



POSITIONING IN SOCCER WITH AI

An automated approach to find heuristic-based solutions for defensive positioning of soccer players.

MPE – ProDEI – Tiago Mendes-Neves – up201406104@fe.up.pt

Motivation

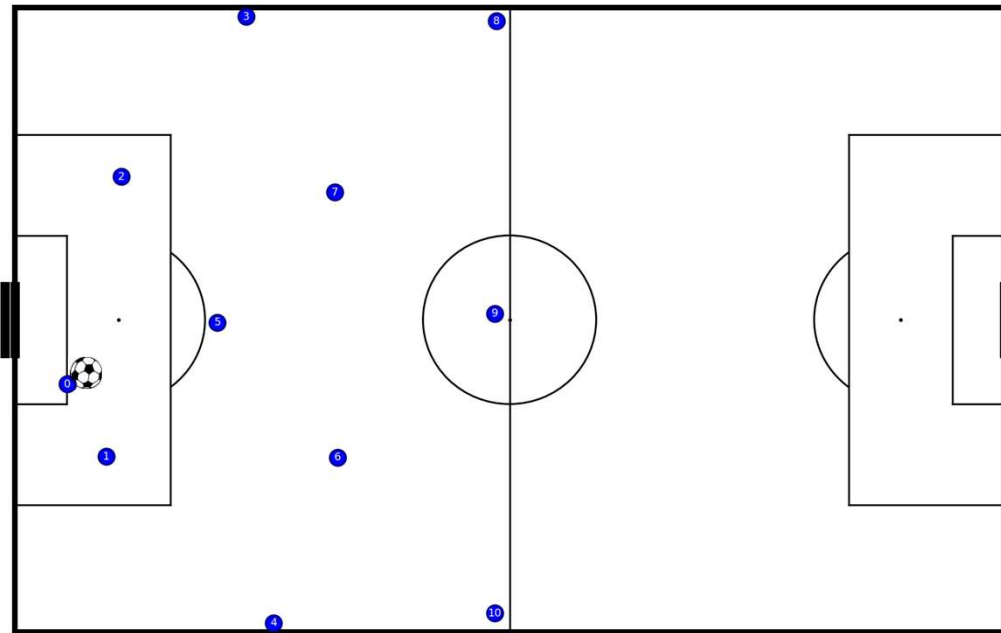
Preparation of a soccer game has a very extensive checklist.

- Defensive positioning in certain game situations is one item.

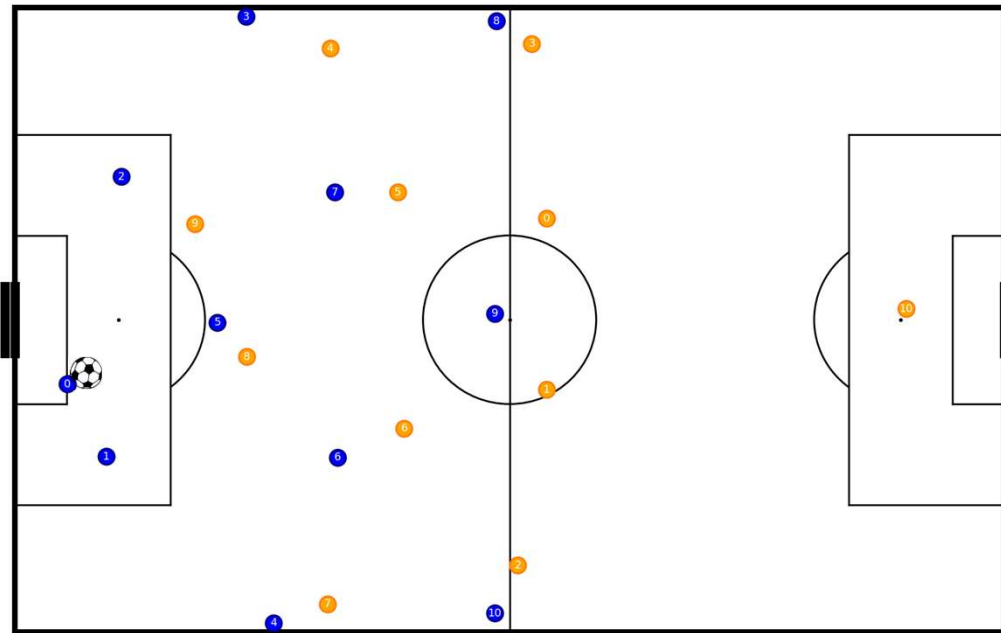


Can AI (Optimization algorithms) help in this regard?

Problem definition



- **11** opposition players
 - Dependent on the use case
 - E.g., Goal kick
 - 10 (x,y) tuples

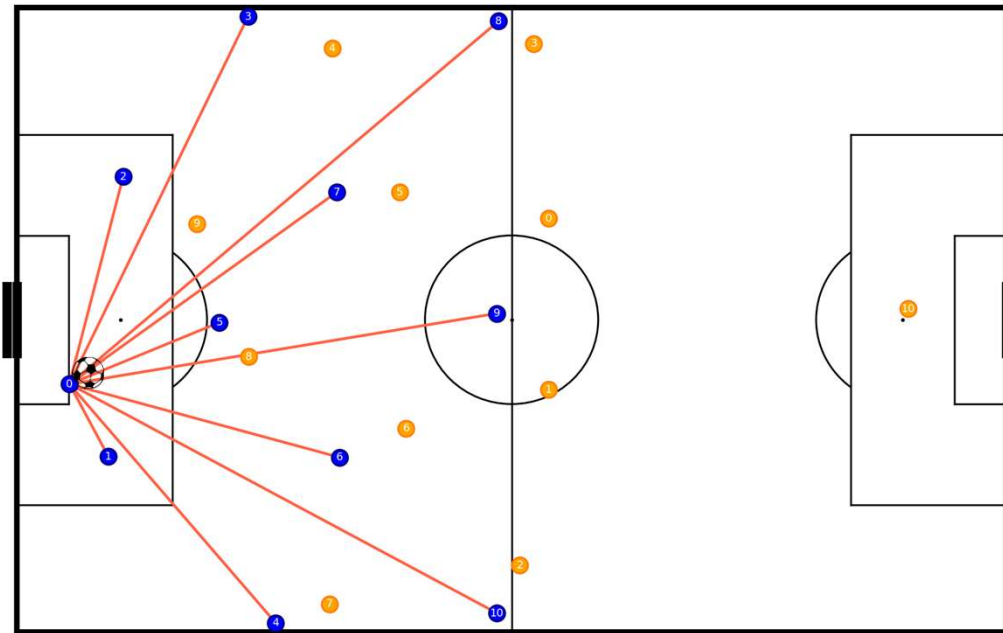


Problem definition

- 11 opposition players
 - Dependent on the use case
 - E.g., Goal kick
 - 10 (x,y) tuples
- We want to position 10 defensive players
 - 10 (x,y) tuples

