

Hypervolume Gradient Ascent for Memetic Building Spatial Design Optimisation

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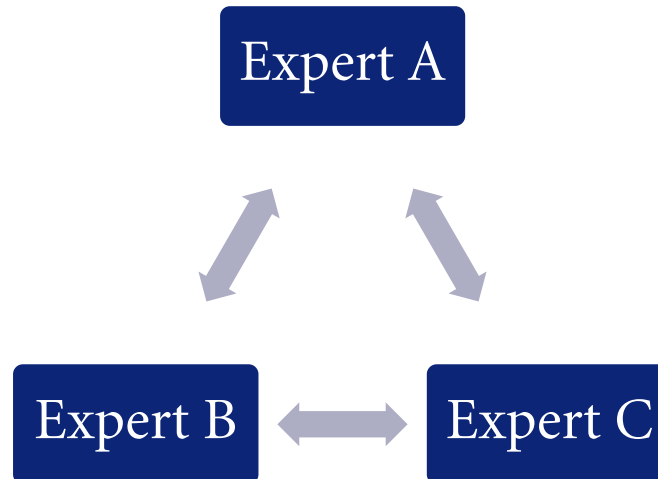
**Universiteit
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Traditional building design

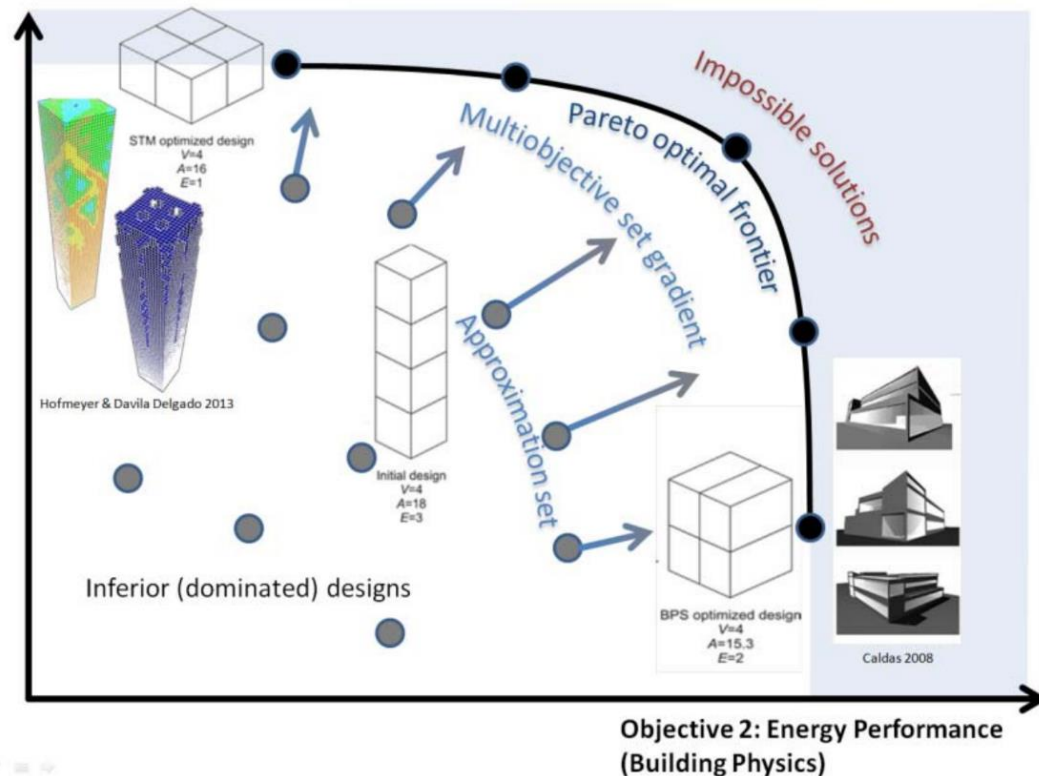
- Many disciplines with different experts
 - E.g. Structural, plumbing, HVAC, etc.
- Issues
 - Sequential
 - Limited communication
- Solution: Automation



Problem description

- Optimise building spatial design (i.e. the shape)
 - Structural performance (compliance)
 - Thermal performance (heating/cooling energy)

Objective 1: Optimal Strain Energy (Structural Design)



Work so far

- Problem representation and constraint functions [1,2]
- Tested with standard algorithms [2,3]
- Constraint satisfaction penalty functions [2]
- Constraint satisfaction by specialised initialisation and mutation operators [3]
- Improved operators and parameter tuning [4]
- Cooperative superstructure and free representation [5]

[1] S. Boonstra, K. van der Blom, H. Hofmeyer, R. Amor, and M. T. M. Emmerich, “Super-structure and super-structure free design search space representations for a building spatial design in multi-disciplinary building optimisation,” in *Electronic proceedings of the 23rd International Workshop of the European Group for Intelligent Computing in Engineering*. Jagiellonian University ZPGK, 2016, pp. 1–10.

