

Lecture 8 Text Entry on Mobile Devices -2

Xiaojun Bi

Stony Brook University

xiaojun@cs.stonybrook.edu

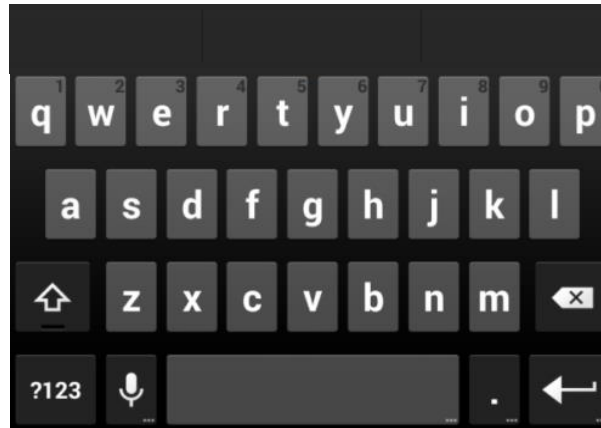
10/3/2019

Smart Touch Keyboard

Typed Word

Keyboard Output

agsim



again

quivj



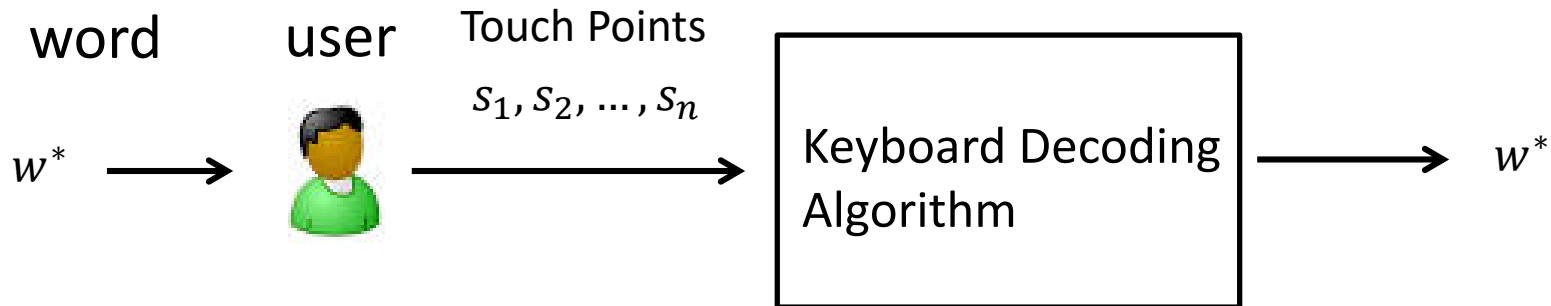
quick

fav



favorite

Text Entry Decoding Algorithm



$$W^* = \arg \max_{W \in L} P(W|S).$$

From Bayes' rule: $P(W|S) = \frac{P(S|W)P(W)}{P(S)}.$

As $P(S)$ is a constant across all the words, we have:

$$W^* = \arg \max_{W \in L} P(S|W)P(W).$$

- Spatial Model:

Assuming that W is comprised of n letters: $c_1, c_2, c_3, \dots, c_n$, S has n touch points, and each tap is independent, we have:

$$P(S|W) = \prod_{i=1}^n P(s_i|c_i).$$

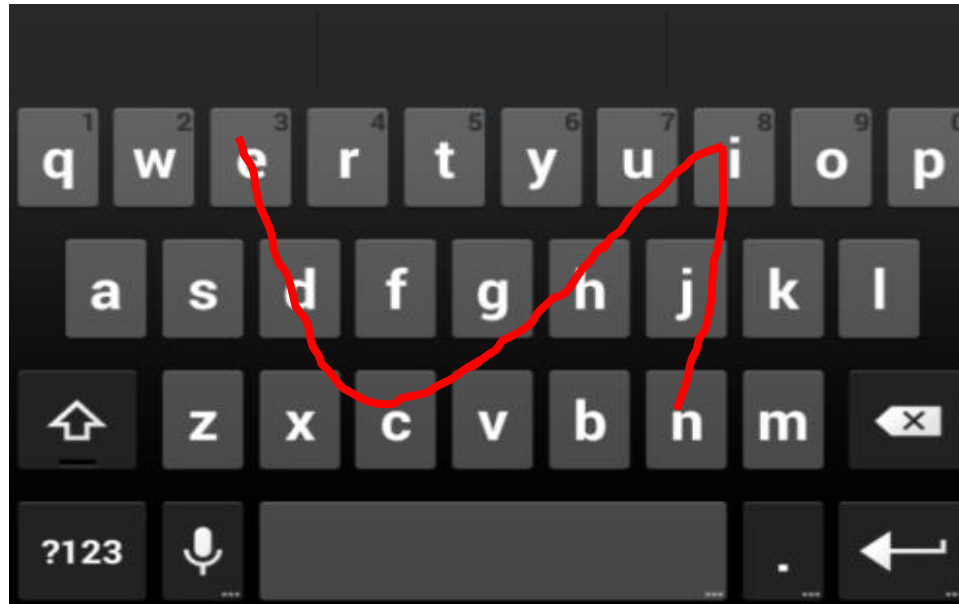
$$P(s_i|c_i) = \frac{1}{2\pi\sigma_{i_x}\sigma_{i_y}\sqrt{1-\rho_i^2}} \exp\left[-\frac{z}{2(1-\rho_i^2)}\right] \quad \text{Gaussian Distribution}$$

- Language Model:

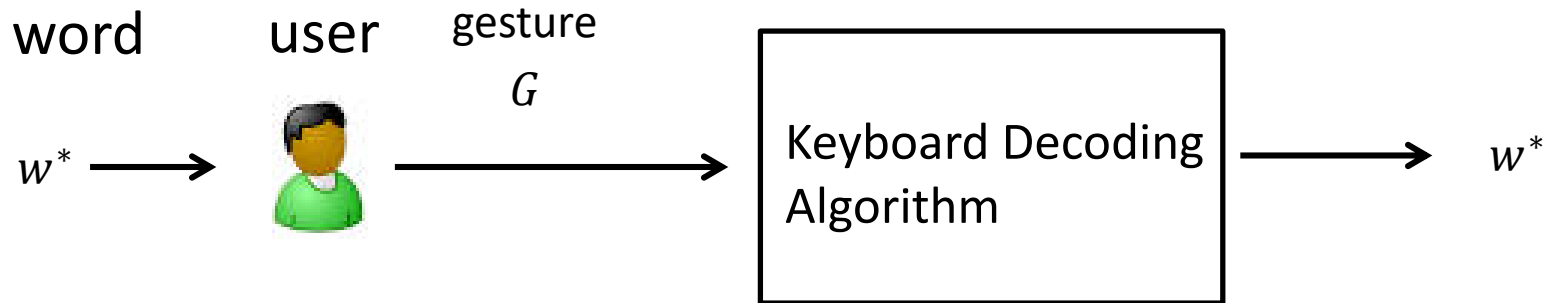
$P(W)$ is obtained from a Language Model (LM)

Gesture Keyboard

Entering *nice*



Gesture Decoder



$$W^* = \operatorname{argmax}_w P(W|G) = \operatorname{argmax}_w \frac{P(G|W)P(W)}{P(G)}$$

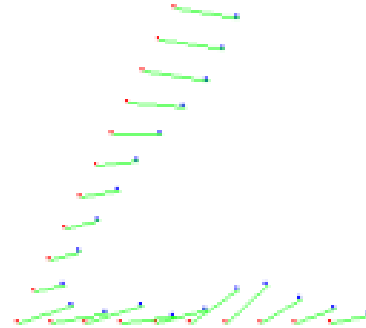
$$W^* = \operatorname{argmax}_w P(G|W)P(W)$$

How to calculate $P(G|W)$?

SHARK² Algorithm

- Location Recognition Channel

$$x_s = \frac{1}{N} \sum_{i=1}^N \|u_i - t_i\|_2$$



- Shape Matching Channel

Per-Ola Kristensson and Shumin Zhai. 2004. SHARK²: a large vocabulary shorthand writing system for pen-based computers. In *UIST '04*. 43-52.

Gesture Keyboard



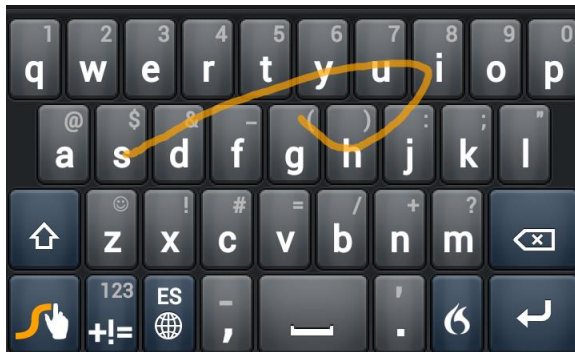
ShapeWriter



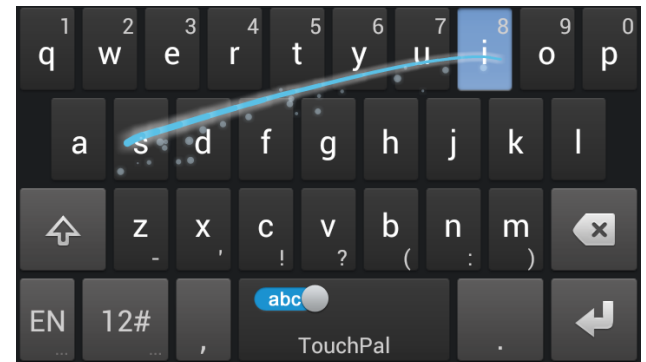
Android

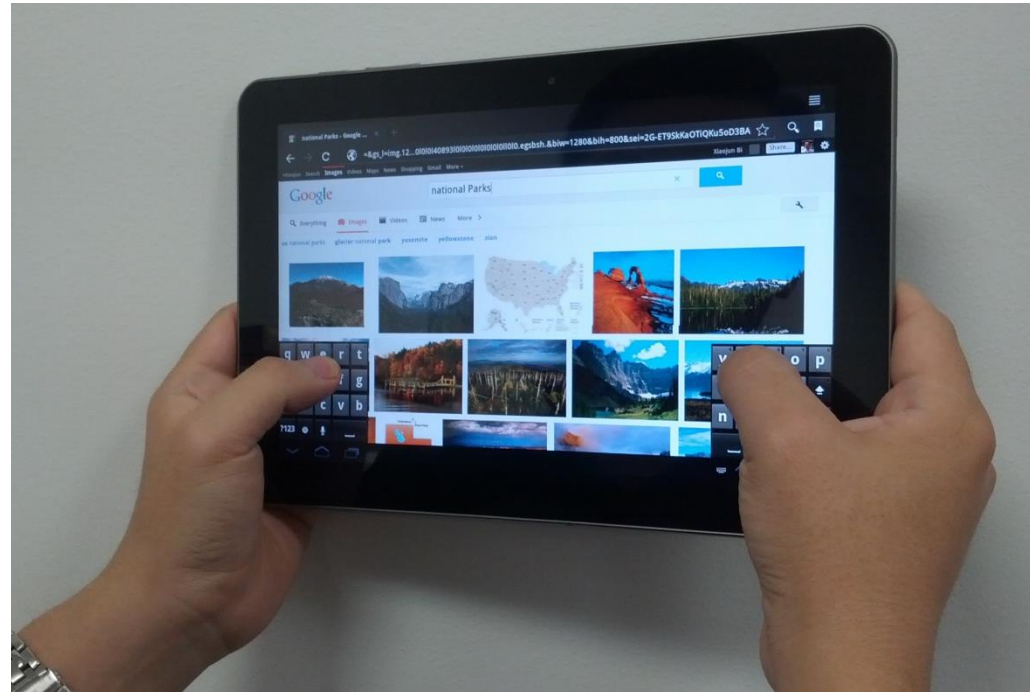
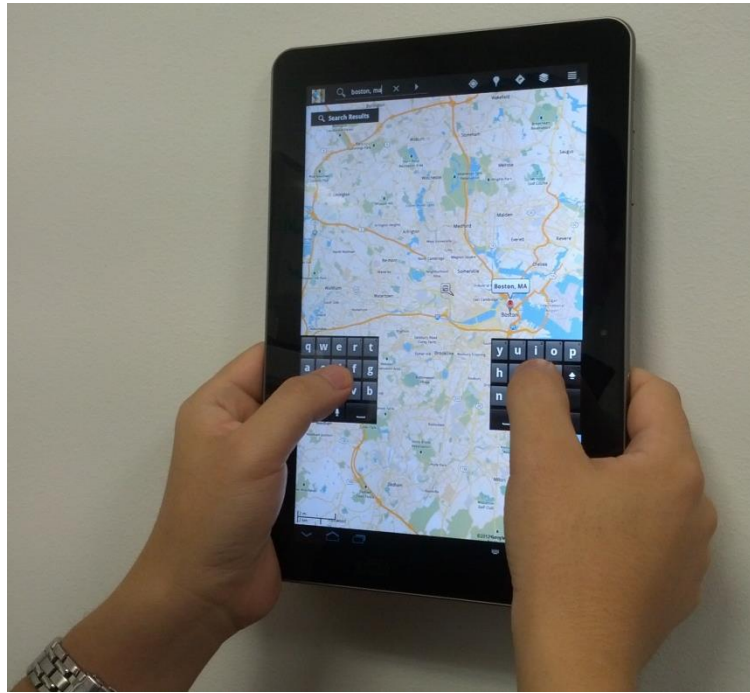