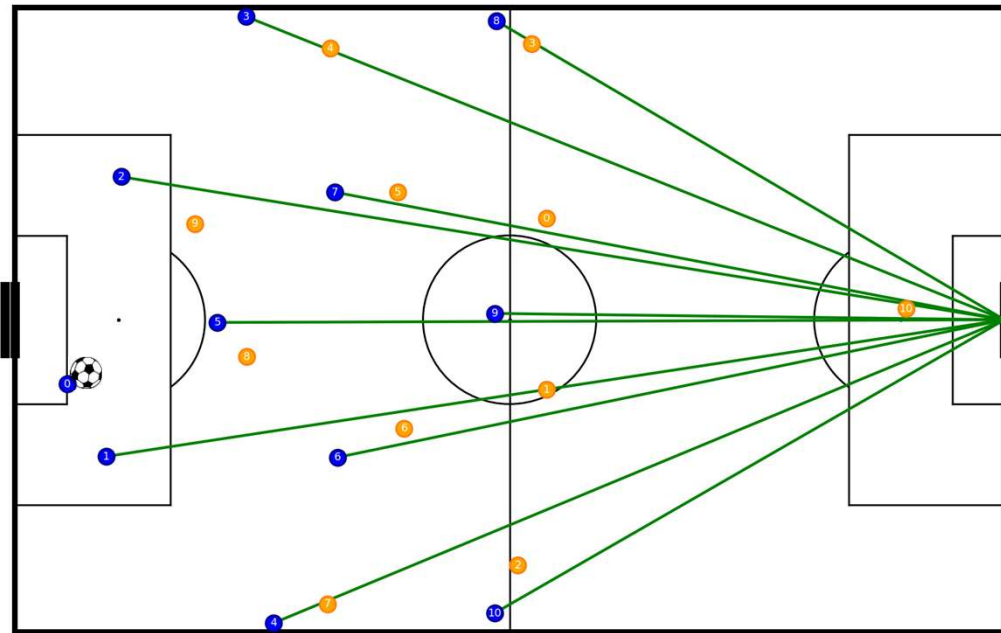
Criteria 1

- Maximize coverage of pass lines

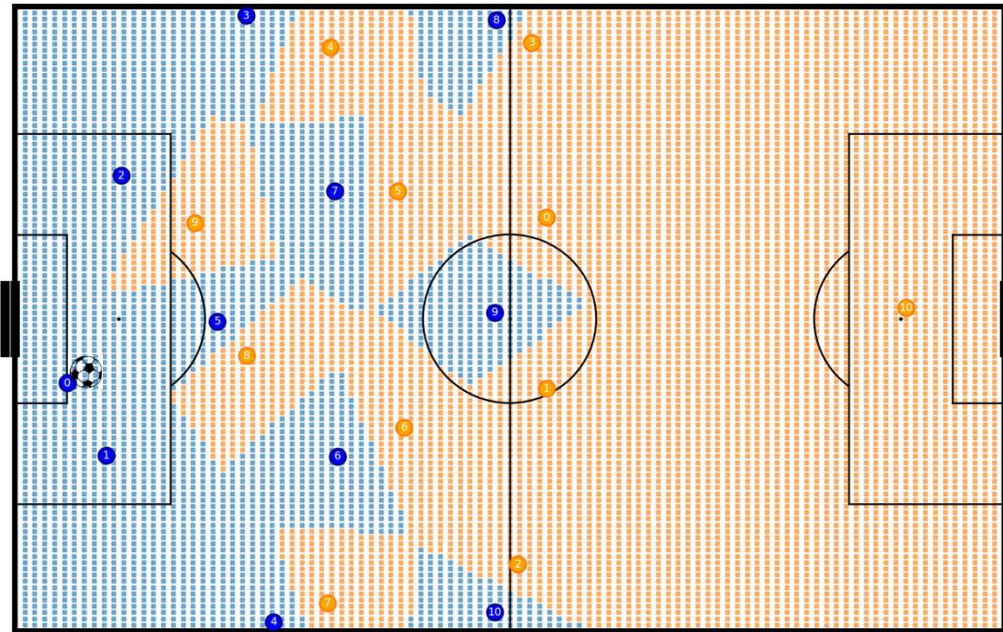


Criteria 2

- Maximize coverage of pass lines
- Maximize coverage of the goal line



Criteria 4



- Maximize coverage of pass lines
- Maximize coverage of the goal line
- Minimize the distance to opponent players
- Maximize pitch control area

Criteria 5



From cafepress.com/+soccer-referee+wall-art

- Maximize coverage of pass lines
- Maximize coverage of the goal line
- Minimize the distance to opponent players
- Maximize pitch control area
- Follow the rules

Heuristic

- Each criterion Cx will have a quantified value (we will normalize them between $[0, 1]$).
- The user will define a weight Wx for each criterion $[0, 1]$.

- $$H(x) = \sum^x Wx * Cx$$

Algorithms



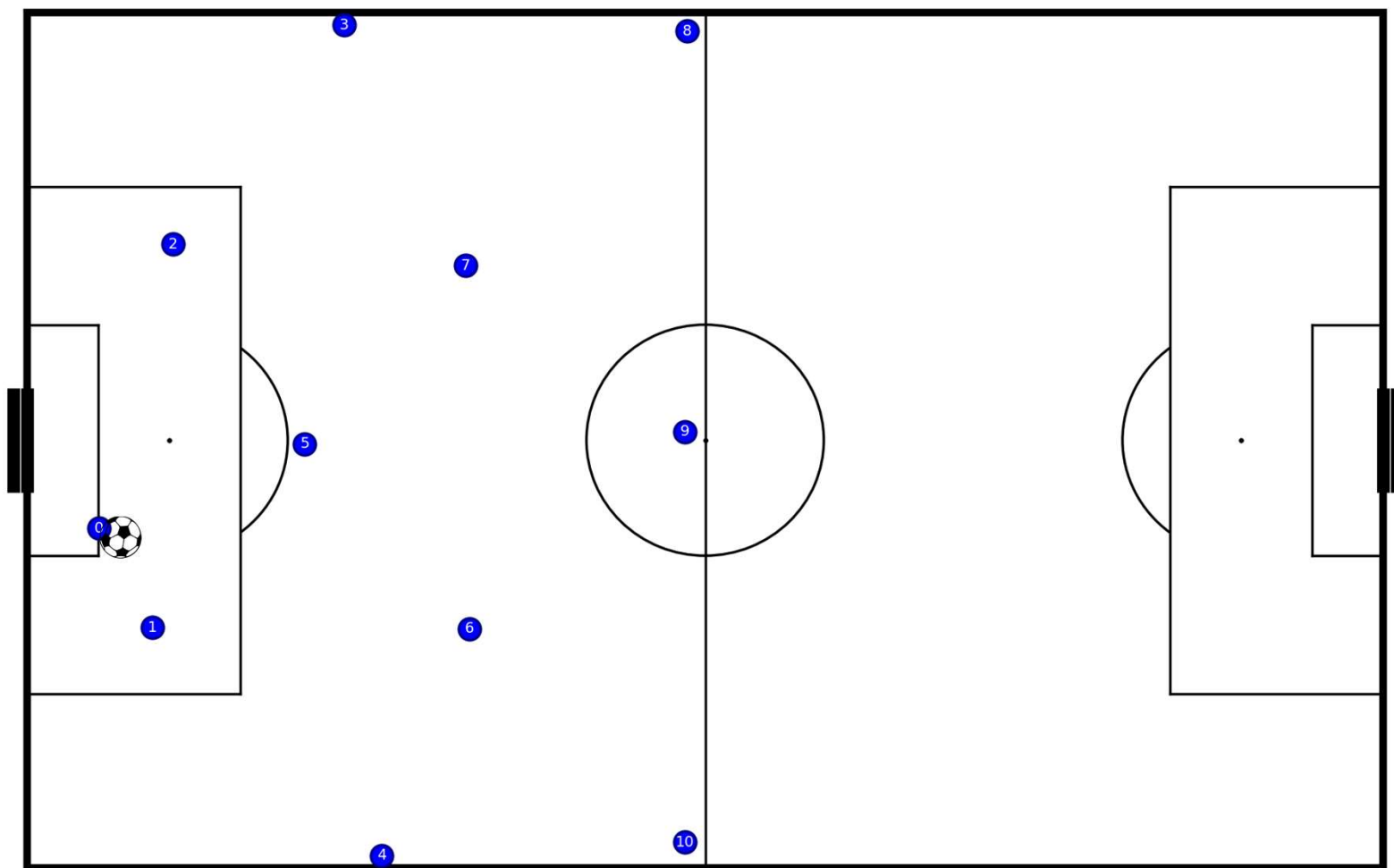
Optuna (TPE/CMA-ES)



