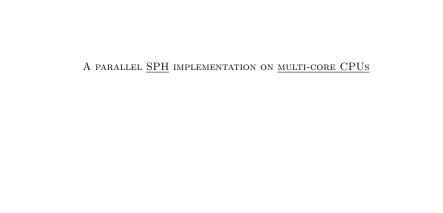
# A PARALLEL SPH IMPLEMENTATION ON MULTI-CORE CPUs

(2010)

Ihmsen, M., Nadir, A., Becker, N., Teschner, M.

presented by Steffen Haug



A PARALLEL SPH IMPLEMENTATION ON MULTI-CORE CPUs

# Applies to particle simulations in general!

(And raytracing, rigid body collision, ...)

(Briefly)

$$A(x) = \int_{\Omega} A(V)\delta(x - V) dV$$

Start with the convolution definition of  $\delta$ .

$$A(x) = \int_{\Omega} A(V) W(x - V) dV$$

Replace  $\delta$ -function with smoothing kernel that "works like"  $\delta$ .

$$A(x) = \int_{\Omega} A(V) W(x - V) dV$$

Key points: 
$$W$$
 has  $compact\ support$  of radius  $h$  
$$\int_{\Omega}W=1\ ({\rm Normalization})$$
 
$$W\longrightarrow \delta\ {\rm as}\ h\longrightarrow 0$$

$$A(x) = \int_{\Omega} A(V) W(x - V) dV$$

TLDR: W is a "bell-like" curve.

(Convolution by bell curve "smoothes" signal, hence the name SPH)

$$A(x_i) = \sum_{i} A(x_i) W(x_i - x_j) V_j$$

Replace integral over  $\Omega$  with sum over samples at N "particles".

$$i, j \in \{1 \dots N\}$$

$$A(x_i) = \sum_{j} A(x_j) W(x_i - x_j) V_j$$

Volume element dV is now the volume  $V_j$  of particle j.

$$A_i = \sum_j A_j W_{ij} V_j$$

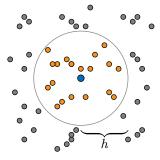
Standard compact notation.

$$A_i = \sum_j A_j W_{ij} V_j$$

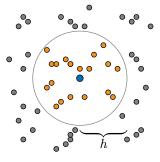
From this you can derive discretizations for  $\nabla A$ ,  $\nabla^2 A$ ,  $\nabla \cdot A$ , and so on...

$$A_i = \sum_j A_j W_{ij} V_j$$
Zero outside the ball  $\frac{B_h(x_i)}{T}$ !

(Compact support of W)



We need to discard the vast majority of particles.



Fast access to particles in  $B_h(x_i)$  is a major optimization!

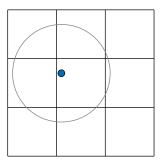
"FIXED-RADIUS NEAR NEIGHBORS" (Classic computational geometry problem)

# The authors present two strategies

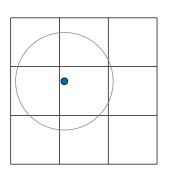
# INDEX SORTING and SPATIAL HASHING

I will present some criticism, especially in the context of GPU.

# Both methods use a tiling of size h



 $B_h(x_i)$  is completely contained in  $3 \times 3$  tile region.



Trivially constructed by rounding the coordinates:

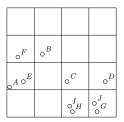
$$i = \left\lfloor \frac{x}{h} \right\rfloor, \quad j = \left\lfloor \frac{y}{h} \right\rfloor$$



1. Create a finite regular grid

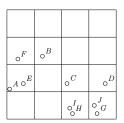
$\circ^F$	$\circ^B$		
$\circ^A \circ^E$		$\circ^C$	$\circ^D$
		${\displaystyle \mathop{\circ}_{O}^{I}}_{H}$	$\circ^J_G$

- 1. Create a finite regular grid
- 2. Calculate the strided grid cell index for each particle





