

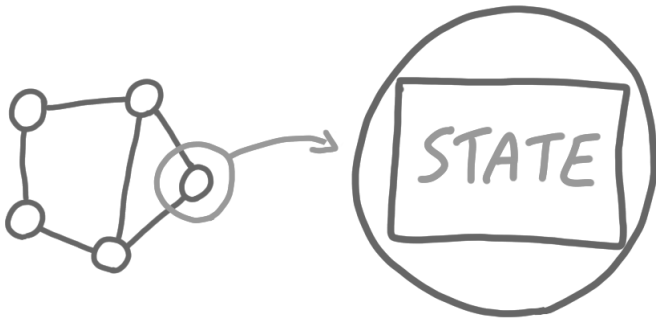
Silent MST approximation for tiny memory

Lélia Blin, Swan Dubois and Laurent Feuilloley

SSS 2020 · 18th Novembrer 2020 · Virtual

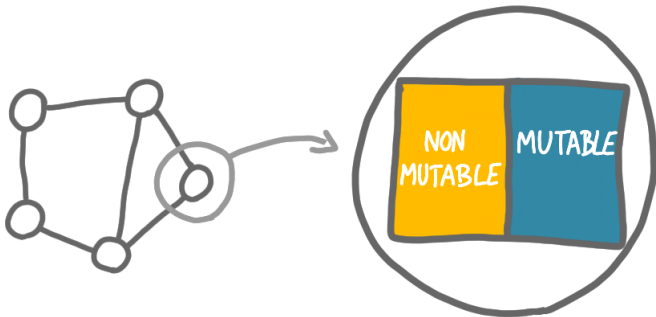
The state model and MST

1. Computation on a graph and every node has a state, with two parts : the mutable and the non-mutable memory.



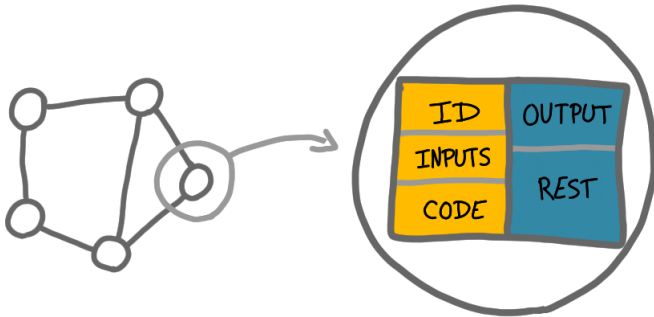
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2. The non-mutable memory contains the ID, inputs and code.
3. At the end the mutable memory should contain the output.

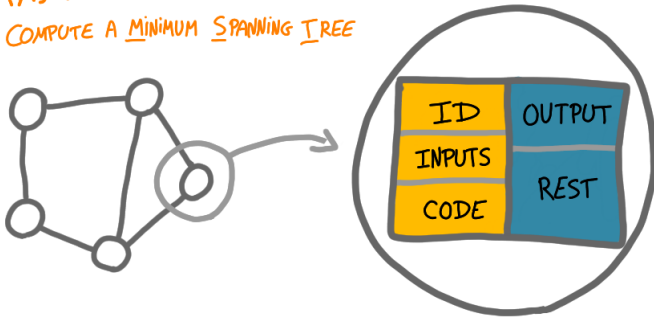


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TASK:

COMPUTE A MINIMUM SPANNING TREE

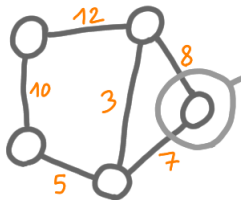


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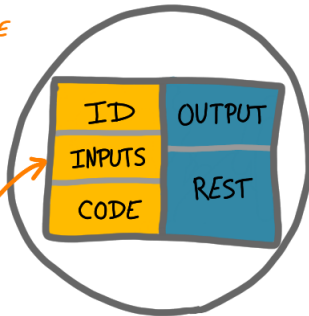
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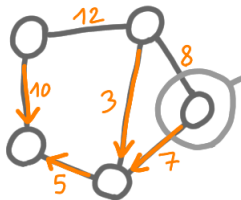


