



ADVANCES IN ALIGNMENTS

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ALIGNMENTS: CHALLENGES

- Various model classes require/allow for model specific techniques
 - Petrinets, Reset/Inhibitor nets, Process trees, Colored Petri nets,
 - Petri nets with IDs?
 - Mixed paradigm models?
- Theoretical Bounds are unclear
- No proper heuristic solutions exist
 - Online approaches do not have a global view,
 - Decomposition approaches do not guarantee to return alignments.

THEORETICAL PROPERTIES

- Finding an alignment for a given model M and trace t for cost function c :
 - Identify the (a) cheapest firing sequence from m_i to m_f in a synchronous product model s , with a cost function c'
- For Reset/Inhibitor nets this problem is *undecidable*, (if $c' = 0$ for all transitions then this is the same as asking: is m_f reachable from m_i)
- For Petri nets this problem is EXPSPACE hard,
- For 1-safe nets, this problem is PSPACE hard,
- For Free-choice Petri nets this problem is NP hard,
- For marked graphs, this problem is polynomial...

PRACTICAL SOLUTIONS

- Assume m_f is reachable from m_i in the synchronous product model s , (lazy soundness)
- Use a shortest-path algorithm building the search space from m_i until m_f is reached, never visiting the same state twice
- Since we know m_f is reachable from m_i , we know that the search space is bounded (decidability is guaranteed), but worst case enormous
- Use an estimator to *underestimate* the remaining cost from m_c to m_f

ESTIMATING: THE MARKING EQUATION

- For any given Petri net \mathcal{N} , with initial marking \mathbf{m}_i , firing sequence s leading to final marking \mathbf{m}_f , holds:
$$\mathbf{m}_i + \mathbf{A} \cdot \mathbf{x} = \mathbf{m}_f \quad (\text{marking equation})$$

with \mathbf{x} the parikh vector of s
- Minimizing $c(\mathbf{x})$ leads to an underestimate for $c(s)$, and hence an underestimate for the cheapest path from \mathbf{m}_i to \mathbf{m}_f . (again an NP-hard problem)
- \mathbf{x} does, in no way, correspond to a realizable firing sequence, but
- For *free-choice nets*: if \mathbf{x} is an integer solution to the marking equation for \mathbf{m}_i and \mathbf{m}_f , then there exists a firing sequence s from \mathbf{m}_i to \mathbf{m}_f .

