

New Models and Algorithm for Throughput Maximization in Broadcast Scheduling

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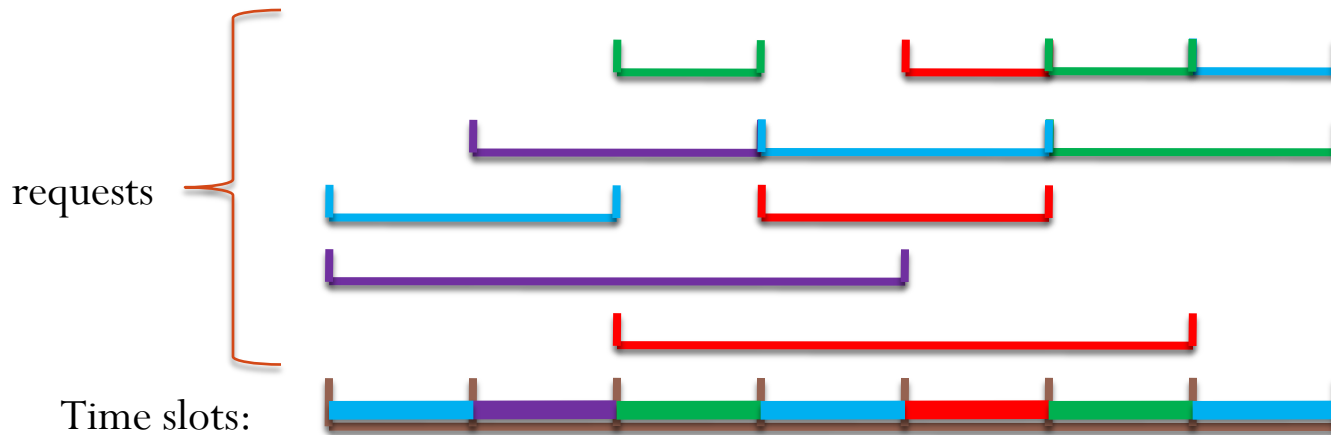
Broadcast Scheduling

Problem definition:

- Given a set of pages $P = \{p_1, p_2, \dots, p_n\}$
- Time is slotted, $T = \{1, 2, \dots, T\}$
- Each client sends a request r for page p , with release time a_r and deadline d_r
- The server broadcasts one page p in a time slot t , and all requests r of page p with $t \in [a_r, d_r]$ can be satisfied

Broadcast Scheduling

- Example:



Several requests may be satisfied by one broadcast.

Broadcast Scheduling

- Traditional objectives.
 - Hard deadlines:
 - Throughput maximization (MAX-THP)
 - ...
 - No deadlines:
 - Minimizing the max response time.
 - Minimizing the flow time (i.e., avg. response time).
 -
 - NP-hardness [Chang et al. 08].

