

## **Day 2: Optimization methods for interface design and adaptation**

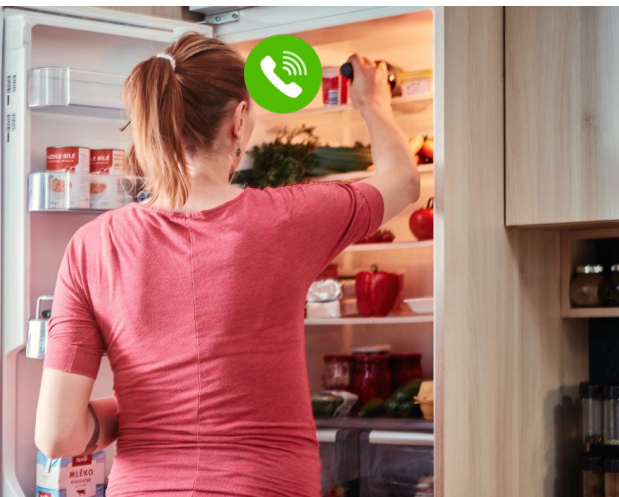
# **Motivation**

Where would you place a button to answer a call?



# Where would you place a button to answer a call?

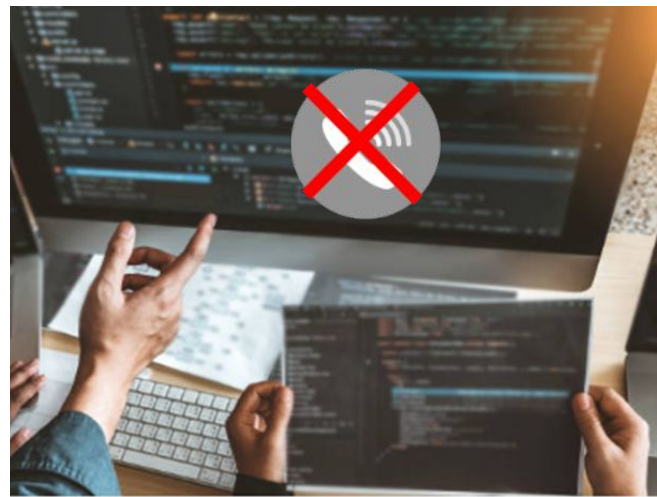
In sight, close to the hand



Avoid overlap with the environment



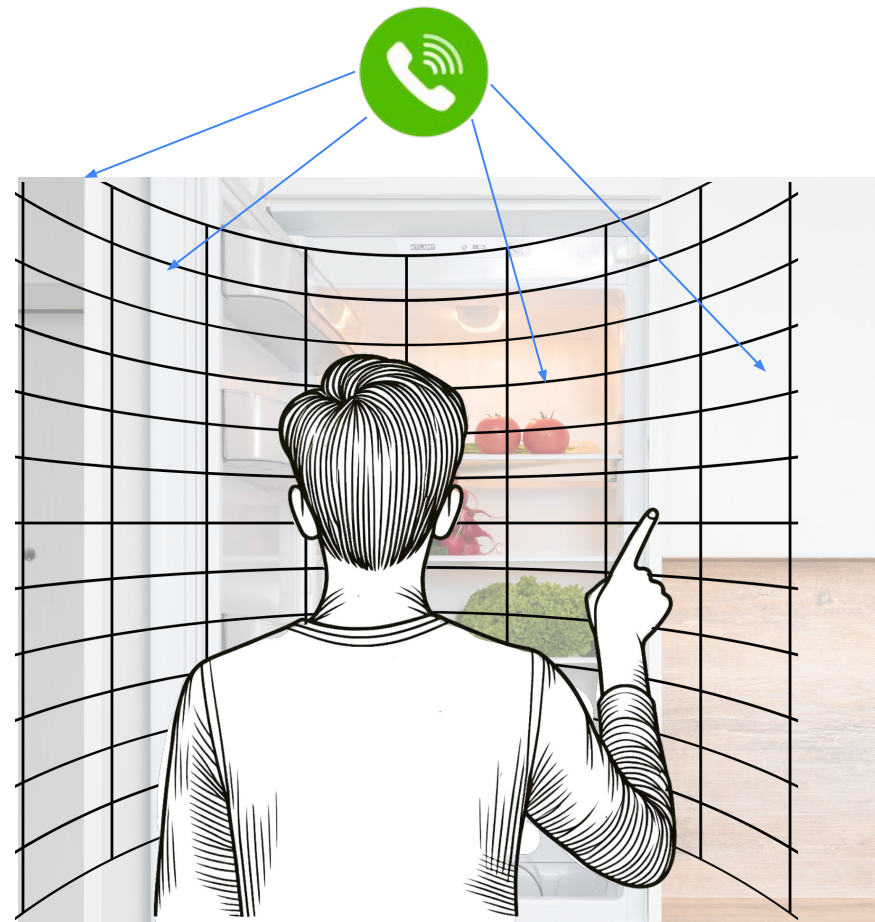
