

Concentration bounds for RIC^2 (Randomized Incremental Construction)

Sandeep Sen

IIT Delhi, India

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²to appear in STACS'19

- 1 Introduction
- 2 A Martingales based framework and Quicksort
- 3 More complex settings
- 4 Lower bounds

Once upon a time ..



Conceived : 1986

Forgotten: 1993

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Randomized Incremental Construction (RIC)

Starting from an empty set

Repeat:

- 1 *Insert the next object*
- 2 *Update the partial construction (data-structures)*

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Expected total time = $\max_{input\ I} \mathbb{E}[T_s(I)]$
 (worst case for any input of size n).

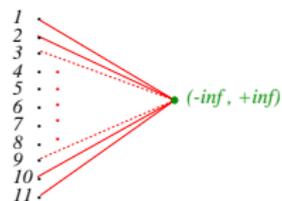
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Quicksort as R.I.C.

Gradual refinement of partition

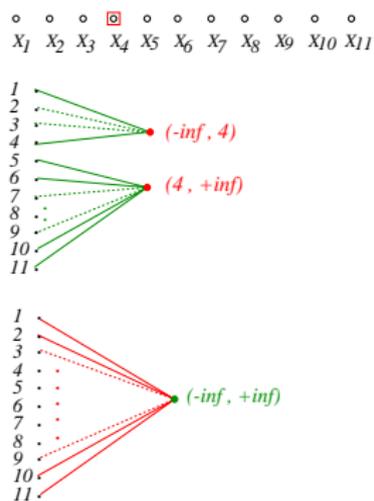
x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_8 x_9 x_{10} x_{11}

1 .
 2 .
 3 .
 4 .
 5 .
 6 .
 7 .
 8 .
 9 .
 10 .
 11 .



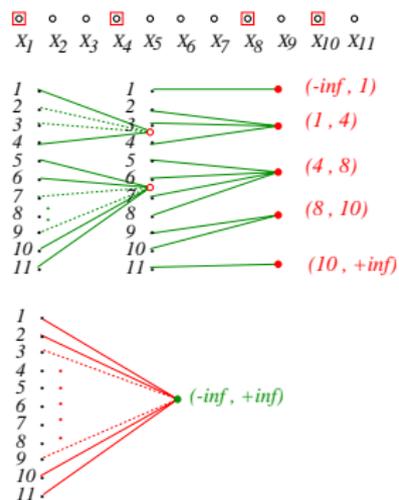
Quicksort as R.I.C.

Conflict graph



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Bounding the maximum sub-problem size

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$l(\sigma)$: size of the subproblem (unchosen elements in σ)

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Example: Quicksort

$$\Pi(n) = \binom{n}{2} \text{ pairs of points}$$

A subproblem is defined by a pair of sample points

$D(\sigma)$ end-points of σ

$I(\sigma)$: unsampled points in σ

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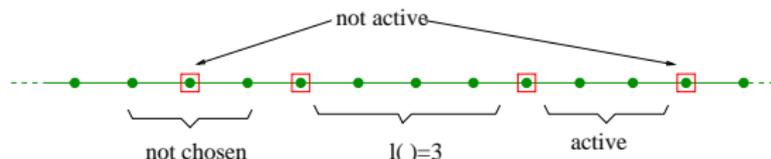
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$\Pi^0(n)$ special significance : $l(\sigma) = 0$

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Bounding the maximum sub-problem

Claim :

$$\Pr\left\{\max_{\text{active } \sigma} l(\sigma) \geq c \frac{n}{r} \log r\right\} \leq \frac{1}{2}$$

R chosen by Bernoulli sampling $p = r/n$

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$p(\sigma, r)$: conditional probability that none of the $k (= l(\sigma))$ elements are selected given $D(\sigma)$ chosen

$$\leq (1 - r/n)^k$$

$$\leq e^{-c \log r} = 1/r^c \quad \text{for } \boxed{k \geq cn/r \ln r}$$

BAD σ

¹³to appear in STACS'19

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$$\begin{aligned} &\leq \frac{1}{r^c} \sum_{\sigma \in \Pi(n)} \Pr[D(\sigma) \subset R] = \frac{1}{r^c} \sum_{\sigma \in \Pi(n)} E[D(\sigma) \subset R] \\ &= \frac{1}{r^c} E[\text{number sub-problems for which } D(\sigma) \subset R] \\ &\quad \text{(linearity of Expectation)} \\ &= r^{O(1)} \text{ for } D(\sigma) = O(1) \end{aligned}$$

