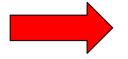
A Deterministic Algorithm for Summarizing Asynchronous Streams over a Sliding Window

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Outline of Talk



Introduction

Algorithm

Analysis

Time

Data stream: (v_1) (v_2) (v_3) (v_4) (v_5)

$$V_1$$



$$t_3$$

For simplicity assume unit valued elements

Most recent time window of duration Wtime Data stream: (v_1) (v_2) (v_3) (v_4) (v_5)

Goal: Compute the sum of elements with time stamps in time window [C-W,C]

$$\sum_{C-W \leq t_i \leq C} V_i$$

Example I: All packets on a network link, maintain the number of different ip sources in the last one hour

Example II: Large database, continuously maintain averages and frequency moments

Synchronous stream

ti: In ascending order

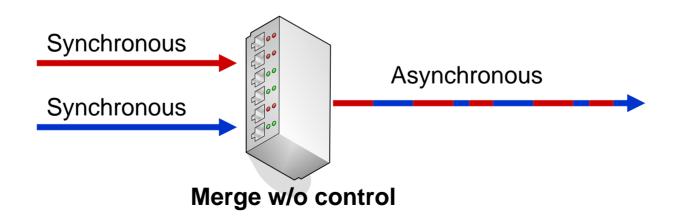
Asynchronous stream

ti: No order guaranteed

Why Asynchronous Data Streams?



Network delay & multi-path routing



Processing Requirements:

- ·One pass processing
- ·Small workspace: poly-logarithmic in the size of data
- ·Fast processing time per element
- · Approximate answers are ok

Our results:

A deterministic data aggregation algorithm

Time:
$$O\left(\log B + \frac{\log W}{\varepsilon}\right)$$

Space:
$$O(\log B \log W \frac{\log W + \log B}{\varepsilon})$$

Relative Error:
$$\varepsilon = \frac{|X - S|}{S}$$

Previous Work:

[Datar, Gionis, Indyk, Motwani. SIAM Journal on Computing, 2002]

Deterministic, Synchronous Merging buckets

[Tirthapura, Xu, Busch, PODC, 2006]

Randomized, Asynchronous

Random sampling

Outline of Talk

Introduction



Algorithm

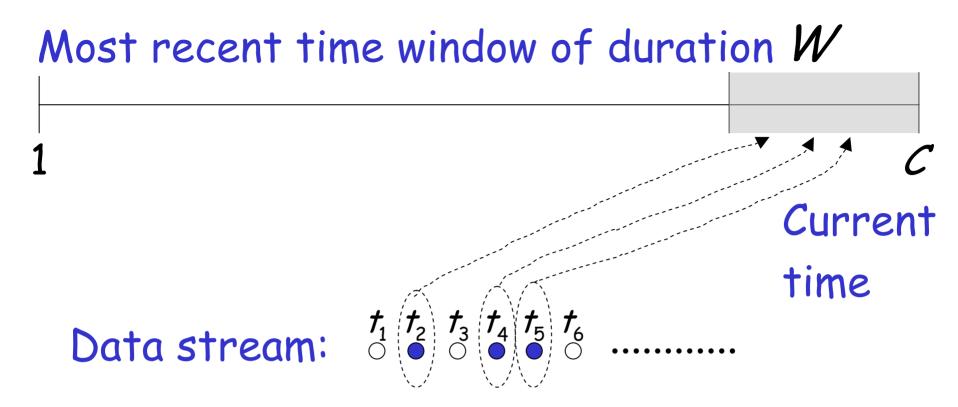
Analysis

Time -

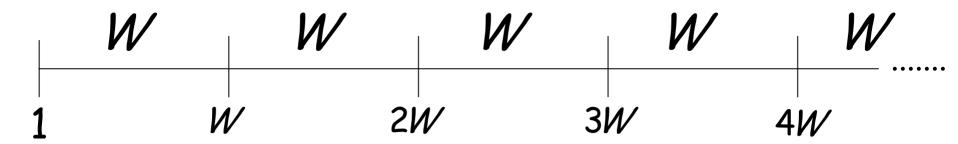
1

Current time

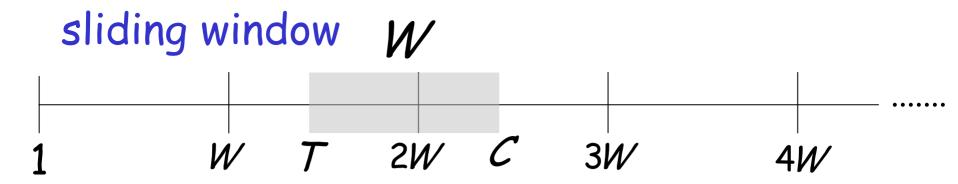
For simplicity assume unit valued elements