Do not interrupt user



# Let's do this more generally

- Given a 2D/3D grid
- A users' position
- The 3D geometry of the environment
- and a UI element

**Which criteria determine the goodness of the placement?**

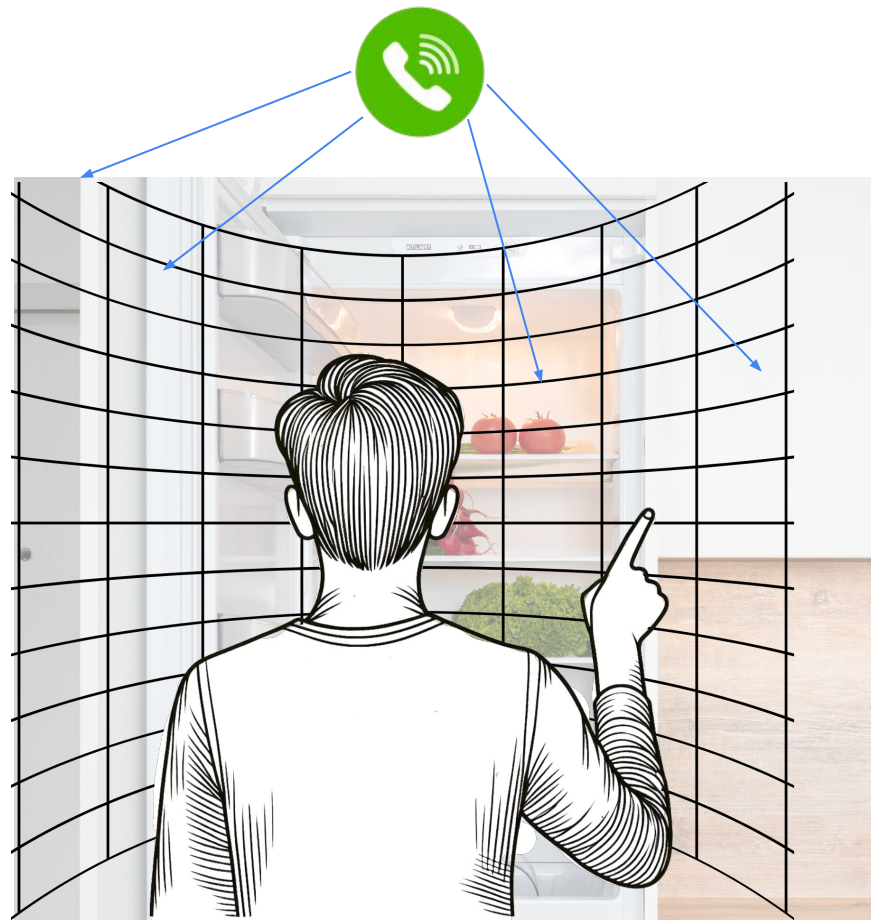


# Let's do this more generally

- Given a 2D/3D grid
- And a UI element

Which criteria determine the goodness of the placement?

