

Intelligent Text Input and Optimization

Learning Goals

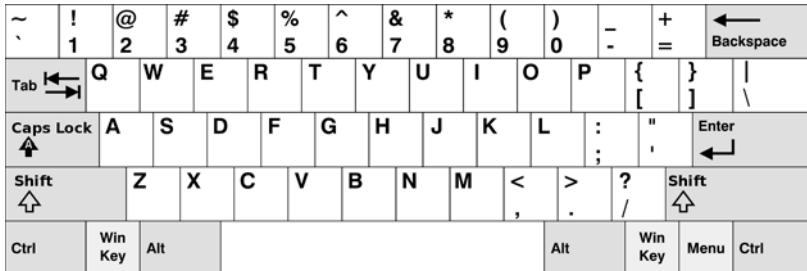
- Combinatorial optimization as a UI design approach
- Probabilistic methods for handling input
- Application example: keyboards

Optimizing User Interfaces

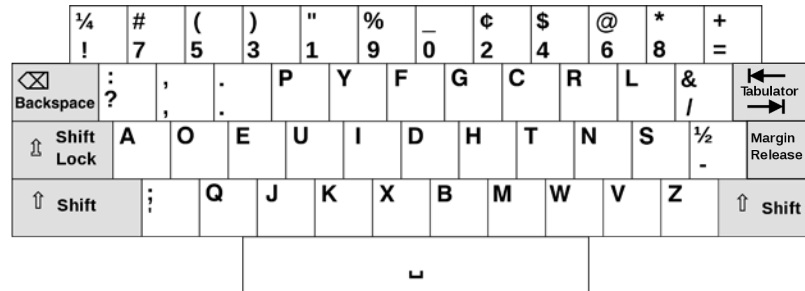
Example: keyboard layout optimization

Motivation: Fast typing without errors

- Are some layouts better than others?
- If so, how do we find the best one?



QWERTY, by Christopher Sholes, 1873



Dvorak, by August Dvorak, 1936



By: https://commons.wikimedia.org/wiki/File:KB_United_States.svg, https://commons.wikimedia.org/wiki/File:KB_United_States_Dvorak.svg

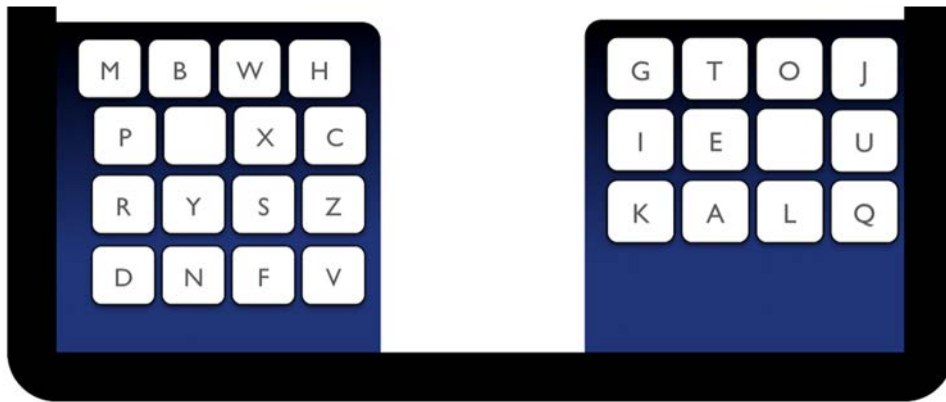
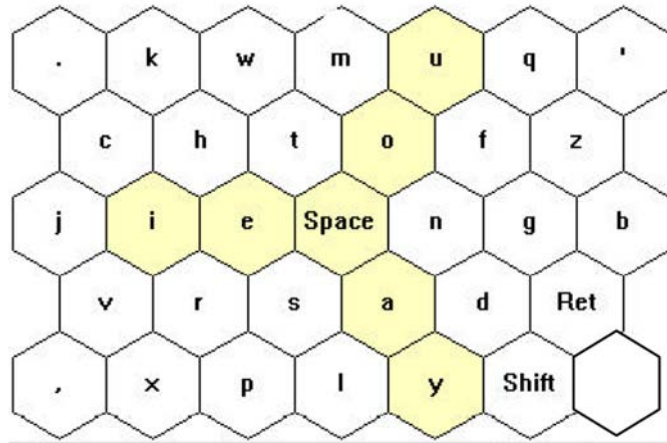
Key Assignment Problem



How many layouts are there? $26! = 4 * 10^{26}$

For comparison - stars in the universe: https://www.esa.int/Science_Exploration/Space_Science/Herschel/How_many_stars_are_there_in_the_Universe

Examples of Optimised Designs



Zhai et al. 2000, Dunlop and Levine 2012, Oulasvirta et al. 2013, Gong et al. 2018

What is „optimal“?

- **Design space:**
Best among which options?
- **Design objective:**
Best for what?
- **Optimizer:**
How to find the best design?
Best with which guarantees?



Design Space, formalised

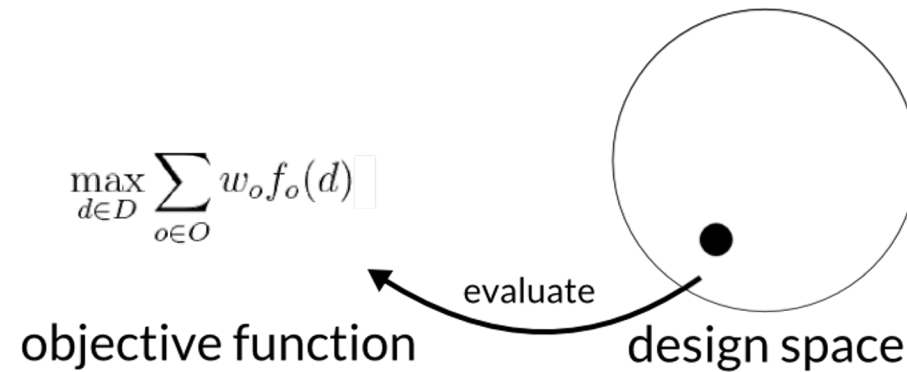
$$d = \begin{pmatrix} d_1 \\ d_2 \\ \vdots \\ d_n \end{pmatrix}$$

Design space ***D***
with ***n*** design variables

Design Space: Set of possible layouts



Objective Function: How to judge a layout?



Objective Function: How to judge a layout?

- Finger movement time (e.g. Fitts' law)

$$t(k_1, k_2) = a + b \log_2 \left(\frac{D}{W} + 1 \right)$$

- Language properties (e.g. bigram frequencies)

