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och Metrologi inom hälsovetenskaperna 2019

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<https://www.hkr.se/pmhealth2019>



**Six decades of measurement using the Rasch model  
in the social sciences –  
Setting the agenda for the next sixty years**

(and a )

*("The future glimmers long before it comes to be")*



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## Personal Background

### Why Measurement in the Social Sciences?

### Numbers and Measurement – The Role of the Measurement Model A Very Simple Approach

#### Requirements of Measurement

#### Does the Very Simple Approach Deliver?

#### Is Measurement in the Social Sciences Different from Measurement in the Natural Sciences?

#### A not Quite so Simple Approach

#### Comparing CTT, IRT and the Rasch Model/RMT

#### A case for the Rasch Model/RMT

#### The Rasch Model: the first 60 years

#### The Rasch Model: the next 60 years

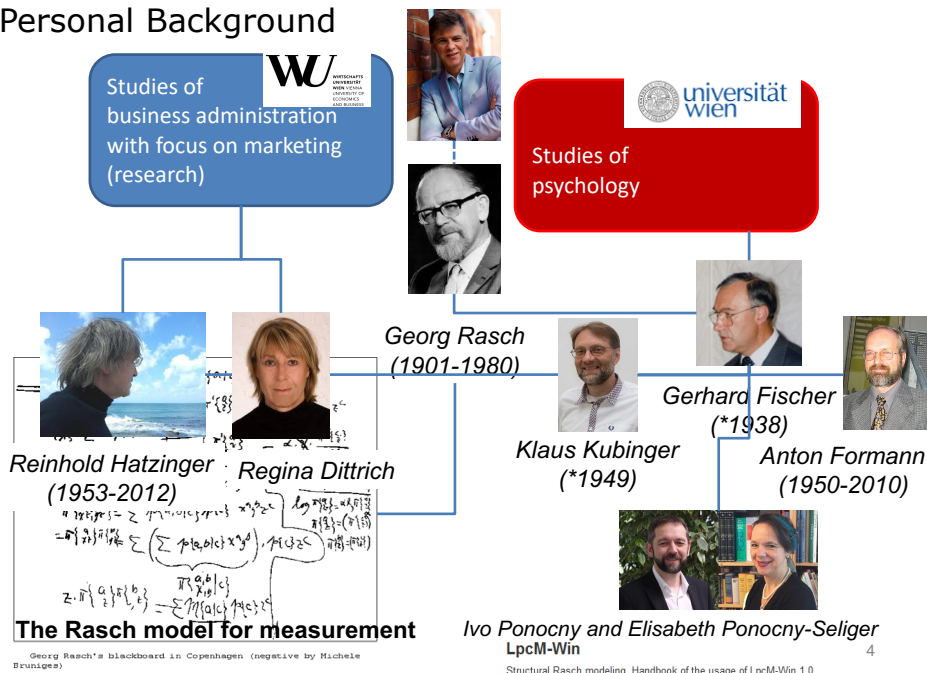
#### The Rasch Model and Metrology

#### Revisiting Feynman

# A G E N D A

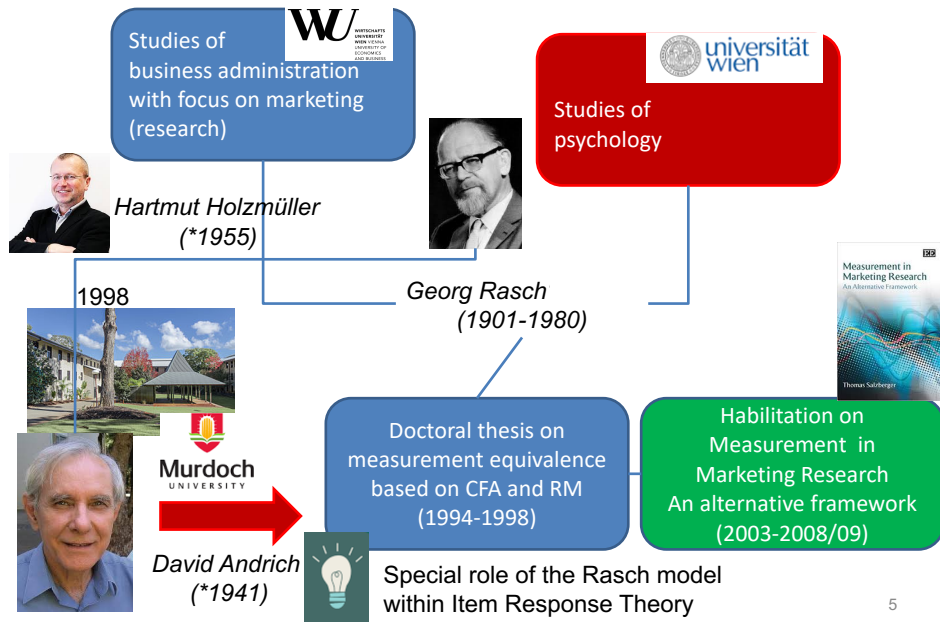
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## Personal Background



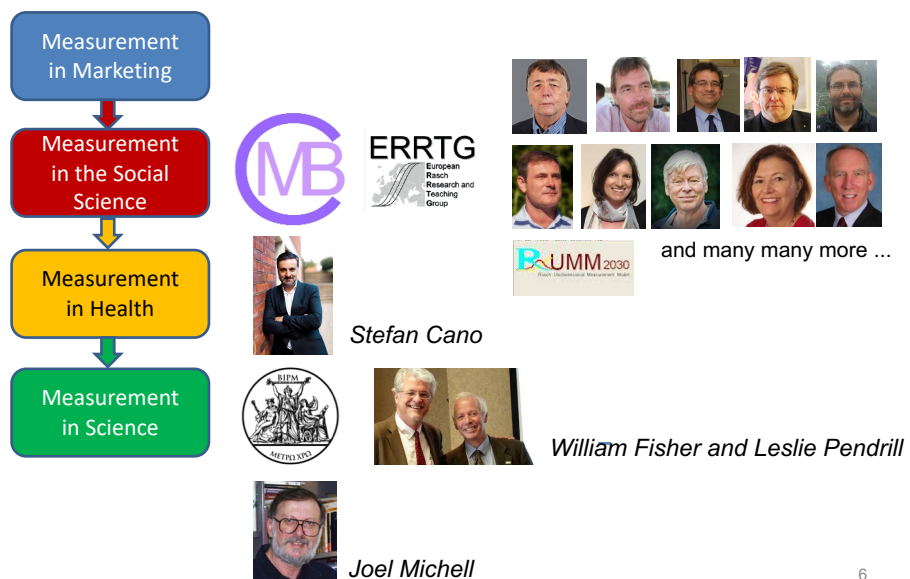
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## Personal Background



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## Personal Background



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## Why Measurement?