- Reachability
- **Performance**
- **Field of View**
- Occlusion
- Ergonomics
- ...

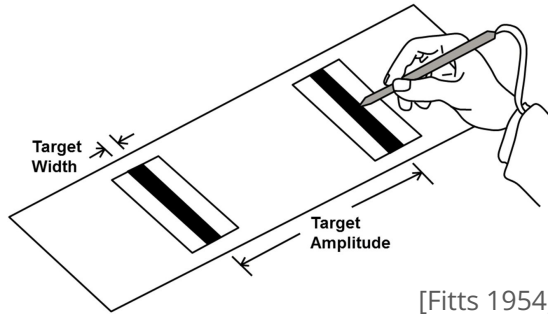




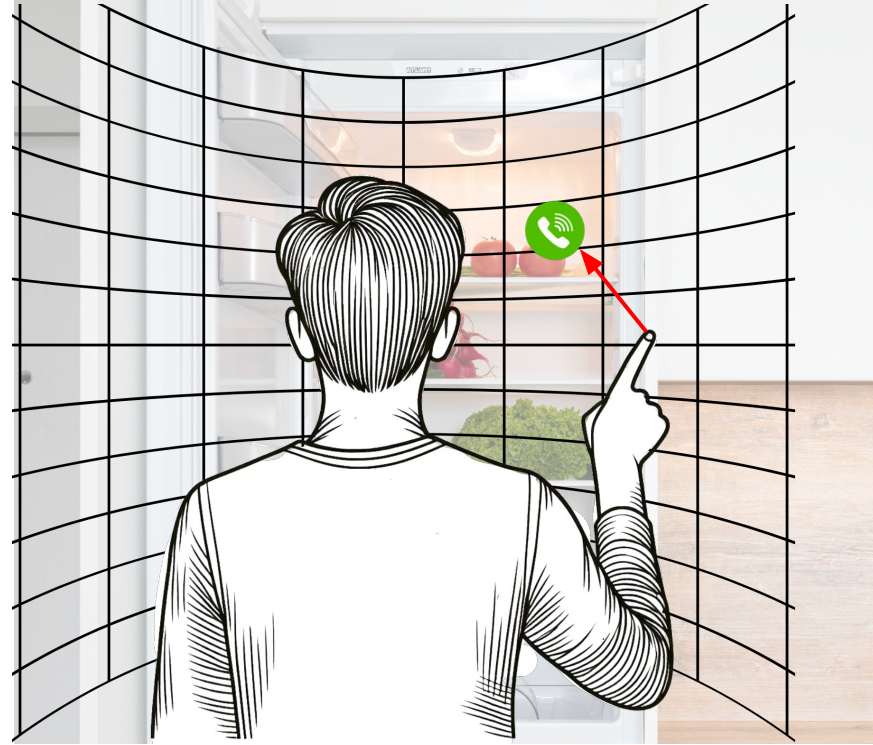
# How can we quantify these two criteria?

