

Use of Meta-Analysis to derive PSF multipliers for Human Reliability Analysis

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PSAM 9

Outline for Today

- ▶ Introduction
- ▶ Method
 - Meta-Analysis
 - PSF Multipliers
- ▶ Results
- ▶ Conclusions
- ▶ Next Steps

What is a PSF?

- ▶ PSF – Performance Shaping Factor
- ▶ Factors which influence human error rates
- ▶ Typically in Human Reliability Analysis (HRA) are used to modify normal error rates
 - e.g. SPAR-H

$$HEP = P_0 \prod_{i=1}^n PSF_i$$

Goals of this research:

- ▶ Use available data on studies looking at sleep deprivation effect on performance for the purpose of developing a framework to aid in the quantification of PSF multiplier for use in HRA



What is Meta-analysis?

► What it is:

- Synthesis of results from available literature about a topic
- Structured format to extract information from selected studies
- Compiles data to quantify an overall effect
- Weighted by both
 - ▶ Sample size
 - ▶ Size of change in variable of interest

► Background:

- Statistical concepts introduced by Pearson
- 1st done by Smith and Glass in 1977

Smith, M., and Glass, G. (1977). "Meta-Analysis of Psychotherapy Outcome Studies." *American Psychologist* **32**: 752–760

Meta-Analysis

5 Steps of Meta-analysis

- ▶ Formulate the problem
- ▶ Collect the data
- ▶ Evaluate the data
(coding)
- ▶ Analyze the data
(calculate Effect Size)
- ▶ Report the findings



What is an Effect Size?

- ▶ Term often used in psychological and biological studies
- ▶ Describes the amount of change in output variable due to changes in the input variable
 - Nominal or control performance vs. test condition performance
- ▶ Many statistical definitions of Effect Size

Common Effect Size Calculations for Cohen's d

$$ES = \frac{(X_1 - X_2)}{S_{pooled}}$$

$$ES = \sqrt{\frac{F(n_1 + n_2)}{n_1 n_2}}$$

$$ES = t \sqrt{\frac{(n_1 + n_2)}{n_1 n_2}}$$

$$S_{pooled} = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 1}}$$

Where,

ES = standardized mean difference effect size

X_1 = mean of control

X_2 = mean of test

S_p = pooled sample deviation

s_1^2 = variance of sample 1

s_2^2 = variance of sample 2

n_1 = control sample size

n_2 = test sample size

F = F value

t = t-test value

d = Cohen's Effect Size

Probability Measure

- ▶ Probability measure defined in terms of the probability of the reaction time being longer under sleep deprivation than with normal sleep
- ▶ Data from multiple studies is used to supply data for calculating probabilities

Define Probability Model (Reaction Time)

- ▶ Model reaction time as random variable
- ▶ Assuming Normal Distributions:

$$P(g < 0) = \Phi\left(0 - \frac{\mu_C - \mu_T}{\sqrt{\sigma_C^2 + \sigma_T^2}}\right) = \Phi\left(0 - \frac{\mu_g}{\sigma_g}\right) = \Phi(-\beta)$$

- ▶ Where:
 - C = control condition reaction time – normal sleep
 - T = test condition reaction time – sleep deprived

$$\mu_g = \mu_C - \mu_T$$

$$\sigma_g = \sqrt{(\sigma_C^2 + \sigma_T^2)}$$

$$\beta = \mu_g / \sigma_g$$

Define Error Region (Reaction Time)

- ▶ Connect Effect Size to error probability
- ▶ Define $g = C-T$ for Reaction Time
 - C – measured value under Control condition
 - T – measured value under Test condition
- ▶ Define error as $T>C$
 - Fatigue degradation -- Test performance takes longer than control
 - i.e. $g<0$
- ▶ Assume T and C are normally distributed
- ▶ Then, g is also normally distributed
- ▶ Probability of Error = $P(g < 0)$

