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# **An alternative MIP formulation for the Military Flight and Maintenance Planning problem**

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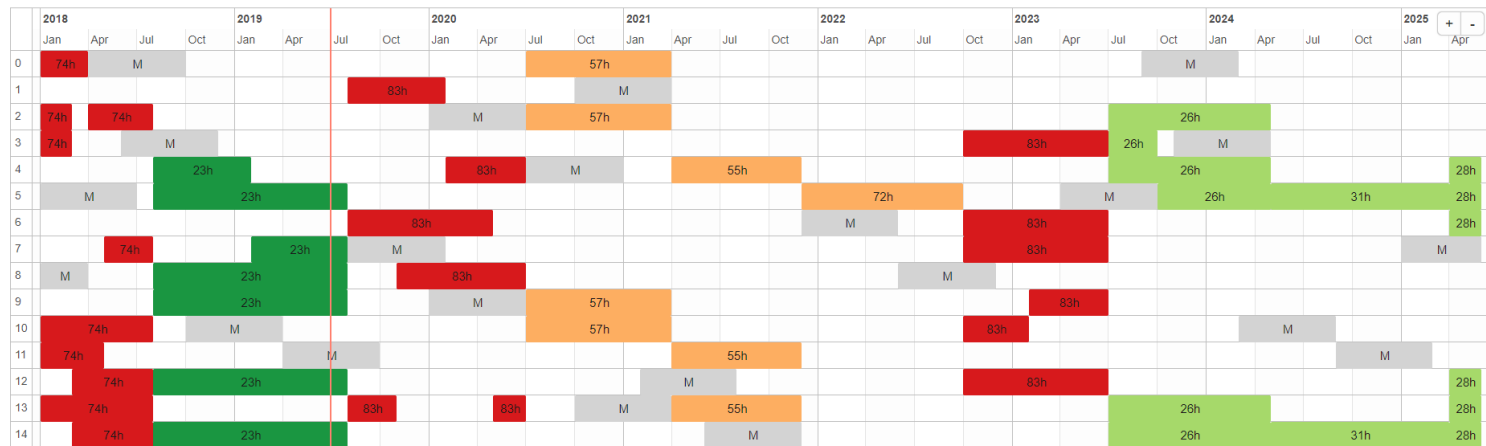
**February, 2020**

# Structure of talk

1. Problem.
2. State of the art.
3. MIP formulations.
4. Machine Learning.
5. Experiments.
6. Conclusions.

# Problem (informally)

Assign both **missions** and **maintenance operations** to a fleet of aircraft in order to maximize availability and minimize costs. **Missions** have fixed start and end times and have particular needs in terms of aircraft and flight hours. **Maintenance operations** have specific rules that govern their frequency and capacity.



# Problem

- A series of  $j \in \mathcal{J}$  pre-scheduled missions are planned along a horizon divided into  $t \in \mathcal{T}$  periods.
- Each mission requires a certain **number**  $R_j$  of aircraft  $i \in \mathcal{I}$  which it employs for  $H_j$  **hours** in each period.
- The heterogeneous fleet requires recurrent preventive maintenance operations (**checks**).
- A check takes exactly  $M^d$  periods and cannot be interrupted. There is a limited **capacity** for maintenances at each period.
- The objective is to **minimize the total number of checks** while **maximizing the status of the fleet** at the end of the horizon.

**Other:** soft constraints on the state of the fleet at each period, initial conditions, default consumption, reduce variance of frequency of checks.

# State of the art

- FMP: Flight and Maintenance Planning problem.
- In [Cho11], US Army aircraft were assigned daily operations over a year to aircraft in order to minimize the maximum number of maintenances.
- In [Koz08], Greek aircraft had monthly assignments of maintenances and flight hours in order to maximize the availability and final state of squadrons.
- In [VVC15], monthly assignments were done and several objectives were taken into account: availability, serviceability and final state.
- In [SY18], a generalization for different types of maintenances and capacities was done.
- In [Pes+20], the problem was proved NP-Complete.