Swype



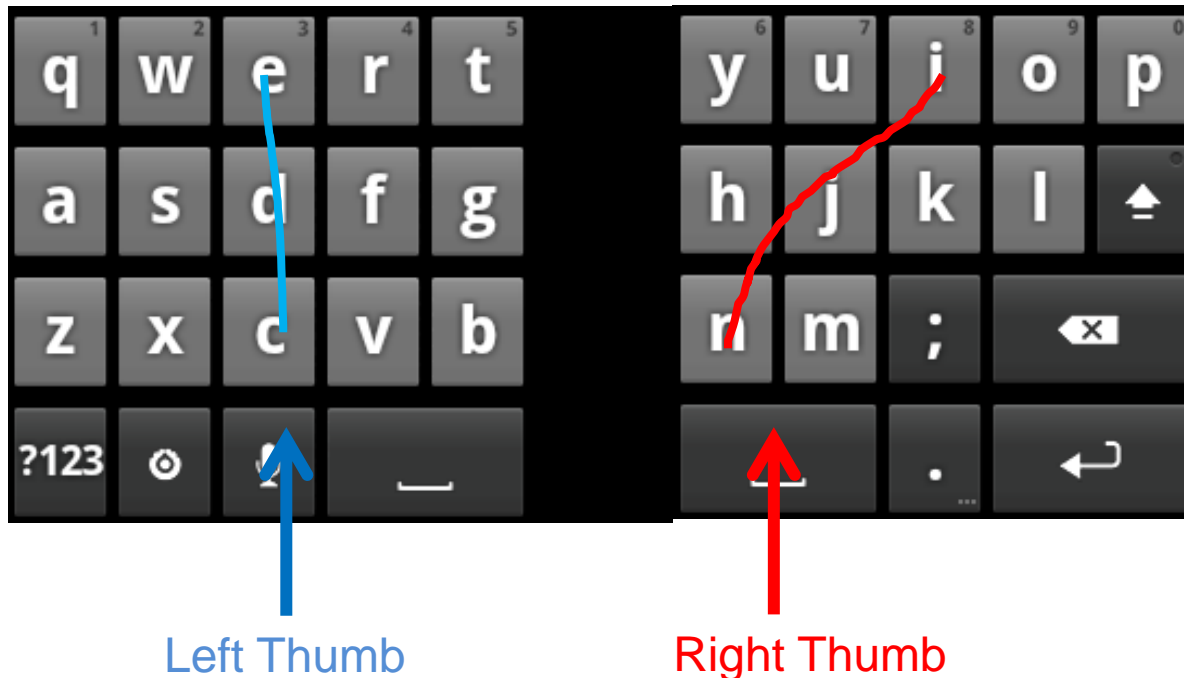
TouchPal





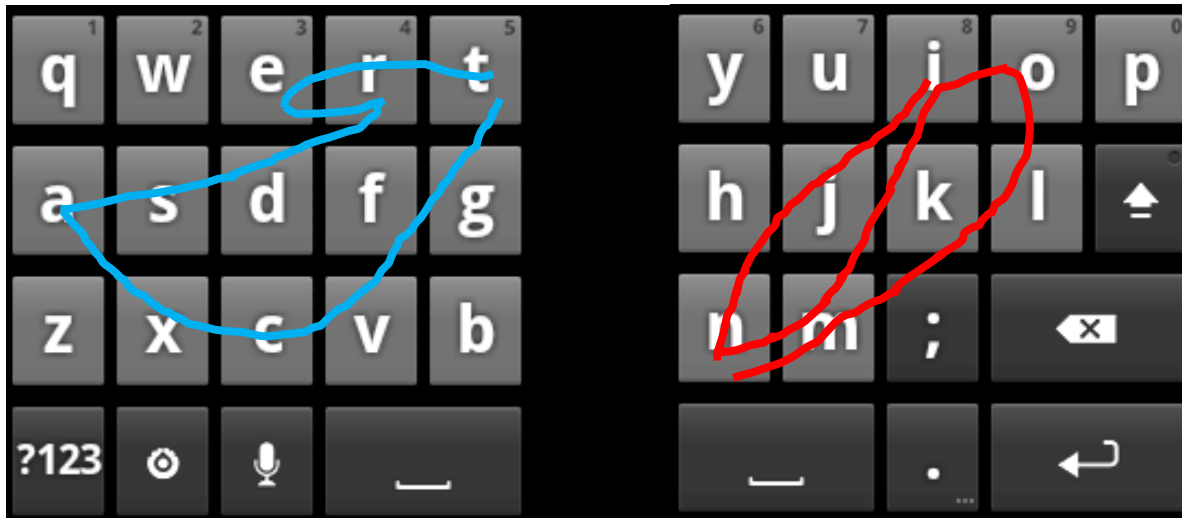
Bimanual Gesture Typing

Entering *nice*

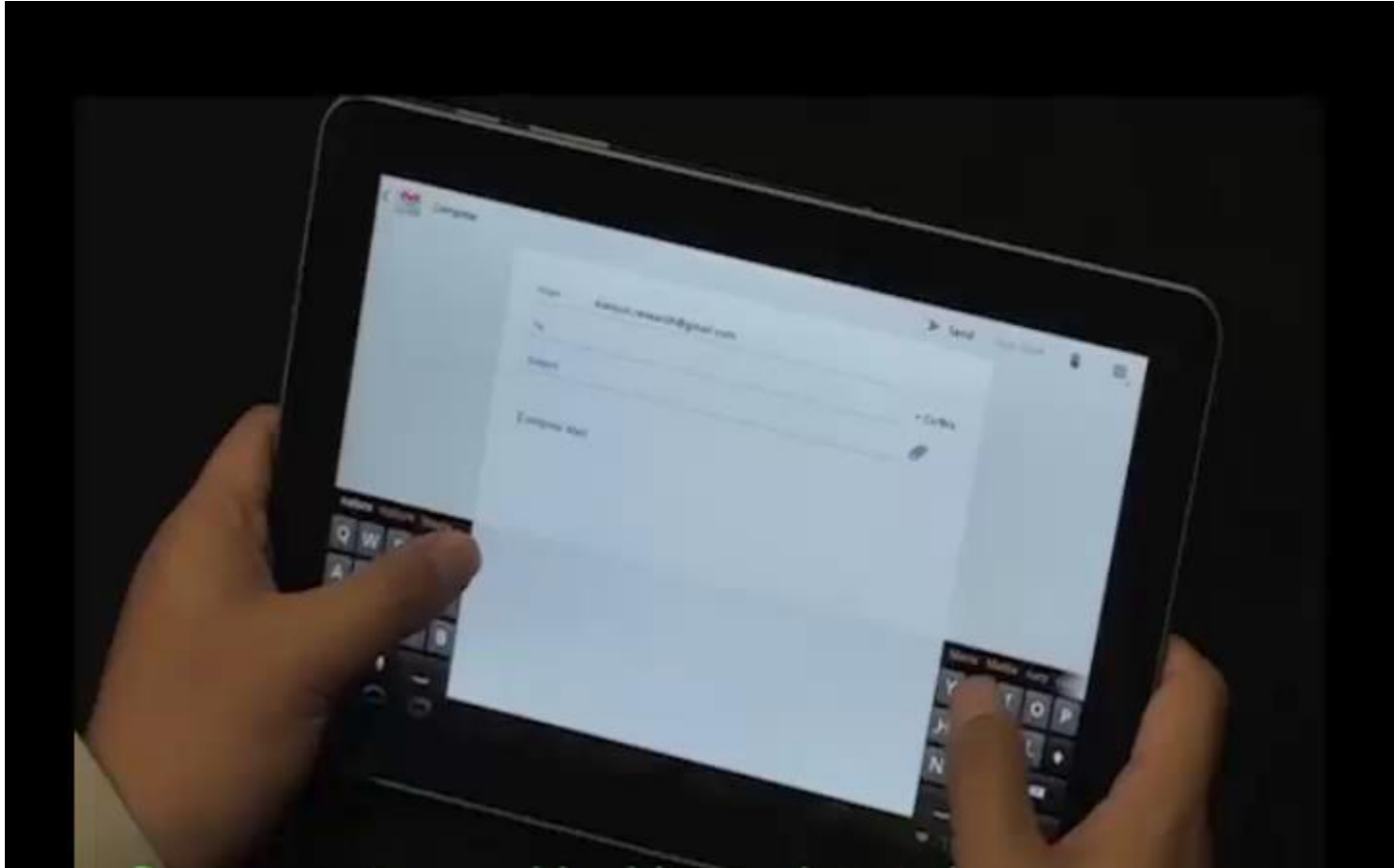


Bimanual Gesture Typing

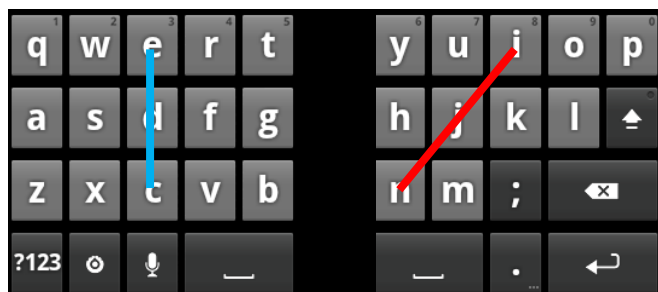
Entering *interaction*



Bimanual Gesture Typing



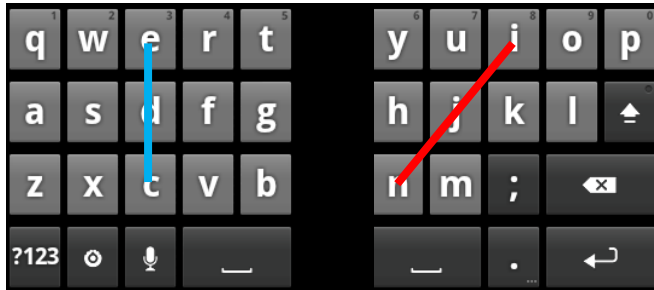
Perfect Templates



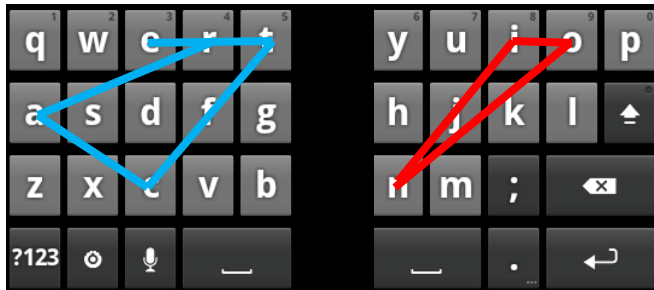
ce ni

nice

Perfect Templates

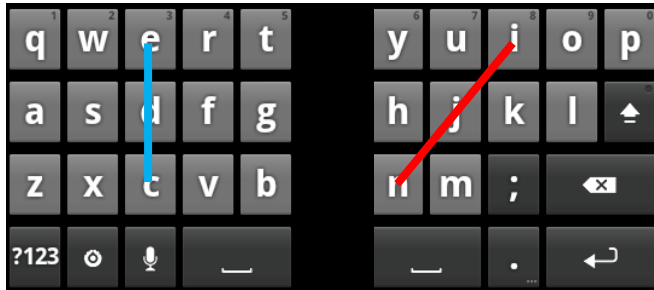


ce ni
nice

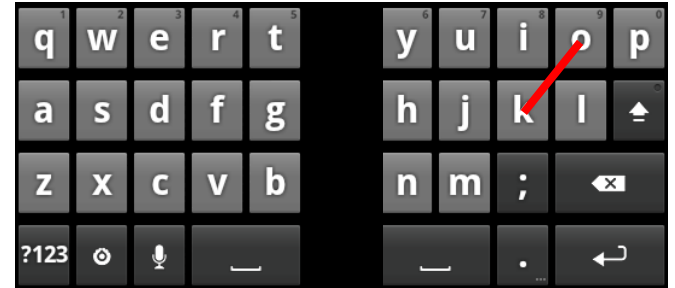


teract inion
