

A Practice-Minded Approach to Computing Motorcycle Graphs

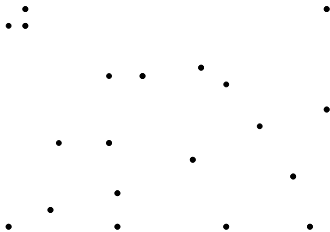
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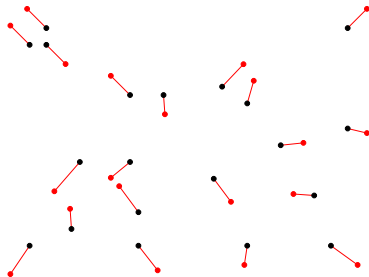
What is a motorcycle graph?

- We define a **motorcycle** m as a triple $(p, s, t^*) \in \mathbb{R}^2 \times \mathbb{R}^2 \times [0, \infty)$, where p is the start point, t^* is the start time and s is the speed vector.
- Consider n motorcycles m_1, \dots, m_n , with $m_i = (p_i, s_i, t_i^*)$. Each motorcycle leaves a trace behind and crashes when reaching the trace of another motorcycle.



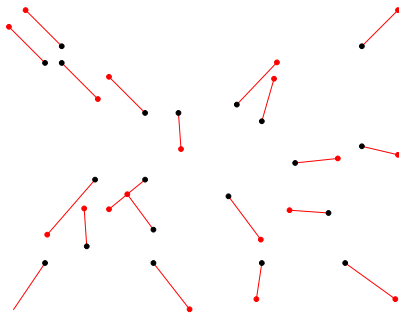
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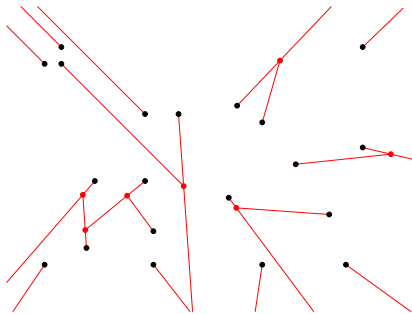
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- Find $O(n)$ crashes in chronological order. Testing each against each takes $O(n^2)$ time for each crash.
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- **Cheng and Vigneron, 2002**

- Induced a partitioning of the plane by $1/\sqrt{n}$ -cuttings and exploited arrangements on each cutting-cell, resulting in an $O(n\sqrt{n} \log n)$ algorithm.
- Too complicated for an actual implementation.

Summary

- No “close-to linear” algorithm is known.
- No sub-quadratic implementation is known.

Basic idea

Replace the $1/\sqrt{n}$ -cutting of Cheng and Vigneron's algorithm by a regular rectangular grid and drop the arrangements.

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Computing motorcycle graphs

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Main question

Consider n motorcycles on a $h \times h$ hash. In the worst case, this leads to $O(n \cdot h)$ crossings of motorcycles with the grid-lines. Do we have a chance to obtain a good performance in practice?

Discrete event simulation of the moving motorcycles:

- 1 Crash event: a motorcycle crashes into a trace.
- 2 Switch event: a motorcycle leaves one grid cell and enters a neighboring grid cell.

In the course of simulation, the algorithm iteratively extracts the next event from a priority queue and processes it.

Input data and data structures

Our input consists of:

- A set $M := \{m_1, \dots, m_n\}$ of motorcycles.
No need to know M a-priori: new motorcycles may emerge, if their start time is in the future.
- A set W of line segments representing walls, where motorcycles may crash against.

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We maintain the following data structures:

- A priority queue Q of pending events.
- For every motorcycle m a binary search trees $C[m]$, where $C[m]$ holds potential future crash events of m .
- A geometric hash H for tracking the motorcycle traces, using a $h \times h$ grid.
- A geometric hash G for the wall-segments, using the same grid geometry as H .

Basic algorithm

```
1: procedure MCGRAPH(motorcycles  $M$ , walls  $W$ )
2:    $Q, C, H \leftarrow$  initialize empty
3:    $G \leftarrow$  geometric hash with all  $w \in W$ 

4:   for all  $m \in M$  do
5:     INSERTMC( $m$ )
6:                                      $\triangleright$  Adds an empty binary tree  $C[m]$  to  $C$ 
7:                                      $\triangleright$  Inserts an initial switch-event of  $m$  to  $Q$ 
8:   end for

9:   while not  $Q.empty()$  do                                      $\triangleright$  Process all events  $e$ 
10:     $e \leftarrow Q.pop()$ 
11:    HANDLE( $e$ )
12:                                      $\triangleright$  Switch-event: attach motorcycle to new grid cell,
13:                                     add next switch-event to  $Q$ , maintain  $C$ .
14:                                      $\triangleright$  Crash-event: clean-up stale future crash-events in  $C$ .
15:   end while
16: end procedure
```

Let k be the maximum number of motorcycles in a hash cell. Processing a switch- resp. crash-event can be done in $O(k \log n)$ time.