# Previous formulation

- $a_{jti} : =1$  if mission  $j \in J$  in period  $t \in \mathcal{T}_j$  is realized with aircraft  $i \in \mathcal{I}_j$ , 0 otherwise.
- $m_{it} : =1$  if aircraft  $i \in I$  starts a check in period  $t \in \mathcal{T}$ , 0 otherwise.
- $u_{it} :$  flown time (continuous) by aircraft  $i \in I$  during period  $t \in \mathcal{T}$ .

$$u_{it} \geq \sum_{j \in \mathcal{J}_t \cap \mathcal{O}_i} a_{jti} H_j \quad t = 1, \dots, \mathcal{T}, i \in \mathcal{I}$$

$$u_{it} \geq U^{\min} (1 - \sum_{t' \in \mathcal{T}_t^s} m_{it'}) \quad t = 1, \dots, \mathcal{T}, i \in \mathcal{I}$$

$$rft_{it} \leq rft_{i(t-1)} - u_{it} + H^M m_{it} \quad t = 1, \dots, \mathcal{T}, i \in \mathcal{I}$$

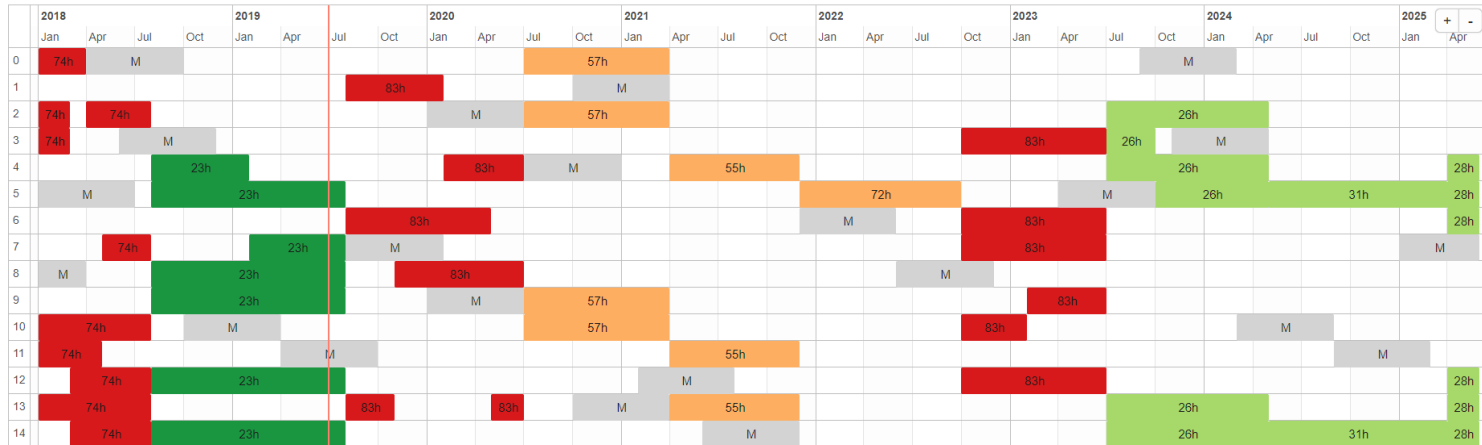
$$rft_{it} \in [0, H^M] \quad t \in \mathcal{T}, i \in \mathcal{I}$$

# New formulation

- $a_{ijtt'}$  : =1 if aircraft  $i$  starts an assignment to mission  $j$  at the beginning of period  $t$  and finishes at the end of period  $t'$ , zero otherwise.
- $m_{ip}$  : =1 if aircraft  $i$  uses check pattern  $p$ , zero otherwise.
  - each pattern  $p$  has a single feasible combination of check starts for an aircraft during the whole planning (usually only 1-2 checks per aircraft).

