

# Information Theory:

**A useful tool to determine channel capacity, the maximum number of stimuli that can be correctly identified, without running multiple experiments**

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# Overview

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- ◆ Why information theory?
- ◆ Information as psychological concepts
- ◆ Information transmission measurement
- ◆ Channel capacity



# **WHY INFORMATION THEORY?**

# How many tactors to put on a belt?

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4?      8?      12?



- ◆ As number of tactors increases, accuracy (percent-correct scores) decreases
- ◆ Objectives:
  - Ⓢ Place maximum number of tactors on the belt, and
  - Ⓢ Achieve a perfect accuracy

➔ **Maximum** number of **perfectly-identifiable** tactors



# **INFORMATION AS PSYCHOLOGICAL CONCEPTS**

# Information & Uncertainty

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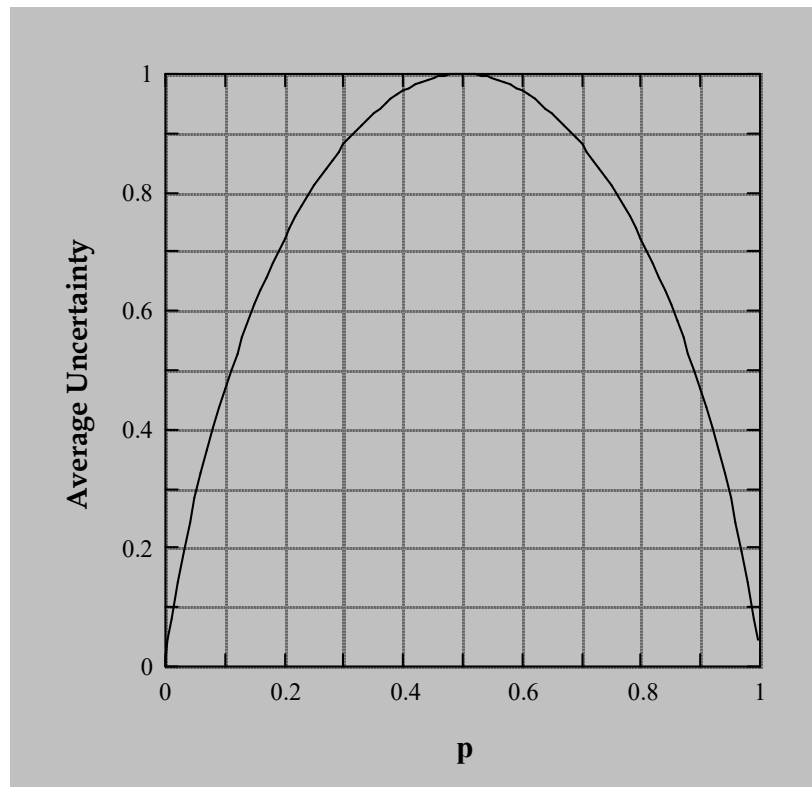
- ◆ Information = reduction in uncertainty
- ◆ Uncertainty of a given outcome  $X_i$ 
  - ⊙  $U_i = \log_2[1/P(X_i)] = -\log_2 P(X_i)$
- ◆ If  $P(X_i) = 1/k$  (equally likely  $X_i$ ,  $i \in [1, k]$ )
  - ⊙ Then  $U_i = \log_2 k$  for all  $i$
- ◆ Average uncertainty:
  - ⊙  $U = \sum P(X_i) U_i$
  - ⊙  $U = -\sum P(X_i) \log_2[P(X_i)]$

# Average Uncertainty

- ◆ Shannon's measure:  $U = - \sum P(X_i) \log_2[P(X_i)]$
- ◆ Unit for uncertainty and information: *bits*