[2] K. van der Blom, S. Boonstra, H. Hofmeyer, and Emmerich M. T. M., A super-structure based optimisation approach for building spatial designs. in *Proceedings of the VII European Congress on Computational Methods in Applied Sciences and Engineering*, Papadrakakis M., Papadopoulos V., Stefanou G., Plevris V., Eds.. National Technical University of Athens, 2016, pp. 3409–3422.

[3] K. van der Blom, S. Boonstra, H. Hofmeyer, and M. T. M. Emmerich, “Multicriteria building spatial design with mixed integer evolutionary algorithms,” in *Parallel Problem Solving from Nature – PPSN XIV*, ser. *Lecture Notes in Computer Science*, J. Handl, E. Hart, P. R. Lewis, M. López-Ibáñez, G. Ochoa, and B. Paechter, Eds., vol. 9921. Cham: Springer International Publishing, 2016, pp. 453–462.

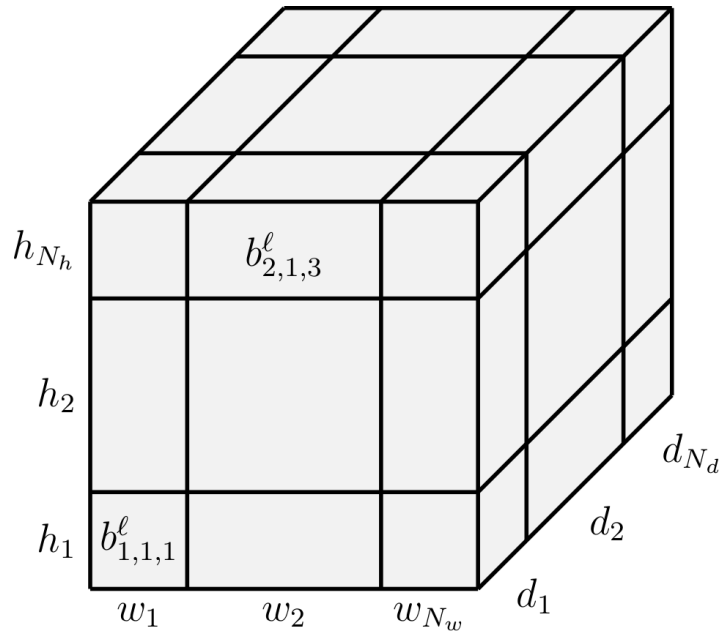
[4] K. van der Blom, S. Boonstra, H. Hofmeyer, T. Bäck, M. T. M. Emmerich, “Configuring advanced evolutionary algorithms for multicriteria building spatial design optimisation,” in *2017 Congress on Evolutionary Computation (CEC)*. IEEE, 2017, pp. 1803–1810

[5] S. Boonstra, K. van der Blom, H. Hofmeyer, M. T. M. Emmerich, “Combined super-structured and super-structure free optimisation of building spatial designs,” in *24rd International Workshop of the European Group for Intelligent Computing in Engineering*, C. Koch, W. Tizani, J. Ninic, Eds.. University of Nottingham, 2017, pp. 23–34

Contributions

- Hypervolume gradient ascent in the real world
- Challenges:
 - Numerical gradients only
 - Box constraints
 - Repair function
- Memetic algorithm
 - Improve local search
 - Relay or alternate?

Problem representation



$$i \in \{1, 2, \dots, N_w\} \quad w_i \in \mathbb{R} \geq 0$$

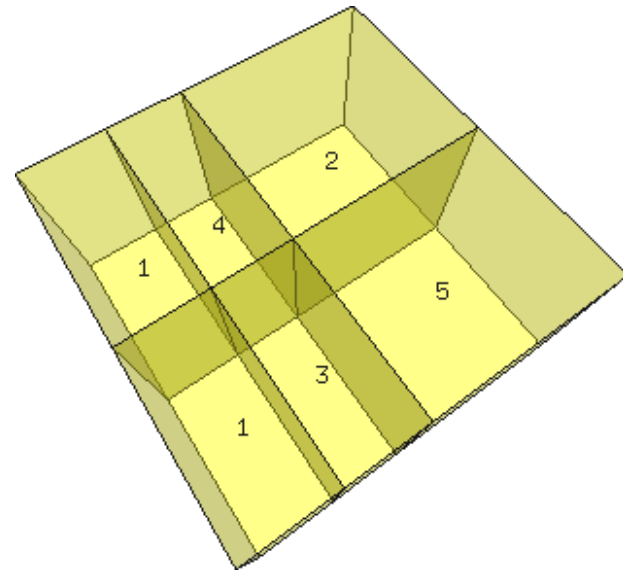
$$j \in \{1, 2, \dots, N_d\} \quad d_j \in \mathbb{R} \geq 0$$

$$k \in \{1, 2, \dots, N_h\} \quad h_k \in \mathbb{R} \geq 0$$

$$\ell \in \{1, 2, \dots, N_{rooms}\}$$

$$b_{i,j,k}^\ell = \begin{cases} 1 & \text{if cell } (i, j, k) \text{ belongs to room } \ell \\ 0 & \text{otherwise} \end{cases}$$

