

Practical Issues in Designing and Conducting an AI Experiment

Five Issues to be Considered

- Range of stimulus parameter (R)
- Number of stimulus alternatives (k)
- Spacing between the k stimuli (*linear vs. log*)
- Total number of trials (n)
- Training procedure

Issue #1: Range of Stimulus Parameter (R)

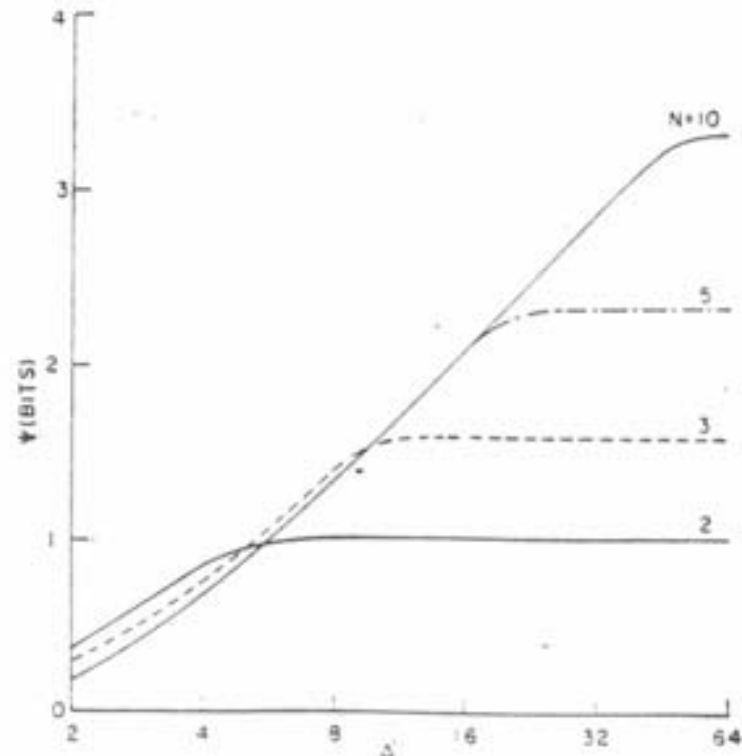


FIG. 1. Dependence of mutual information Ψ on the total sensitivity Δ' for various values of the number N of stimuli. (See text for assumptions.)

- **Problem**

- ◆ **At small R , where $R = \log(I_{\max}/I_{\min})$, information transfer < channel capacity**

- **Strategy**

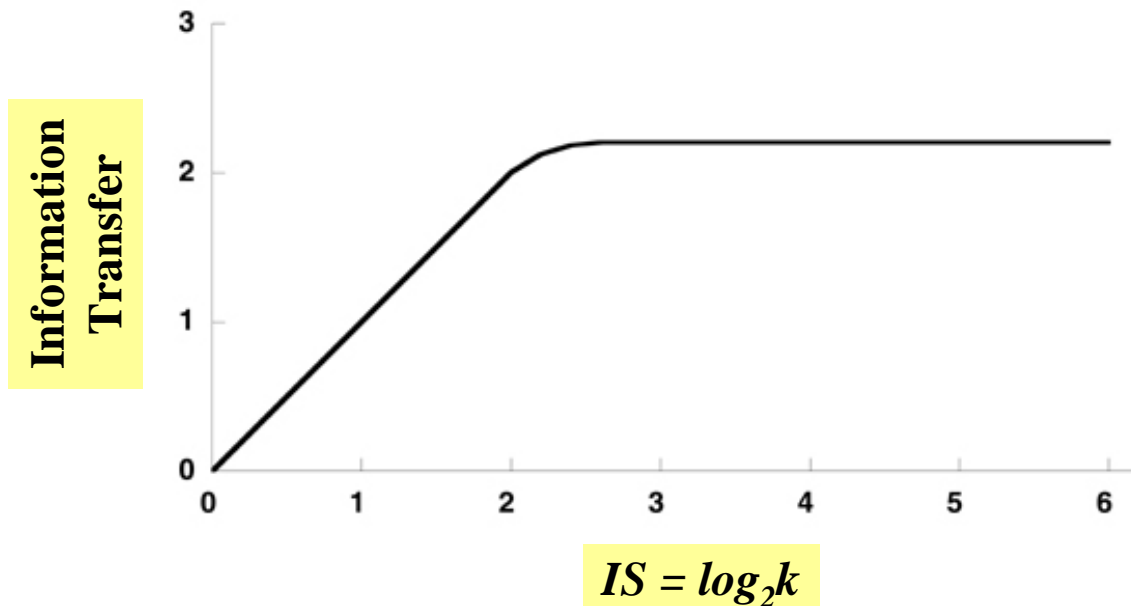
- ◆ **Use largest possible range given the experimental setup**