Hill Climbing,
Simulated annealing

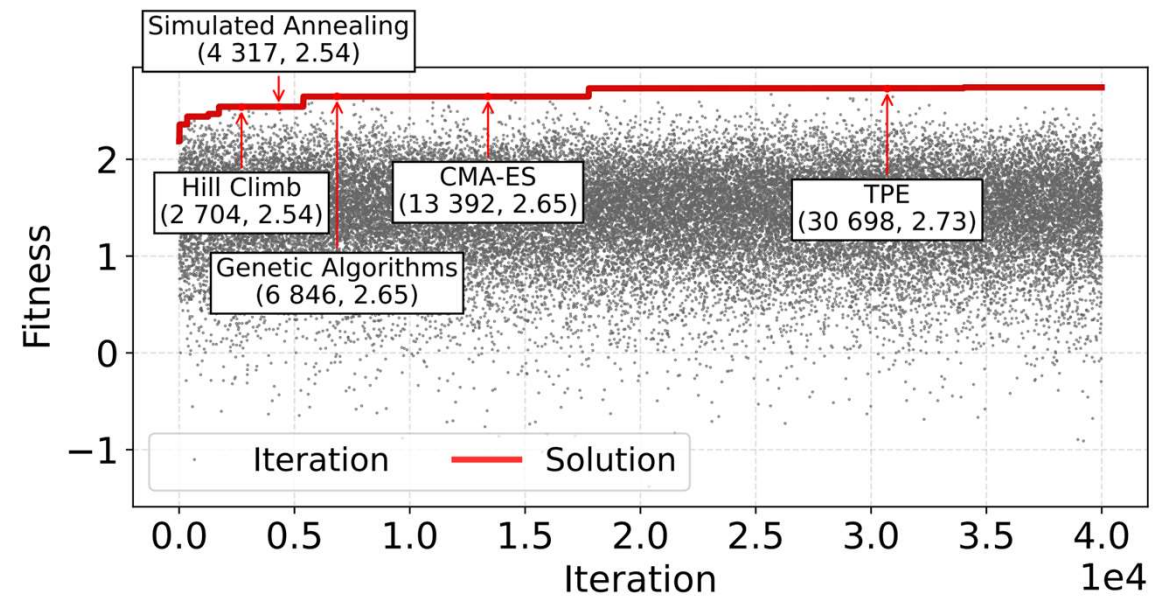


Genetic algorithms

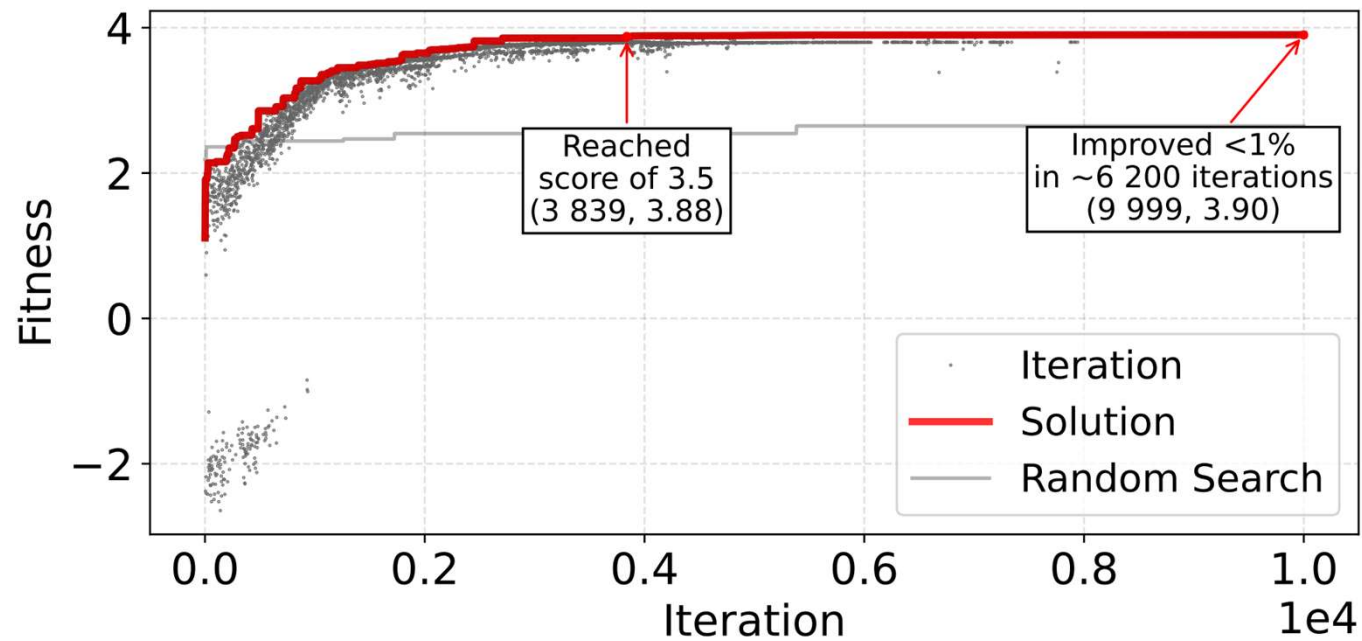


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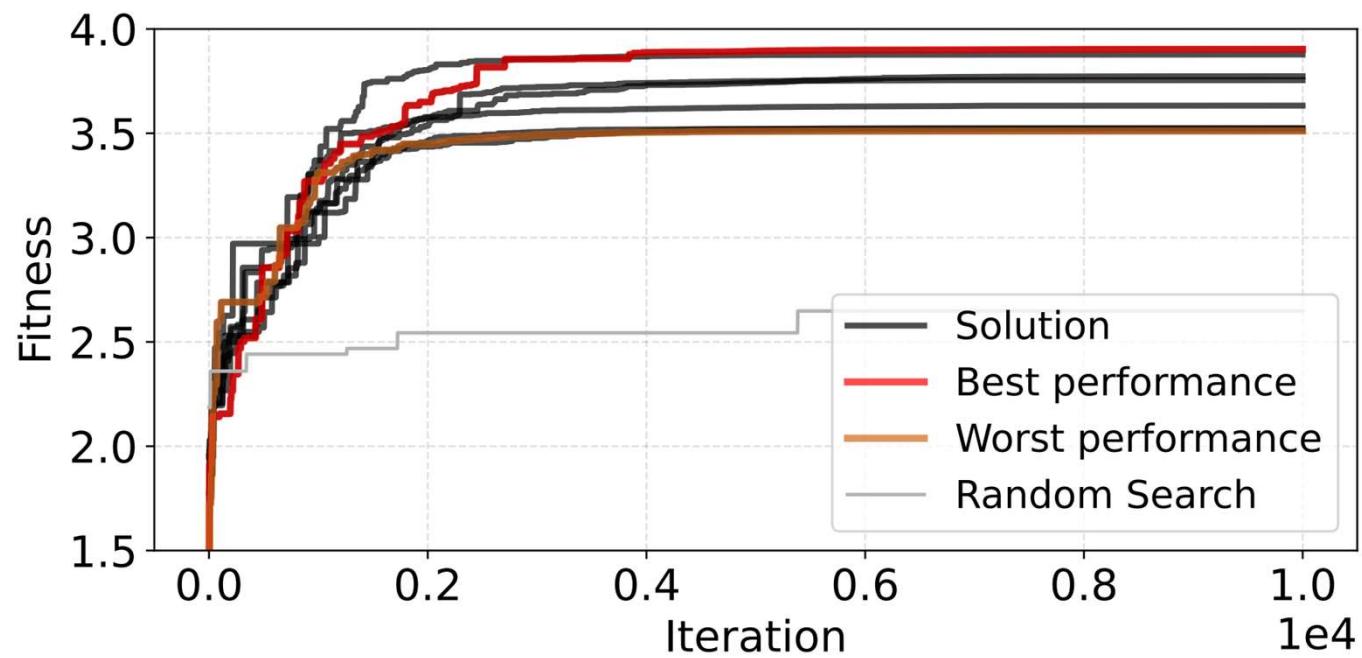
RANDOM SEARCH