Performance

$$MT = a + b \cdot \log_2\left(\frac{A}{W} + 1\right)$$

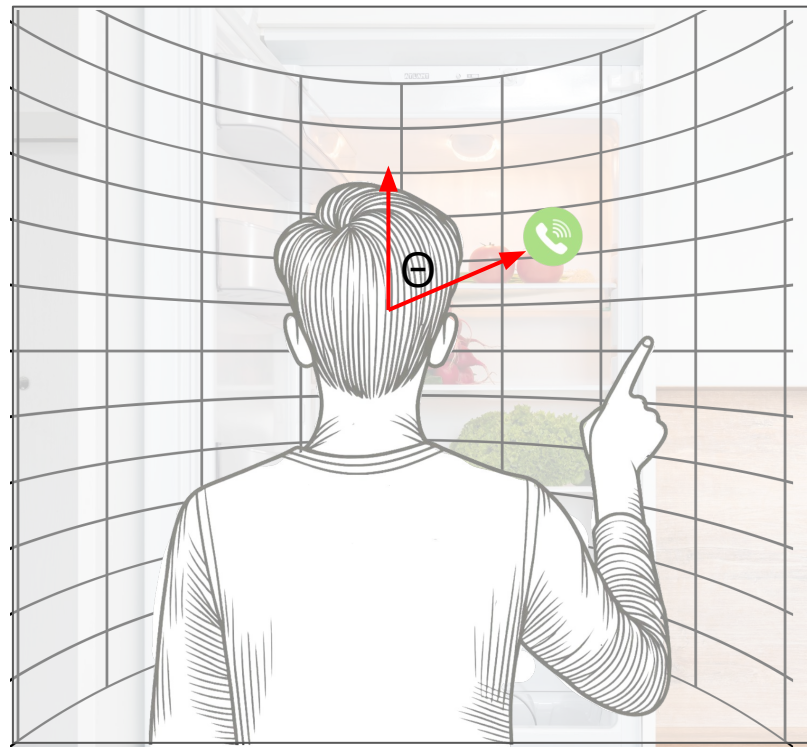
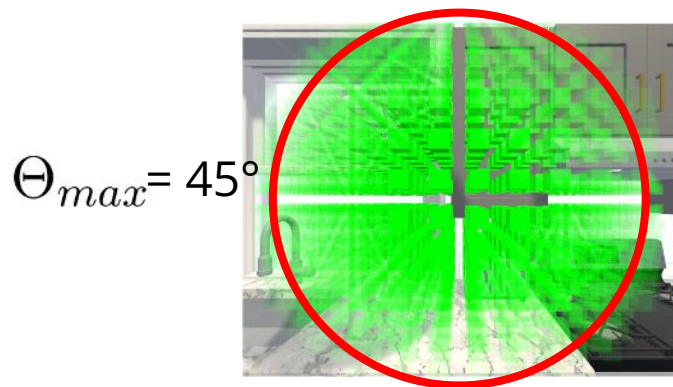


[Fitts 1954, MacKenzie 1992]



# How can we quantify these two criteria?

$$c_{FoV}(\Theta) = \frac{\Theta}{\Theta_{max}}$$



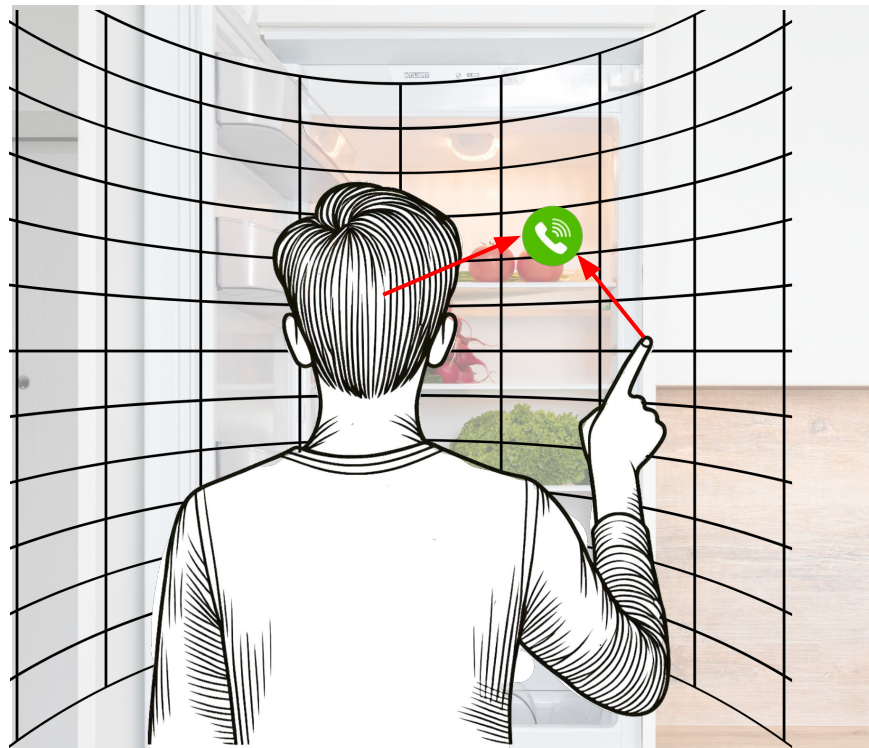
See also: Belo et al. AUIT – the Adaptive User Interfaces Toolkit for Designing XR Applications. UIST 2022. <https://dl.acm.org/doi/10.1145/3526113.3545651>

