

Rating Experiment

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Outline

- Procedure
- Decision Model
- Data Analysis

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Procedure

- Same as the procedure for 1-I experiments, *except that*
- **There are > 2 admissible responses**
- Subjects decide not only which one of the two stimuli was perceived to have been presented, but their confidence in such decisions.

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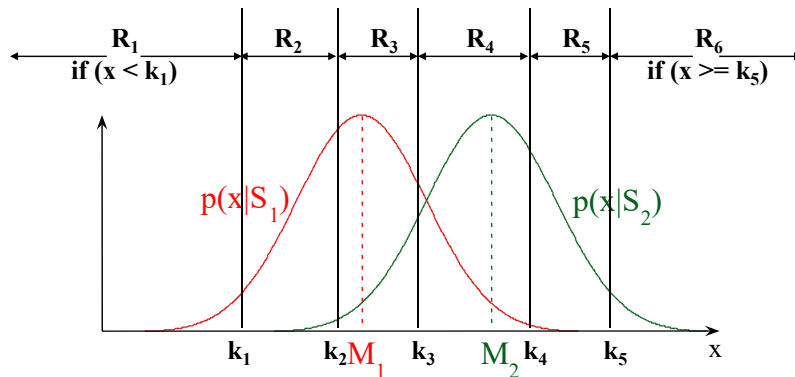
Procedure (*cont.*)

- Three types of response sets
 - ◆ Numerals – simplest
 - ◆ Verbal categories
 - ◆ 2 sub-responses
 - ☞ R_1 or R_2 , then
 - ☞ Grade the certainty of response with numerals or verbal categories
- All three types are equivalent in functionality

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Decision Model



$$d' = \frac{M_2 - M_1}{\sigma}$$

$$c_i = \frac{1}{\sigma} \left(k_i - \frac{M_1 + M_2}{2} \right)$$

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Data Analysis

- Stimulus-response matrix for experimental data
- Calculate d'
- Derive ROC (iso-sensitivity curve)
- Calculate response bias c_i
- **The key is to find the appropriate pair of (H, F) values, then**
 - ◆ $d' = z(H) - z(F)$
 - ◆ $c = - [z(H) + z(F)]/2$

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An Example (from M&C, chap.3)

- Word recognition (old vs. new, *high-frequency vocabulary*)
- “Detection” of old words
 - ◆ Old is S₂, New is S₁

	“old 3”	“old 2”	“old 1”	“new 1”	“new 2”	“new 3”	Total
Old	49	94	75	60	75	22	375
New	8	37	45	60	113	113	376

(from Table 3.1, M&C, Chap. 3, p. 1)

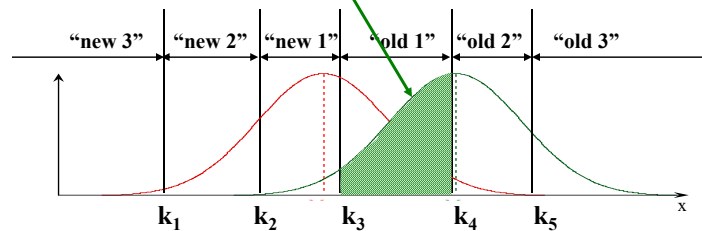
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Proportions

	“old 3”	“old 2”	“old 1”	“new 1”	“new 2”	“new 3”	Total
New (S ₁)	.021	.098	.120	.160	.301	.301	1.00
Old (S ₂)	.131	.251	.200	.160	.200	.059	1.00

(from Table 3.2, M&C, Chap. 3, p. 3)



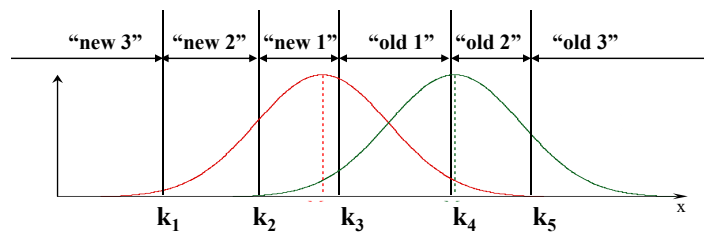
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Cumulative Proportions: Hit and False-Alarm Rates

	“old 3” (k_5)	“old 2” (k_4)	“old 1” (k_3)	“new 1” (k_2)	“new 2” (k_1)	“new 3” ($k = -\infty$)	Total
F	.021	.119	.239	.399	.700	1.00	1.00
H	.131	.382	.582	.742	.942	1.00	1.00