Quantitative approach to science proved enormously successful in the natural sciences.



*"The world is built upon  
the power of numbers."*

Pythagoras

*"All is number."*

Explaining the world using  
mathematics.

Mathematics: "that which is  
learned"

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## Why Measurement in the Social Sciences?

Quantitative science has become the role model for the social sciences

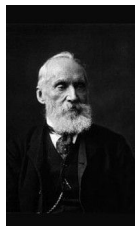
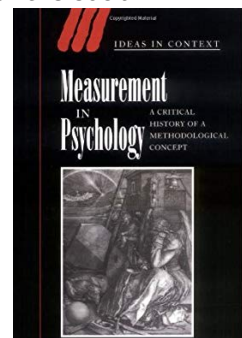
Quantitative imperative (Joel Michell)



*"Measure what can be measured,  
and make measurable what cannot be  
measured"*

- Galileo Galilei (perhaps)

*"Two things are infinite: the universe and human ingenuity  
to make up quotes; and I'm not sure about the universe."*  
Max Planck



I often say that when you can measure what you are speaking about, and express it in numbers, you know something about it; but when you cannot measure it, when you cannot express it in numbers, your knowledge is of a meagre and unsatisfactory kind: it may be the beginning of knowledge, but you have scarcely, in your thoughts, advanced to the stage of science, whatever the matter may be.

(Lord Kelvin)



## What Does Measurement Provide?

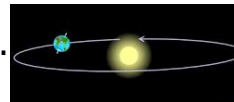


Measurement is an important **epistemological window** into reality

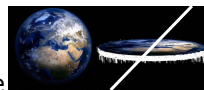
Effectiveness and trustworthiness of **quantitative science** hinges on proper measurement and substantive theory

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Today, the Earth Orbits the Sun.  
It wasn't always so. Or was it?



Roman Catholic doctrine:  
Earth at the centre of the universe  
Echoed Greek geographer Ptolemy (2<sup>nd</sup> century AD)



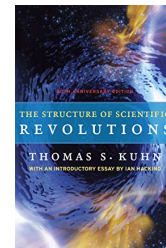
Even though Aristarchus of Samos (3<sup>rd</sup>/2<sup>nd</sup> century BC) proposed that the Earth orbits the Sun

But no empirical evidence was available  
Theory on its own is pointless



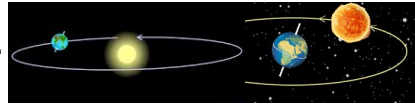
Heliocentric theory only revived in the  
16<sup>th</sup> century by Copernicus  
New empirical evidence: Mathematics &  
Observations confirmed heliocentric theory

And yet did Copernicus dare to publish his results  
only in 1543 - the year of his death



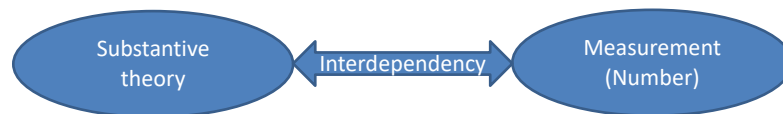
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Today, the Earth Orbits the Sun.  
It wasn't always so. Or was it?



Galilei collected further evidence of heliocentric theory  
Galilei's contribution (among others) was the use of mathematics in physics

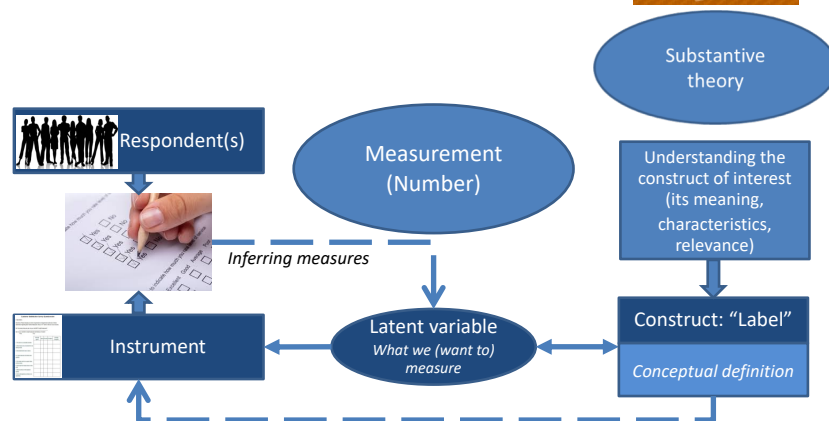
Withdrew due to massive resistance  
from the Roman Catholic Church  
(think of different paradigms!)



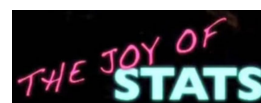
Measurement as quantification involves **numbers**  
*The reverse is not necessarily true ...*

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How Do We Get the Numbers in the  
Social Sciences?



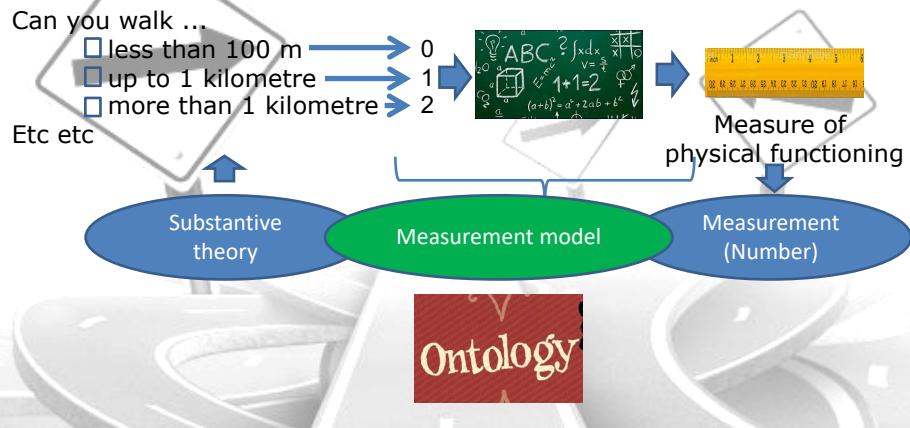
How exactly do we infer numbers?  
We need a measurement model.