Goal: Compute the sum of elements with time stamps in time window [C-W,C]



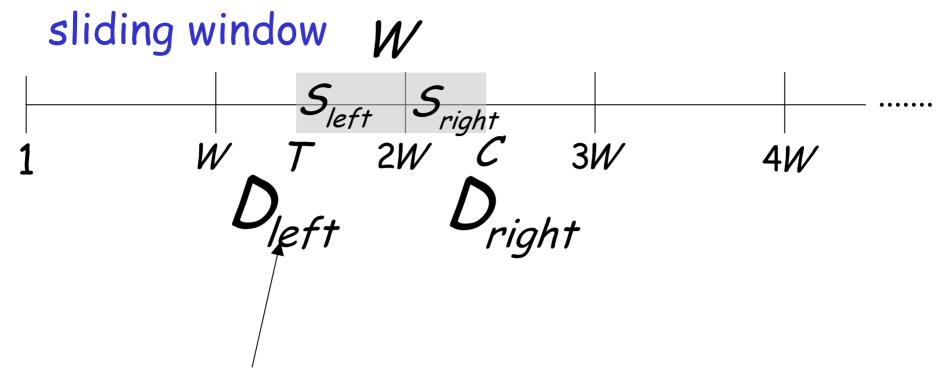
Divide time into periods of duration W



The sliding window may span at most two time periods

$$S = S_1 + S_2$$

Sum can be written as two sub-sums In two time periods



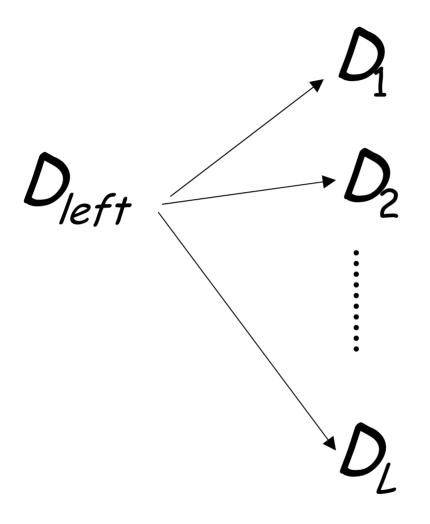
Data structure that maintains an estimate of S_{left} In left time period

 $\frac{\mathcal{S}_{left}}{\mathcal{T}}$

 D_{left}

Without loss of Generality, Consider data structure D_{left} in time period [1, W]

Data structure consists of various levels



 2^{L} is an upper bound of the sum in a period

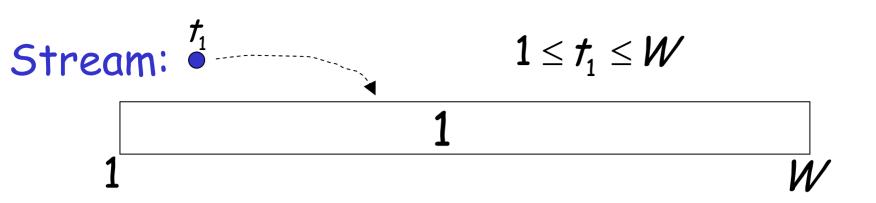
Consider level D_i

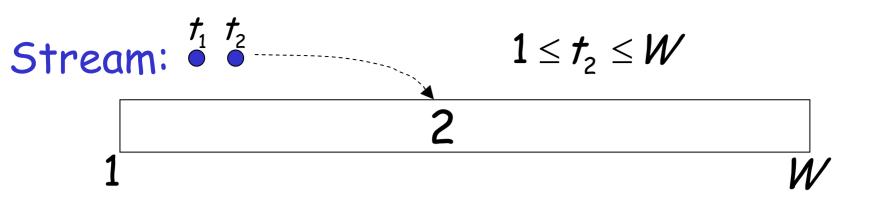
Bucket at Level
$$i+1$$

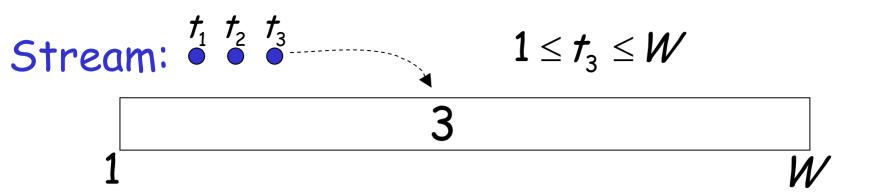
$$0$$

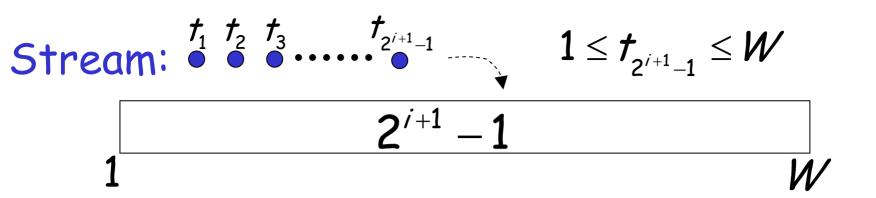
$$1 \leftarrow Time period \longrightarrow W$$