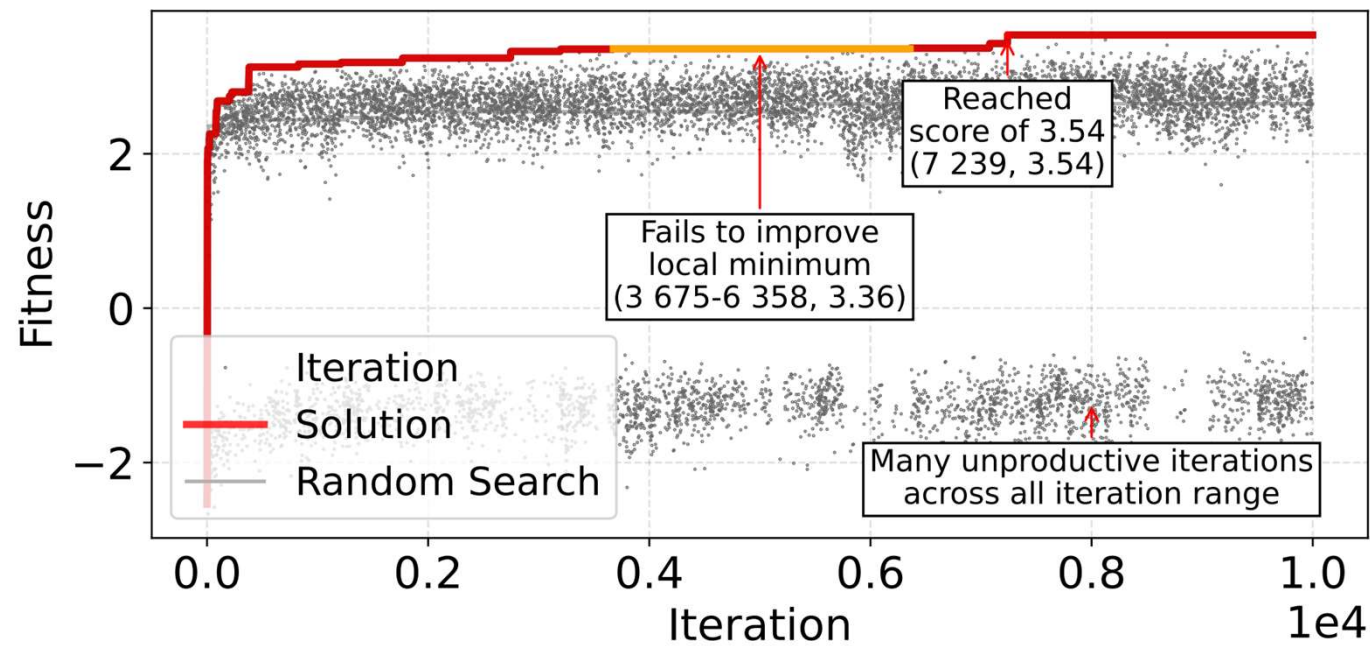
CMA-ES



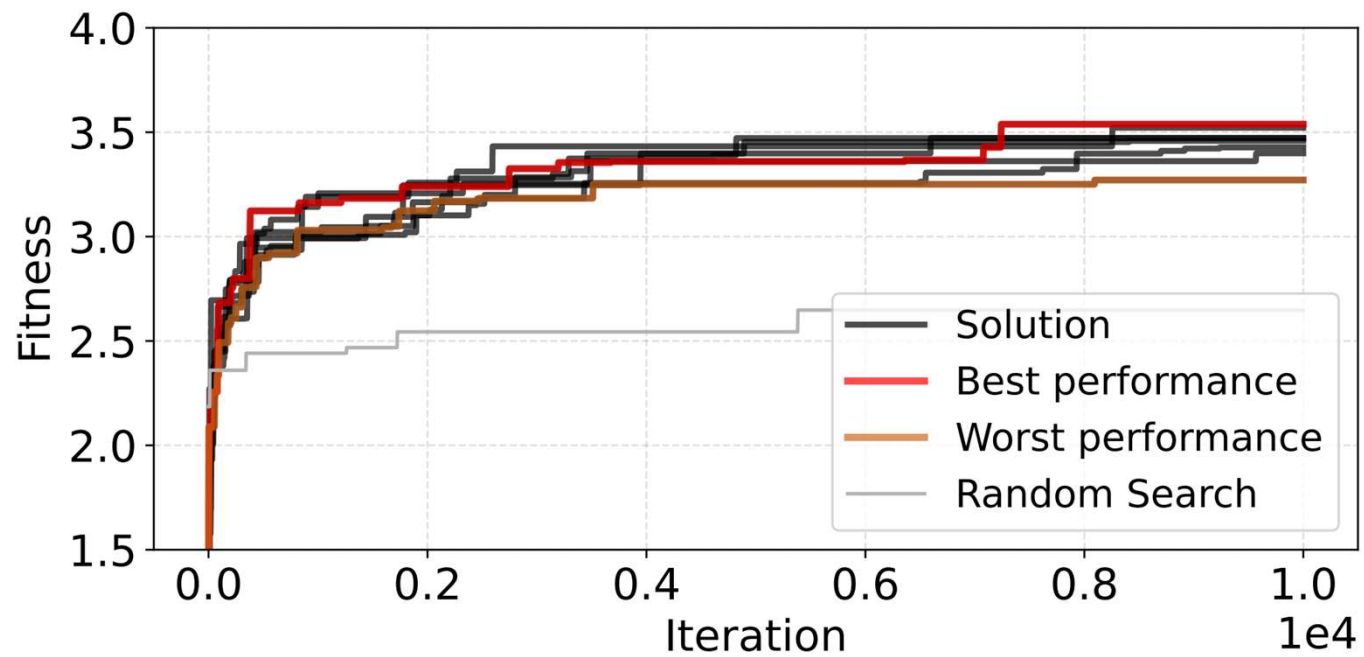
CMA-ES



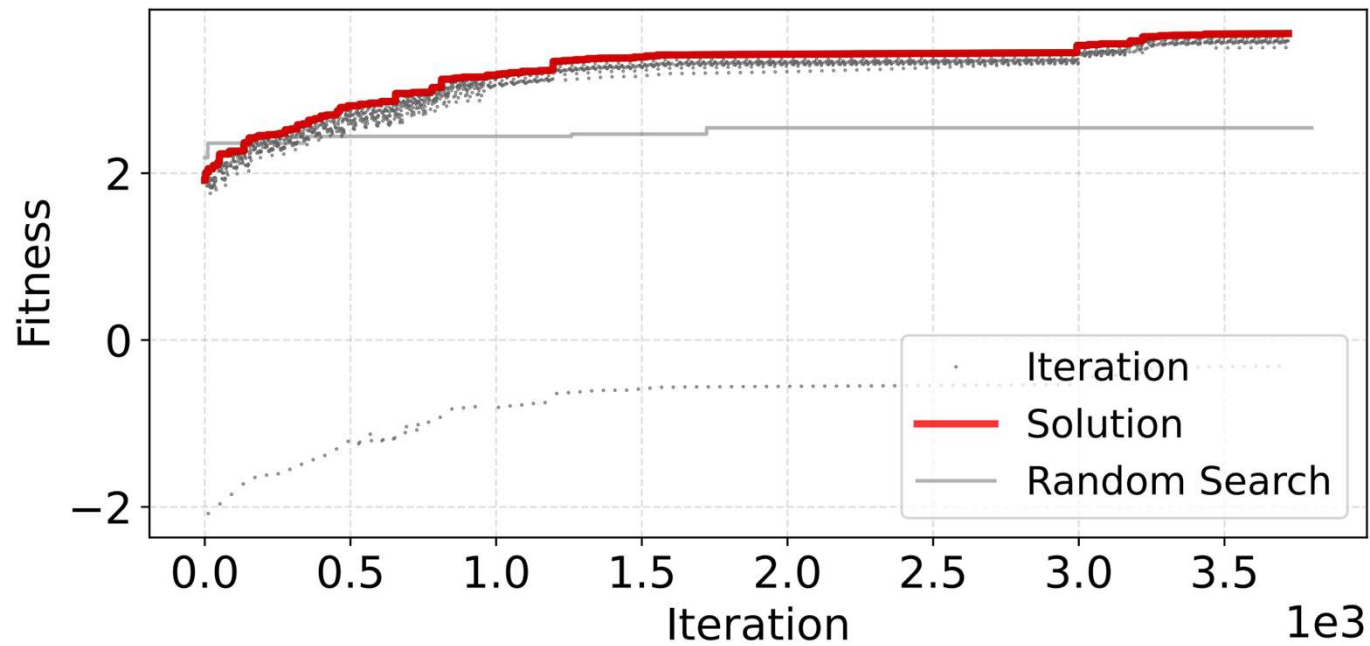
TPE



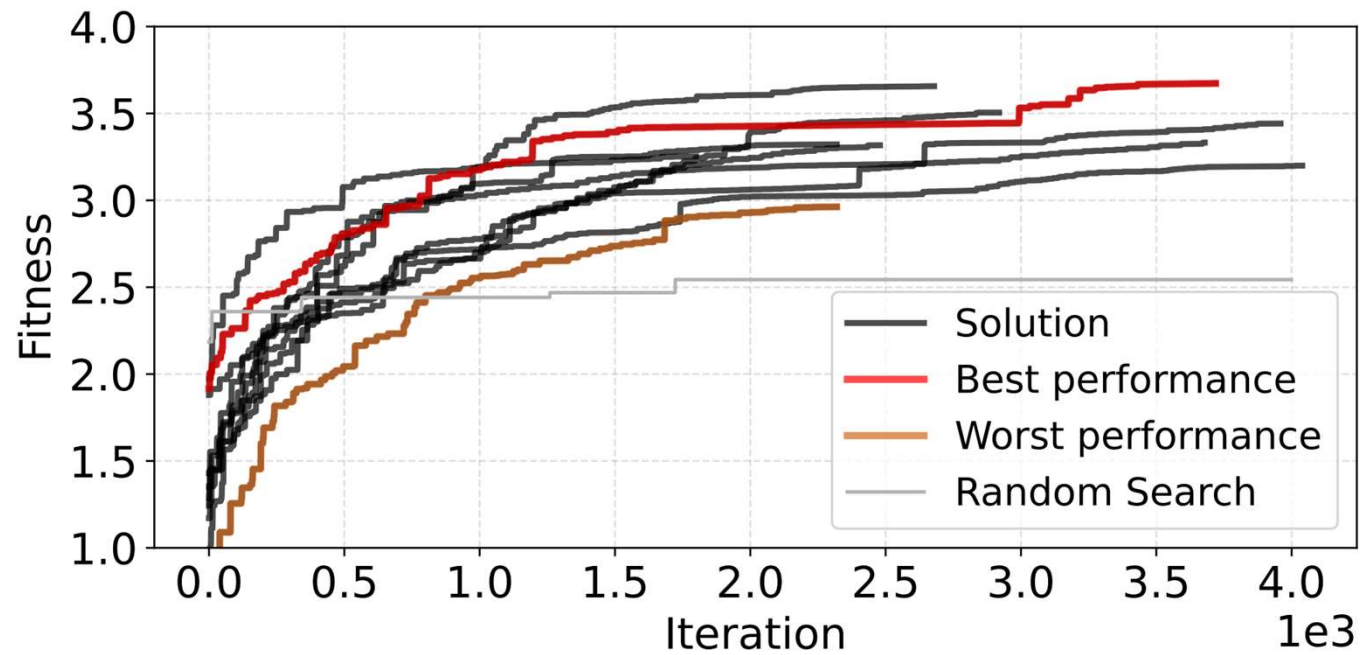
TPE



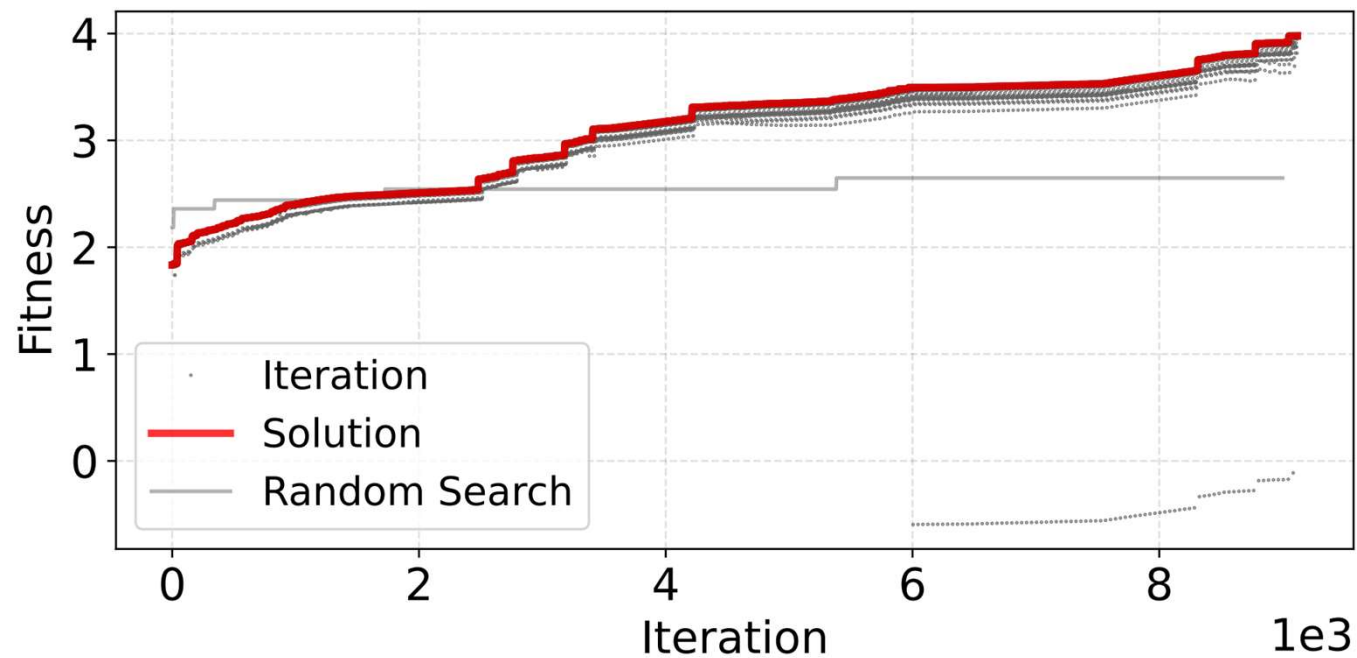
HILL CLIMBING



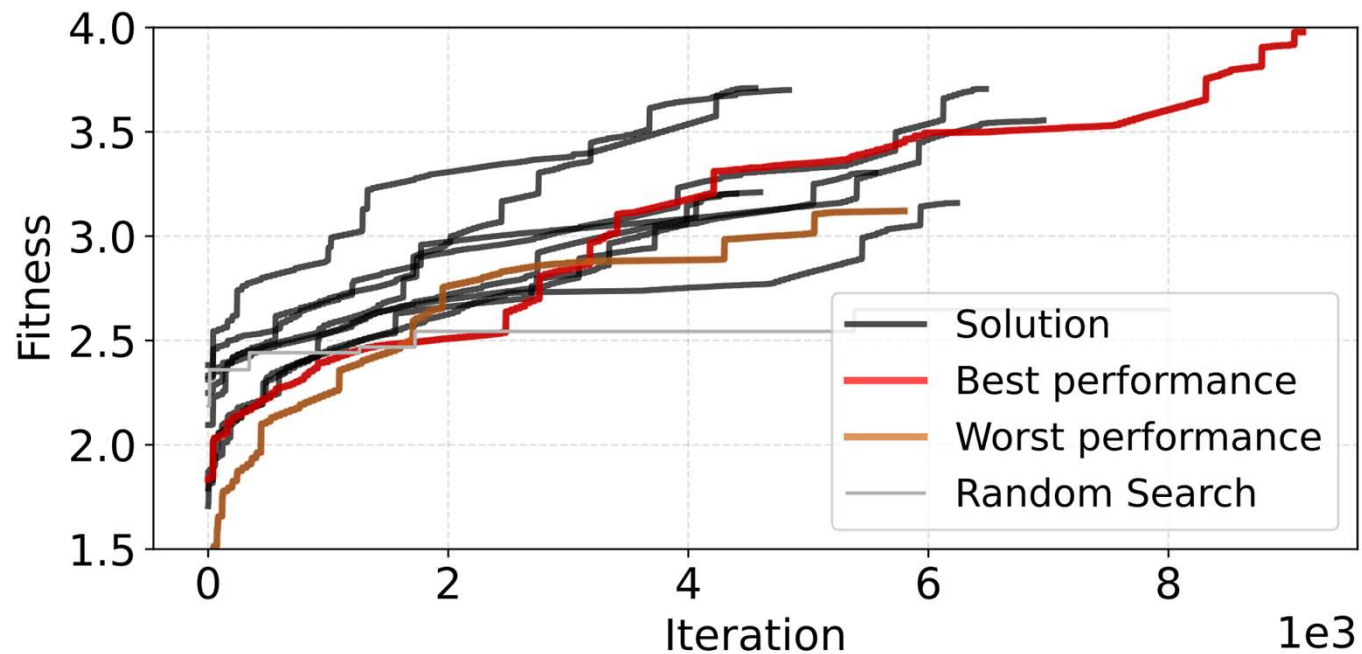
HILL CLIMBING