e.g. $p("n"|"e") = 0.001$

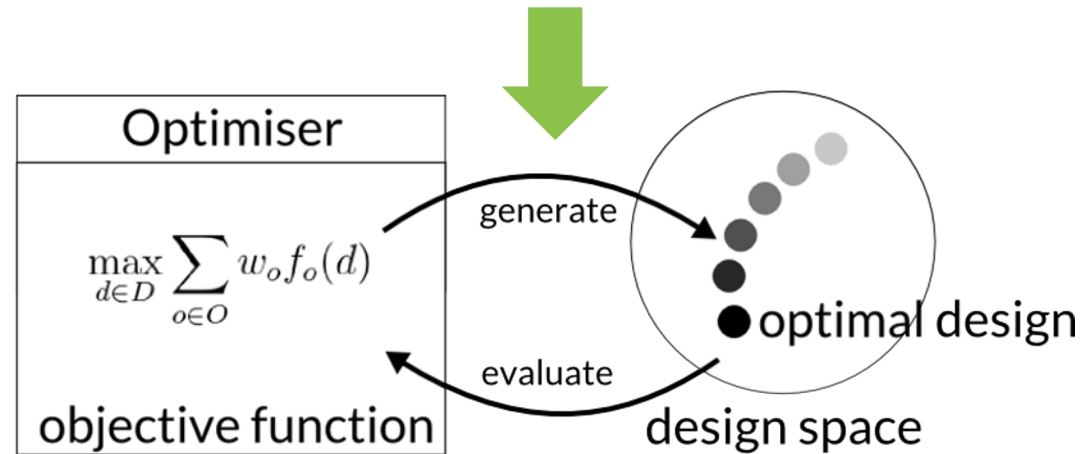


- Combined: mean time between two key presses

$$f(d) = \sum_{k_1 \in K} \sum_{k_2 \in K} p(d(k_2)|d(k_1)) t(k_1, k_2)$$

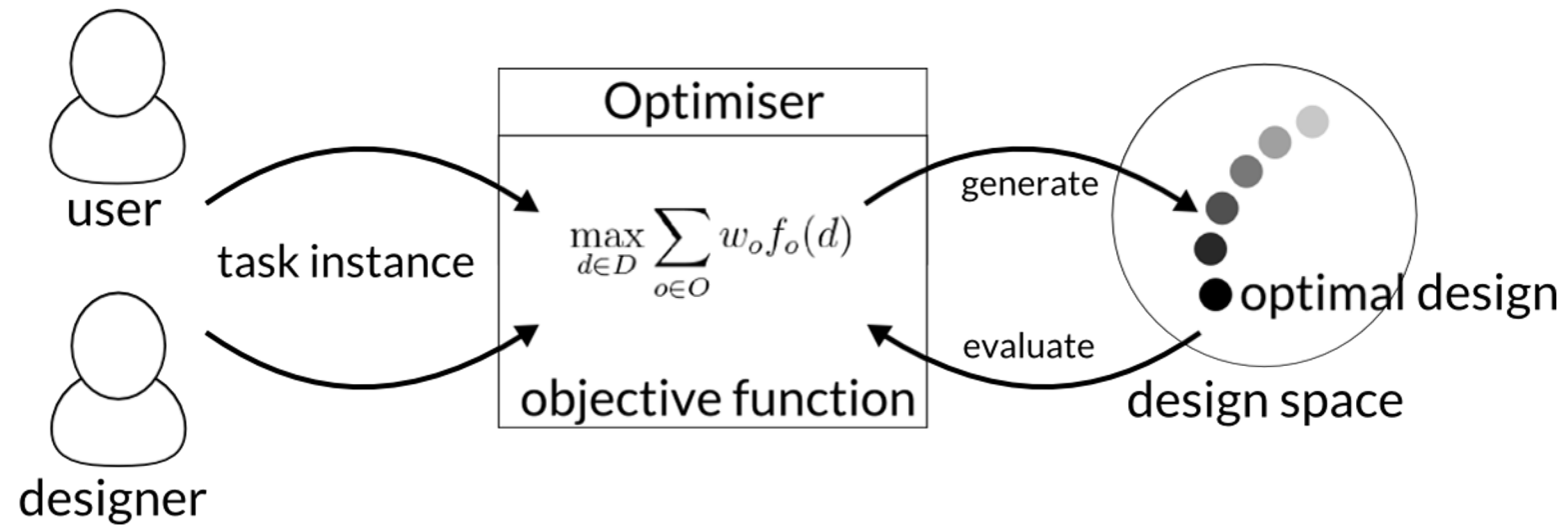
where the design d maps from keys to characters

Optimizer: How to pick layouts?



Design Task

e.g. keyboard layout optimization

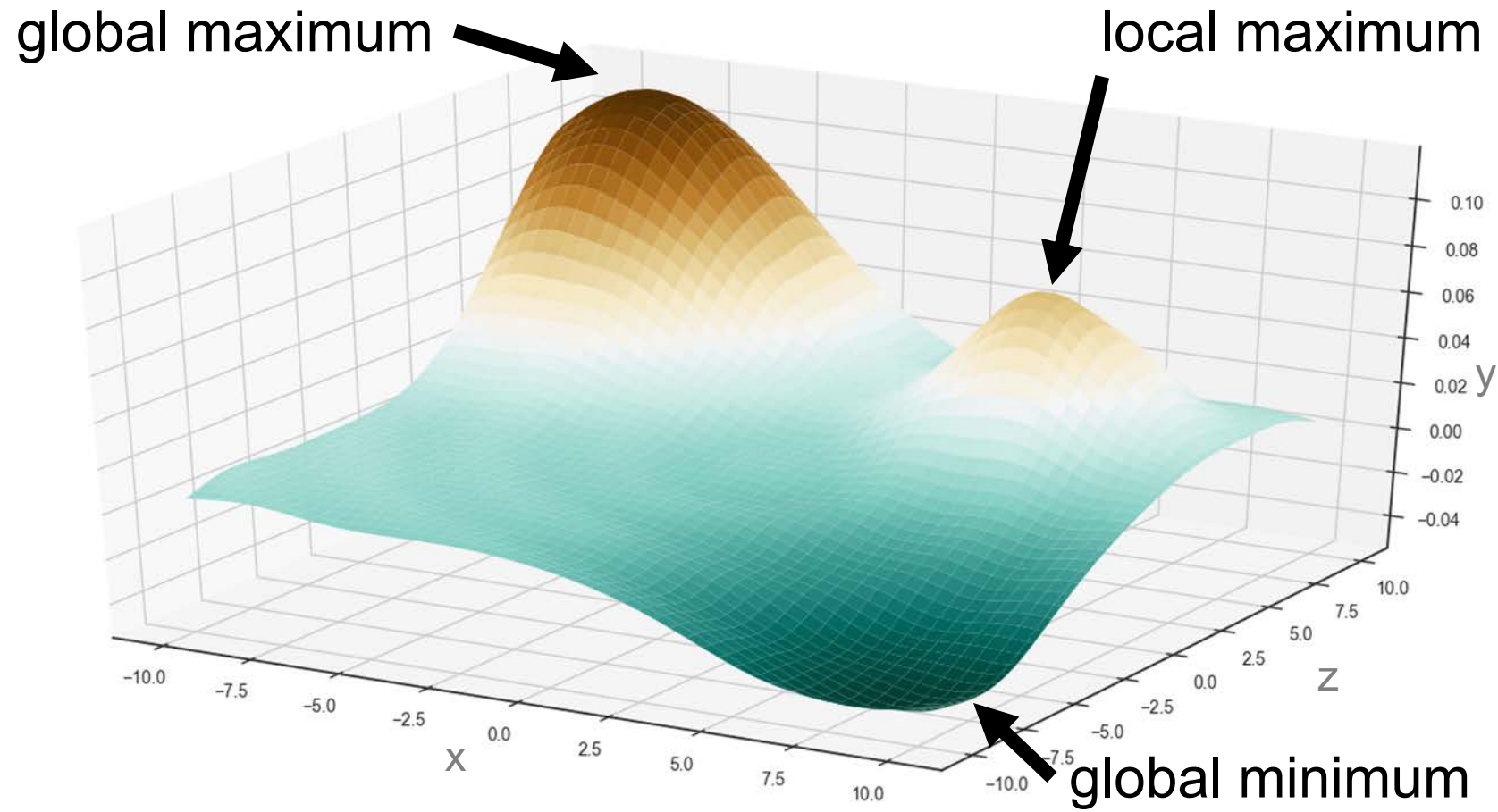


A Simple Optimizer

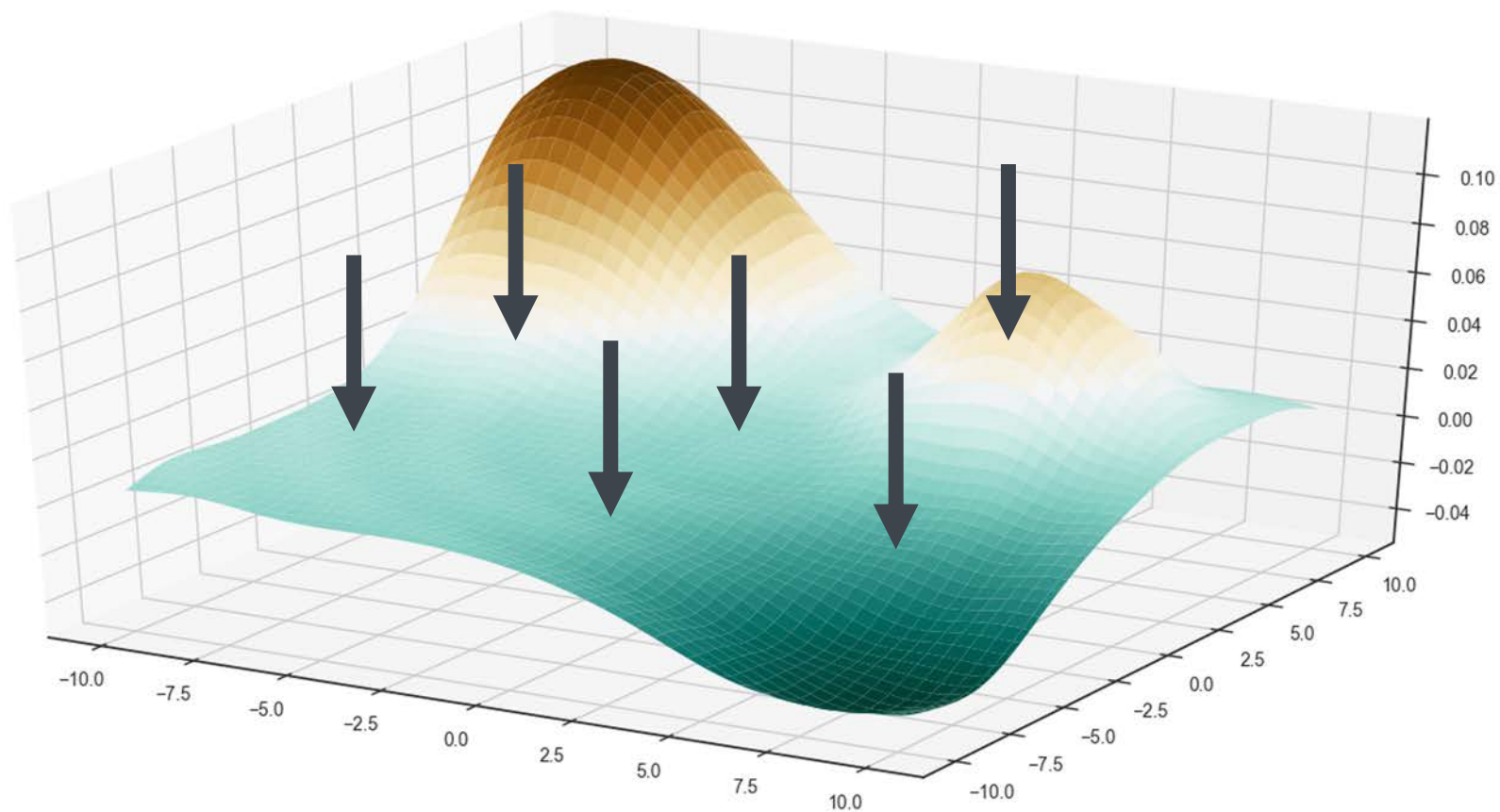
- Can you think of a trivial optimizer?
- Random Search:
 1. Generate random design
 2. Keep if better than current best design
 3. Repeat