*dichotomous  
distribution  
with  $p$  and  $q=1-p$*

*average  
uncertainty:  
 $-p\log_2 p - q\log_2 q$*





# **INFORMATION TRANSMISSION MEASUREMENT**

# Absolute Identification Experiment

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- ◆ One-interval experiment
- ◆ Stimuli:  $S_i, i \in [1, k]$  ( $k > 2$ )
- ◆ Responses:  $R_j, j \in [1, k]$
- ◆ One-to-one mapping ( $S_i \Leftrightarrow R_i$ )
- ◆ On each trial, one of the stimuli  $S_i$  is presented with an *a priori* probability of  $P(S_i)$
- ◆ Subject makes a response with  $R_j$
- ◆ Trial-by-trial correct-answer feedback is optional

# S-R Confusion Matrix (e.g., $k = 5$ )

	$R_1$	$R_2$	$R_3$	$R_4$	$R_5$	
$S_1$	14	3	2	0	1	20
$S_2$	0	13	2	3	1	19
$S_3$	4	3	11	1	0	19
$S_4$	2	0	2	15	1	20
$S_5$	5	3	2	0	12	22
	25	22	19	19	15	100

# of times  $S_2$  was Presented

# of times the joint event ( $S_3, R_4$ ) occurred

# of times  $R_5$  was called

# IS and IR

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## ◆ IS (*Information in Stimulus*)

⌚ IS is the average uncertainty in stimulus

$$IS = -\sum_{i=1}^k P(S_i) \log_2 P(S_i)$$

◆ *If all stimuli are equally likely, then*

$$IS = \log_2 k$$

## ■ IR (*Information in Response*)

◆ *IR is the average uncertainty in response*

$$IR = -\sum_{j=1}^k P(R_j) \log_2 P(R_j)$$

# IT (Information Transfer)

- ◆ Also called “mutual information”
- ◆ IT = reduction in uncertainty
- ◆ For a particular  $(S_i, R_j)$  pair:
  - Ⓢ  $U(S_i)$  before:  $-\log_2 P(S_i)$ 
    - ❖ Assuming that  $P(S_i)$  is constant throughout the experiment
  - Ⓢ  $U(S_i)$  after:  $-\log_2 P(S_i | R_j)$
  - Ⓢ  $IT(S_i, R_j) = -\log_2 P(S_i) - [-\log_2 P(S_i | R_j)]$

$$IT(S_i, R_j) = \log_2 \frac{P(S_i | R_j)}{P(S_i)}$$

- Average IT =  $\sum \sum P(S_i, R_j) IT(S_i, R_j)$
- IT: the degree of correlation between S's and R's



# **DATA ANALYSIS**

	$R_1$	$R_2$	$R_3$	$R_4$	$R_5$	
$S_1$	14	3	2	0	1	20
$S_2$	0	13	2	3	1	19
$S_3$	4	3	11	1	0	19
$S_4$	2	0	2	15	1	20
$S_5$	5	3	2	0	12	22
	25	22	19	19	15	100

# Estimation of IT — $IT_{est}$

- ◆ Average information transfer:

$$IT = \sum_{j=1}^k \sum_{i=1}^k P(S_i, R_j) \log_2 \frac{P(S_i | R_j)}{P(S_i)}$$

- *Its maximum-likelihood estimate:*

$$IT_{est} = \sum_{j=1}^k \sum_{i=1}^k \left( \frac{n_{ij}}{n} \right) \log_2 \left( \frac{n_{ij} \cdot n}{n_i \cdot n_j} \right) \quad \text{where} \quad \begin{aligned} n_{ij} &= \sum_{j=1}^k n_{ij} & n_j &= \sum_{i=1}^k n_{ij} \\ n &= \sum_{j=1}^k \sum_{i=1}^k n_{ij} = \sum_{i=1}^k n_i = \sum_{j=1}^k n_j \end{aligned}$$

- *Interpretation of  $2^{IT}$  or  $2^{IT_{est}}$  (compare with  $k=2^U$ )*

# Percent-correct scores and $IT_{est}$

$$IT_{est} = \sum_{j=1}^k \sum_{i=1}^k \left( \frac{n_{ij}}{n} \right) \log_2 \left( \frac{n_{ij} \cdot n}{n_i \cdot n_j} \right)$$

(A)

25	25
25	25

50%  
0 bits

(B)