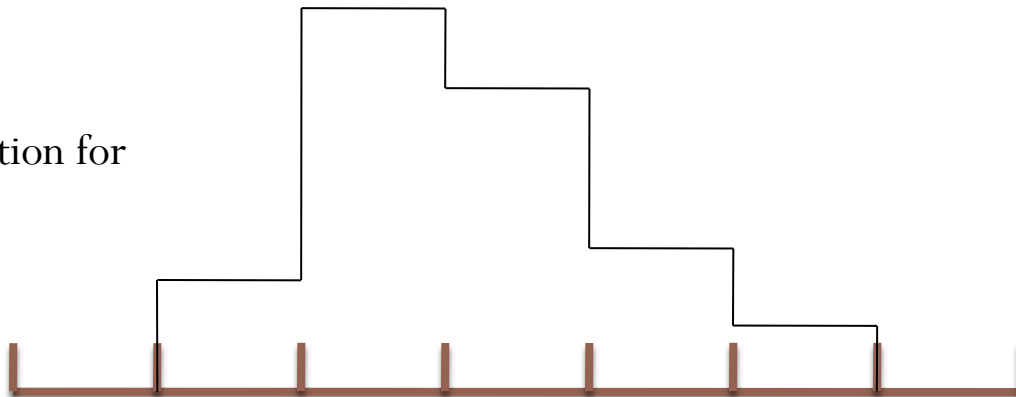
Motivation

- Each client request the reading of some sensor at some time. The server can probe one sensor in a time slot.
 - A client requests the temperature reading at 5:30PM. She may be satisfied with a reading at 5:33PM. A reading at 5:40PM may be still useful, but not as much. But a reading in 6:00PM is useless.
- Traditional objectives are not sufficient in this example.
 - Minimizing response time ignores deadlines.
 - Minimizing throughput ignores the latency of satisfied requests.
- We capture this in two approaches.
 - A general time-dependent profit function.
 - Tradeoff between completeness and latency.

Profit Maximization

- A generalization of throughput maximization: Profit Maximization (MAX-PFT)
 - A time –dependent profit function $g_r(t)$ for each request r .
 - If a request is satisfied multiple times, we take the maximum one.
 - A more nuanced view of “satisfying” a request.

The profit function for
a request:



Our Results

- Offline setting.
 - A $(1-1/e)$ -approximation for MAX-PFT.
 - A $3/4$ -approximation for MAX-PFT when the profit functions are unimodal.
 - MAX-THP offline: A $3/4$ -approximation [Gandhi et al. '06].
- Online settings.
 - An s -speed $(1+1/s)$ -competitive algorithm for MAX-PFT.
 - MAX-THP online: A $1/2$ -competitive algorithm [Kim et al. '04].

Our Results

- Minimizing latency subject to completeness requirement.
 - A $(3/4, 1)$ -approximation for the (completeness, latency) pair.
 - Note that both ratios are *in expectation*.
- Throughput Maximization with Relaxed Time Windows.
 - Suppose there is a fractional solution that satisfies all requests. We can find a schedule in polynomial time such that each request can be satisfied by right (or left) shifting the window by at most its length.