Optimization Landscape

Here: objective function (y) across two design parameters (x, z)



Random Guessing



Optimizers

- **Heuristic methods** (e.g. Simulated Annealing)
 - + Flexible
 - Not guaranteed to find global optimum
- **Exact methods** (e.g. Integer Programming)
 - + Guarantees
 - Less flexible objectives

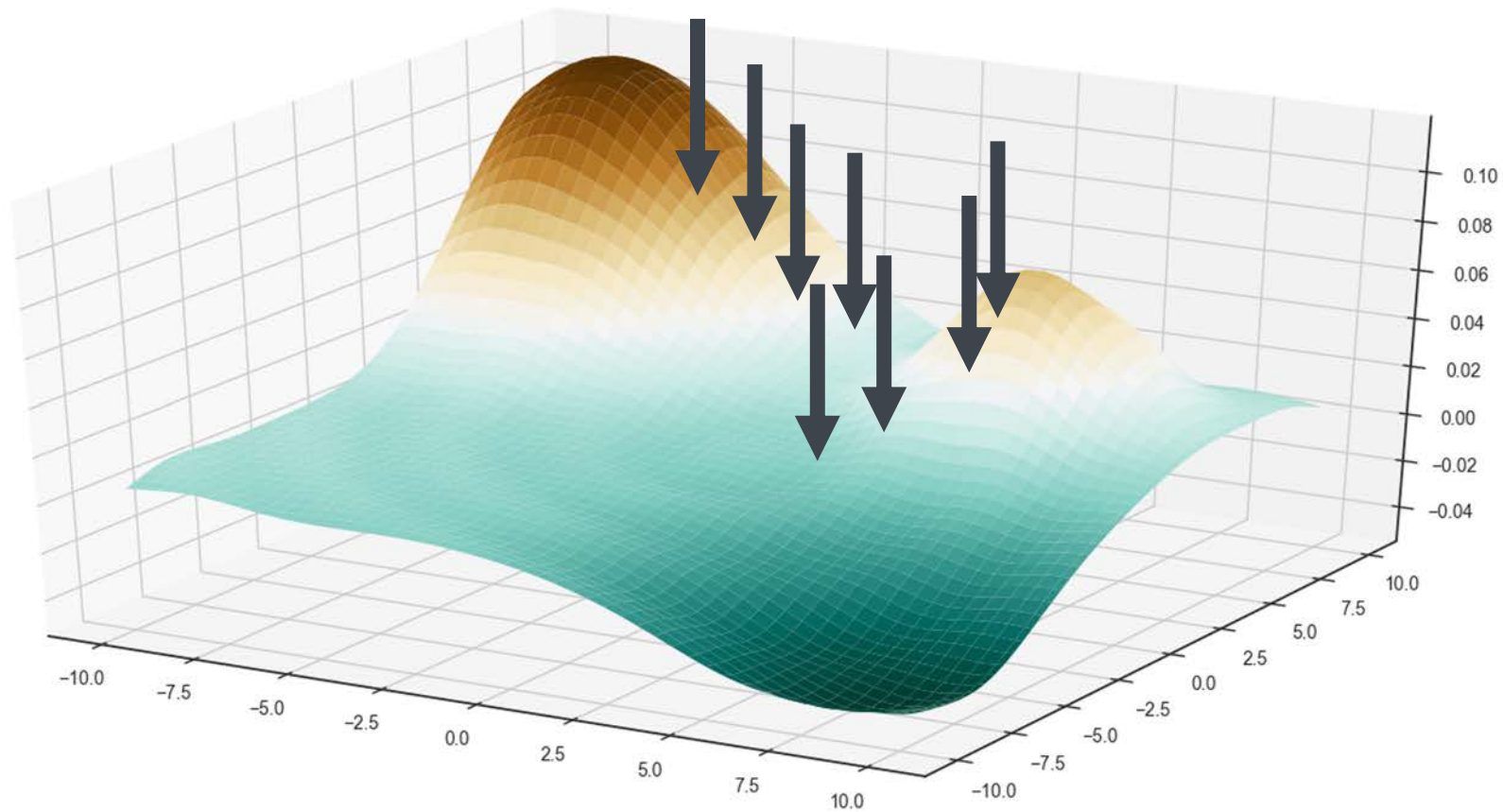
Example: Simulated Annealing

- Metaphor: shaping hot metal
- Flexible at beginning (exploration)
- Gradually more rigid as it “cools down” (exploitation)



```
For i=0 to N:  
  reduce temperature T  
  generate neighbor design  
  if better: go to neighbor  
  else: still go with chance relative to T
```

Simulated Annealing



Example Results

“boxpum”

found with random search

WPM: 31.97

j k z f b o x p u m

d n i t c y r l v

h a e s g q w

“aero”

found with Simulated Annealing

WPM: 36.61

x y c h t i n d k z

b l a e r o f g j

v p m s u w q

Challenge – can you find a better layout than “aero”?
Use the provided python notebook as a starting point.

Example Results

With a modified objective function

“chat”

found with Simulated Annealing

WPM: 34.56

z x y v p l d u s m

q j b f r e o n i

k w g c h a t

What was this layout optimized for?

→ Typing with right thumb, reduce thumb stretching

Potential of Optimization-based Design

- Obtaining information on the design problem and a formal specification
 - Exploring a large design space comprehensively
 - Improving quality and robustness of designs
 - Estimating possible improvements
 - Supporting human designers
 - Optimization during use, personalised UIs
-
- Requires: Models of user behaviour, formal problem definition / objective function, computational capacity, ...

Adaptive and Predictive Keyboards

Motivation: Fast typing without errors

Here: mobile devices

- „Inviscid entry rate“:
Bottleneck is not the
text entry UI but coming
up with the text
- Estimated as 67 WPM

→ Try to reach this on
your phone without errors,
e.g. in an online typing
speed test.

Text entry method	Highest reported entry rate (wpm)
<i>Estimate of the inviscid entry rate</i>	<i>67</i>
Physical thumb keyboards	60 [3]
Gesture keyboards	45 [9]
Optimized on-screen keyboards	45 [12]
QWERTY on-screen keyboards	40 [12]
KALQ thumb keyboard	37 [14]
Half-QWERTY	35 [13]
Twiddler	35 [11]
WalkType	31 [5]
ContextType	28 [6]
Disambiguating keypads	26 [7]
Unconstrained handwriting recognition	25 [8]
Dasher	20 [21]
Mobile speech	18 [18]
Quikwriting	16 [15]
Unistrokes	16 [1]
TiltText	14 [22]
Multi-tap	12 [23]
Graffiti	11 [1]
EdgeWrite	7 [24]

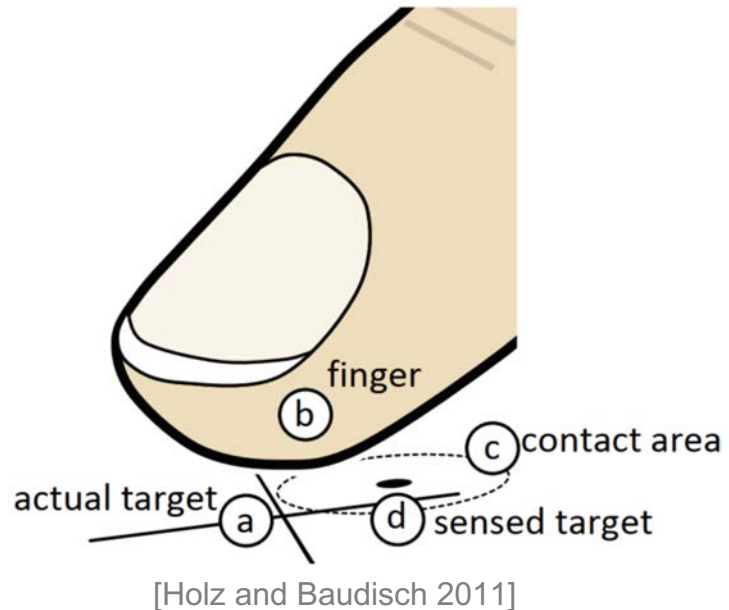
[Kristensson and Vertanen 2014]

Challenges for Mobile Typing

Why is it inaccurate?

