

# IN4MATX 133: User Interface Software

**Lecture:**  
**Modeling human performance**

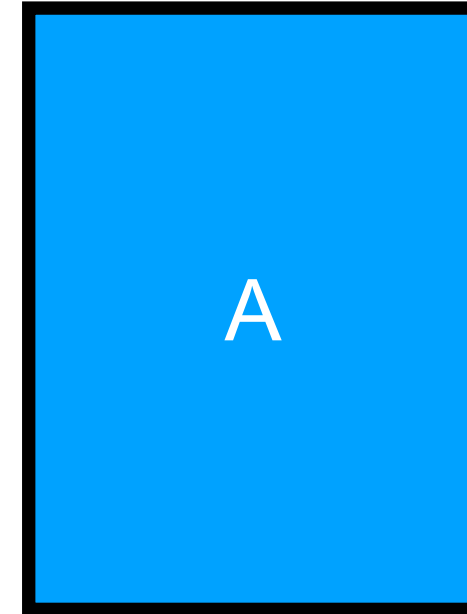
# Today's goals

**By the end of today, you should be able to...**

- Describe the major components of Fitts's Law
- Explain how Fitts's Law impacts how interfaces should be designed
- Describe approaches for correcting systematic errors in touch performance

# Question

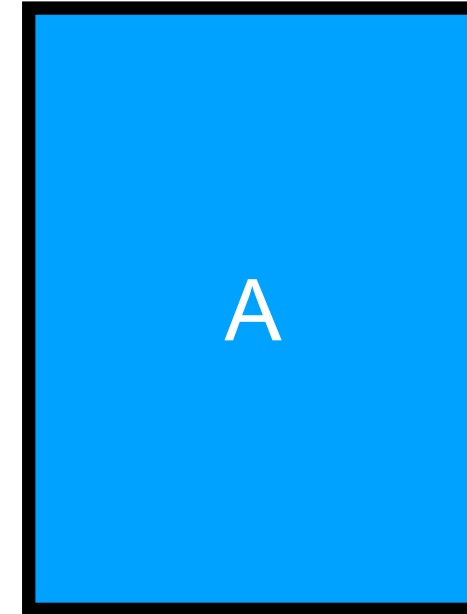
Which button would be faster to click on?



- ☒ A
- ☐ B
- ☐ C Roughly equal
- ☐ D
- ☐ E

# Question

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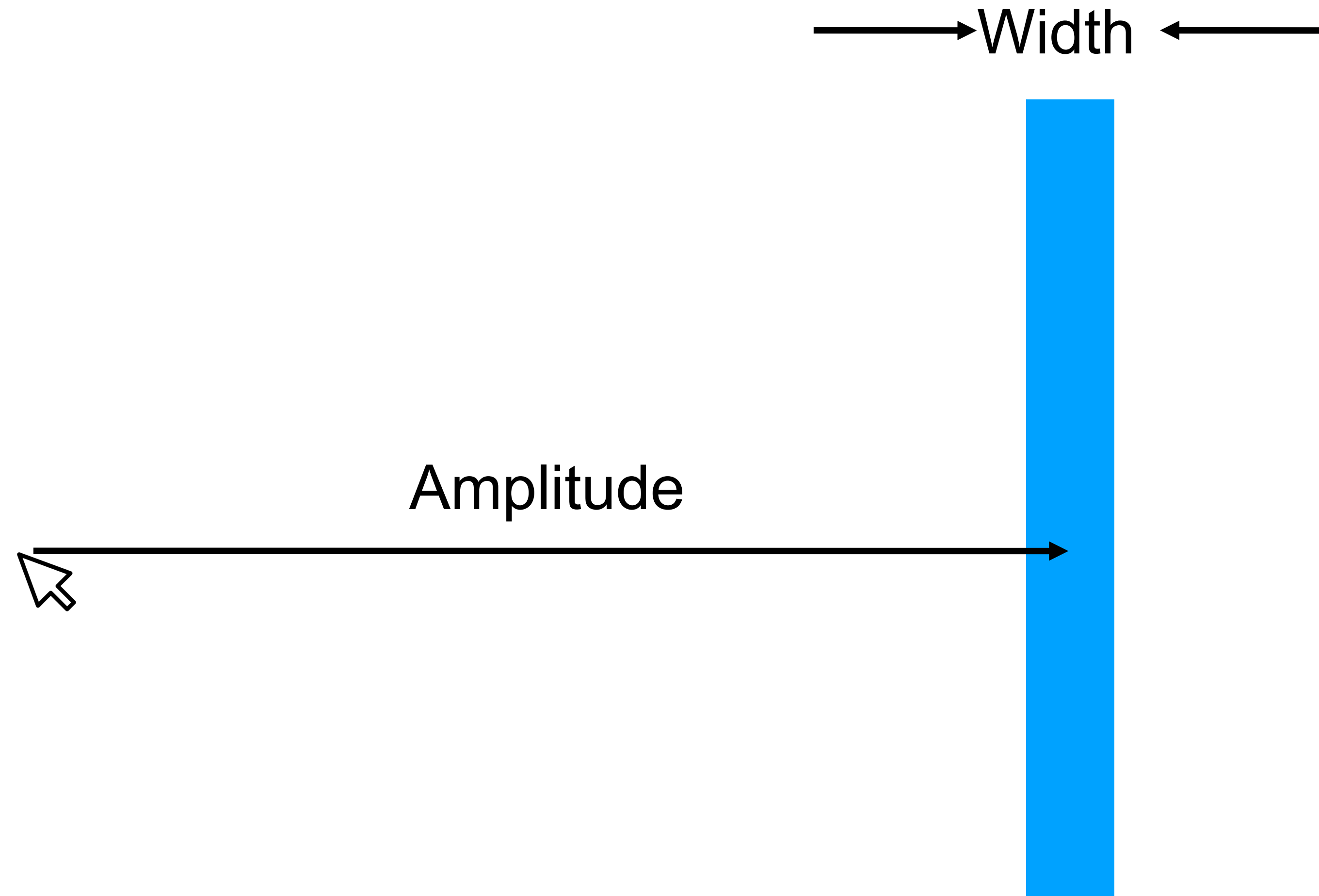
# Fitts's Law (1954)

- Models time to acquire targets in aimed movement
  - Reaching for control in a cockpit
  - Moving across a dashboard
  - Pulling defective items from a conveyor belt
  - Clicking on icons using a mouse

# Fitts's Law (1954)

- Very powerful, widely used
  - Holds for many circumstances (e.g., under water)
  - Allows for comparison among different experiments
  - Used both to measure and predict

# Point-select task



# Fitts's Law

- $MT = a + b \log_2(A / W + 1)$ 
  - What kind of equation does this look like?



# Fitts's Law

- $MT = a + b \log_2(A / W + 1)$ 
  - What kind of equation does this look like?
- $y = mx + b$
- $MT = a + bx$ , where  $x = \log_2(A / W + 1)$ 
  - $x$  is called the Index of Difficulty (ID)
  - As “A” goes up, ID goes up
  - As “W” goes up, ID goes down

# Movement Time (MT)

- $MT = a + b \log_2(A / W + 1)$
- Time, in seconds, to acquire the target (e.g., click on the button)

# Index of Difficulty (ID)

- $\log_2(A / W + 1)$ 
  - Fitts's Law claims that the time to acquire a target increases linearly with the log of the ratio of the movement distance or amplitude (A) to target width (W)

# Index of Difficulty (ID)

- $\log_2(A / W + 1)$ 
  - Fitts's Law claims that the time to acquire a target increases linearly with the log of the ratio of the movement distance or amplitude (A) to target width (W)
- Why is it significant that it is a ratio?
  - Units of A and W don't matter
  - Allows comparison across experiments

# Index of Difficulty (ID)

- $\log_2(A / W + 1)$ 
  - Fitts's Law claims that the time to acquire a target increases linearly with the log of the ratio of the movement distance or amplitude (A) to target width (W)
- ID units typically in “bits”
  - Because of association with information capacity and somewhat arbitrary use of base-2 logarithm

# Index of Performance (IP)

- $MT = a + b \log_2(A / W + 1)$ 
  - $b$  is slope
- $1/b$  is called Index of Performance (IP)
  - If  $MT$  is in seconds,  $IP$  is in bits/second
- Also called “throughput” or “bandwidth”
- **$a$  and  $b$  depend on the input device**