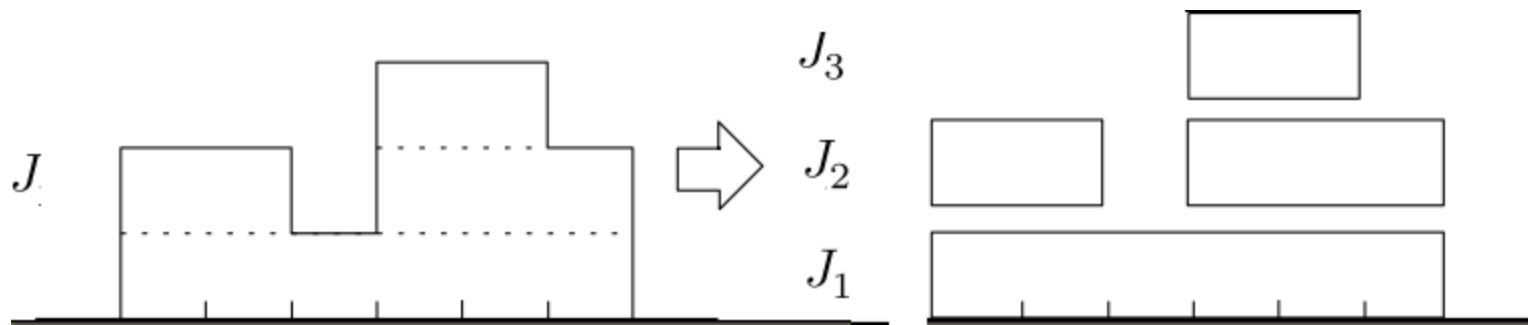
Relaxed window

Our Results

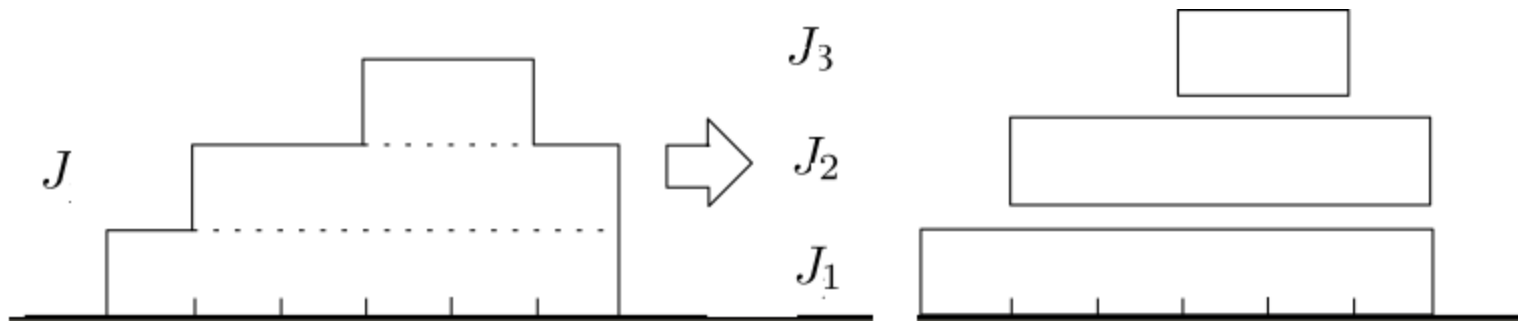
- Offline MAX-THP:
 - 2-speed 1-approximation.
 - Such a result was known only if all request can be scheduled in a fractional solution [Chang et al. 08].
 - This directly implies a 2-approximation for MAX-THP.
- Minimizing the max response time
 - A $(2-\epsilon)$ -lower bound for randomized algorithms in the oblivious adversary model.
 - The same bound was only known for deterministic algorithm [Bartal et al. 00, Chang et al. 08].
 - Note that FIFO is 2-competitive [Bartal et al. 00, Chang et al. 08, Chekuri et al. 09].

Offline – Profit Maximization

- The slicing trick: Convert MAX-PFT to weighted MAX-THP



A general profit function



A **unimodal** profit function

Offline – Profit Maximization

THM: A $3/4$ –approximation for MAX-PFT when the profit functions are unimodal.

Proof: The slicing trick and the $3/4$ -approximation for weighted MAX-THP.

THM: A $(1-1/e)$ –approximation for MAX-PFT with general profit functions.

Proof 1: A simple independent rounding schema.

Offline – Profit Maximization

Proof 2:

(submodular maximization subject to a matroid constraint)

- $f: 2^N \rightarrow \mathbb{R}$ is a **submodular function** if

$$f(A + x) - f(A) \leq f(B + x) - f(B) \quad \forall B \subseteq A, x \in N$$

- Let N be $\{(p, t)\}_{p, t}$. The set of feasible solutions is a **partition matroid**.
- Let $Profit(S)$ be the profit obtained by schedule S ($S \subseteq N$). $Profit(\cdot)$ is submodular.
- Submodular function maximization subject to a matroid constraint: $(1-1/e)$ -approximation [Calinescu et al. '07, Vondrak '08, Chekuri et al. '10].

Online – Profit Maximization

- **Maximum Additional Profit First (MAPF):**
 - At any time t , broadcast s pages which give the maximum additional profits.

THM: MAPF is an s -speed $(1+1/s)$ -competitive online algorithm for MAX-PFT.

The analysis is **tight**.