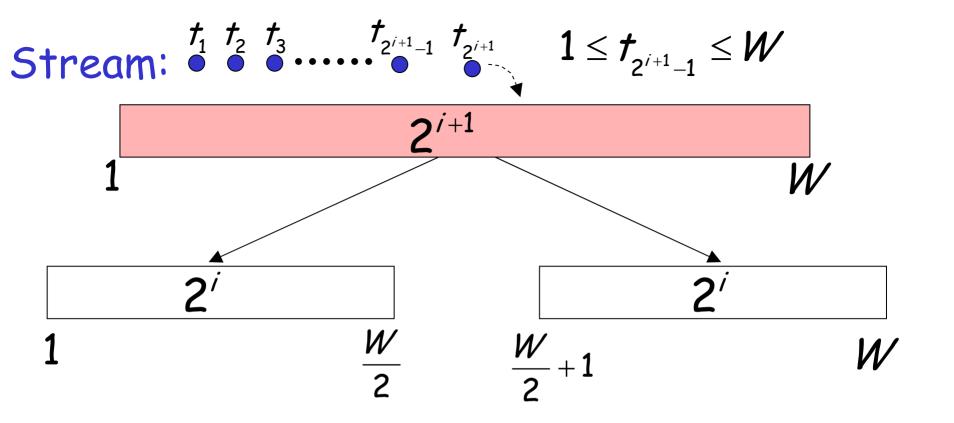
Counts up to 2^{i+1} elements











Split bucket

Counter threshold of 2^{i+1} reached

Stream:
$$t_1 t_2 t_3 t_4 t_{2^{i+1}-1} t_{2^{i+1}-1} \le W$$

$$\begin{array}{c|cccc}
2^{i} & & & 2^{i} \\
\hline
1 & & \frac{W}{2} & & \frac{W}{2} + 1
\end{array}$$

New buckets have threshold also 2^{i+1}

Stream:
$$t_1, t_2, t_3, \dots, t_{2^{i+1}-1}, t_{2^{i+1}}, t_{2^{i+1}+1}$$

$$1 \le t_{2^{i+1}+1} \le \frac{W}{2}$$

$$1 \le t_{2^{i+1}+1} \le \frac{W}{2}$$

$$1 \le t_{2^{i+1}+1} \le \frac{W}{2}$$

Stream:
$$t_1, t_2, t_3, \dots, t_{2^{i+1}-1}, t_{2^{i+1}}, t_{2^{i+1}+1}, t_{2^{i+1}+2}, \dots, \frac{W}{2} \le t_{2^{i+1}+2} \le W$$

$$2^{i} + 1$$

$$1$$

$$\frac{W}{2}$$

$$\frac{W}{2} + 1$$

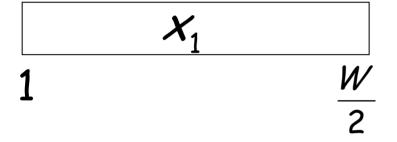
$$W$$

Stream:
$$t_1 t_2 t_3 t_{2^{i+1}-1} t_{2^{i+1}} t_{2^{i+1}+1} t_{2^{i+1}+2} t_{2^{i+1}+3} 1 \le t_{2^{i+1}+3} \le \frac{W}{2}$$

$$2^{i} + 2 2^{i} + 1 W$$

Stream:
$$\overset{t_1}{\bullet}$$

$$\frac{W}{2}+1\leq t_{m}\leq \frac{W}{2}$$



$$\frac{2^{i+1}}{2}W$$

$$2^{i}$$

$$2^{i}$$

$$2^{i}$$

Split bucket

$$\frac{W}{2}+1$$
 $\frac{3W}{4}$ $\frac{3W}{4}+1$

Stream: $\overset{t_1}{\bullet}$.

