



THE MYTHOLOGY OF RISK ASSESSMENTS

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Prediction

“Half of my patients would go straight out in the community and commit a major violent crime. The problem is that I do not know what half....”

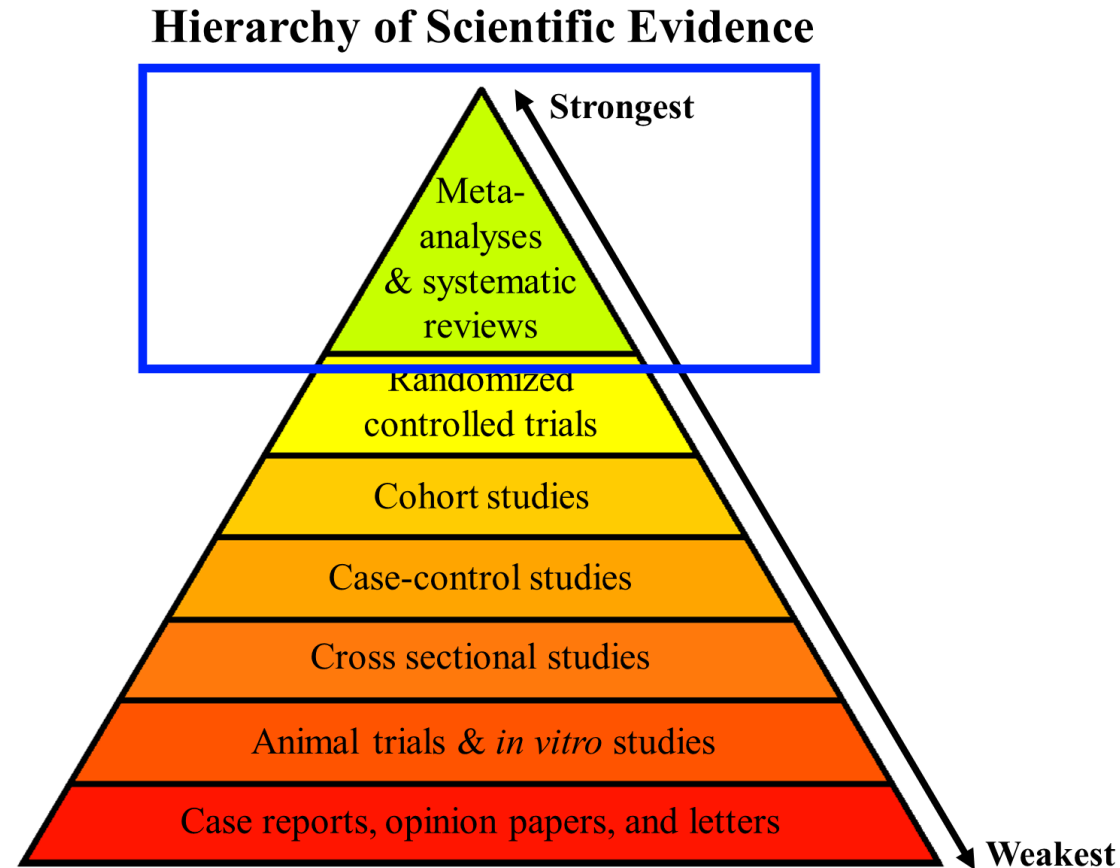
// Unknown clinician at Maudsley Special Hospital