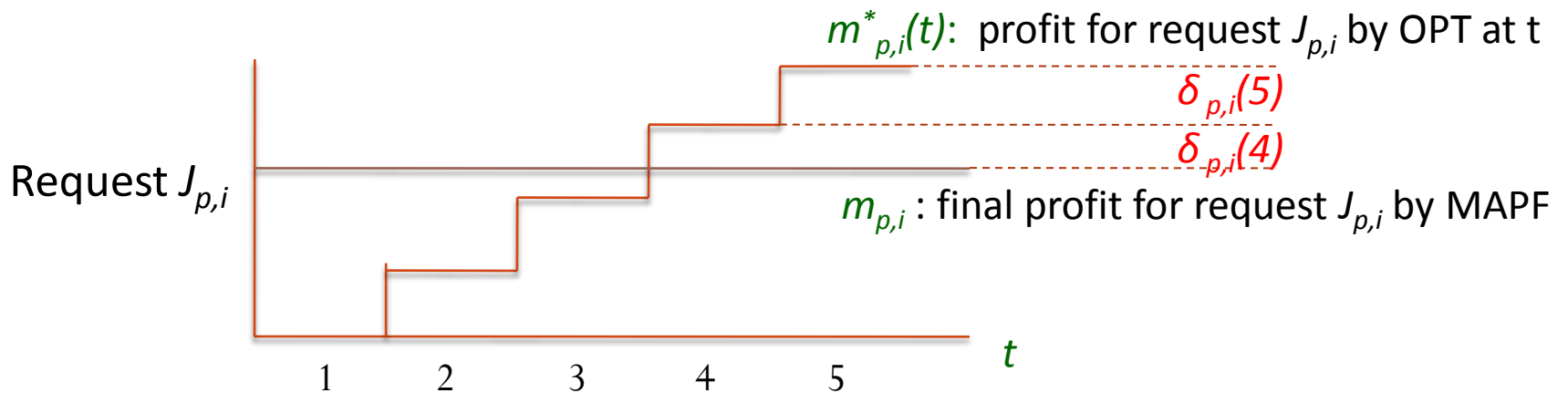
THM: For any $\epsilon > 0$ and $s \geq 1$, MAPF is not s -speed $(1+1/s-\epsilon)$ -competitive.

Online – Profit Maximization

Proof (sketch):

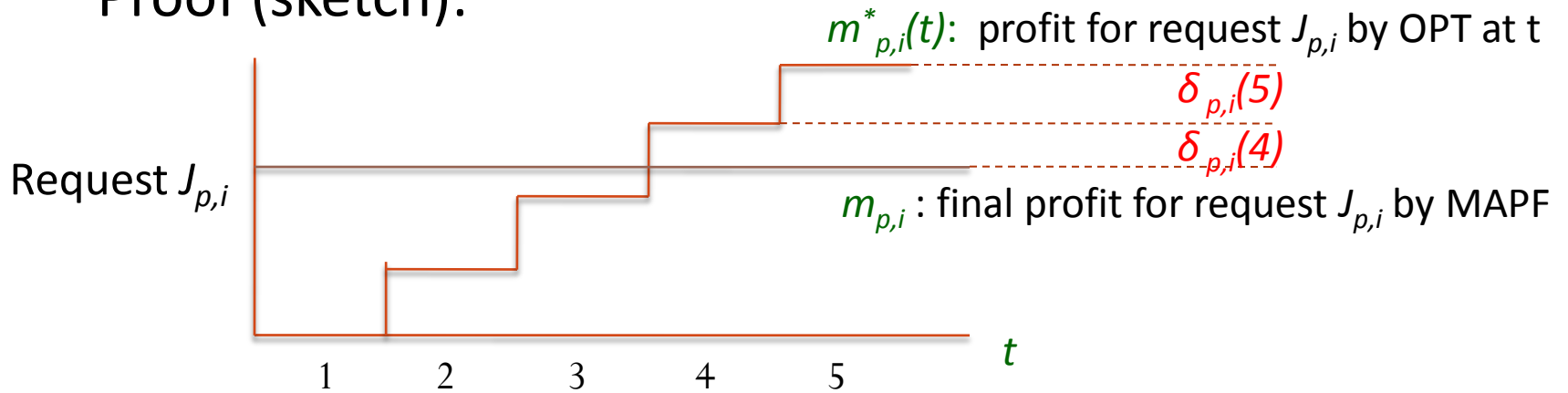
$\Delta(t)$: the increase of the so-far-gained profit by OPT over the final profit by MAPF.