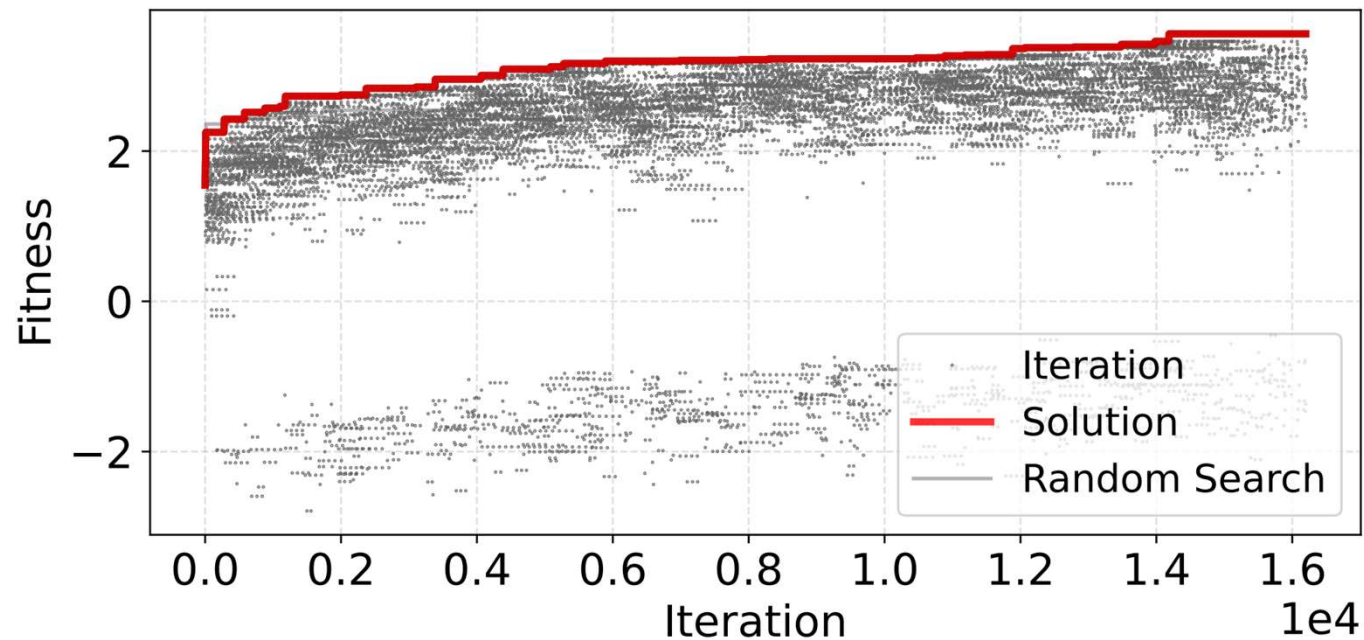
SIMULATED ANNEALING



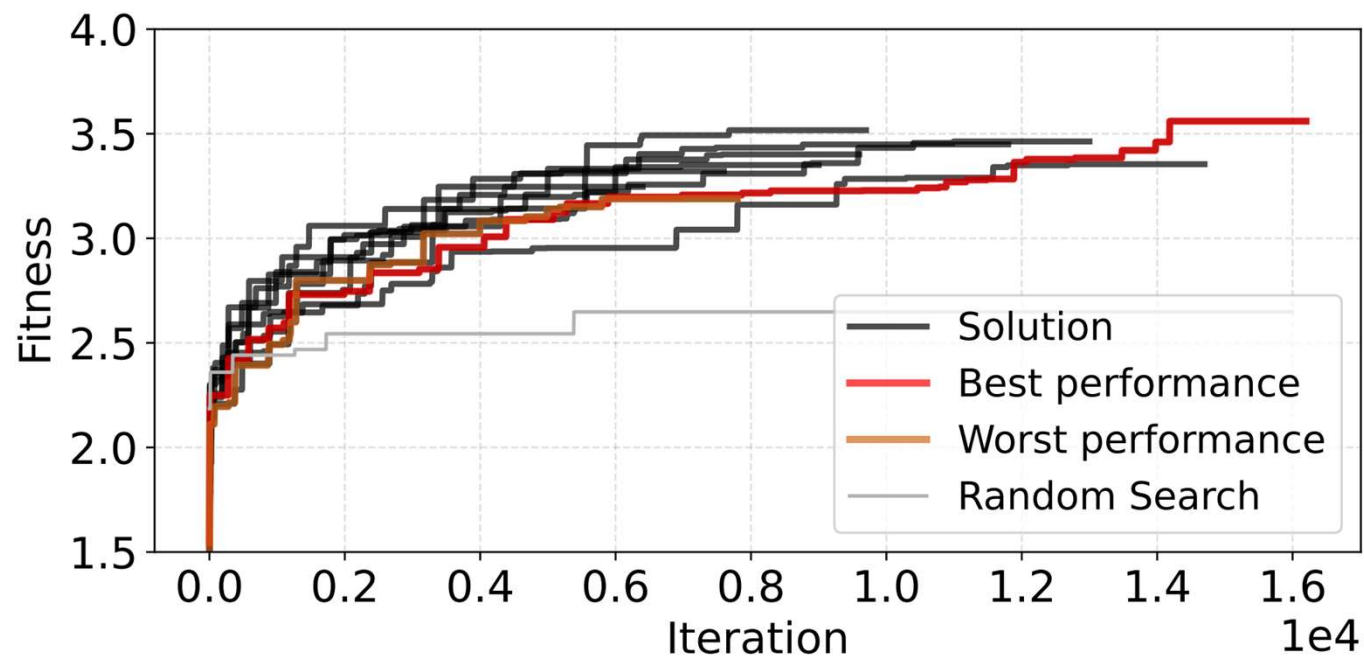
SIMULATED ANNEALING



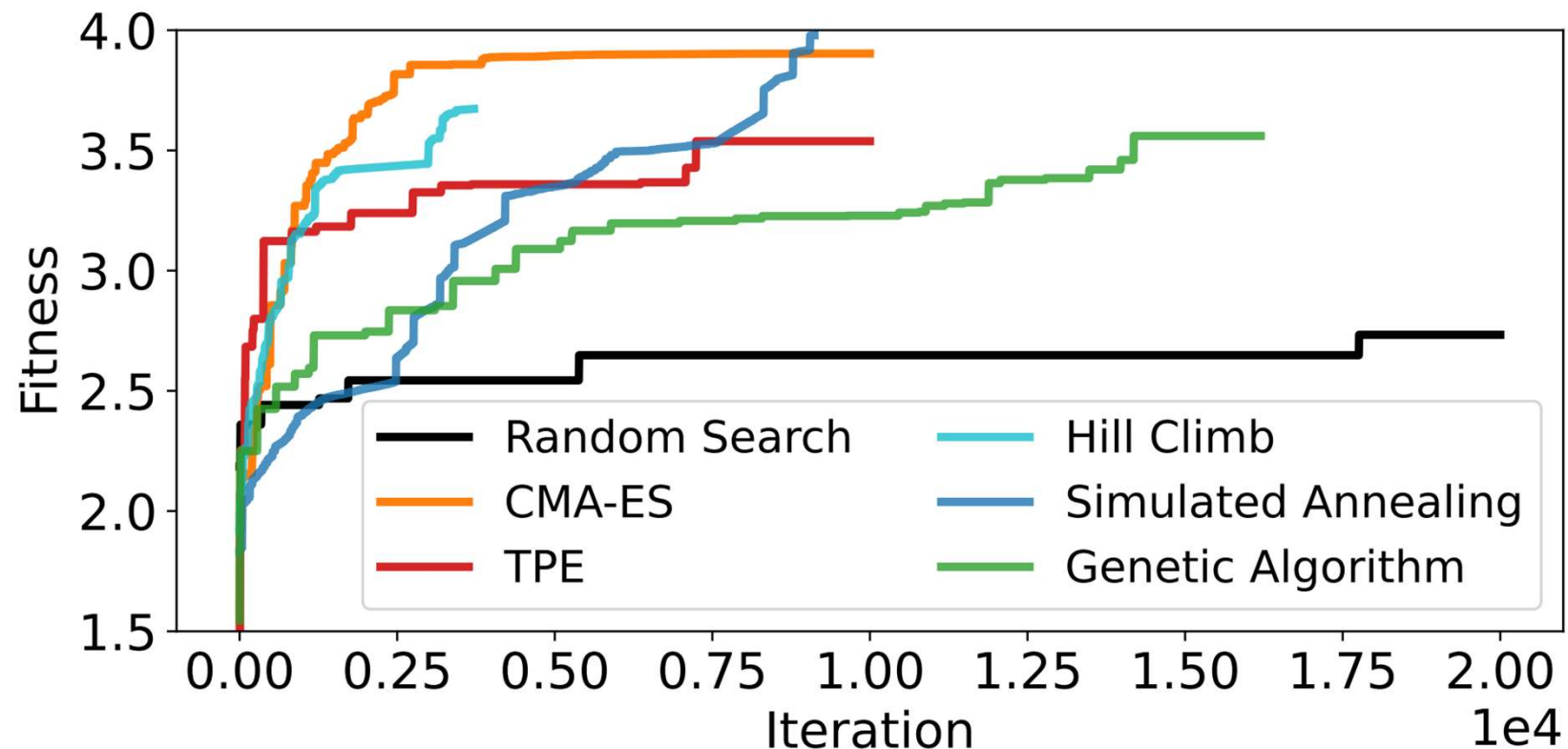
GENETIC ALGORITHM



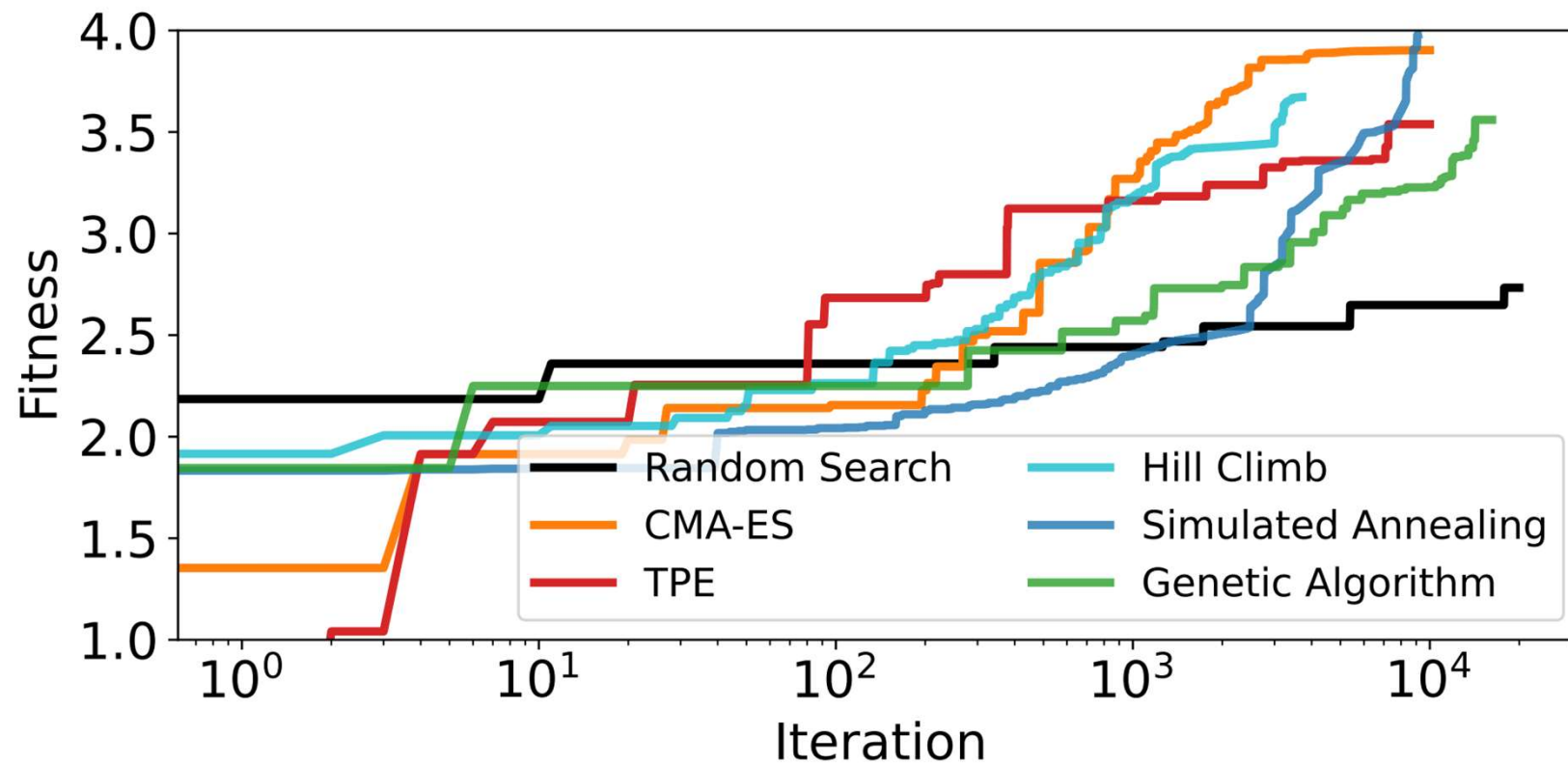
GENETIC ALGORITHM



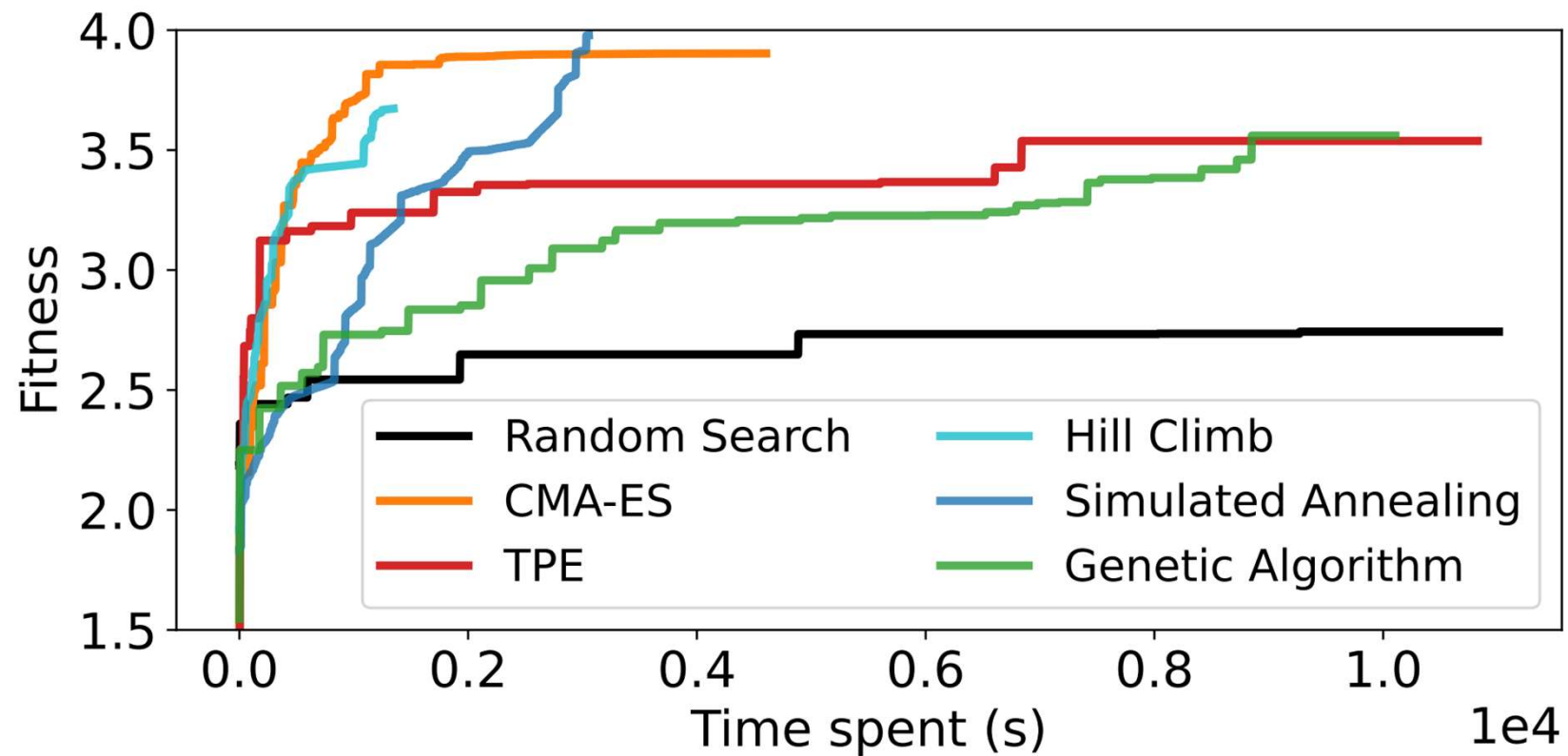