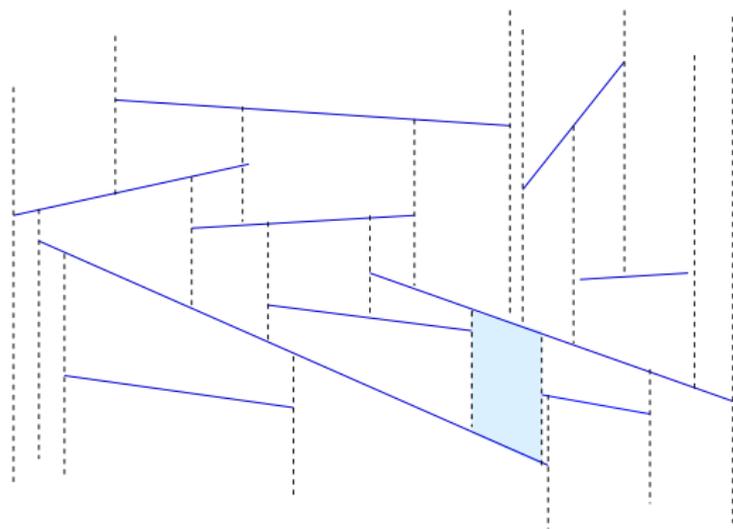
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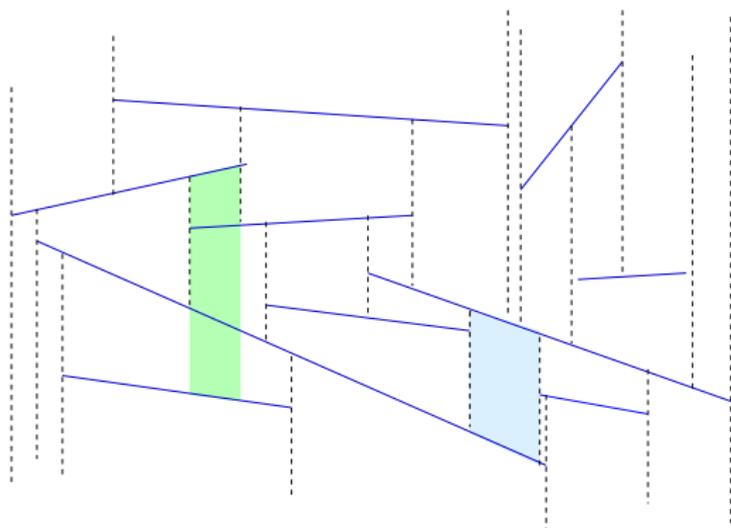
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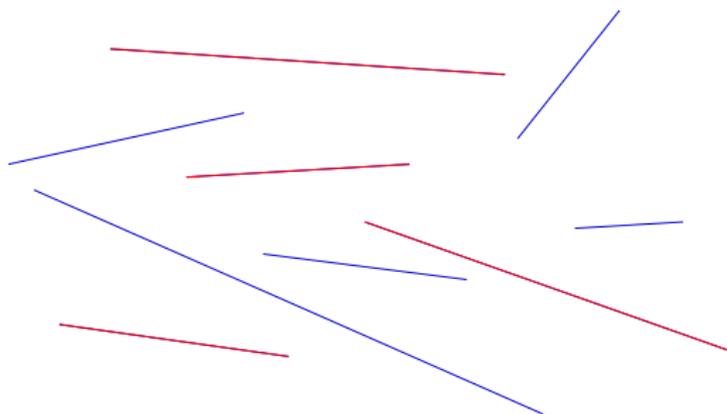
More general : Trapezoidal Map



