

Time Complexity of Link Reversal Routing

Talk is based on ...

work with Bernadette Charron-Bost (CNRS, LIX),
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Josef Widder (TU Wien)

(Sirocco, '11), (ACM Trans. on Algorithms, '15).

Algorithms on Graphs

Graph rewriting algorithms:

G_0, G_1, G_2, \dots

Link reversal algorithms: on link directions

Applications:

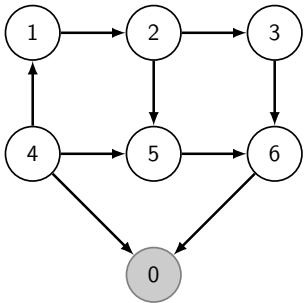
- ▶ Routing algorithms
- ▶ Behavior of asynchronous circuits
- ▶ Behavior of Physarum (?)

The Routing Problem

Requirements:

- ▶ Computationally efficient algorithm.
- ▶ Adapt to failures/mobility fast.

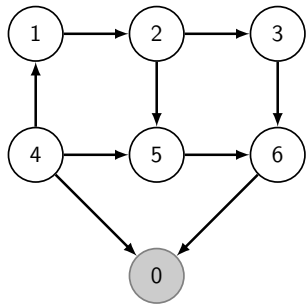
Routing to a destination



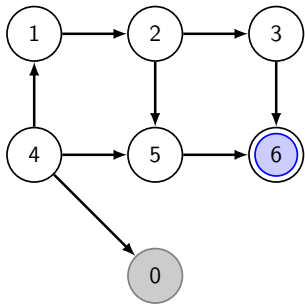
acyclic, destination oriented

Acyclic and
destination oriented \Rightarrow
Routing is simple.

Routing to a destination

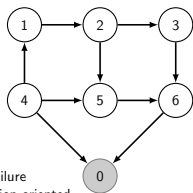


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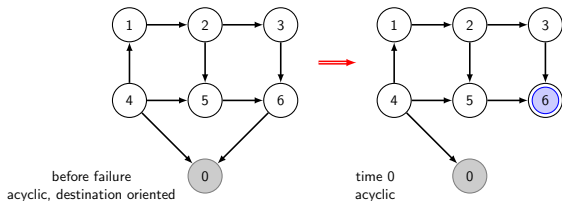
acyclic

Full Reversal routing (Gafni and Bertsekas, 1981)