Theoretical aspects

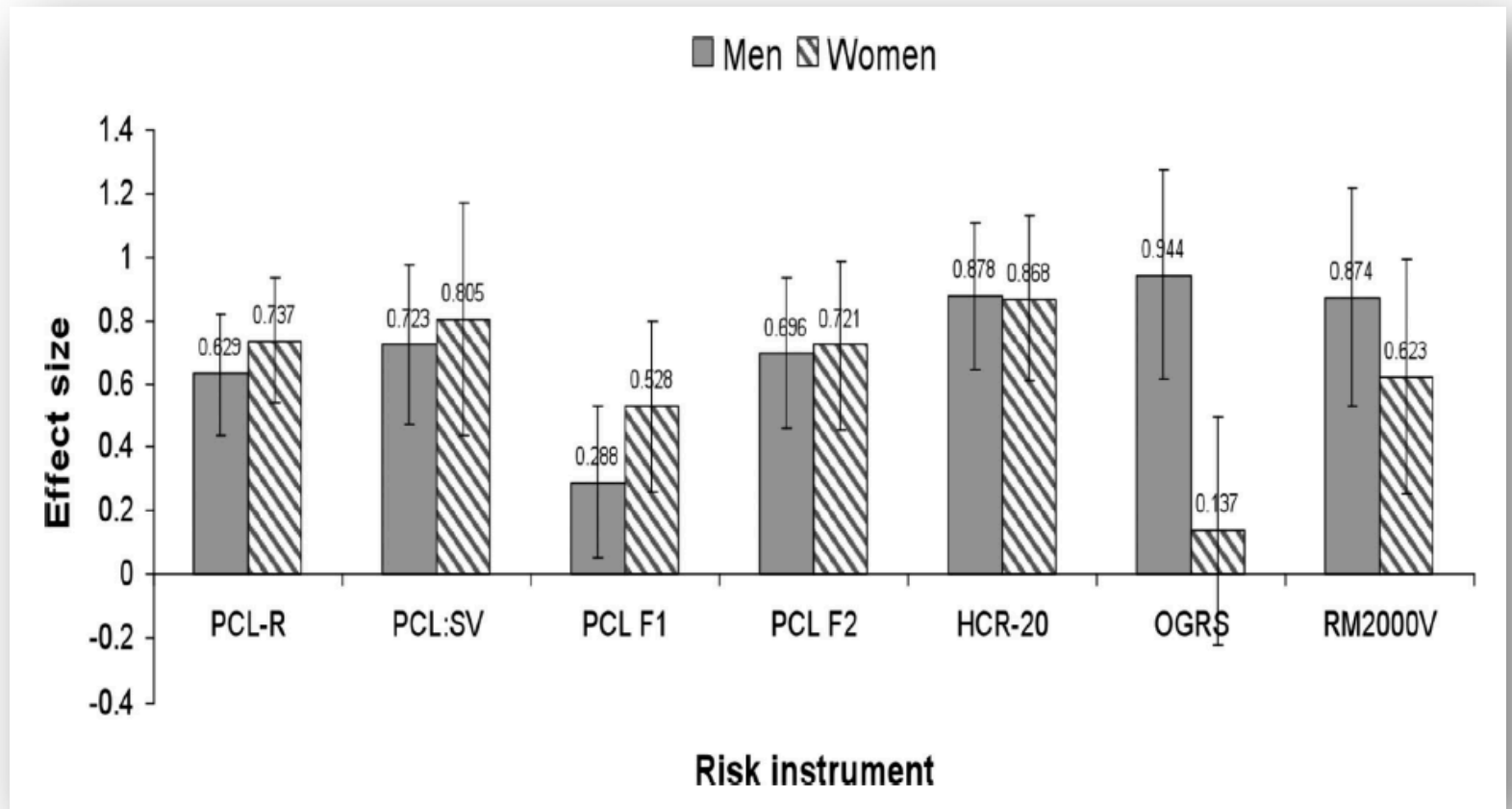
According to Paul Falzer (2013) there are three aspects of interest:

- The practical aspect (are they better than chance?)
- The differential aspect (is one method superior to others?)
- The epistemological aspect (how accurate is accurate enough?)

What can recent meta-analytical work tell us?



Are risk assessments better than chance?



The Efficacy of Violence Prediction: A Meta-Analytic Comparison of Nine Risk Assessment Tools

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Actuarial risk assessment tools are used extensively to predict future violence, but previous studies comparing their predictive accuracies have produced inconsistent findings as a result of various methodological issues. We conducted meta-analyses of the effect sizes of 9 commonly used risk assessment tools and their subscales to compare their predictive efficacies for violence. The effect sizes were extracted from 28 original reports published between 1999 and 2008, which assessed the predictive accuracy of more than one tool. We used a within-subject design to improve statistical power and multilevel regression models to disentangle random effects of variation between studies and tools and adjust for study features. All 9 tools and their subscales predicted violence at about the same moderate level of predictive efficacy with the exception of Psychopathy Checklist-Revised (PCL-R) Factor 1, which predicted violence only at chance level among men. Approximately 25% of the total variance was explained by methodological features (age, length of follow-up, different types of violent outcome, sex, and sex-related interactions). Sex-differentiated efficacy was found for a small number of the tools. If the intention is only to predict future violence, then the 9 tools are essentially interchangeable; the selection on its efficacy in predicting violence. The moderate level of predictive accuracy of these tools suggests that they should not be used solely for some criminal justice decision making that requires a very high level of accuracy such as preventive detention.