interaction

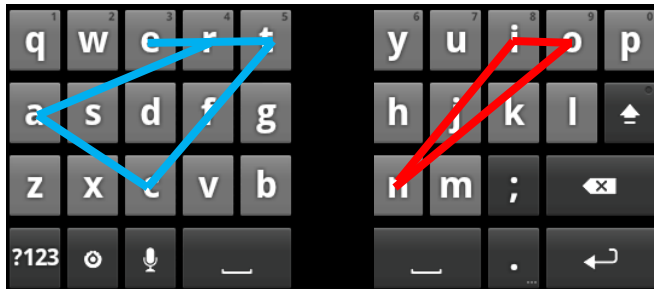
Perfect Templates



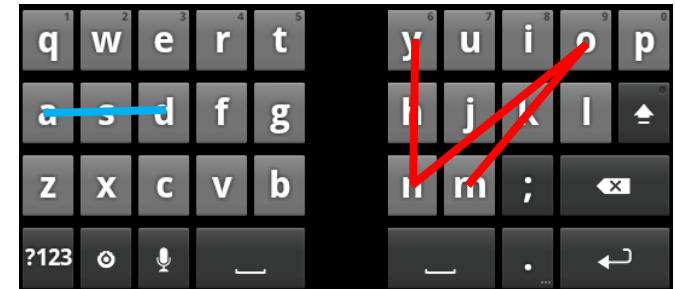
ce ← nice → ni



ok → ok

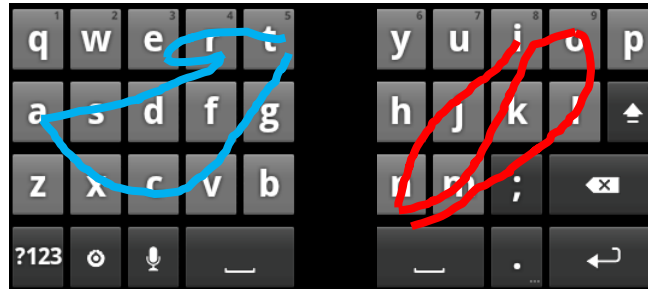


teract ← inion → interaction

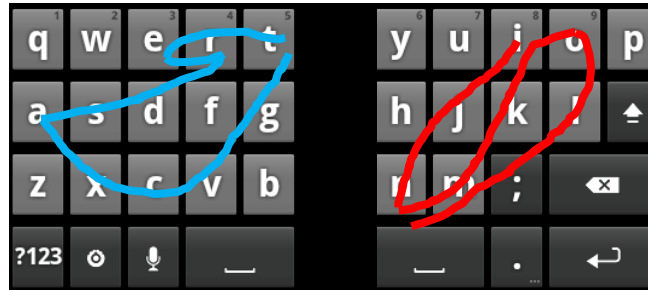


da ← mony → Monday

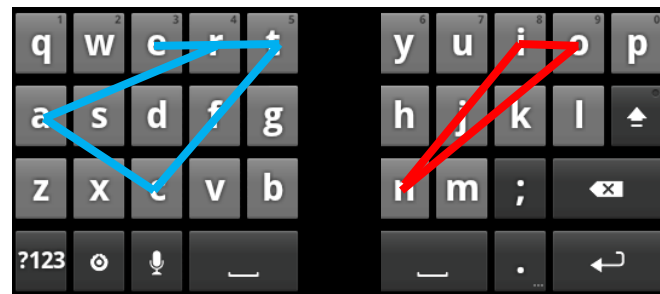
Bimanual Gesture Recognition



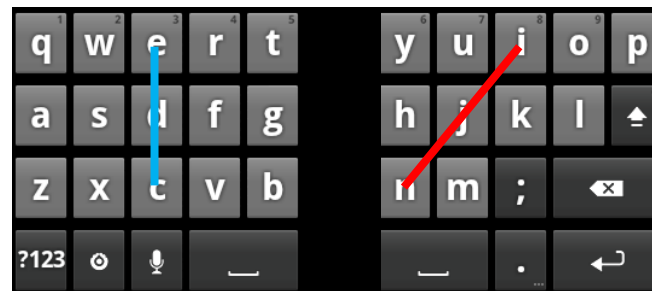
Bimanual Gesture Recognition



Best Match



interaction

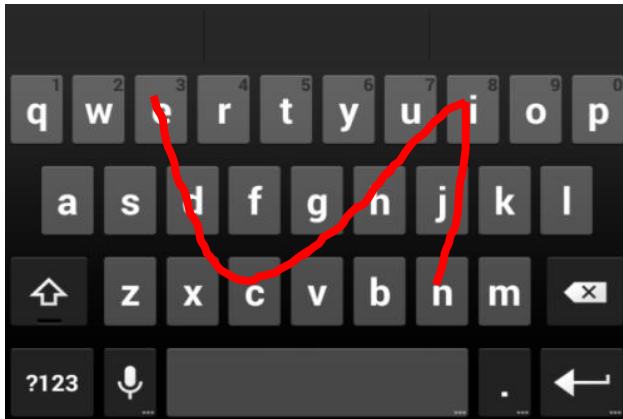


nice

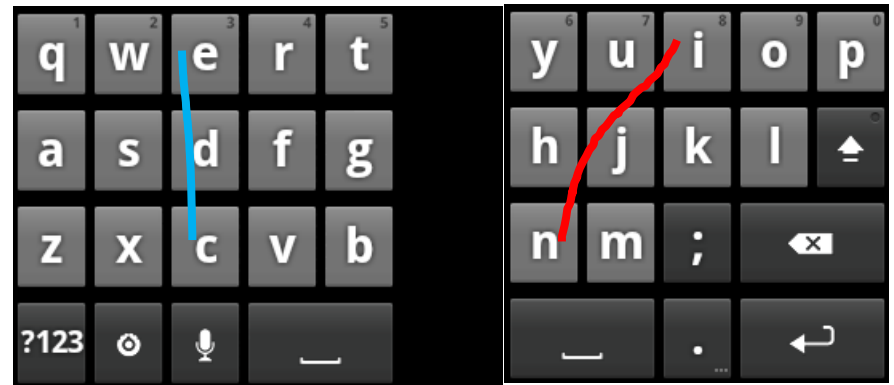
...