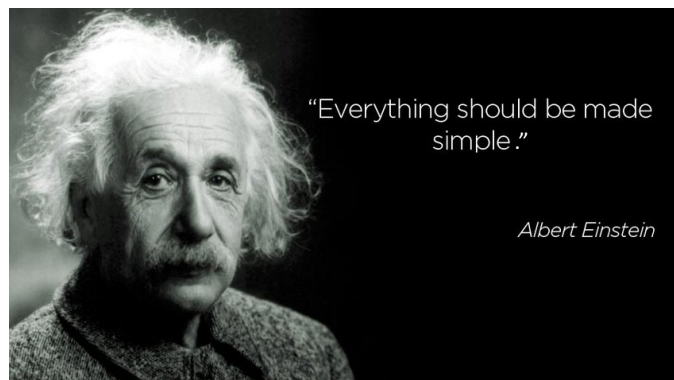
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## Measurement Models

Manifest response (qualitative) -> transformed into count by scoring (ordinal) -> inference of measures by measurement model (number)



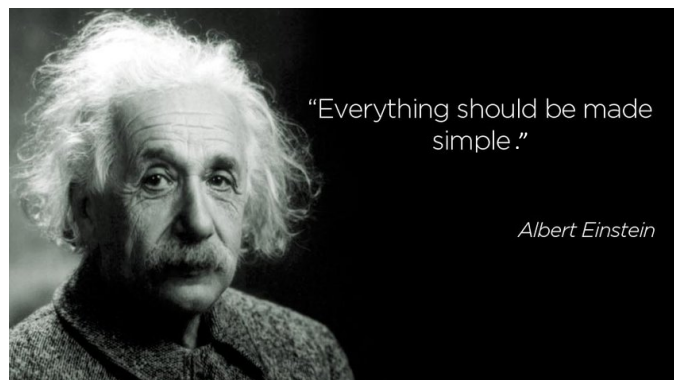
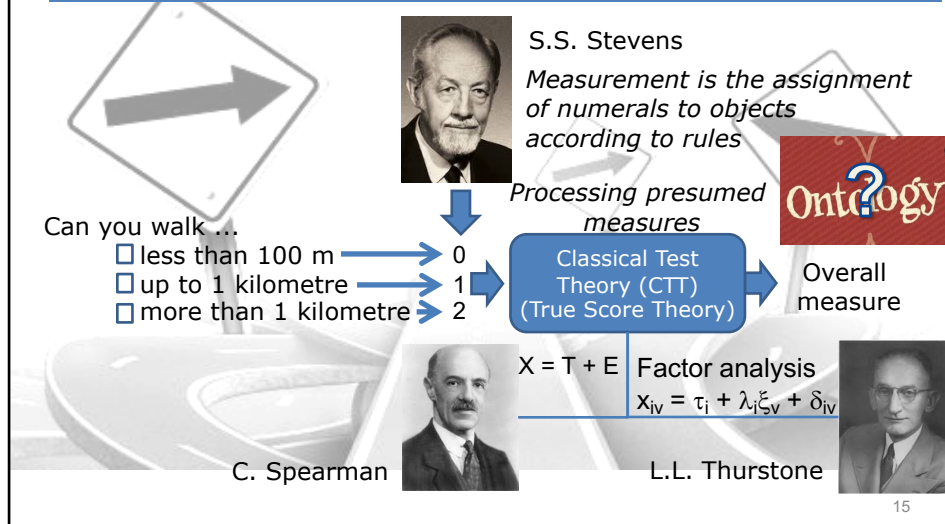
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## Measurement Models

Manifest response (qualitative) -> transformed into count by scoring (ordinal) -> inference of measures by measurement model (number)



## Numbers and Measures

What we end up with:  
a number



... which can be many different things

A proper measure,  
An amount expressed in an uncertain  
or varying metric,  
A count,  
A summary of measures for some  
purpose,  
Merely a numeral disguised as a  
measure

What we need:  
a measure



Justification of the scoring  
(raw score sufficiency),  
Invariant ruler (invariance,  
specific objectivity),  
Stable metric (unit of  
measurement),  
Measure as an interpretable  
amount of the property  
measured (link to substantive  
theory),  
Statement of uncertainty  
(measurement error),  
Transcending dependency on  
specific instrument

**Treating numbers as measures, requires a strong justification**

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## Measurement and Sufficiency



Measurement is an exhaustive,  
unambiguous, parsimonious  
and meaningful description of a  
respondent

*We cannot know more of the  
person with respect to the  
attribute measured*

Combining observations  
(adding up ratings to a score)  
has to be based on a minimally  
sufficient statistic

*Raw score sufficiency is not  
merely a statistical convenience  
but a requirement for  
measurement*

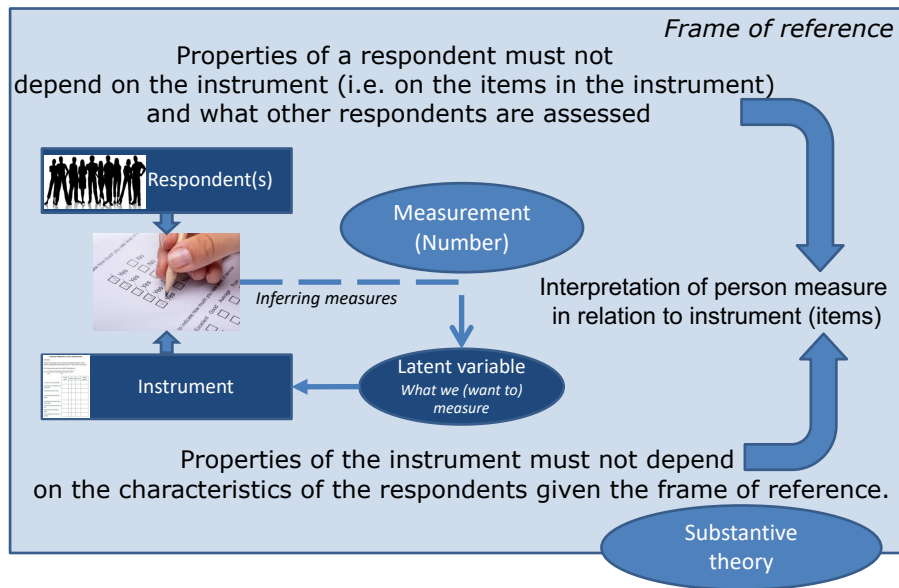
Summary scores (indexes) that combine multiple attributes and  
describe them by one number

have no unambiguous meaning

have a different ontological status (social construction)

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## Measurement, Invariance and the Frame of Reference



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## Measurement and the Unit of Measurement



Measures must have a stable unit of measurement (linearity)

*Raw scores are non-linear*

Unit of measurement must be invariant

*Measurement based on models assuming a particular person distribution (typically normal) lacks an invariant unit*

Unit of measurement must to be interpretable

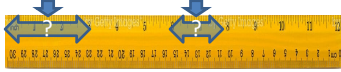
*Factor scores merely express percentile ranking at best*

Unit of measurement ought to transcend a particular instrument

*Requires theory-based unit of measurement*

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## Measurement and Uncertainty



Uncertainty of measurement must be invariant with respect to the sample of persons

Precision of a person measurement must not depend on who else has been assessed

Uncertainty of measurement must reflect actual precision of the measurement instrument along the measurement continuum

*One and the same standard of error across the whole range is implausible*

*Paradox of consistency of extreme scores*

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## Does CTT/TTT Deliver Measurement?