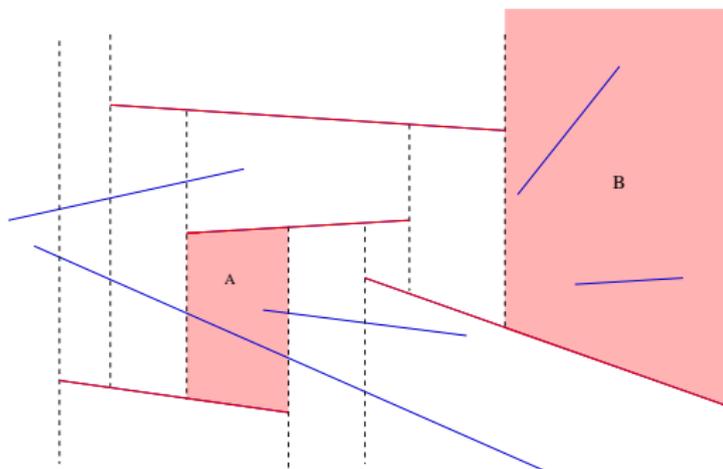
Trapezoidal Map : Ranges



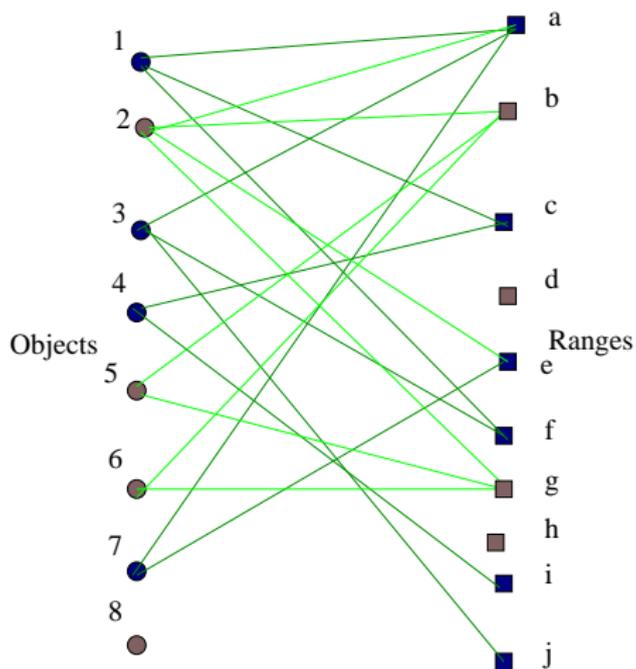
Randomized Incremental Construction



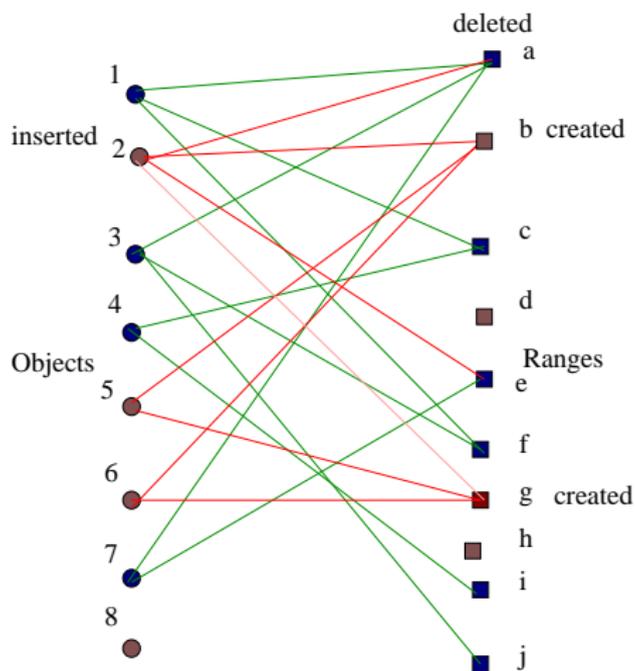
Objects (segments) and ranges (trapezoids)



A more general scenario



Modifications caused by insertion of an object



A general bound for RIC [CI-Sh]

Total (amortised) cost = $O(\text{edges created in conflict graph})$

Edges can be deleted at most once

General Step: $R \leftarrow R \cup s$ (both random subsets)

Expected work (#edges created in the conflict graph) =

$$\sum_{\sigma \in \Pi^0(R \cup s)} I(\sigma) \cdot \Pr\{\sigma \in \Pi^0(R \cup s) - \Pi^0(R)\}$$

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From *backward analysis* this probability is the same as deleting a random element from $R \cup s$ which is $\frac{d(\sigma)}{r+1}$. Substituting

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Sum bound

$$\sum_{\sigma \in \Pi^0(R \cup s)} I(\sigma) = \frac{n}{r} \cdot E[\Pi^0(R \cup s)]$$

A general bound on Expected running time of RIC

$$= O\left(\frac{d(\sigma)}{r} \cdot \frac{n}{r} E[\Pi^0(R \cup s)]\right)$$

A common scenario $E[\Pi^0(R)] = O(r)$.

$$\text{Total expected cost of RIC} = \sum_{r=1}^{r=n} O\left(\frac{d}{r} \cdot n\right)$$

$= O(n \log n)$ (also applicable to convex hulls)

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Tail estimates in the general case

²⁵to appear in STACS'19

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Tail estimates in the general case **without independent repetitions**

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Sum of subproblem sizes

Def: c -order conflict $\binom{l(\sigma)}{c}$, for some $c \geq 0$

Let $T_c = \sum_{\sigma \in \Pi^0(R)} \binom{l(\sigma)}{c}$

Remark For technical reasons it is not $l(\sigma)^c$. $T_0 = |\Pi^0(R)|$. $T_1 =$ sum of subproblems.