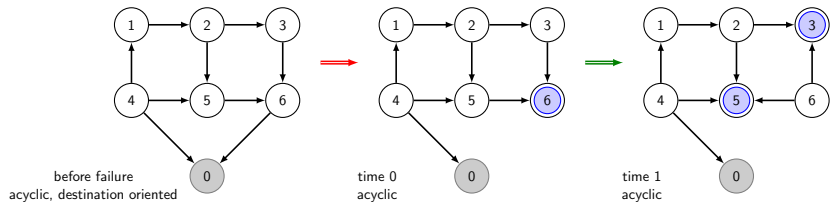


before failure
acyclic, destination oriented

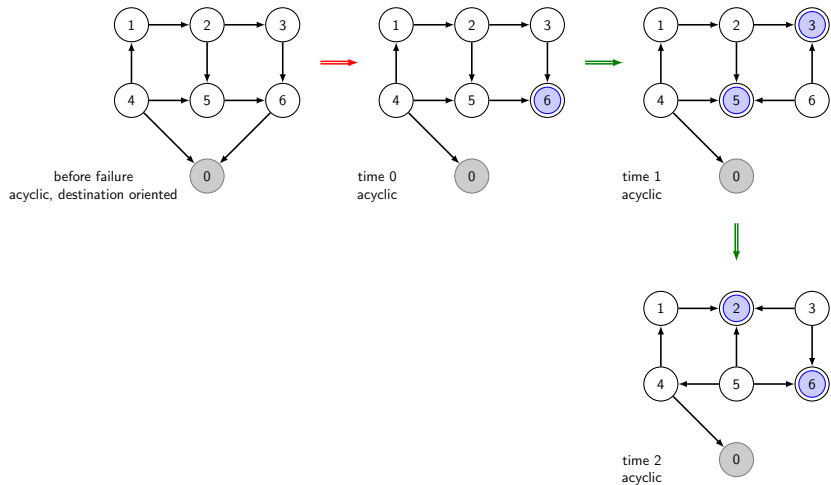
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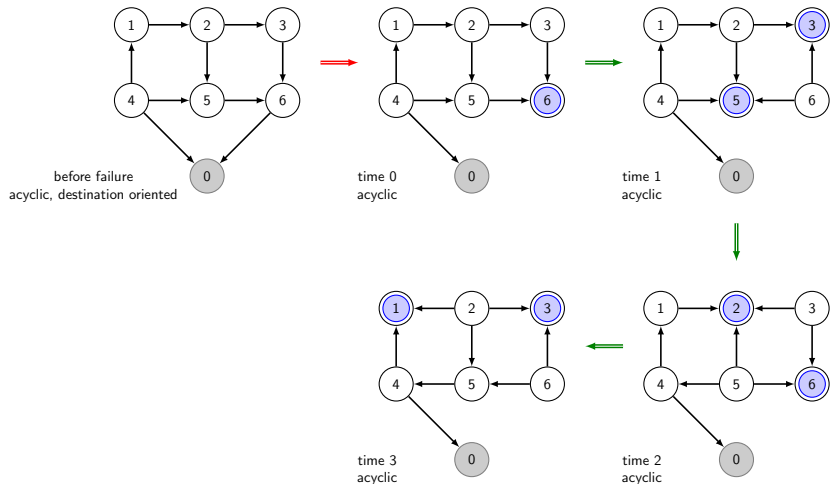
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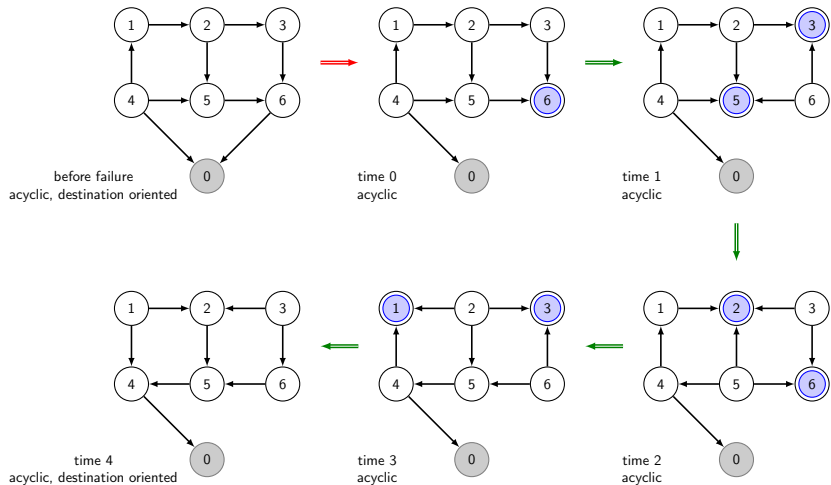
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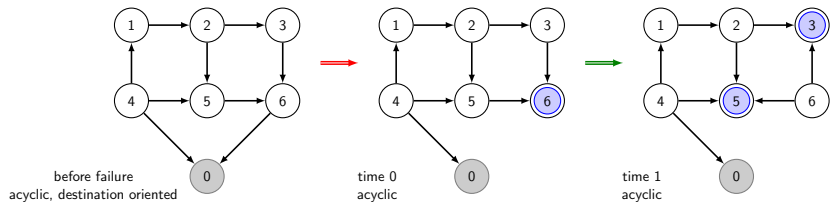
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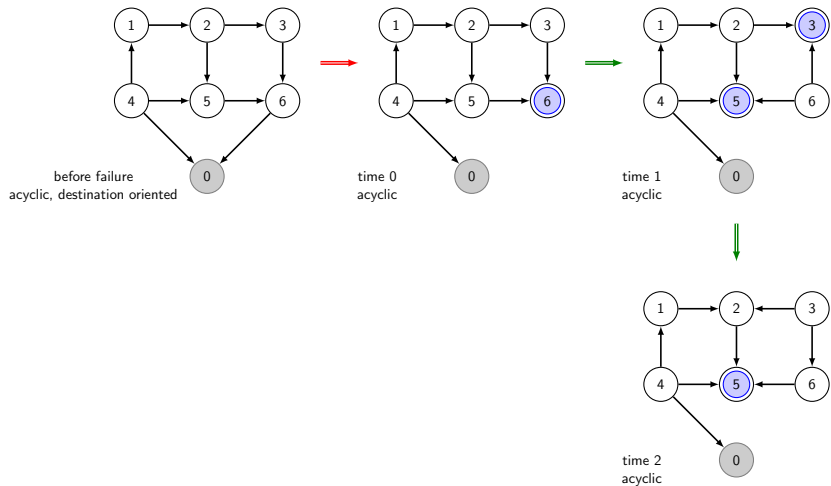
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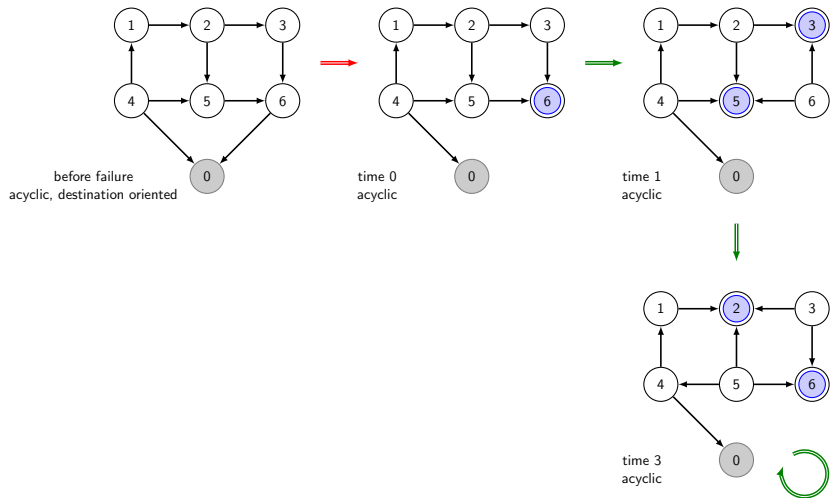
Full Reversal routing - alternative execution