- **Examples of “largest possible range”**

- ◆ **Sound levels: AL to “too loud”**
- ◆ **Curvature: straight line to arc of the smallest circle that can be drawn**
- ◆ **Weight: AL to “too heavy”**

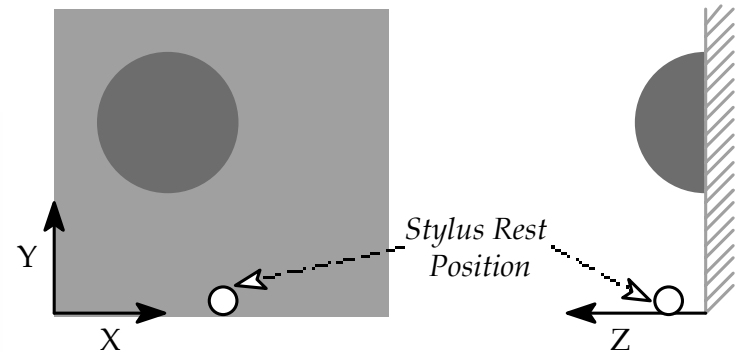
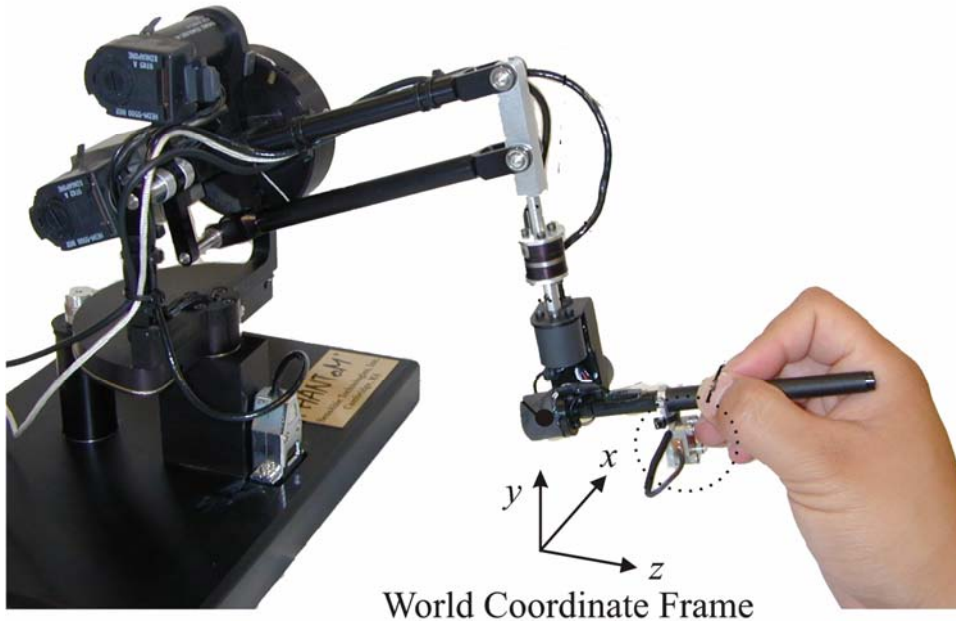
Issue #2:

Number of Stimulus Alternatives (k)



- Small k limits IT_{est}
- Large k requires too many trials
- One strategy: increase k until IT_{est} asymptotes