COMPARING ALL ALGORITHMS



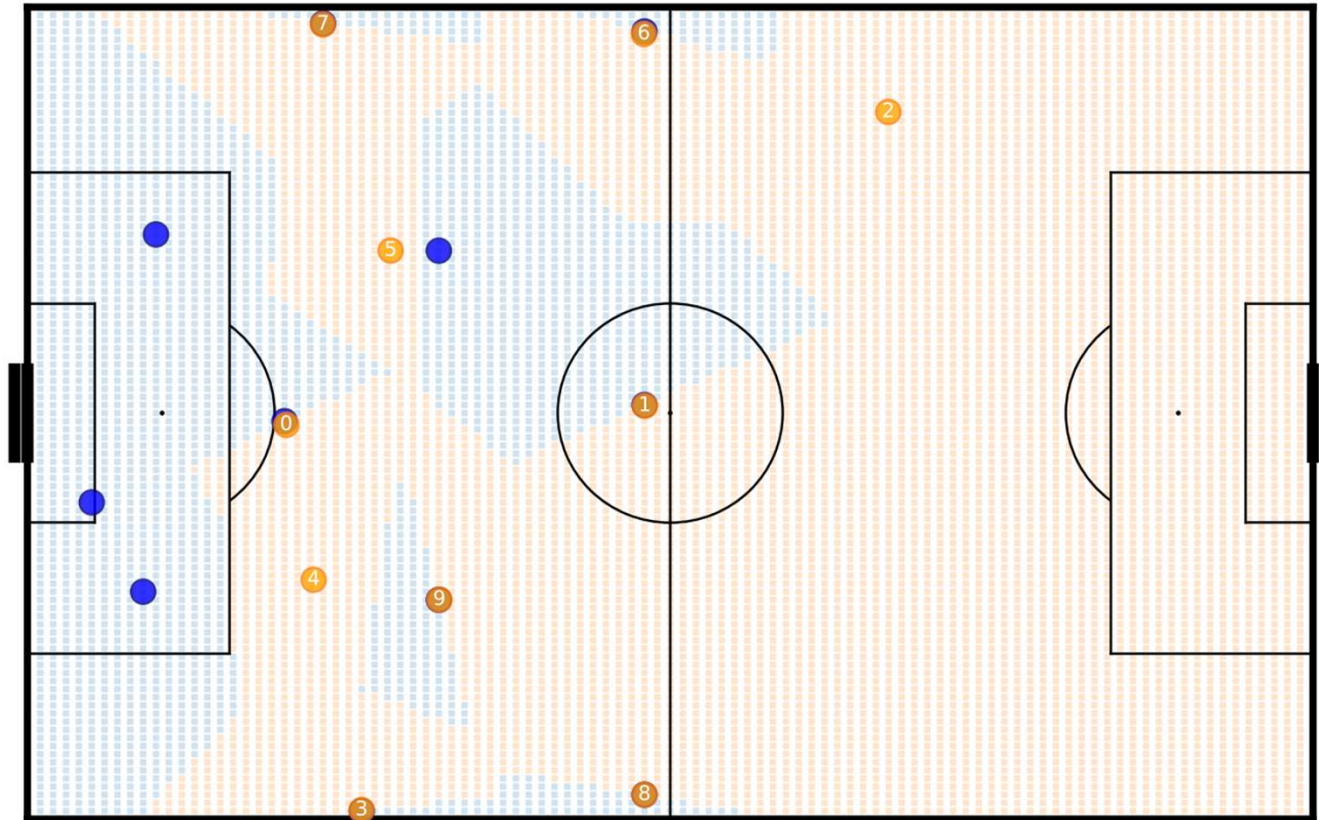
COMPARING ALL ALGORITHMS



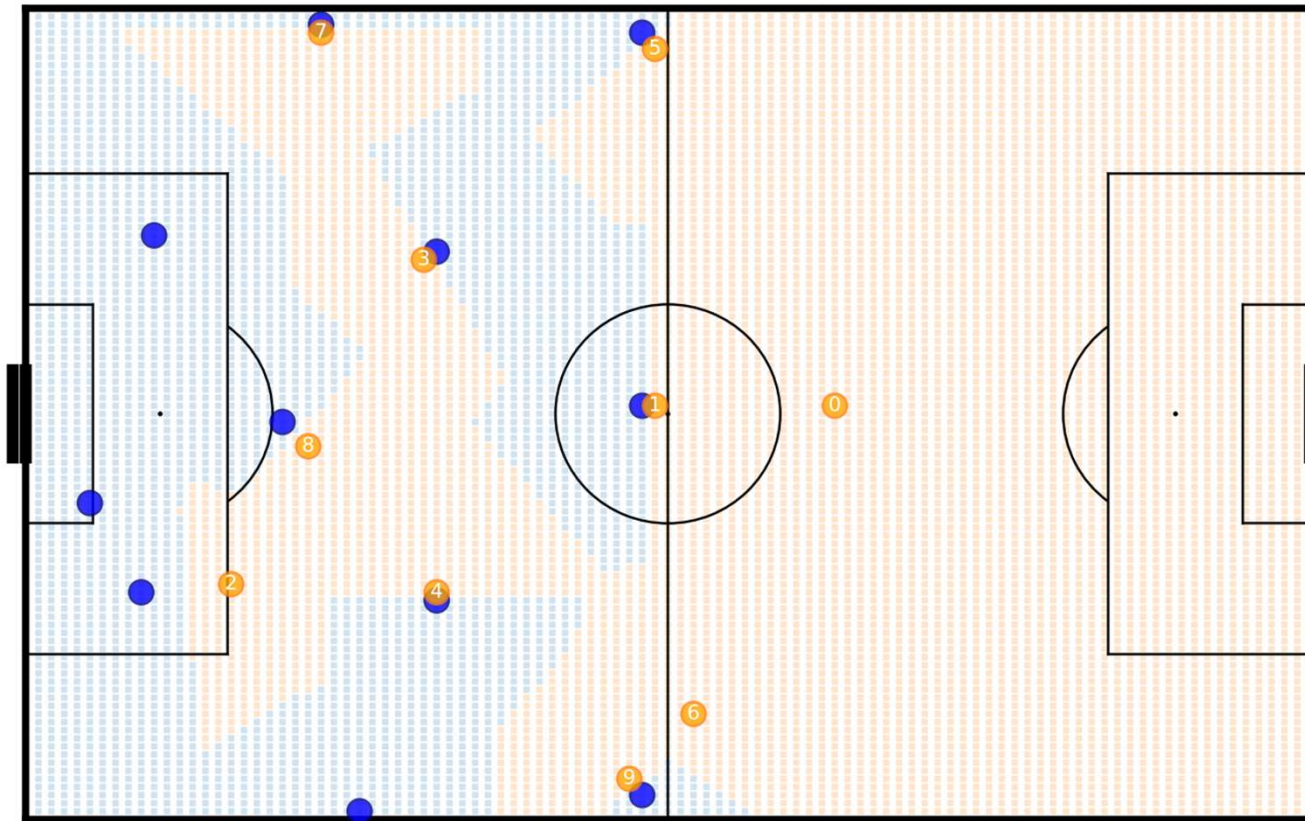
COMPARING ALL ALGORITHMS



SOLUTION CMA-ES



SOLUTION SIMULATED ANNEALING



Conclusions

- CMA-ES & Simulated annealing are the best approaches for our problem.
 - CMA-ES wins in consistency and under restrict time constraints.
 - Simulated annealing wins otherwise (except against Grid Search if time available $\rightarrow \infty$).
- Local methods make a lot of sense:
 - Teams usually have predefined formations which will serve as a good starting point.
- Fitness function is the most important factor \rightarrow knowing what to optimize is the hardest thing to solve.

Future Work

- Test more scenarios to check if the algorithms generalize properly.
- Test on real-time scenarios.

Q&A

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github.com/nvsclub/MarkingWithAI