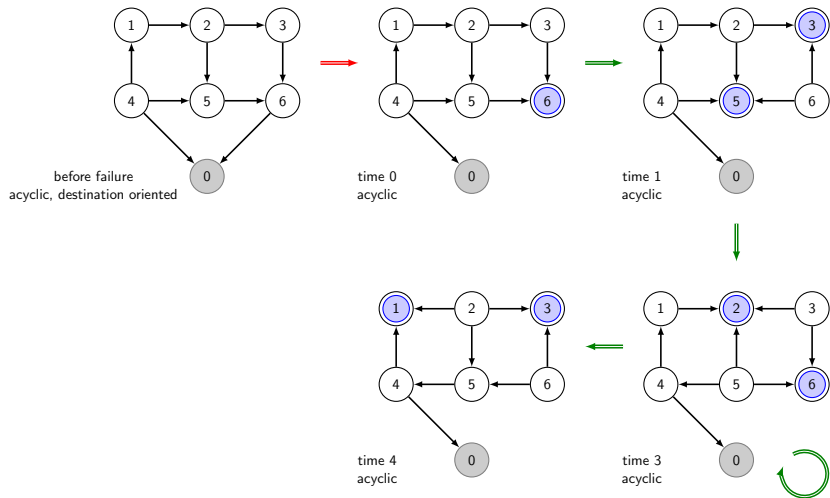
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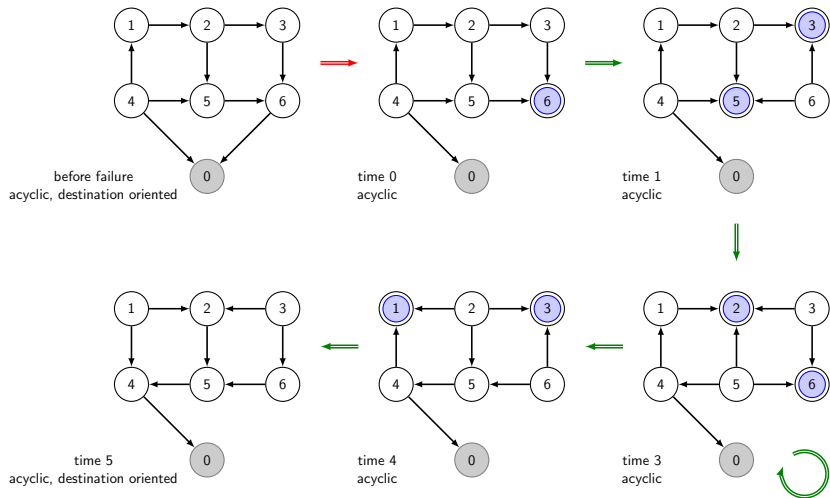
Full Reversal routing - alternative execution