“Overall, our results showed that all of the nine tools predicted violence at above-chance levels, with medium effect sizes...thus the instruments included in this study demonstrated medium effects for predicting violence risk.” Yang et al (2010) p. 754 & 757

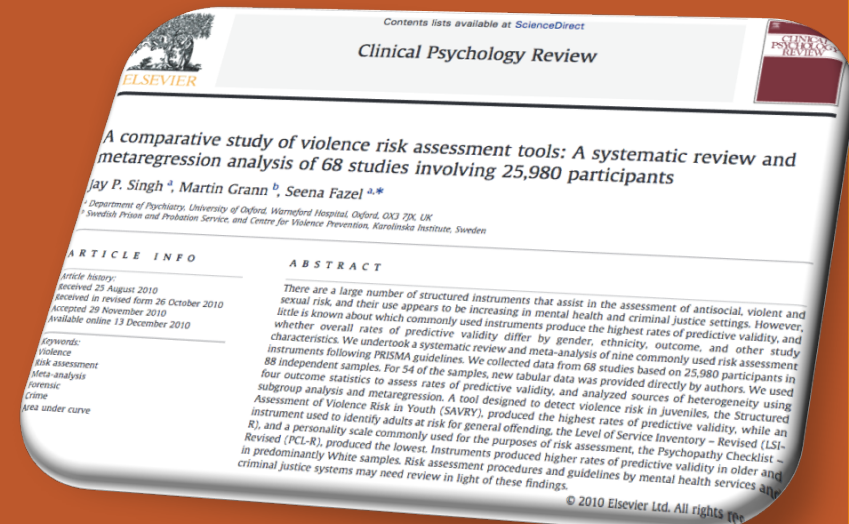
Is one method superior to others?

Table 3

Median area under the curve produced by nine risk assessment tools ranked in order of strength.

Tool	n	k	Median AUC	IQR
SVR-20	380	3	0.78	0.71–0.83
SORAG	1599	6	0.75	0.69–0.79
VRAG	2445	10	0.74	0.74–0.81
SAVRY	915	8	0.71	0.69–0.73
HCR-20	1320	8	0.70	0.64–0.76
SARA	102	1	0.70	–
Static-99	8246	12	0.70	0.62–0.72
LSI-R	856	3	0.67	0.55–0.73
PCL-R	2645	10	0.66	0.54–0.68

Note. n = sample size; k = number of samples; AUC = area under the curve; IQR = interquartile range.



“Our study found no evidence that, compared with SCJ tools, actuarial instruments produced better levels of predictive validity.” Singh et al. 2011

“...and no tool predicted violence significantly better than any other. In sum, all did well, but none came first.” Yang et al p. 757

Author bias?

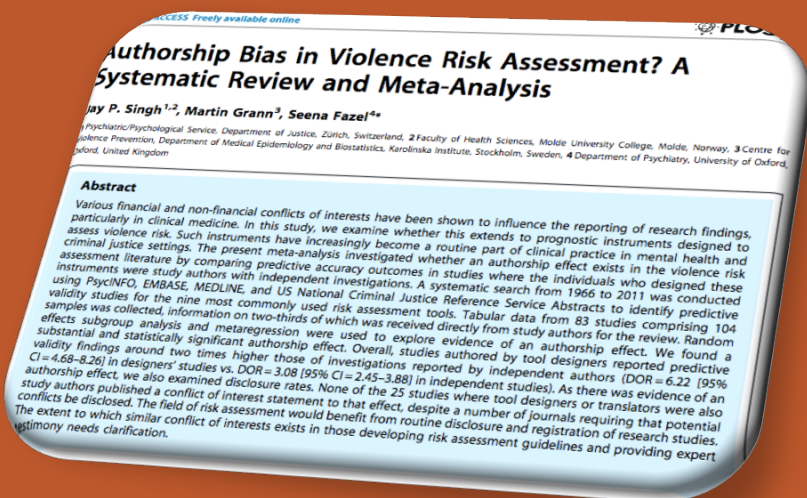
Table 3. Subgroup and metaregression analyses of diagnostic odds ratios (DORs) produced by nine commonly used risk assessment tools when a tool designer was a study author versus independent investigations.