### Scoring

First of all, we have to abandon Stevens' definition of measurement. Otherwise everything becomes tautological.

*CTT does not justify the scoring.*

*In factor analysis, the weighting of item scores is sample-dependent, hence no invariance.*

### Invariance

*CTT assumes a normal distribution of persons.*

*Reliance on inter-item-correlations, which are sample-dependent.*

### Stable unit of measurement

*CTT relates a non-linear (at best) score by a linear transformation to a measure, which therefore remains non-linear.*

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## Does CTT/TTT Deliver Measurement?

### Interpretable measures

*CTT raw score measures are non-linear and remain unjustified.  
Factor scores related to person ranking.  
Strong focus on item-interrelationship, which is almost  
impossible to predict based on theory or explain meaningfully  
post-hoc. No item hierarchy.*

### Uncertainty

*Measurement precision ("reliability") stated at the instrument  
level but dependent on distributional assumptions and the  
sample.  
Standard error derived from reliability applied to any level of  
the measurement continuum.*

### Transcending dependency on specific instrument and instance of application

*Given the limitations of CTT measures, attempts are  
questionable to begin with.  
Equating remains sample dependent.*

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## But Maybe Measurement in the Social Sciences is Different Altogether

Measurement in the social sciences certainly differs in many ways  
from measurement in the natural sciences.

- Interaction of a very special object (subject capable of cognitions and meta-cognitions) and an instrument (see Pendrill, 2014)
- Replicability of measurement, etc.



But measurement does not differ in terms  
of the intended nature of the outcome and its interpretation



One number that exhaustively and unambiguously represents the amount of the property  
the person possesses

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## Measurement in the Social Sciences and the Natural Sciences

Epistemological difference

*How we observe*

*How we process observations*

*Different "measurement machinery"*

Same ontological claim

### Physical measurement

Properties exist independently of their being measured

Measurement does not create properties

Properties as a function of fundamental laws of nature

Wide frame of reference (universe, the world we live in), but possibly not unbounded

### Social measurement

Properties exist independently of their being measured

Measurement does not create properties

(At least some) properties are a function of society

Narrow frame of reference (e.g., properties may be limited to a particular society, or pertain to time)

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## The Rasch Model as an Alternative to CTT/TTT (& IRT)

The Rasch Model for Measurement

For dichotomous data  
(Rasch, 1960)

$$P(a_{vi} = 1) = \frac{e^{\beta_v - \delta_i}}{1 + e^{\beta_v - \delta_i}}$$

For polytomous data (ordered categorical data)  
(Andrich, 1978, 1988)

$$P(a_{vi} = x | \beta_v, \tau_{ij}, j = 1 \dots m, 0 < x \leq m) = \frac{e^{\left(\sum_{j=1}^x -\tau_{ij}\right) + x \cdot (\beta_v - \delta_i)}}{\Upsilon}$$

By its form, the Rasch model is an Item Response theory (IRT) model

But fundamentally different philosophy underpinning non-Rasch IRT and the Rasch Model/RMT

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## The Rasch Model & IRT

$$P(a_{vi} = 1) = \frac{e^{\beta_v - \delta_i}}{1 + e^{\beta_v - \delta_i}}$$

Rasch Model



$$P_i(X_i = 1 | \theta_j) = \frac{\exp[Da_i(\theta_j - b_i)]}{1 + \exp[Da_i(\theta_j - b_i)]}$$

Birnbaum Model (2pl)  
(Graded Response Model)



Allan Birnbaum &  
Frederic Lord

Based on fundamental principles  
of measurement

Aiming at best accounting for the  
data based on some assumptions of  
the respondent distribution

Prescriptive model

Descriptive model

Invariance (specific objectivity)

No invariance

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## The Rasch Model, CTT and IRT



CTT



IRT (incl. 1pl)



RMT

Scoring/  
Invariance

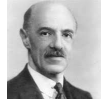
Unweighted item  
score not justified,  
interpreted as  
measure;  
weighted factor score  
(FA) sample dependent;  
widespread habit of  
using the raw score  
as a measure  
essentially implies  
data fitting the RM

Unweighted score  
is not a sufficient  
statistic,  
joint estimation of  
item and person  
properties based on  
distributional  
assumptions;  
meaningfulness  
hinges on the model  
fitting the data  
and data meeting the  
distributional  
assumptions;  
no unequivocal transformation  
of unweighted score  
to a measure  
(despite many papers claiming so  
and implicitly assuming a RM)

Unweighted score  
is a sufficient statistic,  
interpreted as a summary  
of what is observed,  
meaningfulness hinges  
on data fitting the model,  
justification of order  
can be tested;  
invariant measures  
based on an unequivocal  
score to measure  
transformation

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## The Rasch Model, CTT and IRT



CTT



IRT (incl. 1pl)



RMT

Unit of  
measure-  
ment/  
Inter-  
pretable  
measures

non-linear unit  
at best (speculative);

population-  
related inter-  
pretation of  
measure

Linear unit but  
speculative,  
depends on data  
meeting the  
distributional  
assumptions,  
and the description  
of unequal item  
unit (by means of  
estimating item  
discrimination)  
being true  
(assuming a frame-of-  
reference RM holds)

measure population-  
related, interpretation  
complicated by the item  
hierarchy varying

Linear unit of  
measurement;

person measures  
interpretable in terms of  
unequivocal hierarchy of  
items

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## The Rasch Model, CTT and IRT



CTT



IRT (incl. 1pl)



RMT

Un-  
certainty

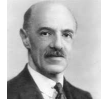
concept of reliability  
confounds  
population  
characteristics and  
precision;  
S.E.M. population/  
sample-dependent

S.E.M. depends  
on available  
information, which,  
however, is  
population-dependent

