



Experiments with MIP versus CP for Paint Production Scheduling

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Paint Production Scheduling

- Multiple stages – dispersion, mixing, packing
- Stages coupled – paint needs a vessel
- Disparate equipment in each stage
- Disparate batches to schedule
- Challenging problem



Contents

- Description of problem
- Initial MIP model
- CP model and enhancements
- Enhanced MIP model
- And back to CP
- Conclusions

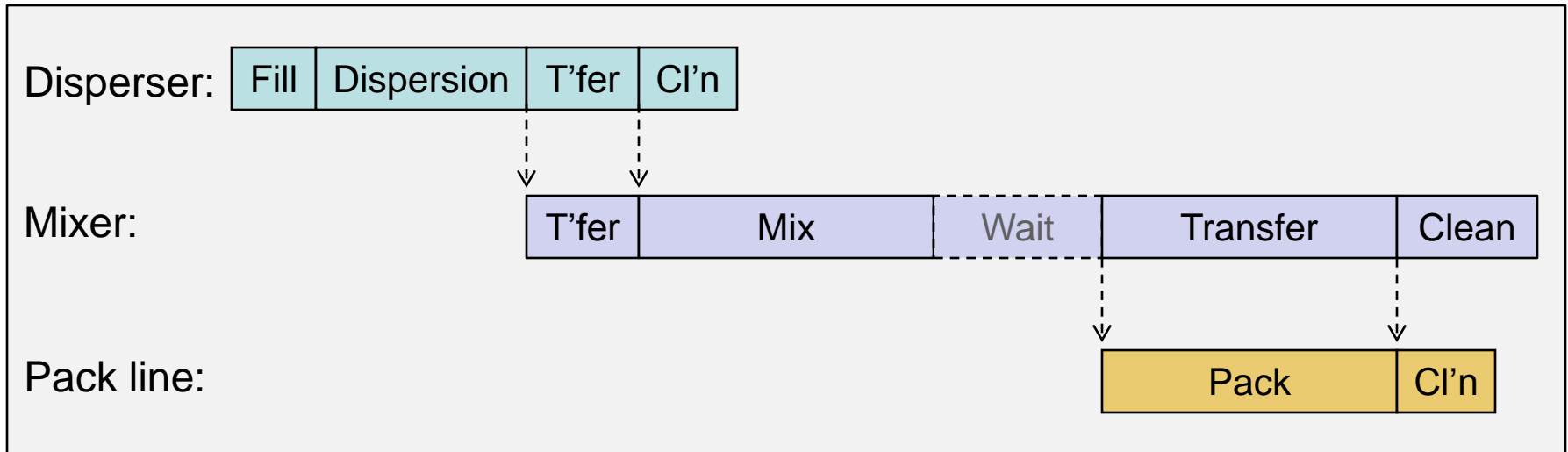


The Problem

- 100 batches to schedule (2 weeks)
 - Pre-processed make-to-order + stock
- Due dates
- Choice of (up to) three dispersers
 - Close to 100% utilisation
- Choice of ten mixing vessels
- Choice of (up to) three pack lines
 - Close to 100% utilisation



Lifecycle of One Batch



- Durations (except for Wait) are fixed
- Different for each batch
- Different for each facility (except Mix)
- No changeover considerations



MIP Model

- Continuous-time model
 - Variables for start times (end times)
- Easy constraints:
 - Selection of facilities for each batch
 - Time offsets
 - Start-end on one facility
 - Start/end from one stage to next
 - Lateness
- Hard: prevent clashes on facilities



MIP Model: Prevent Clashes 1

- Disjunctive constraint?
 - Variables: batch a is before b or after b
 - (b after a) \Rightarrow (b starts after a ends)
 - Complicated by facility choice
 - Weak bounds
- Sequencing (vehicle routing)?
 - Variables: batch b follows a on facility
 - Flow constraints
 - Sub-tour elimination
 - Changeover is not an issue – timing is



MIP Model: Prevent Clashes 2

■ Indexing

- Variables: batch b runs at index i on facility f
- Relate batch index start (end) to facility index start (end)
- Facility index start $i+1 \geq$ facility index end i
- Extra start variables by (batch, facility, index)
- Lower bounds on start by analysis of length (sort shortest first)
- Then push out pack bounds by offsets
- Leads to better objective bound and branch decisions



Indexing Formulation

- $\text{index_run}_{bfi} \in \{0,1\}$

$$\text{index_start}_{bfi}$$

$$\text{facility_start}_{fi}$$

$$\text{batch_start}_{bs}$$

b	batch
f	facility
i	index
s	stage

- $\text{index_start}_{bfi} \geq \text{Earliest_start}_{fi} \cdot \text{index_run}_{bfi}$