25	25	25	25
25	25	25	25
25	25	25	25
25	25	25	25

25%  
0 bits

(C)

25	0	0	0
0	25	0	0
0	0	25	0
0	0	0	25

100%  
2 bits

(D)

0	0	0	25
0	0	25	0
0	25	0	0
25	0	0	0

0%  
2 bits



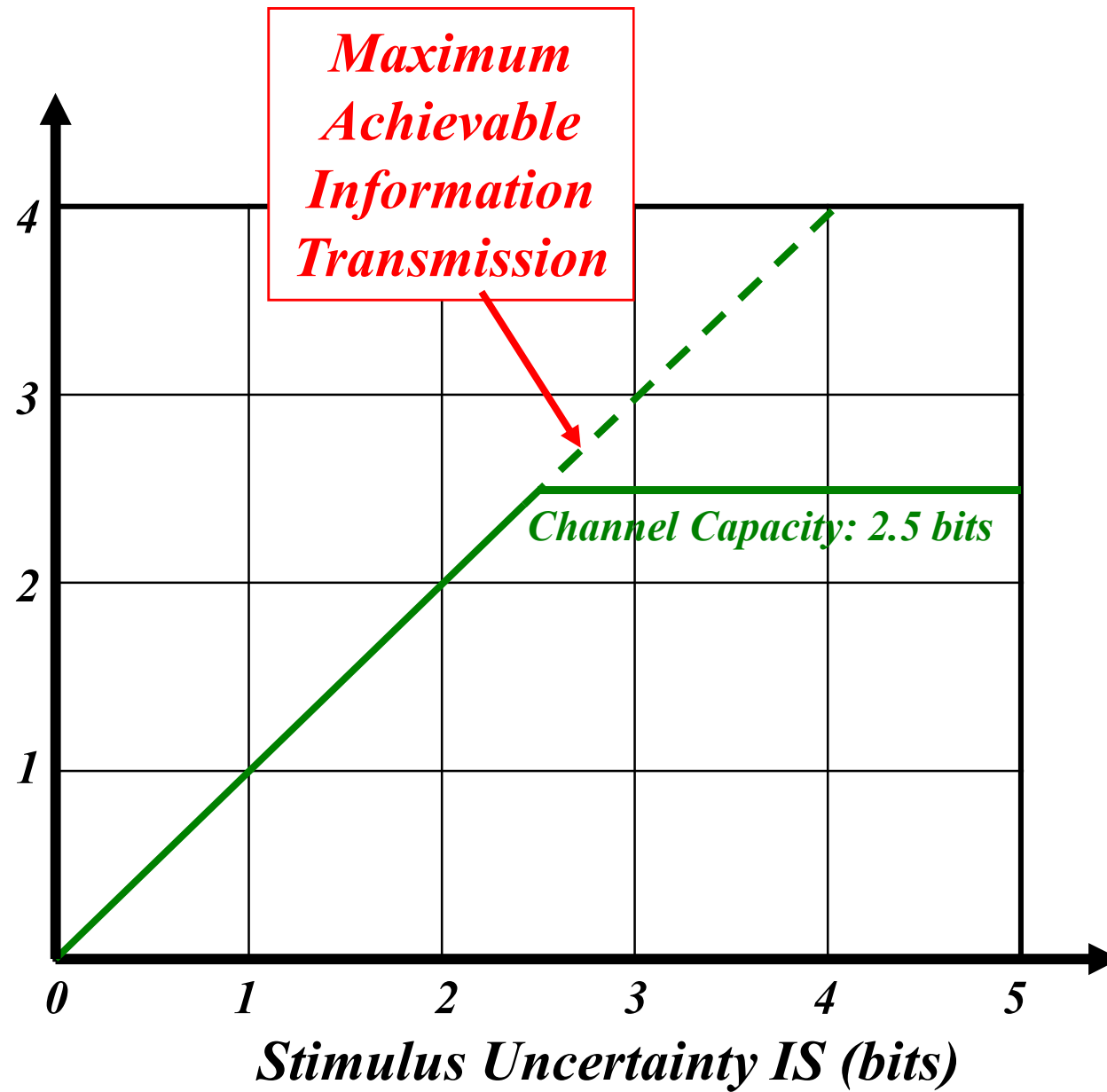
# **CHANNEL CAPACITY**

# Maximum Information Transmission

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- ◆ Mathematically,  $IT \leq IS$ .
- ◆ Intuitively, if the input and output are perfectly correlated, then  $IT = IS (= IR)$ .
- ◆ Assume that there exists a *maximum* information transmission
  - ⊗ For small values of  $IS$ ,  $IT = IS$ .
  - ⊗ As  $IS$  increases,  $IT = \text{constant}$  regardless of the value of  $IS$ .
- ◆ This maximum  $IT$  is accepted as the **channel capacity**

*Information  
Transmission  
IT (bits)*



# The “Magic Number” $7 \pm 2$

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- ◆ The “magic number” is derived from an *IT* range of 2.3 – 3.2 *bits*
- ◆ The “magic number” summarizes the typical **channel capacity** for uni-dimensional stimuli
- ◆ Uni-dimensional stimuli
  - ⌚ Only one physical variables (**target**) is manipulated to form the stimulus set
  - ⌚ Other physical variables (**background**) are either held constant or randomized

G. A. Miller, “The magical number seven, plus or minus two: Some limits on our capacity for processing information,” *The Psychological Review*, vol. 63, pp. 81-97, 1956.



# **EXAMPLES OF HAPTIC CHANNEL CAPACITY**

# Finger-span Length Identification

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Durlach, Delhorne, Wong, Ko, Rabinowitz, & Hollerbach (1989). Manual discrimination and identification of length by the finger-span method. *Perception & Psychophysics*, 46(1), 29–38.

