

# Measuring Specific, Rather than Generalized, Cognitive Deficits, and Maximizing Discriminating Power in Studies of Cognition and Cognitive Change

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# Overview

- Specific versus generalized deficit
- Strategies for avoiding confounds resulting from a generalized deficit
- Optimizing effect size in between-groups comparisons: reliability, within-group variation and between-group variation
- Summary: Tradeoffs

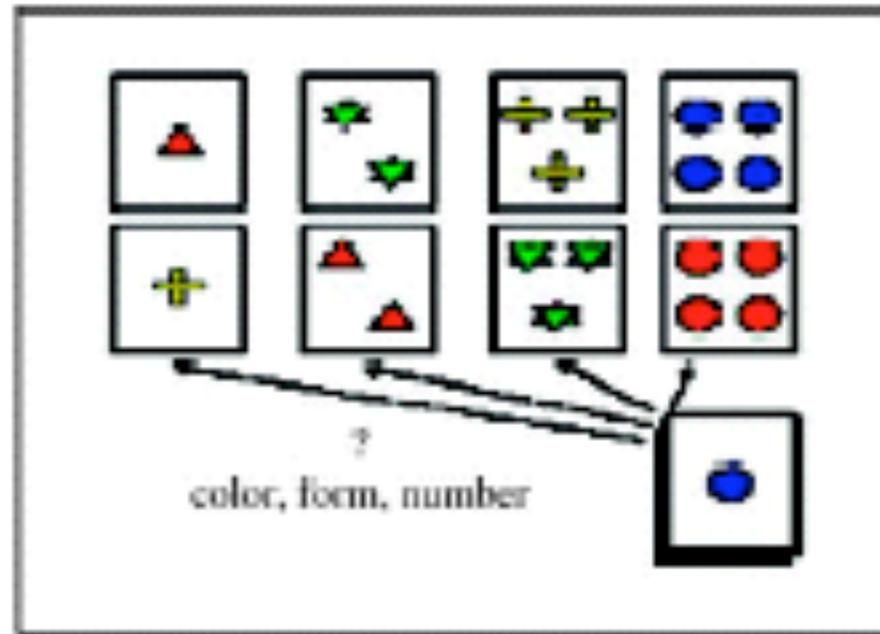


# Obstacles to Isolating Specific Impairments

- Neuropsychological tests are generally confounded by multiple cognitive processes.
- Poor performance can be due to a variety of cognitive and non-cognitive factors.
- Differences in psychometric properties of tests can affect our interpretation of cognitive abilities.

# Example of Multifactorial Nature of Neuropsychological Test

(from C. Carter, 2005, Scz. Bull)



Feedback Processing

Executive Functions

Error Based Learning

Set Shifting

Selective Attention

Working Memory

Response Selection/Inhibition

- Multifactorial tests can be represented as:

$$- z_j = a_{j1}s_1 + a_{j2}s_2 + \dots + a_{jp}s_p + \dots + a_{jm}s_m + e_j E_j$$