- $\text{index_start}_{bfi} \leq \text{Latest_start}_{fi} \cdot \text{index_run}_{bfi}$

- $\text{facility_start}_{fi} = \sum_b \text{index_start}_{bfi}$

- $\text{batch_start}_{bs} = \sum_{f \in S, i} \text{index_start}_{bfi}$

- $\text{facility_start}_{fi} \geq \text{facility_start}_{fi-1} + \sum_b \text{Length}_{bf} \cdot \text{index_run}_{bfi} - \dots$



Problem Size

- All methods lead to quadratic model size
- Impossible to solve in one go
- Solve in steps:
 - From n earliest-due unfixed batches
 - Select m batches, assign facilities and sequence
 - Fix the selected m batches, repeat
- Choice of n, m ?

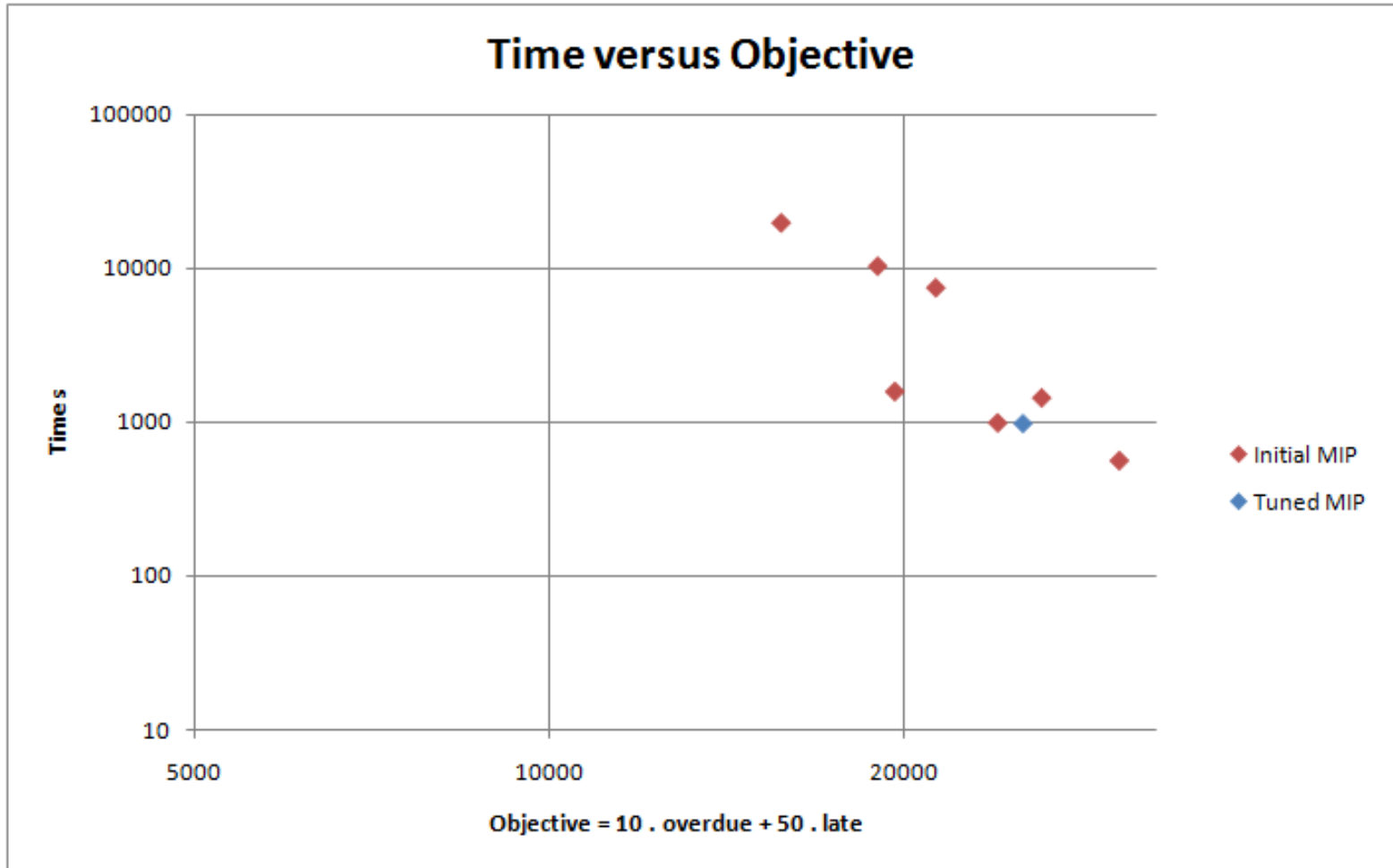


Mosel Implementation

- One .mos file
- Define all variables at the start
- In each step:
 - `reset(Problem), with Problem do`
 - Define binaries, fix old choices
 - Constraints including fixed and new batches
 - Solve – Presolve eliminates fixed part
 - Extract choices