# [Fitts's law demo]

<http://simonwallner.at/ext/fitts/>

# “Beating” Fitts’s law

- It is the law, right?
  - $MT = a + b \log_2(A/W + 1)$
- So how can we reduce movement time?
  - Reduce amplitude (A)
  - Increase width (W)



# “Beating” Fitts’s law

- Put targets closer together
- Make targets bigger
- Make cursor bigger
- Make impenetrable edges

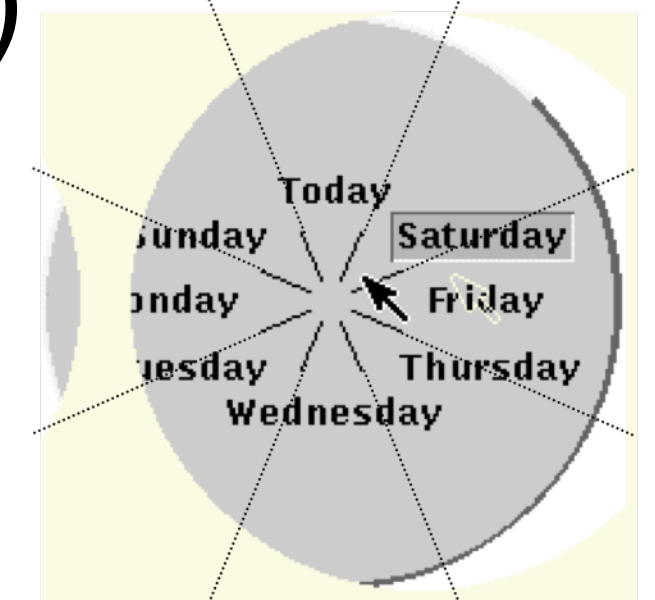
# Question

Which menu will be faster on average?

(A) Pop-up Linear Menu



(B) Pop-up Pie Menu



☐ A

☐ B

☐ C Roughly equal

☐ D

☐ E

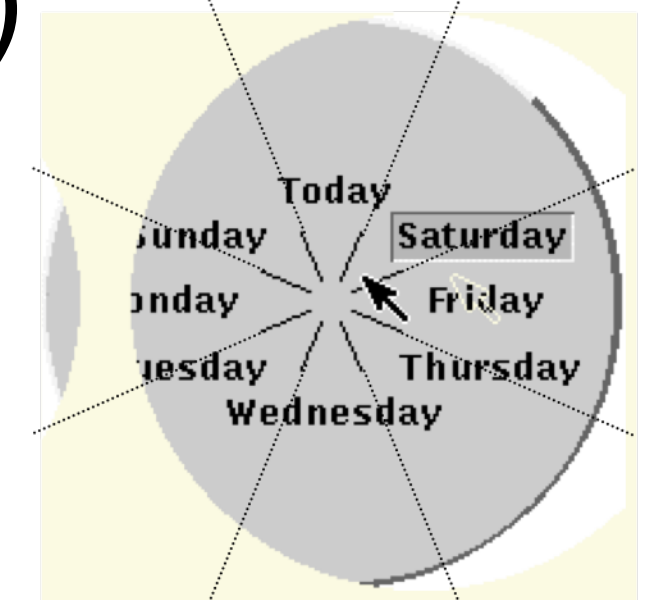
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☐ A

☒ B

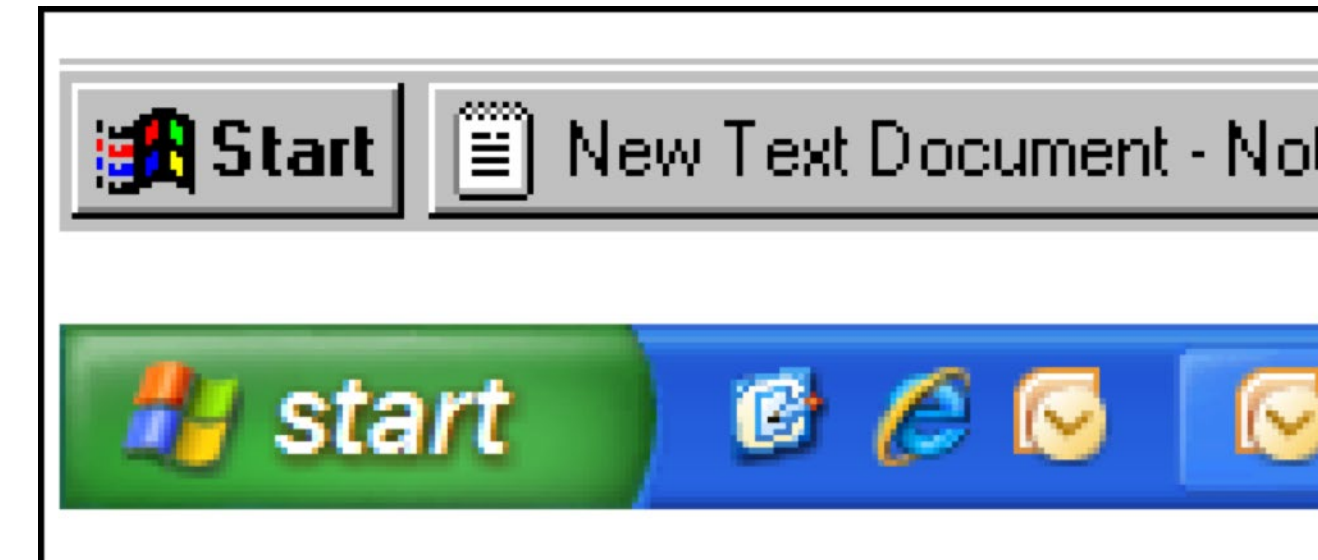
☐ C Roughly equal

☐ D

☐ E

# Fitts's Law in windowing

- Windows 95: missed by a pixel
- Windows XP: good to the end
- Corners and edges make great targets
  - Do not have to move precisely to trigger them
  - They have “infinite” width



# Fitts's Law in other domains

- How would Fitts's Law apply to using touch input on a phone?
  - Shorter distances (smaller screen)
- All things being equal, movement times *should* be lower
  - Shorter distances, faster to move your finger than a mouse

# Fitts's Law in other domains

- But in practice, touchscreens on mobile tend not to be much faster
  - Buttons are smaller
  - People tend to be slower near the edges of touchscreens

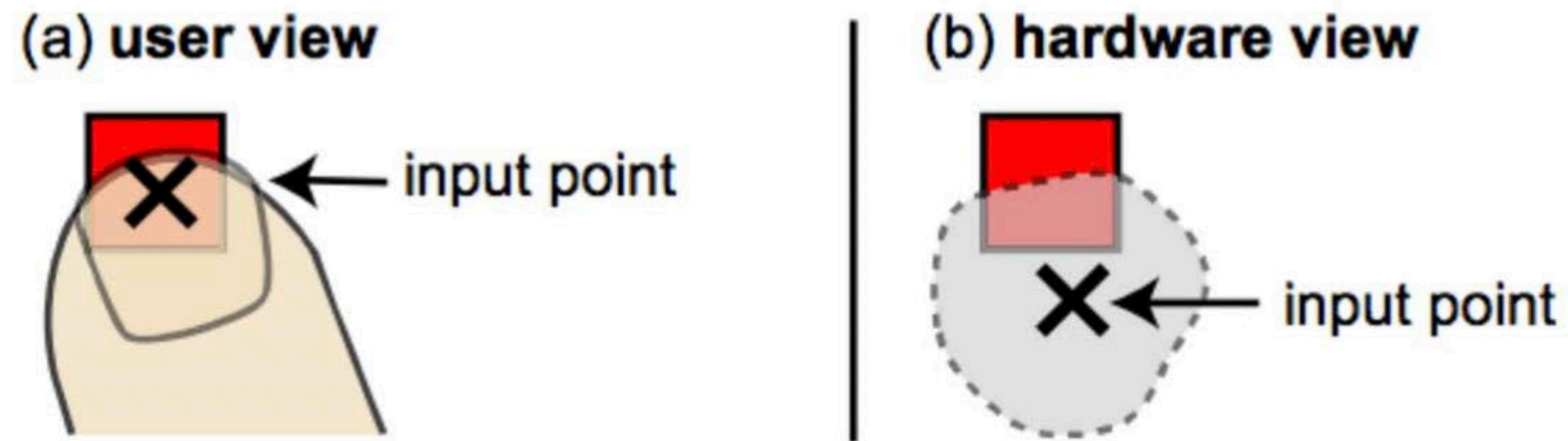
# Modeling input

# Modeling mouse position

- Mouse pointer is relatively small
- We model it via  $X$ ,  $Y$  position on the screen
- See whether that  $X$ ,  $Y$  overlaps with a button, for example
  - Targets are usually large enough that “exact” position does not matter