Full Reversal routing - alternative execution



Full Reversal routing - alternative execution



Executions

Execution: G_0, G_1, G_2, \dots

Two extremes:

- ▶ **Greedy** execution: all sinks in G_i make step
- ▶ **Lazy** execution(s): one sink in G_i makes a step

... and many executions in between.

Distributed Algorithm

Local graph rewriting: $\dots, G_i, G_{i+1}, \dots$

Local mutex: no two sinks are neighbors

Asynchronous execution possible

Some Facts

Algorithm always **terminates destination oriented**.

At all steps, graph is **acyclic**.

Algorithm steps are **commutative**:

- ▶ Final graph is the same.
- ▶ Each node performs same number of steps.

Complexity

For an initial graph:

work complexity of node i

- ▶ number of steps made by node i in **any** execution
- ▶ first exact expression established in (Busch et al., 2003)
- ▶ work complexity for more general algorithms in (Charron-Bost et al., 2009)

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Due to concurrency, understanding of work complexity is not sufficient to get exact time complexity.

A dynamical system

System state $\vec{W}(t) = (\vec{W}_0(t), \vec{W}_1(t), \dots, \vec{W}_N(t))$ at time t .

$\vec{W}_i(t)$... number of steps of node i up to time t .

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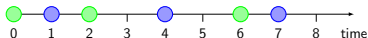
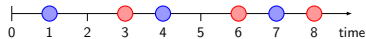
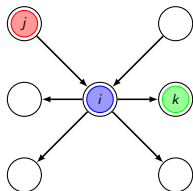
$\vec{W}_i(0) = 0$ for all nodes i

$\vec{W}_0(t) = 0$ for all times t

Looking for a function F such that

$$\vec{W}(t) = F(\vec{W}(t-1))$$

The influence of links



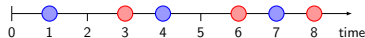
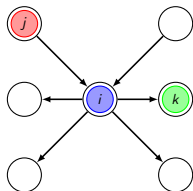
Proposition

Between two consecutive steps by a node i , each neighbor of i takes exactly one step.

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In any FR execution in which i takes a step, before the first step by i , each node $j \in In_i$ takes no step and each node $k \in Out_i$ takes exactly one step.

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⇒ Links induce strict alternation.