MA Effect Size results and β values

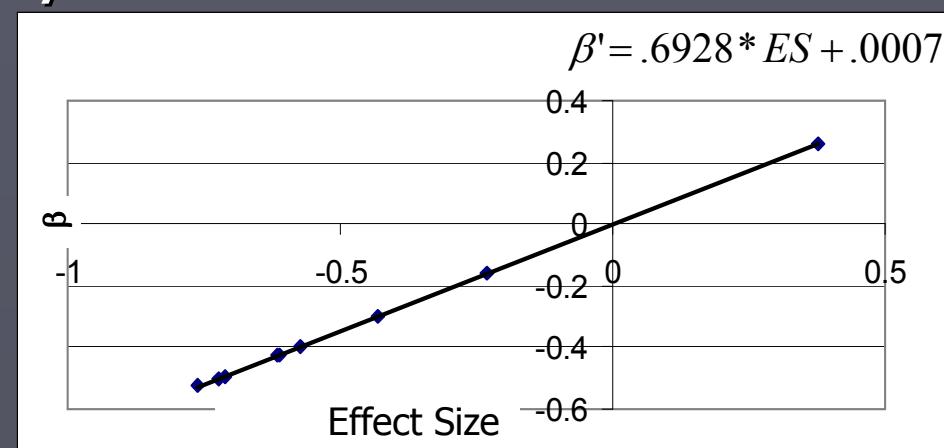
- ▶ Calculate ES
- ▶ Calculate $\beta = (\mu_g / \sigma_g)$

Study	Sleep Dep (hrs)	Effect Size	Beta
Chee_2004	24	-0.615	-0.426
Chee_2004	24	-0.572	-0.396
Chee_2004	24	-0.230	-0.159
Nilsson_2005	31.5	-0.763	-0.526
Choo_2005	24.4	-0.713	-0.493
Choo_2005	24.4	-0.724	-0.500
Choo_2005	24.4	-0.432	-0.299
Choo_2005	24.4	-0.613	-0.424
Thomas_2000	24	0.376	0.262

- ▶ Plot Effect Size and β to find relationship

Method to calculate β' equation from ES and β

- ▶ Estimate a linear relationship between Effect Size and β
- ▶ Used to approximate a β when only Effect Size data is present in study



This equation is valid only for this data set

Connecting probabilities from study data to PSF Multipliers

- ▶ Proposing a connection between error probability and PSF multipliers
- ▶ In this example
 - When $C=T$ the probability of ($g<0$) = 0.5
 - This is the condition of no change or a PSF = 1
 - Divide the probability of β by 0.5
$$\text{Multiplier} = \Phi(-\beta)/0.5$$