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Claim $E[T_c] = O\left(\left(\frac{n}{r}\right)^c E[\Pi^c(R)]\right)$

For constant c , $E[\Pi^c(R)] = O(E[\Pi^0(R)])$ implying that average conflict size is very close to $\frac{n}{r}$

Sum of subproblem sizes

$$T_c = \sum_{\sigma \in \Pi(N)(R)} \binom{l(\sigma)}{c} I_{\sigma,R} \text{ where } I_{\sigma,R} = 1 \text{ if } \sigma \in \Pi^0(R).$$

$$E[T_c] = \sum_{\sigma \in \Pi(N)} \binom{l(\sigma)}{c} p^{d(\sigma)} \cdot (1-p)^{l(\sigma)} \text{ for } l(\sigma) \geq c.$$

$$= \sum_{\sigma \in \Pi(N)} \binom{l(\sigma)}{c} p^{d(\sigma)+c} \cdot (1-p)^{l(\sigma)-c} \cdot \left(\frac{1-p}{p}\right)^c$$

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since $\Pr\{\sigma \in \Pi^c(R)\} = \Pr\{d(\sigma) \text{ objects chosen and } c \text{ out of } l(\sigma) \text{ not chosen}\}$

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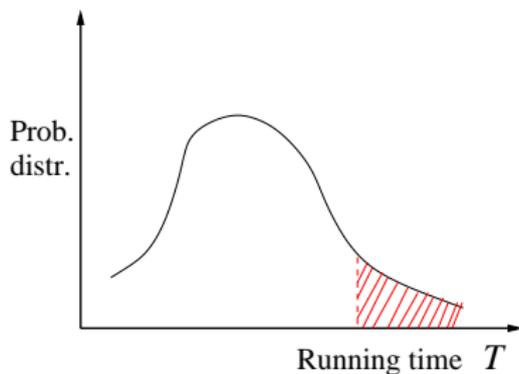
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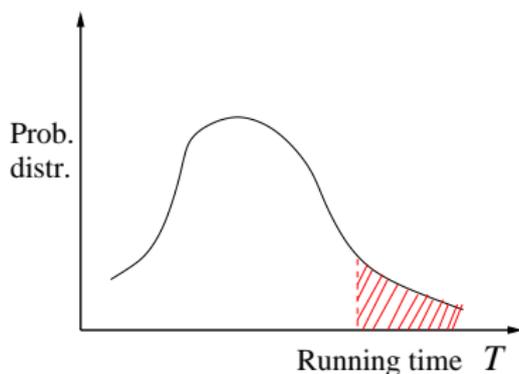
Standard Tools

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Probabilistic Inequalities :

- Markov $\Pr[X > kE[X]] < 1/k$
- Chernoff $\Pr[X > A] \leq G_X(s) \cdot s^{-A}$
- Chebychev $\Pr[|X - E[X]| > r] \leq \sigma^2/r^2$

Notation : $\tilde{O}(\cdot) \stackrel{\text{def}}{=} O(\cdot)$ with prob. $1 - \frac{1}{n}$ Inv polynomial

³¹to appear in STACS'19

A martingale framework

(Ω, \mathcal{U}) : all permutations of n objects \mathcal{U} uniform probability distribution.
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Deviation from Expectation

Z_0 : expected running time of ric and Z_n : actual running time

Concentration bound : $\Pr[|Z_0 - Z_n| \geq \lambda] < ???$

³³to appear in STACS'19

A martingale inequality

Theorem [Freedman 75]

Let $X_1, X_2 \dots X_n$ be a sequence of random variables and let Y_k , a function of $X_1 \dots X_k$ be a martingale sequence, i.e., $\mathbb{E}[Y_k | X_1 \dots X_k] = Y_{k-1}$ such that $\max_{1 \leq k \leq n} \{|Y_k - Y_{k-1}|\} \leq M_n$. Let

$$W_k = \sum_{j=1}^k \mathbb{E}[(Y_j - Y_{j-1})^2 | X_1 \dots X_{j-1}] = \sum_{j=1}^k \text{Var}(Y_j | X_1 \dots X_{j-1})$$

where Var is the variance using $\mathbb{E}[Y_j] = Y_{j-1}$. Then for all λ and $W_n \leq \Delta^2$, $\Delta^2 > 0$,

$$\Pr[|Y_n - Y_0| \geq \lambda] \leq 2 \exp\left(-\frac{\lambda^2}{2(\Delta^2 + M_n \cdot \lambda/3)}\right)$$

Martingale inequality

Extended Freedman inequality

Let $\Pr[\max_{1 \leq k \leq n} \{|Y_k - Y_{k-1}|\} \geq M_n, W_n \geq \Delta^2] \leq \frac{1}{f(n)}$, then,

$$\Pr[|Y_n - Y_0| \geq \lambda] \leq 2 \exp\left(-\frac{\lambda^2}{2(\Delta^2 + M_n \cdot \lambda/3)}\right) + O(1/f(n))$$