MIP Results





Implementation in Kalis 1

- Variables are simpler and fewer
 - Batch is on facility → Facility that batch is on
 - But cannot use strings as identifiers
 - Have to `setname` explicitly
 - Data rounded to whole hours so all `cpvar`
- Constraints mostly easy to write
 - `equiv(vBatRun(bn) >= 1, vBatFac(bn, s) >= 1)`
 - `implies(vBatIndex(bn, fn) = i, vBatStart(bn, s1) = vFacStart(fn, i))`
 - `distribute` expects an array of variables?
 - Use multiple `occurrences` instead



Implementation in Kalis 2

- Have to use main model/submodels
- Not intelligent about optimisation
 - $vLate(bn) + vOverdue(bn) \geq vBatEnd(bn, "PAC") - iDueTime(bn)$
 - $vOver2(bn) = vBatEnd(bn, "PAC") - iDueTime(bn)$
 - $vOverdue(bn) = \text{maximum}(\{vOver2(bn), vZero\})$
- Single step, 6 batches, did not finish



Implementation in Kalis 3

- Vital to specify strategy as well
 - Something like priorities in MIP ...

```
cpbStrat(1) := assign_var(KALIS_INPUT_ORDER,  
    KALIS_MAX_TO_MIN, vBatRun)
```

```
VarSet := {}
```

```
forall (bn in BatFlex)
```

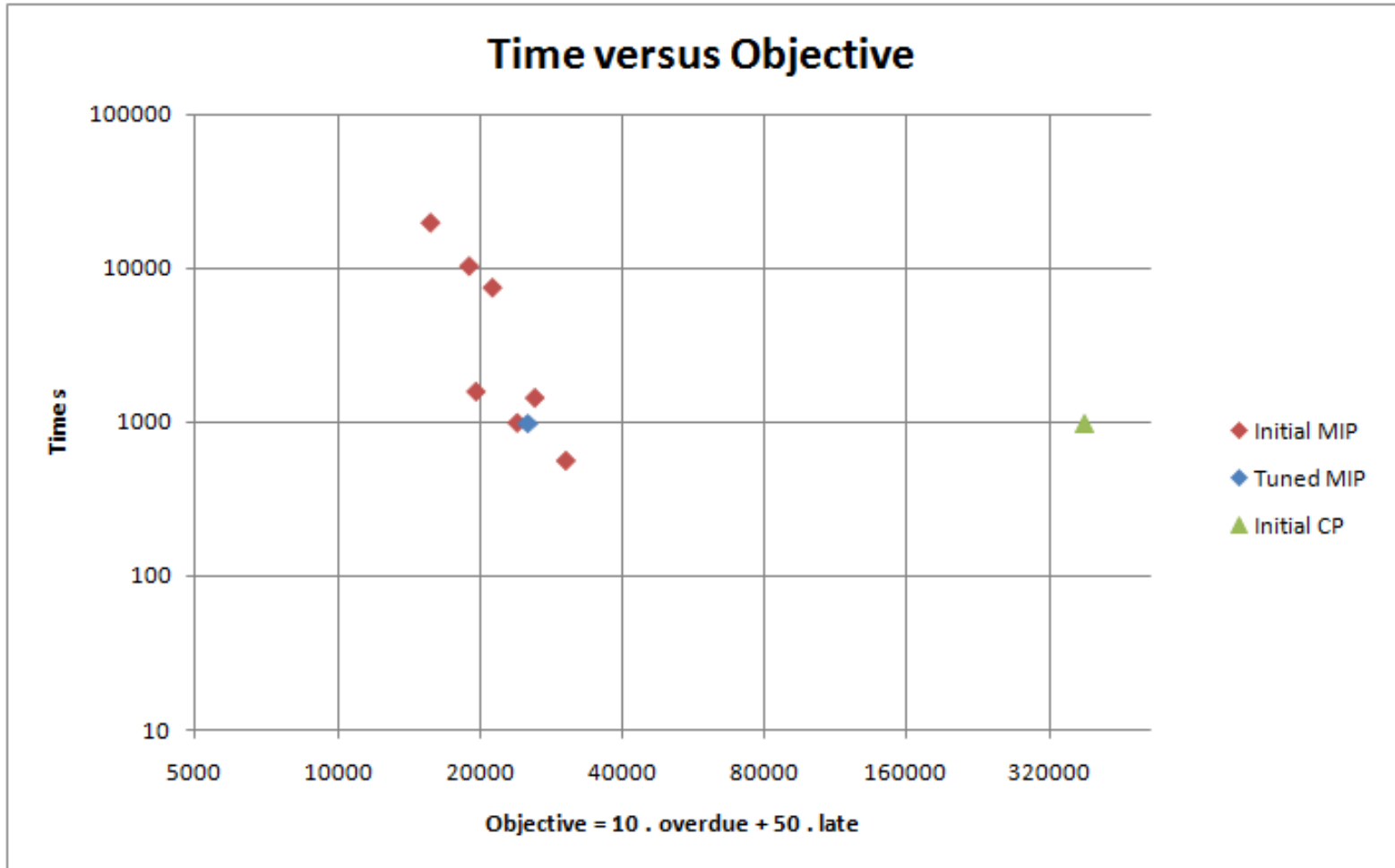
```
    VarSet += {vBatFac(bn, "DIS")}
```

```
cpbStrat(2) :=
```

```
    assign_and_forbid(KALIS_SMALLEST_DOMAIN,  
        KALIS_RANDOM_VALUE, VarSet)
```