*t*_n

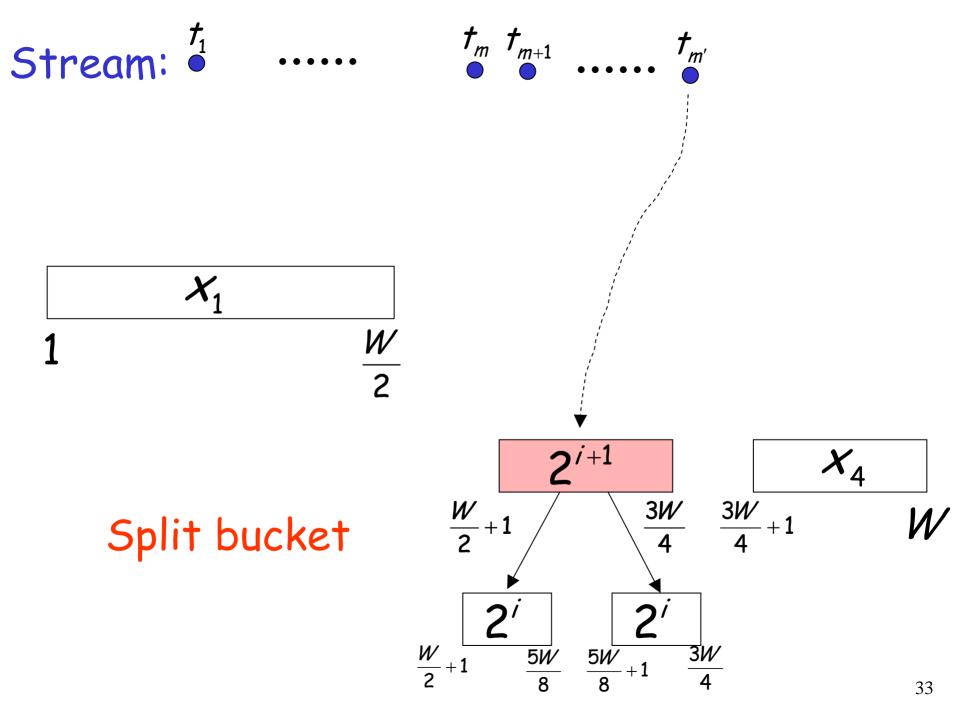
$$\frac{X_1}{2}$$

Stream:
$$t_1 t_m t_{m+1} \frac{W}{2} + 1 \le t_{m+1} \le \frac{3W}{4}$$

$$\frac{X_1}{2}$$

$$\frac{W}{2} + 1 \frac{2^i}{4}$$

$$\frac{W}{2} + 1 \frac{3W}{4} \frac{3W}{4} + 1 W$$



Stream: ^{t₁}

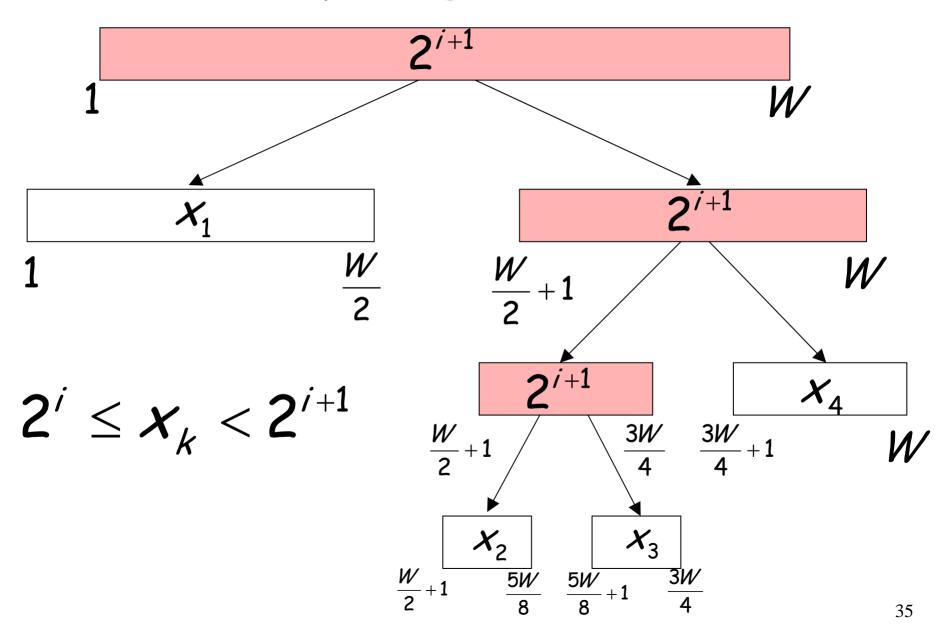