Entering *nice*

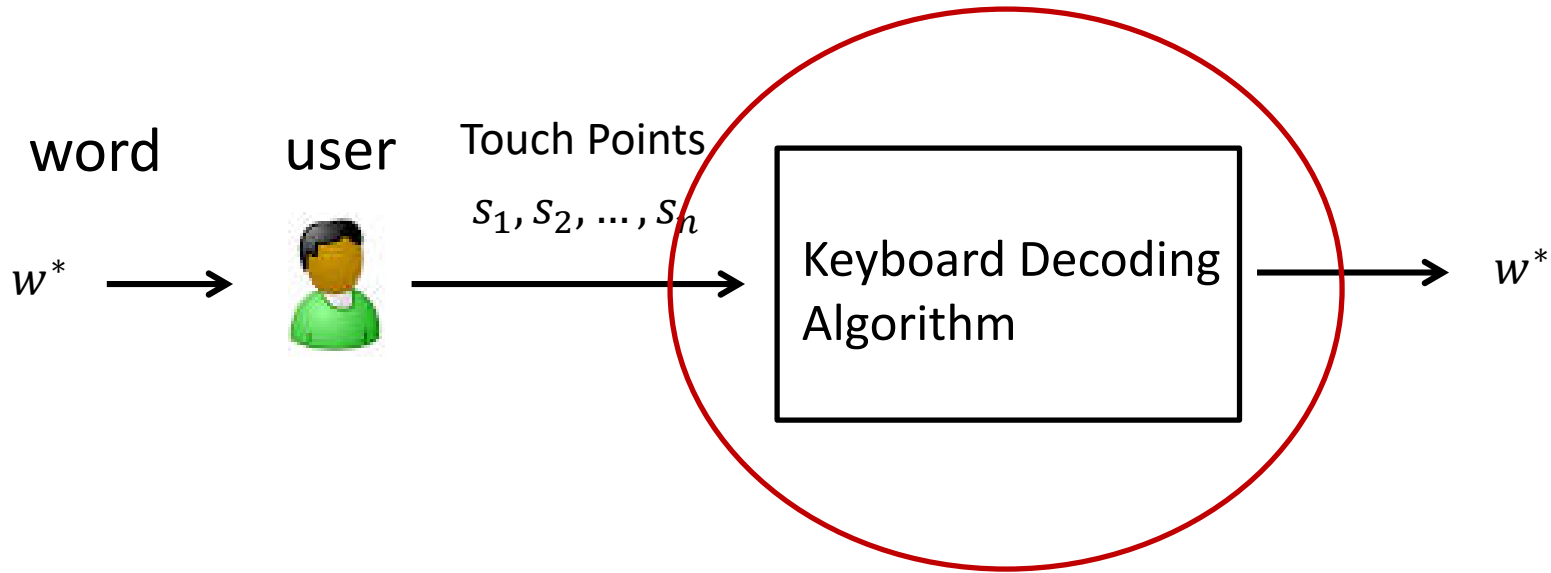
Unimanual Gesture



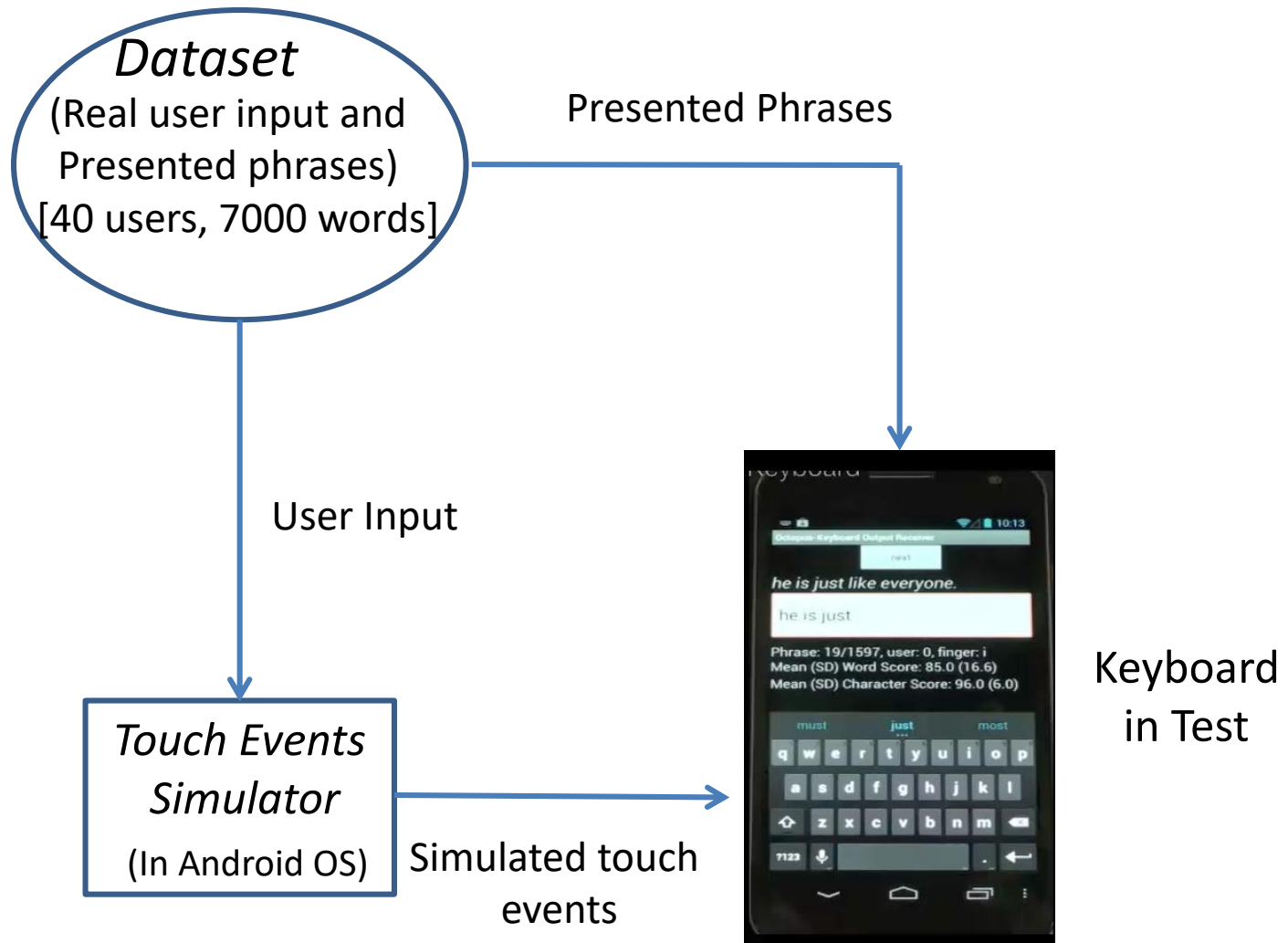
Bimanual Gesture

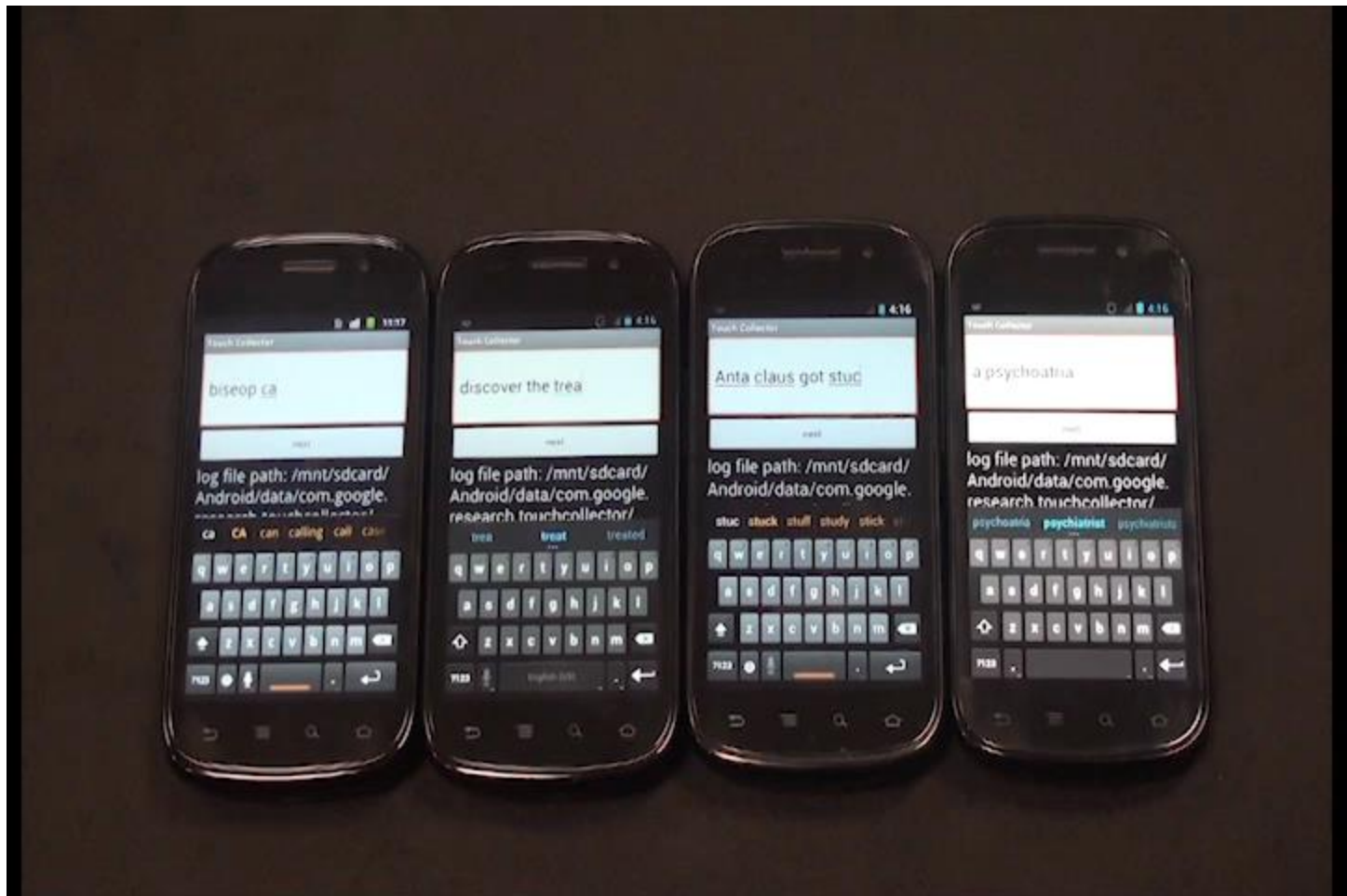


Optimization and Evaluation of Decoding Algorithm



Remulation: **Replicating** prior user study data with **simulation**.



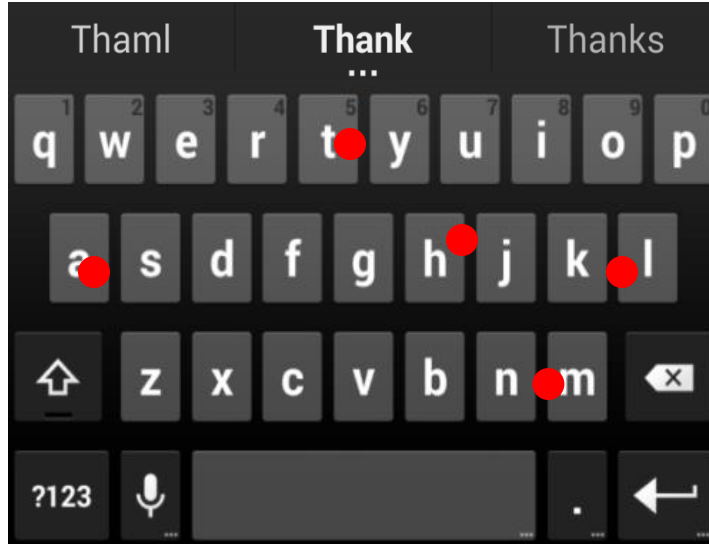


[Bi, Azenkot, Partridge, Zhai. *Octopus: Evaluating Touchscreen Keyboard Correction and Recognition Algorithms via Remulation*. ACM CHI2013]

Smart Touchscreen Keyboard

Correction

Thaml → Thank



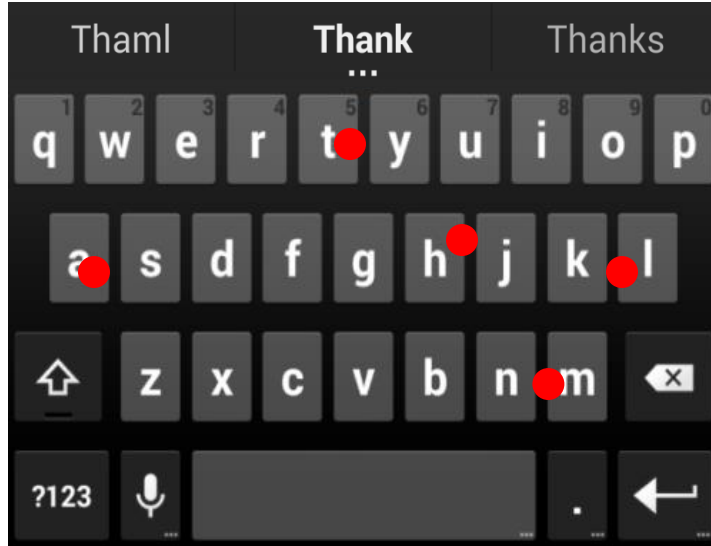
$$\text{Success Rate}(W) = \frac{\text{Correct Words}}{\text{Total Words}}$$

[Bi, Ouyang, Zhai. *Both Complete and Correct? Multi-Objective Optimization of Touchscreen Keyboard*. ACM CHI2014]

Smart Touchscreen Keyboard

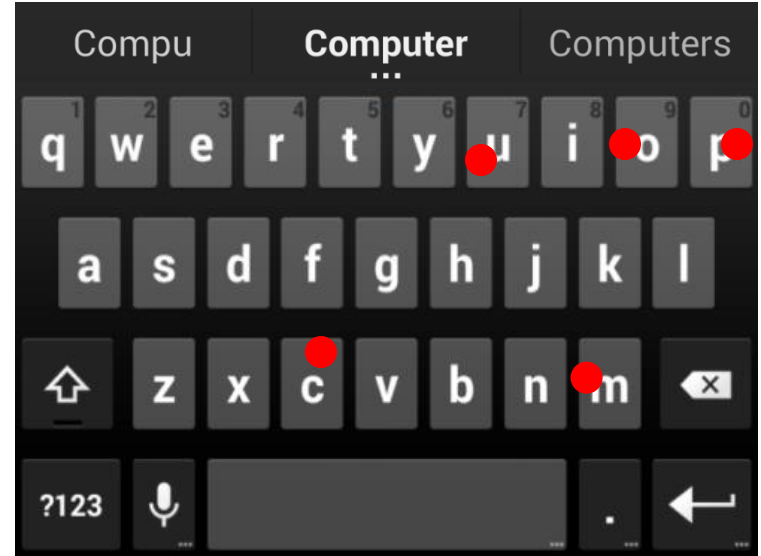
Correction

Thaml → Thank



Completion

Compu → Computer



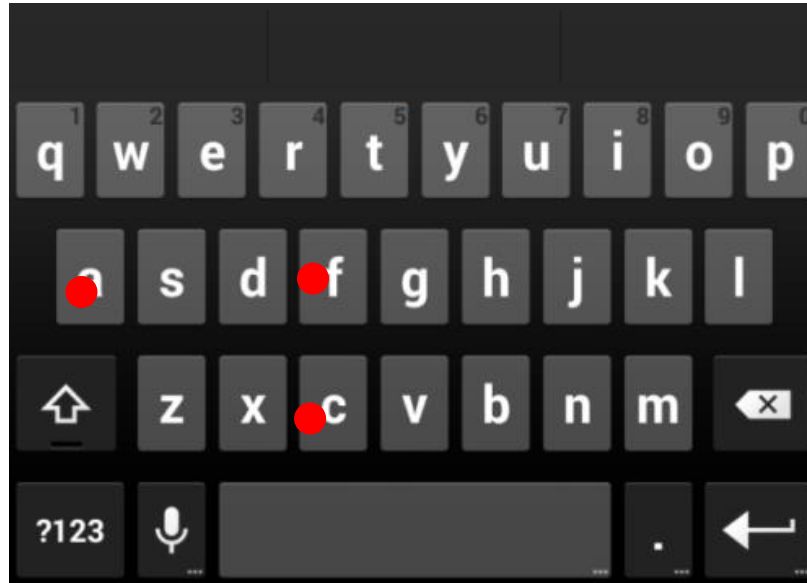
$$\text{Success Rate}(W) = \frac{\text{Correct Words}}{\text{Total Words}}$$

$$\text{Keystroke Saving}(S) = \frac{\text{Saved Keystrokes}}{\text{Maximum Keystrokes}}$$

[Bi, Ouyang, Zhai. *Both Complete and Correct? Multi-Objective Optimization of Touchscreen Keyboard*. ACM CHI2014]