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Azuma-Hoeffding

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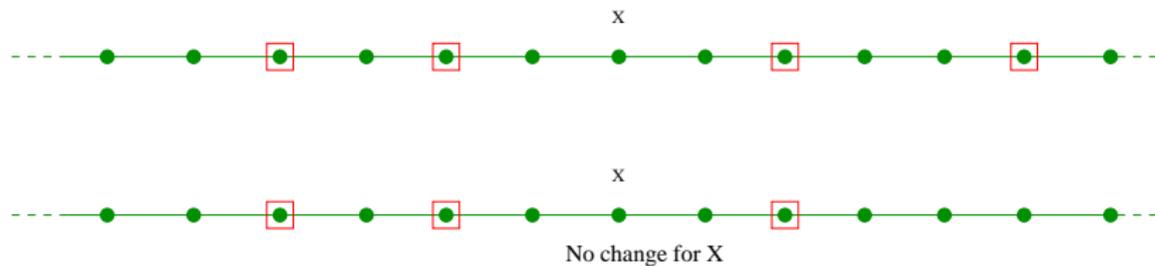
Mehlhorn-Sharir-Welzl [93] obtained a special kind of Martingale concentration bound that was effective for some cases of line segment intersection RIC.

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Application to quicksort

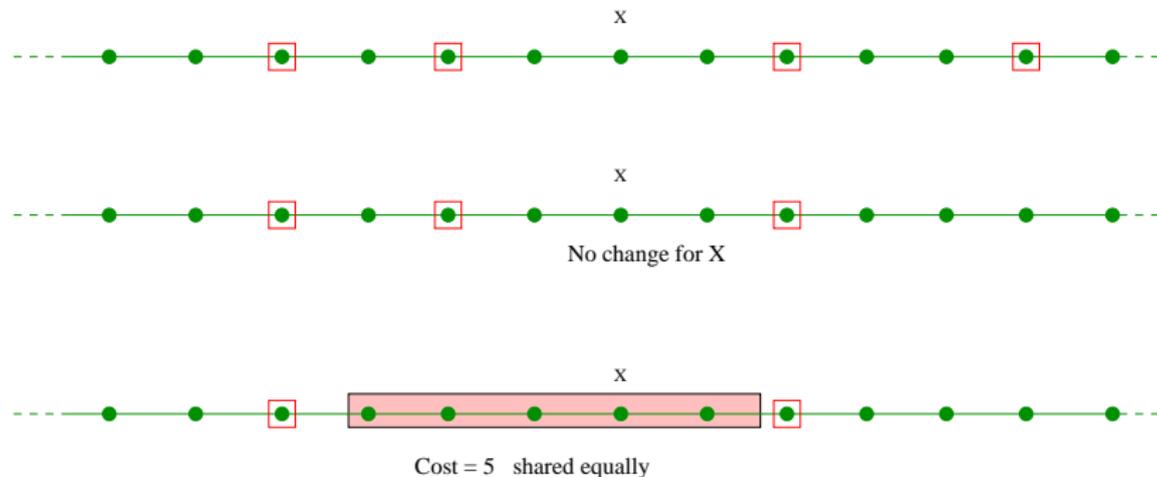


Application to quicksort



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Application to quicksort



$$I_j^x = \begin{cases} 1 & \text{if interval containing } x \text{ changes in step } j \\ 0 & \text{otherwise.} \end{cases}$$

Analysis

$$\mathbb{E}\left[\sum_j I_j^x\right] = \sum_j \Pr[I_j^x = 1] = \sum_j \frac{2}{j} = \log n$$

Analysis

$$\mathbb{E}\left[\sum_j l_j^x\right] = \sum_j \Pr[l_j^x = 1] = \sum_j \frac{2}{j} = \log n$$

$M_n = 1$ since maximum charge on X is 1 for any pivot.

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$$\begin{aligned} Y_j - Y_{j-1} &= w(v_{j-1}, v_j) + \left(\sum_{k=j+1}^n \mathbb{E}[I'_k] \right) - \left(\sum_{k=j}^n \mathbb{E}[I_k] \right) \\ &= I_j - E[I_j] \text{ assuming } I_j, I'_j \text{'s have the same distribution} \end{aligned}$$

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$$\mathbb{E}\left[\sum_j I_j^X\right] = \sum_j \Pr[I_j^X = 1] = \sum_j \frac{2}{j} = \log n$$

$M_n = 1$ since maximum charge on X is 1 for any pivot.

$$\begin{aligned} Y_j - Y_{j-1} &= w(v_{j-1}, v_j) + \left(\sum_{k=j+1}^n \mathbb{E}[I'_k] \right) - \left(\sum_{k=j}^n \mathbb{E}[I_k] \right) \\ &= I_j - E[I_j] \text{ assuming } I_j, I'_j \text{'s have the same distribution} \end{aligned}$$

$$\begin{aligned} \mathbb{E}_j[(I_j - E[I_j])^2] &= \mathbb{E}_j[I_j^2] - \mathbb{E}_j^2[I_j] \leq \mathbb{E}_j[I_j^2] - \frac{4}{(n-j)^2} \\ &\leq \frac{2}{n-j} I_j^2 \text{ also 0-1 indicator rv} \end{aligned}$$

$$\sum_{j=1}^n \mathbb{E}_{j-1}[(Y_j - Y_{j-1})^2] \leq \sum_{j=1}^n \frac{2}{n-j} \leq 2 \log n = W_n$$