First full CP Result





Kalis Enhancement 1

- Analysis of search tree
 - Zoom by selection rectangle
 - Branch path by double click
- Almost all time exploring equivalent mixer assignments
- So shortcut by fixing mixers in code
- Greedy algorithm based on disperser assignments is (almost) optimal
- But how to control call?

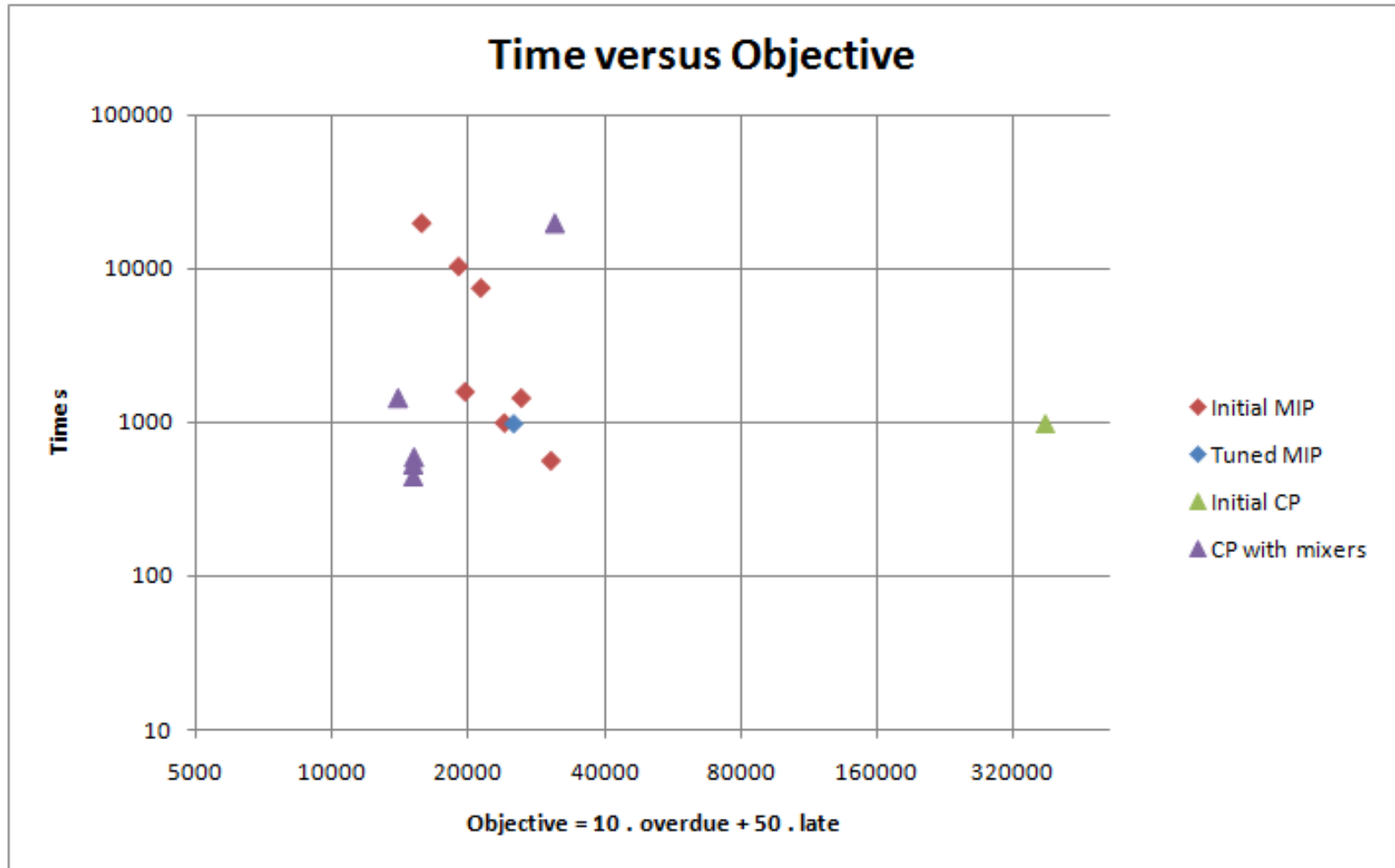


Kalis Enhancement 2

- Call routine as a user-defined strategy ... after disperser strategies
- Fix mixer variables with `setval`
- End with `cp_propagate` (OK if fail)
- Do not return a branching variable
- Must not run again in a descendent node
- Compare current depth to mix fix depth
- Have to create callbacks to know what depth is! (`cp_set_branch_callback`)



Improved CP Result





Fixing Mixers in MIP 1

- Would the same approach work in MIP?
- Apply as a cut manager callback
 - Use `setlb` and `setub`
 - Bounds are carried down tree
- When to apply – not automatic
 - Check for all dispersers at integer values
 - Cannot wait for all dispersers fixed
- Use auxiliary variable as a flag
 - Must have (tiny) positive entry in objective
 - Bound up to indicate mixers fixed - `getlb`