S.E.M. depends on  
available information  
(the items actually  
administered),  
information is  
population-independent  
by virtue of invariance

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## The Rasch Model, CTT and IRT



CTT

linking instruments  
by equipercntile  
equating,  
sample/population-  
dependent



IRT (incl. 1pl)

sample/  
population-dependent



RMT

establishing a common  
metric by co-calibration of  
instruments,  
sample-independent

Tran-  
scending  
instru-  
mentation  
and  
instance  
of  
application

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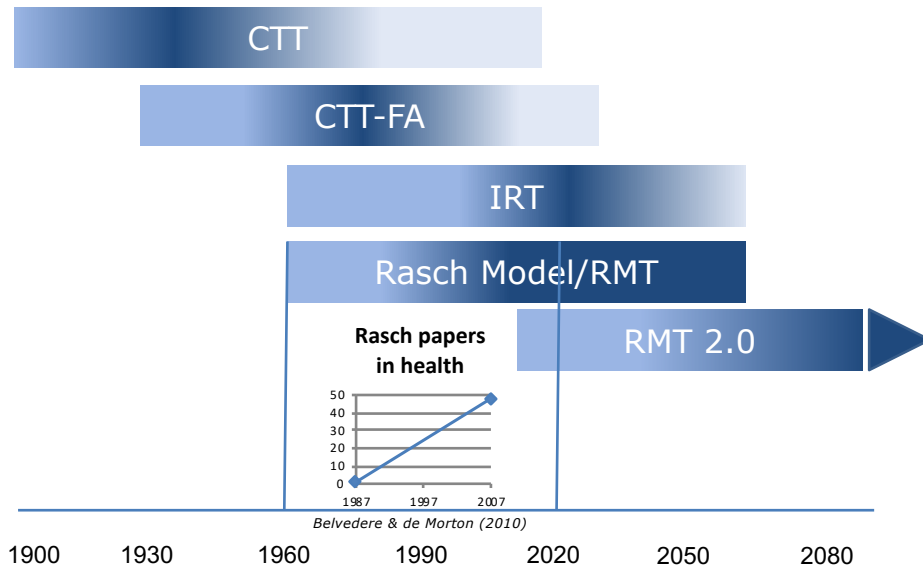


*Rasch. What else ?*

**R**

*Rasch Measurement Theory*  
Specifically objective since 1960.

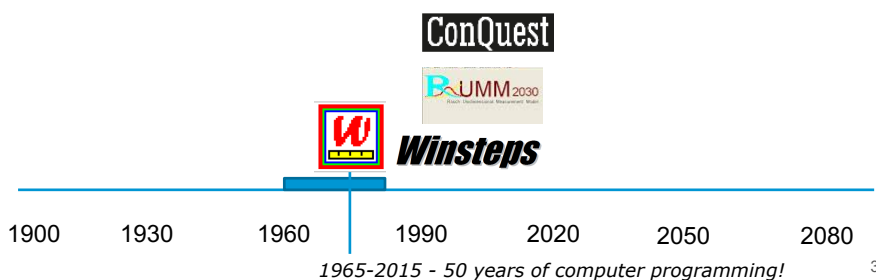
## The First 60 Years of the Rasch Model



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## Fundamental Contributions

- The Rasch model for dichotomous responses (Rasch, 1960, 1961)
- Specific objectivity (Rasch, 1977)
- Justifying polytomous item scoring (Andersen, 1977; Andrich, 1978)
- The Rasch model for polytomous responses (Andrich, 1978; Masters, 1982)



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## Recent Contributions

### Understanding the Rasch Measurement Model

Formal requirements, how the model works, understanding and handling of problems

- *Longitudinal measurement* (Olsbjerg & Christensen, 2015; Andrich, 2016).
- *Local dependence* (Marais & Andrich, 2008; Andrich & Kreiner, 2010; Andrich et al., 2012).
- *Frame of reference: impact on unit* (Humphry, 2005; Humphry & Andrich, 2008)
- *Differentiating true and artificial differential item functioning* (Andrich & Hagquist, 2012, 2015; Hagquist & Andrich, 2015)
- *Scoring polytomous items/threshold structure* (Andrich, 2013)
- *Guessing in multiple choice items* (Andrich et al., 2012; Andrich & Marais, 2014)

1900 1930 1960 1990 2020 2050 2080

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## Recent Contributions

### Understanding fit statistics

- *Tests of fit and their interpretation* (Kubinger et al., 2011)
- *Insight into distributional properties* (Christensen et al., 2017; Marais, 2013; Müller, 2015, 2018)
- *Insight into the role of the sample size* (Hagell & Westergren, 2016; Draxler & Alexandrowicz, 2015; Draxler, 2010; Hobart et al., 2012; Kubinger et al., 2009)



1900 1930 1960 1990 2020 2050 2080

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## Psychometric Justification of the Rasch Model

### Theory → Data



Rasch Model embodies necessary requirements of measurement in the social sciences

- Statistical measurement model: Formal theory of measurement
- Suitable to falsify measurement attempt due to its rigorous prescriptive character
- Provides measures expressed in a stable unit of measurement (implicit unit)
- Provides estimate of uncertainty
- Provides strong evidence of validity

### *Rasch Measurement Theory*

(Andrich, 2017; Ewing, Salzberger, Sinkovics, 2005)

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## Understanding the Concept to be Measured

### Data → Conceptual Theory



- Understanding of the characteristics of the attribute measured
- Understanding of the hierarchy of items
- Understanding of the meaning of unit of measurement

### Challenge

- Developing more advanced conceptual theories

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## The Rasch Model and Metrology – the Science of Measurement and its Application (towards RMT 2.0)

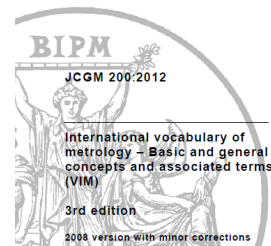
Metrological  
traceability

Uncertainty

### 2.41 (6.10) metrological traceability

property of a **measurement result** whereby the result can be related to a reference through a documented unbroken chain of **calibrations**, each contributing to the **measurement uncertainty**

NOTE 1 For this definition, a 'reference' can be a definition of a **measurement unit** through its practical realization, or a **measurement procedure** including the measurement unit for a non-ordinal **quantity**, or a **measurement standard**.