# Modeling touch position



# Modeling touch position

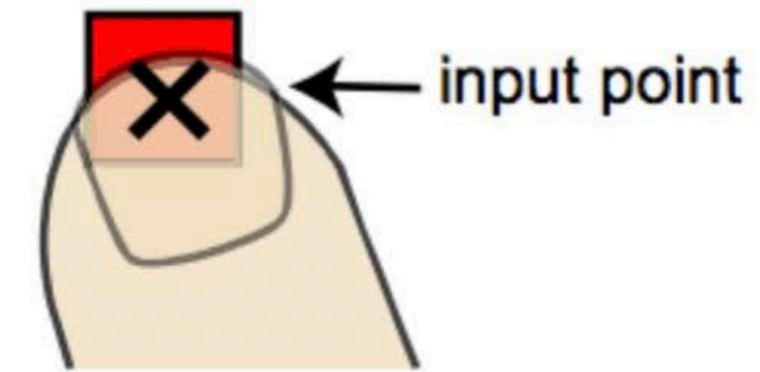
- One interpretation of the problem:  
*our fingers are fat*
  - We should use tiny styluses to make our selection more accurate
- Another interpretation:  
*our model of touch position is inaccurate*
  - We should make our model better



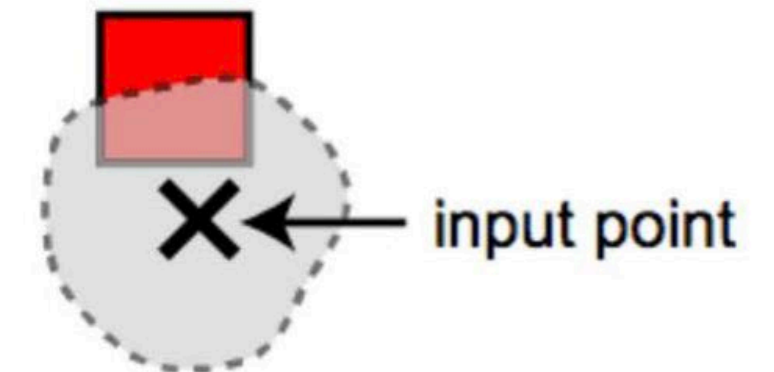
# Modeling touch position

- How can we improve our model?
- Make the hardware view more closely match the user view

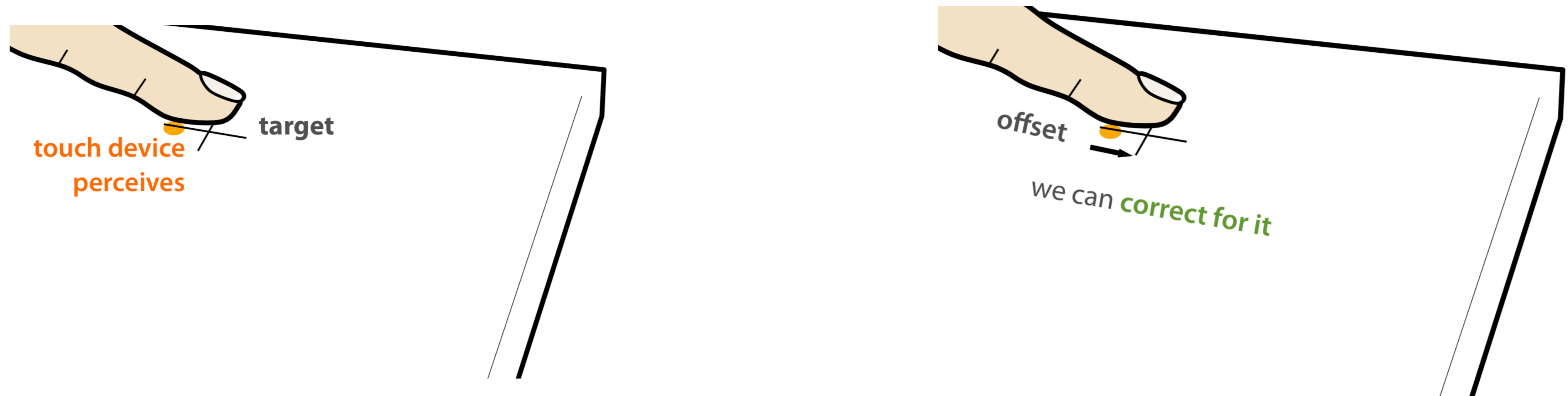
(a) **user view**



(b) **hardware view**

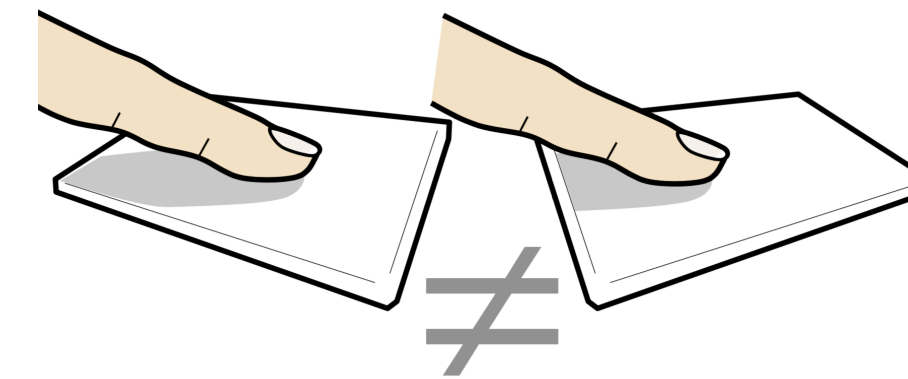


# Modeling touch position

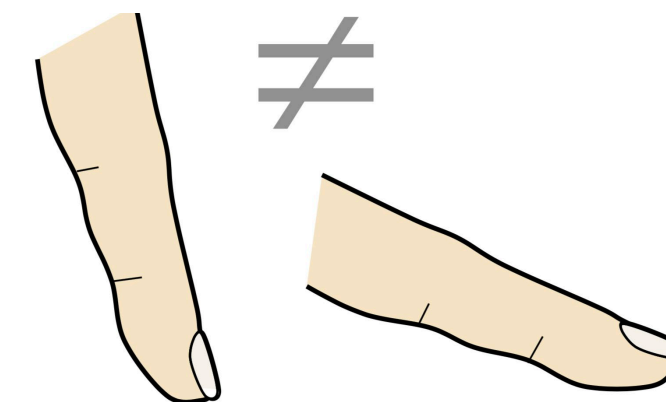


# Modeling touch position

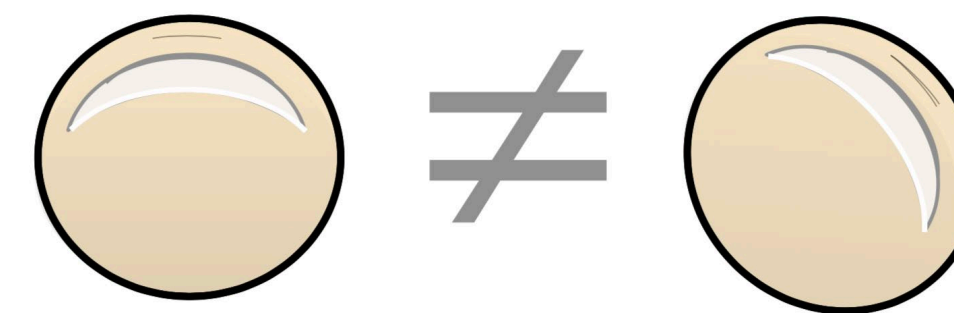
- Hypothesis: yaw, pitch, and roll all impact touch position
- Additionally, for each person, finger size/shape and mental model impact touch position