	A	B	C	D
1				
2				
3				
4				
5				
6				
7				



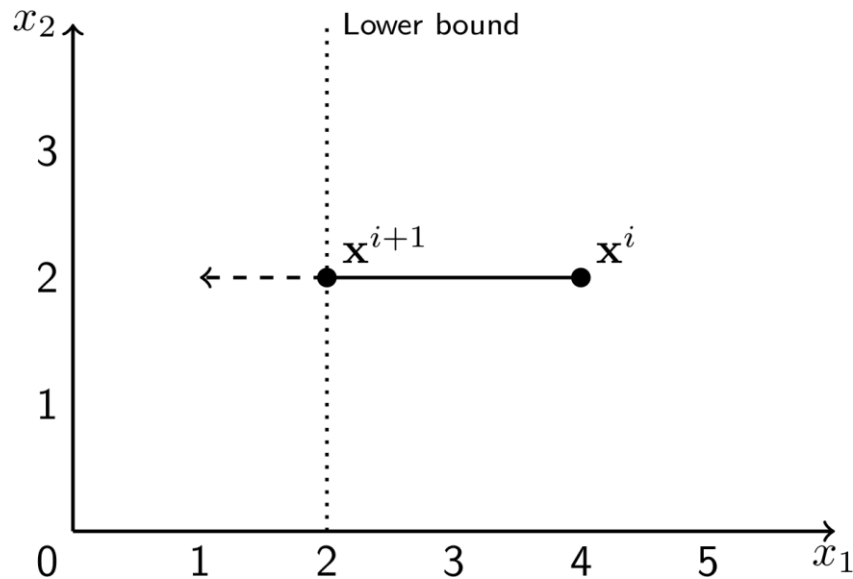
Constraints on binary variables

- Active – every room has at least one active cell
- No overlap – cell (x, y, z) is active for at most one room
- Cuboid shape – all cells active for a room together form a cuboid (3D rectangle)
- No floating cells – every cell has ground or another cell below it