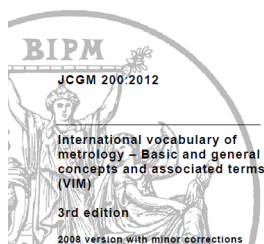
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## The Rasch Model and Metrology – the Science of Measurement and its Application (towards RMT 2.0)

Metrological  
traceability

Uncertainty

*Precision based on actually available  
information  
uncertainty expressed in a stable unit*



### 2.26 (3.9) measurement uncertainty

uncertainty of measurement  
uncertainty

non-negative parameter characterizing the dispersion of the **quantity values** being attributed to a **measurand**, based on the information used

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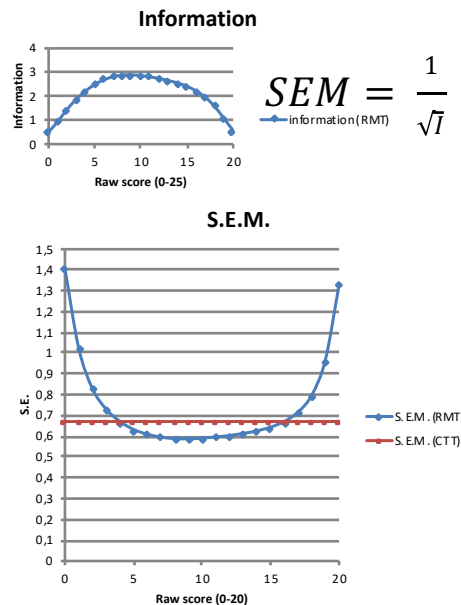
## Uncertainty: Reliability and the S.E.M. in RMT and CTT

$$Rel = \frac{VAR_T}{VAR_X}$$

$$SEM = SDx \times \sqrt{1 - Rel}$$

Example of actual data

	value (w/o extremes)
Cronbach's $\alpha$	0.91 (0.88)
PSI	0.88 (0.87)
Person SD	2.24 (1.83)
Person VAR	5.00 (3.34)
n	2434



## Expressing RMT's S.E.M. in the Reliability "Metric"

Idea:

Stating the S.E.M., which is expressed in an unfamiliar metric, at a given level in the more familiar reliability metric of 0 to 1

~ Local reliability

What would reliability be, if the sample distribution would have a particular observed variance and the SEM would be the same for all respondents.

Example of actual data

	value (w/o extremes)
Cronbach's $\alpha$	0.91 (0.88)
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Person SD	2.24 (1.83)
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n	2434

## Expressing RMT's S.E.M. in the Reliability "Metric"

$$Rel = 1 - \frac{SEM^2}{1}$$



Thissen (2000)

$$Rel = 1 - \frac{SEM^2}{VAR_x}$$

Assumption that observed variance = 1

Definition of the metric in IRT by setting the person true variance to 1

Locally  $SEM^2$  might be bigger than 1 resulting in a nonsensical reliability  $< 0$

Example of actual data

	value (w/o extremes)
Cronbach's $\alpha$	0.91 (0.88)
PSI	0.88 (0.87)
Person SD	2.24 (1.83)
Person VAR	5.00 (3.34)
n	2434

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## Expressing RMT's S.E.M. in the Reliability "Metric"

$$Rel = 1 - \frac{SEM^2}{1 + SEM^2}$$



Ramsay (2000)

Assumption that true variance = 1  
IRT: true variance = 1

$$Rel = 1 - \frac{SEM^2}{VAR_T + SEM^2}$$

RMT: item discrimination = 1  
True variance is estimated

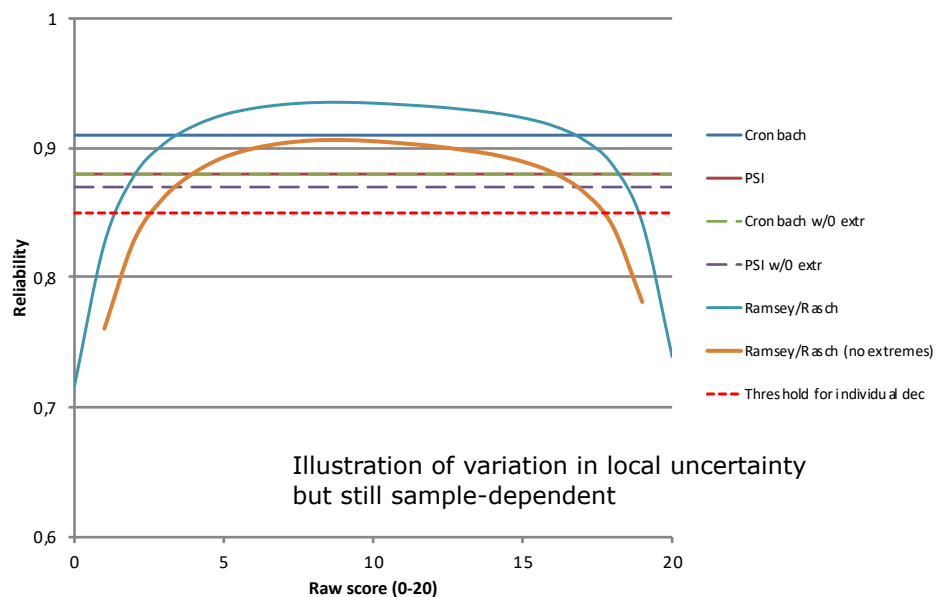
Ramsey/Rasch

Example of actual data

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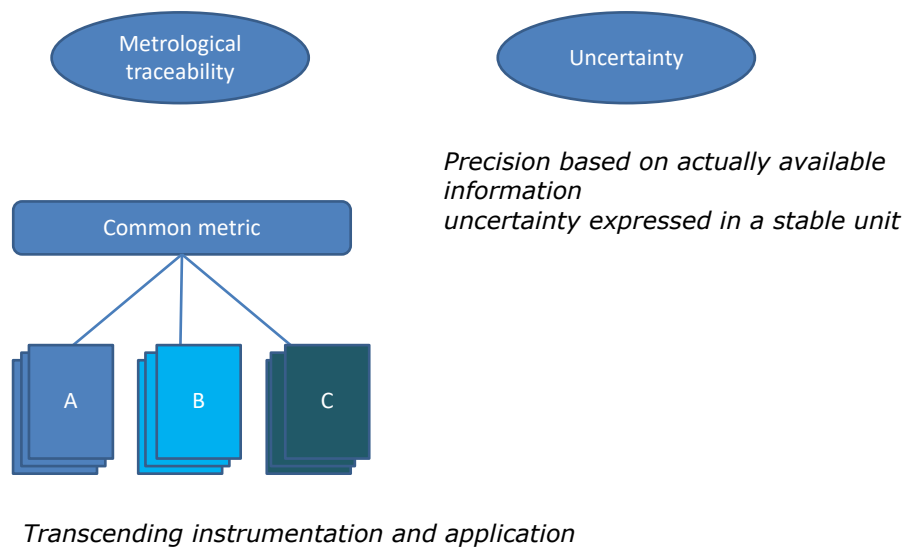
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## Expressing RMT's S.E.M. in the Reliability "Metric"



