1 in 8 writes pass the distance test)

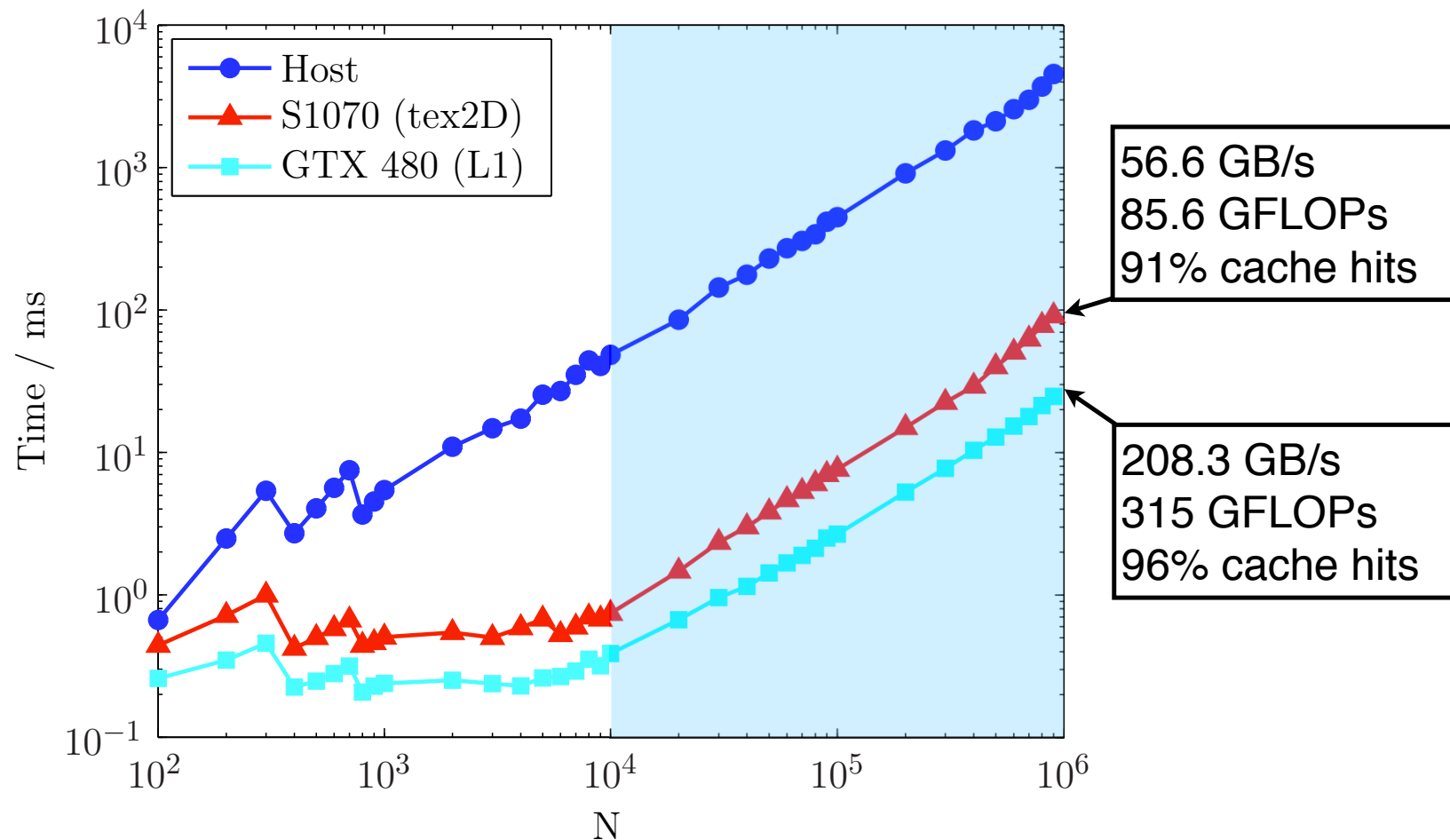
G200

```
for each particle i in parallel
  load position pos[i]
  compute cell index ci
  for each nid from 0 to 27
    cn=tex2D adj_list[ci][nid]
    for each particle p in cn
      tex2D cell_list[cn][p]
      if distance < rcut
        append to n_list[][i]
```