Inverse polynomial bound for quicksort

$$\Pr[|Y_n - Y_0| \geq c \log n] \leq \exp\left(-\frac{4c^2 \log^2 n}{2(\log n + c \log n/3)}\right) \leq \frac{1}{n^c}.$$

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⁴²to appear in STACS'19

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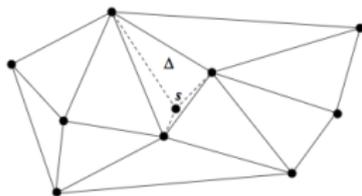
Azuma-Hoeffding

$$\Pr[|Y_n - Y_0| \geq t] \leq \exp\left(\frac{-t^2}{\sum_{i=1}^n M_i^2}\right)$$

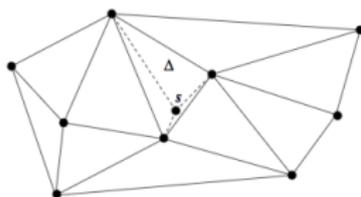
For $M_i = 1$, there is no meaningful bound for this setting.

⁴²to appear in STACS'19

Incremental Delaunay triangulation



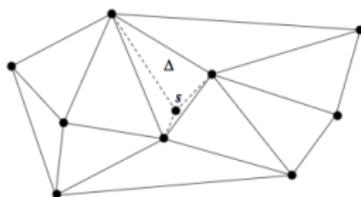
Incremental Delaunay triangulation



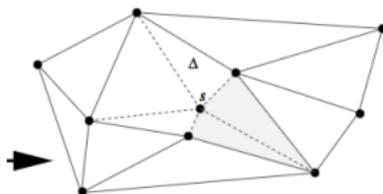
Total size is $O(i)$ but the degree of a vertex can be large

⁴⁴to appear in STACS'19

Incremental Delaunay triangulation



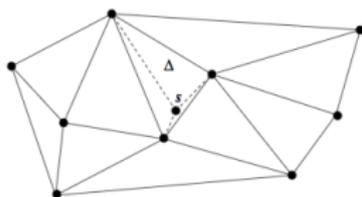
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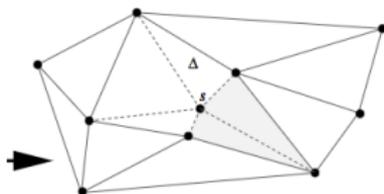
Number of new Δ 's is 5 and can be as large as i but ..

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Incremental Delaunay triangulation



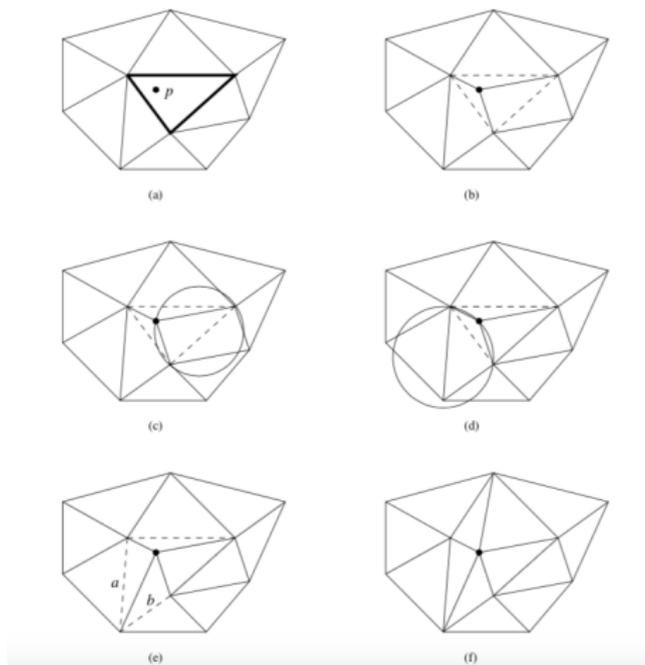
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Number of new Δ 's is 5 and can be as large as i but ..
the average degree of a planar graph is $O(1)$.