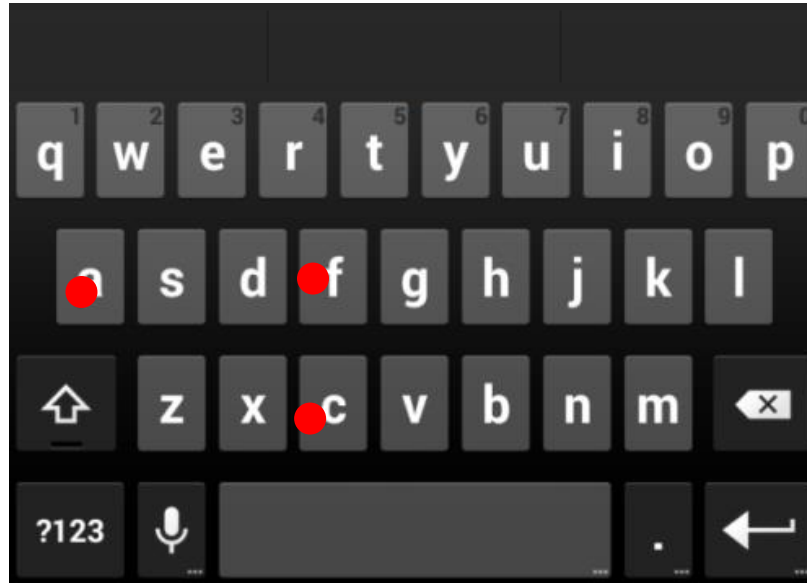
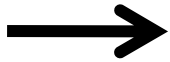
Correction vs. Completion

fac →



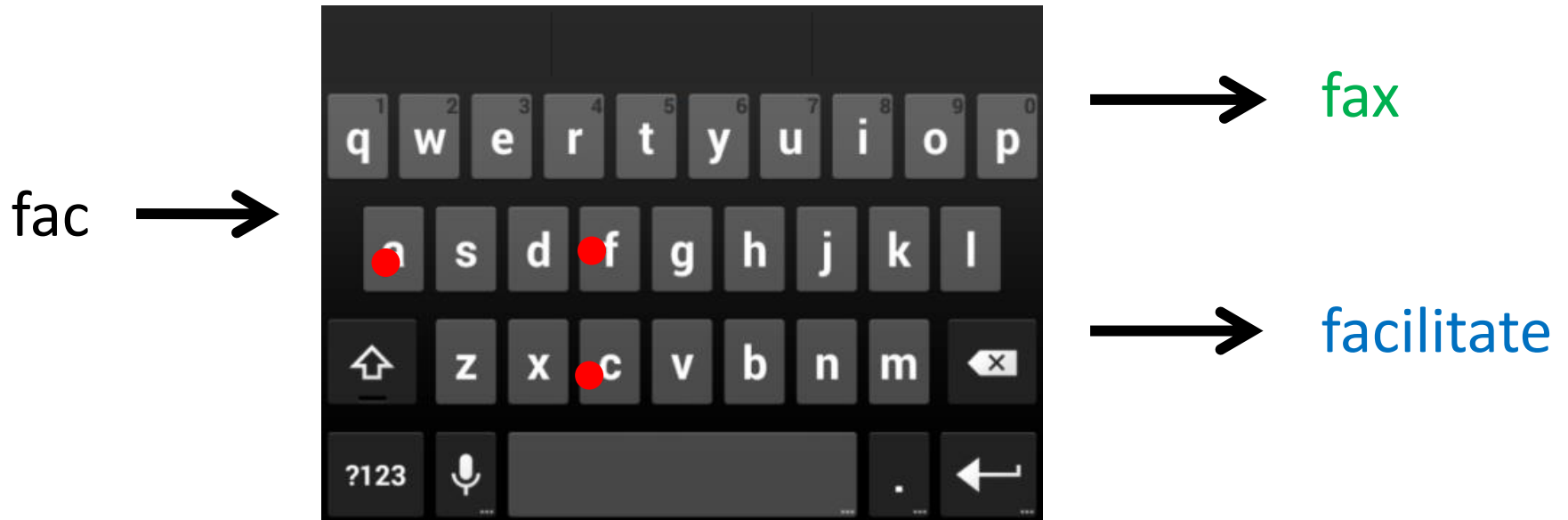
Correction vs. Completion

fac

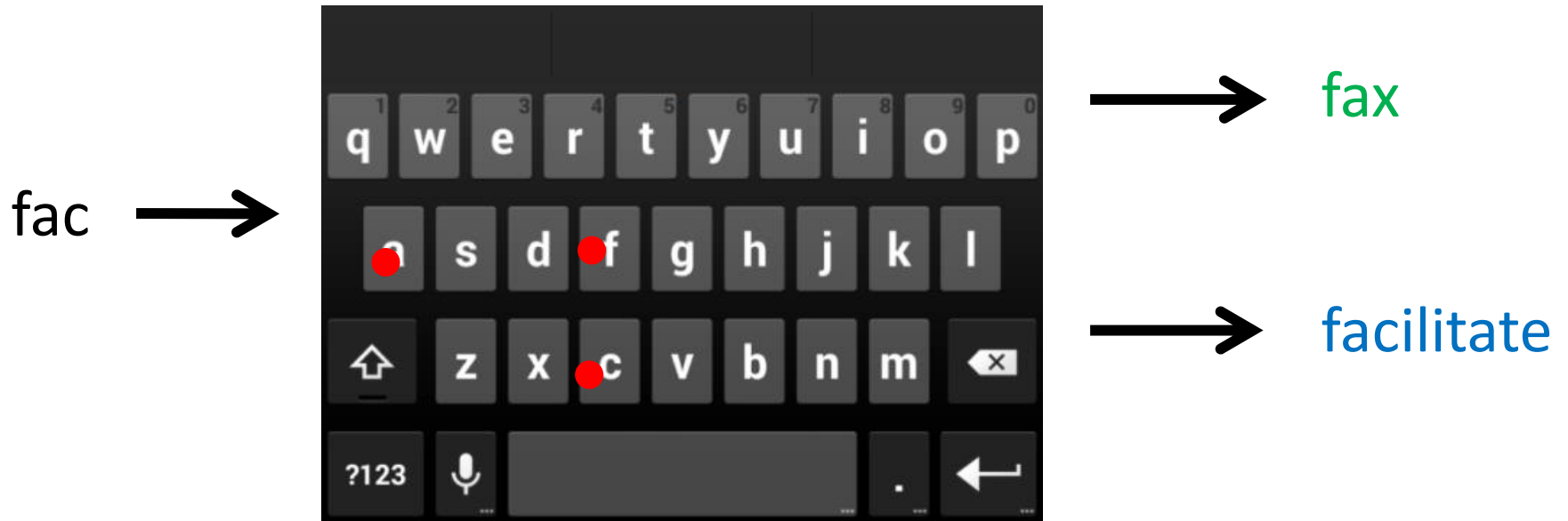


fax

Correction vs. Completion



Correction vs. Completion



Relationship between Correction and Completion

?

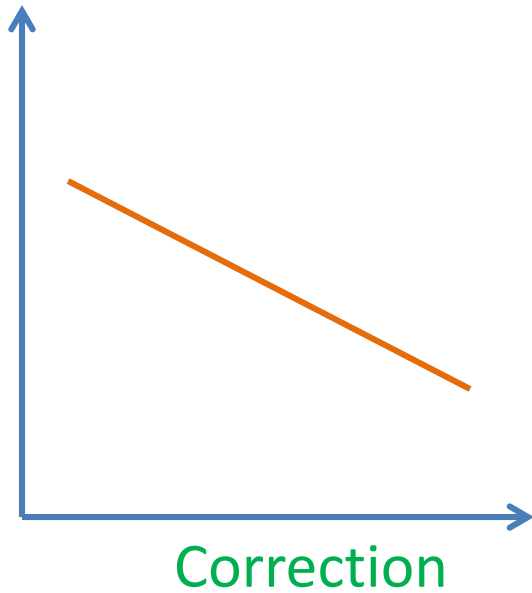
Correction vs. Completion

Possible Relationships

Correction vs. Completion

Possible Relationships

Completion



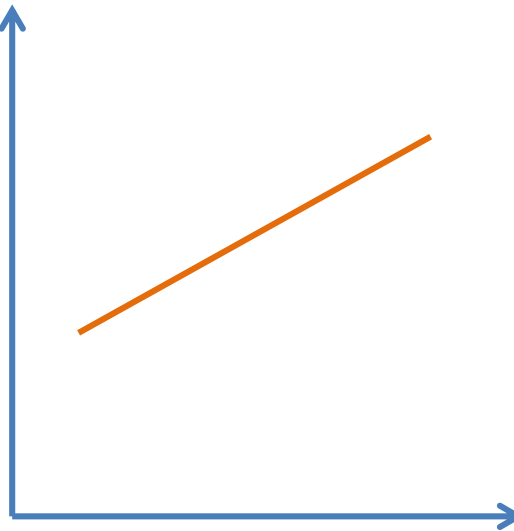
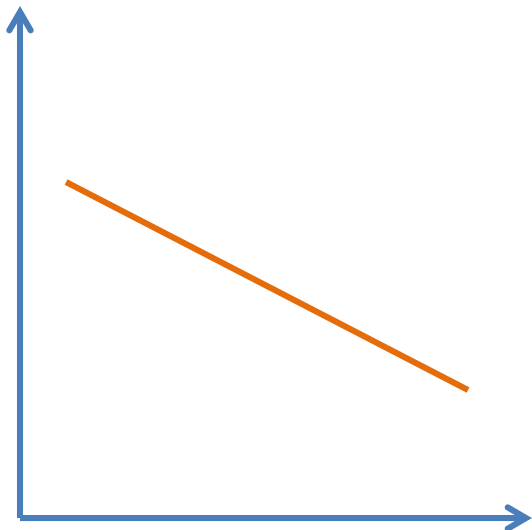
[Bi, Ouyang, Zhai. *Both Complete and Correct? Multi-Objective Optimization of Touchscreen Keyboard*. ACM CHI2014]

Correction vs. Completion

Possible Relationships

Completion

Completion



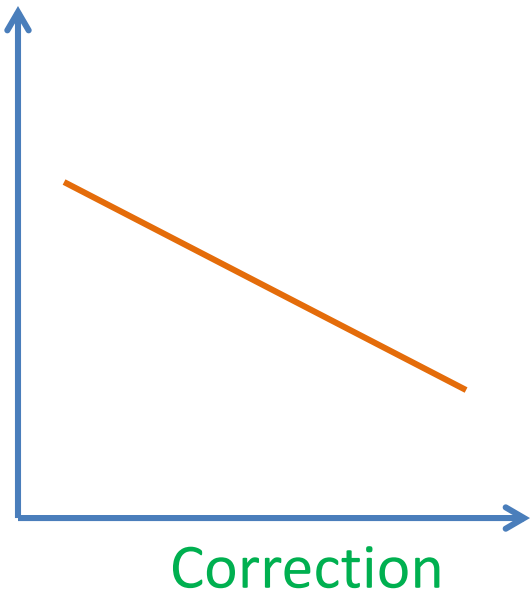
Correction

Correction

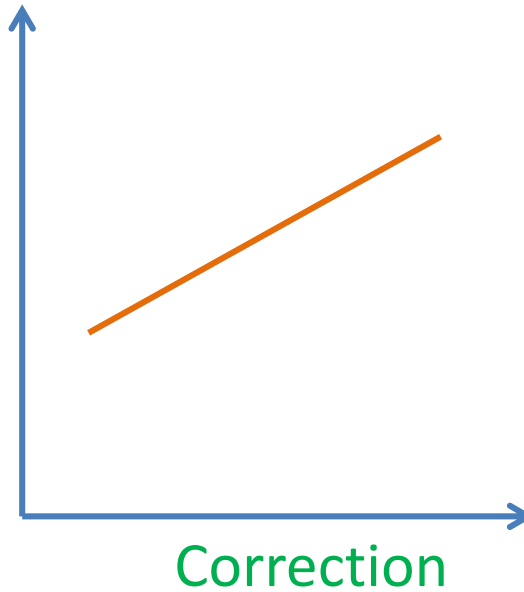
Correction vs. Completion

Possible Relationships

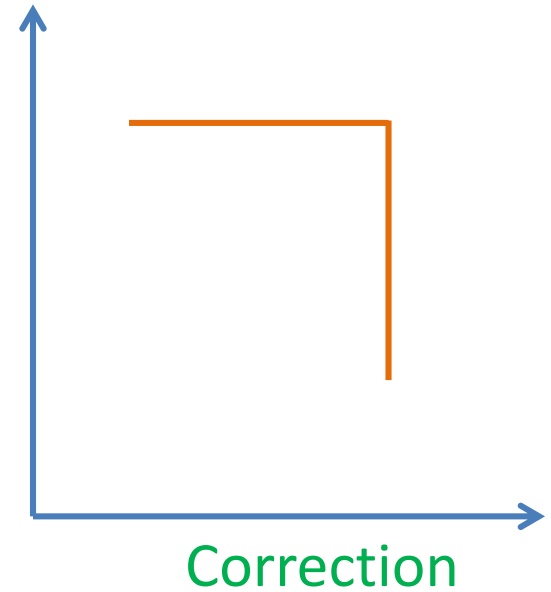
Completion



Completion



Completion

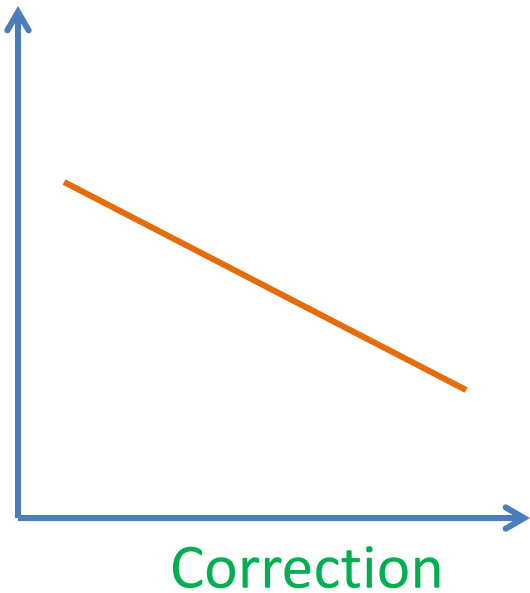


Correction vs. Completion

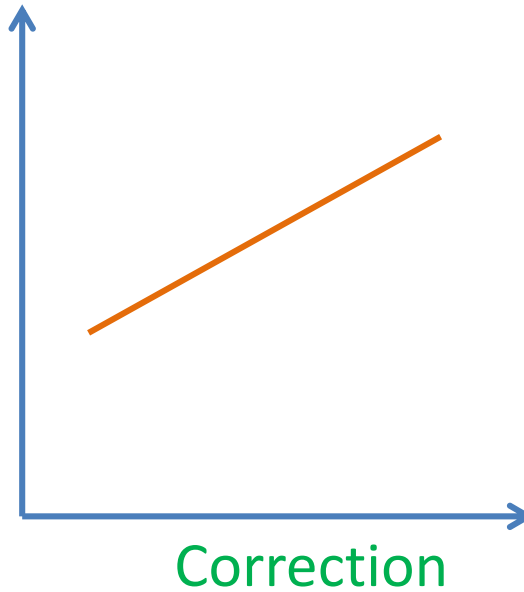
Possible Relationships

?