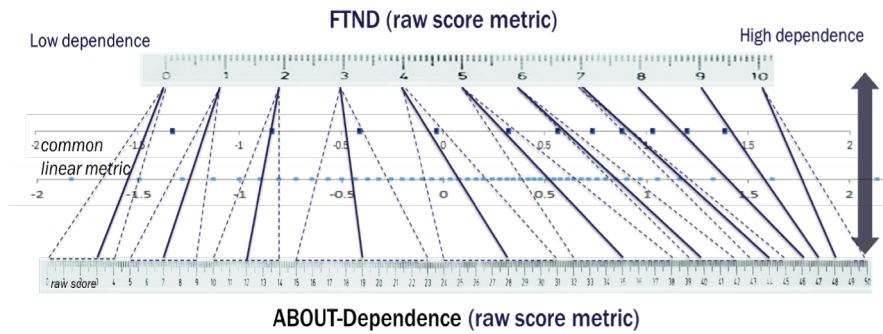
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## The Rasch Model and Metrology



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## Co-calibration of Dependence Instruments (Salzberger et al., 2019)



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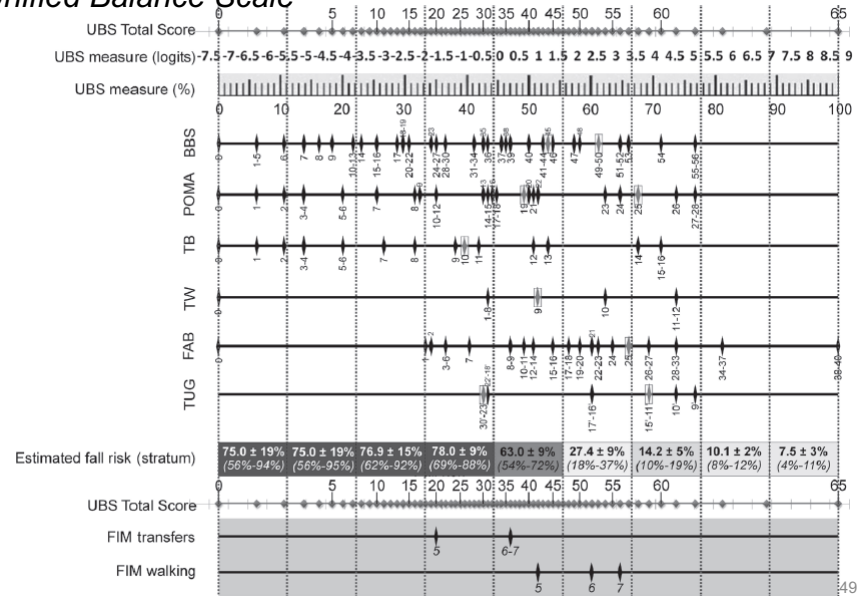
## Enhanced Interpretation of Measurement

Raw score	Rach composite measure				Most likely domain (raw score)				Domain EXTENT OF USE				Domain FEELING				Domain DOING			
	EXCELSIS	WALK	POCKY	DOING	FEELING	EXTENT	How soon after you wake up (00)	How long before going to bed (00)	strong desire (00)	difficult completely quit (00)	had to have one (10)	function normally (0)	hard to control the need/urge (10)	use more than intended (00)	stop what you are doing (10)	use in situations not supposed to (10)	snack off to use product (10)	avoid an activity (17)		
0	-2.00	0	0	0	0	0	<3 hrs	<3 hrs	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
1	-1.90	1	0	0	0	1	<3 hrs	<3 hrs	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
2	-1.80	2	0	1	1	2	<3 hrs	<3 hrs	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
3	-1.70	3	0	2	2	3	<3 hrs	<3 hrs	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
4	-1.60	4	0	3	3	4	<3 hrs	<3 hrs	RARELY	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
5	-1.50	5	0	4	4	5	<3 hrs	<3 hrs	RARELY	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
6	-1.40	6	0	5	5	6	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
7	-1.30	7	0	6	6	7	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
8	-1.20	8	0	7	7	8	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
9	-1.10	9	0	8	8	9	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
10	-1.00	10	0	9	9	10	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
11	-0.90	11	0	10	10	11	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
12	-0.80	12	0	11	11	12	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
13	-0.70	13	0	12	12	13	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
14	-0.60	14	0	13	13	14	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
15	-0.50	15	0	14	14	15	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
16	-0.40	16	0	15	15	16	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
17	-0.30	17	0	16	16	17	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
18	-0.20	18	0	17	17	18	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
19	-0.10	19	0	18	18	19	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
20	0.00	20	0	19	19	20	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
21	0.10	21	0	20	20	21	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
22	0.20	22	0	21	21	22	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
23	0.30	23	0	22	22	23	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
24	0.40	24	0	23	23	24	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
25	0.50	25	0	24	24	25	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
26	0.60	26	0	25	25	26	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
27	0.70	27	0	26	26	27	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
28	0.80	28	0	27	27	28	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
29	0.90	29	0	28	28	29	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
30	1.00	30	0	29	29	30	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
31	1.10	31	0	30	30	31	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
32	1.20	32	0	31	31	32	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
33	1.30	33	0	32	32	33	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
34	1.40	34	0	33	33	34	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
35	1.50	35	0	34	34	35	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
36	1.60	36	0	35	35	36	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
37	1.70	37	0	36	36	37	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
38	1.80	38	0	37	37	38	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
39	1.90	39	0	38	38	39	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
40	2.00	40	0	39	39	40	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
41	2.10	41	0	40	40	41	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
42	2.20	42	0	41	41	42	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
43	2.30	43	0	42	42	43	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
44	2.40	44	0	43	43	44	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
45	2.50	45	0	44	44	45	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
46	2.60	46	0	45	45	46	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
47	2.70	47	0	46	46	47	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
48	2.80	48	0	47	47	48	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
49	2.90	49	0	48	48	49	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER
50	3.00	50	0	49	49	50	<3 hrs	<3 hrs	SOMETIMES	RARELY	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER	NEVER