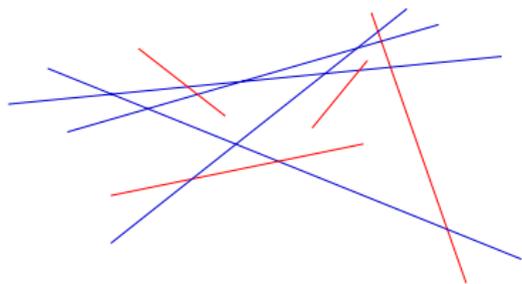
Expected number of triangles that appear over the course of RIC = $O(n)$
⁴⁴ to appear in STACS'19

Flip algorithm [Green and Sibson]



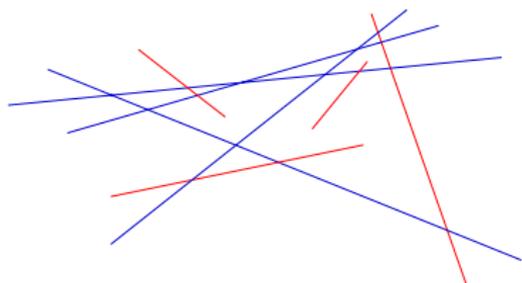
Courtesy: Lischinski[93]

Line segment intersections

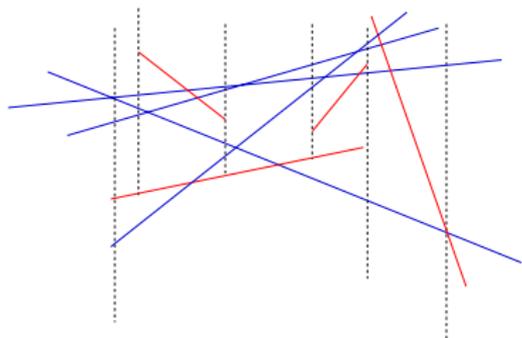


No intersections between red. All blue segments intersect.

Line segment intersections



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Intermediate structures (trapezoidal maps) can have huge variance

Expected number of intersections at stage i

Probability intersection (s_j, s_k) appear in stage $i =$ Probability s_j, s_k have been chosen.

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Ideal bound for segment intersections : $O(m + n \log n)$

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Summary of results

Problem	Exp Run time	Tail estimates
Quicksort	$O(n \log n)$	$O(c\gamma n \log n)$ w.p. $\geq 1 - n^{-c}$
Delaunay Triangulation	$O(n \log n)$	$O(c\gamma n \log n)$ w.p. $\geq 1 - 2^{-c}$
Segment intersections/ Trapezoidal maps	$O(n \log n + m)$ *no conflict list*	$O(n \log n + m)$ w.p. $1 - \exp - \left(\frac{m+n \log n}{n\alpha(n)} \right)$
	*using conflict list**	w.p. $\geq 1 - \exp - \left(\frac{m}{n \log^2 n} \right)$

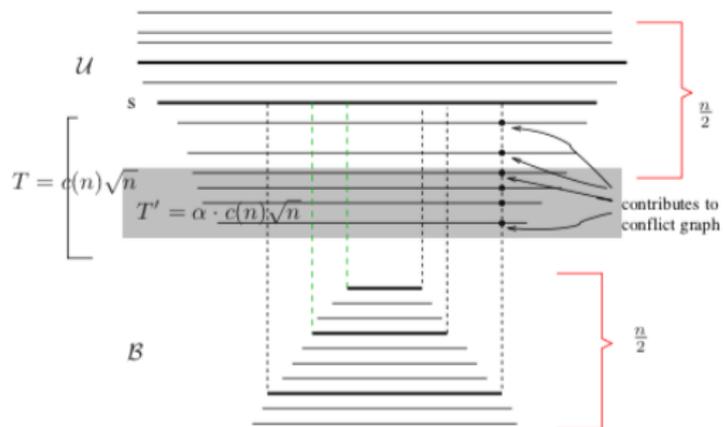
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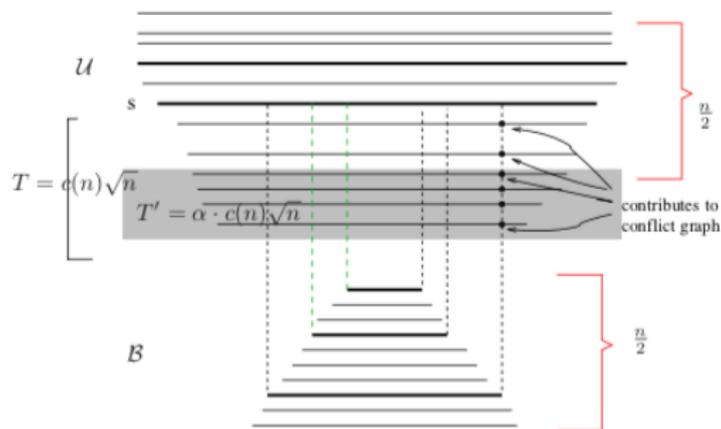
The non-conflict bound is nearly inverse polynomial for $m = 0$, i.e.,
 $\exp - (\log n / \alpha(n))$