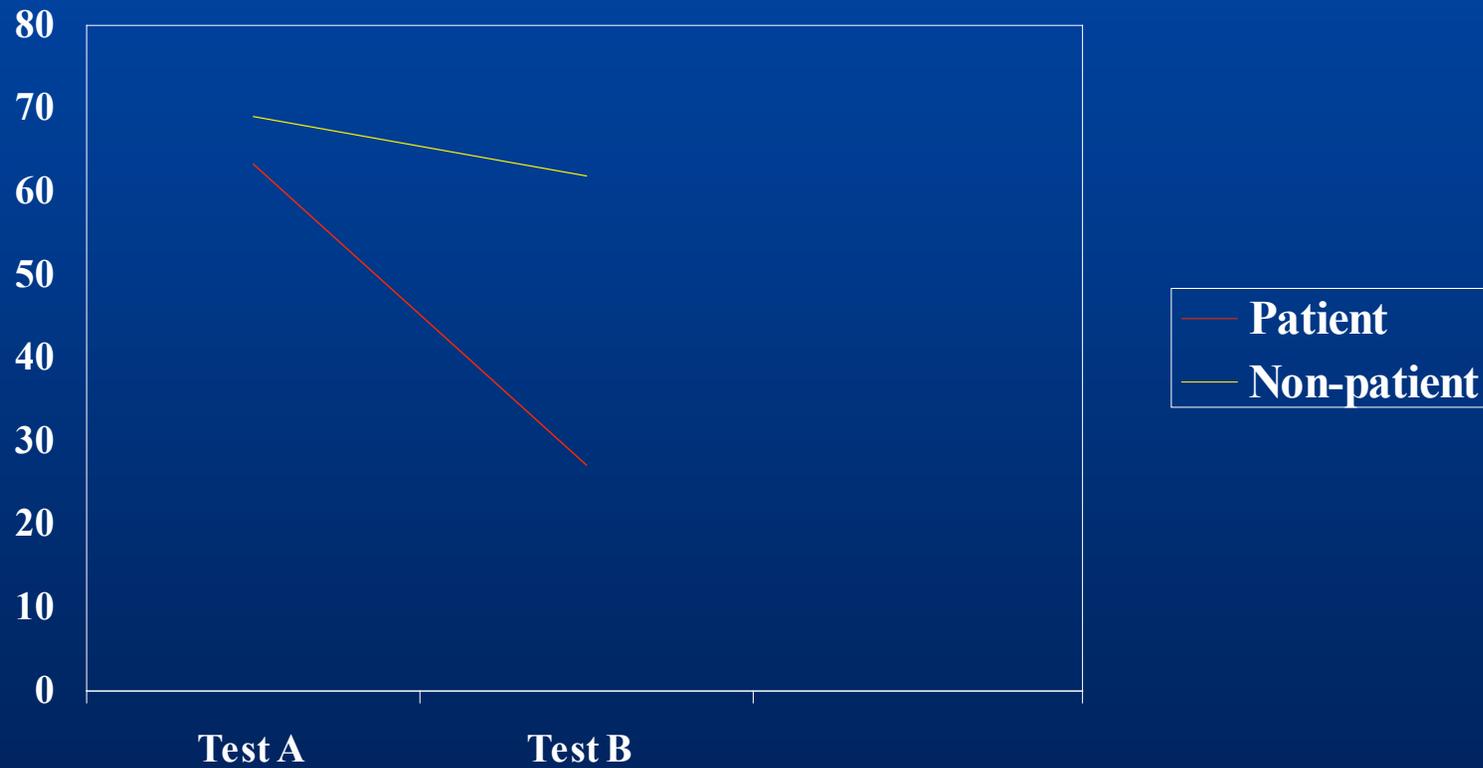
- $z_j$  = individual's standardized score on test  $j$
- $s_p$  = true score for source of variance  $p$
- $a_{jp}$  = influence of variance source  $p$  on test  $j$
- $E_j$  = sources of measurement error on  $z_j$
- $e_j$  = influence of  $E_j$  on  $z_j$  (Neufeld, 1984)

- We want:  $z_j = a_{jp}s_p + e_jE_j$
- We need to either:
  - eliminate all ‘non-specific’ sources of true score variance (s),  
or
  - minimize effects of these sources  
(a) on test scores

# Strategies to Isolate Cognitive Deficits



# Differential Deficit



## But....

- A differential deficit could be due to greater discriminating power of 1 of the tests.
- A test that is more reliable, and/or more difficult will discriminate between subjects better than a less reliable or less difficult test.