Yaw: angle of touch device



Pitch: angle of finger

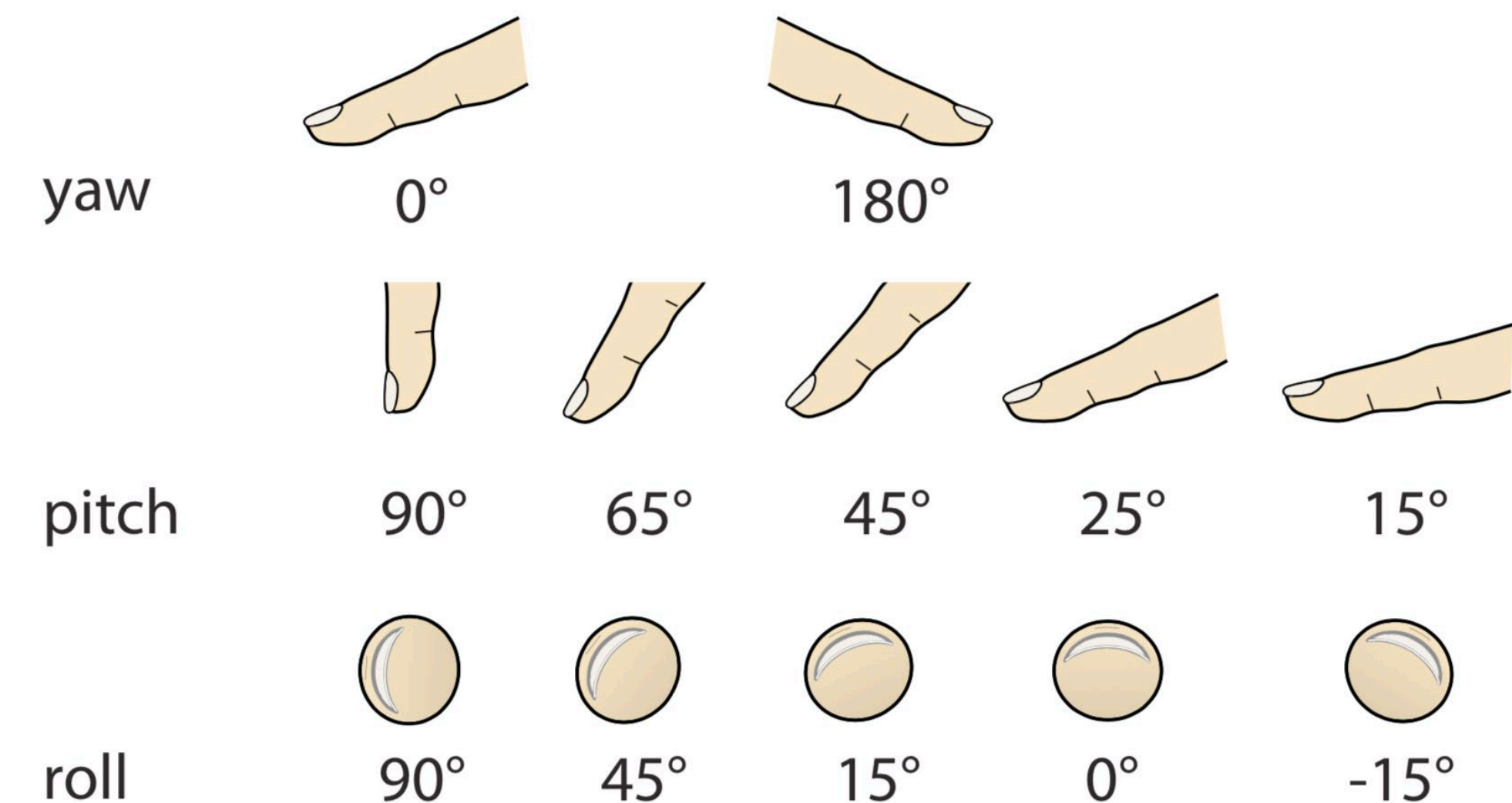


Roll: rotation of finger



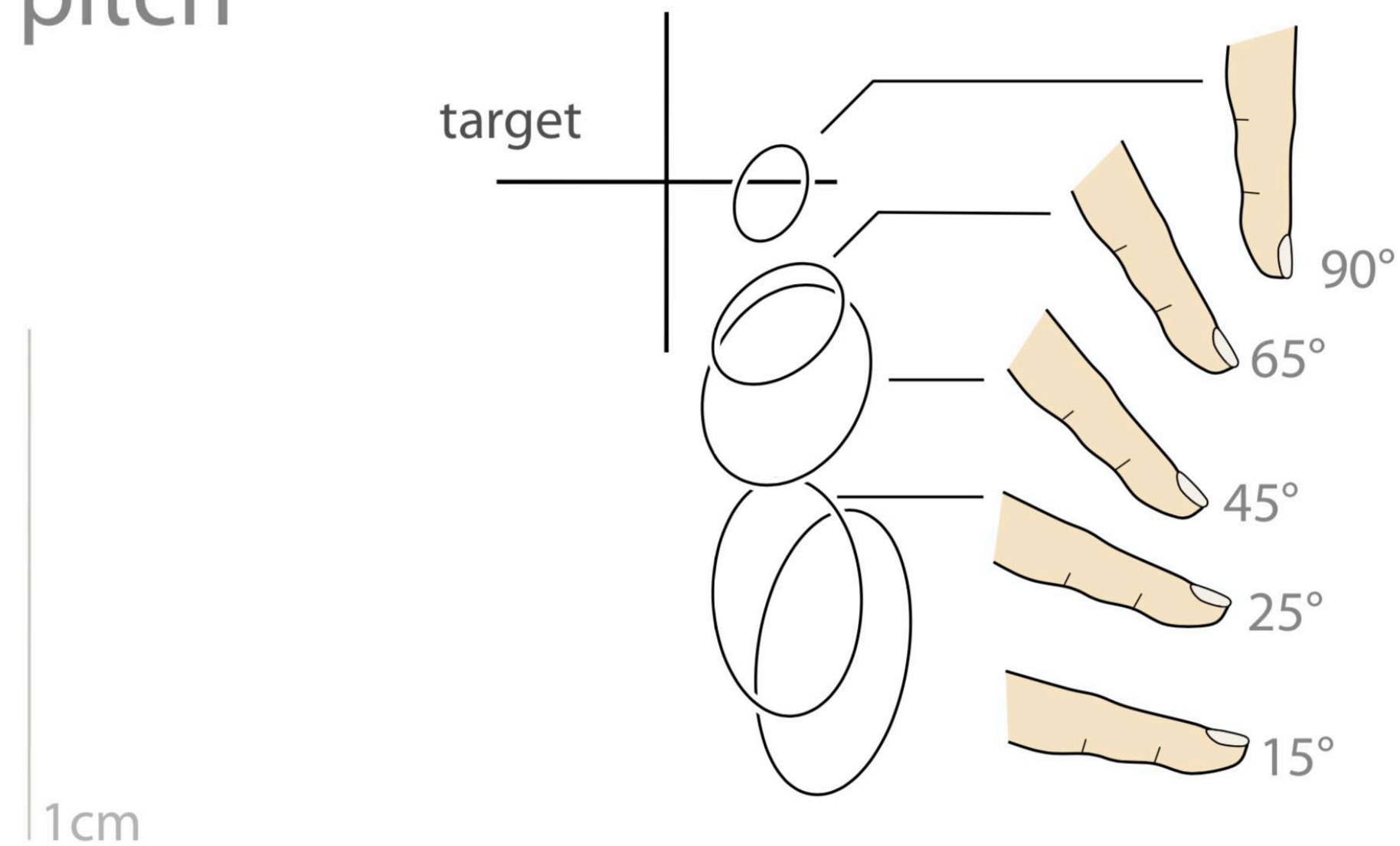
# Modeling touch position

- Ran a study
  - 12 participants touched 600 points each
  - Varied yaw, pitch, and roll