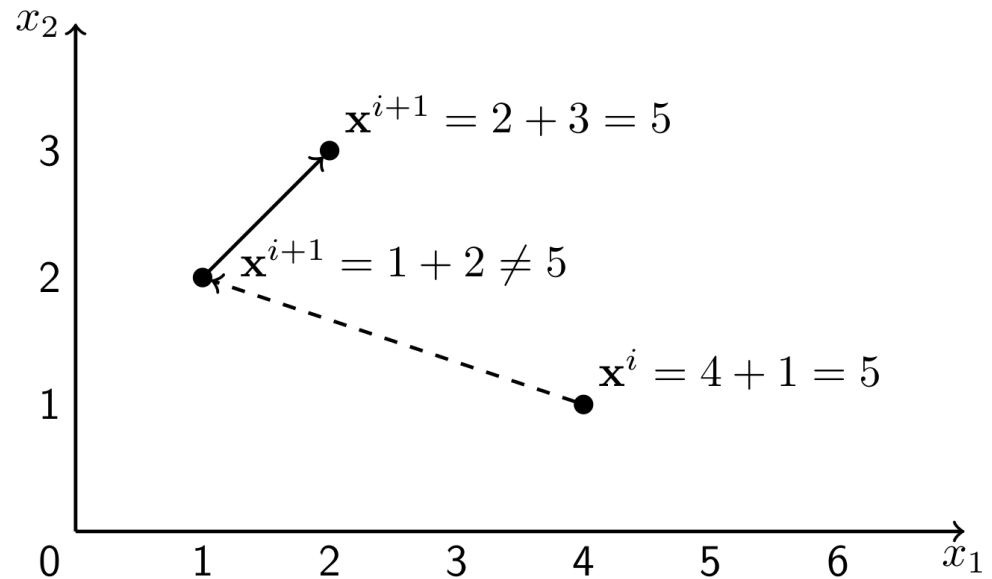
Problem specific challenges

- Gradients can only be found numerically
- Constraints on variables

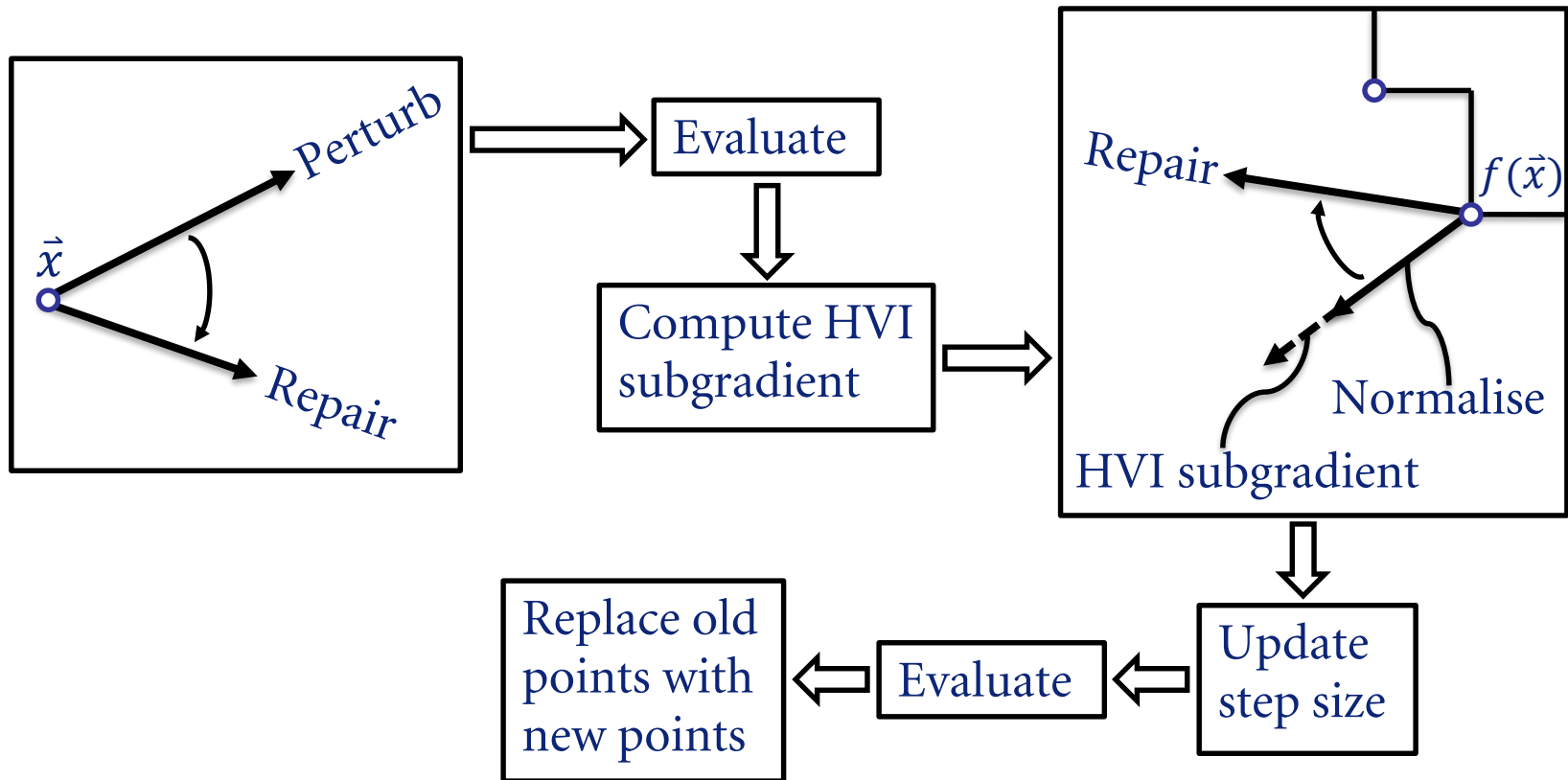
Box constraint



Rescale



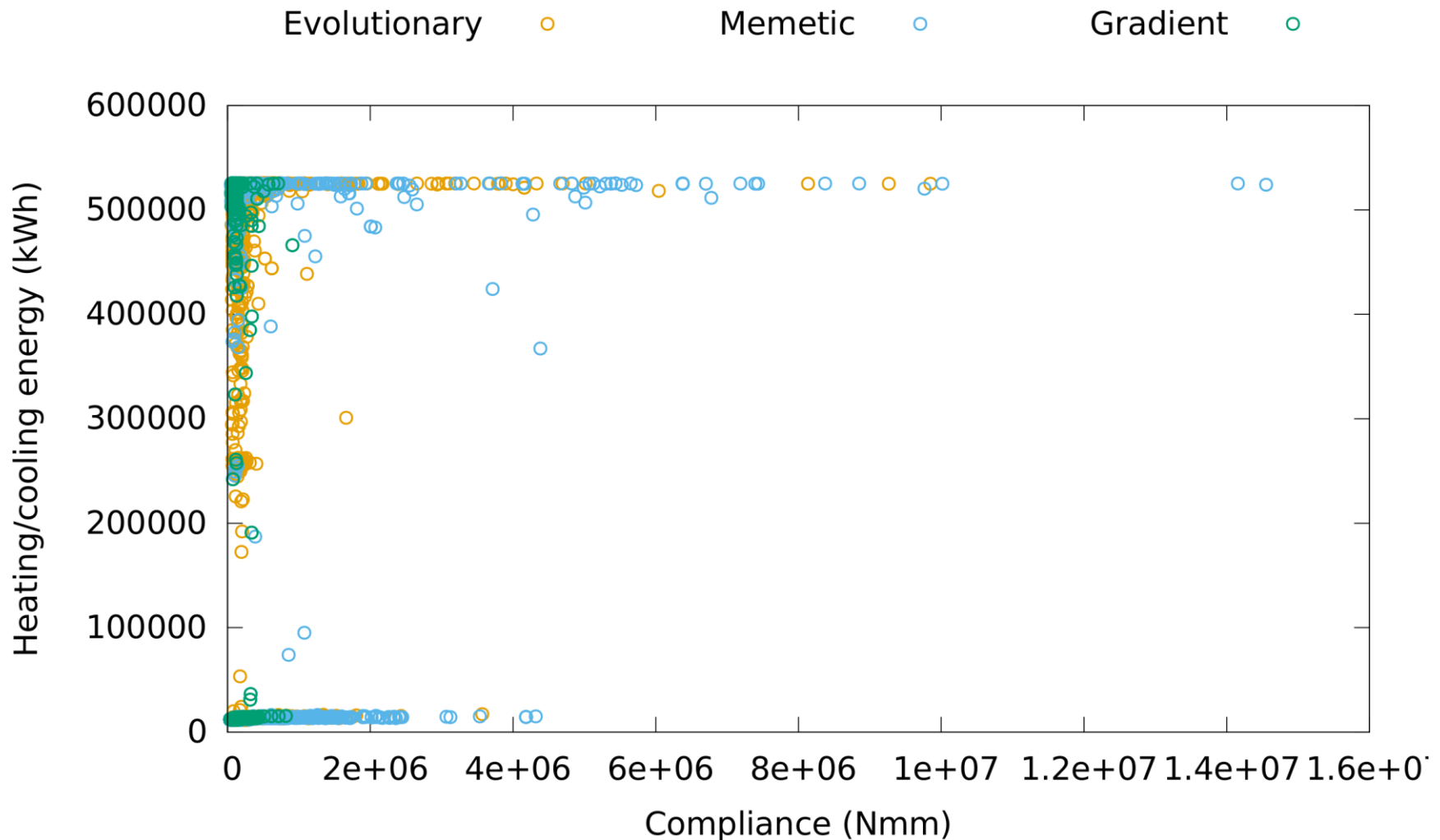