## A differential deficit is only meaningful if:

- the patient group achieves superior performance on 1 of the tests.
- differences between groups are greater on the less discriminating task, and/or
- both tests have equivalent reliability and difficulty levels (Chapman & Chapman, 1978; Strauss, 2001)

# Problems with Task Matching

- Matching on reliability and difficulty does not ensure construct validity (process specificity)
- Matching on difficulty level is a problem for cognitive neuroscience tasks where parameter manipulations change difficulty levels
- Matching does not maximize between-groups discriminating power (Knight & Silverstein, 2001)

# Reliability and Discriminating Power

- Reliability:  $r_{xx} = \sigma_t^2 / \sigma_o^2$  or (=)  $\sigma_t^2 / [\sigma_t^2 + \sigma_{me}^2]$
- Reliability of a test can be increased by:
  - reducing measurement error ( $\sigma_{me}^2$ )
  - increasing true score variance ( $\sigma_t^2$ )
- Reducing  $\sigma_{me}^2$  will reduce within-group variance, and increase sensitivity to between-groups sources of variance.

- Increasing  $\sigma_t^2$  will increase within-group variance/discrimination, but if it does not also increase between-groups discrimination, power will decrease (Neufeld, 1984).
- It has been shown that, for 2 tests of the same construct that differ by as much as 3x in  $\sigma_t^2$ , the test with higher  $\sigma_t^2$  was associated with a lower between-group effect size, due to  $\sigma_t^2$  being increased by mainly focusing on processes that increase within group variation but that are not related to between group discrimination.



- Magnitude of between-group difference can be expressed as  $(c\tau + \beta) / (\tau + e)$ , where
  - $\beta$  is the effect of a variable unique to group membership
  - $\tau$  represents effects of other variables that generate variance within-groups,
  - $c$  represents overlap between  $\tau$  and  $\beta$  (Neufeld, 2007)
- In standardization sample,  $c$  and  $\beta$  are irrelevant, within group discrimination =  $\tau / (\tau + e)$ , and we want to maximize  $\tau$ .
- But, “a measure becomes less group-discriminating as its standardization-group psychometric precision goes up” (Neufeld, 2007; also Cohen, 1988).

$$(c\tau + \beta) / (\tau + e)$$

- Where group separation is a function primarily of  $\beta$ , power goes up as  $\tau$  goes down.
- As  $\tau$  increases, power goes up as  $c$  goes up.
- But, increasing  $\tau$  is only beneficial to between-group discrimination when  $\beta < c^*e$ .
- Less reliable tests with higher  $c$  values can be more (between-group) discriminating than more reliable tests with low  $c$  values.

# Similar Issue With Increasing Task Length

- Adding trials to a task may increase test-retest reliability, but can reduce between-group discrimination if new items are associated with sources of within-group variance that are independent of  $\beta$ .
- Increasing task length is OK only if the test is unifactorial, or covariance structure of the task does not change with added items.
- However, this can add significant time and cost to clinical trials.

- Neither matching on reliability and difficulty, nor maximizing within-groups true score variance (i.e., individual differences) ensures either that a specific process is being measured, or that between-groups discriminating power is maximized.



# Alternative Strategies - I

- ANCOVA
  - typically not appropriate as a control for another cognitive process as represented by a second task score.
  - assumes independence of covariate and IV (group)
  - most appropriate when there is random assignment to groups. It was designed to reduce within-groups variance rather than between-groups variance.
- IRT
  - requires large samples to construct measures
  - cannot resolve the issue that a focus on  $\tau$  and  $e$  cannot ensure a match on group discriminating power.
  - Assumes that item parameters do not differ across groups.

