## 1. EXPERIMENTAL EVALUATION

This set of experiments comprises of comparing leafsize and fanout effect on EVERGREEN variations and traversal type.

# 1.1 General Setup

# 1.1.1 Hardware setup

Laptop, Intel CORE i7, NVIDIA CUDA 5.5. Thrust library (for parallel sorting during building phase), 8 GB  $\rm RAM$ 

# 1.1.2 Algorithm parameters

Unless specified for concrete experiment, parameters are the following:

• Fanout: 3

• Leaf size: 100

• Epsilon: 5

• Datasets: Random Samples generated from data about Axons and Dendrites. Every experimental value is a mean of 5x5 random samples. For example, for calculating total time of joining 1000 axons with 1000 dendrites, firstly 5 random samples of axons and 5 random samples of dendrites are generated, then they are joined (25 results) and a mean result is reported.

Local join algorithm: SGH with dynamic grid  $\left(\left(\frac{Node\ MBR\ volume}{Average\ Assigned\ Object\ volume}\right)^{1/3}\right)$ 

 $\bullet$  First tree assignment probability (For dTOUCH):

$$P_{T_A}^B = e^{-\alpha \times \frac{l}{L_{T_A}}} \times (1 - \frac{l}{L_{T_A}})$$

# 1.2 Leaf size

## 1.2.1 Methodology

Leaf size is a parameter of the experiment. The default value used in other experiments is 100. In this experiment we variate leaf size from 50 to 300. Number of objects in each dataset is 2000.

### 1.2.2 Observation

Figure 1 shows the total runtime of all the four EVER-GREEN variations. We exclude TOUCH because of distracting (worse) performance. This plot suggests:

- reTOUCH and re\*TOUCH: we should run for leafsize larger than 300 to have more clue.
- dTOUCH: for smaller leafsize, i.e. taller tree, performs faster.
- cTOUCH: almost independent or a leafsize between 100 and 200 is already good enough when fanout is 3.

Figure 2 shows the total filtering, i.e.  $F_{total}^{-1}$ . This plot suggests:

- dTOUCH, reTOUCH and re\*TOUCH: shorter tree has more filtering ability. However, we should run for leaf-size larger than 300 to have more clue.
- cTOUCH: almost independent when fanout is 3.

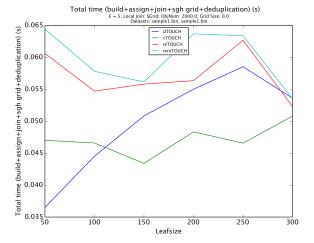


Figure 1: leafsize comparison

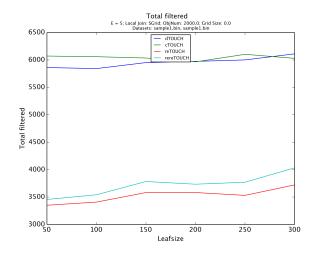


Figure 2: leafsize comparison

#### 1.2.3 Conclusion

The leafsize has very small effect on the runtime of all the variations except dTOUCH.

### 1.3 Fanout

#### 1.3.1 Methodology

Fanout is a parameter of the experiment. The default value used in other experiments is 3. In this experiment we variate leaf size from 2 to 10. Number of objects in each dataset is 2000.

### 1.3.2 Observation

Figure 3 shows the total runtime of all the four EVER-GREEN variations. We again exclude TOUCH because of distracting performance. We can make following observations:

• reTOUCH and re\*TOUCH: we should run for fanout smaller 4 to have more clue.

<sup>&</sup>lt;sup>1</sup>explained in table 1

 dTOUCH and cTOUCH: the performance does not depend from fanout significantly in the observed range.