Recurrence

Strict alternation \Rightarrow

$$\vec{W}_i(t) \leq \vec{W}_j(t-1) + 1 \text{ for } j \in In_i$$

$$\vec{W}_i(t) \leq \vec{W}_k(t-1) + 0 \text{ for } k \in Out_i$$

Theorem

In a greedy FR execution, for any node i other than 0 and any $t \geq 1$,

$$\vec{W}_i(t) \leq \min \left\{ \vec{W}_j(t-1) + 1, \vec{W}_k(t-1) + 0 : j \in In_i, k \in Out_i \right\}.$$

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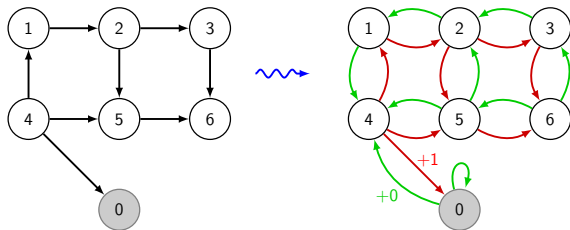
\Rightarrow discrete *linear* dynamical system in min-plus algebra

Reduction to graph properties

- ▶ computing: $\vec{W}(t) = A^t \otimes \vec{0} = (A^{t/2})^2 \otimes \vec{0}$

Reduction to graph properties

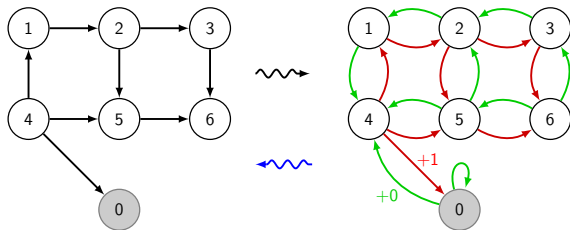
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- ▶ but:
matrix A quite similar to initial graph's adjacency matrix



- ▶ $W_i(t) = \min \{ \text{weight}(p) : p \text{ is path to } i \text{ of length } t \}$

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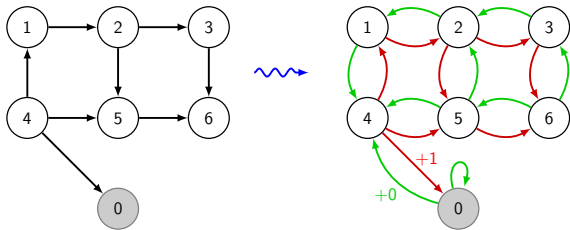
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- ▶ $W_i(t) = \min \{ \text{weight}(p) : p \text{ is path to } i \text{ of length } t \}$
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Simple work complexity proof

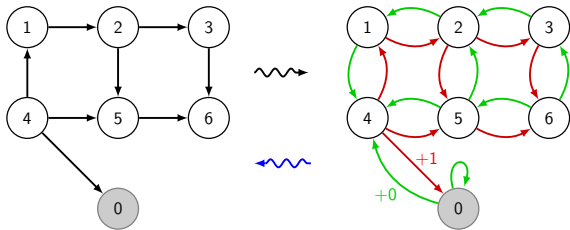
$$w_i = \lim_{t \rightarrow \infty} W_i(t) = \lim_{t \rightarrow \infty} \min \{ \text{weight}(p) : p \text{ is path to } i \text{ of length } t \}$$



- ▶ A path long enough to reach 0, will loop there for free.
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Can we say something about time complexity?