$$\Delta(t) = \sum_{p,i} \delta_{p,i}(t).$$



Online – Profit Maximization

Proof (sketch):



$$\Delta(t) = \sum_{p,i} \delta_{p,i}(t).$$

$$OPT \leq MAFP + \sum_t \Delta(t) \quad \text{-- by definition of } \Delta.$$

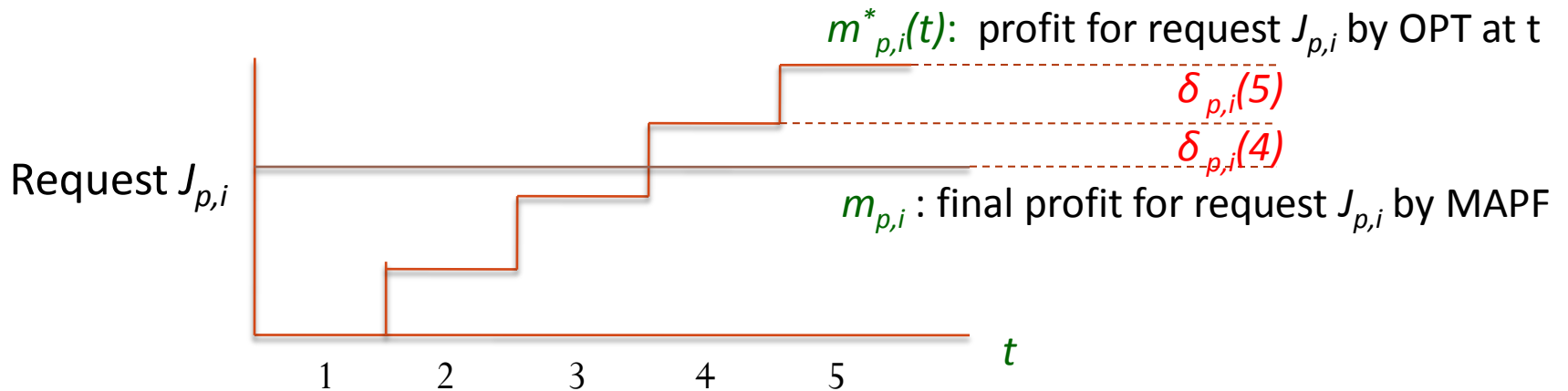
$$\leq MAFP + \sum_t (1/s) \sum_{p,i} (m_{p,i}(t) - m_{p,i}(t-1)) \quad \text{-- next slides.}$$

$$\leq (1 + 1/s) MAFP$$

Additional profit
obtained by MAFP

Online – Profit Maximization

Proof (sketch):



It suffices to show $\Delta(t) \leq (1/s) \sum_{p,i} (m_{p,i}(t) - m_{p,i}(t-1))$

- Assume OPT broadcast q at time t and $\Delta(t) > 0$.
- We can show MAPF does not broadcast q . O.w. $\Delta(t) = 0$.
- $\Delta(t) \leq \sum_i (m_{q,i}(t) - m_{q,i}(t-1))$.
- $(m_{q,i}(t) - m_{q,i}(t-1)) \leq (m_{p,i}(t) - m_{p,i}(t-1))$ if MAPF broadcast p .

Open Problems

- Is it possible to get a $4/3$ -speed 1-approximation for MAX-THP. Note this would imply a $3/4$ -approximation for MAX-THP (matching the bound by Gandhi et al. '06).
- Derandomizing the $(3/4, 1)$ -approximation for the (completeness, latency) pair.
- A better understanding of the completeness-latency tradeoff.

Thanks

Offline – Profit Maximization

Proof 1: (LP rounding)

$Y_p^{(t)} = 1$: The server broadcasts p at time t .