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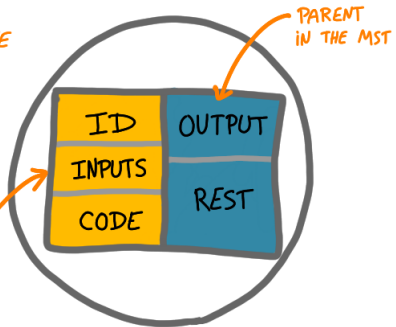
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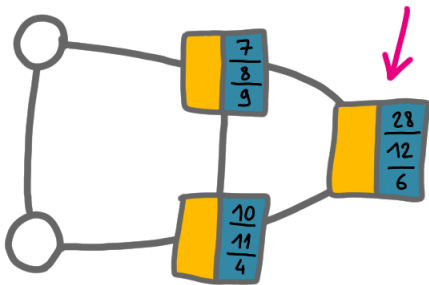
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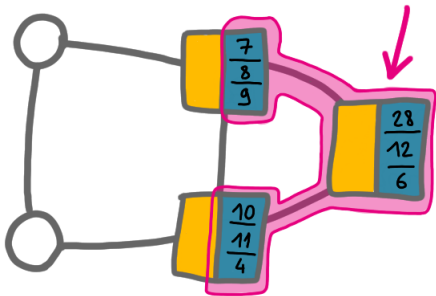


Computation in the state model



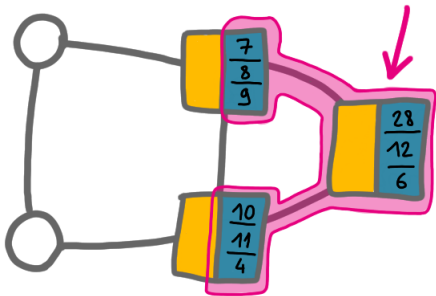
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- ▶ Local view, modeling messages.



Computation in the state model

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- ▶ Local "if" rules, activated/non-activated.



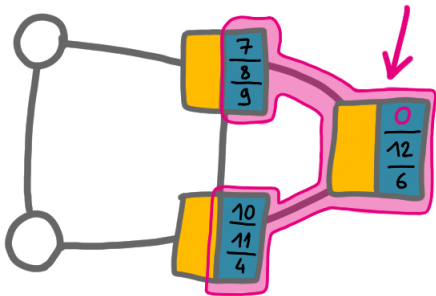
CODE:

RULE1: IF var1=28
& ID is ...
& neighbors var1
& weight ...
then var1:= 0

RULE2:

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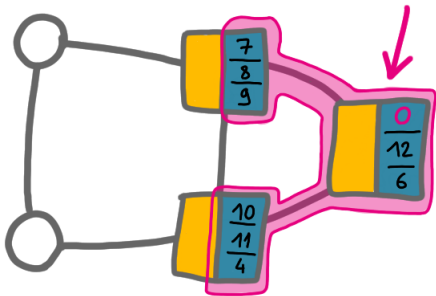
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Computation in the state model

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- ▶ Local "if" rules, activated/non-activated.
- ▶ Adversary scheduler : chooses the (active) nodes taking steps.



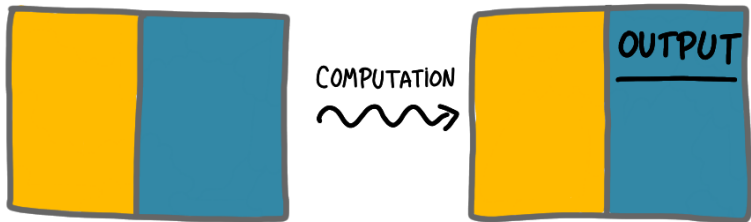
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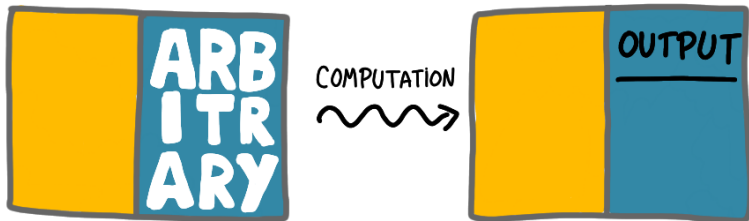
Silent self-stabilization