Notes

- Semi-random memory reads performed from 2D tex cache

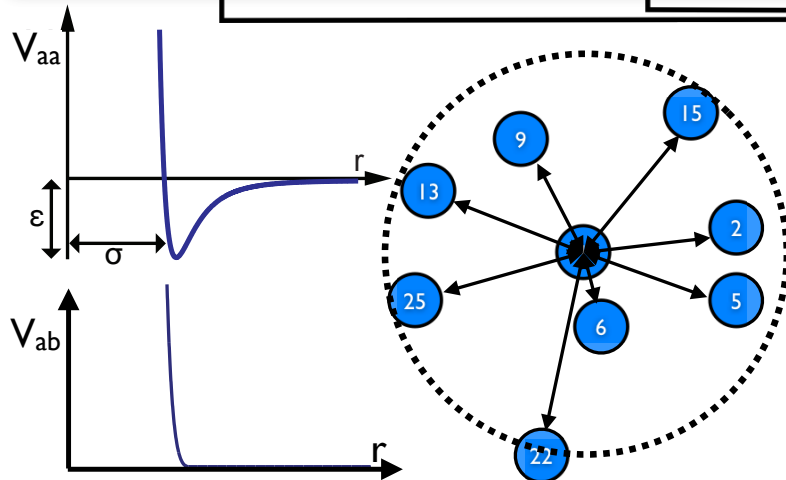
Neighbor list performance



Computing pair forces

Position array

x_1, y_1	x_2, y_2	x_3, y_3	...								
------------	------------	------------	-----	--	--	--	--	--	--	--	--



```

for each particle i in parallel
  load position pos[i]
  for each neighbor n
    j = n_list[n][i]
    load pos[j]
    load coeffs for typei, typej
    compute interaction i, j
    write total interaction on i
    
```

Neighbor list

25	1	7	3	18
13	6	9	8	20
...	...	...	...	...

1	2	3	4	5
---	---	---	---	---

--	--	--	--	--

Computed forces and energies

-
-
-
-
-
-
-
-
-
-



Pair forces on Fermi vs. Tesla

GF100

identical



G200

```
for each particle i in parallel
  load position pos[i]
  nextj = ld.global n_list[n][i]
  for each neighbor n
    curj = nextj
    nextj=ld.global n_list[n+1][i]
    tex1Dfetch pos[curj]
    ld.shared coeffs[typei][typej]
    compute interaction i,j
  write total interaction on i
```