# How good is this placement?

$$MT = a + b \cdot \log_2\left(\frac{A}{W} + 1\right)$$

+

$$c_{FoV}(\Theta) = \begin{cases} 0 & \text{if } \Theta \leq \Theta_{in}, \\ \frac{\Theta - \Theta_{in}}{\Theta_{out} - \Theta_{in}} & \text{if } \Theta_{in} < \Theta \leq \Theta_{out}, \\ 1 & \text{if } \Theta > \Theta_{out}. \end{cases}$$





# How good is this placement?

$$MT = a + b \cdot \log_2\left(\frac{A}{W} + 1\right)$$

We can use **models** to evaluate a user interface

$$c_{FoV}(\Theta) = \begin{cases} 0 & \text{if } \Theta \leq \Theta_{in}, \\ \frac{\Theta - \Theta_{in}}{\Theta_{out} - \Theta_{in}} & \text{if } \Theta_{in} < \Theta \leq \Theta_{out}, \\ 1 & \text{if } \Theta > \Theta_{out}. \end{cases}$$



# How do we find the best position?

**Exhaustive search:** place the element at each possible solution and compute the cost

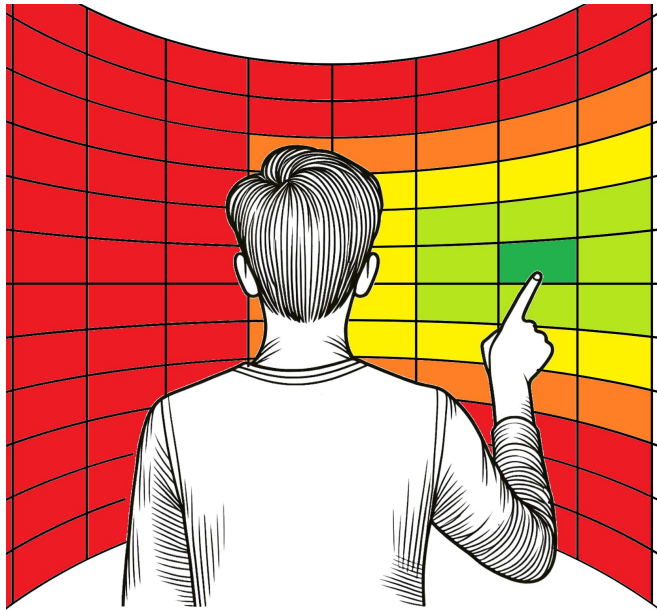


See also: Belo et al. XRgonomics: Facilitating the Creation of Ergonomic 3D Interfaces. CHI 2021.