 t_m t_{m+1} $t_{m'}$

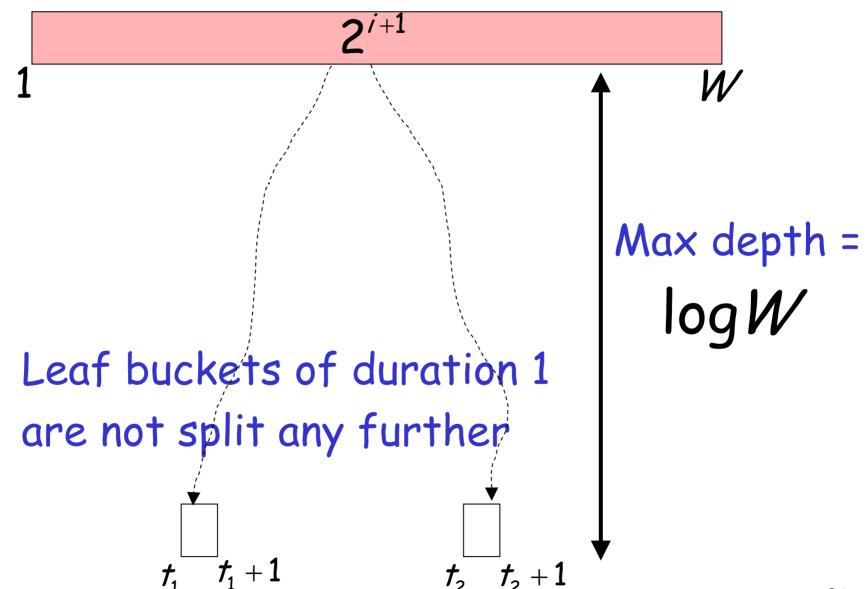
$$\frac{x_1}{1}$$

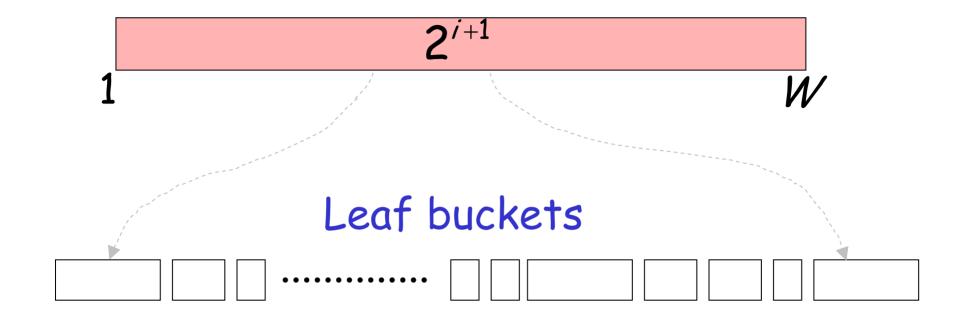
 $\frac{3W}{4}+1$ W

$$\frac{\mathcal{U}}{2} + 1 \qquad \frac{5\mathcal{W}}{8} \qquad \frac{5\mathcal{W}}{8} + 1 \qquad \frac{3\mathcal{W}}{4}$$