Dual of $W_i(t)$ is $T_i(w)$

- ▶ $W_i(T_i(w)) = w$ $\Rightarrow W_i(\theta_i) = w_i$
- ▶ $W_i(T_i(w) - 1) = w - 1$ $\Rightarrow W_i(\theta_i - 1) = w_i - 1$

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Theorem

The *termination time* θ_i of any node i that takes a step is equal to

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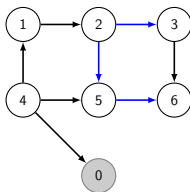
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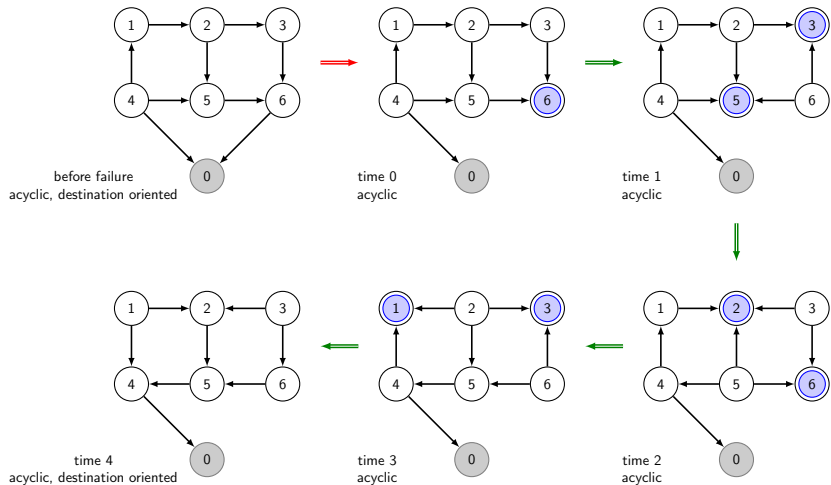
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$$w_3 = 2$$

$$\theta_3 = 4$$

Full Reversal routing (Gafni and Bertsekas, 1981)



Immediate results

Corollary

There is an N node graph family with FR time complexities scaled from $\Theta(N)$ to $\Theta(N^2)$

Corollary

In any tree with $N + 1$ nodes, the FR time complexity is at most equal to $2N - 1$.

Corollary

In general FR time complexity is unstable. Adding one link can increase it from $\Theta(N)$ to $\Theta(N^2)$.

FR only for routing?

Distributed algorithm where **no two sinks are neighbors**.

⇒ Distributed scheduling with local mutex.

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Work until termination: $w_i = \min \{r(c) : c \text{ is chain from } 0 \text{ to } i\}$.

⇒ What if we remove the special node 0?

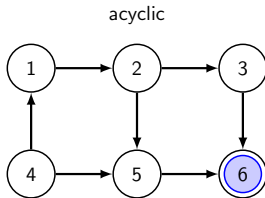
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Will run forever.

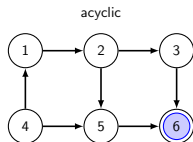
Distributed scheduling

Scheduling frequency $\lim_{t \rightarrow \infty} W_i(t)/t$

Distributed scheduling

Again a dynamical system $\vec{W}(t)$

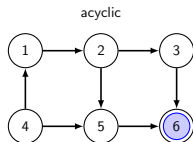
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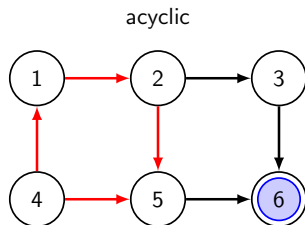


\Rightarrow eventually c will find a cycle γ with minimal $r(\gamma)/\ell(\gamma)$.

Distributed scheduling

Theorem

$$\lim_{t \rightarrow \infty} W_i(t)/t = \min \{r(\gamma)/\ell(\gamma) : \gamma \text{ is cycle}\}$$



$$\lim_{t \rightarrow \infty} W_i(t)/t = 1/4$$

Beyond Full Reversal

LR $\hat{=}$ Reverse only **some** links.

Proof idea:

- ▶ Not necessarily linear in N -dimensional system
- ▶ But: simulate nodes with one or two nodes (transformed graph).
- ▶ Relate chains in transformed graph to chain in original graph.

\Rightarrow analogous results with other potentials Π .

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