Notes

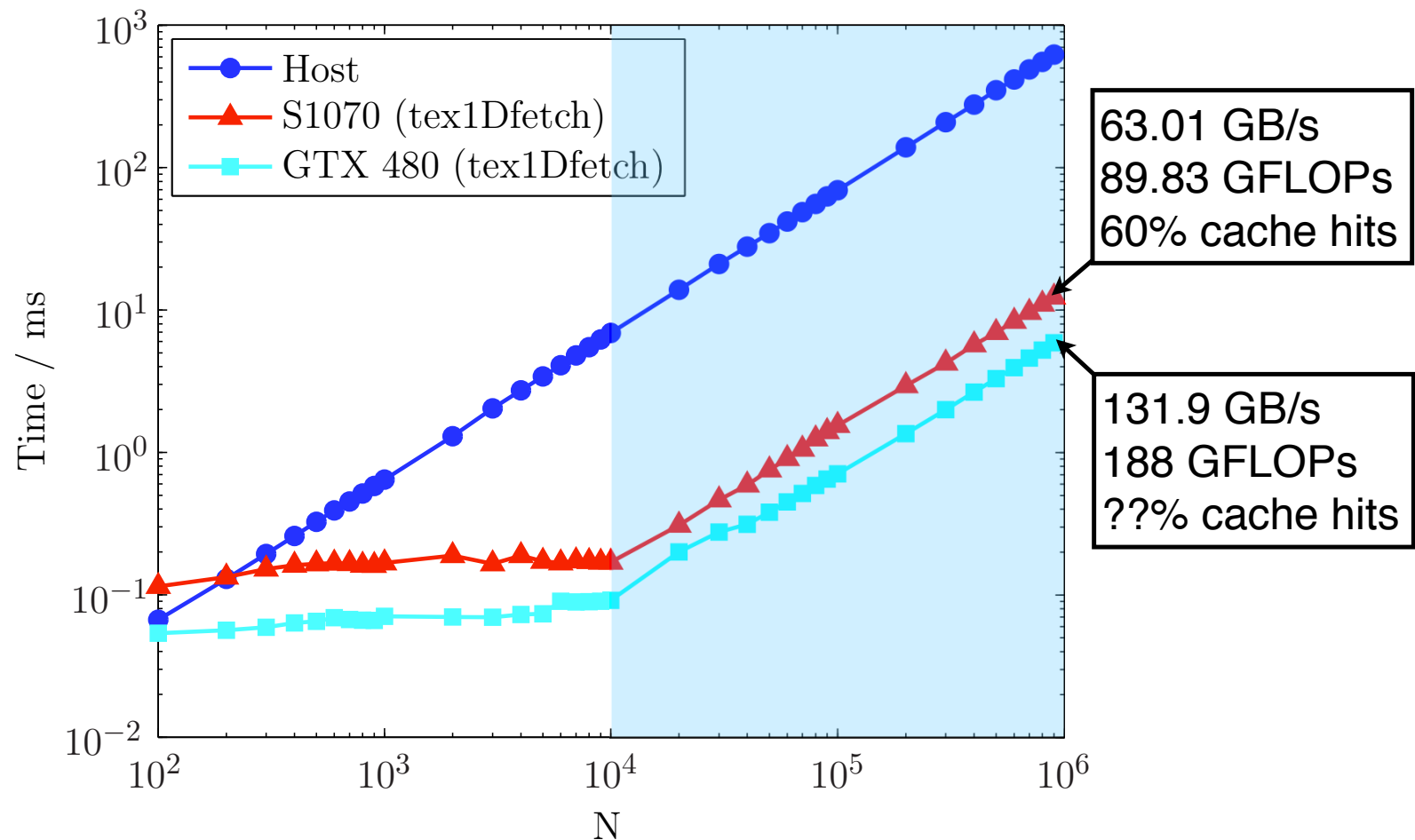
- Switching the *pos[curj]* read to use L1 reduces performance
- I'm not sure why..... 48k L1 >>> 8k tex cache
- Have tried a number of transformations without success
- The coefficient **ld.shared** can be converted to **ld.global** with no performance hit

Notes

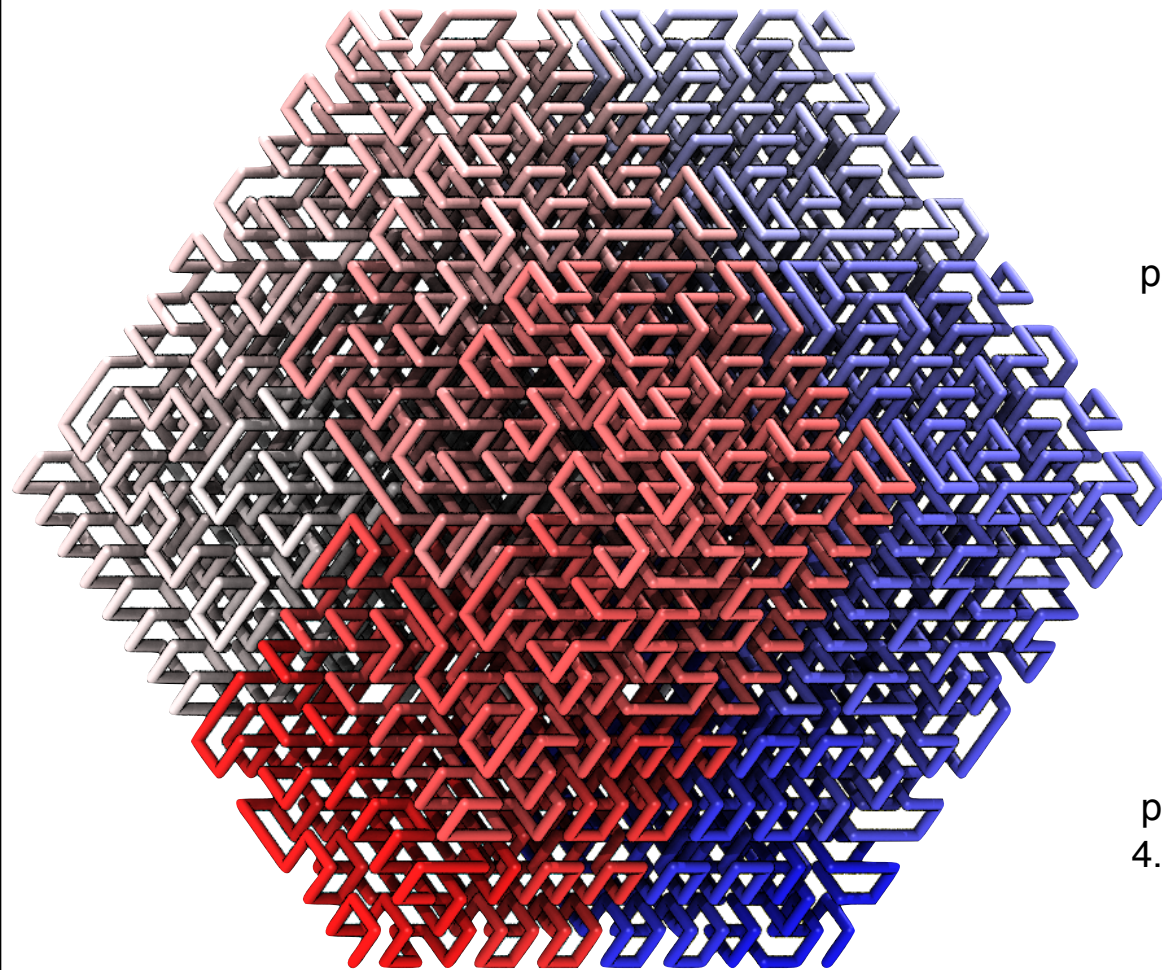
- Semi-random memory reads performed via tex1Dfetch