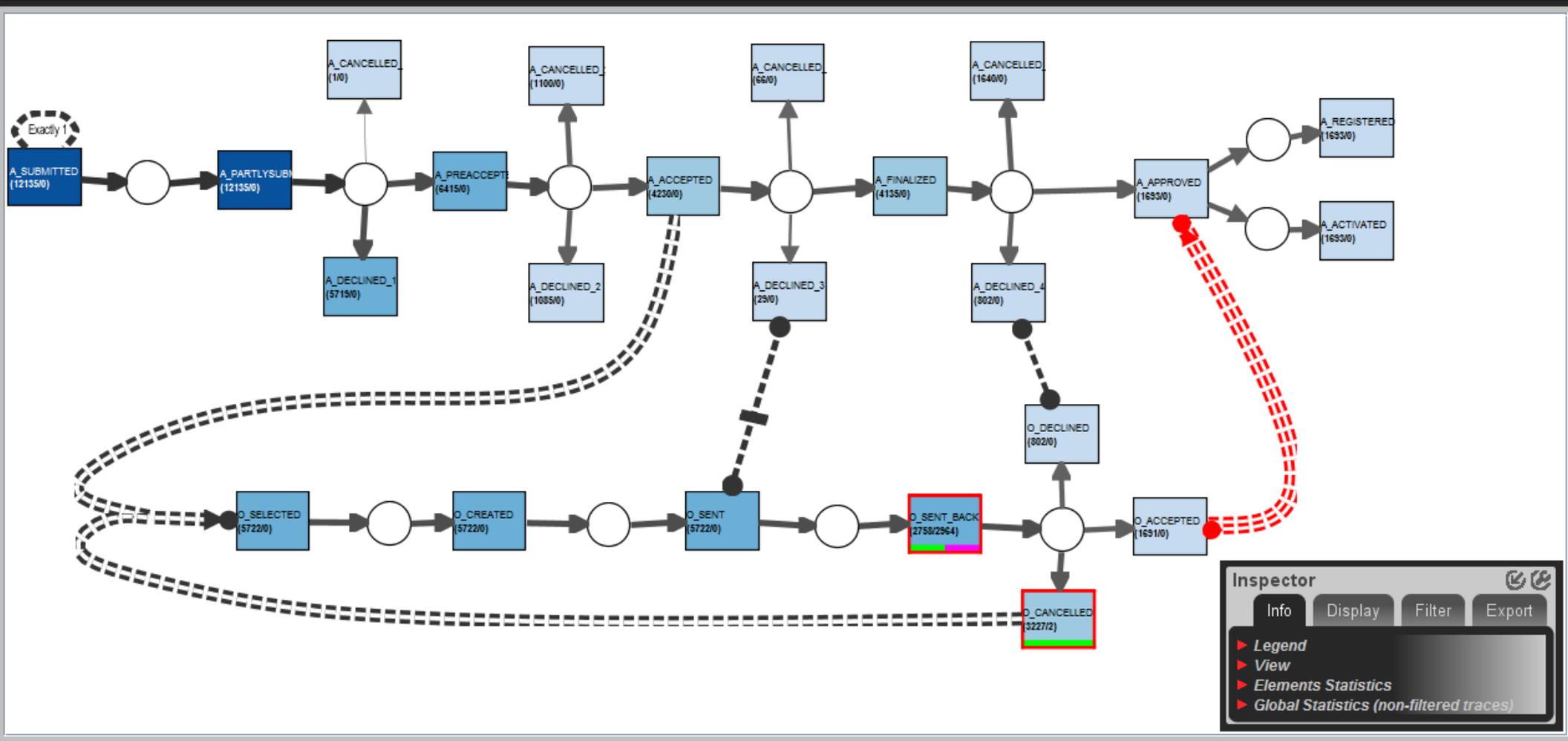
PROPERTIES OF MODELS

- Process trees are 1-safe, free-choice Petri nets with nice structural properties, so finding alignments on them is easier, but still NP hard.
- Restricting the behavior in a Petri net (with inhibitor arcs, data guards, declarative constraints, etc.) makes the marking equation a worse and worse estimator, but it remains an *underestimating* estimator
- For mixed-paradigm models, an implementation now also exists allowing to assign costs to violating declarative constraints (work with Claudio Di Ciccio)

MIXED PARADIGM MODELS

ProM 6

Replay result - log BPI Challenge 2012 (AO complete ca



HEURISTICS FOR ALIGNMENTS

- Decomposition approaches typically give an underestimate of the costs, but no alignment (actual path) is returned,
- Online approaches provide alignments overestimating the costs, but do not guarantee termination in the model (and finishing the model is again an alignment problem),
- We propose a new technique that:
 - Keep a global view on both trace and model,
 - Provides proper alignments,
 - In predictable time (and memory),
 - Overestimating the optimal alignment costs.

AGAIN, THE MARKING EQUATION

- For any given Petri net \mathcal{N} , with initial marking \mathbf{m}_i , firing sequence s leading to final marking \mathbf{m}_f , holds:

$$\mathbf{m}_i + \mathbf{A} \cdot \mathbf{s} = \mathbf{m}_f$$

- The firing rule of Petri nets can also be translated in this form, namely, if t is enabled in marking \mathbf{m} :

$$\mathbf{m} + \mathbf{A}^- \cdot \mathbf{1}_t \geq \mathbf{0}$$

Where \mathbf{A}^- is the consumption matrix of the Petri net and a $\mathbf{1}_t$ vector with value 1 for transition t and 0 otherwise.