# Alternative Strategies - II

- Profile analysis
  - this vulnerable to same psychometric artifacts as differential deficit strategy
- Aggregation of scores into cognitive subdomains
  - exacerbates effects of  $\sigma_{me}^2$  and  $\tau$
- PCA, Factor analysis, and cluster analysis
  - Tests with the same confound may load on the same factor/cluster, confounding interpretation
  - Can be useful for understanding factor structure of single tests

# Alternative Strategies - III

- Partially ordered classification models\* (Jaeger, et al., 2006, *Schizophrenia Bulletin*)
  - Useful with neuropsychological battery data
  - Assumes that tests are multifactorial and accommodates this by organizing test scores into a conceptual network, based on the cognitive functions that are shared between tests, and functions that are unique to tests. Patients are then classified as belonging to 1 functional state in this network, based on their test scores, and Bayesian analysis techniques are used to determine the likelihood that these assignments are correct.
  - Would not be necessary with unifactorial tests

# Simplest Poset: 2 States

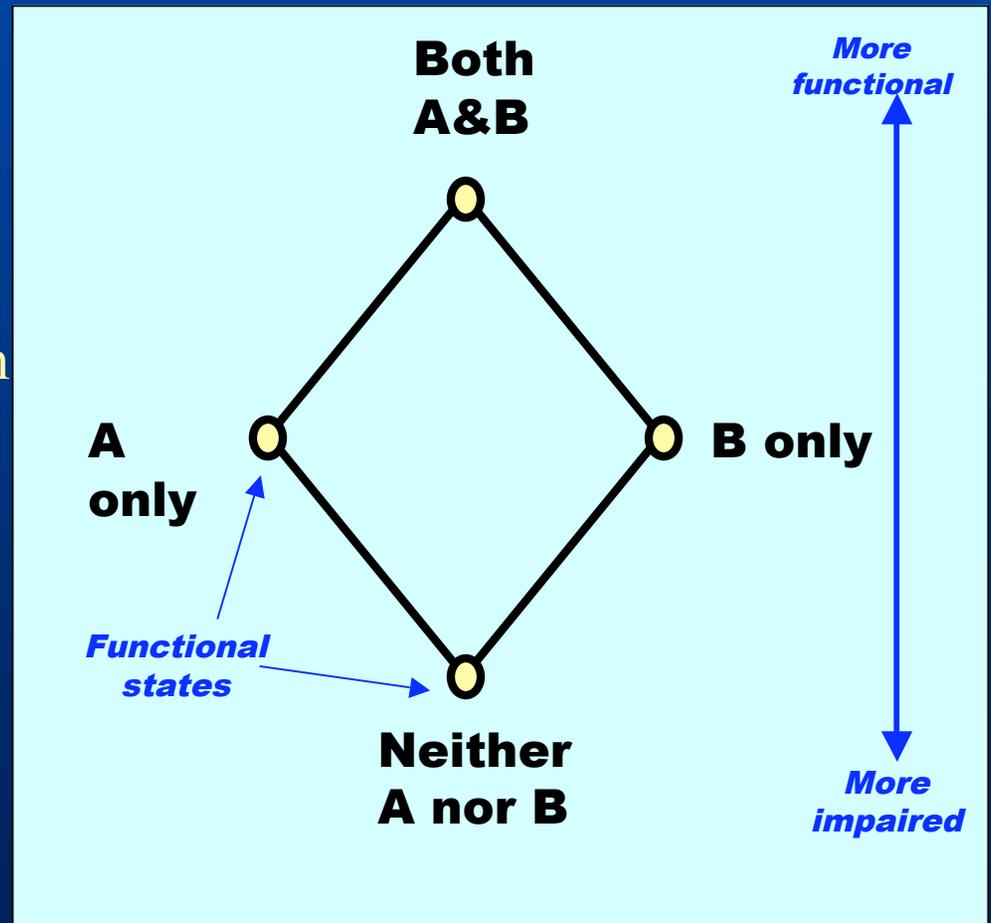
(this slide contributed by Judith Jaeger)

- These states can be viewed as belonging to a partially ordered set (i.e. poset)
- Some states have higher (cognitive) functionality than others. Others are not directly comparable.
- In typical application, more tests are used and more network states are present.