- ▶ Fault-free setting vs. self-stabilizing setting



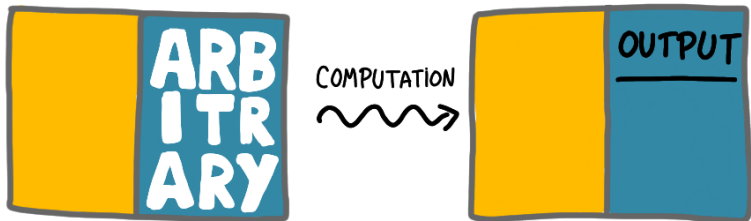
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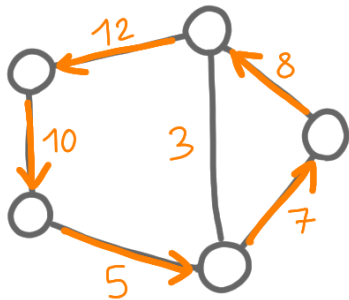
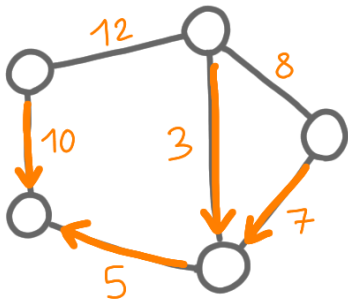
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- ▶ Silent self-stabilization : at the end, no activated nodes



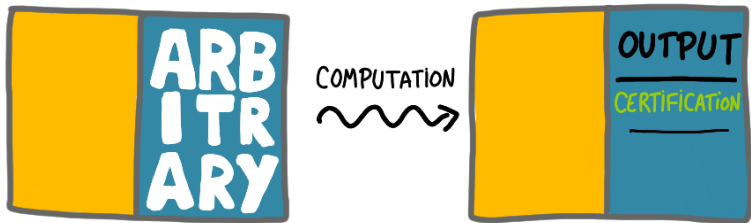
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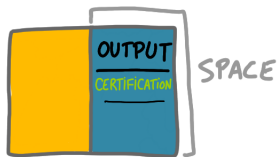
Silent self-stabilization

- ▶ Fault-free setting vs. self-stabilizing setting
- ▶ Silent self-stabilization : at the end, no activated nodes
- ▶ The output is not enough (for most tasks).
- ▶ Certification has been studied on its own (*Proof-labeling schemes* Korman, Kutten and Peleg)



Certification size and MST

- ▶ Focus on the space (keeping polynomial time).
- ▶ → minimize size of certification and do not use more space.



Theorem [Korman, Kutten] : The optimal certification size for minimum spanning tree is $\Theta(\log n \times s)$, when weights are encoded on s bits.

→ For weights in a $poly(n)$ range $s = \log n$, and the optimal size is $\Theta(\log^2 n)$.

Our results

Theorem 1 : A (full) self-stabilizing MST algorithm with optimal space $O(\log n \times s)$.

Proof : First : build a non-certified MST. Second : build the optimal certification on top.

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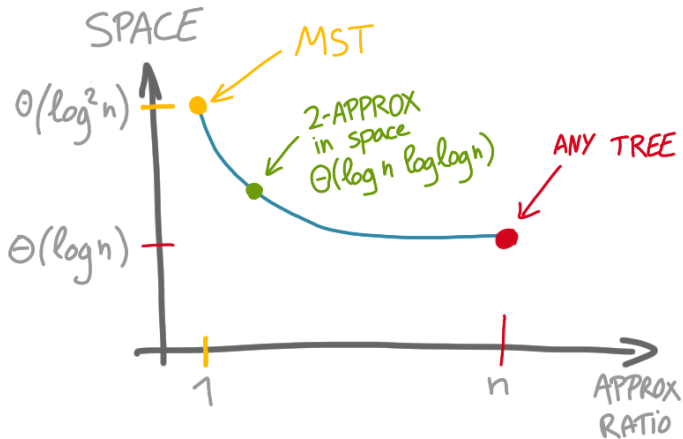
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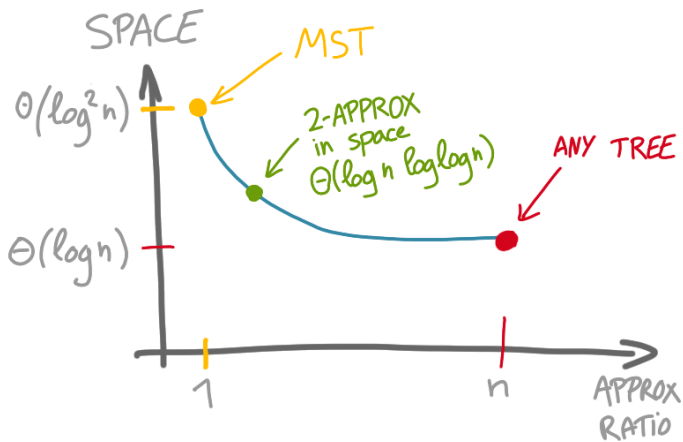
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Theorem 2 : We can parameterize the algorithm such that there is **tradeoff between space complexity and quality of the solution**.