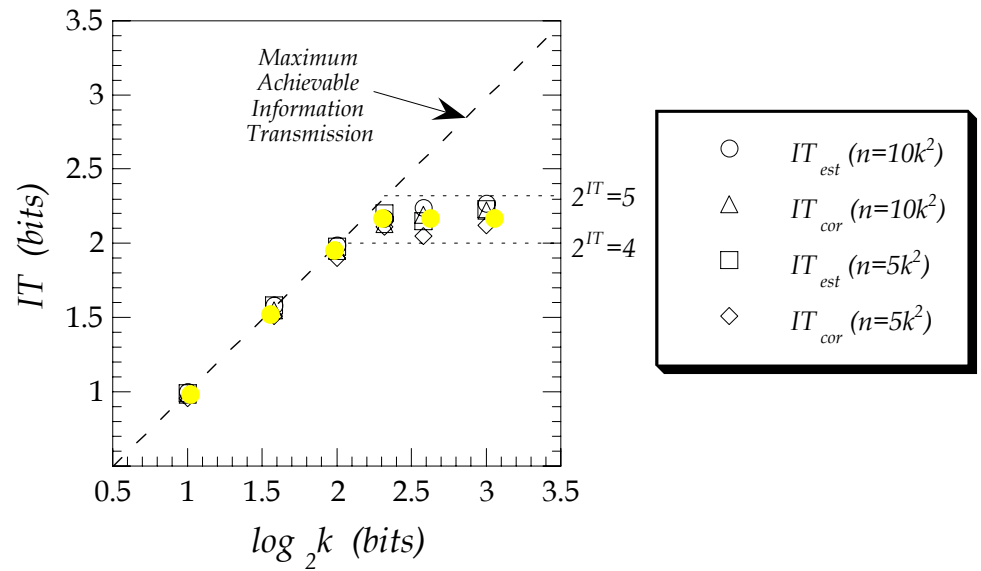
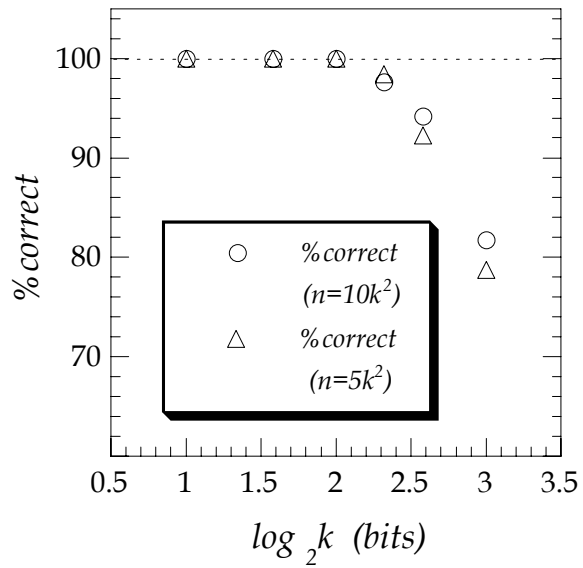
An Example (Tan, 1997)



(a) Front view.

(b) Side view.

**Range of radius:
10 – 80 mm
For different numbers
Of stimulus alternatives (k)**

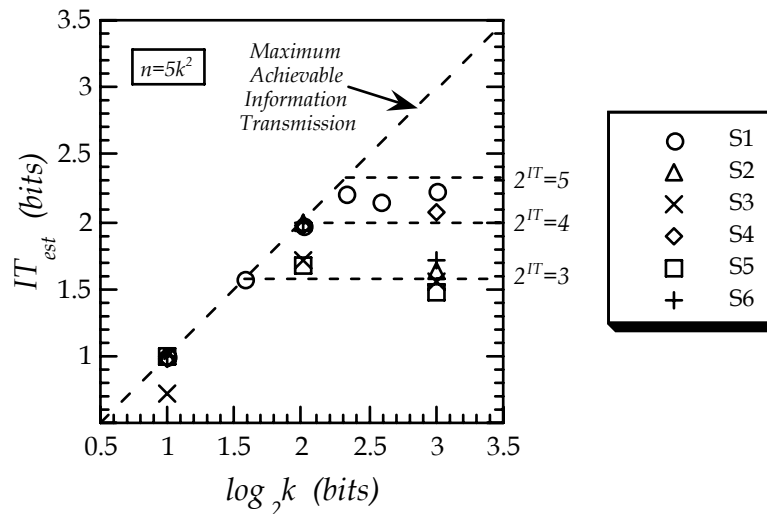


Issue #3: *Linear or Log Spacing?*

- **Objective**
 - ◆ **Equal perceptual distance between adjacent stimuli**
 - ◆ **If Weber's law applies, logarithmic spacing is preferred**
- **Problem**
 - ◆ **Many discrimination experiments are required before an absolute identification exp can be designed and conducted**
- **Lucky Solution**
 - ◆ **In most cases, *information transfer* is not sensitive to stimulus spacing**