Analysis	Subcategory	Authorship status	DOR (95% CI)	Metaregression
Overall	Translators not included as “designers”	Tool designer as study author	6.22 (4.68–8.26)	$\beta = 0.83$, $SE = 0.36$, $p = 0.02$
		Tool designer not study author	3.08 (2.45–3.88)	
	Translators included as “designers”	Tool designer as study author	4.45 (3.06–6.47)	$\beta = 0.39$, $SE = 0.26$, $p = 0.13$
		Tool designer not study author	3.04 (2.36–3.91)	
Type of tool ^a	Actuarial	Tool designer as study author	5.38 (3.82–7.58)	$\beta = 0.78$, $SE = 0.48$, $p = 0.11$
		Tool designer not study author	2.56 (1.98–3.30)	
	SCJ	Tool designer as study author	8.60 (5.15–14.35)	$\beta = 0.59$, $SE = 0.51$, $p = 0.26$
		Tool designer not study author	5.07 (3.27–7.84)	
Publication source ^a	Journal	Tool designer as study author	6.13 (4.59–8.20)	$\beta = 0.79$, $SE = 0.38$, $p = 0.04$
		Tool designer not study author	3.09 (2.39–3.98)	
	Gray literature	Tool designer as study author	8.73 (2.06–36.94)	$\beta = -1.03$, $SE = 1.05$, $p = 0.34$
		Tool designer not study author	3.07 (1.93–4.90)	

Note. β = unstandardized regression coefficient; SE = standard error; SCJ = structured clinical judgment; DOR = diagnostic odds ratio; CI = confidence interval; Gray literature = doctoral dissertations, Master’s theses, government reports, and conference presentations.

^aAuthorship operationally defined as being an author of the English-language version of the instrument under investigation.
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”Having explored this issue in the growing violence risk assessment literature, we have found evidence of both an authorship effect as well as lack of disclosure by tool designers and translators.”
Singh et al. (2013)



An example from the real world

Table 1
PCL-R Score Assessment

PCL-R
Total
Factor 1[#]
Facet 1 [†]
Facet 2 [‡]
Factor 2⁺
Facet 3 [§]
Facet 4
[#] Item 1, 2, 4
[†] Item 16.
[‡] Item 15.
[§] Item
Item

