



max planck institut
informatik

Improvements to Keyboard Optimization with Integer Programming

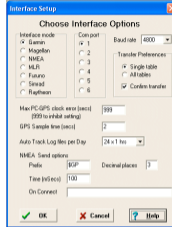
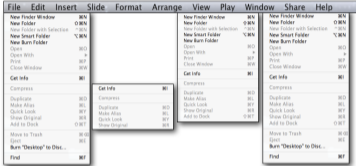
Andreas Karrenbauer¹ Antti Oulasvirta^{1,2}

¹Max Planck Center for Visual Computing and Communication,
Saarbrücken, Germany

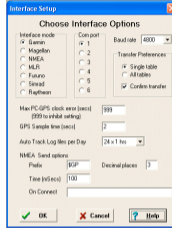
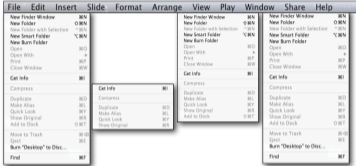
²Aalto University, Helsinki, Finland

ACM UIST 2014

User Interface Optimization



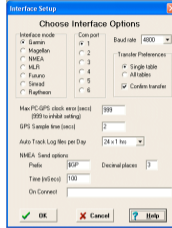
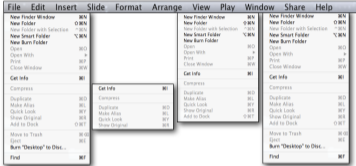
User Interface Optimization



Brute Force

- Explicit enumeration
- Nonpractical for UI
(immense design spaces $> 10^{26}$)
- No guarantees when stopped prematurely

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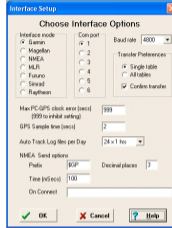
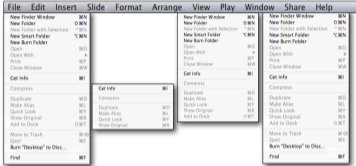
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Random Search Heuristics

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- **Practical**
- **But by chance and no guarantees**

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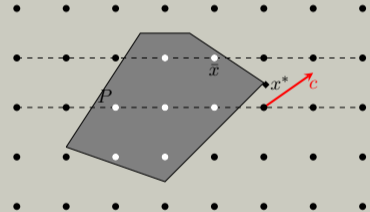
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Integer Programming

Given a set of constraints, decide whether an integer point satisfies all of them.

Benefits

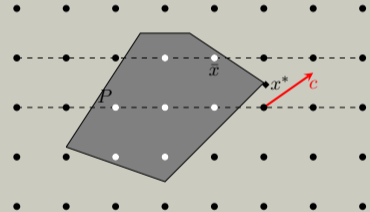


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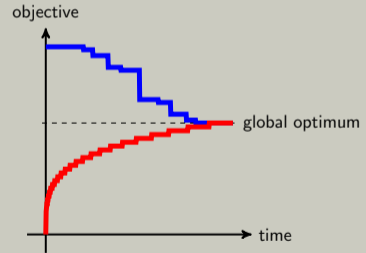


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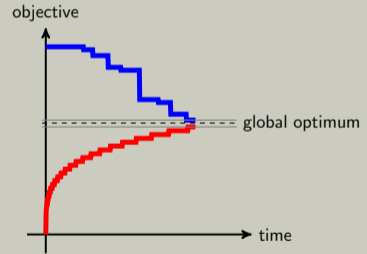


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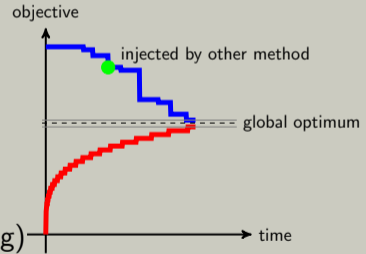


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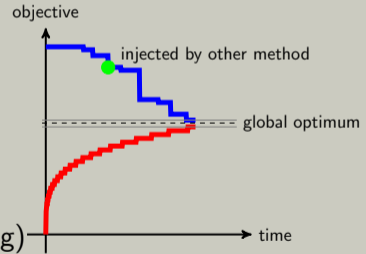


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6. Practical



Overview of the Approach

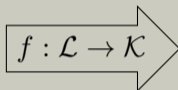
Steps

1. Definition of constraints and objective
2. Formalization with decision variables
3. Integer Programming Model
4. Refinement by reformulations and linearizations
5. IP-Solver

The Letter Assignment Problem – Constraints and Objective

Constraints: find an assignment (1-to-1 correspondence)

$$\mathcal{L} = \left\{ \begin{array}{l} ABCDEFGHI \\ JKLMNOPQR \\ STUVWXYZ \end{array} \right\}$$



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PSY

$$f : \mathcal{L} \rightarrow \mathcal{K}$$



Objective: maximize typing speed, i.e., minimize the expected inter-key intervals

$$\sum_{k, \ell \in \mathcal{L}} p_{k, \ell} \cdot t_{f(k), f(\ell)}$$

bigram probabilities $p_{k, \ell}$

inter-key intervals $t_{i, j}$

The Letter Assignment Problem – Decision Variables

Model assignment by binary decision variables

$$x_{lj} = \begin{cases} 1 & \text{if letter } \ell \text{ is assigned to keyslot } j, \text{ i.e., } f(\ell) = j \\ 0 & \text{otherwise} \end{cases}$$