Parallax

eye – finger - screen



Mobile use

1-2 fingers, small keys, body movement

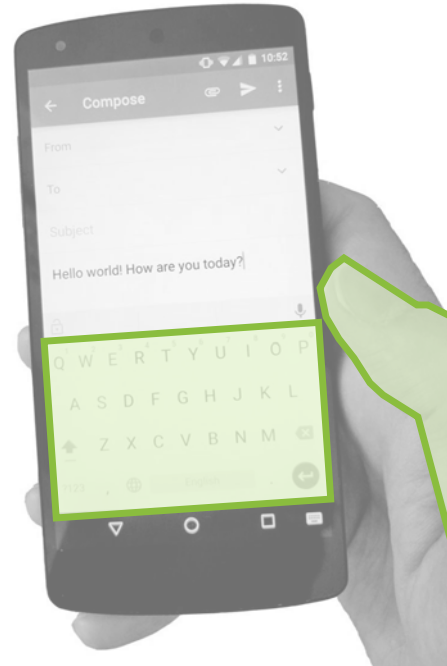
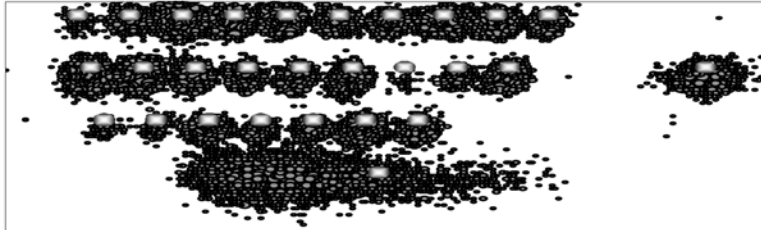


Photo by Ketut Subiyanto

Variance in Touchscreen Keypresses

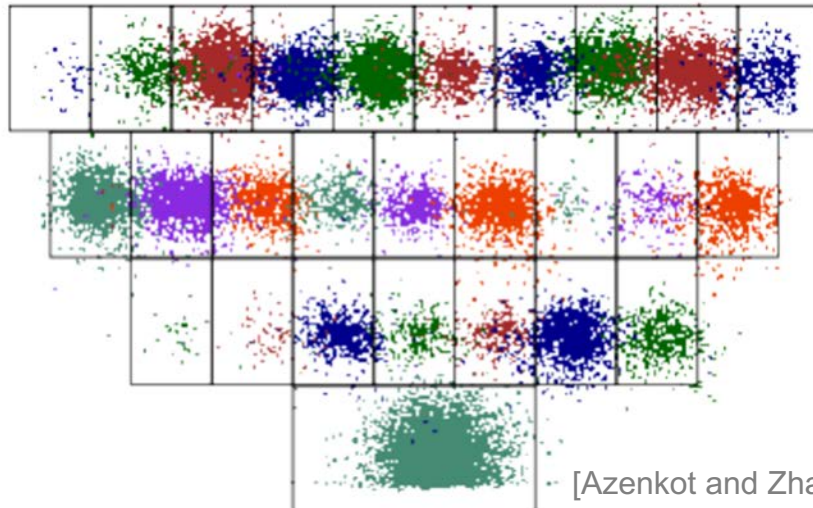
Spread of x,y touch locations around key centres

PDA



[Goodman et al. 2002]

Smartphone



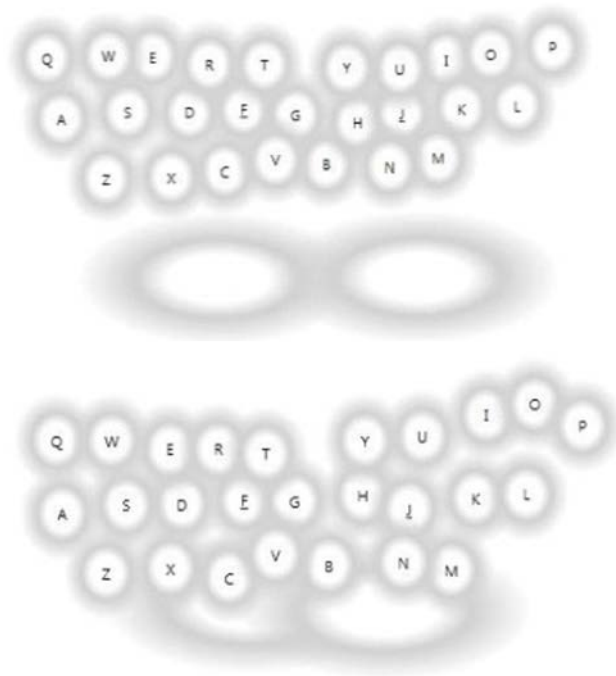
[Azenkot and Zhai 2012]



<https://www.microsoft.com/en-us/swiftkey>

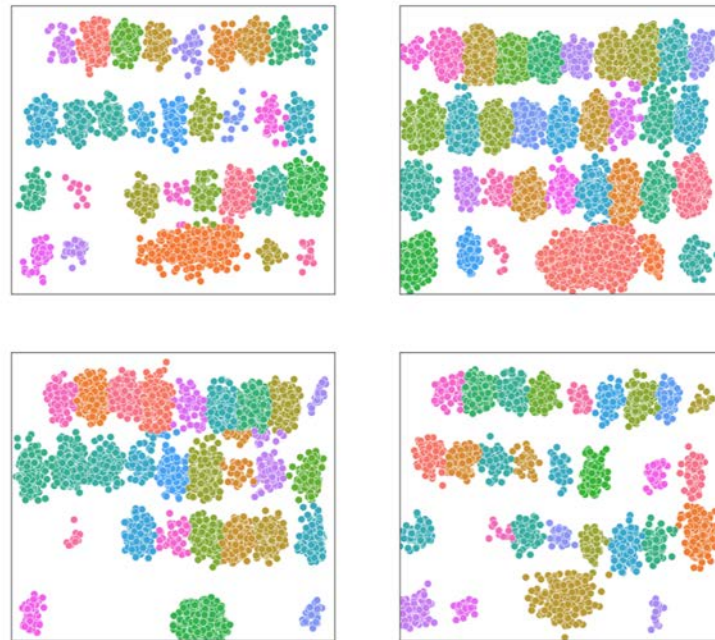
Individual Typing Behaviour

Tabletop



[Findlater and Wobbrock 2012]

Smartphone



[Buschek et al. 2018]

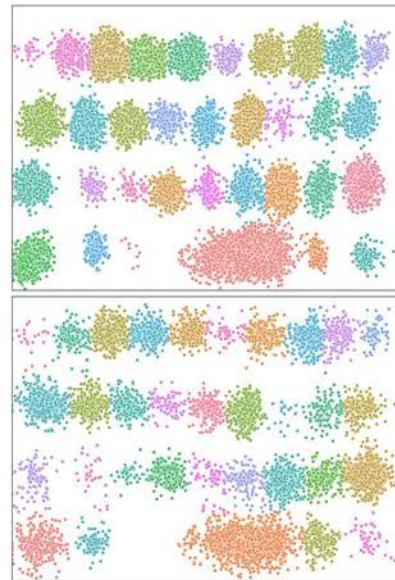
Adapting Keyboards to Typists

Overview

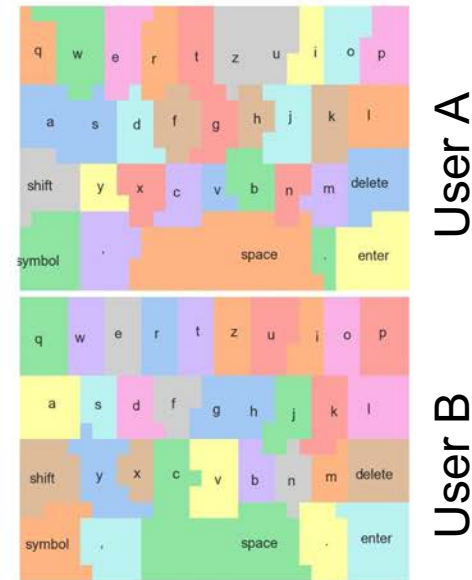
Visible keyboard



Collect touches

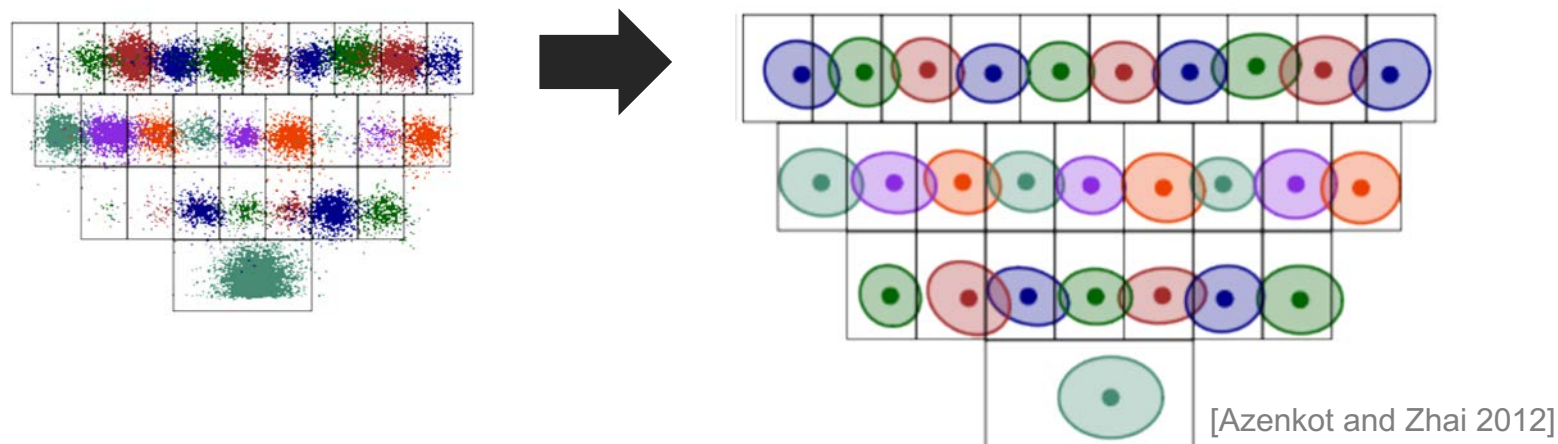


Adapt underlying key regions



Modelling Touchscreen Keypresses

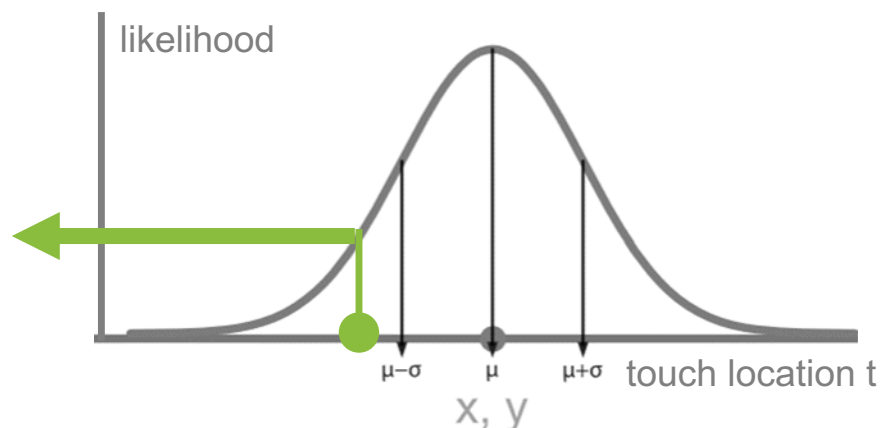
From x,y touch points to one Gaussian per key