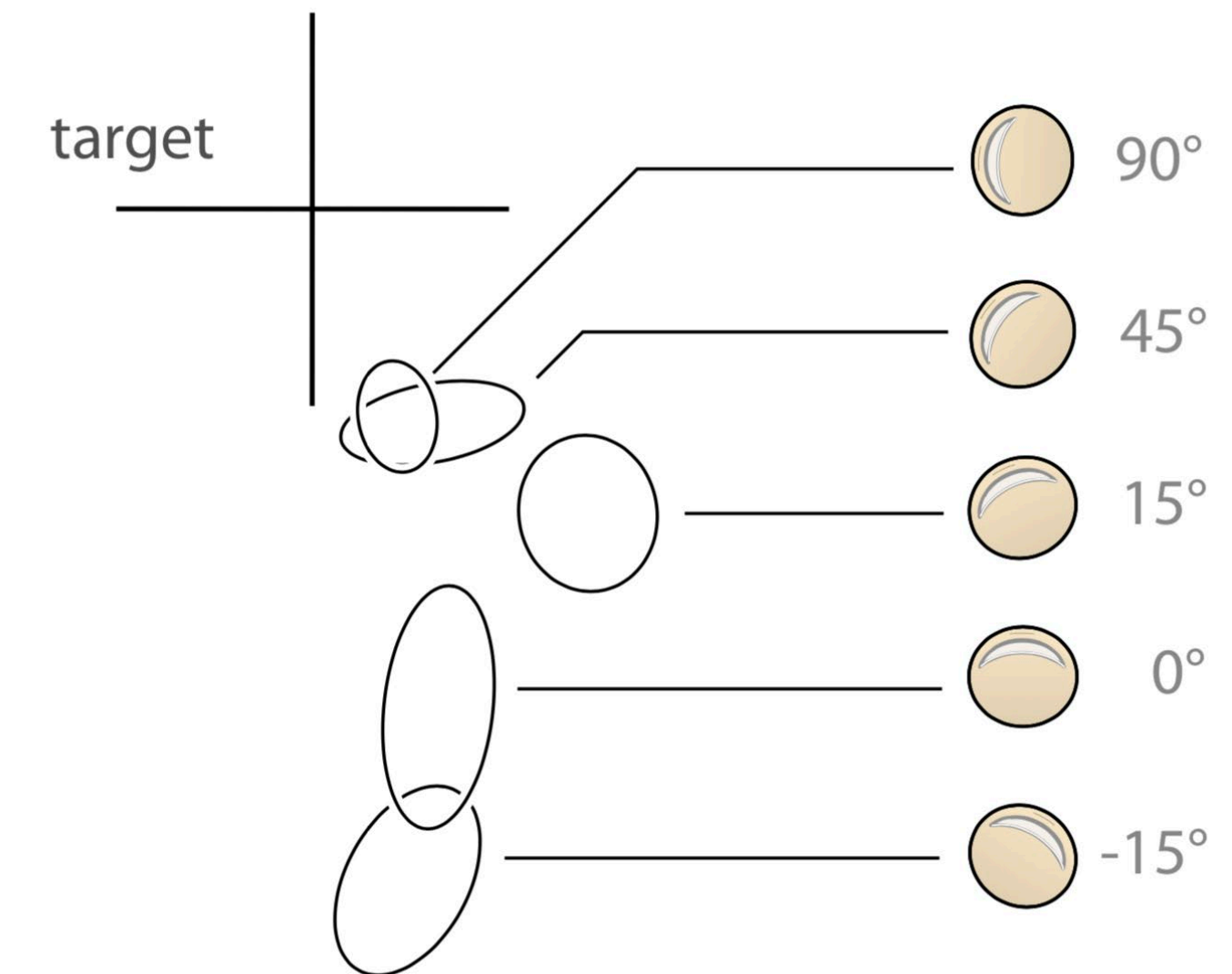


# Modeling touch position

pitch



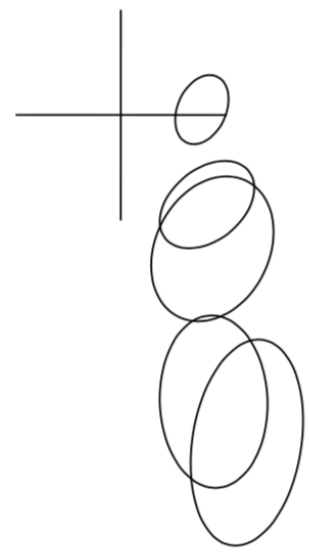
roll



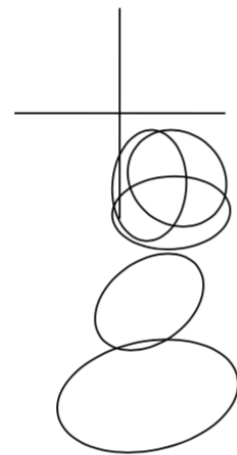
# Modeling touch position

user

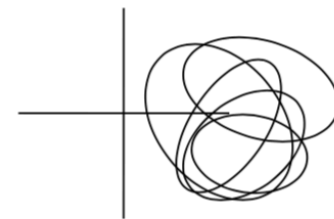
#1



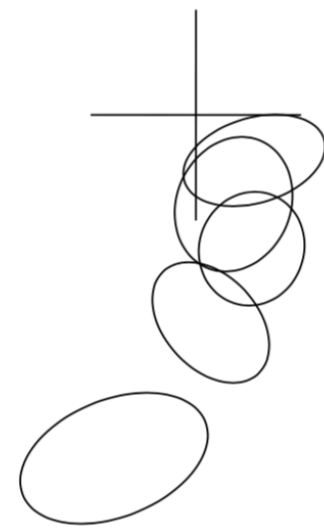
#2



#3



#4



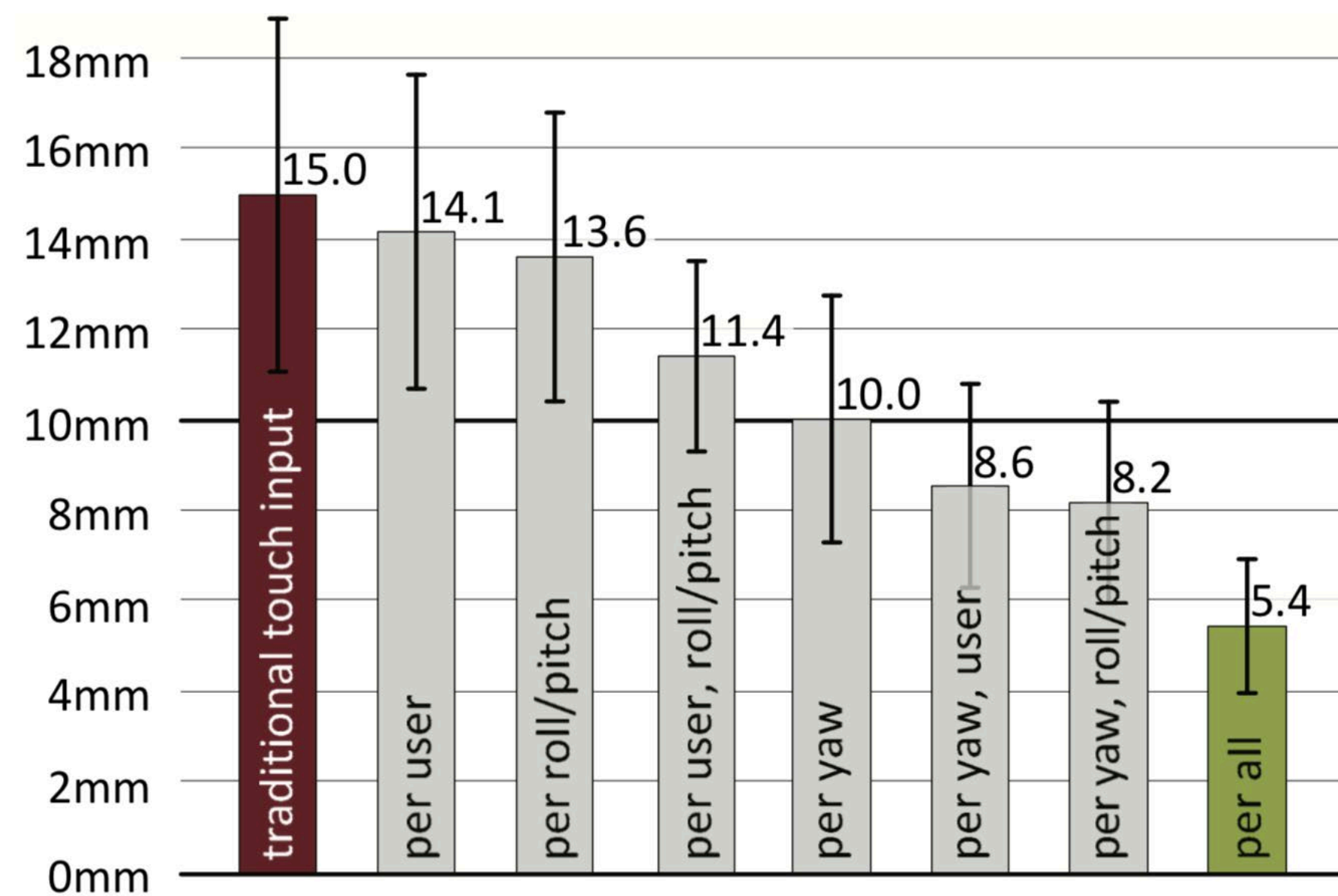
pitch

1cm



# Modeling touch position

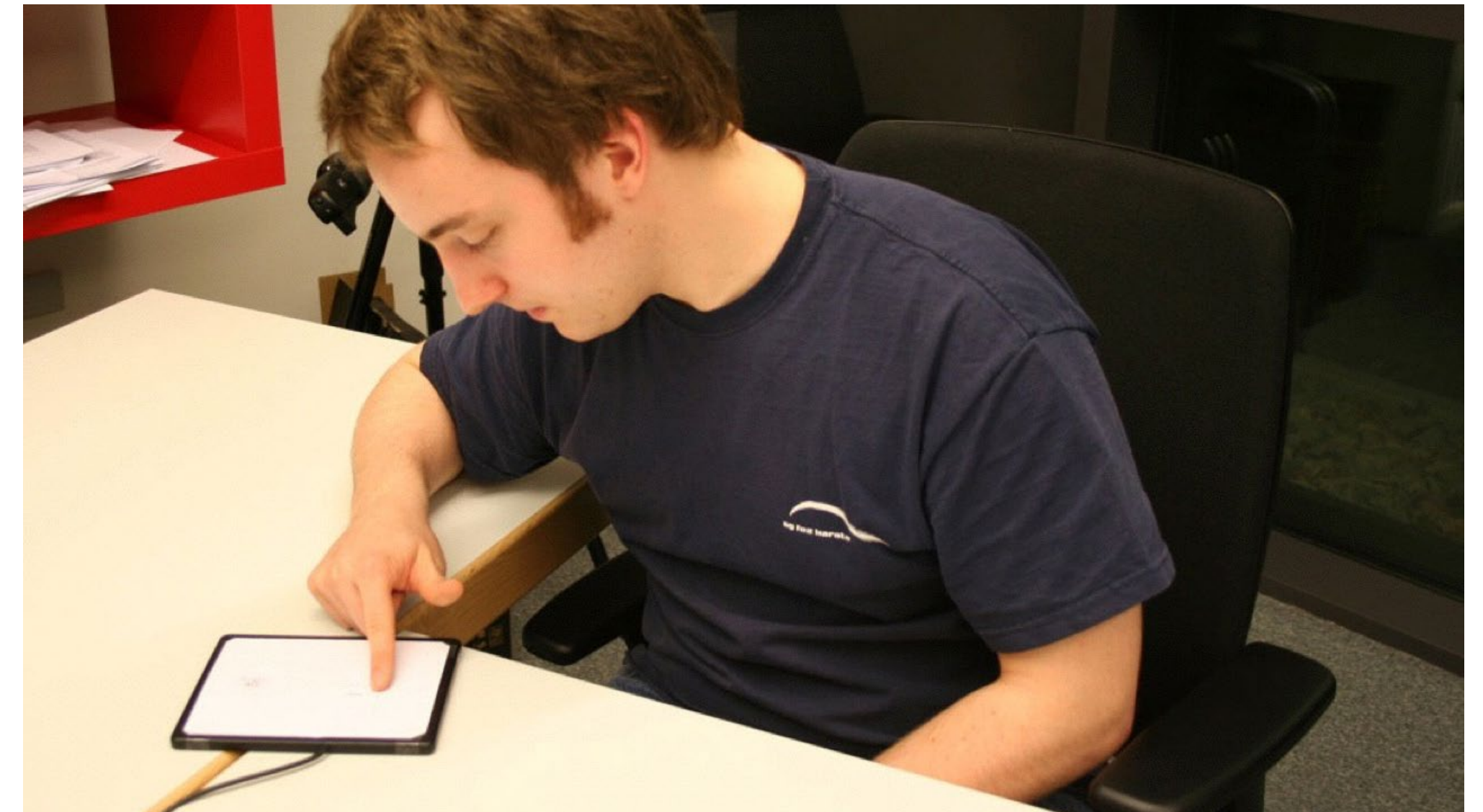
minimum button size



Improving the model means that buttons can be **3x** smaller and not be any harder to click

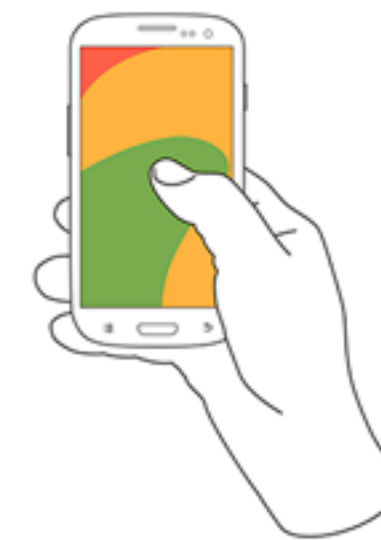
# Modeling touch input

- Study was *very* controlled
  - Participant sat in a chair, the screen was on a desk
- How about the other ways that people use their phones?



# Modeling phone grip

- People grip their phones in different ways
- Grip changes with phone size, hand size
  - Situational changes (e.g., walking, holding something)
- Can we detect phone grip and update our model?