Approximation and trade-off



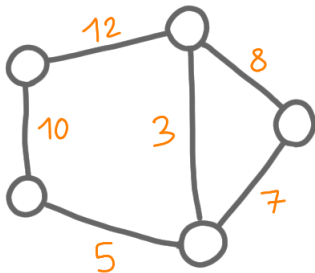
Approximation and trade-off



Approximation in certification : Censor-Hillel, Paz and Perry (2017), and Emek and Gil (2020).

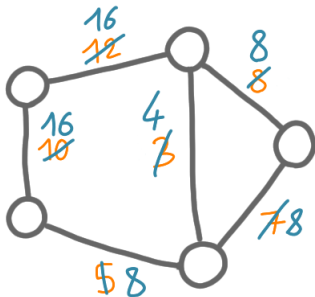
Example : 2-approximation

- ▶ Round every weight to the next power of 2.



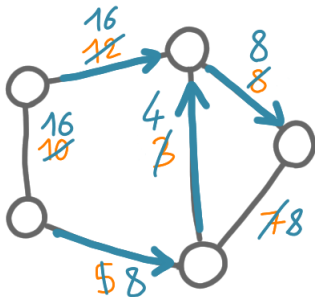
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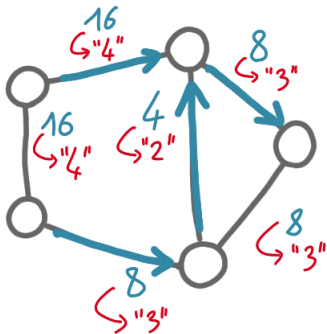
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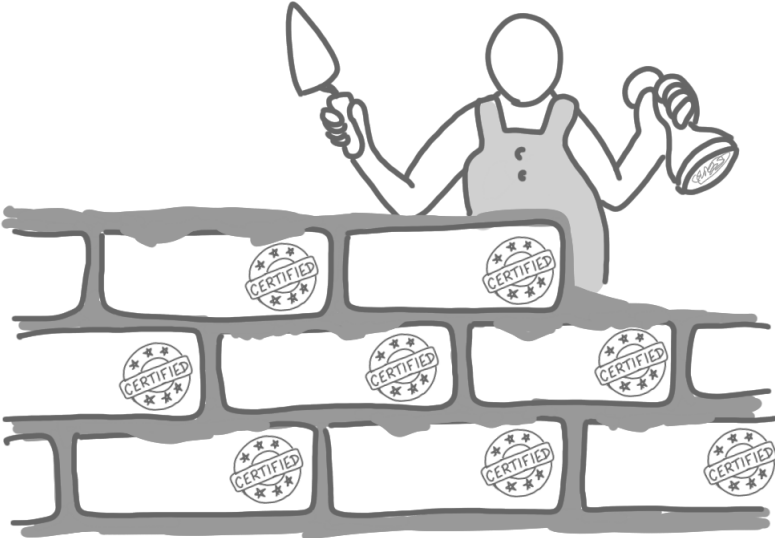
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- ▶ Round every weight to the next power of 2.
- ▶ An MST on these new weights is a 2 -approximation.
- ▶ Weights can be encoded in a compact way : 2^p encoded as p .
- ▶ This reduces the weight exponentially $\log n \rightarrow \log \log n$.
- ▶ Using Theorem 1, we get a 2-approx in space $\log n \log \log n$.



To finish : build vs certify

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Thanks for your attention !