Gaussian key model:

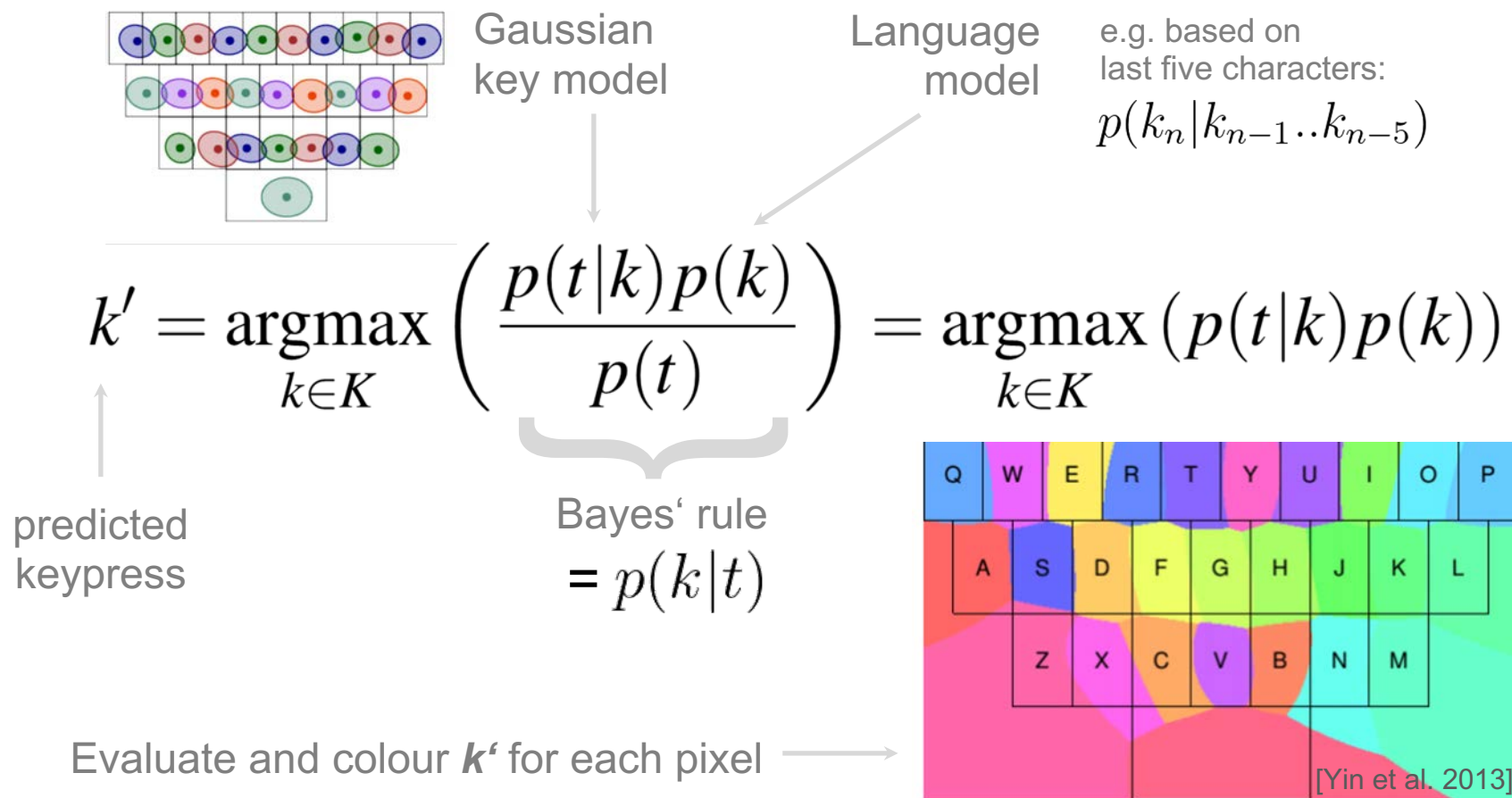
(shown in 1D here)

$$p(t|k) = \mathcal{N}(\mu_k, \sigma_k^2)$$



Probabilistic Keyboard Model

Which key does the user intend to press? i.e. „input decoding“



DIY: Probabilistic Keyboard Model

```
touchX = ... // touch X coordinate
touchY = ... // touch Y coordinate
num_keys = ... // number of keys on keyboard
means = [...] // list of all key means (2D key locations)
variances = [...] // list of key variances (real values) or covariances (2x2 matrices)

probs = [] // list to store the likelihoods of each key being pressed
sum = 0 // variable to store sum of likelihoods for normalisation (see below)

for k = 0 to num_keys: // iterate over all keys
    // evaluate touch location under distribution of the key*:
    prob_t_given_k = multinormal_pdf(touchX, touchY, means[k], variances[k])
    // likelihood of key without touch info; uniform (here), or based on language*:
    prob_k = 1/num_keys
    // store product and add it to the sum of all likelihoods*:
    probs[k] = prob_t_given_k * prob_k
    sum = sum + probs[k]

// normalise, so that the likelihoods add up to 1*:
probs = probs / sum //note: "/" is element-wise division

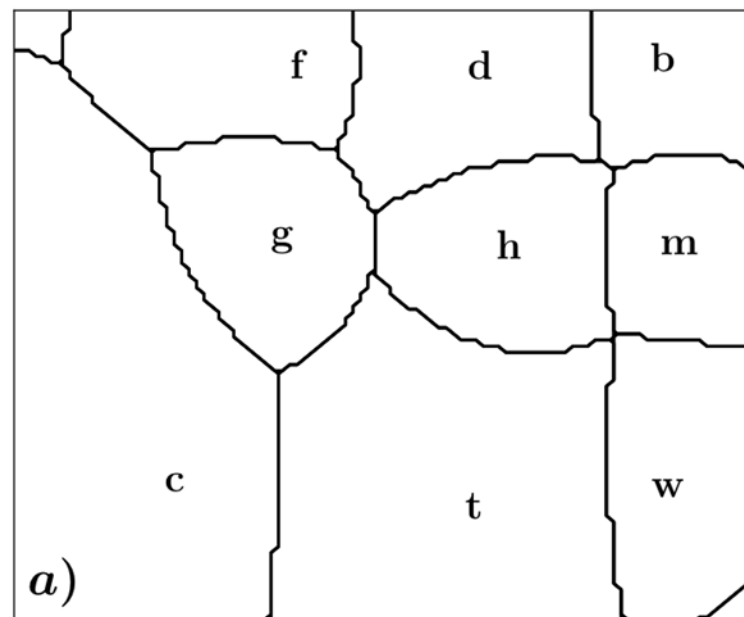
// find most likely key:
pressed_key_index = argmax(probs)
// TODO for adaptation: update means and variances with new touchX and touchY
```

* in real implementation use logarithm and corresponding operations for numerical stability

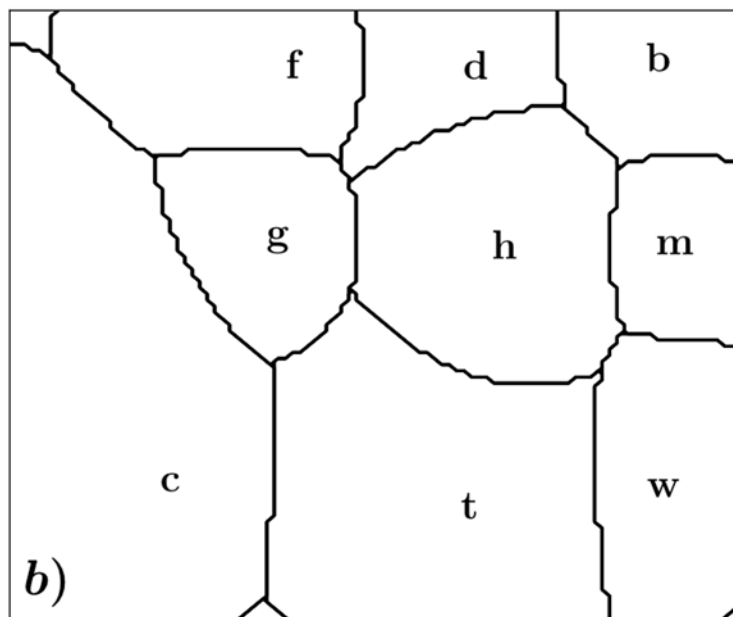
Language Model Influence

Example: bigram model for English

After „n“:



After „t“:



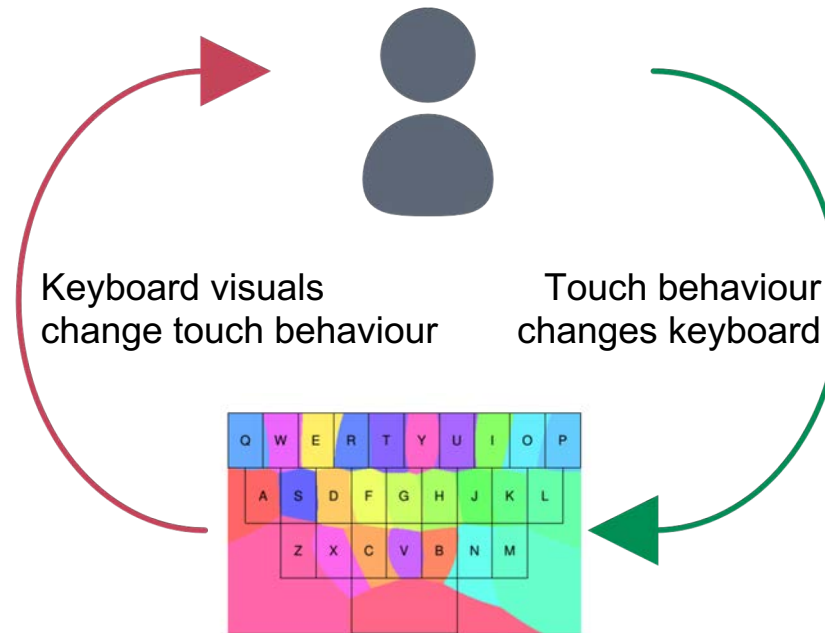
Adaptation in the Background

Why do our keyboards not look like this?



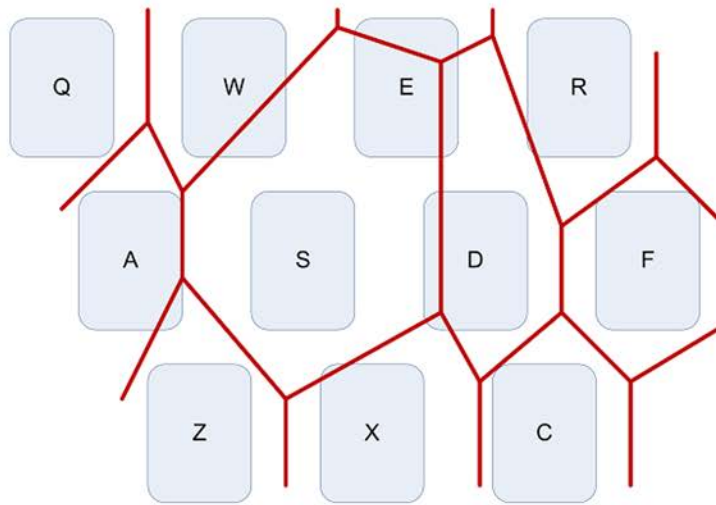
[Yin et al. 2013]

→ Avoid co-adaptation of user and system



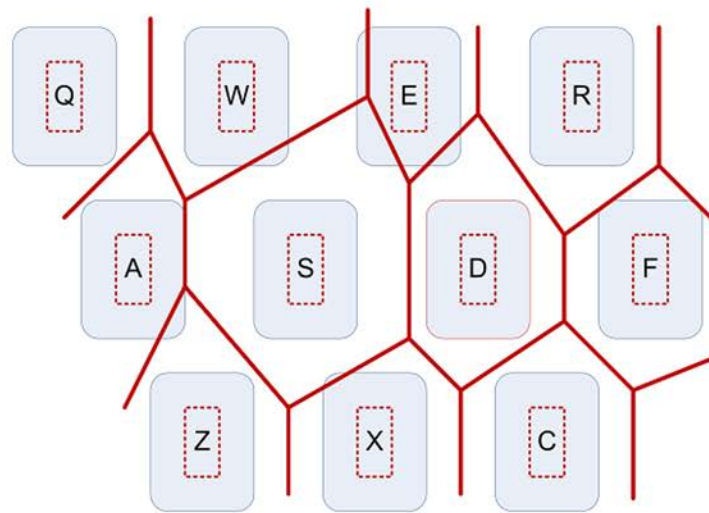
Adaptation vs Distortion

Unlimited adaptation



Here: (almost) impossible to type „e“!

With protected key region

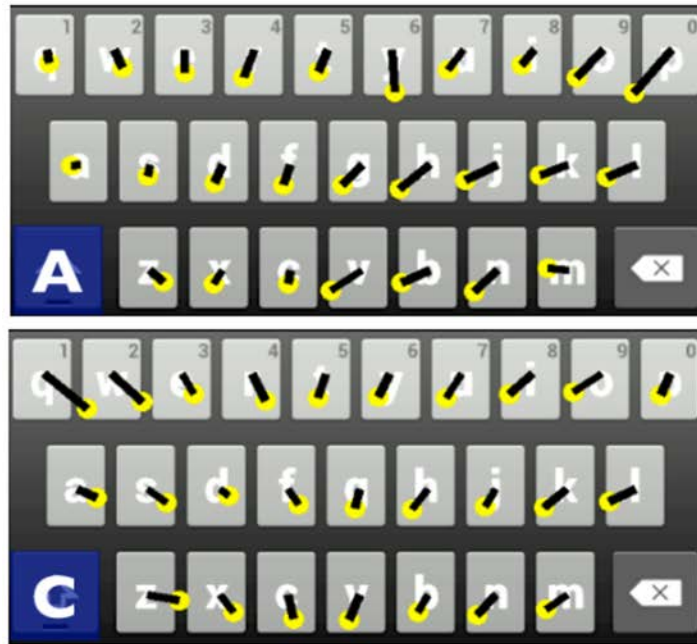


[Gunawardana et al. 2010]

Context Adaptations

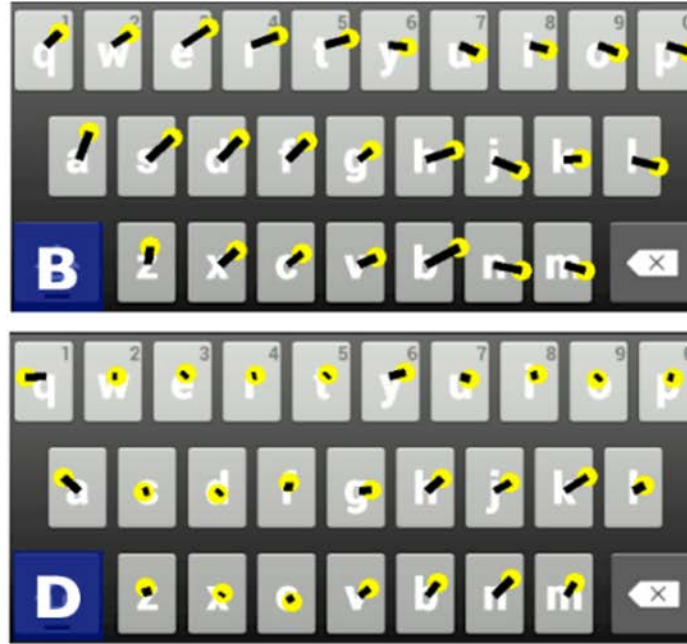
e.g. hand posture – „ContextType“, Goel et al. 2013

Left thumb

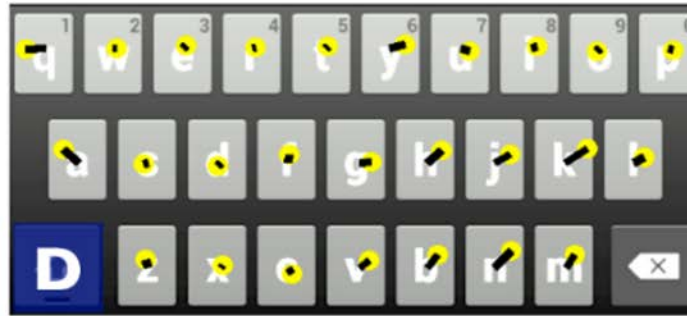


Index finger

Right thumb



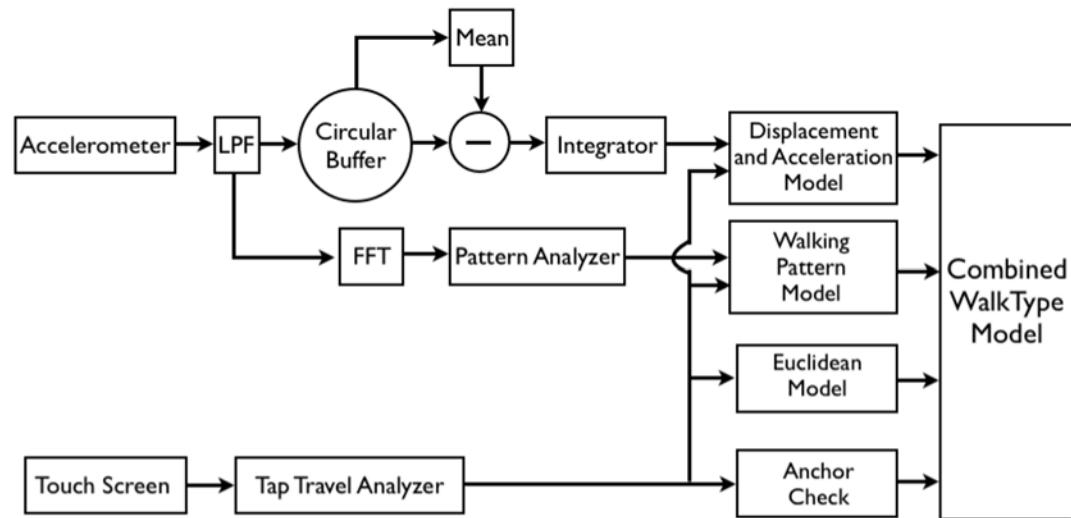
Two thumbs



[Goel et al. 2013]