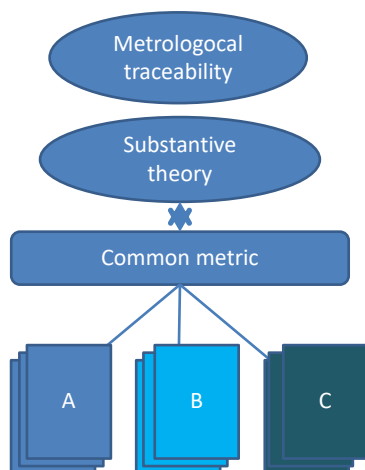
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## Common metric for Balance Scales (LaPorta et al., 2011)

### Unified Balance Scale



## Challenges for the Next 60 Years



Causal Rasch Model (Stenner et al., 2013)

Theory-based item hierarchy

Explicit, theory-based unit of measurement rather than emerging, implicit unit

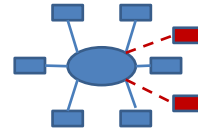
*Transcending instrumentation and application*



## Advancing the Conceptual theory

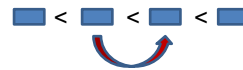
### Qualitative theory of the attribute

- Suggests item content
- Updated based on empirical evidence
- *Descriptive theory*



### Theory suggesting order of items (ordinal theory)

- General principle governing the order of items?
- Updated based on empirical evidence
- *(Partly) descriptive theory*



### Quantitative theory of the attribute

- Specifies measurement mechanism that explains relationship between items (quantitative predictions with respect to item location)



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## Example: Measuring Cognitive Dissonance

Table 7.1 A theoretical construct map of cognitive dissonance in marketing

		Level of dissonance	Cognitive characteristics	Emotional characteristics
<div style="display: flex; align-items: center;"> <div style="writing-mode: vertical-rl; transform: rotate(180deg);"> <div style="text-align: center;">weak</div> <div style="flex-grow: 1; border-left: 2px solid blue; position: relative;"> <div style="position: absolute; top: 0; left: -5px;">↑</div> <div style="position: absolute; bottom: 0; left: -5px;">↓</div> </div> <div style="text-align: center;">strong</div> </div> <div style="margin-left: 10px;"> <div style="writing-mode: vertical-rl; transform: rotate(180deg);">Weak cognitive dissonance</div> <div style="writing-mode: vertical-rl; transform: rotate(180deg);">Strong cognitive dissonance</div> </div> </div>	levels	Very mild dissonance	Possibly not having had a sufficient look-around	Uneasiness, denial of being happy
		Mild dissonance	Lack of certainty about the purchase; possibly not enough information asked for	
		Noteworthy dissonance	Regretting the money spent on the product purchased	Denial of having a good feeling
		Moderate dissonance	Increased uncertainty whether purchase was sensible	Annoyance
		Considerable dissonance	Expecting negative surprises during use of product; impression that product does not fit	Denial of pleasure given by the product in the future
		Strong dissonance	Wouldn't buy product again; feeling that they were spinning me a line	Unpleasant feeling when explaining purchase to friends; belief that product won't go down well with friends

characteristics  
(cognitive,  
emotional)



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$$\square < \square < \square < \square$$

$$\square < \square < \square < \square$$


Figure 7.17 Comparison of expected item endorsability and actual item location estimates (overall)

## Advancing Measurement in Health (the Social Sciences)

Improve definition of the concept to be measured and establish measurement standards (not instrument dependent)

"Thinking like physicists" (Stenner & Burdick, 2011)\*

"Concept-driven measurement" (Cano et al., 2018)

Quantitative theory that exposes the measurement mechanism, causal Rasch models (Stenner et al., 2013)

Towards an explicit unit and measurement that transcends a particular instrument (Humphry, 2005; Humphry & Andrich, 2008; Humphry et al., 2014)

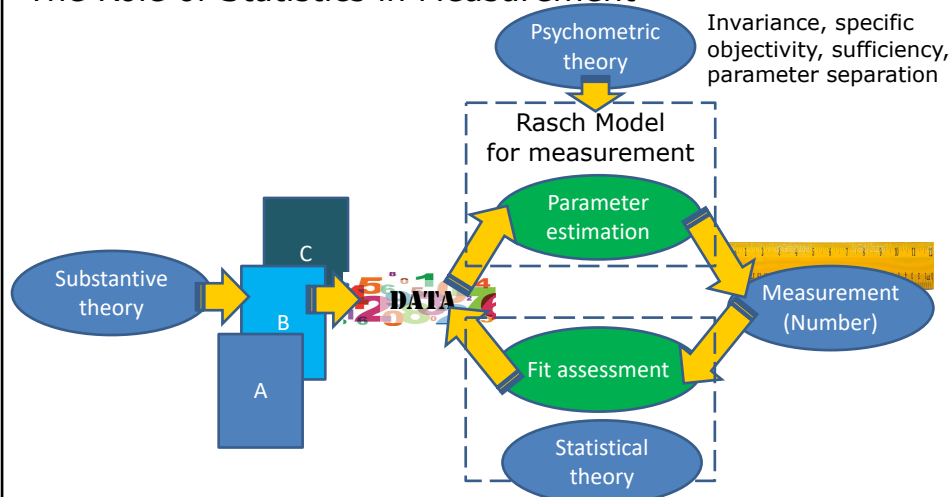
Linking metrology and psychometrics (Pendrill, 2017, 2014; Cano et al., 2016; Cano et al., 2017; Cano et al., 2018, 2019; Pendrill & Fisher, 2013)



-> Universal interpretation of measures (see Browne & Cano, 2019)

- Reese, T.W. (1943). *The Application of the Theory of Physical Measurement to the Measurement of Psychological Magnitudes, with Three Experimental Examples. Psychological Monographs*, 55, 1-89.  
"The conclusion was drawn that none of the attempts at measurement, used so far by psychologists, meet the necessary criteria for fundamental measurement."

## The Role of Statistics in Measurement



*Fit assessment not to be based on – in the end arbitrary – cut-off values (e.g. p values), theory-informed decision making, but statistical evidence need to be accurate and trustworthy*

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## Why RMT 2.0?

- Overcoming the threat of just “do[ing] so and so and so forth”, of just “follow[ing] the form”
- Finding laws requires theorizing about laws



*"Because of the success of science, there is a kind of a pseudo-science. Social science is an example of a science which is not a science. They **follow the forms**. You gather data, you **do so and so and so forth**, but they don't get any laws, they haven't found out anything. They haven't got anywhere – yet. Maybe someday they will, but it's not very well developed."*

*Richard Feynman, BBC Interview, 1981*

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