$$\sum_{(j,t,t') \in \mathcal{JTT}_{ic}} a_{ijtt'} H'_{jtt'} + U'_{tc} \leq H^M + M(1 - m_{ip})$$

$$i \in \mathcal{I}, p \in \mathcal{P}, c \in \mathcal{C}_p$$





# Formulation

$$\text{Max} \sum_{i \in \mathcal{I}, p \in \mathcal{P}} m_{ip} \times W_p$$

$$\sum_{i \in \mathcal{I}, p \in \mathcal{P}_t} m_{ip} \leq C^{\max} \quad t \in \mathcal{T}$$

$$\sum_{i \in \mathcal{I}_j, (t_1, t_2) \in \mathcal{T}_{jt}} a_{ijt_1 t_2} \geq R_j \quad j \in \mathcal{J}, t \in \mathcal{T}_{\mathcal{J}_j}$$

$$\sum_{p \in \mathcal{P}_i} m_{ip} + \sum_{j \in \mathcal{J}_i \cap \mathcal{J}_i} \sum_{(t_1, t_2) \in \mathcal{T}_{jt}} a_{ijt_1 t_2} \leq 1 \quad t \in \mathcal{T}, i \in \mathcal{I}$$

$$\sum_{(j, t, t') \in \mathcal{JTT}_{ic}} a_{ijtt'} H'_{jtt'} + U'_{tc} \leq H^M + M(1 - m_{ip})$$

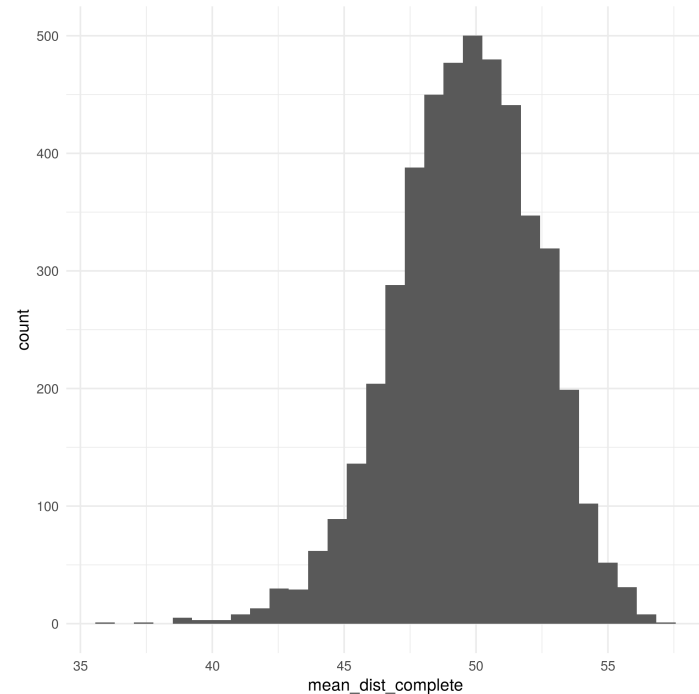
$$i \in \mathcal{I}, p \in \mathcal{P}, c \in \mathcal{C}_p$$

# Is this better?

1. It uses 3 times the number of constraints and 3 times the number of variables.
  - variables: 11000 => 28000.
  - constraints: 13000 => 48000.
2. It is still better. Better lineal relaxation, better performance.

# Distance between maintenances

- The distance between maintenance has a maximum of  $E^M$  periods.
- Depending on the instance, the optimal distance can be shorter.
- This distance conditions the total number of patterns to create.



# Forecasting + Optimization

## We want to:

1. Train a statistical model to predict the mean distance between maintenances for any given instance.
2. Use this information to limit all possible combinations of patterns to generate.