$X_{p,i} = 1$: The i th request of p is satisfied.

$$\text{maximize } \sum_{p,i} w_{p,i} X_{p,i}$$

$$\text{subject to } \sum_{t \in \mathcal{T}_{p,i}} Y_p^{(t)} \geq X_{p,i} \quad \forall p, i,$$

$$\sum_p Y_p^{(t)} \leq 1, \quad \forall t,$$

$$X_{p,i} \in \{0, 1\}, \forall p, i, \quad Y_p^{(t)} \in \{0, 1\}, \forall p, t$$

Offline – Profit Maximization

Proof 1: (LP rounding)

Let $x_{p,i}, y_p^{(t)}$ be the optimal fractional solution.

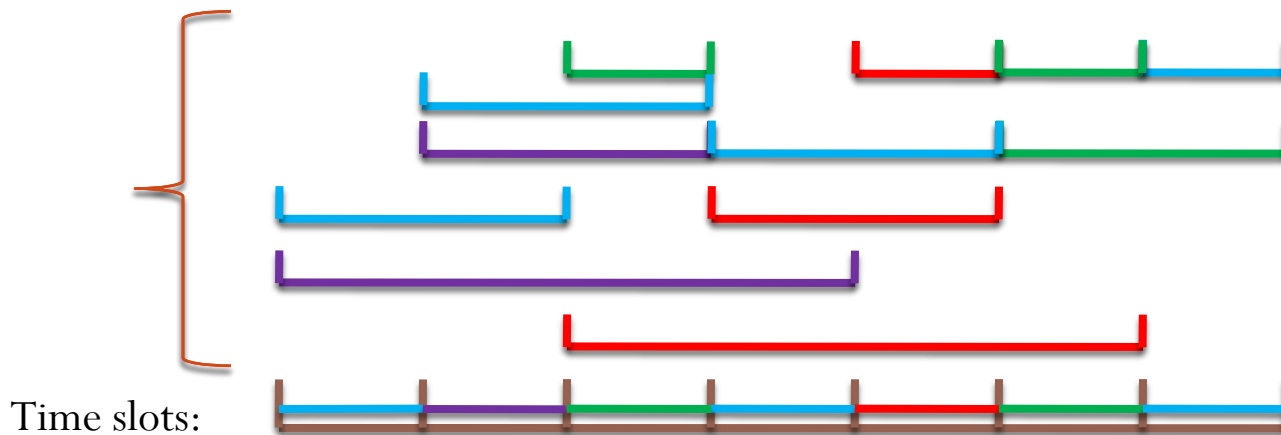
Algorithm: (Independent rounding)

- At time t , choose p to broadcast with prob. $y_p^{(t)}$

It is not hard to show that $Pr(\text{request } J_{p,i} \text{ is satisfied}) \geq (1-1/e) x_{p,i}$

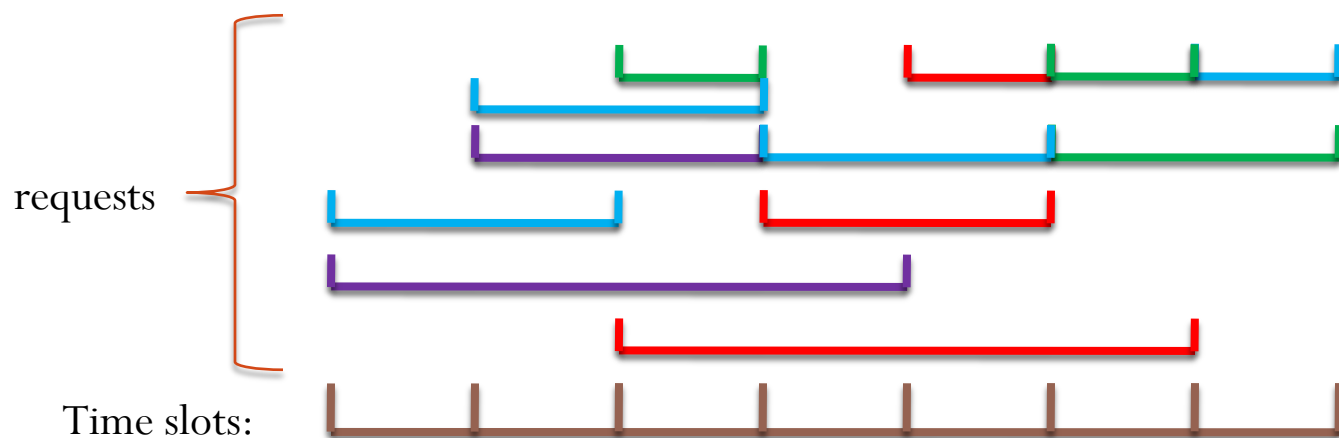
Throughput Maximization with Relaxed Time Windows