<https://doi.org/10.1145/3411764.3445349>

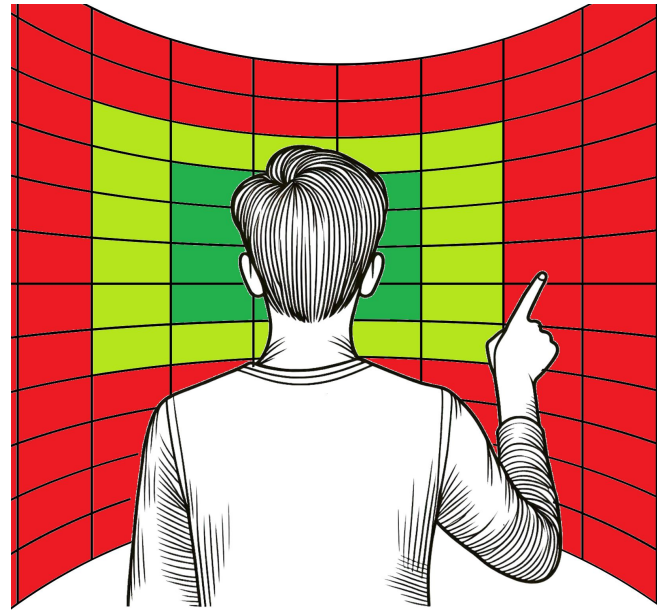
# How do we find the best position?

**Exhaustive search:** place the element at each possible solution and compute the cost



Pointing

+



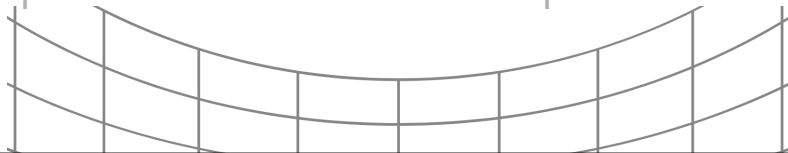
Field of view

See also: Belo et al. XRgonomics: Facilitating the Creation of Ergonomic 3D Interfaces. CHI 2021.

<https://doi.org/10.1145/3411764.3445349>

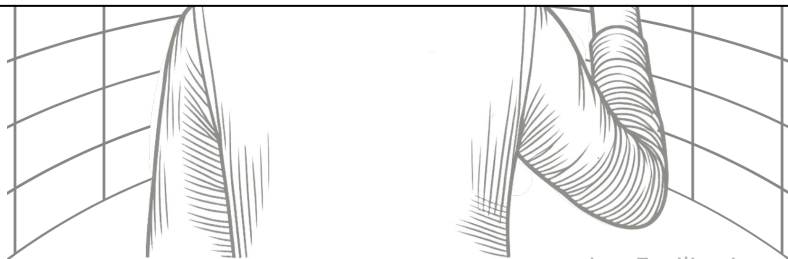
## How do we find the best position?

**Exhaustive search:** place the element at each possible solution and compute the cost



We can use **algorithms** to generate all possible user interfaces

and use models to evaluate them and choose the best one



See also: Beio et al. xrgonomics: Facilitating the Creation of Ergonomic 3D Interfaces. CHI 2021.

<https://doi.org/10.1145/3411764.3445349>





Now let's look at optimization a bit more systematically and implement this problem in code.

Switch to the notebook.

## **Day 2: Optimization methods for interface design and adaptation**

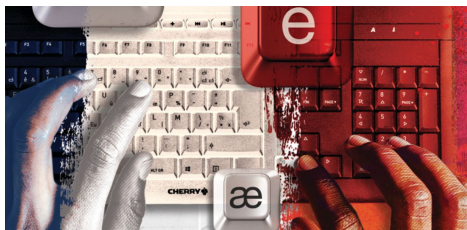
# **Summary and Discussion**

# What have you learned today?

- How to formulate optimization problems
  - Decision variables
  - Constraints
  - Objective functions
- How to solve them
  - Exhaustive search
  - Genetic Algorithms
- Multi-objective optimization
  - Weighted sum
  - Pareto frontier



# Optimization for UI Design vs. Adaptation+Personalization



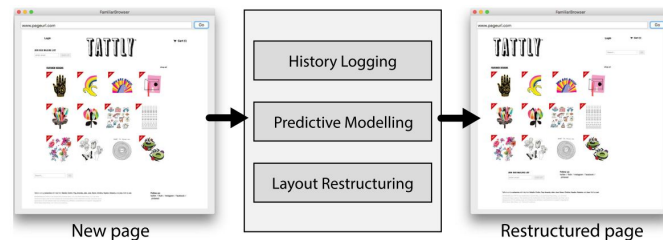
Keyboard layouts, e.g. [Feit, 2021](#)