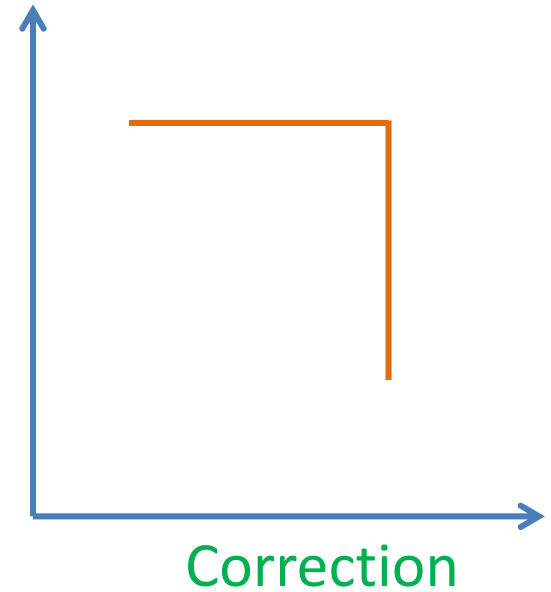
Completion



Completion

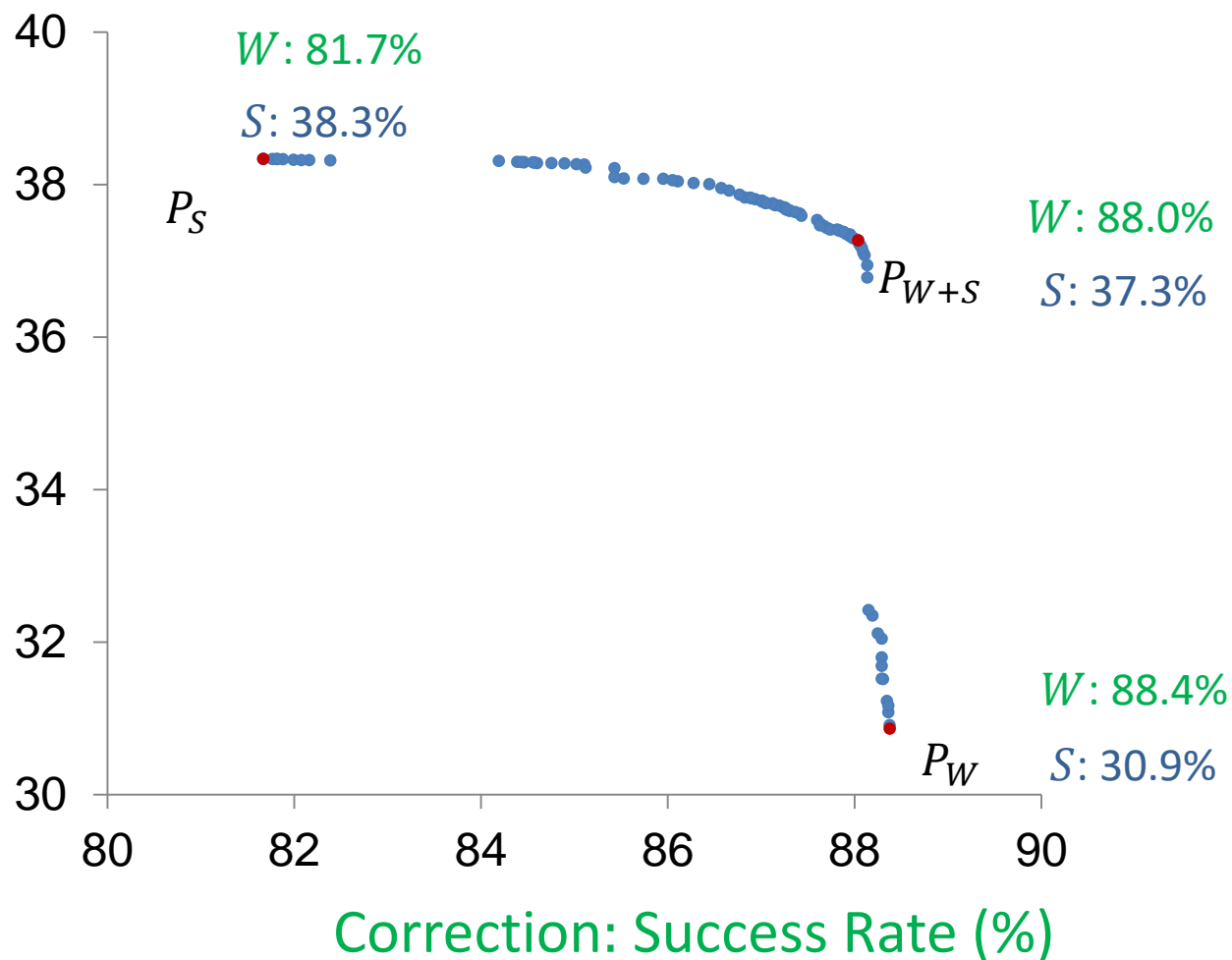


Completion



Pareto Frontier

Completion: Keystroke Saving (%)



[Bi, Ouyang, Zhai. *Both Complete and Correct? Multi-Objective Optimization of Touchscreen Keyboard*. ACM CHI2014]

Outline

- Smart Touch Keyboard
- Gesture Typing
- Optimizing Keyboard Layouts

Qwerty Layout

Q	W	E	R	T	Y	U	I	O	P
A	S	D	F	G	H	J	K	L	
	Z	X	C	V	B	N	M		

Qwerty is inefficient for one finger typing.

Optimization Objective Function

- Fitts' Law (Fitts 1954):

$$MT_{ij} = a + b \log_2 \left(\frac{D_{ij}}{W} + 1 \right)$$

MT_{ij} : Movement Time from Key i to Key j

D_{ij} : Distance from Key i to Key j

W : Key Width

Optimization Objective Function

- Fitts' Law (Fitts 1954):

$$MT_{ij} = a + b \log_2 \left(\frac{D_{ij}}{W} + 1 \right)$$

MT_{ij} : Movement Time from Key i to Key j

D_{ij} : Distance from Key i to Key j

W : Key Width

- Average time of typing a letter:

$$t = a + b \sum_i^{26} \sum_j^{26} P_{ij} \log_2 \left(\frac{D_{ij}}{W} + 1 \right)$$