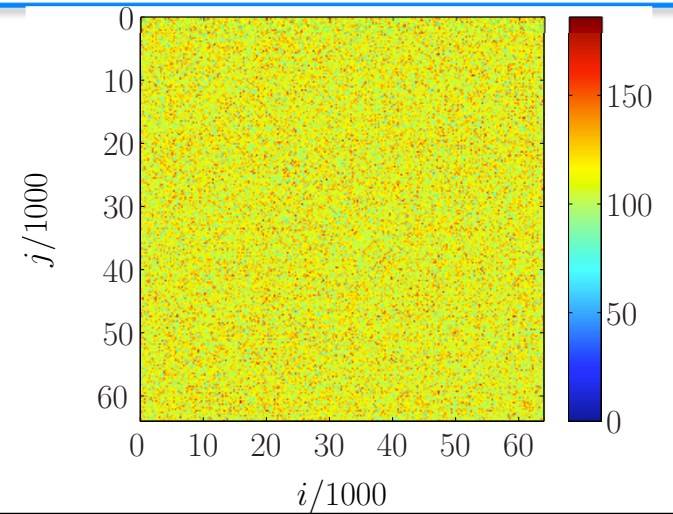
Pair force performance



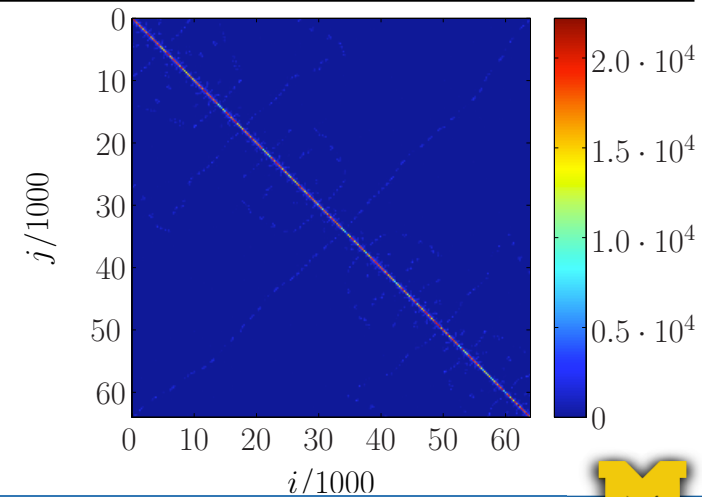
Spatial sorting reorders particles



Random
pair: 50.4 ms



Sorted
pair: 12.3 ms
4.2x speedup!