# Length Identification Results

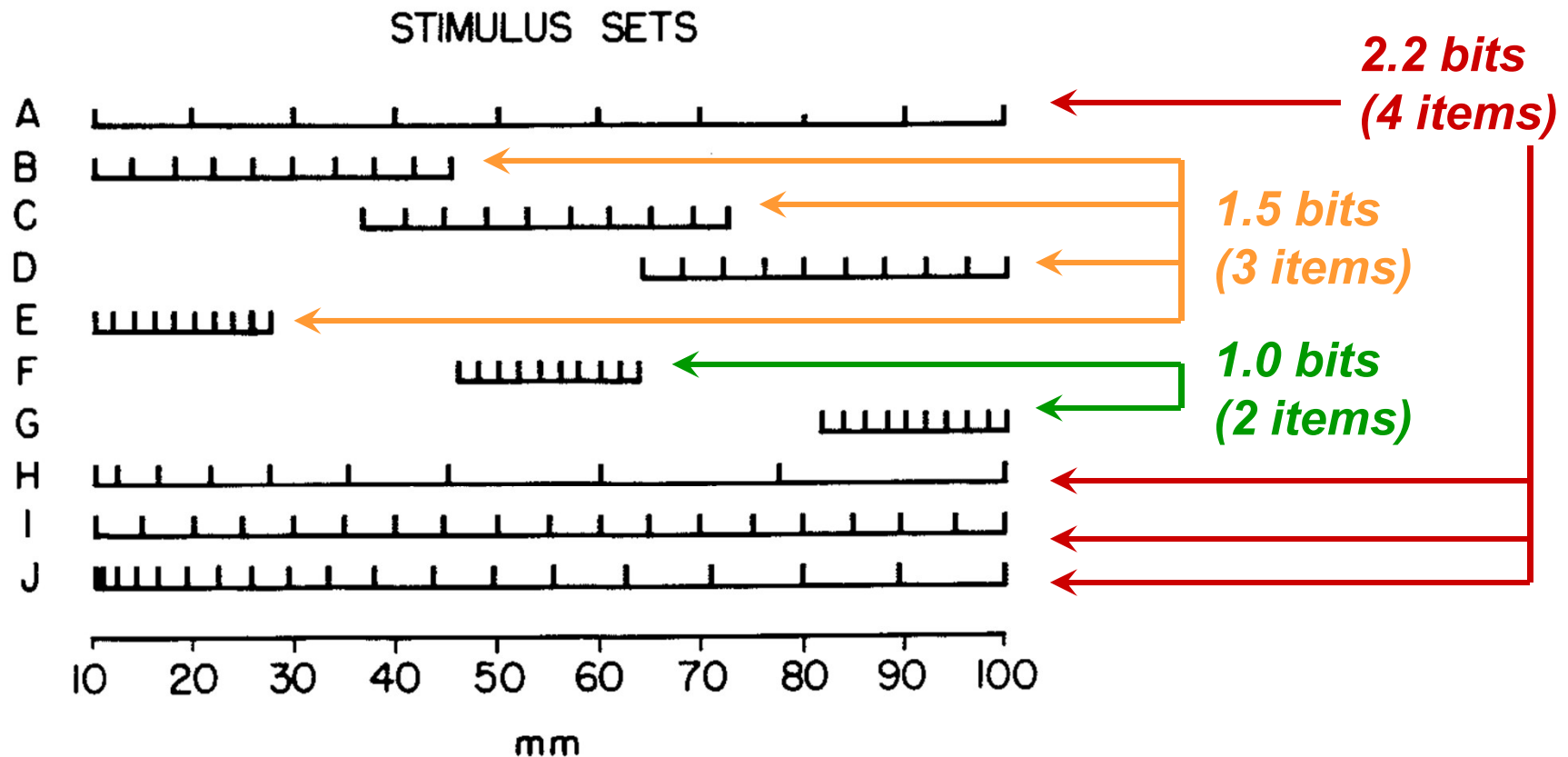