⁵¹to appear in STACS'19

A lower bound construction



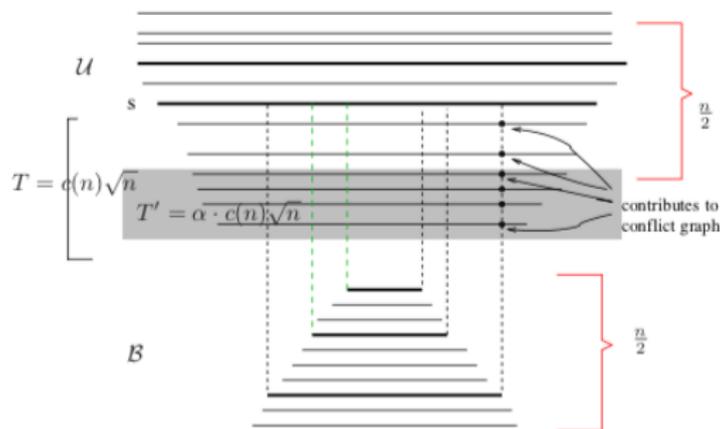
A lower bound construction



$$|\mathcal{U}| = |\mathcal{B}| = \frac{n}{2}$$

Among first $3\sqrt{n}$ insertions, $\Pr[\mathcal{B} \geq \sqrt{n}] \geq 1 - 2^{-\sqrt{n}}$.

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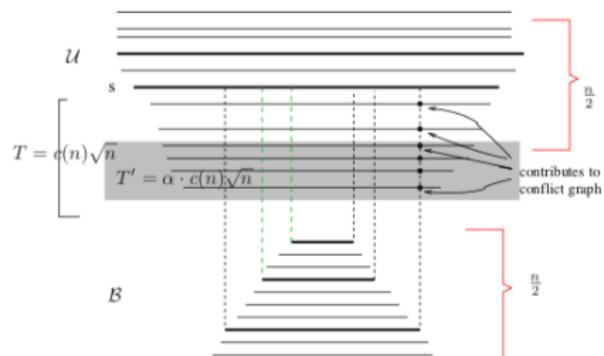


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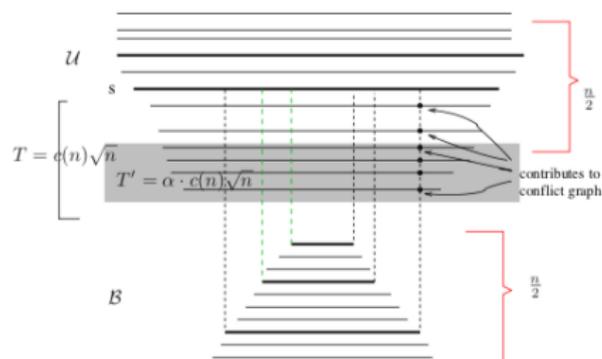
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$$\Pr[|T| \geq c(n)\sqrt{n}] \geq 4^{-3c(n)}$$

Construction cont.

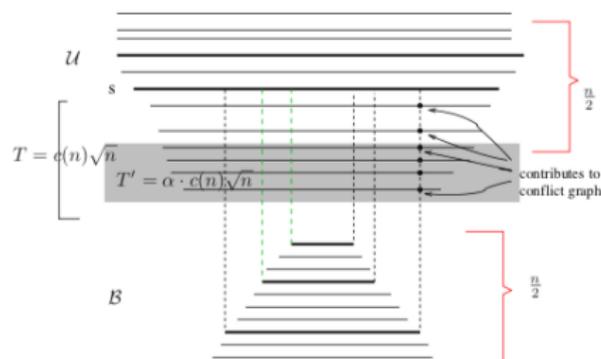


Construction cont.



Number of times s moves down without touching T' (shaded region) = r
 \Rightarrow RIC incurs $r \cdot c(n)\sqrt{(n)}\sqrt{(n)}$ cost

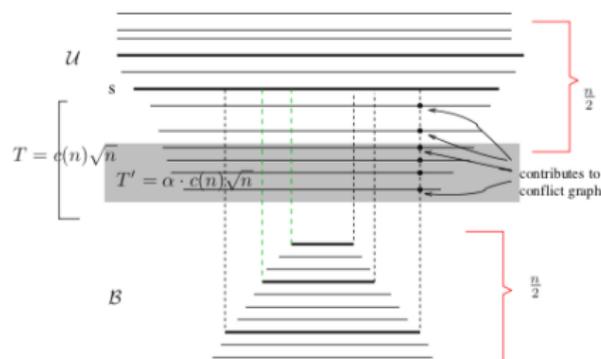
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$$\Pr[X \geq n \log n \log \log n] \geq \frac{1}{\sqrt{n}} \quad \text{for } c(n) = \Omega(\log n)$$

Future directions

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- Inverse polynomial bounds open for Delaunay, Line segments and many others

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Future directions

- Inverse polynomial bounds open for Delaunay, Line segments and many others
- More lower bound constructions
- Distinct variations of RICs (rebuild ?) may have different performance including data structures like conflict lists (without).

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