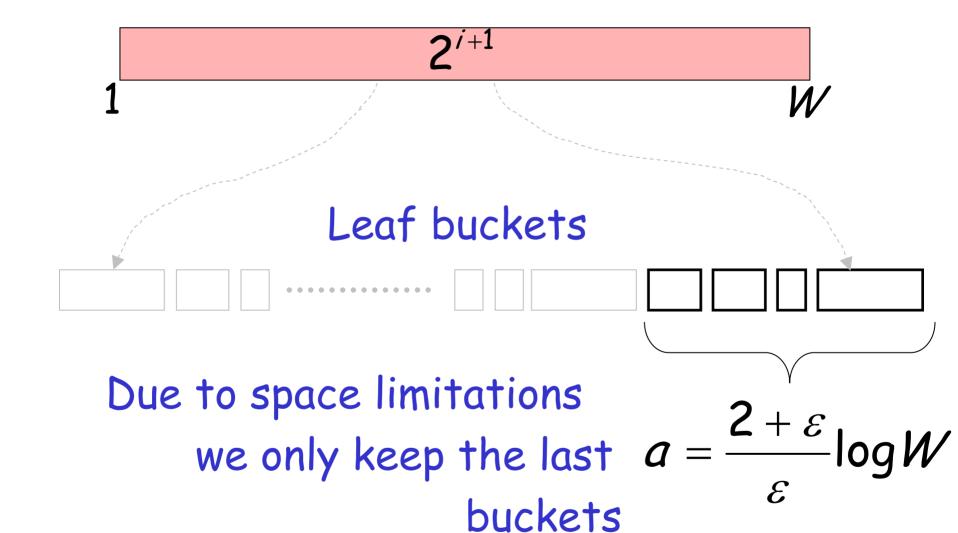
Splitting Tree





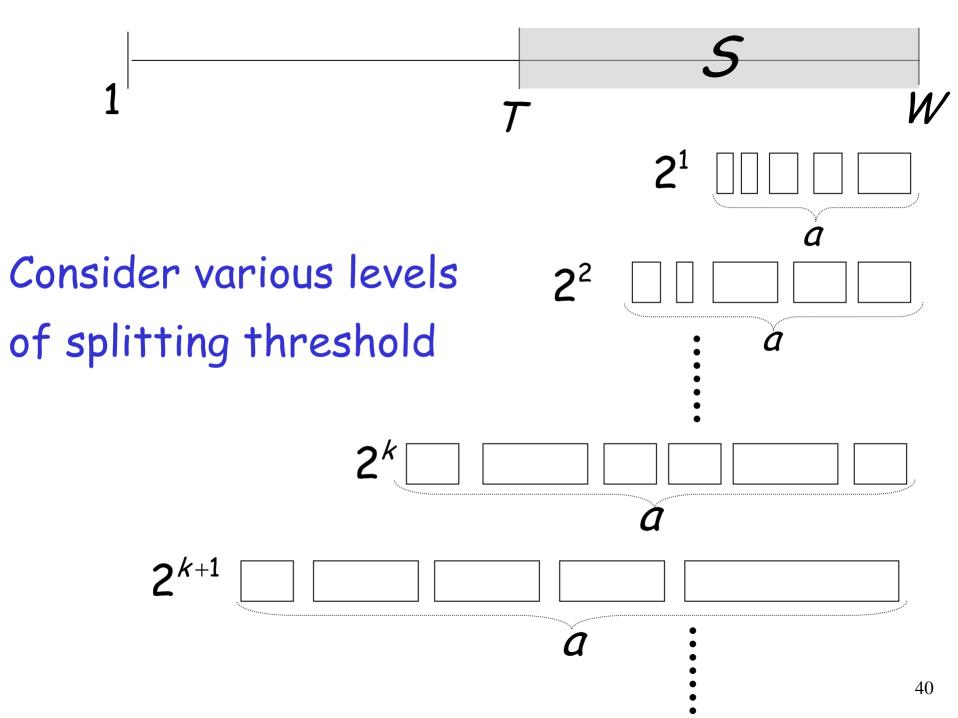


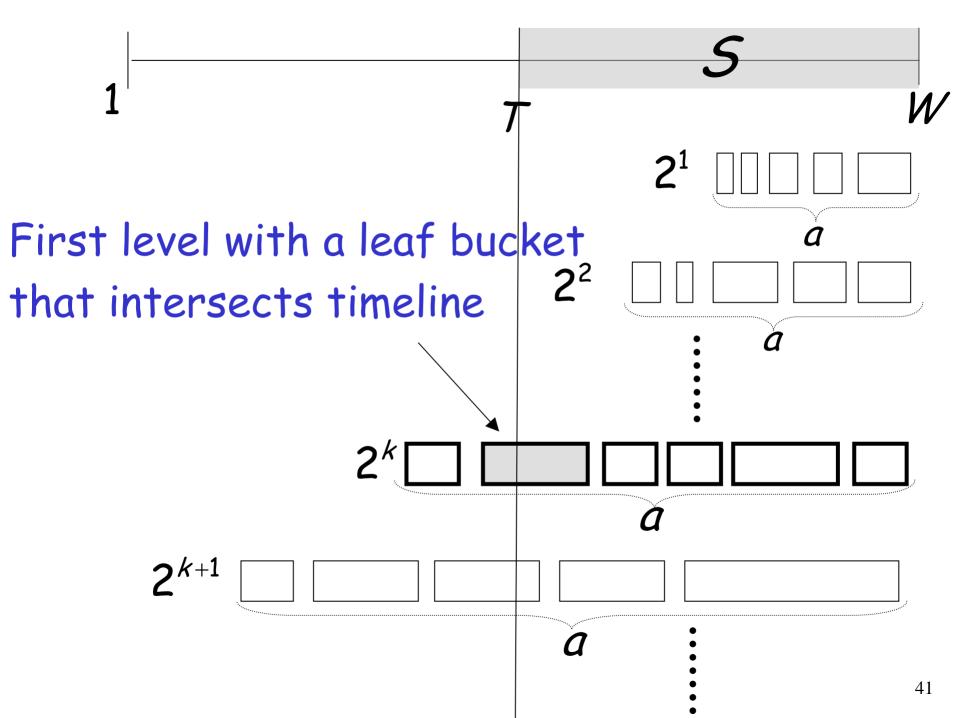
The initial bucket may be split into many buckets