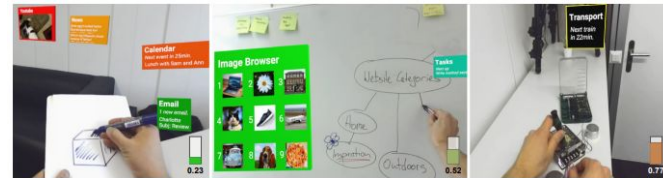
Multi-device, multi-user interfaces, [Park, 2018](#)

Interactive optimization of menus, [Bailly 2013](#)

Functionality selection, [Oulasvirta 2017](#)



Familiarisation [Todi, 2019](#)



Cognitive load [Lindlbauer, 2019](#)

Adapting XR interfaces, [Belo 2022](#)

Interactive adaptation with multiple objectives, [Johns 203](#)

# Optimization for UI Design versus Adaptation

- Computational complexity
  - One-shot optimization, vs
  - continuous real-time optimization

Affects objective function, problem formulation, choice of solver, ...

- Data source
  - static user models and simulations, offline data collection, etc. vs.
  - real-time data, e.g. for online parameter inference, about user's context, etc.
- Interaction with designers/developers versus end-users
  - Optimization as part of an iterative and user-centred design process [\[Feit 2021\]](#)
  - Additional considerations such as *when* the optimization is triggered, *how* the UI transitions to a new state, etc. see [\[Belo 2022\]](#), what is the adaptation cost for users



# Optimization can be used for more

Interaction Design, e.g. [gestural interaction](#), [haptic feedback](#)

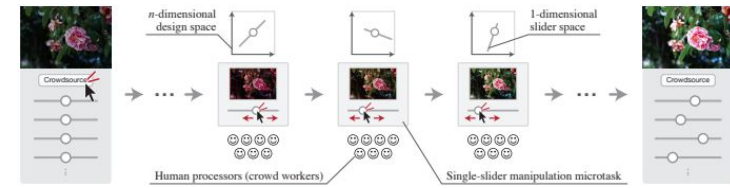
Interactive human-in-the loop optimization for [photo editing](#), [music](#), etc.

Optimization of [data plots](#)

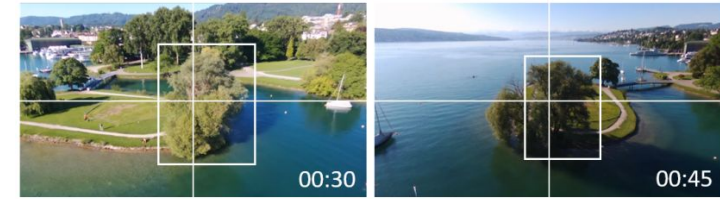
Providing [design suggestions](#), helping in [design ideation](#), or suggesting [alternative designs](#)

[Drone videography](#) and [interactive path planning](#)

Fabrication for [personalization](#) or [compliance](#)



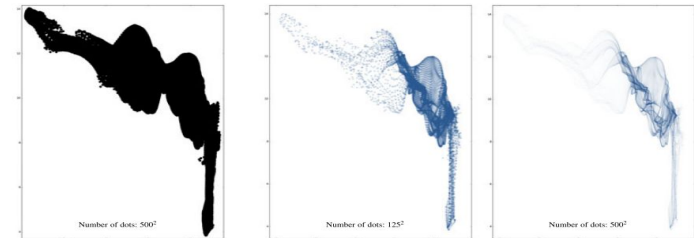
[Koyama 2017](#)



[Gebhardt, 2021](#)



[Hofmann, 2023](#)



[Micallief, 2017](#)