P_{ij} : Frequency of an ordered letter pair i, j

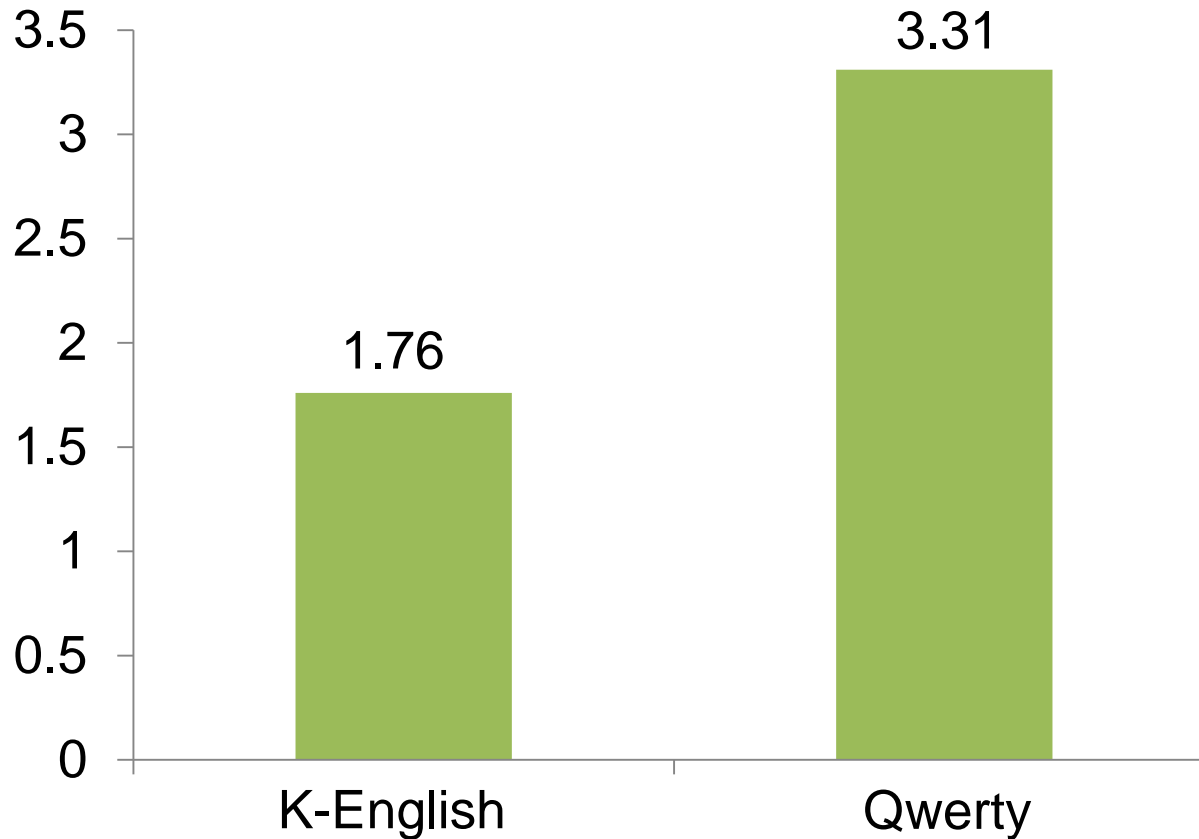
Layout Optimized for English

K-English

Z	J	D	G	K	
Y	L	N	I	C	
F	O	A	T	H	W
B	U	R	E	S	
Q	P	M	V	X	

Average Finger Travel Distance

Key Width



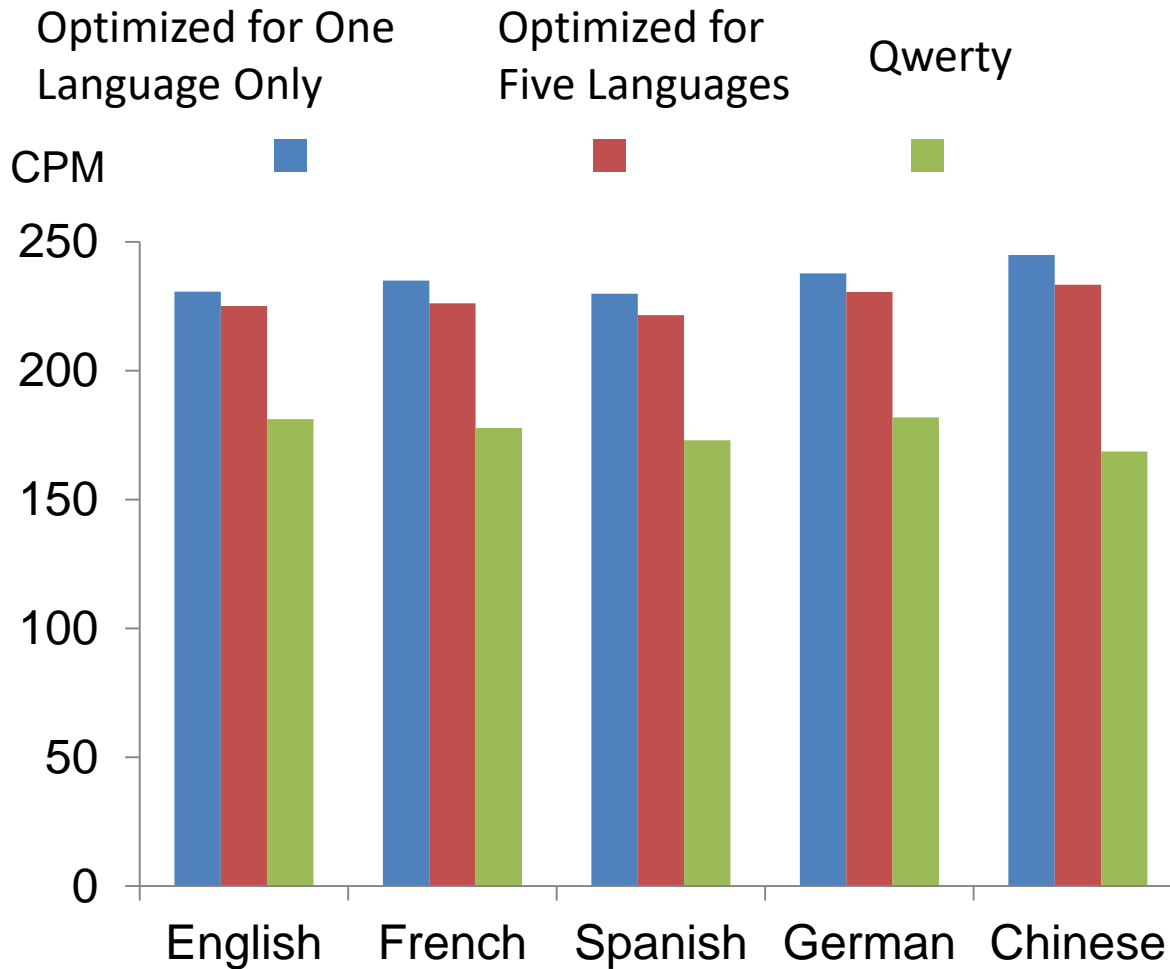
Layout Optimized for Five Languages

English, French, German, Spanish, and Chinese *Pinyin*

K5

	K	J	Z	X	
	F	C	H	T	W
Q	U	O	I	S	P
Y	M	A	N	E	R
	B	L	G	D	V

Typing Speed



Optimized Layout for Gesture Typing

Qwerty

Q	W	E	R	T	Y	U	I	O	P
A	S	D	F	G	H	J	K	L	
	Z	X	C	V	B	N	M		

Optimized Layout for Gesture Typing

or vs. our

Qwerty

Q	W	E	R	T	Y	U	I	O	P
A	S	D	F	G	H	J	K	L	
	Z	X	C	V	B	N	M		

GK-T (Speed, Clarity, Similarity to Qwerty)

Q	D	W	S	O	I	Y	U	J	P
Z	R	F	A	T	N	G	K	L	
	C	E	X	H	V	M	B		

Optimized Layout for Gesture Typing

or vs. our

Qwerty

Q	W	E	R	T	Y	U	I	O	P
A	S	D	F	G	H	J	K	L	
	Z	X	C	V	B	N	M		

GK-T (Speed, Clarity, Similarity to Qwerty)

Q	D	W	S	O	I	Y	U	J	P
Z	R	F	A	T	N	G	K	L	
	C	E	X	H	V	M	B		

COMPASS: Rotational Keyboard on Non-Touch Smartwatches



Xin Yi, Chun Yu, Weijie Xu, Xiaojun Bi, and Yuanchun Shi. 2017. COMPASS: Rotational Keyboard on Non-Touch Smartwatches. In Proceedings of the 2017 CHI Conference on Human Factors in Computing Systems (CHI '17). ACM, New York, NY, USA, 705-715.



Smartwatches are popular

But what if...



COMPASS



- Circular Keyboard
- Multi-cursor paradigm
- Bezel rotation
- Physical button
- Dynamic Cursor Placement
- Flick to delete



Multi-Cursor Paradigm



1- Cursor
 $dis('M') = 11$



3-Cursors
 $dis('M') = 4$

Algorithm of COMPASS

Similar as an ambiguous keyboard

Candidate word

Users' input

$$P(W|I) \propto P(I|W) \times P(W)$$

$$P(I|W) = \begin{cases} 1 & \text{if } I \text{ matches with } W \\ 0 & \text{otherwise} \end{cases}$$

Top 15,000 words and
frequencies
in American National Corpus

Visual Cues



1. For each character c , calculate $P(c)$ as the **probability of it being the next character** given the input prefix.
2. Each key is **highlighted** according to $P(c)$.
3. Impossible characters are **dimmed**.

Cursor Adjustment



Locations of cursors **dynamically adjust** upon each character selection.

Choose the locations that would minimize the *Expected Next Rotation Distance*

$$ENRD = \sum_{c \in \chi} dis(c) \times P(c)$$

“Bad Case”

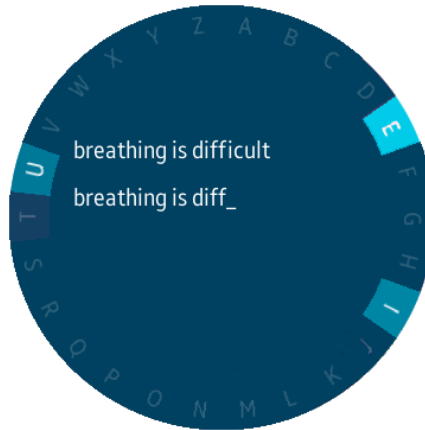


Naïve

Optimization

$$\begin{cases} dis(E) = 0 \\ dis(I) = 0 \\ dis(U) = 0 \end{cases}$$

#possible next key = 5

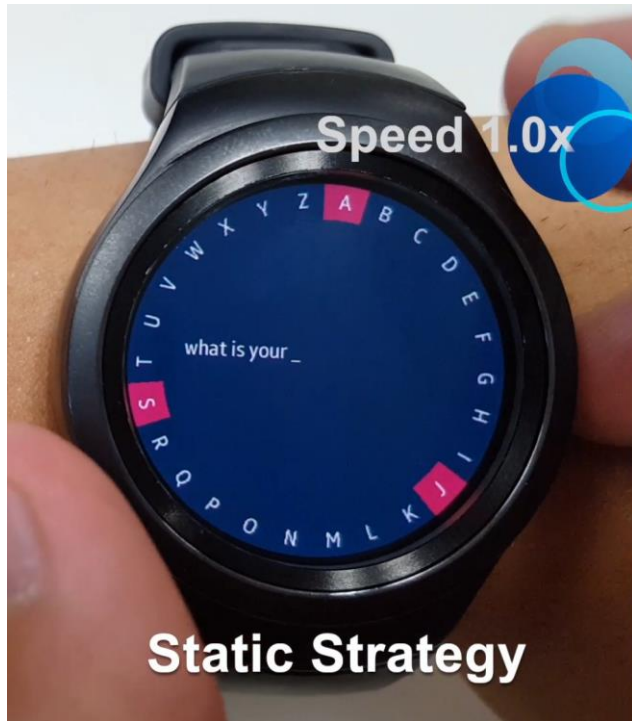


Restriction Applied

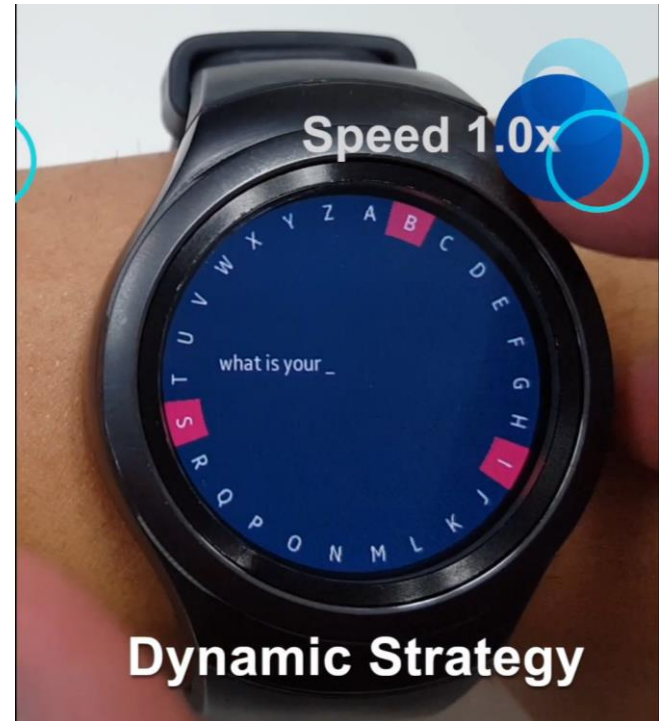
$$\begin{cases} dis(E) = 0 \\ dis(I) = 1 \\ dis(U) = 1 \end{cases}$$

#possible next key = 1

S-COMPASS & D-COMPASS



*S(Static)-
COMPASS*



*D(Dynamic)-
COMPASS*

Number of Cursors (N)

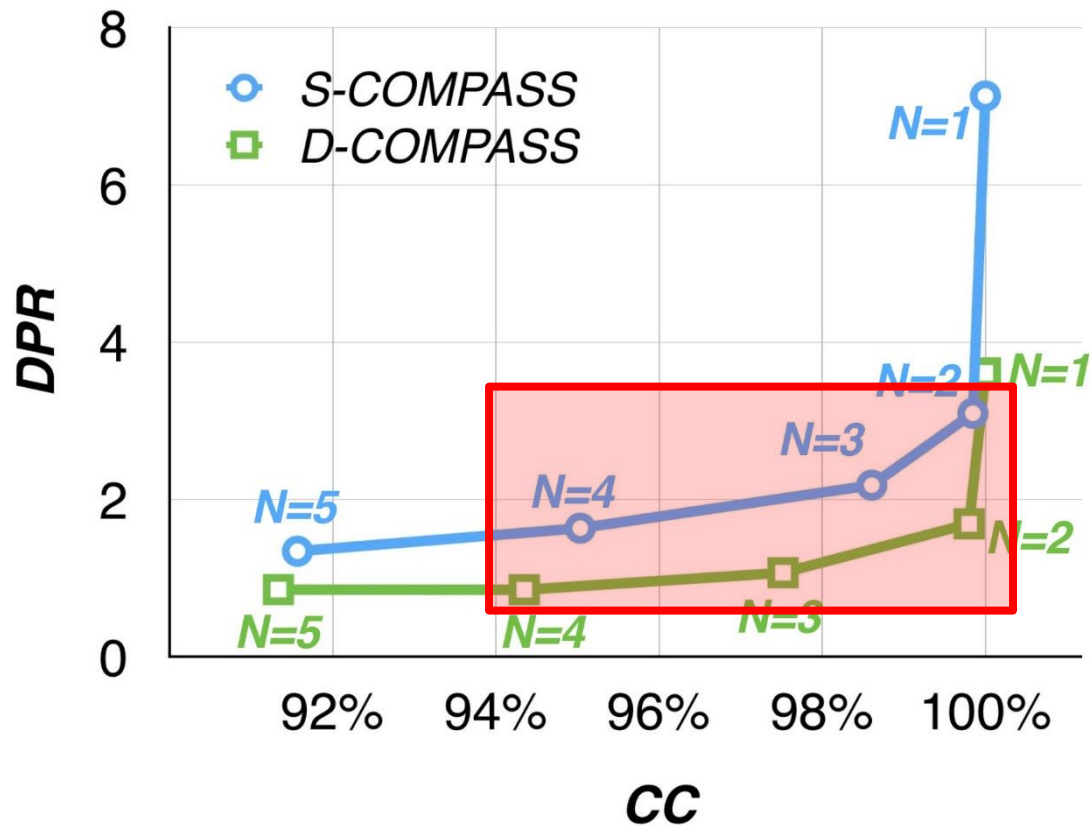


- Simulated all 15,000 words in the corpus
- Assumed perfect user input
- $1 \leq N \leq 5$

DPR (Distance Per Rotation):
Average **distance** of each
rotation

CC (Candidate Coverage):
Ratio of words that appear in
the **top-3 candidates** given
users' perfect input