 $\frac{5}{\tau}$

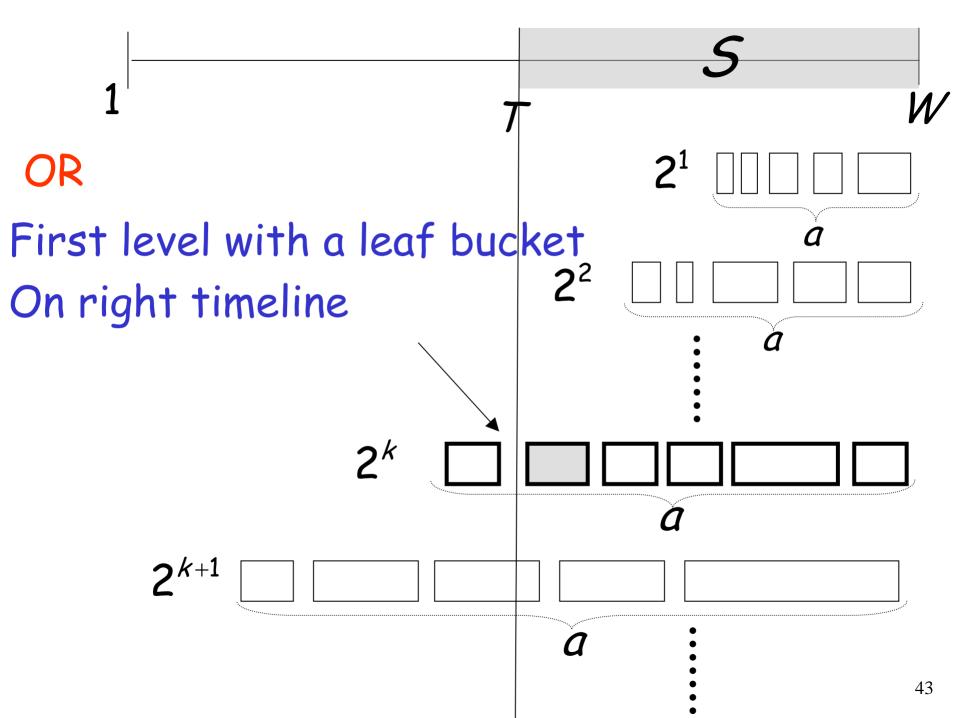
Suppose we want to find the sum S of elements in time period [T,W]





Estimate of S: $X = X_1 + X_2 + \cdots + X_z$

Consider buckets on right of timeline



Outline of Talk

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Algorithm



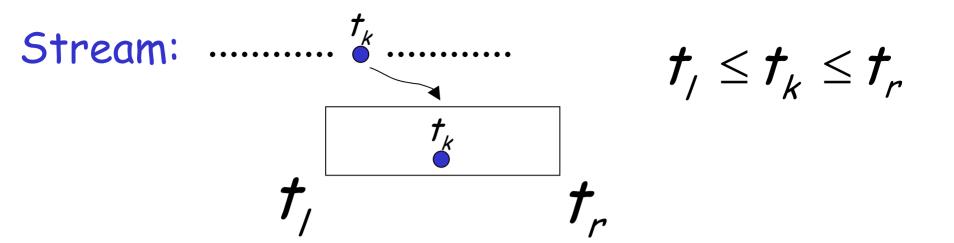
 $\frac{5}{\tau}$

Suppose that we use level 2^{i+1} in order to compute the estimate

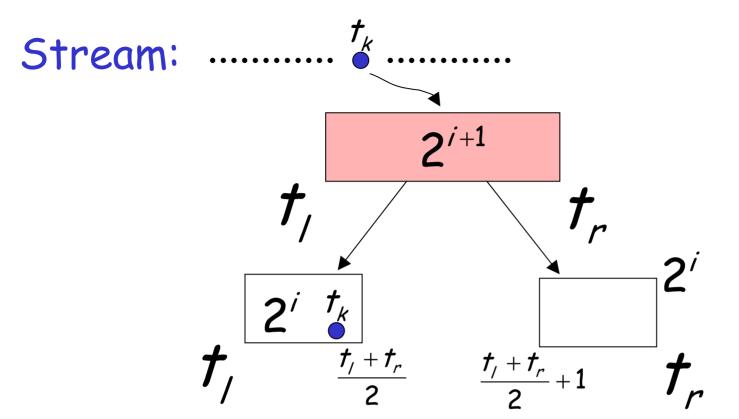
Stream:
$$x_b \rightarrow x_b + 1$$
 $t_b \rightarrow t_b$

Consider splitting threshold level 2^{r+1}

A data element is counted in the appropriate bucket



We can assume that the element is placed in the respective bucket



We can assume that when bucket splits the element is placed in an arbitrary child bucket

If:
$$t_1 \le t_k \le \frac{t_1 + t_r}{2}$$
 GOOD!