(from Table 3.3, M&C, Chap. 3, p. 4)



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Z-scores

	“old 3” (k_5)	“old 2” (k_4)	“old 1” (k_3)	“new 1” (k_2)	“new 2” (k_1)	“new 3” ($k = -\infty$)	Total
z(F)	-2.035	-1.180	-0.705	-0.255	0.525	-	1.00
z(H)	-1.125	-0.300	0.205	0.645	1.575	-	1.00

(from Table 3.4, M&C, Chap. 3, p. 5)

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Calculation of d'

- Two conceptually different methods
 1. Collapse the data matrix into a 2×2 matrix, by combining responses for the same stimulus

	"old 3"	"old 2"	"old 1"	"new 1"	"new 2"	"new 3"	Total
Old	49	94	75	60	75	22	375
New	8	37	45	60	113	113	376

(from Table 3.1, M&C, Chap. 3, p. 1)

	"old"	"new"
Old (S_2)	218	157
New (S_1)	90	286

	R_1	R_2
S_1	286	90
S_2	157	218

$$d' = 0.910$$

$$c = 0.25$$

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Calculation of d' (cont.)

- Two methods (*cont.*)
 2. Calculate d' from the many (H, F) values

(from Table 3.4, M&C, Chap. 3, p. 5)

	"old 3" (k_5)	"old 2" (k_4)	"old 1" (k_3)	"new 1" (k_2)	"new 2" (k_1)	"new 3" ($k = -\infty$)	Total
$z(F)$	-2.035	-1.180	-0.705	-0.255	0.525	-	1.00
$z(H)$	-1.125	-0.300	0.205	0.645	1.575	-	1.00
d'	0.910	0.880	0.910	0.900	1.050	-	-

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Data Analysis

- Stimulus-response matrix for experimental data
- Calculate d'
- Derive ROC iso-sensitivity curve
- Calculate response bias c_i
- The key is to find the appropriate pair of (H, F) values, then
 - ◆ $d' = z(H) - z(F)$
 - ◆ $c = - [z(H)+z(F)]/2$

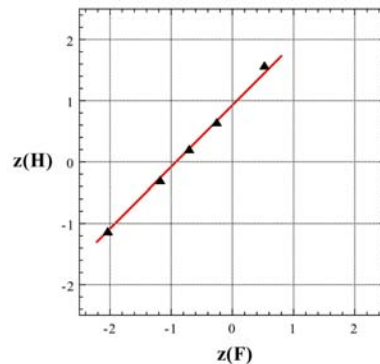
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Plotting the ROC

	k_5	k_4	k_3	k_2	k_1
$z(F)$	-2.035	-1.180	-0.705	-0.255	0.525
$z(H)$	-1.125	-0.300	0.205	0.645	1.575

(from Table 3.4, M&C, Chap. 3, p. 5)



$$z(H) = z(F) + d'$$

$$d' = 0.93$$

(from Figure 3.1, M&C, Chap. 3)

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More on Estimation of $\sigma_{d'}$

- This is the “other” way to construct a ROC curve for estimating $\sigma_{d'}$
 - ◆ See lectures notes on “how to calculate $\sigma_{d'}$ ”
 - ◆ Fitting based on minimum rms error is not appropriate because both abscissa and ordinate are *dependent* variables
 - ◆ Use *maximum likelihood estimation* method
 - ◆ See M&C, chap.3, p. 13, for reference to the Dorfman & Alf’s algorithm

Data Analysis

- Stimulus-response matrix for experimental data
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- Derive ROC iso-sensitivity curve
- Calculate response bias c_i
- The key is to find the appropriate pair of (H, F) values, then
 - ◆ $d' = z(H) - z(F)$
 - ◆ $c = - [z(H)+z(F)]/2$

Calculation of c_i

(from Table 3.4, M&C, Chap. 3, p. 5)

	“old 3” (k_5)	“old 2” (k_4)	“old 1” (k_3)	“new 1” (k_2)	“new 2” (k_1)	“new 3” ($k = -\infty$)	Total
z(H)	-1.125	-0.300	0.205	0.645	1.575	–	1.00
z(F)	-2.035	-1.180	-0.705	-0.255	0.525	–	1.00
d'	0.910	0.880	0.910	0.900	1.050	–	–
c_i	1.58	0.74	0.25	-0.195	-1.05	–	–

Reading

- Chap. 3 of Macmillan and Creelman's book