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Constraints

- Each letter l is assigned to exactly one keyslot $\leadsto \forall l \in \mathcal{L} : \sum_{j \in \mathcal{K}} x_{l,j} = 1$
- Each keyslot j contains exactly one letter $\leadsto \forall j \in \mathcal{K} : \sum_{l \in \mathcal{L}} x_{l,j} = 1$

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Objective

expected inter-key interval $\sum_{k, l \in \mathcal{L}} p_{k, l} \cdot t_{f(k), f(l)}$ \rightsquigarrow $\sum_{i, j \in \mathcal{K}} \sum_{k, l \in \mathcal{L}} p_{k, l} \cdot t_{i, j} \cdot x_{k, i} \cdot x_{l, j}$

The Letter Assignment Problem – IP Model

Quadratic Assignment Problem

$$\min \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n \sum_{l=1}^n p_{kl} t_{ij} x_{ki} x_{lj}$$

$$\text{s.t. } \sum_{l=1}^n x_{lj} = 1 \quad \forall j \in \{1, \dots, n\}$$

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The Letter Assignment Problem – IP Model → IP-Solver

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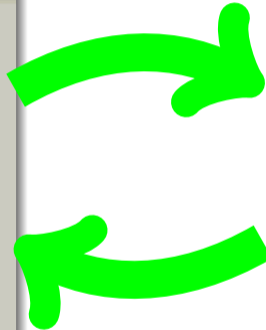
The Letter Assignment Problem – IP Models \leftrightarrow IP-Solver

Reformulations

$$\begin{aligned} \min & \sum_{i=1}^n \sum_{j=1}^n \sum_{k=1}^n \sum_{\ell=1}^n p_{k\ell} x_{ki} x_{\ell j} \\ \text{s.t.} & \sum_{\ell=1}^n x_{\ell j} = 1 \quad \forall j \in \{1, \dots, n\} \\ & \sum_{j=1}^n x_{\ell j} = 1 \quad \forall \ell \in \{1, \dots, n\} \\ & x_{\ell j} \in \{0, 1\} \quad \forall \ell, j \in \{1, \dots, n\} \end{aligned}$$

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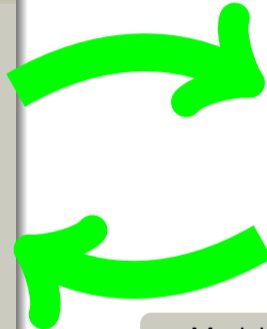
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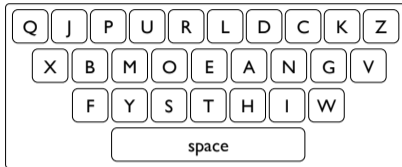
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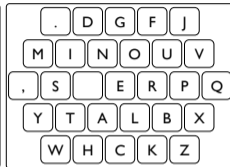


- Models can be solved in parallel
- Complementary to other approaches such as random search heuristics

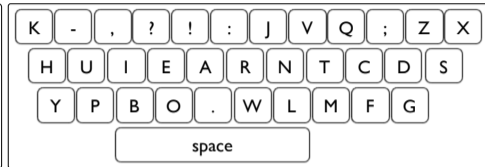
Results



Virtual 3-row trapezoid
26 keys
Fitts' model

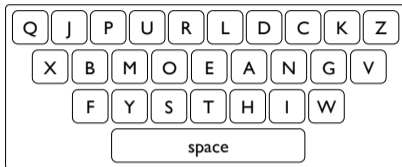


Virtual hexagonal
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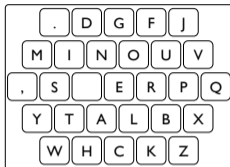


Physical 3-row standard
34 keys
Hiraga et al. 10-finger model

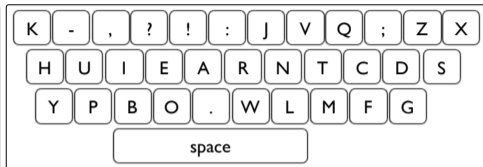
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Keyboards (wpm)

qwerty (33.65)
DSK (33.97)

This (43.10)

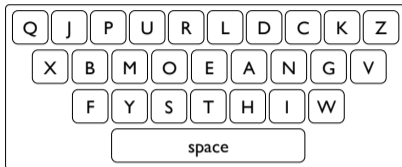
metropolis (42.04)
justhci (42.38)

This (42.65)

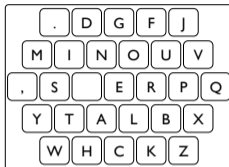
qwerty (73.21)
DSK (76.59)

This (78.36)

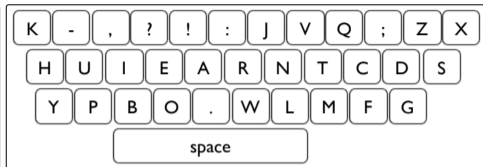
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any (< 45.55)

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any (< 79.95)

Summary

Take-home messages

- Integer Programming proved useful for UI Optimization
- IP provides rigorous mathematical guarantees for the optimum
- IP is complementary to other methods (e.g., random search heuristics)

`http://resources.mpi-inf.mpg.de/keyboardoptimization/`
(by Oct 15)



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Thank you for your attention!

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