Flexible pair potentials

```
template< class evaluator> __global__ void
gpu_compute_pair_forces_kernel(float4 *d_force, float4 *d_pos, gpu_nlist_array nlist,
                             typename evaluator::param_type *d_params,
                             ...)
{
    extern __shared__ typename evaluator::param_type s_params[]
    // load data from d_params into s_params ...
    __syncthreads();
    unsigned int idx = blockIdx.x * blockDim.x + threadIdx.x;
    // load in data for particle idx ...
    for (int neigh_idx = 0; neigh_idx < n_neigh; neigh_idx++)
    {
        // access current neighbor ...
        // calculate dr^2 (with periodic boundary conditions) ...
        float rsq = dx*dx + dy*dy + dz*dz;
        unsigned int typpair = typpair_idx(__float_as_int(posi.w), __float_as_int(posj.w));
        typename evaluator::param_type param = s_params[typpair];

        evaluator eval(rsq, rcutsq, param);
        eval.evalForceAndEnergy(force_divr, pair_eng, energy_shift);

        // tally results into force ...
    }
    d_force[idx] = force;
}
```

Evaluator functor

```
class EvaluatorPairLJ
{
public:
    typedef Scalar2 param_type;

    DEVICE EvaluatorPairLJ(Scalar _rsq, Scalar _rcutsq, const param_type& _params)
        : rsq(_rsq), rcutsq(_rcutsq), lj1(_params.x), lj2(_params.y) { }

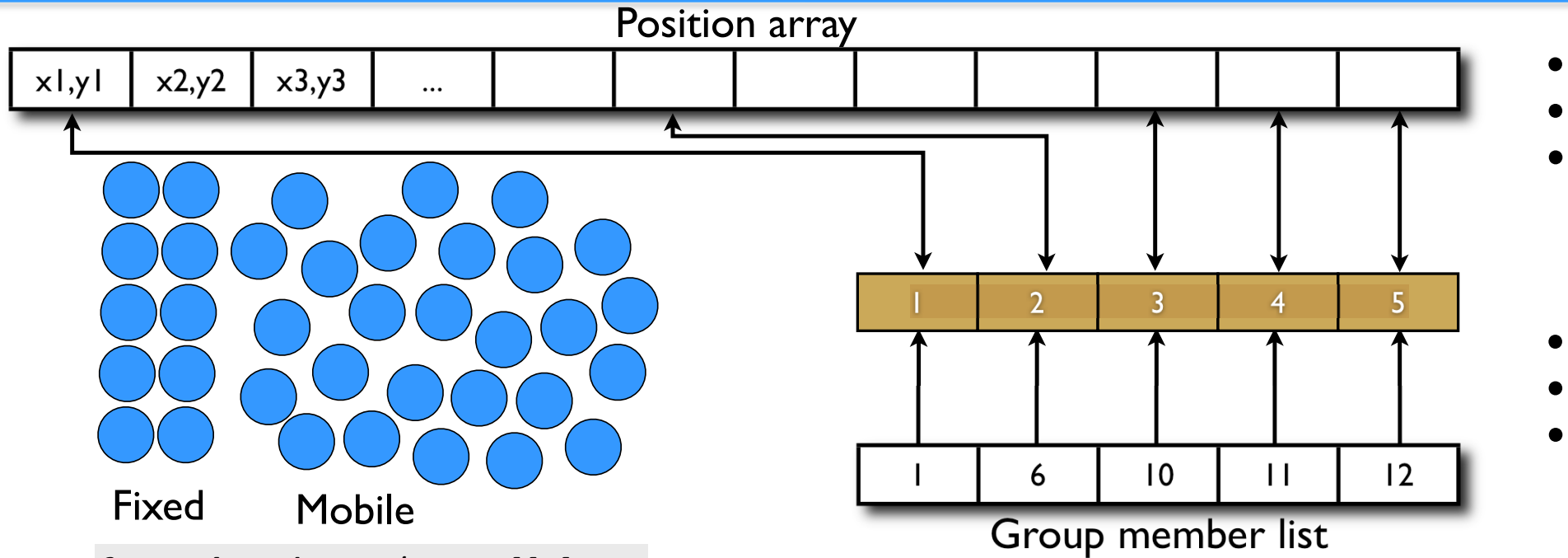
    DEVICE void evalForceAndEnergy(Scalar& force_divr, Scalar& pair_eng)
    {
        if (rsq < rcutsq && lj1 != 0)
        {
            Scalar r2inv = Scalar(1.0)/rsq;
            Scalar r6inv = r2inv * r2inv * r2inv;
            force_divr= r2inv * r6inv * (Scalar(12.0)*lj1*r6inv - Scalar(6.0)*lj2);

            pair_eng = r6inv * (lj1*r6inv - lj2);
        }
    }

protected:
    Scalar rsq, rcutsq, lj1, lj2;
};

class EvaluatorPairGauss
{
    //...
```