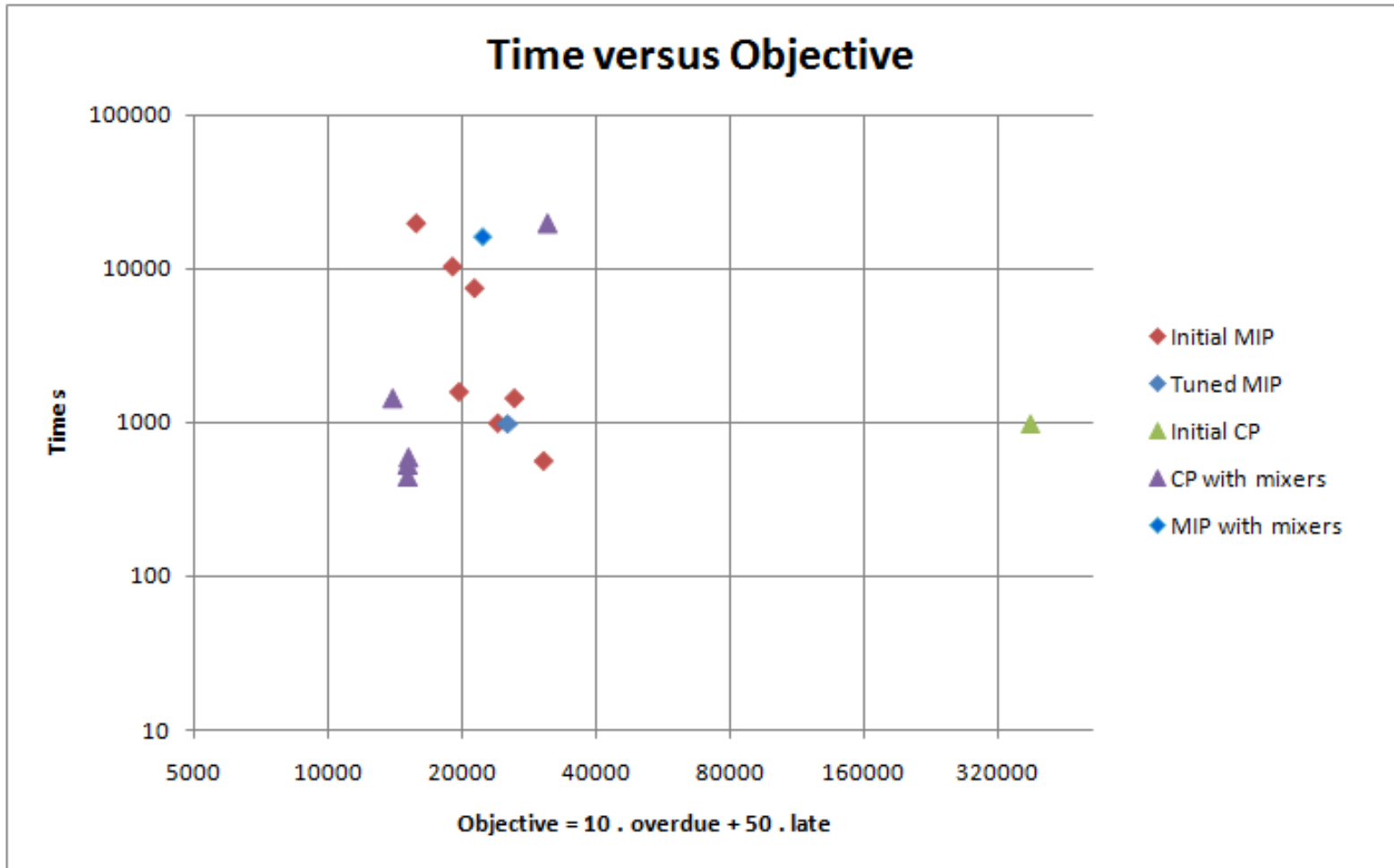


Fixing Mixers in MIP 2

- Errors!
 - Variables no longer existing
 - Conflicting bounds set already
 - Variables changing identity?
- Recommendation with custom cuts:
 - No presolve
 - No MIP presolve
 - No heuristics
 - No cut generation
- Observation – problems caused by:
 - Presolve (any options, even everything switched off)
 - MIP presolve - reduced cost fixing