An Example (Tan, 1997)

- Identification of sphere size:
 - ◆ Range of radius (fixed): 10.0 to 80.0 mm
 - ◆ Linear (e.g., $k=3$): 10.0, 45.0, 80.0 mm
 - ◆ Logarithmic (e.g., $k=3$): 10.0, 28.28, 80.0 mm



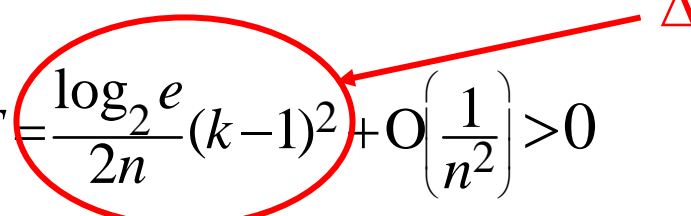
S1, S2, S3: Linear
S4, S5, S6: Log

Issue #4: How Many Trials?

- The issue:
 - ◆ IT_{est} is subject to statistical fluctuations
 - ◆ IT_{est} is biased ($E[IT_{est}] > IT$)
 - ◆ bias > sampling errors
- Need sufficient number of trials to overcome bias and sampling errors

Miller's (1954) Formula

- IT_{est} is an over-estimate of IT

$$E[IT_{est}] - IT = \frac{\log_2 e}{2n} (k-1)^2 + O\left(\frac{1}{n^2}\right) > 0$$


- With large n ($> 5k^2$), Δ is small (**0.14 bit**)
- With small n , Δ can over-correct

i.e., $IT_{est} - \Delta < IT$

Miller's (1954) Formula (*cont.*)

- When performance level is high, Δ over-corrects

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

$$P(C) = 100\%$$

$$IS = IR = IT_{\text{est}} = IT = 2 \text{ bits}$$

yet $\Delta \neq 0$

- Bottom line: $n \geq 5k^2$ is needed.

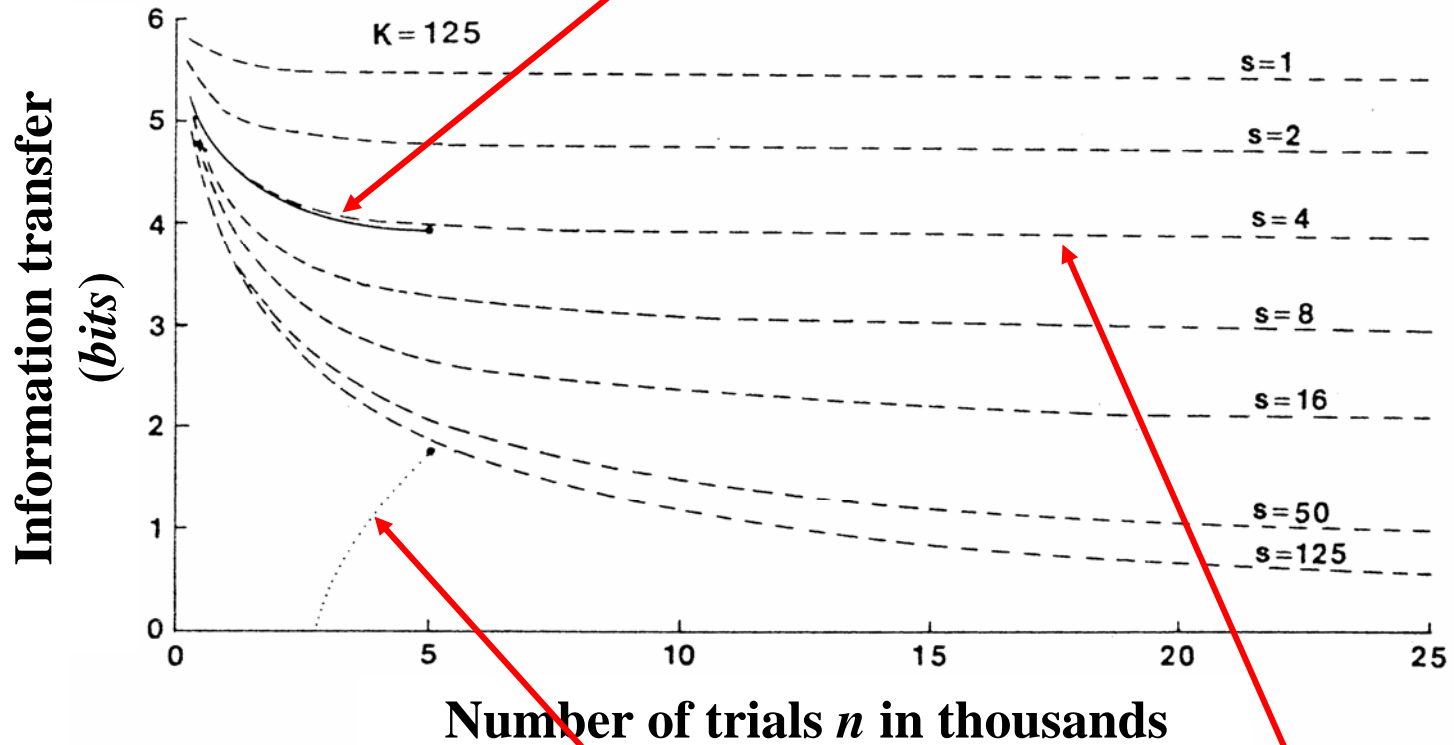
An Experiment where $k=125$ (Rabinowitz et al., *JASA*, 1987)