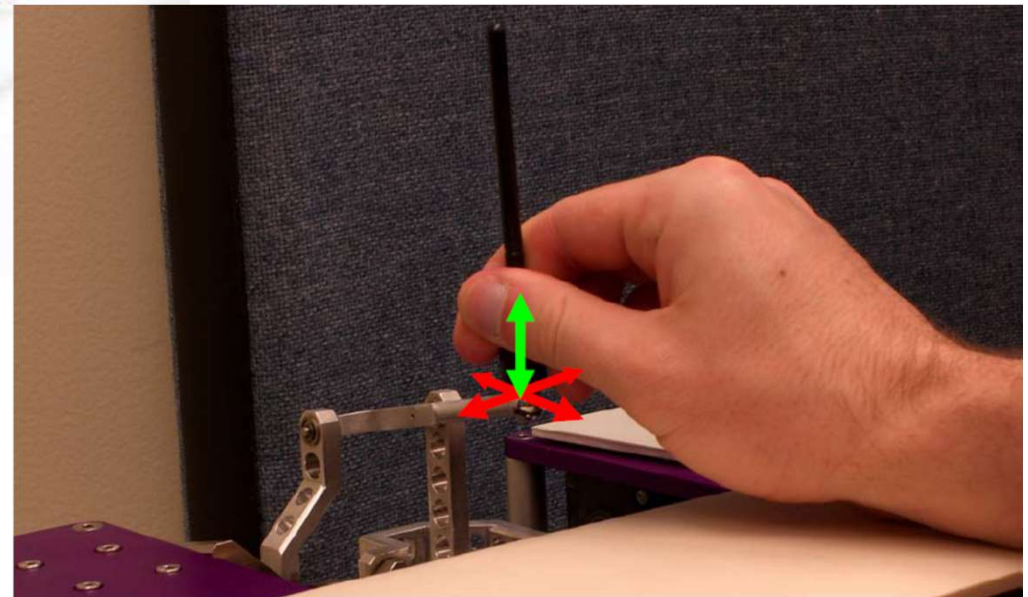
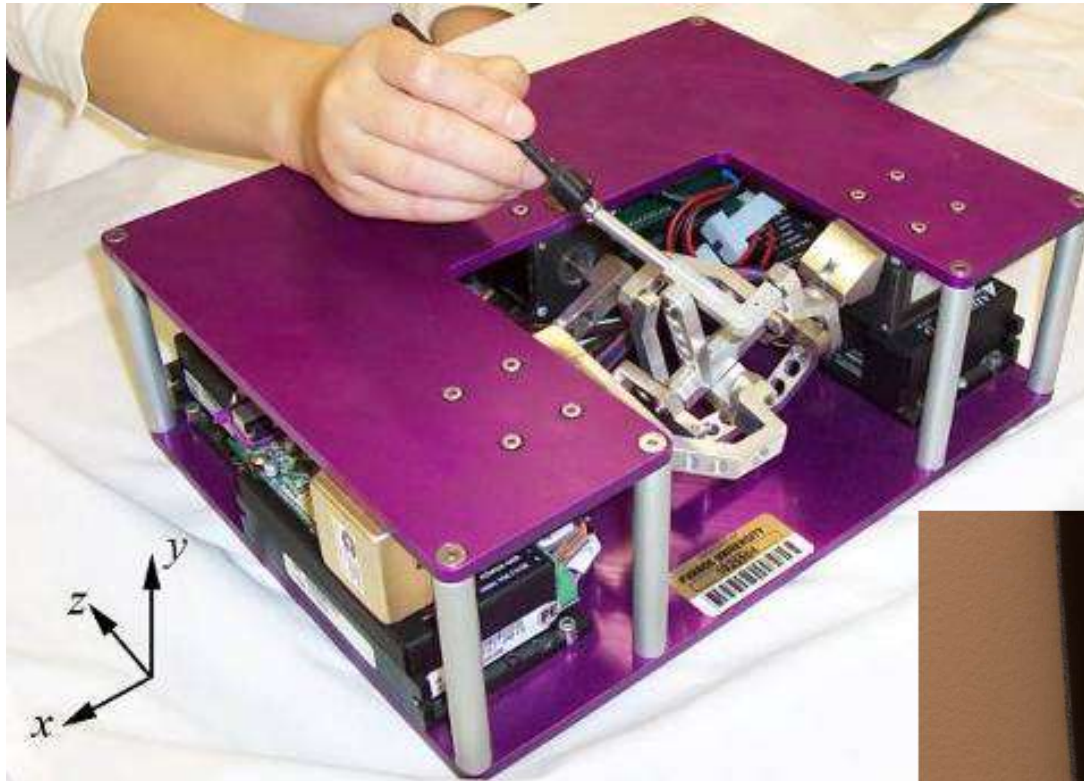