Example: A & B are attributes

Let A=Memory

Let B=Attention



# Process-Oriented Strategies

(Knight, 1984, 1992; Knight & Silverstein, 1998, 2001 *J. Abnormal Psychology*)

- Guided by theoretical models that make specific, falsifiable predictions, that can be tested against other hypothesis.
- Tasks typically include multiple conditions where specific parameters are varied to probe the integrity of an underlying process.
- Adequacy of the target process is understood in terms of the pattern of scores across conditions, or the pattern of psychophysiological correlates.
- Superiority and relative superiority are strongest findings.

# Example of a Process-Oriented Task Involving a Relative Superiority Prediction

(Silverstein et al., 1996 *J of Abnormal Psychology*)

- Different patterns of RT predicted for schizophrenia inpatients with poor premorbid functioning compared to other patients
- Example of relative insensitivity to perceptual organization reflected in a display size effect, in contrast to other groups.

# Examples of Stimuli in Target Detection Task

Condition 1

Condition 2

Condition 3

Condition 4

Condition 5

7 7

7 7 7

7 7

7 7 7

7 7

7

7 7

7 7

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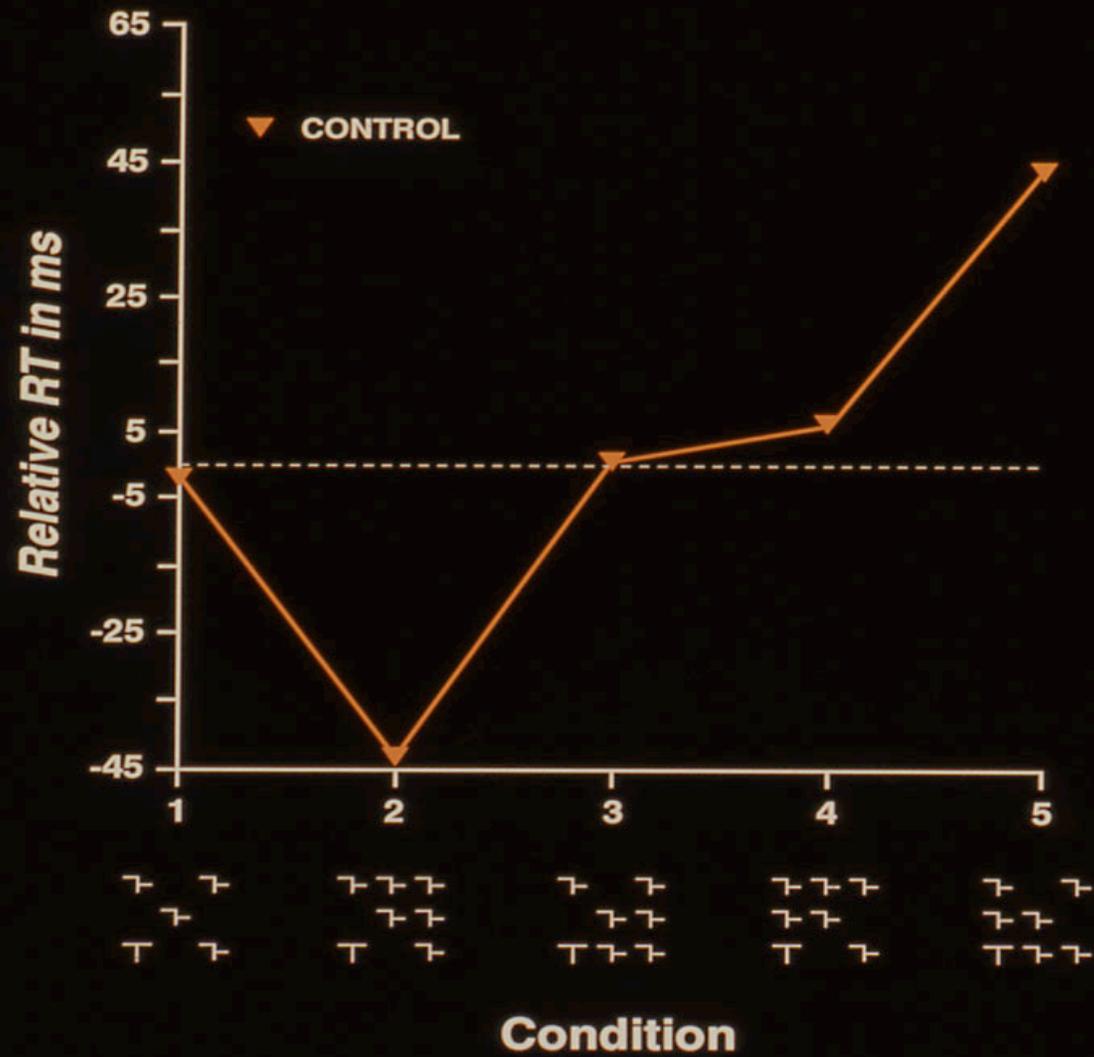
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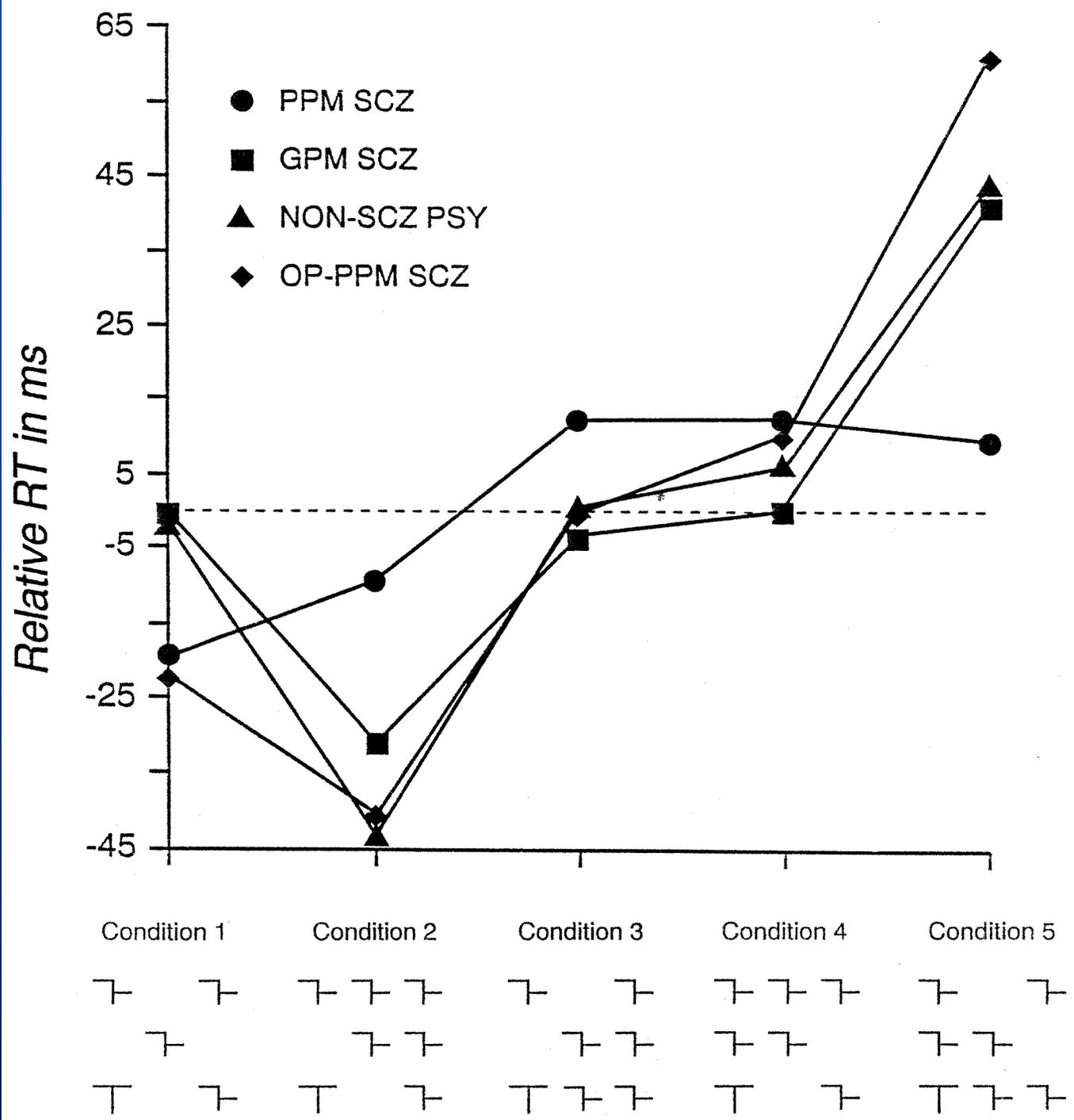
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## RT Pattern Predicted for Control Groups in Target Detection Task