# Some optimization methods / solvers

Method	Gradient	Solution	Variables	Implementation
Exhaustive Search	zero-order	exact	discrete	See the notebook
Branch and Bound	zero-order	exact	discrete	<a href="#">Python-MIP</a>
Simulated Annealing	zero-order	approximate	cont.+discr.	<a href="#">Scipy</a>
Genetic Algorithms	zero-order	approximate	cont.+discr.	<a href="#">Pymoo</a>
Gradient Descent	first-order	approximate	continuous	<a href="#">Scikit learn</a>
Interior Point optimization	second-order	approximate	cont.+discr.	<a href="#">Ipopt</a> ( <a href="#">python bindings</a> )
Bayesian Optimization	zero-order	approximate	mostly cont.	<a href="#">BOtorch</a>

# Critical for success

- The formulation of the optimization problem
  - There are many types of [standard problems](#) (e.g. assignment problem) in operations research. Casting your problem as a standard problem gives you access to suitable algorithms
- The data and models used as objectives (more on modeling in the coming days)
- Suitable trade-offs between multiple objectives

# Benefits of using optimization in HCI

- *Explicitness*: uncovers assumptions, trade-offs, design choices, ...
- *Data -driven*: data and observations directly influence design and steer adaptation
- *Quality*: Can give guarantees on the goodness of designs, can cover the whole design space
- *Automation*: reduces manual effort and design time, designer can focus on creative aspects and problem formulation
- *Flexibility*: enables interfaces to optimally adapt to changing context, users, devices, preferences, etc.
- *Knowledge transfer*: Design knowledge can be transferred to other devices or contexts

# Challenges in using optimization

- *Ramp-up costs*: formal definition and development of search methods can be laborious and risky
- *Hard problems*: real-world cases can be hard to model and solve, collaborate with optimization experts
- *Reliance on data and models*: defining meaningful objective functions (e.g., aesthetics, learnability) and gathering data or developing models to quantify them may require support from domain experts, cognitive scientists, machine learning experts etc.



*We are always happy to collaborate.  
Reach out if you are interested in  
a specific problem*

# Research opportunities

- Computational design and adaptation tools for standard design problems
- "Participatory optimization": optimization tools that
  - Integrate into an iterative user-centred design process
  - That can be used by non-experts (w.r.t. optimization)
  - Carrying out diagnostics and visualizations to understand the results, trade-offs, decisions
  - Interaction techniques for end-users to participate in an adaptation process
- Models and simulations of user behavior
- Online learning and parameter inference
- Empirical evaluation and testing of adaptation methods
- ...

# Further reading

**More on integer programming and the assignment problem:** Feit, A.(2019). [Integer Programming for UI Optimization](#), Lecture notes from 5th Summer School on Computational Interaction, New York

**An introduction to combinatorial optimization:** Oulasvirta, A.(2019). [Introduction to Combinatorial Optimization](#), Lecture notes from 5th Summer School on Computational Interaction, New York

## More research papers:

Oulasvirta, A. & Karrenbauer, A. (2018). Combinatorial Optimization for Interface Design, in [Computational Interaction](#). Oxford University Press.

Feit, A. M. (2018). [Assignment Problems for Optimizing Text Input](#). Aalto University publication series Doctoral Dissertations.

Oulasvirta, A., Dayama, N. R., Shiripour, M., John, M., & Karrenbauer, A. (2020). [Combinatorial Optimization of Graphical User Interface Designs](#). Proceedings of the IEEE

Lindlbauer, D., Feit, A. M., & Hilliges, O. (2019). [Context-aware Online Adaptation of Mixed Reality Interfaces](#). In Proc UIST'19

Evangelista Belo, J. M., Lystbæk, M. N., Feit, A. M., Pfeuffer, K., Kán, P., Oulasvirta, A., & Grønbæk, K. (2022). [AUII-the Adaptive User Interfaces Toolkit for Designing XR Applications](#). In Proc. UIST'22