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
- lacksquare Matrix of points per day per player, $p_{n imes n}$

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)
- lacksquare 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)

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

Integer Linear Programming Model: Constraints



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

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

$$|W(p)| = \frac{n-1}{2} \quad \forall r \in [1, \text{ Rounds}], \forall p \in P \quad \text{(3)}$$

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

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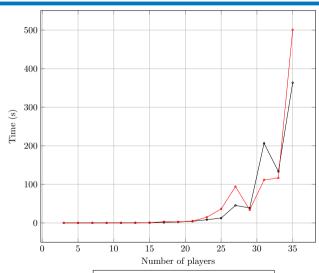
Integer Linear Programming Model: Redundant constraints



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

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

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



 \longrightarrow ILP non-redundant \longrightarrow ILP redundant



Greedy cost function

$$q(c, day) = c.points_per_day[day]$$

Algorithm Greedy algorithm

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

Metaheuristics: Local Search



Algorithm Local Search

- 1: for i in 0..days do
- 2: best swap points $\leftarrow 0$
- 3: best_swap \leftarrow i
- 4: **for** j in 0..days **do**
- 5: change = EvaluateRestSwap(i,j)
- 6: **if** change > best_swap_points **then**
- 7: best_swap_points \leftarrow change
- 8: best_swap \leftarrow j
- 9: $rests[i] \leftrightarrow rests[best_swap]$

Metaheuristics: GRASP



Algorithm constructRCL(day)

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

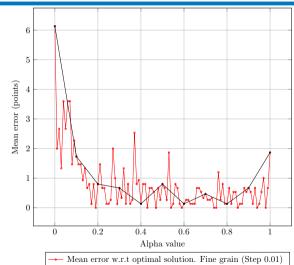
Algorithm GRASP

- 1: $rests \leftarrow \{\}$
- 2: for day in 0..days do
- 3: $RCL \leftarrow constructRCL(day)$
- 4: select $p \in RCL$ randomly
- 5: rests[day] ← p

GRASP: Parameter tuning



- Hemos probao no se que no se cuantos
- Muchos alphas nos valen porque hay muchos 0 calvo de mierda

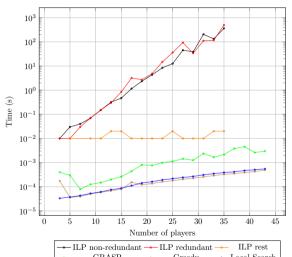


→ Mean error w.r.t optimal solution. Coarse grain (Step 0.1)

Results: Time



- ILP rest blabla
- texto



GRASP Greedy -- Local Search

Results: Quality of solutions



- ILP rest blabla
- texto