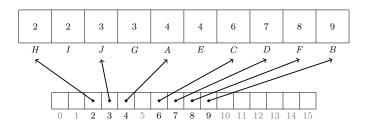
- 1. Create a finite regular grid
- 2. Calculate the strided grid cell index for each particle
- 3. Sort particles by grid index



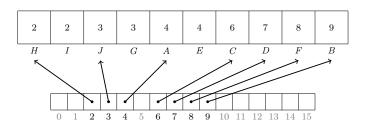
2	2	3	3	4	4	6	7	8	9
LI	T	7	C	4	E.	C	D	E.	D

# Index sorting

- 1. Create a finite regular grid
- 2. Calculate the strided grid cell index for each particle
- 3. Sort particles by grid index
- 4. Identify where the edges of each grid cell is in the particle buffer



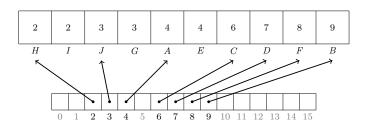
- 1. Create a finite regular grid
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Domain must be finite: Unique index maps to finite set of buckets.

# Index sorting

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Possible optimization: Order grid cells by space filling curve.

# Criticism

- 1. Finite domain is an annoying restriction
- 2. High number of vacant grid cells
- 3. Full sort can be very expensive

#### Criticism

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- 2. High number of vacant grid cells
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# Advantages

- 1. No dynamic memory allocation required
- 2. Minimal synchronization between threads required

Verdict: Reasonably well-suited for GPU

(COMPACT) SPATIAL HASHING

1. Create a (possibly infinite) regular tiling

$\circ^F$	o <sup>B</sup>		
o <sup>A</sup> o <sup>E</sup>		o <sup>C</sup>	<b>o</b> <sup>D</sup>
		${}^{I}_{o}$	$\circ^J_G$

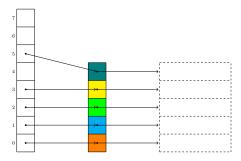
- 1. Create a (possibly infinite) regular tiling
- 2. Hash the grid cell index for each particle (Muliple grid cells can hash to the same bucket!)



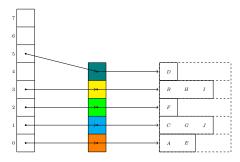


- Create a (possibly infinite) regular tiling
- Hash the grid cell index for each particle (Muliple grid cells can hash to the same bucket!)
- 3. For each populated bucket, allocate a particle buffer in a secondary structure

(Initial capacity k, grow with amortized constant time)



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- Hash the grid cell index for each particle (Muliple grid cells can hash to the same bucket!)
- For each populated bucket, allocate a particle buffer in a secondary structure
   (Initial capacity k, grow with amortized constant time)
- 4. Copy each particle into its respective particle buffer



#### Criticism

- 1. Uses dynamic memory allocation in multiple places
- 2. Forced reallocation of vectors in multple critical places
- 3. Contention between threads on particle buffers (Imagine 10K threads mutating the same vectors...)
- 4. Hash collisions cause performance degradation
- 5. Particle buffers live in separate (non-coherent) heap allocations (The bane of dynamic vectors is that it fragments the heap)

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#### ADVANTAGES

- 1. No restriction on domain size
- 2. Avoids complete sort
- 3. Can more easily be incrementally updated



Verdict: Absolutely useless on GPU

## Results

method	construction	query	total
basic uniform grid	25.7 (27.5)	38.1 (105.6)	<b>63.8</b> (133.1)
index sort [Gre08]	35.8 (38.2)	29.1 (29.9)	<b>64.9</b> (77.3)
Z-index sort	16.5 (20.5)	26.6 (29.7)	<b>43.1</b> (50.2)
spatial hashing	41.9 (44.1)	86.0 (89.9)	127.9 (134.0)
compact hashing	8.2 (9.4)	32.1 (55.2)	<b>40.3</b> (64.6)

**Table 2:** Performance analysis of different spatial acceleration methods with and without (in brackets) reordering of particles. Timings are given in milliseconds for CBD 130K and include storing of pairs.