Numerical HIGA-MO



Experiment

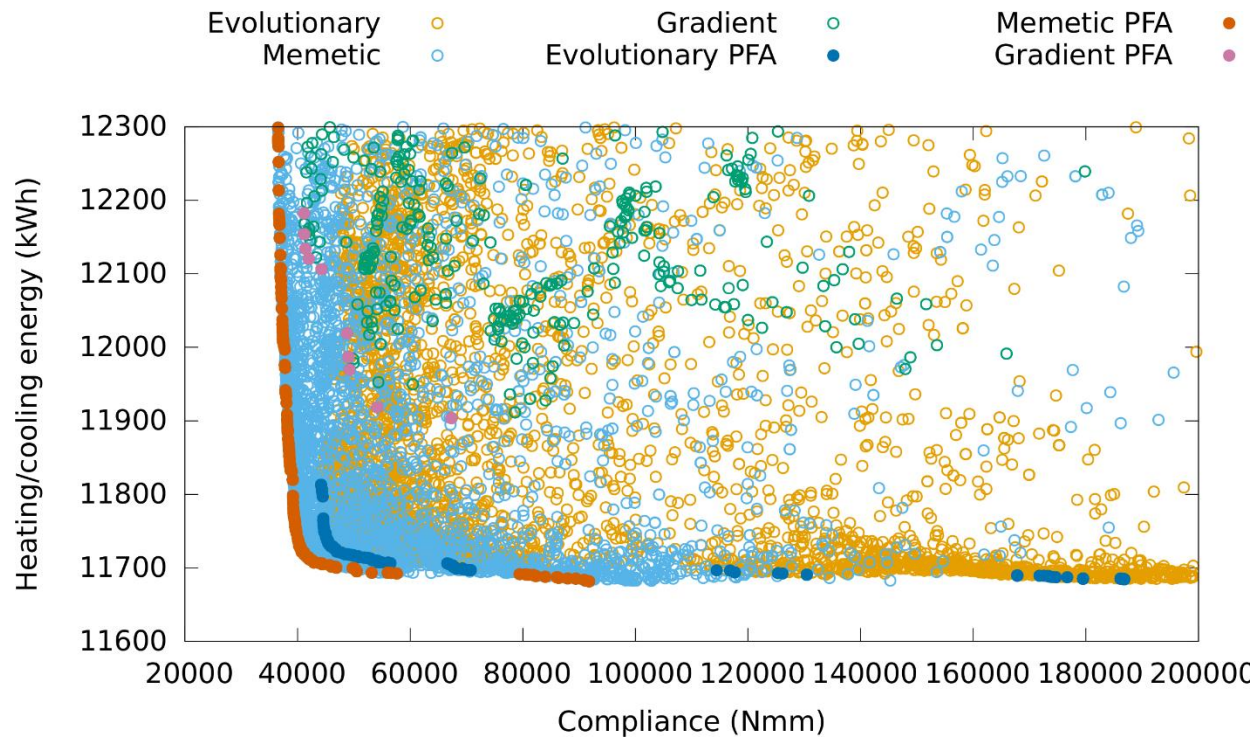
- Compare
 - Evolutionary (SMS-EMOA + problem specific operators)
 - Hypervolume indicator gradient ascent (HIGA)
 - Memetic: Evolutionary + HIGA in relay (switch half way)
- Setup
 - 2 objectives
 - 81 binary variables
 - 9 continuous variables
 - 10,000 evaluations

Results – overview objective space



Pareto Front

- Gradient only is not sufficient
- In this example: Memetic \succ Evolutionary before gradient search starts (in discrete space)