Element counted in correct bucket

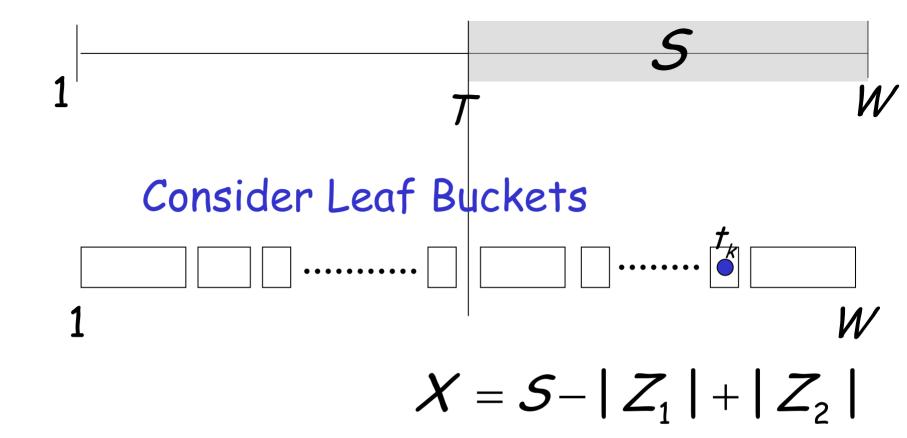
If:
$$\frac{t_{/}+t_{r}}{2}+1\leq t_{k}\leq t_{r}$$
 BAD!

Element counted in wrong bucket

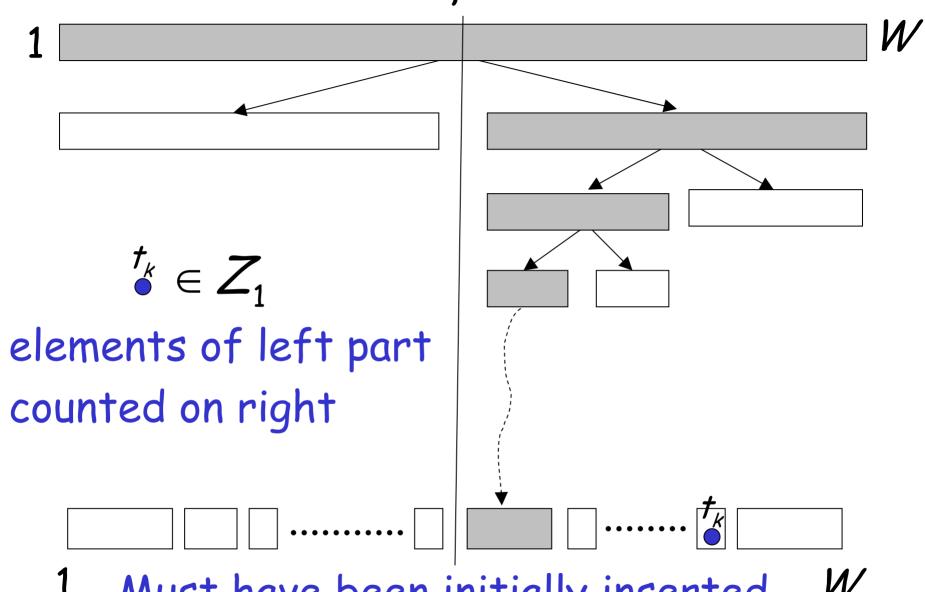
If
$$T \leq t_k \leq W$$
 GOOD!

If
$$t_k < T$$
 BAD!

Element counted in wrong bucket



- Z_1 : elements of left part counted on right
- Z_2 : elements of right part counted on left



Must have been initially inserted W in one of these buckets

Since tree depth $\leq \log W$

$$|Z_1| = \mathcal{O}(2^i \log W)$$

Since tree depth $\leq \log W$

$$|Z_1| = \mathcal{O}(2^i \log W)$$

Similarly, we can prove

$$|Z_2| = O(2^i \log W)$$

Therefore:
$$|X - S| \le ||Z_1| - |Z_2|| = O(2^i \log W)$$

Since
$$a = \frac{2 + \varepsilon}{\varepsilon} \log W$$

It can be proven
$$\varepsilon \cdot S = \Omega(2^i \log W)$$

Since
$$a = \frac{2 + \varepsilon}{\varepsilon} \log W$$

It can be proven
$$\varepsilon \cdot S = \Omega(2^i \log W)$$

Combined with $|X - S| = O(2^i \log W)$

We obtain relative error:
$$\frac{|X-5|}{5} \le \varepsilon$$