- **For $k = 125$, $5k^2 = 78,125$ total trials!!**
- **3-D stimulus set — pulsed sinusoidal vibration**
 - ◆ **Five values of intensity**
 - ◆ **Five values of contact area**
 - ◆ **Five values of frequency**
- **One-interval AI paradigm with feedback**
- **3-tuplets as responses (e.g., 111, 254, etc.)**
- **Data: 125-by-125 confusion matrix!!**

Houtsma's Computer Simulation (JASA, 1983)

- Assumption
 - ◆ 1-D experiment with $k=125$
- Procedure
 - ◆ Randomly pick an S from 1-125; $S \in [1, 125]$
 - ◆ $R_{raw} = S + noise (\pm s)$
 - ◆ R is reset to 1 or 125 if R_{raw} is too small or too large; $R \in [1, 125]$
 - ◆ Collect enough number of “trials”, n
 - ◆ Estimate IT_{est} as a function of n
- The value of s is used to control the asymptotic value of IT_{est}

Experimental data



Dotted line: $IT_{est} - \Delta$

True IT value

So How Many Trials are Enough?

- Collect $n \geq 5k^2$ trials if possible
- For one-dimensional stimuli, k is usually reasonable (7 ± 2).
- For multi-dimensional stimuli,
 - ◆ Additivity: $IT(\text{multi-D}) = \Sigma IT(ID)$?
Usually, $IT(A, F) < IT(A) + IT(F)$
 - ◆ A general additivity law (Durlach *et al.*, 1989)

Issue #5: Training

- **Training is usually needed for AI paradigms**
- **Criterion for termination of training**

References

- H. Z. Tan, “Identification of sphere size using the PHANToM™: Towards a set of building blocks for rendering haptic environment,” in *Proceedings of the ASME Dynamic Systems and Control Division*, vol. 61. Dallas, TX: American Society of Mechanical Engineers, 1997, pp. 197–203.
- G. A. Miller, “Note on the bias of information estimates,” in *Information Theory in Psychology*, H. Quastler (Ed.), 1954, pp. 95-100.

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- **W. M. Rabinowitz, A. J. M. Houtsma, N. I. Durlach, and L. A. Delhorne, “Multidimensional tactile displays: Identification of vibratory intensity, frequency, and contactor area,” *Journal of the Acoustical Society of America*, vol. 82, pp. 1243-1252, 1987.**
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- N. I. Durlach and L. D. Braida, “Intensity perception I. Preliminary theory of intensity resolution,” *The Journal of the Acoustical Society of America*, vol. 46, pp. 372–383, 1969.