Flexible integration



```
for each member g in parallel
  i = load group_idx[g]
  load pos[j]
  load vel[j]
  load force[j]
  compute updated quantities
  write pos[j]
  write vel[j]
```

Notes

- Member list is maintained in a sorted order
- This reduces the number of wasted memory transactions

Feature sheet

Integration

- NVT (Nosé-Hoover)
- NPT
- Brownian Dynamics
- Dissipative Particle Dynamics
- NVE
- FIRE energy minimization

Bond forces

- harmonic
- FENE

Angle forces

- harmonic
- CGCMM

Dihedral/Improper forces

- harmonic

Simulation types

- 2D and 3D
- Replica exchange (via script)

Snapshot formats

- MOL2
- DCD
- PDB
- XML

Pair forces

- Lennard Jones
- Gaussian
- CGCMM
- Morse
- Table (arbitrary)
- Yukawa
- PME (*in development*)

Many-body forces

- EAM (*in development*)

Hardware support

- All recent NVIDIA GPUs
- Multi-core CPUs via OpenMP

Script example

Documentation online: <http://codeblue.umich.edu/hoomd-blue/doc>

```
from hoomd_script import *

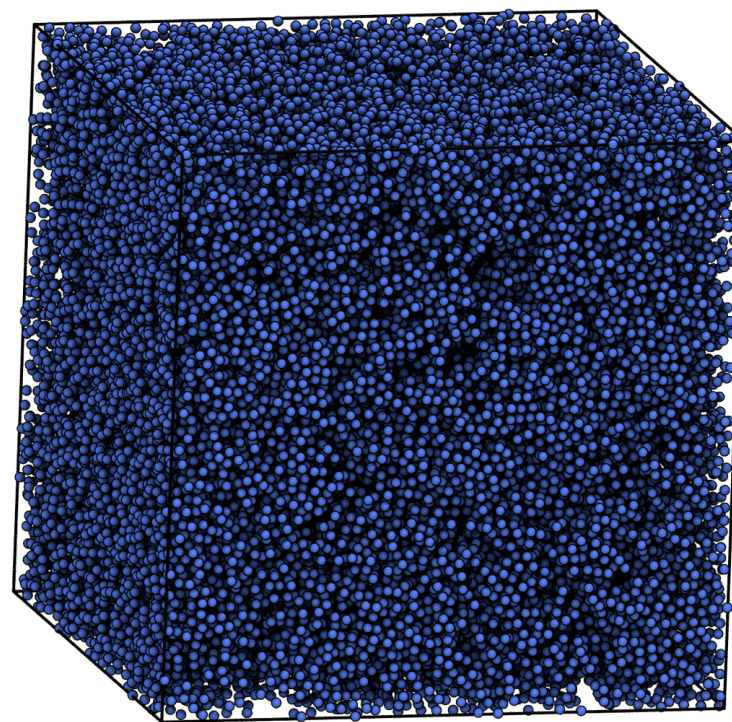
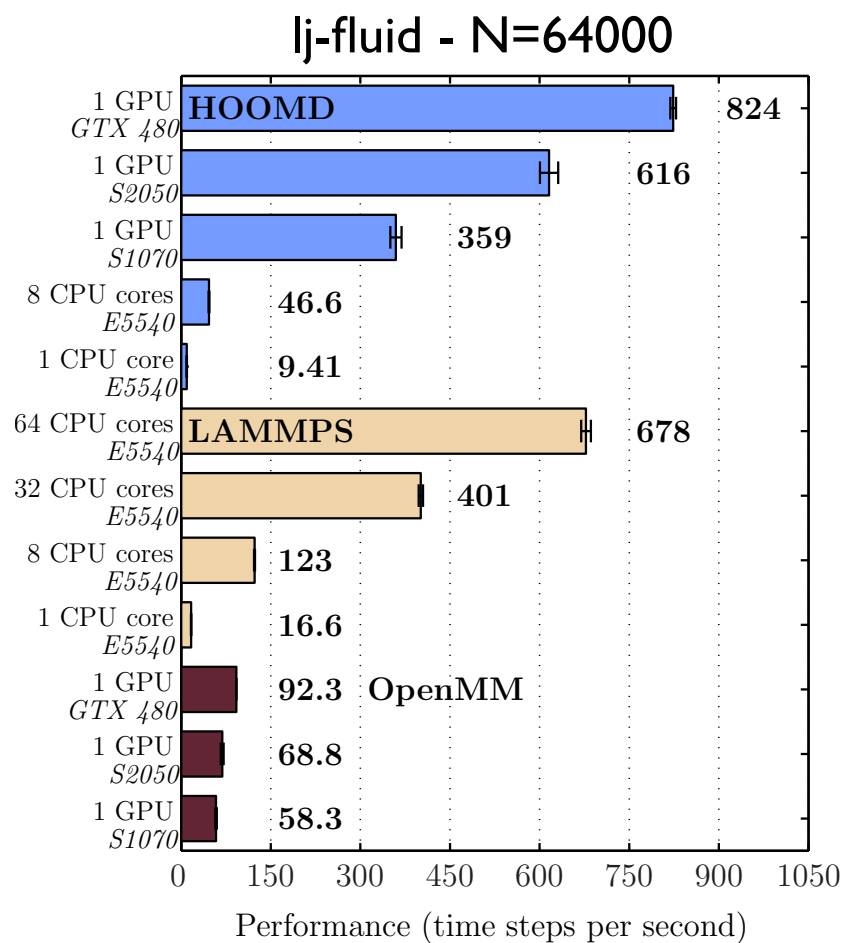
init.read_xml(filename='init.xml')

lj = pair.lj(r_cut=2.5)
lj.pair_coeff.set('A', 'A', epsilon=1.0, sigma=1.0)

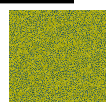
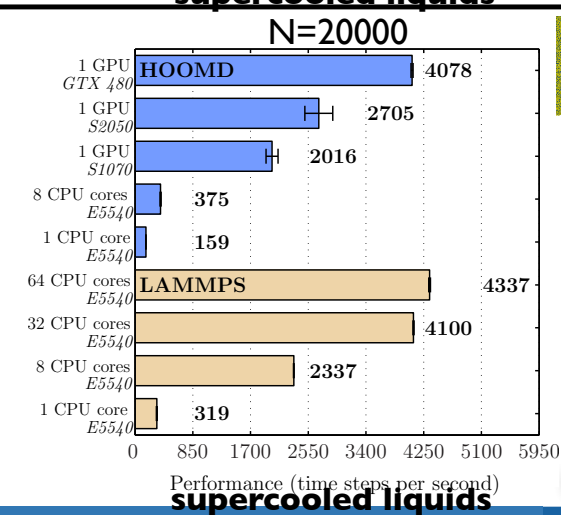
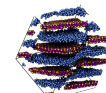
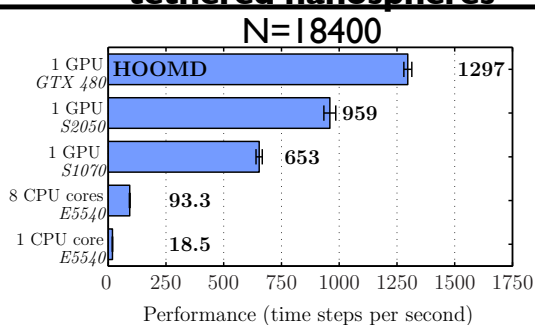
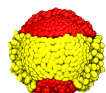
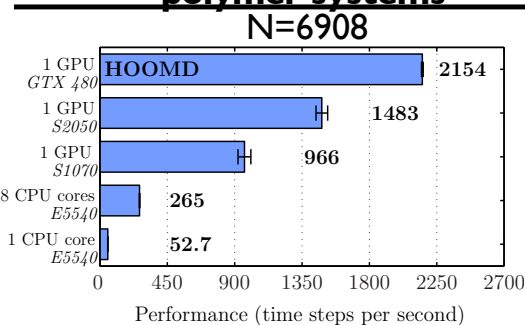
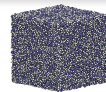
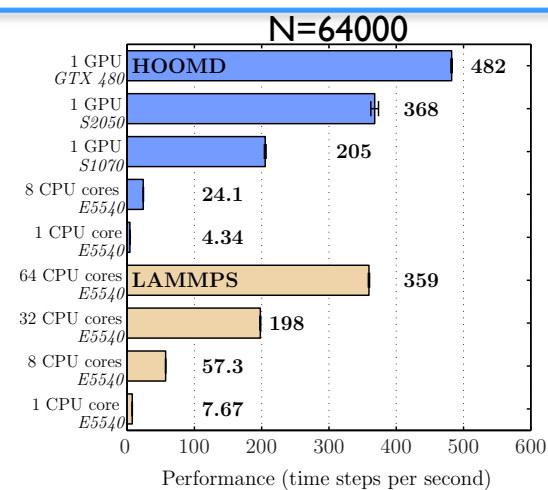
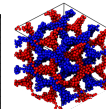
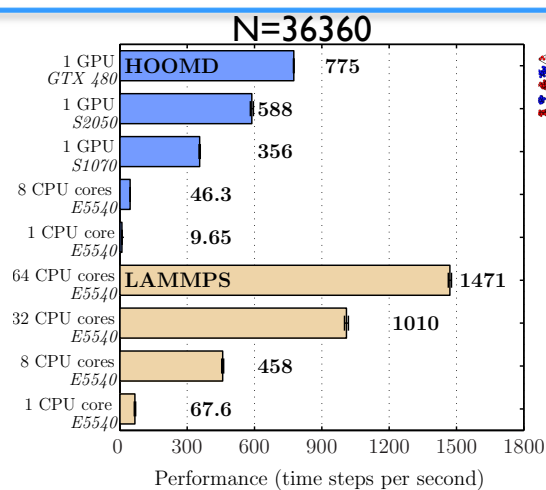
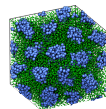
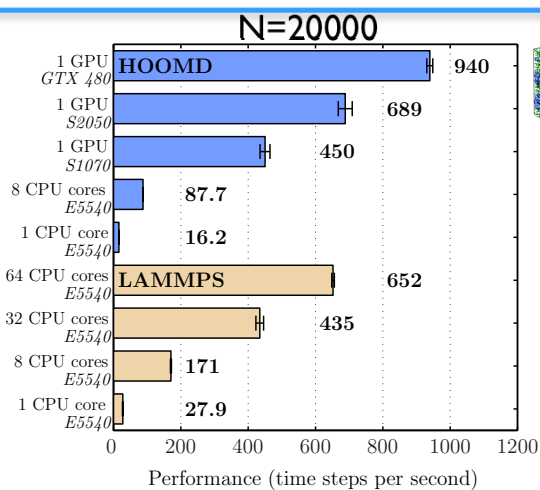
integrate.mode_standard(dt=0.005)
integrate.nvt(T=1.2, tau=0.5)

run(10e3)
```