- The fractional solution corresponds to a flow.
- There is integral flow.



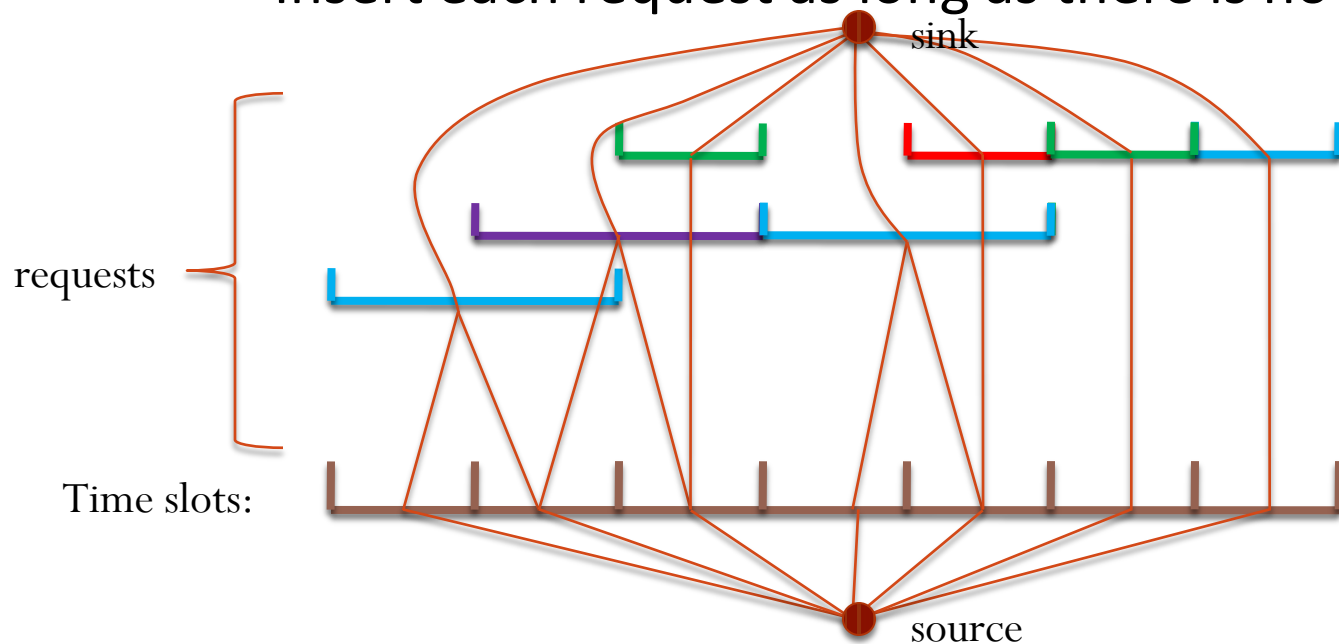
Throughput Maximization with Relaxed Time Windows

- Suppose there is a fractional solution that satisfies all requests.



Throughput Maximization with Relaxed Time Windows

- For each page
 - Order the requests for page p in non-decreasing window length.
 - Insert each request as long as there is no overlap



- The fractional solution corresponds to a flow.

Throughput Maximization with Relaxed Time Windows

- The fractional solution corresponds to a flow.
- There is integral flow.