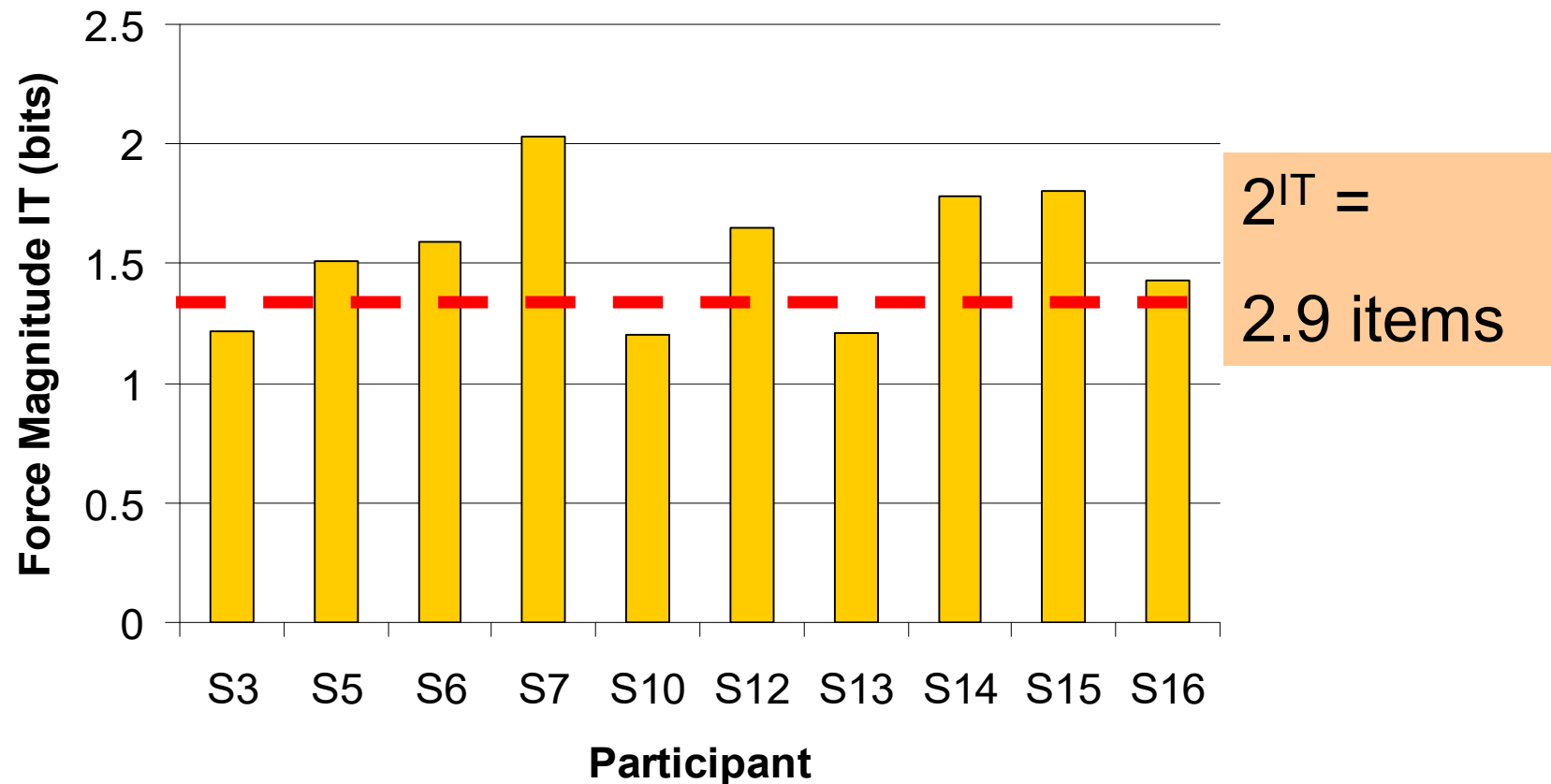
Figure 6. Stimulus sets used in the identification experiments (Experiments 2A and 2B). Sets A–G and I are linear; Sets H and J are approximately logarithmic. Each of Sets A–H have 10 elements; each of Sets I and J have 19 elements.