**49%**  
one handed

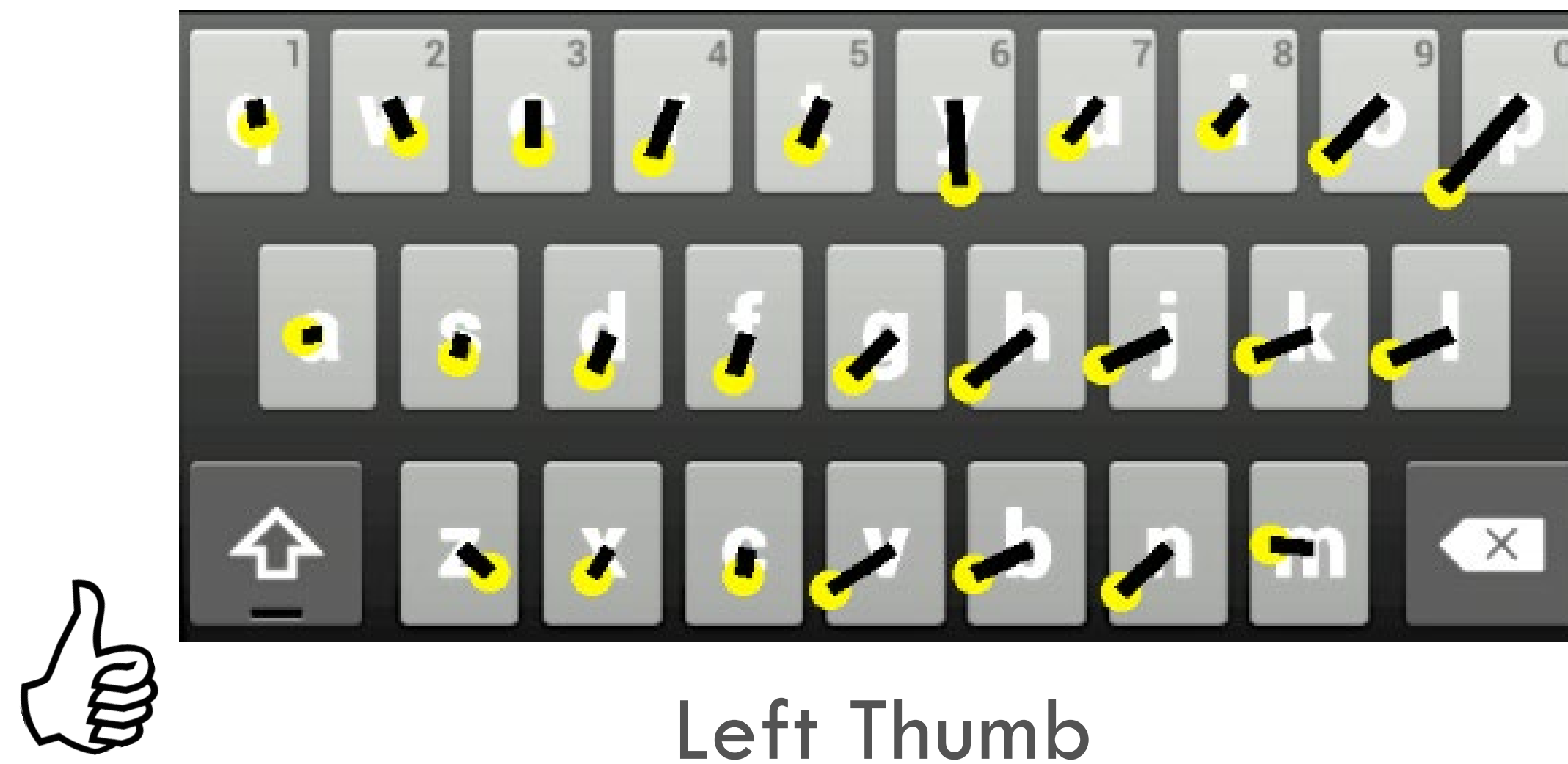


**36%**  
cradled



**15%**  
two handed

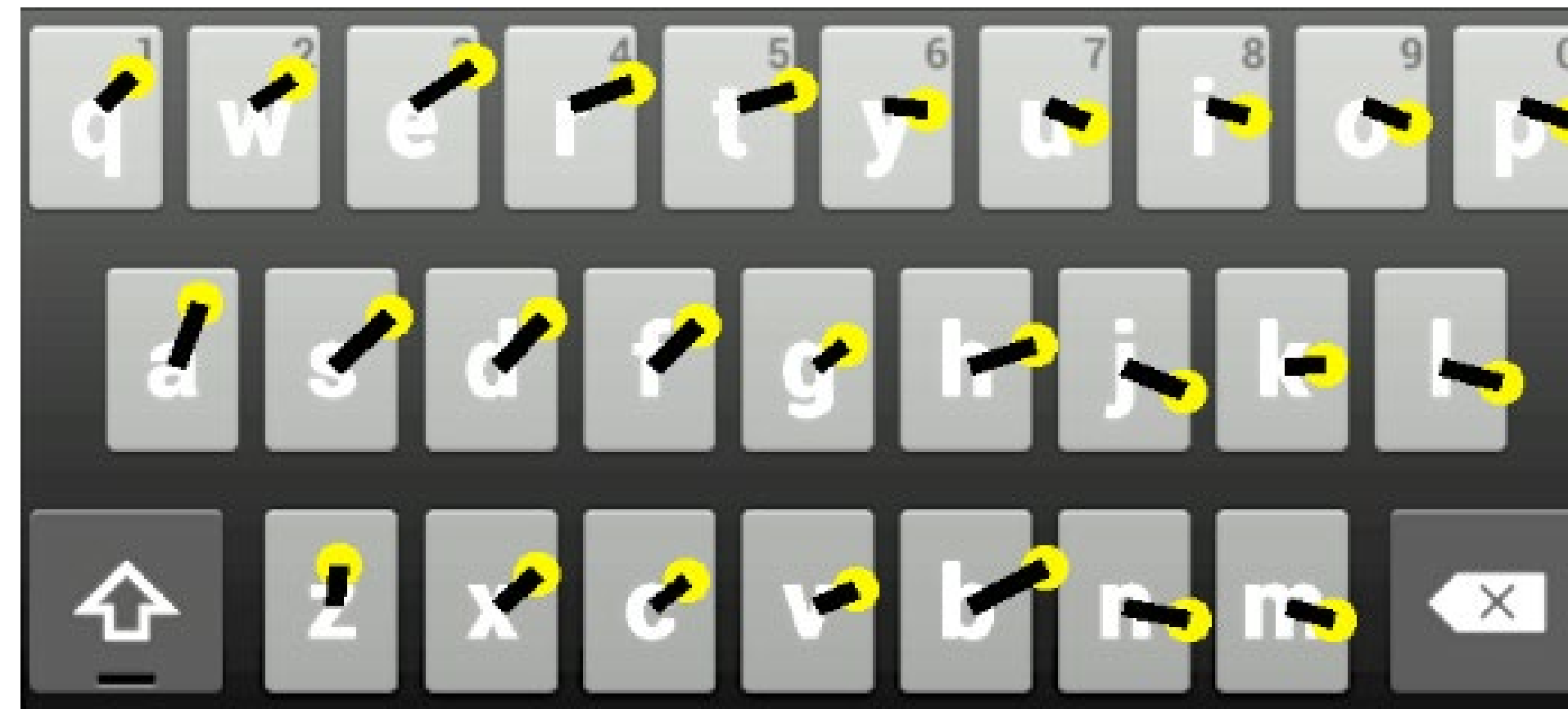
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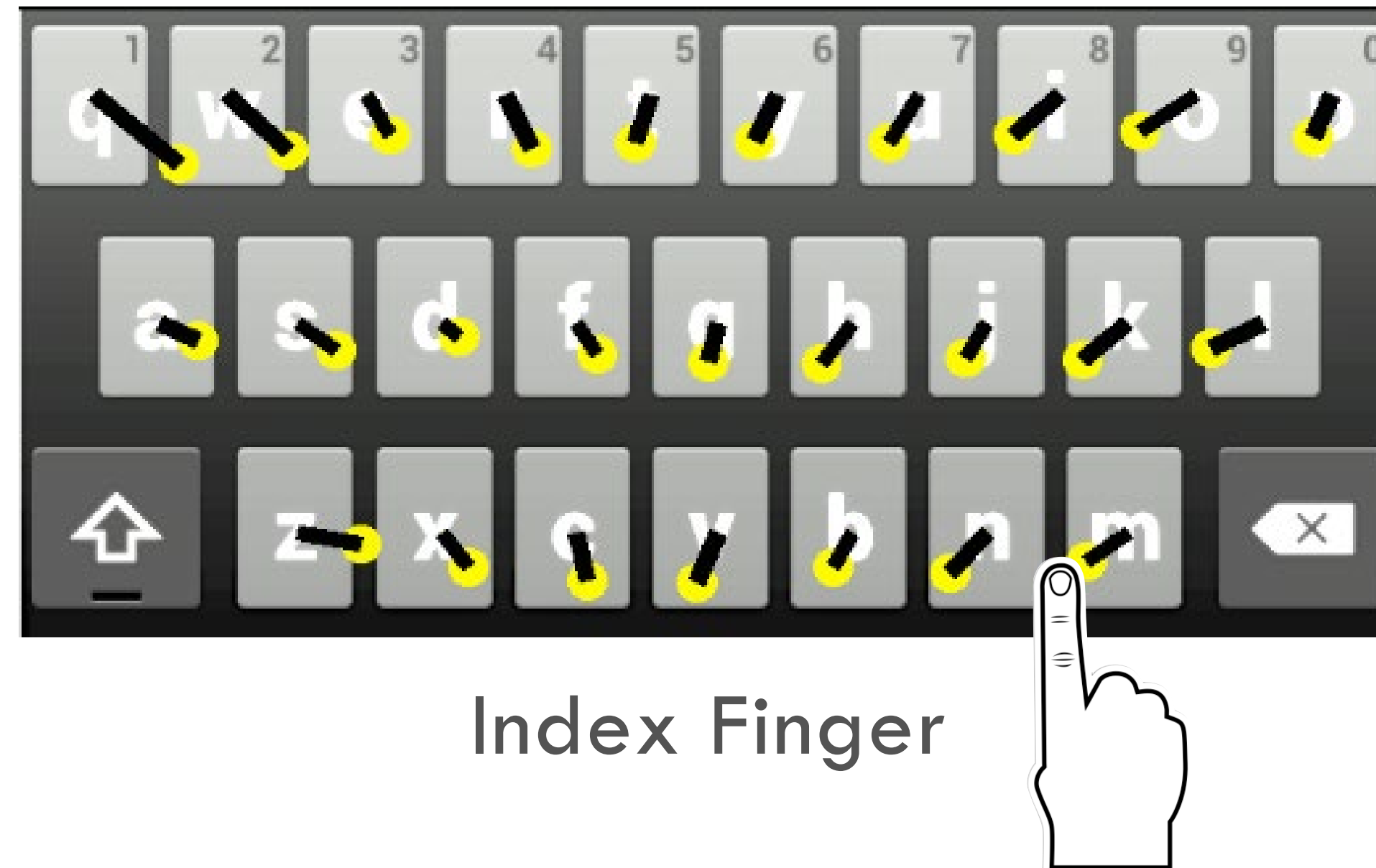