Context Adaptations

e.g. walking – „WalkType“, Goel et al. 2012



[Goel et al. 2012]

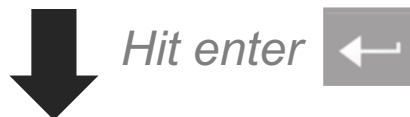
Notebook Example

Visualising adaptive keyboards

Decoding Typing Sequences

- Infer intended input after entering whole word or sentence
 - + More evidence for inference
 - + No need for user to pay attention to intermediate output
 - No intermediate feedback
- Example (sentence-based decoding):

„pleaseforwarxmetheattachement“



„Please forward me the attachement.“

[Vertanen et al. 2015]

Sequence Decoding, formally

- Sequence of user's intended keys/letters (i.e. unknown to system)

$$s = \{k_1, k_2, \dots, k_n\}$$

- Sequence of user's touches (i.e. system's observations)

$$o = \{t_1, t_2, \dots, t_m\}$$

- Can we infer s from o ?
$$s' = \operatorname{argmax}_s (p(o|s)p(s))$$

Note: Same approach as for a single touch but now we have sequences of touches and keys

$$k' = \operatorname{argmax}_{k \in K} \left(\frac{p(t|k)p(k)}{p(t)} \right) = \operatorname{argmax}_{k \in K} (p(t|k)p(k))$$

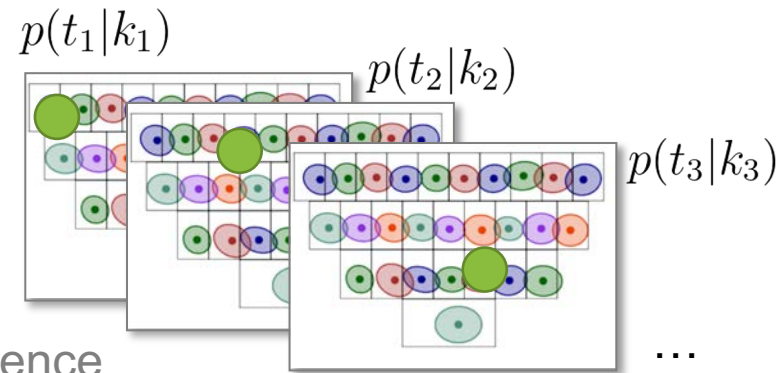
Probabilistic Model for Sequences

See [notebook](#) for an [example implementation](#)

Prior over letter sequences,
i.e. probability of s in a language,
i.e. a language model

$$s' = \operatorname{argmax}_s (p(o|s)p(s))$$

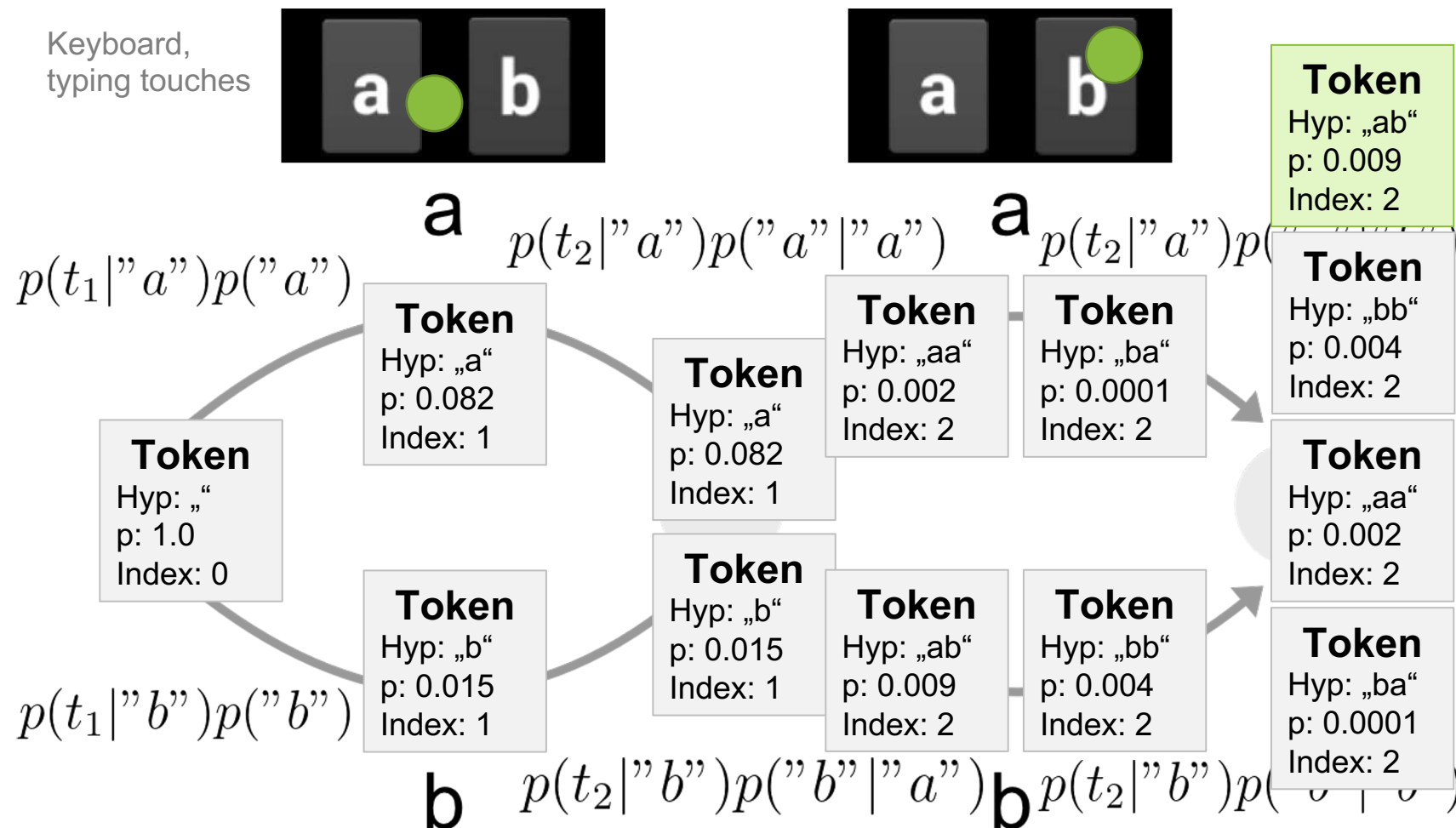
$$p(o|s) = \prod_{i=1}^m p(t_i|k_i)$$



Gaussian key model, now for a touch sequence
(i.e. likelihood of observing the touch sequence o under the assumption that the user intended to write s)