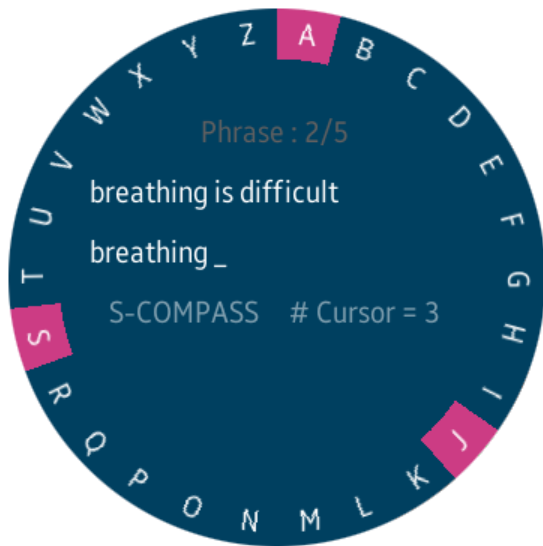
Simulation Result



DPR (Distance Per Rotation):
Average **distance** of each **rotation**

CC (Candidate Coverage):
Ratio of words that appear in the **top-3 candidates** given users' perfect input

User Study



12 participants

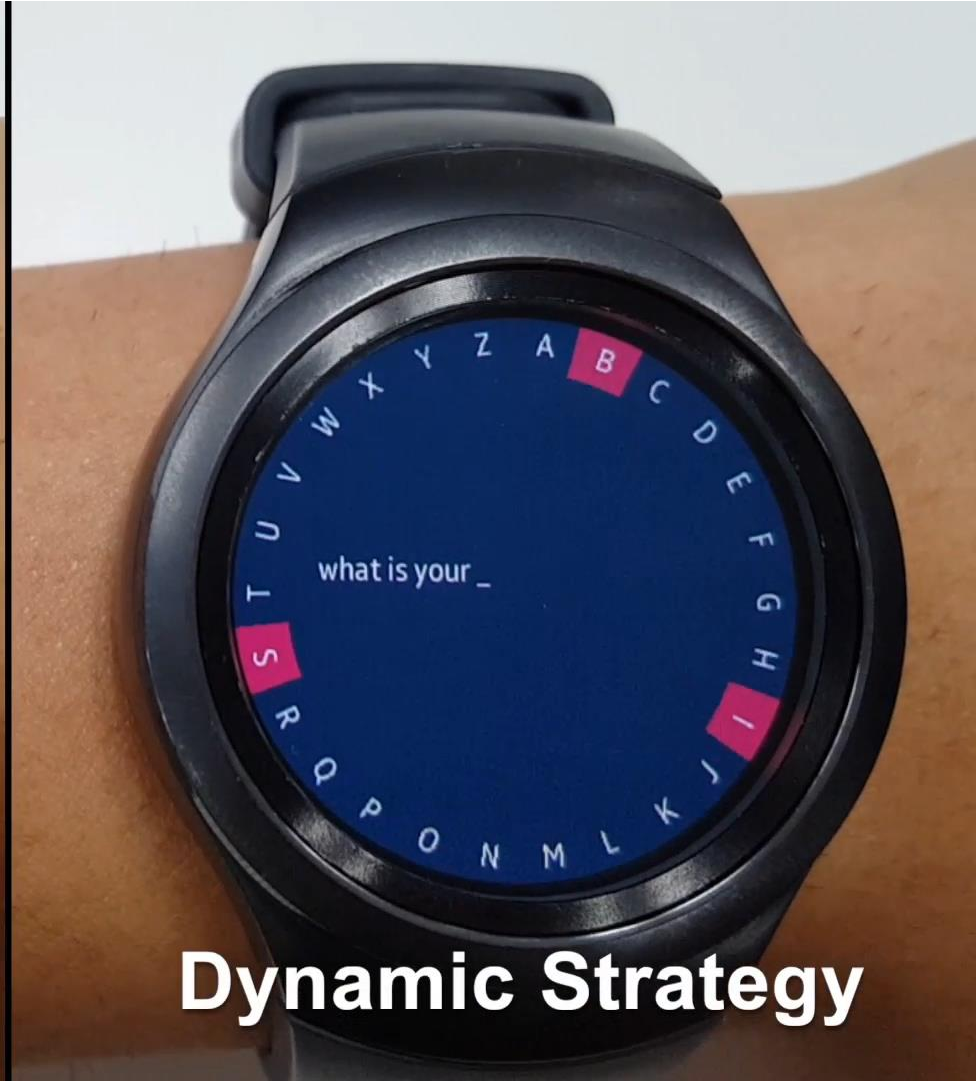
× 2 techniques (*S-COMPASS* and *D-COMPASS*)

× 3 $N(2, 3 \text{ and } 4)$

× 5 phrases

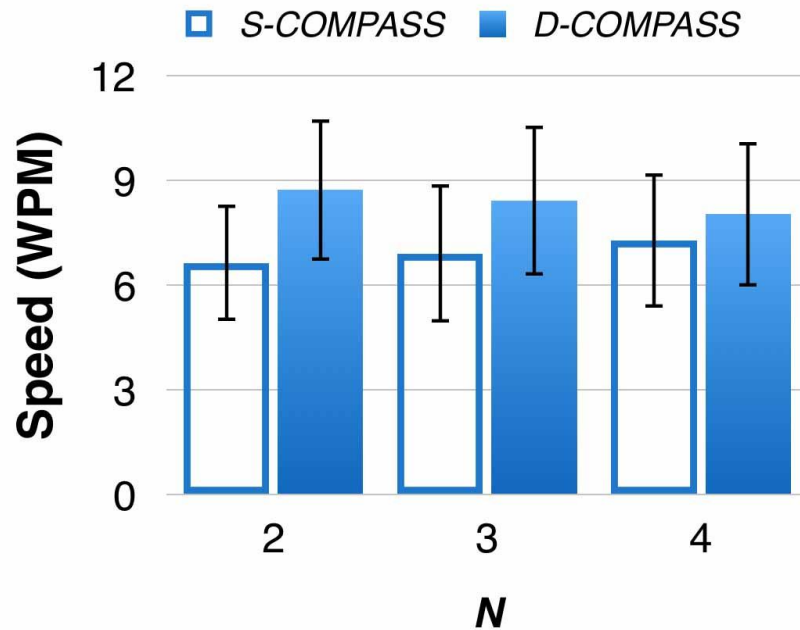


Static Strategy



Dynamic Strategy

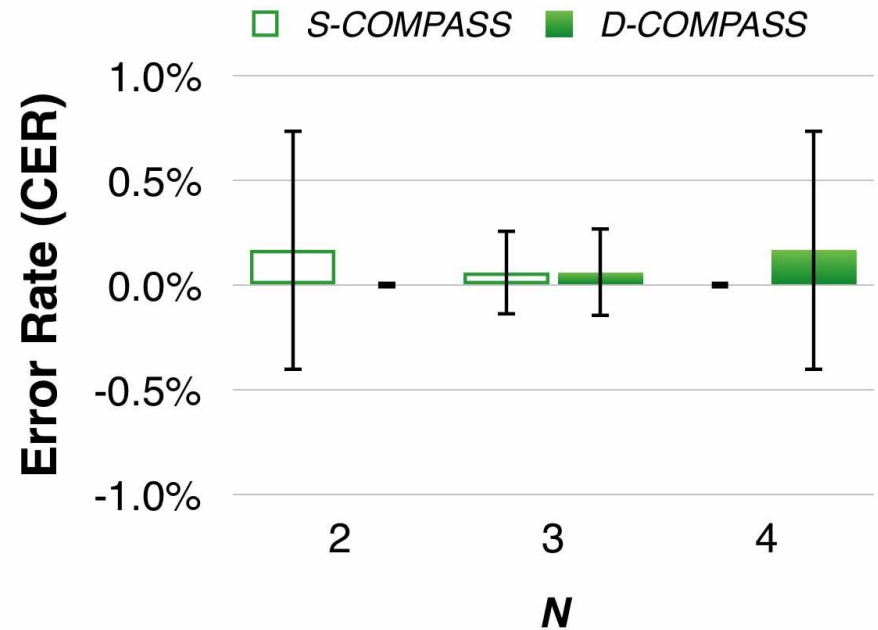
Speed & Error Rate



N-Speed

$F_{2,22} = 3.42, p = .05$ (*S-COMPASS*)

$F_{2,22} = 7.04, p < .01$ (*D-COMPASS*)



N- Error Rate

$F_{2,22} = 0.66, n.s.$ (*S-COMPASS*)

$F_{2,22} = 1.00, n.s.$ (*D-COMPASS*)

Final Design



D-COMPASS

$N = 3$

Auto-completion

Auto-Completion

$$I = I_1 I_2 \cdots I_n$$

$$W = W_1 W_2 \cdots W_n W_{n+1} \cdots W_m$$

$$P(I|W) = \begin{cases} 1 & \text{if } I \text{ matches with } W \\ 0 & \text{otherwise} \end{cases}$$

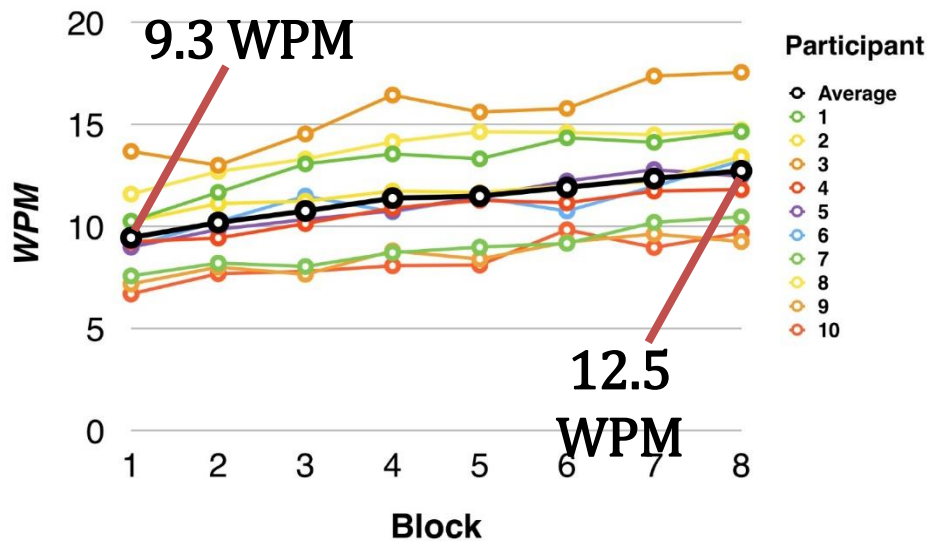


$$P(I|W) = \begin{cases} \alpha^{m-n} & \text{if } I \text{ matches with the prefix of } W \\ 0 & \text{otherwise} \end{cases}$$

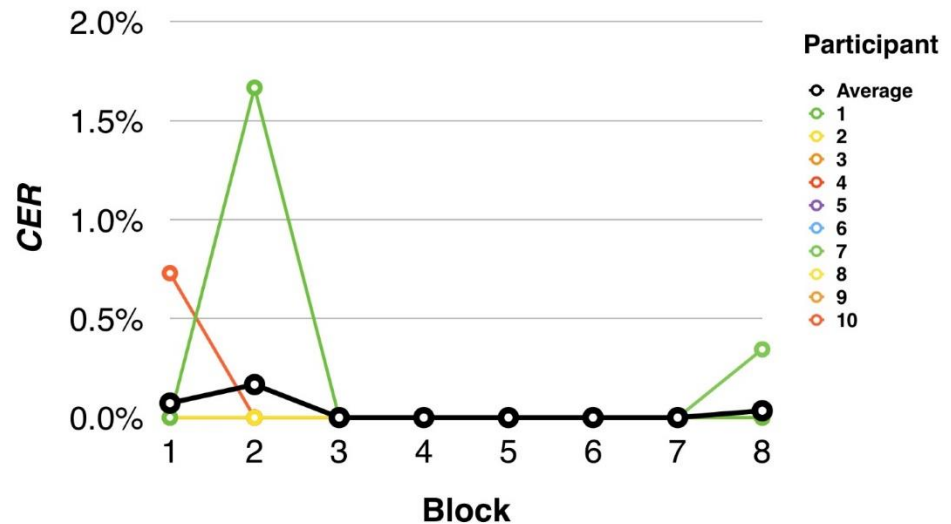
$\alpha = 0.7$ as the penalty of looking ahead

Speed & Error Rate

10 participants × 8 blocks × 10 phrases



Block - Speed
 $F_{7,63} = 48.0, p < .0001$



Block - Error Rate
 $F_{7,63} = 0.81, n. s.$

Applications