Right Thumb



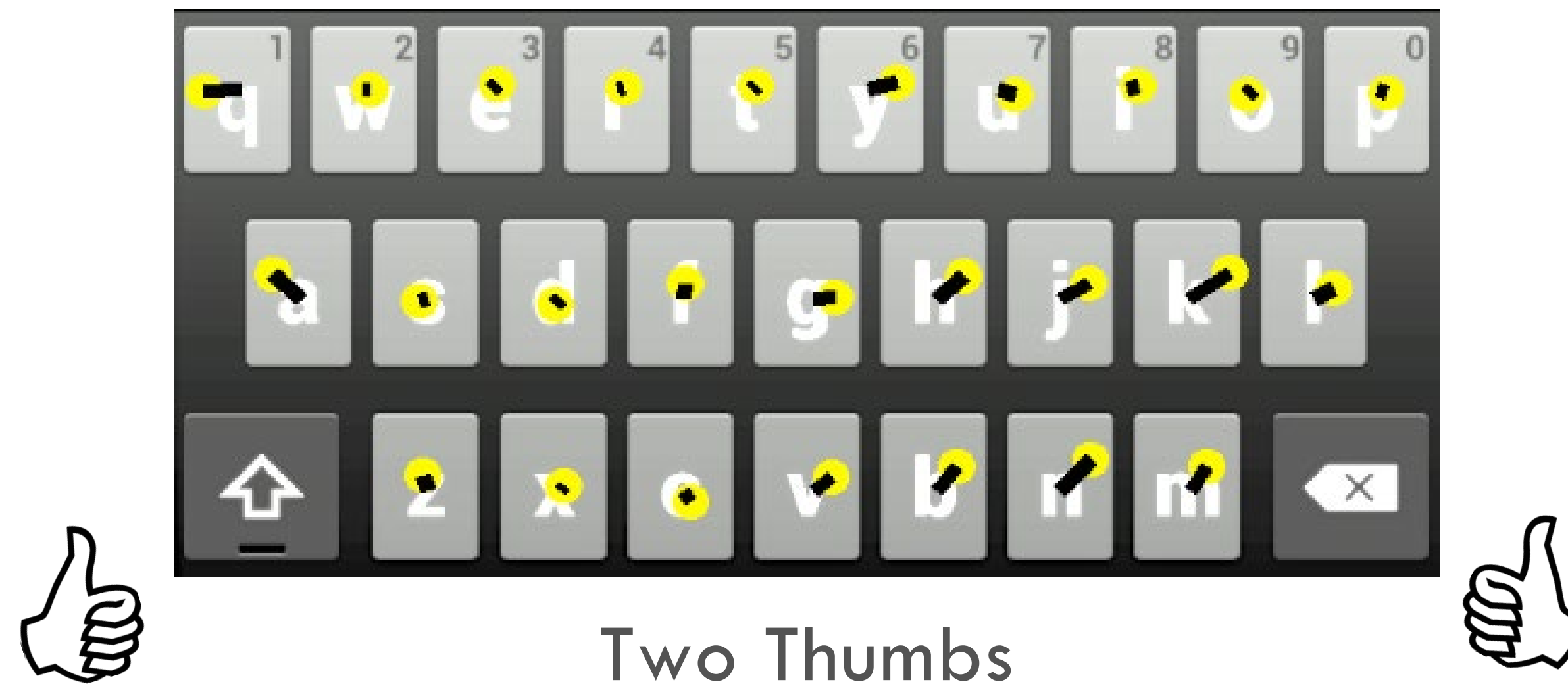
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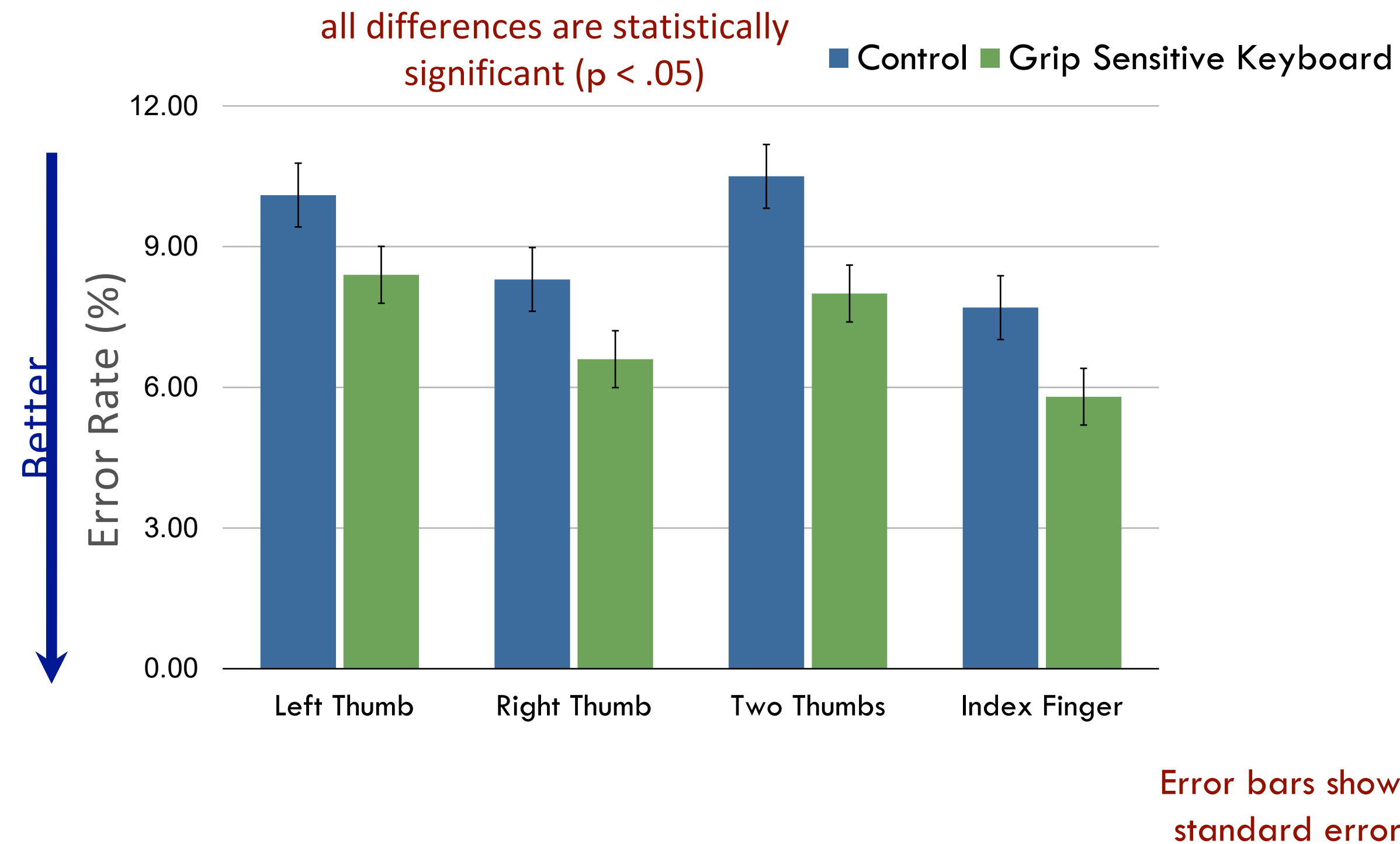
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# Summary

- Modeling helps us measure and predict whether a tool or approach is beneficial for a task
- Fitts's law models time taken to click on a target
  - Demonstrates that larger, nearer buttons reduce time taken
- Improved models lead to higher accuracy
  - Adjust for finger angle and rotation rather than assuming that a user intends to touch with the center of their finger
  - Infer grip using phone sensors to improve typing accuracy

# Today's goals

**By the end of today, you should be able to...**

- Describe the major components of Fitts's Law
- Explain how Fitts's Law impacts how interfaces should be designed
- Describe approaches for correcting systematic errors in touch performance