# AMMM - Course project

Master in Research and Innovation in Informatics

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#### Problem Statement



#### Main requirements

- 1. Each contestant will play exactly once against each of the other contestants.
- 2. Each round will consist of  $\frac{n-1}{2}$  matches.
- 3. Players will play 50% of their games as white, 50% will be played as black.

### Subtle requirements

- A player can only play up to 1 game per round
- A player can't play against himself

# Problem Statement: Inputs & Outputs



### Inputs

- Number of contestants, n. Has to be odd
- $\blacksquare$  Matrix of points per day per player,  $p_{n\times n}$

### Outputs

■ Schedule with the set of pairings  $\{\{r_1, p_i, p_j\}, \ldots, \{r_n, p_k, p_h\}\}$  that maximizes total score. Ensured to be optimal if it's obtained through the ILP.

### **Problem Statement: Definitions**



In order to specify the constraints, we need to specify the sets and variables we're going to work with:

- M(x,y) matches played among x and y (1)
- $\blacksquare$  R(r) matches played at round r (2)
- $lackbox{ }W(p)$  matches played by player p as white (3)
- lacksquare B(p) matches played by player p as black (3)
- lacksquare G(p,r) games played by p at round r (4)
- F(r) free players at round r (5)

- Each contestant will play exactly once against each of the other contestants.
- 2. Each round will consist of  $\frac{n-1}{2}$  matches.
- 3. Players will play 50 % of their games as white, 50 % will be played as black.
- 4. Players can play up to 1 match per round
- Objective function



Every set will be constructed from a boolean multidimensional array. matches[w][b][r] will be 1 whenever player w plays player b in round r, and 0 otherwise.

#### Set constructions

```
\begin{split} &M(x,y) = \{\{x,y,r\},\ r \in [1,Rounds] \mid \mathsf{matches}[x][y][r] = 1 \lor \mathsf{matches}[y][x][r] = 1\} \\ &F(r) = \{p,\ p \in [1,n] \mid \mathsf{matches}[p][o][r] = 0 \land \mathsf{matches}[o][p][r] = 0 \quad \forall o \in [1,n]\} \\ &W(p) = \{\{p,b,r\},\ r \in [1,Rounds], b \in [1,n] \mid \mathsf{matches}[p][b][r] = 1\} \\ &B(p) = \{\{w,p,r\},\ r \in [1,Rounds], w \in [1,n] \mid \mathsf{matches}[w][p][r] = 1\} \\ &R(r) = \{\{w,b,r\},\ w,b \in [1,n] \mid \mathsf{matches}[w][b][r] = 1\} \\ &G(p,r) = \{\{o,p,r\},\ o \in [1,n] \mid \mathsf{matches}[p][o][r] = 1 \lor \mathsf{matches}[o][p][r] = 1\} \end{split}
```

# Integer Linear Programming Model: Constraints



$$|M(x,y)| = 1 \quad \forall x, y \in P \mid x \neq y$$
 (1)

$$|R(r)| = \frac{n-1}{2} \quad \forall r \in [1, \text{ Rounds}]$$
 (2)

$$|W(p)| = \frac{n-1}{2} \quad \forall p \in P \tag{3}$$

$$|G(p,r)| \le 1 \quad \forall p \in P, r \in [1, \text{ Rounds}]$$
 (4)

- 1. Each contestant will play exactly once against each of the other contestants.
- 2. Each round will consist of  $\frac{n-1}{2}$  matches.
- 3. Players will play  $50\,\%$  of their games as white,  $50\,\%$  will be played as black.
- 4. Players can play up to 1 match per round

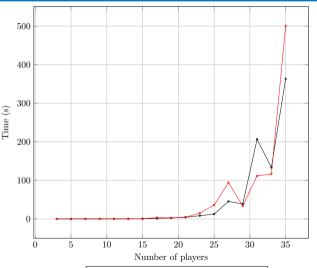
## Integer Linear Programming Model: Redundant constraints



Redundant constraints might appear to make the model faster but they seem make it slower in the long run

$$|M(x,x)| = 0 \quad \forall x \in P$$

$$|B(p)| = \frac{n-1}{2} \quad \forall p \in P$$



 $\longrightarrow$  ILP non-redundant  $\longrightarrow$  ILP redundant



## **Greedy cost function**

$$q(c, day) = c.points\_per\_day[day]$$

### Algorithm Greedy algorithm

- 1: Players  $\leftarrow$  Set of Players
- 2:  $rests \leftarrow \{\}$
- 3: for day in 0..days do
- 4:  $playersToRest \leftarrow filter Players(p) | p.hasNotRested$
- 5: sortedPlayers  $\leftarrow$  sort playersToRest(p) by q(p,day) (DESC)
- 6:  $rests[day] \leftarrow sortedPlayers.first()$

### Metaheuristics: Local Search



### Algorithm Local Search

- 1: for i in 0..days do
- best swap points  $\leftarrow 0$
- best swap ← i
- for j in 0..days do 4:
- change = EvaluateRestSwap(i,j)6:
- if change > best\_swap\_points then
- best swap points ← change 7:
- $best_swap \leftarrow j$ 8:
- rests[i] ↔ rests[best\_swap] 9:

### Metaheuristics: GRASP



## **Algorithm** constructRCL(day)

- 1:  $q_{max} \leftarrow \text{sortedPlayers.first().points[d]}$
- 2:  $q_{min} \leftarrow \text{sortedPlayers.last().points[d]}$
- 3:  $RCL_{max} \leftarrow \{p \in sortedPlayers \mid p.points[d] >= q_{max} \alpha \cdot (q_{max} q_{min})\}$

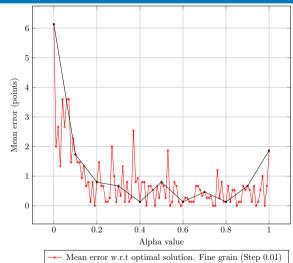
### **Algorithm** GRASP

- 1:  $rests \leftarrow \{\}$
- 2: for day in 0..days do
- 3:  $playersToRest \leftarrow filter Players(p) \mid p.hasNotRested$
- 4: sortedPlayers  $\leftarrow$  sort playersToRest(p) by q(p,day) (DESC)
- 5:  $RCL \leftarrow constructRCL(day)$
- 6: select  $p \in RCL$  randomly
- 7:  $rests[day] \leftarrow p$

## **GRASP:** Parameter tuning



- Figure shows the arithmetic mean error with respect to the optimal solution for each of the instances.
- We keep the smallest alpha that gives the minimum mean error.



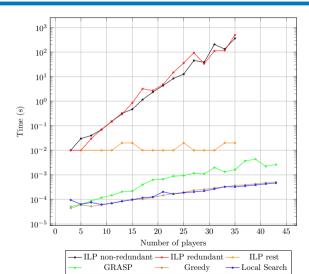
Mean error w.r.t optimal solution. Fine grain (Step 0.01)

Mean error w.r.t optimal solution. Coarse grain (Step 0.1)

#### Results: Time



- Greedy and Local Search need approximately same time to reach the solution. They're fastest but their quality of the solution is too low.
- GRASP sacrifices some runtime performance to improve the quality of the solution.
- ILPs obtain the optimal solution at cost of being several orders of magnitude slower due to the complexity of creating valid pairings.
- ILP rest obtains the optimal solution just computing the rest day for each player. It is not much slower than GRASP and for large instances it can be even faster.



## Results: Quality of solutions



- Greedy offers the worst solution for every instance
- Greedy + Local Search improves the quality of the solution taking practically the same time as Greedy.
- GRASP commonly reachs the optimal solution, having a good improvement over Local Search.