- This can be combined into a larger ILP problem for k of x prefix alignments.

k OF x PREFIX ALIGNMENTS

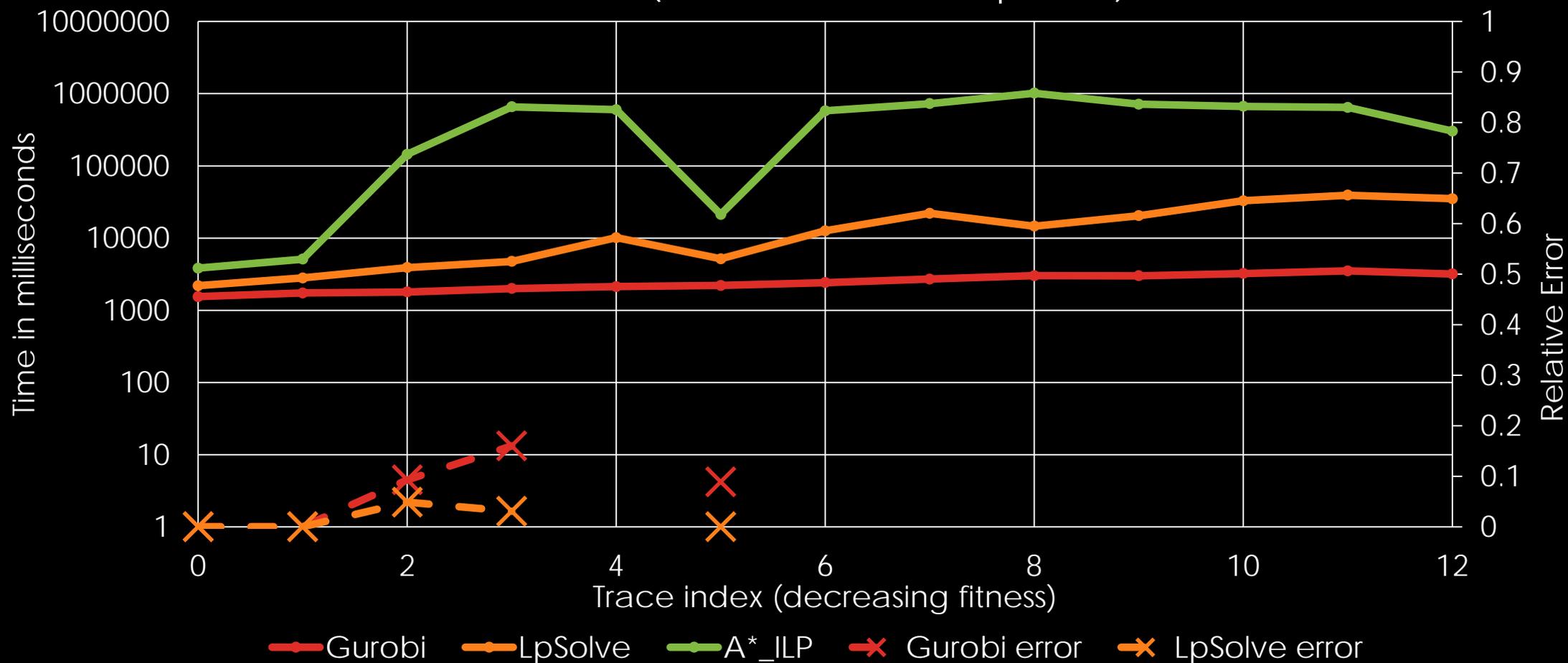
- Consider a firing sequence: $\langle t_1, t_2, t_3, t_4, t_5, \dots \rangle$
- We can construct an ILP, that guarantees that:
 - The first x transitions can be executed in that order (*using the firing rule equations*)
 - The remainder of the trace is a solution to the marking equation (*and hence for free choice nets, the final marking is reachable from the marking reached after firing the first x transitions*)
 - k out of the first x transitions correspond to an event (*log-move or sync-move*)
 - The cost function is minimal.
- Then iteratively, we keep the x transitions and iteratively solve the problem for the remaining elements of the trace.