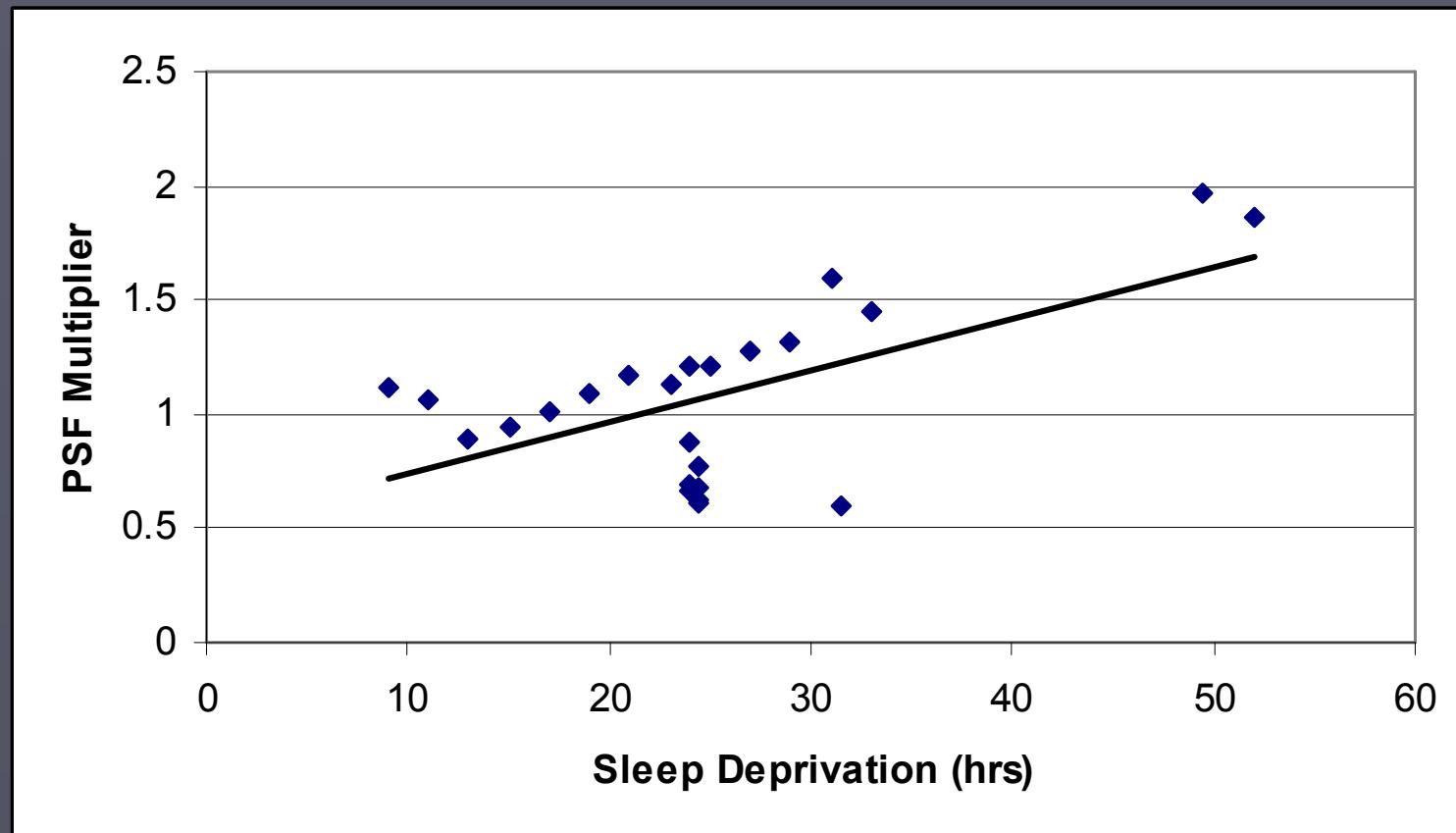
Calculate β' then Derive PSF Multipliers

Study	SD (hrs)	ES	β'	Multiplier
Marmuff_2005	9	0.2	0.146	1.116
Marmuff_2005	11	0.1	0.076	1.061
Marmuff_2005	13	-0.2	-0.132	0.895
Marmuff_2005	15	-0.1	-0.062	0.950
Marmuff_2005	17	0	0.007	1.006
Marmuff_2005	19	0.16	0.118	1.094
Marmuff_2005	21	0.29	0.208	1.165
Marmuff_2005	23	0.23	0.166	1.132
Marmuff_2005	25	0.38	0.270	1.213
Marmuff_2005	27	0.49	0.346	1.271
Marmuff_2005	29	0.59	0.416	1.322
Marmuff_2005	31	1.19	0.831	1.594
Marmuff_2005	33	0.85	0.596	1.449
Killgore_2006	49.5	3.216	2.235	1.975
Kobbeltvedt_2005	52	2.102	1.463	1.857

- ▶ $\beta' = .693 * ES + .007$
- ▶ Multiplier = $\Phi(\beta) / .05$

Results

Hours of SD vs. Derived PSF Multipliers



Procedure Summary

- ▶ Use sleep deprivation effect on performance data from various sources
- ▶ Find ES from individual studies
- ▶ Connect Effect Size to error probability
 - Relate ES to β
 - Calculate β' in absence of β data
- ▶ Propose a connection between error probability and PSF multipliers
- ▶ Method for basing PSF multipliers on the empirical data used to calculate the ES and probabilities

Next Steps

- ▶ Look at other psychomotor performance measures
- ▶ Accuracy
 - Different Limit state definition for probability model
 - Define error ($g < 0$) as $T < C$
 - Fatigue degradation -- Test accuracy is less than that of Control
- ▶ Expand data set to reach a population β' equation