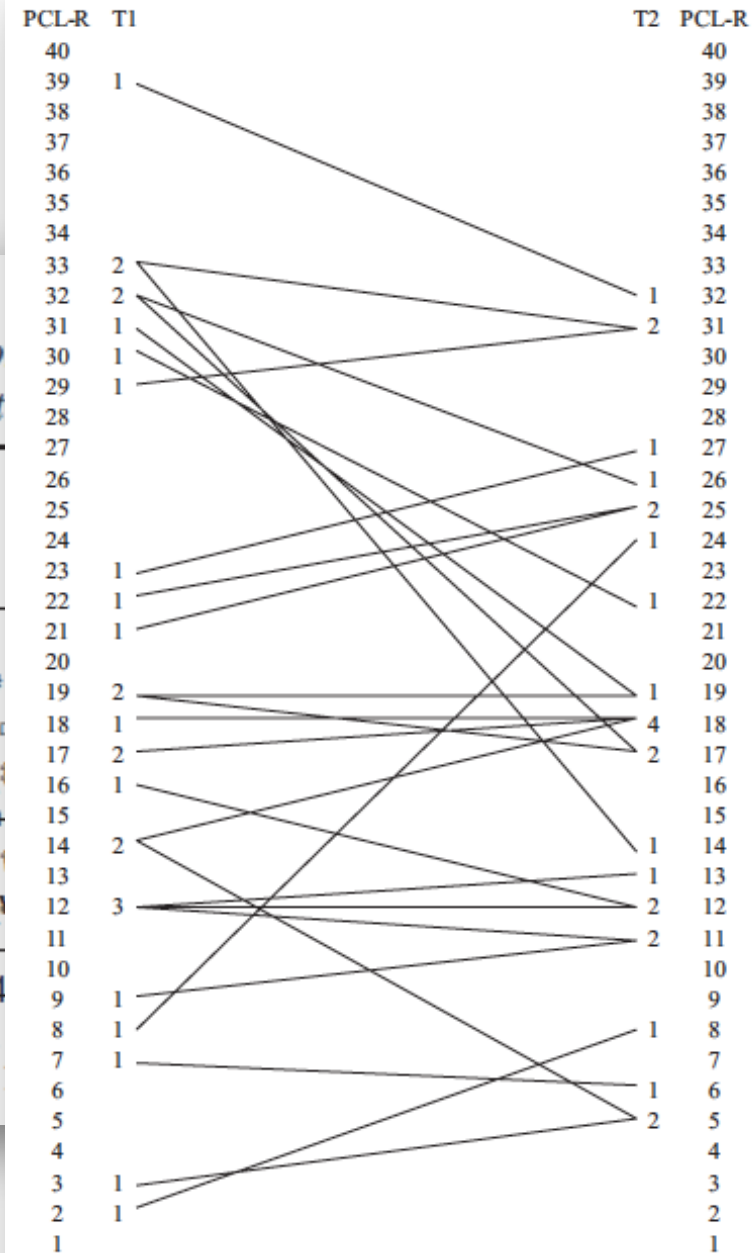


Figure 1. Galton Squeeze Diagram of PCL-R Scores at Time 1 and Time 2.

Field Reliability of the Psychopathy Checklist-Revised Among Life Sentenced Prisoners in Sweden

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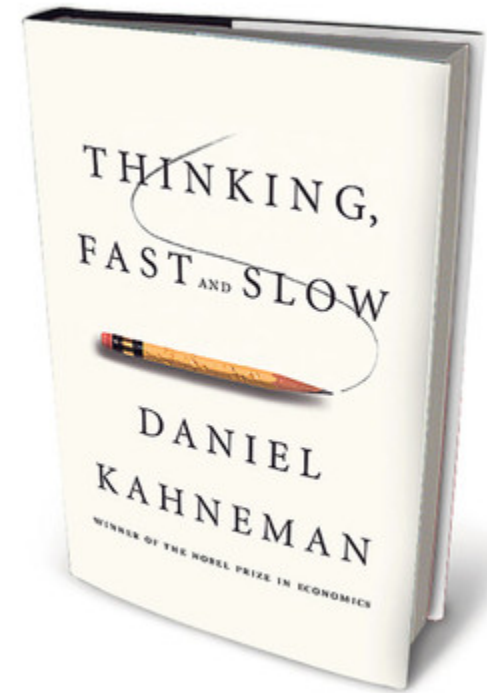
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Other biases...

There are a number of biases that influence our assessments:

- Anchoring bias
- Confirmation bias
- Overconfidence
- Recency bias
- Etc.



So, where are we?

All in all median AUC is .75

=

25% better than chance or
25% less than perfect prediction?



Theoretical aspects revisited...

Falzers (2013) three aspects of interest:

- Are risk assessments better than chance? Yes
- Is one risk assessment method superior to others? No
- How accurate is accurate enough? Not only a scientific question!

What is practice/science/politics?

Within the field of risk assessments it is extra important to part the areas of science and moral:

- Science can tell us what is true!
- Moral (or politics) can tell us what is right!



A possible solution?

- With the current status of risk assessments should clinicians avoid situations where the individual assessed is at risk to be incarcerated due to what he or she is likely to do
- I argue that it is more acceptable and justifiable to conduct risk assessments when: I) they can lead to something positive (e.g. time determined life sentence, previously laid conditional release, etc), II) standardized, valid and reliable (as reliable as possible) methods are used, and III) that the assessor is well aware of the recent scientific literature...



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