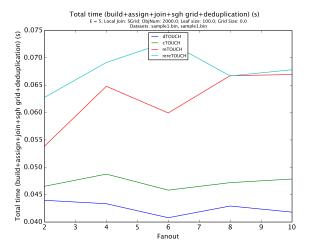


Figure 3: fanout comparison

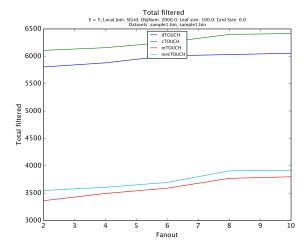


Figure 4: fanout comparison

For all algorithms fanout increases slightly with fanout.

### 1.3.3 Conclusion

Fanout does not influence the performance of algorithms in selected fanout range. The biggest change can be observed for cTOUCH, that is also quite small.

# 1.4 Traversal types

### 1.4.1 Methodology

For reTOUCH 4 cases of tree traversal during joining step were implemented. Here are the explanation of these cases:

 Case 1: Top-Down travesal. This traversal type was used originally in TOUCH and other variations of EV-ERGREEN. Grids are precalculated just before joining step. Every object in every node is joined with all objects of those nodes below, where the overlap of current object and MBR of descendant node is not empty.

- Case 3: Top-Down traversal without filtering. For each node in the tree we create a grid and for each descendant node we join all assigned objects of the descendant node with created grid.
- Case 3 with filtering: For each node in the tree we create a grid and for each descendant node we join all assigned objects if MBR of descendant node overlaps with a node with created grid.
- Case 4: Bottom-up traversal. Using depth-first tree traversal for each node create a grid and for each ancestor of this node join all assigned objects with created grid.

Experiments were conducted with constant grid resolution and dynamic grid resolution to state if observed performance is a result of resolution equation or traversal type. Also, additionally to Axon/Dendrite dataset, one more experiment was conducted using Random distribution with additional object random expansion. This was done because of the observation that bigger deviation of object size leads to emphasizing the specifications of the algorithm.

#### 1.4.2 Observation

Figure 5 shows total time of joining axons with dendrites using logarithmic scale. Case 1 shows better performance than all others, despite of outstanding result if number of objects is 4000. This point (objNum = 4000) can be explained by bad equation of SGH. The reason of bad total time is not a joining phase (as it should be if the tree structure is the bottleneck), but assignment step and SGH building step (see Figure 7 and Figure 8). Probably, some average size of assigned objects leads to extreme case of SGH resolution. At the same time, the ComparedMax value (see Terminology table) proves the best Case 1 performance (Figure 9).

Figure ?? is the explanation of this good performance of case 1. It shows the number of comparisons during the joining step. Additional filtering during the joining phase using object's MBR make total number of comparisons during the assignment step smaller.

For excluding the factor of bad SGH equation, two more experiments were conducted. One of them is Axon/Dendrite join with constant resolution and another is Uniformly randomly distributed objects join that was also randomly expanded (for bigger deviation). First experiment result is illustrated on Figure 10 and second (Random distribution) on Figure 11. Constant grid on Axon/Dendrite datasets shows slightly different results with different tree traversal types, however Case 1 is also the winner. Random distribution with high deviation was expected to emphasize the difference and we see the much bigger increase in performance if we use Case 1.

#### 1.4.3 Conclusion

Case 1 is suggested for further experiments.