There are $O(n)$ crash-events and $O(n \cdot h)$ switch-events and we choose $h \in \Theta(\sqrt{n})$. Hence, the worst case complexity is

$$O(nkh \log n) \subseteq O(n^2 \sqrt{n} \log n).$$

Worst case

$\Omega(n)$ motorcycles cross $\Omega(h)$ hash-cells in a narrow strip that is $O(1)$ cells thick. Further, no other motorcycle is allowed to cross this strip before.

... what is the expected runtime?

We denote by $S := [0, 1]^2$ the unit square covered by a $h \times h$ grid.

Lemma

Let $R = (p, \varphi) \in S \times [0, 2\pi)$ be a uniformly distributed ray, starting at p , with direction angle φ . Further, let C be a cell of a $h \times h$ grid on S . The probability that R intersects C is $\Theta(1/h)$.

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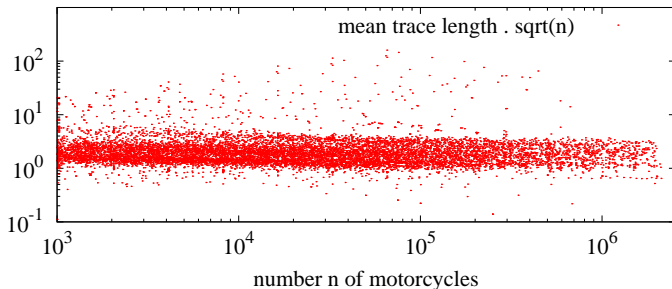
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Theorem

Consider n random rays distributed within S . The expected number of rays intersecting a specific cell is in $\Theta(n/h)$.

Expected runtime

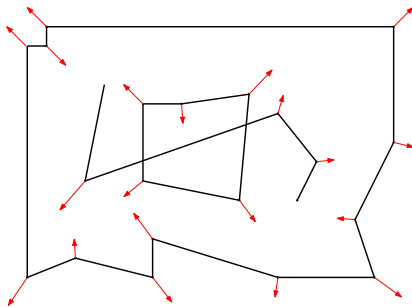


Observation

Consider n random motorcycles within S and let $h \in \Theta(\sqrt{n})$. A motorcycle trace has a mean length proportional to $1/\sqrt{n}$. Hence, it intersects $O(1)$ cells in average. This leads to an $O(n \log n)$ expected runtime.

Experimental setup

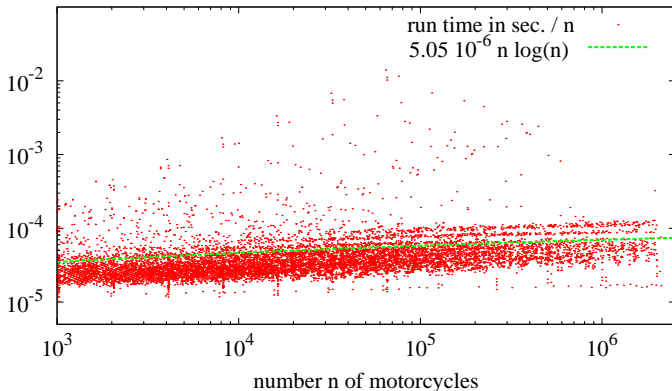
- A data set consist of polygonal chains.
- For every inner vertex of a chain, we define a motorcycle in “straight skeleton” manner. The chains are considered being walls.
- We ran our implementation MOCA on 22 000 thousand data sets, consisting of real-world¹ data and contrived data.



¹Medical scans, GIS maps, outlines of fonts, CAD/CAM models, etc. >

Experimental results

Actual runtime in seconds divided by the number n of motorcycles for each of the 22 000 data sets.



Least-square fit reveals an average run time of $5.05 \cdot 10^{-6} n \log n$ seconds on our computer.

- Easy-to-implement algorithm.
- Surprisingly good performance and competitive in practice.
- Algorithm can be extended easily to more general motorcycle graph problems: motorcycles running out of fuel, curved traces, partial/temporal/conditional invisible traces, etc.