Decoding Typing Sequences

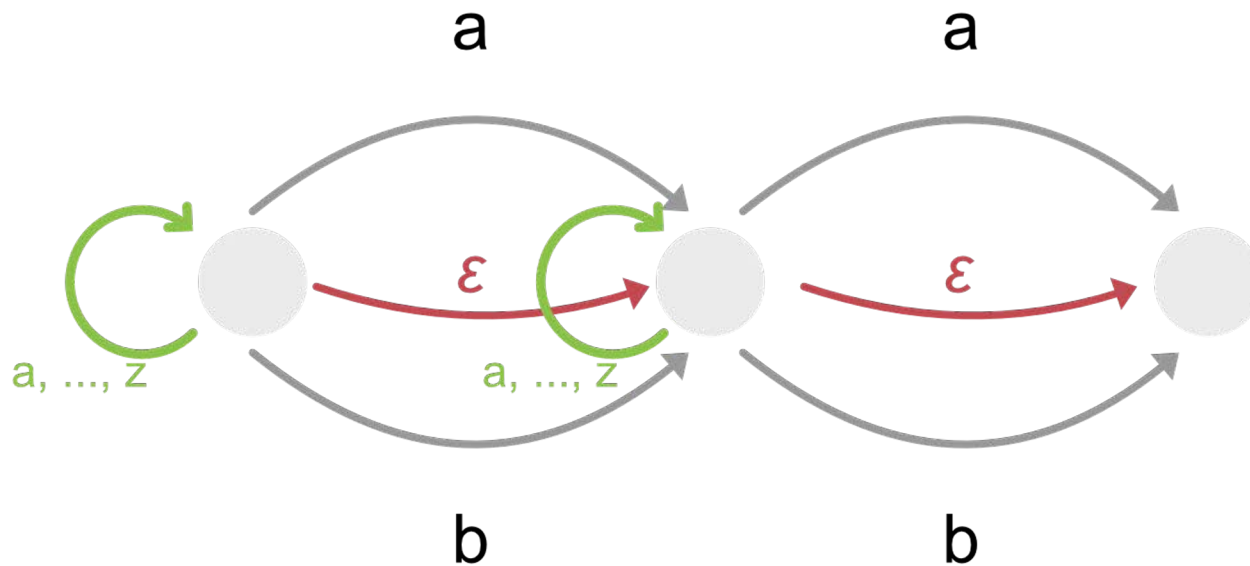
Token passing algorithm



Decoding Typing Sequences

With insertion and deletion

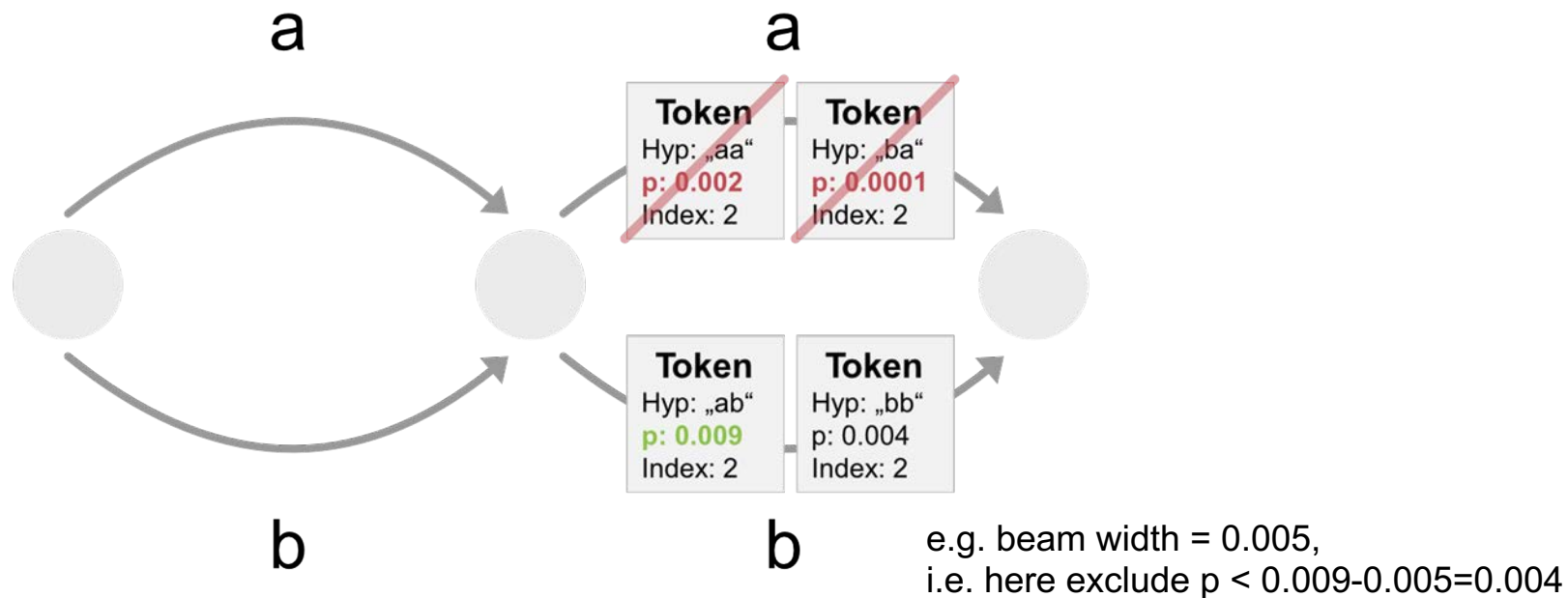
- Previous slide: Substitution-only decoder
- Extensions: **insertion** and **deletion** transitions, with „penalty“



Decoding Typing Sequences

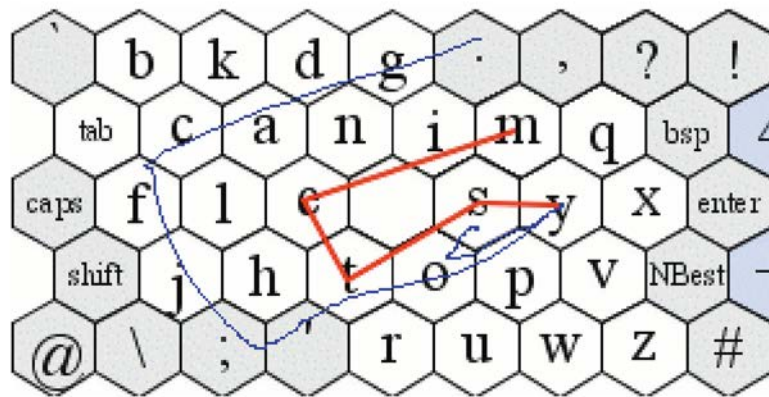
With beam search / pruning

- Problem: Large search space
Substitution-only \rightarrow exponential, Insertion \rightarrow infinite
- Solution: **Beam search / pruning**
Per index, only propagate tokens that are within a certain range (=„beam width“) of the probability of the most likely token.

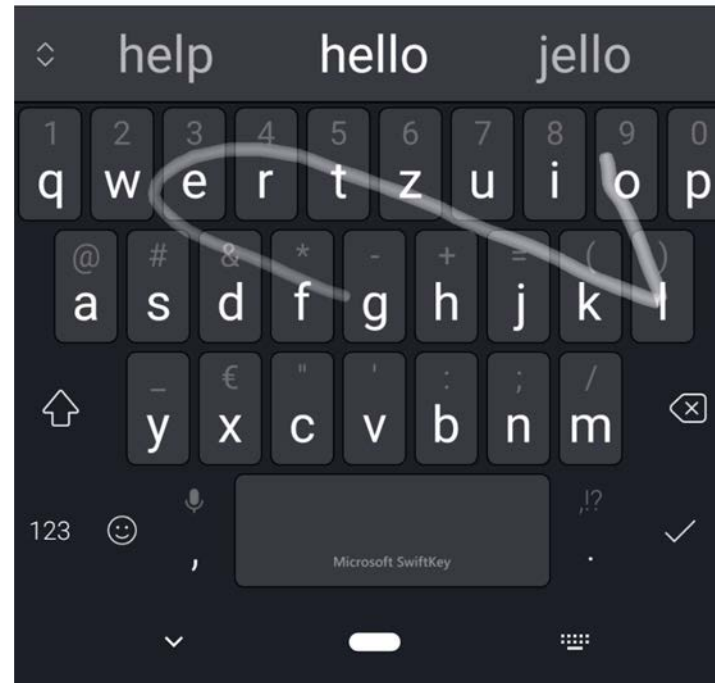


Gesture-based Decoding

Infer intended word from shape of finger trace on the keyboard



„SHARK²“ [Kristensson and Zhai 2004]



Microsoft SwiftKey (screenshot Nov 2020)

Gesture-based Decoding

$$w' = \operatorname{argmax}_{w \in W} (p(\text{trace} | w) p(w))$$

Shape model Language model

Stored template (ideal) shapes
for all words in dictionary W



**Distance
metric**

e.g. see
Kristensson and Zhai,
2004



User's touch trace

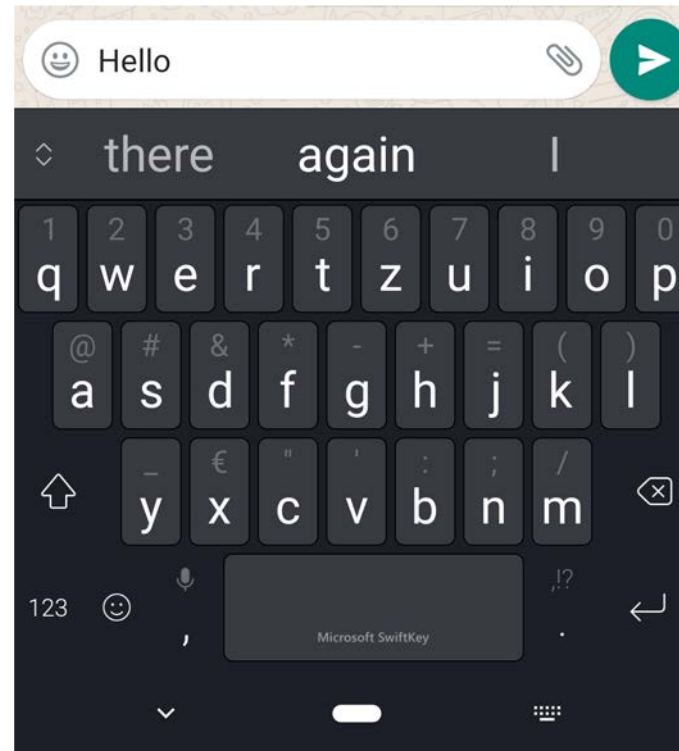


Word Prediction

- So far: Inference used touch input
- Now: Predict *next* word that user has not yet started to type, only using language context

$$p(w_t | w_{t-n} \dots w_{t-1})$$

- E.g. n-gram word models, i.e. context of last n-1 words
- More recently: Deep Learning to include larger context



Summary

- Improving keyboards by probabilistically combining input information with language information
- **Adaptation:**
 - Individual input behaviour → adaptation to typist
 - Further sensors → adaptation to context
- **Prediction/Decoding:**
 - Single touch + language context → current key
 - Touch sequences + language context → current word/sentence
 - Language only → next word(s)

Notebook Example 2

Sequence decoding

References (Part 1)

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Further Reading:

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Further Reading:

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