## 2. EXPERIMENT TERMINOLOGIES

Epsilon	Distance between objects such that closer objects are considered as interacting
#A	Number of objects used from first dataset
#B	Number of objects used from second dataset
infile A	Path to first dataset
infile B	Path to second dataset
LocalJoin Alg	Algorithm used for local join (join inside nodes)
Fanout	Fanout (capacity) of a node in a tree
Leaf size	Size (fanout) of leaf nodes
gridSize	Number of cells per dimention for Grid Hash (Local Join).
Resolution	gridSize/Universe (all used sapce)
Compared #	Total number of comparisons (comparison function calls)
Compared %	Compared # / (#A*#B) * 100%
ComparedMax	Compared # in case of using Nested Loop as local join algorithm
Duplicates	Number of duplicate result pairs in case of Grid Hash
Results	Number of interacting pairs
Selectivity	Results / (#A*#B) * 100%
$F_A$	Number of filtered objects of first dataset
$F_B$	Number of filtered objects of second dataset
$F_total$	$\int F_A + F_B$
t loading	Time for loading and parsing data
t init	Time for initializing Grid Hash
t build	Time for assignment steps
t probe	Time for probing function, total joining step
t comparing	Time for Local Joins
t partition	Time for creating nodes
t total	Total time of Spatial Join class existance (including I/O)
t deDuplicating	Time for removing duplicate objects in Grid Hash
t analyzing	Time for analyzing a tree structure after assignment step
t sorting	Time used for sorting objects during creation of tree structure
t gridCalculate	Time for precalculating all Grid Hash for every node (cTOUCH, re*TOUCH)
t sizeCalculate	Time for calculating size statistics on the nodes
EmptyCells(MaxObj	Number of maximum assigned objects of one type to one node
AveObj	Average number of objects in a node
StdObj	Standart deviation of number of objects in a node
repA	Number of replications of type A using Grid Hash (if one object is assigned to more than one cell of Grid Hash)
repB	Number of replications of type B usign Grid Hash (if one object is assigned to more than one cell of Grid Hash)
max level	Parameter that defines whether object is assigned to the first tree in dTOUCH
gridProbe	(if SGH) time for calculating grid and probing together
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Terminology table

Algorithm

Main algorithm name

Table 1: Terminologies used in the experiments.

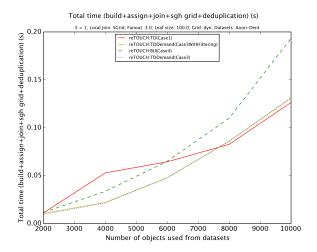


Figure 5: Traversal type comparison: Total time

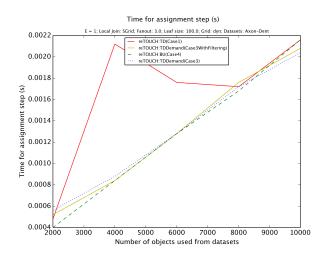


Figure 7: Traversal type comparison : Assignment time

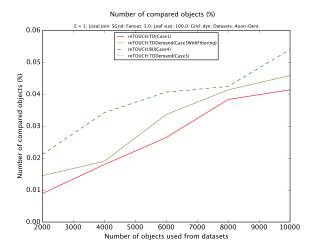


Figure 6: Traversal type comparison: Number of compared objects (%)

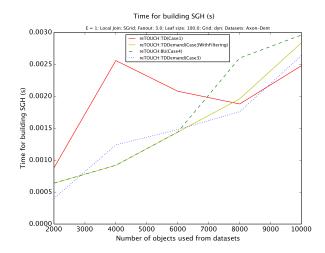


Figure 8: Traversal type comparison : SGH builind time

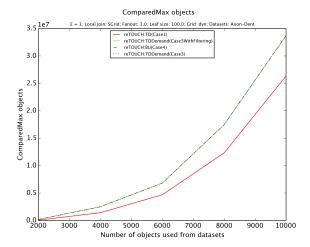


Figure 9: Traversal type comparison : Number of compared objects if  $\operatorname{NL}$  is used

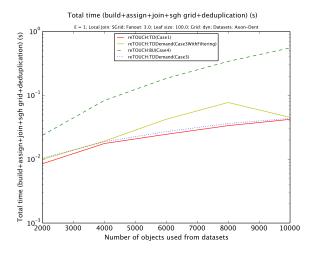


Figure 10: Traversal type comparison : Total time of joining Axon/Dendrite samples using constant SGH resolution

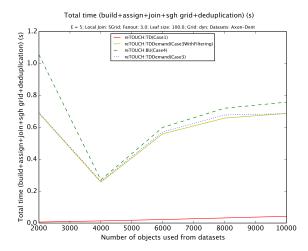


Figure 11: Traversal type comparison : Total time of joining UniRandom Expended samples