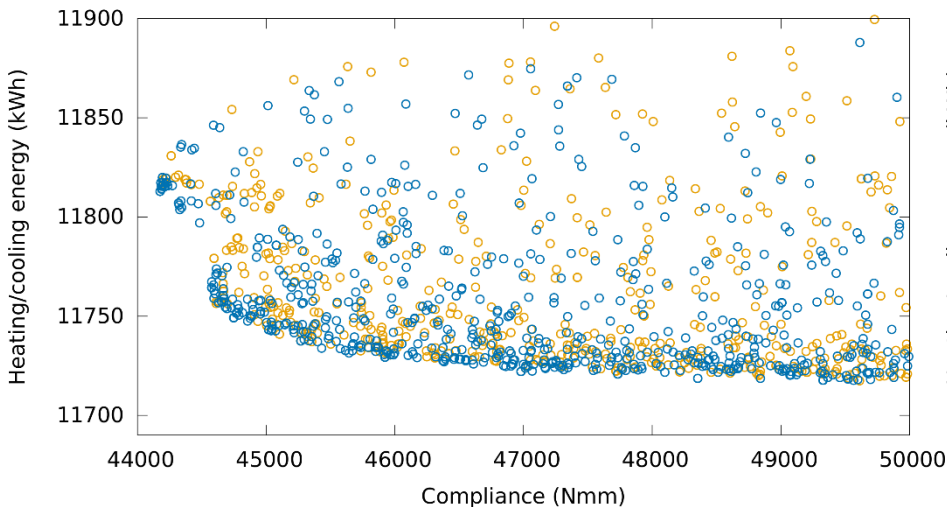


Local search effectiveness

- Chaotic search vs. search focused on the PF

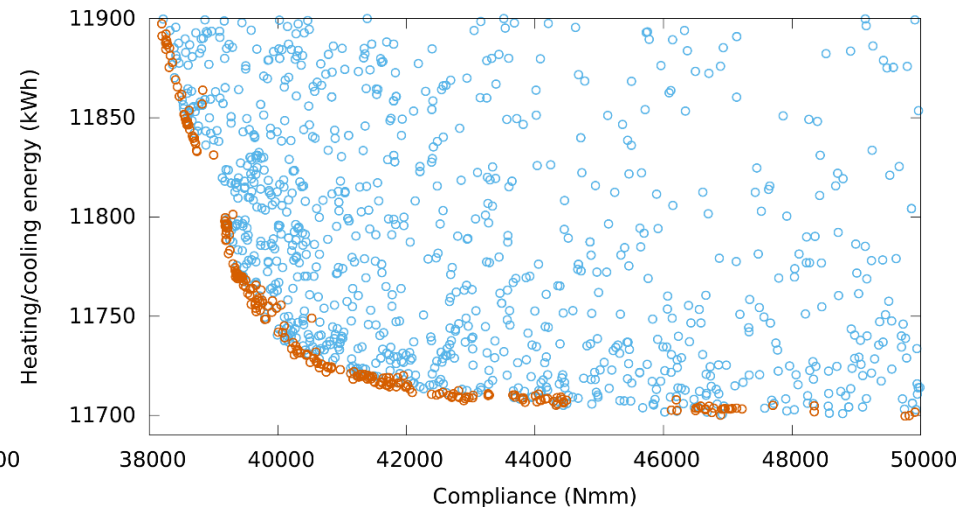
Evolutionary

Evaluations 1-5000 ○ Evaluations 5001-10000 ○



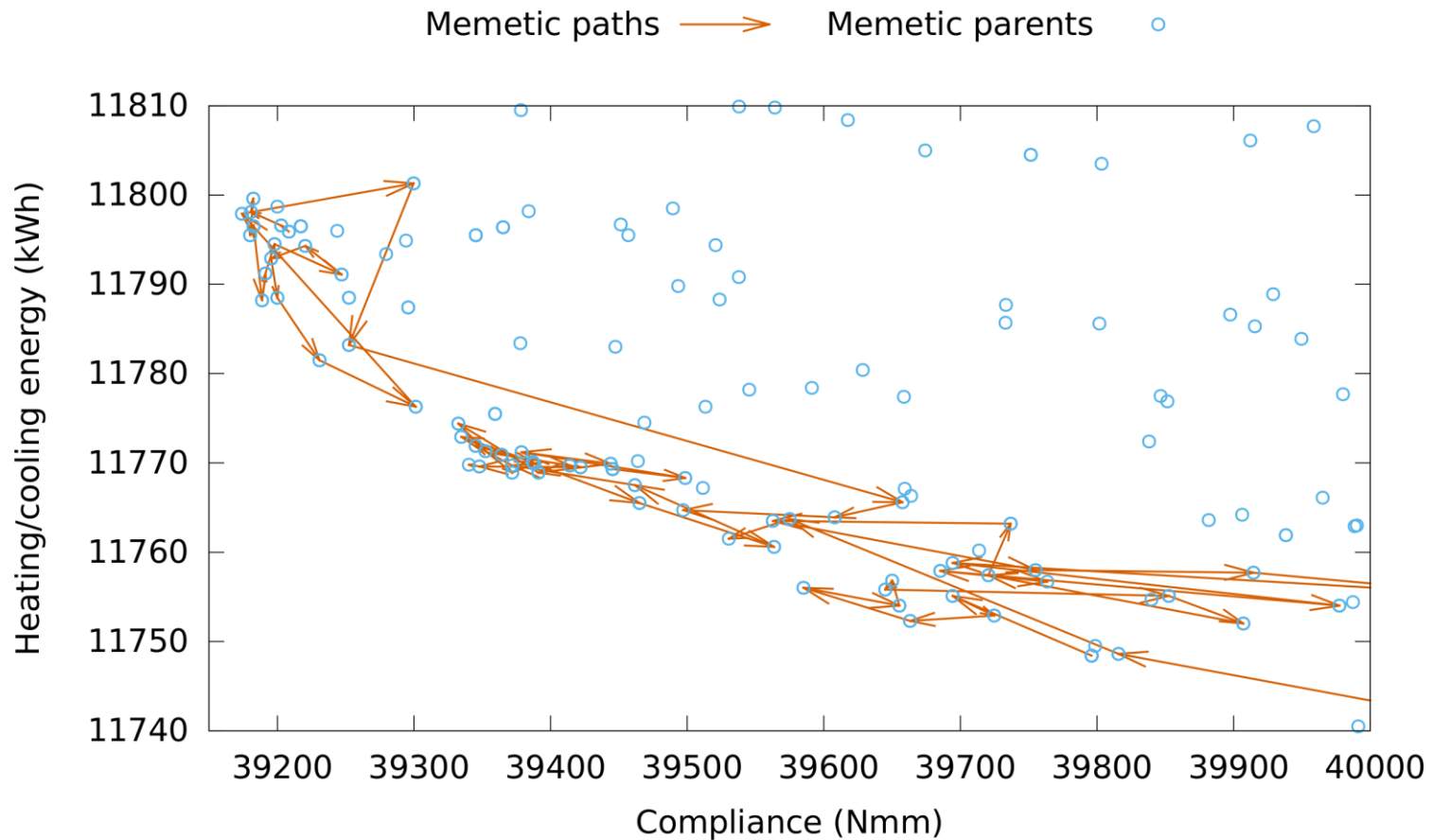
Memetic

Evaluations 1-5000 ○ Evaluations 5001-10000 ○



Local search zoomed in

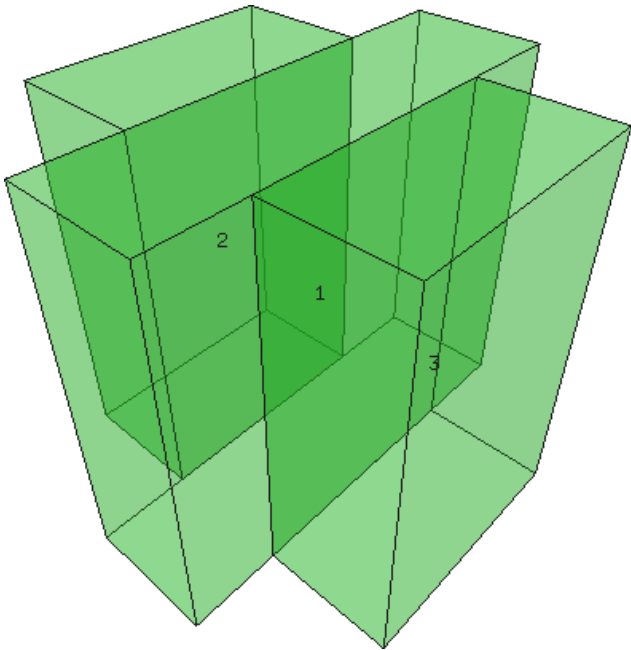
- Gradient search improves the PF



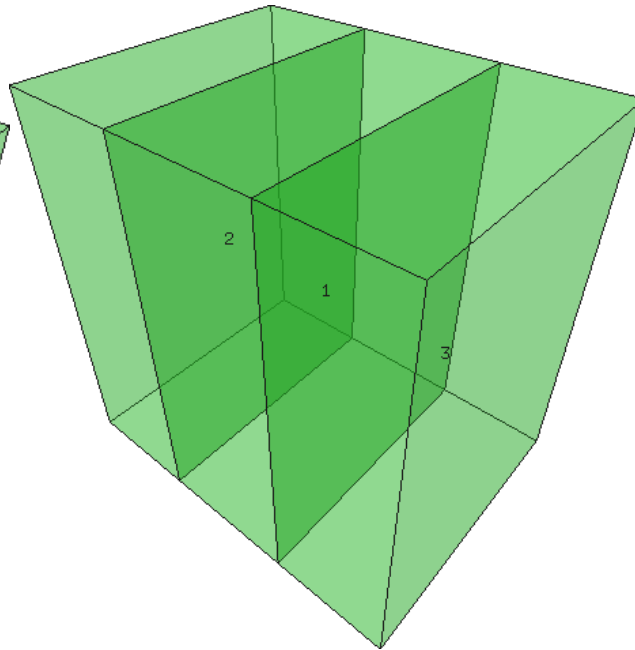
Visualisation

- Trade-off between objectives

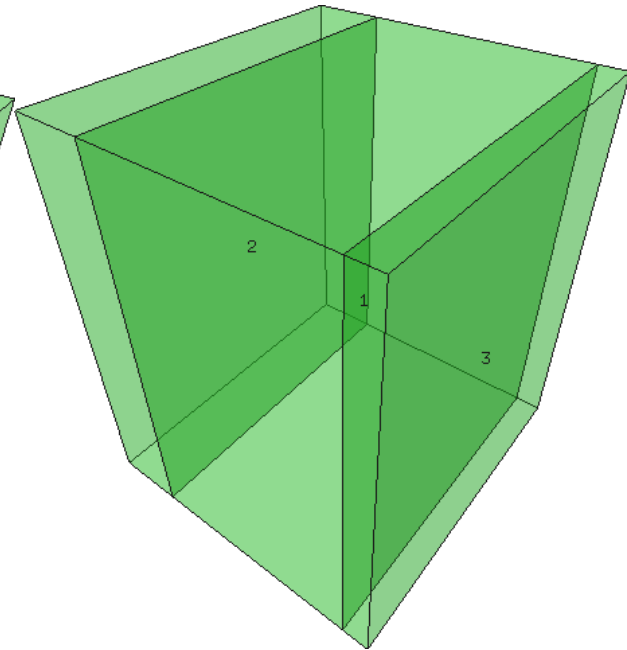
Structural



Kneepoint



Thermal



Conclusion

- Optimising buildings spatial design performance
 - Structural
 - Thermal
- Memetic approach combining
 - Evolutionary
 - Hypervolume gradient ascent
- HIGA-MO in the real world
 - Box constraints
 - Repair functions
 - Etc.

Future work

- More executions to improve confidence in current observations
- Investigate influence of settings, e.g. step size
- Detailed analysis of the influence of the different constraints on HIGA-MO and what can be done to alleviate any problems resulting from this
- Local search of binary space?
- Datamining on solutions

Questions?

Acknowledgements

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