# Identification of Force/Stiffness: Towards Haptic “Glyphs”



# Force Identification Results

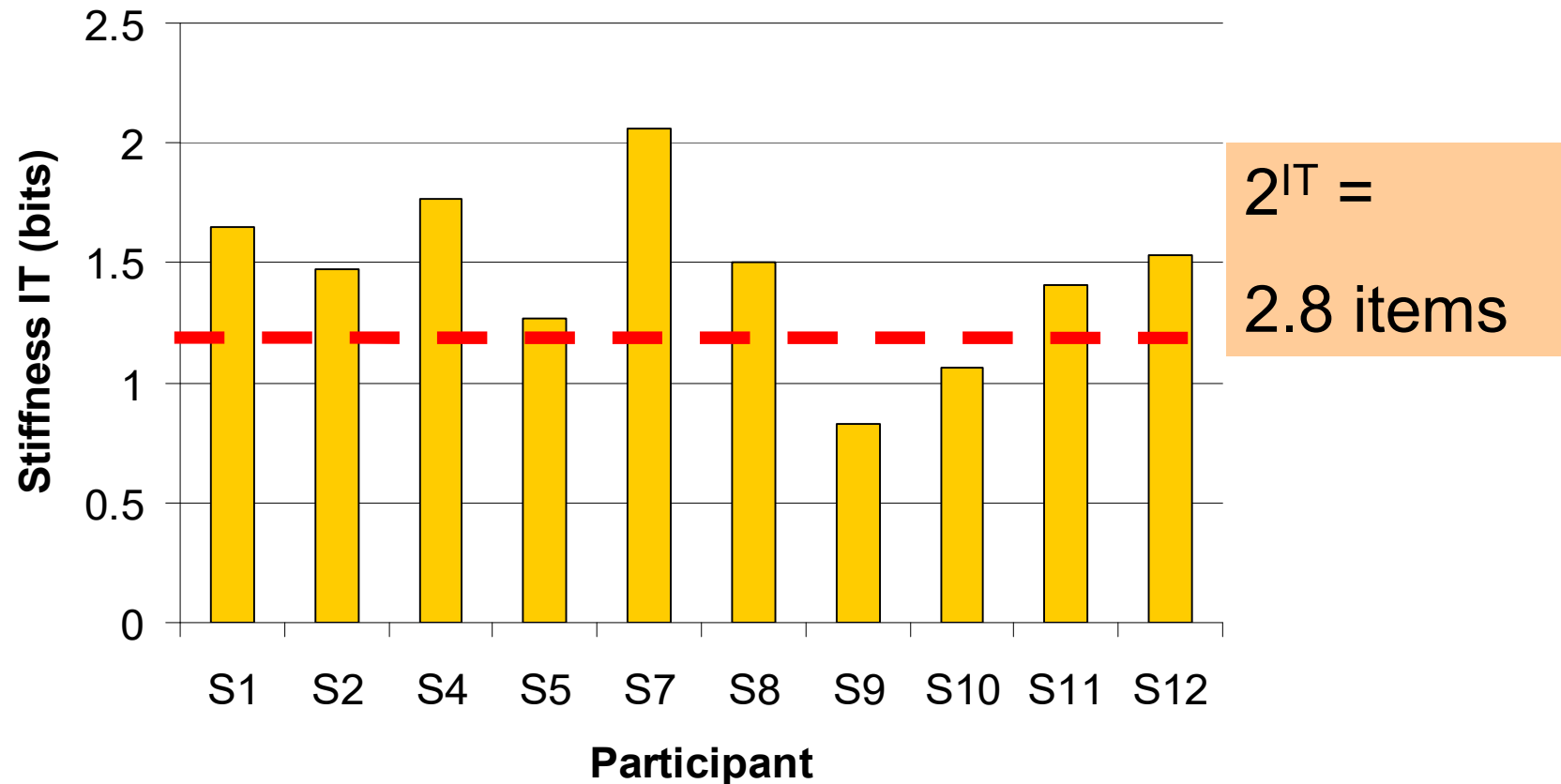
Force range: 0.1 - 5 N



Cholewiak, Tan, and Ebert (2008). "Haptic identification of stiffness and force magnitude," *Proceedings of the Symposium on Haptic Interfaces for Virtual Environment and Teleoperator Systems*, pp. 87-91.

# Stiffness Identification Results

Stiffness range: 0.2 - 3 N/mm



Cholewiak, Tan, and Ebert (2008). "Haptic identification of stiffness and force magnitude," *Proceedings of the Symposium on Haptic Interfaces for Virtual Environment and Teleoperator Systems*, pp. 87-91.

# Summary

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For tasks involving the **identification** of one of many stimulus alternatives,

- ❖ Conduct an absolute identification experiment using a number of alternatives ( $k$ ) that is expected to lead to errors (i.e., beyond channel capacity)
- ❖ Compute the stimulus-response confusion matrix for **each** participant
- ❖ Average estimated IT values across participants
- ❖ Compute  $2^{IT_{est}}$ , the total number of stimuli that can be **perfectly** identified; i.e., channel capacity

Then design your interface accordingly. There is no need to conduct more than one experiment.

# Reference

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Jones, L. A., & Tan, H. Z. (2013). Application of psychophysical techniques to haptic research. *IEEE Transactions on Haptics*, 6(3), 268-284.

# Contact Information

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