EXPERIMENTAL RESULTS

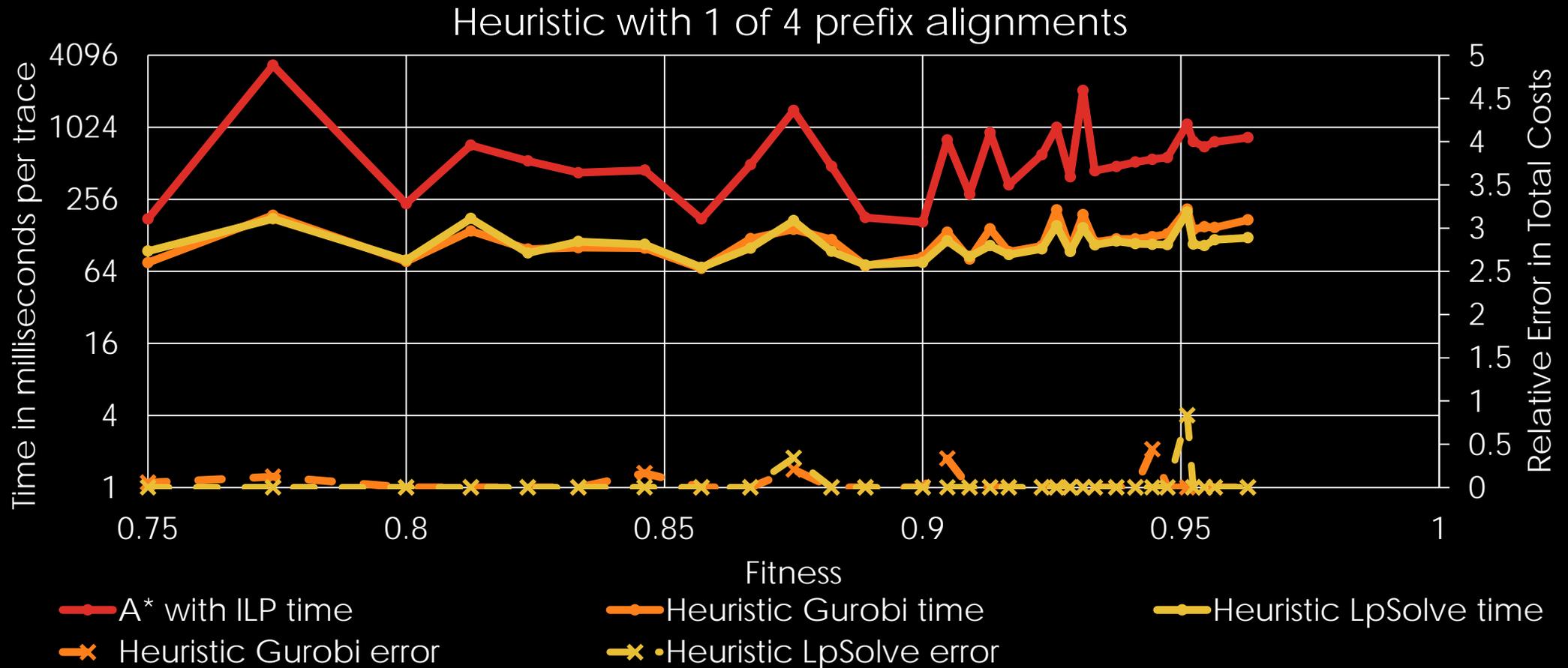
- We used A^* as a baseline, and two solvers, LpSolve and Gurobi,
 - We implemented backtracking in case deviation from expected costs becomes too high, and
 - We used various values for x (and k).
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- We used both real-life logs and models and artificial ones,
 - With traces on the entire fitness spectrum $[0..1]$,
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- Traces are processed one by one (to eliminate caching effects),
 - All multi-threading is removed (to make the comparison fair).

ARTIFICIAL MODEL

Model F (299 transitions, 362 places)



REAL MODEL (FELIX' SEPSIS MODEL)



CONCLUSIONS

Computing Alignments is hard, but

- Current implementations are fairly efficient, but
- They occasionally fail on specific trace/model combinations, and
- Their performance is hard to predict.

So, we provide a heuristic algorithm, which

- Provides sub-optimal alignments,
- In predictable time and memory,

CHALLENGES

- Handling models with lots of tau transitions,
 - Tau transitions are hard to predict!
- In many structured loops
 - Structured loops imply transition invariants which are very bad for the matrix equations
- Handling partially ordered traces like in the regular alignments
 - Purely an implementation issue