Overall performance



... more performance



surfactant coated surfaces

polymer nanocomposites

supercooled liquids



Acknowledgements

HOOMD-blue is open source!

- Contributions from around the world
 - Joshua Anderson, Aaron Keys, Trung Dac Nguyen, Carolyn Phillips (University of Michigan)
 - Rastko Sknepnek (Northwestern)
 - Alex Travesset (Iowa State University)
 - Axel Kohlmeyer, David Lebard, and Ben Levine (Temple)
 - Igor Morozov, Kazennov Andrey, Bystryi Roman (Russian Academy of Sciences)

Cell list on GF100 and G200

GF100

```
for each particle i in parallel
  load position pos[i]
  compute cell index ci
  cur_size = atomicInc length[ci]
  write (ci, pos[i]) to
    cell_list[ci][cur_size]
```

Notes

- atomicInc in L2 cache - *fast!*
- Typically ~30 possible collisions
- Spatial sorting (later) *increases* chances that collisions occur within the same block/warp

G200

x,y,z	x,y,z	x,y,z	x,y,z	x,y,z
-------	-------	-------	-------	-------

Read from global
memory identify cell

Store cell,index pairs in
shared memory

(3,0)	(1,1)	(3,2)	(2,3)	(1,4)
-------	-------	-------	-------	-------

Sort the pairs using cell
as the sort key

Done in parallel with a
bitonic sort

(1,2)	(1,4)	(2,3)	(3,0)	(3,2)
-------	-------	-------	-------	-------

Identify common
sequences

(1,2)	(1,4)	(2,3)	(3,0)	(3,2)
-------	-------	-------	-------	-------

- Only one atomicAdd per unique sequence is needed