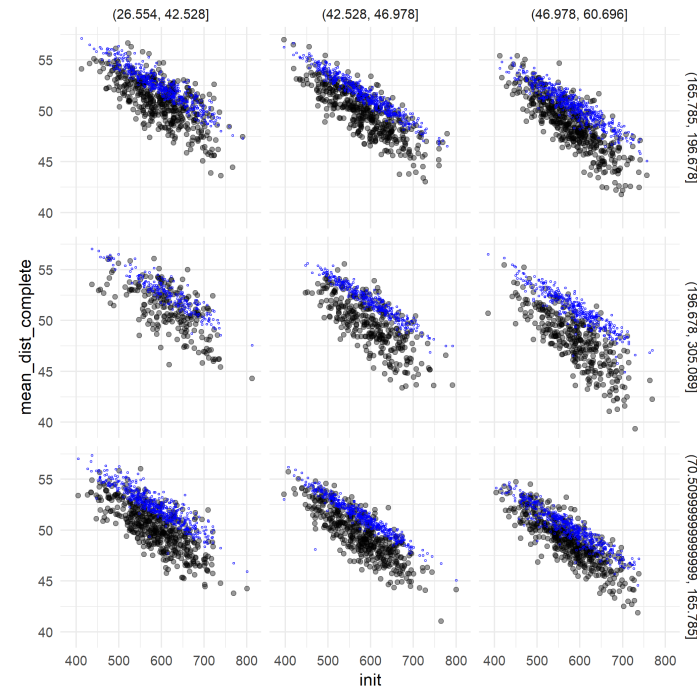
## Benefits:

1. **Performance:** a smaller model is easier to solve.
2. **User feedback:** direct feedback about the solution without needing to solve any model.
3. **More stable solutions:** Every aircraft flies an amount that is closest to the mean of the fleet.

**The better we're able to predict the optimal distance between maintenances for the whole fleet, the less optimality we will lose**

# Prediction model

- **Technique:** *Quantile regressions* to estimate upper and lower bounds.
- **Training:** 5000 small instances.
- **Input features:**
  - mean flight demand per period,
  - total remaining flight hours at start (init),
  - variance of flight demand,
  - demand of special missions,
  - number of period where flight demand is cut in two.
- **Output features:** mean distance between maintenances.



# Experiments

- Number of instances: medium (1000), large (1000) and very large (1000).
- Time limit at 3600 seconds.
- We seeded instance generation for better comparison.
- CPLEX running 1 thread.

Largest instances have 60 aircraft, 90 periods, ~30 missions (4 active missions at any given time).

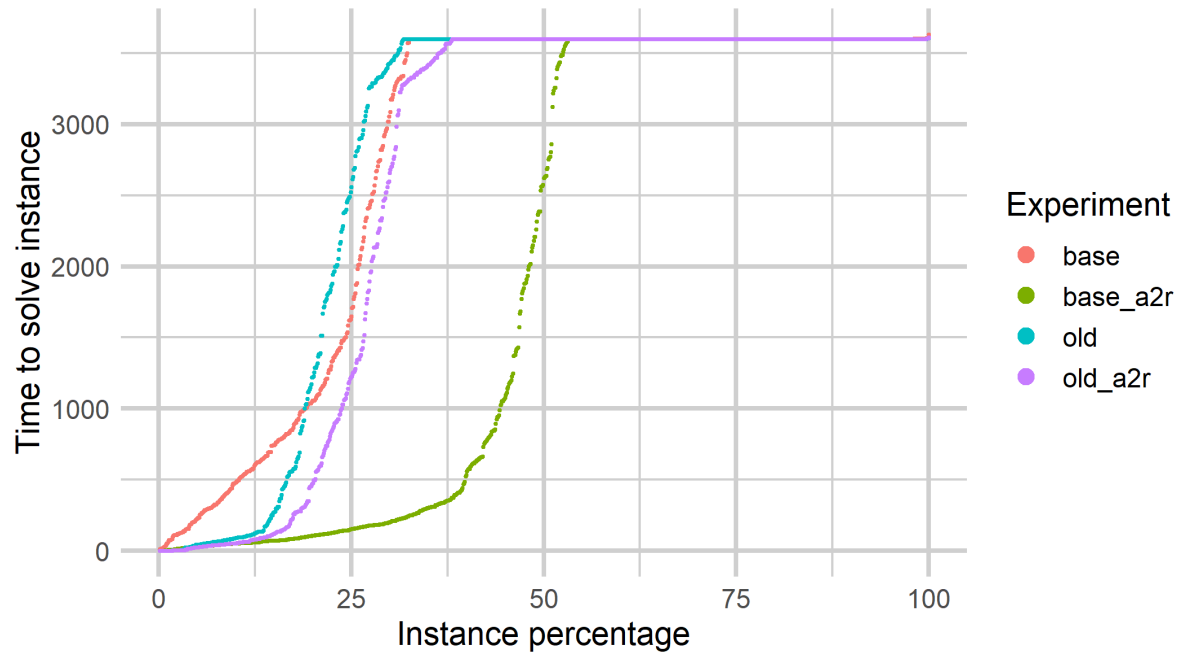
1. Create forecasting model based in 5000 small instances.
2. Use forecasting model to predict bounds on distance between maintenances:  $\hat{\mu}_{t'-t}^{lb}$ ,  $\hat{\mu}_{t'-t}^{ub}$
3. Implement the pseudo-cut:

$$\begin{aligned} m_{ip} = 0 & \quad p_{t'} - p_t < \hat{\mu}_{t'-t}^{lb} - tol \\ m_{ip} = 0 & \quad p_{t'} - p_t > \hat{\mu}_{t'-t}^{ub} + tol \end{aligned}$$

1. Recycling.

# How good is it (performance)

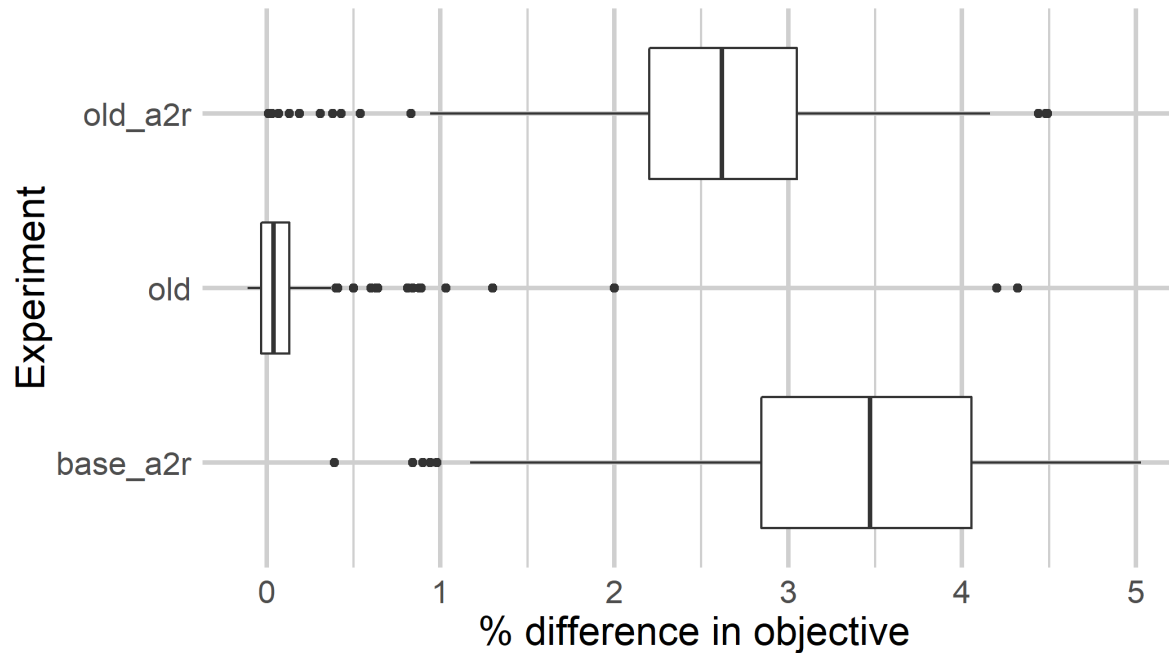
Faster solutions, more solutions.



# How good is it (optimality)

For instances where an optimal solution was found (optimum degradation):

- 95% of instances had less than 4% gap with real optimal.





# Further steps

- **Better predictions** with better features, or predicting several characteristics of optimal solutions.
- **Predict a distribution** and sample patterns from the distribution instead of predicting patterns.
- **Warm-start Column Generation** with a selected subset of potentially good patterns.
- **Automatize prediction** so it can be easily integrated in other problems.

# References

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