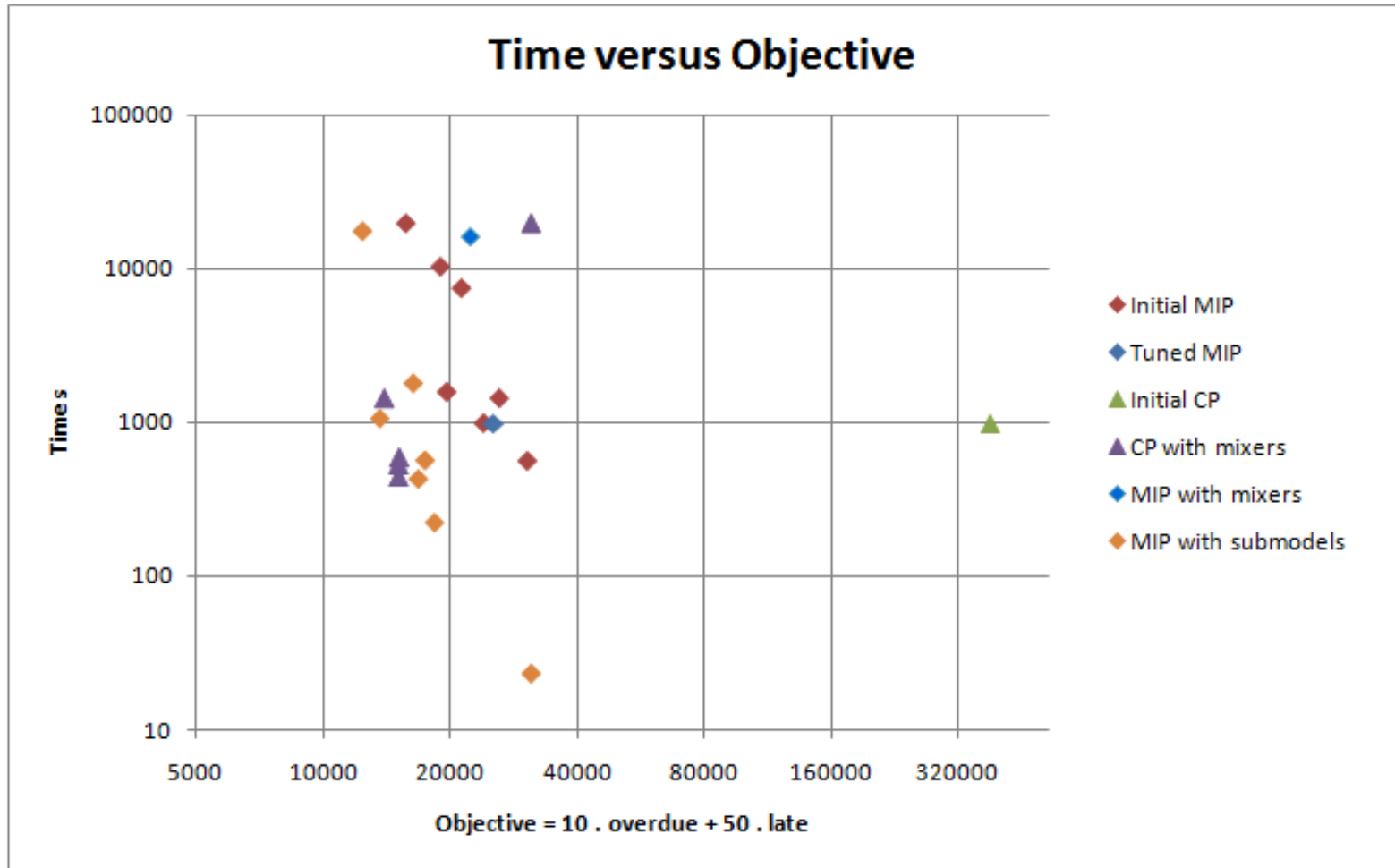


MIP with Fixed Mixers



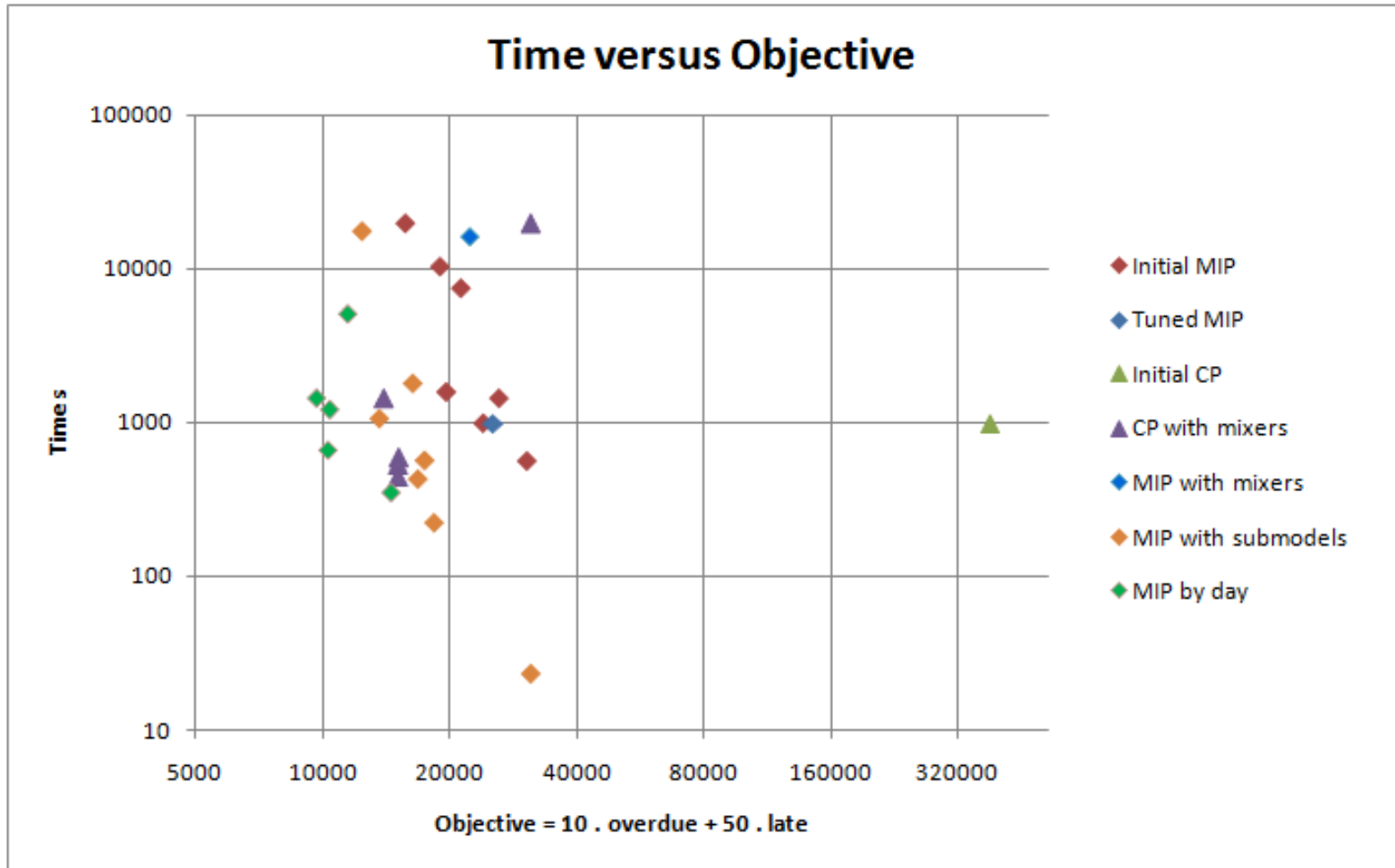


MIP with Submodels



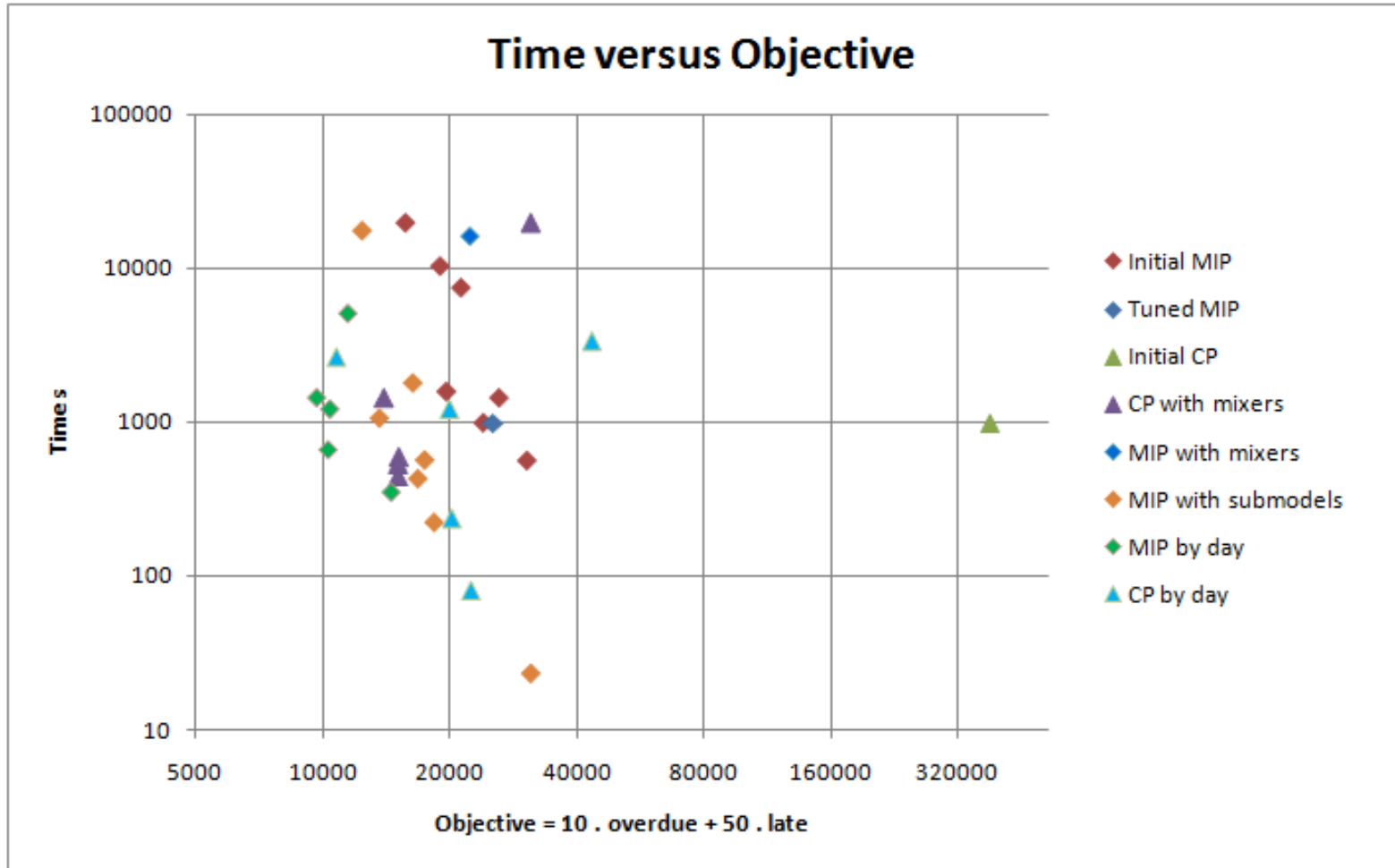


MIP Solving by Day





CP Solving by (Partial) Day





Conclusions

- CP
 - Conceptually easier but traps for the unwary
 - Kalis less well integrated in Mosel
 - Custom strategy vital to get results
 - Exponential time means exponential
- MIP
 - Decent solutions without tuning/customisation
 - Much better than exponential in practice
 - Custom algorithms face disadvantages
 - Hidden complexity – unexpected behaviour