- Examples of superiority or relative superiority are found in multiple cognitive domains [e.g., latent inhibition, working memory (AX-CPT), language (increased semantic priming, reduced negative priming, greater disambiguation for low-probability sentence endings), auditory and visual perception (reduced flanker interference effects, reduced perceptual grouping leading to more accurate judgements about features, etc.)]
- Development of more process-oriented tasks, in more cognitive domains, will allow for greater process specificity, and stronger cognition-neurobiology links.

# An Issue in Multiple Condition Comparisons: The Use of Difference Scores

- Reliability of gain scores:  $\rho_{gg'} = \rho_{xx'} - \rho_{12} / 1 - \rho_{12}$ 
  - $\rho_{xx'}$  = average reliability of pretest and posttest measures
  - $\rho_{12}$  = correlation between the pre- and post-tests  
(Lohrman, 1999).
- It was assumed that adequate validity required high  $\rho_{12}$  (trait stability), so low  $\rho_{gg}$ .
- When there is little change among people, or if all people change to a similar degree, the reliability of difference scores will be low.

- However, when there is heterogeneity in true change:
  - » There is low or moderate  $\rho_{12}$
  - » Reliability of difference scores can be high

$$\rho_{gg'} = \rho_{xx'} - \rho_{12} / 1 - \rho_{12}$$

↓                      ↓

High	→	.75 = (.8 - .2) / (1 - .2)
Low	→	.33 = (.8 - .7) / (1 - .7)

# Issues With Reliability of Change Scores

(Willett, 1989, 1994, 1997)

- Differences between conditions may be heterogeneous across people, even when a test is perfectly construct valid
- Under these conditions, the reliability of a difference score can be higher than the reliabilities of the individual scores that make up the index.
- The critical issue is whether we can understand/model the change in terms of relevant processes.

# Increasing Sensitivity to Change

- Characterization of change across more than 2 conditions, via slope, non-linear functions, or other multivariate methods (e.g., slope, mean, variability around trend line\*), will increase sensitivity
- Standard errors are reduced
- Reliability of change measurement is increased as measurement points are added (Willett, 1989, 1994, 1997)
- Appropriate modeling of covariance structure further increases sensitivity
- Cluster analysis can be useful to identify subgroups of subjects in 3-D space\*, to identify factors responsible for heterogeneity in degree of change (either across conditions within a task, or across time with multiple testing points).

# Summary: Tradeoffs

- Increased measurement sensitivity via increasing number of test conditions vs. ensuring adequate numbers of trials for within-condition measurement
- Measurement of full range of construct vs. optimizing discriminating power in each condition
- Individual difference discrimination vs. between-group discrimination
- Test-retest reliability/stability vs. sensitivity to change
- Construct validity vs. test-retest reliability
- Process-oriented designs vs. task/condition-matching
- Staircase procedures vs. standardized trial presentation

