

How to Measure Anything in Project Management

Hubbard Decision Research 2 South 410 Canterbury Ct Glen Ellyn, Illinois 60137 www.hubbardresearch.com



- This is an introduction to basic concepts for measuring Project Management (PM).
- We will take a critical look at PM, how it is measured and then we will describe some solutions.
- We will treat this as an example of applying the ideas described in How to Measure Anything: Finding the Value of Intangibles in Business (HTMA).
- Some of the content are methods you can start using right away, others are aspirational and you should start on a path to master them.

I'm going to tend to focus on topics I don't think are covered quite enough in PM.



Hubbard Decision Research Background

In 200+ major analysis projects,

HDR has been able to show

that no matter how difficult the measurement and monetization problem appears to be, we find a way to evaluate it and communicate the results.

- The benefits and risks of *dams on the Mekong River*
- Risks and benefits of *Environmental policy* for US farmers
- The benefits of *Educational assistance* in inner city schools
- The benefits of roads, schools and hospitals in Haiti and how to prioritize them for the *United Nations*
- The relative value of *R&D portfolios* in aerospace, biotech, and pharma
- Logistics forecasts for the battlefield and the effectiveness of training for the US Military

(A Few) Information Technology Examples

- Risk of software development
- Value of better information access
- Cybersecurity risk
- Risk of obsolescence and optimal technology upgrades
- Performance metrics for the business value of applications





Question: Why Measure Projects?

Answer: The risk of doing it wrong is high!

- They are large and time-consuming initiatives.
- Sometimes they have high visibility.
- Failure rates of projects from various sources: at least 10% maybe over 30% (failure includes cancellation but, in some cases, not meeting cost, schedule and deliverable goals).
- The cost of a failure can be greater than just wasted effort (loss in productivity, risk of customer loss, etc.).



Introduction

Types of Measurement Methods





Do "Scores" and "Scales" Work?

The Current Most Popular Method





The Analysis Placebo

Confidence in decision making methods is detached from performance

Organizational Behavior and Human Decision Processes 107, no. 2 (2008): 97– 105.

Journal of Behavioral Decision Making 3, no. 3 (July/ September 1990): 153–174.

Law and Human Behavior 23 (1999): 499-516.

Organizational Behavior and Human Decision Processes 61, no. 3 (1995): 305–326.

Interaction with Others Increases Decision Confidence but Not Decision Quality: Evidence against Information Collection Views of Interactive Decision Making

Heath and Gonzalez

<u>Abstract</u>

<u>A</u>

A

We present three studies of *interactive decision making*, where decision makers interact with others before making a final decision alone. Because the theories of lay observers and social psychologists emphasize the role of information collection in interaction, we developed a series





Do "Scores" and "Scales" Work?

Unintended consequences of simple scoring methods

Researchers uncovered several unintended consequences of simple ordinal scales and using words for probabilities.



 David Budescu and Dick Heuer (separately) researched the "illusion of communication" regarding interpretations of verbal labels for probabilities.

Highly Likely	
Likely	
Probable	888
Unlikely	

10% 20% 30% 40% 50% 60% 70% 80% 90%

Craig R. Fox showed how arbitrary features of how scales are partitioned effects responses.

Example:

If "1" on a 5-point impact scale means "less than \$1 million loss", the share of that response is affected by the partition of *other* choices.





The Only Risk Matrix You Need





Experts vs. Algorithms

What the research says about statistical methods vs. Subject Matter Experts



Paul Meehl assessed 150 studies comparing experts to statistical models in many fields (sports, prognosis of liver disease, etc.).



"There is no controversy in social science which shows such a large body of qualitatively diverse studies coming out so uniformly in the same direction as this one."



Philip Tetlock tracked a total of over 82,000 forecasts from 284 experts in a 20year study covering politics, economics, war, technology trends and more.



"It is impossible to find any domain in which humans clearly outperformed crude extrapolation algorithms, less still sophisticated statistical ones."





So Why Don't We Use More Quantitative Methods?

Commonly stated reasons for not using quantitative methods





CONCEPT of Measurement	The definition of measurement itself is widely misunderstood.
OBJECT of Measurement	The thing being measured is not well defined.
METHOD of Measurement	Many procedures of empirical observation are misunderstood.



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• What Measurement Really Means





• What Measurement Really Means

It's not a point value.

<u>Measurement:</u> a quantitatively expressed reduction in uncertainty based on observation.





Calibrated Experts

"Overconfident professionals sincerely believe they have expertise, act as experts and look like experts. You will have to struggle to remind yourself that they may be in the grip of an illusion."

Daniel Kahneman, Psychologist, Economics Nobel



- Decades of studies show that most managers are statistically "overconfident" when assessing their own uncertainty.
- Studies also show that measuring *your own* uncertainty about a quantity is a general skill that <u>can be taught</u> with a *measurable* improvement.



Measuring Overconfidence



- HDR trained over 2,000 individuals in subjective estimation of probabilities.
- Almost everyone is overconfident on the first benchmark test.



Measuring Calibration Training



- Training improves the ability to provide calibrated estimates.
- This improves real-world estimates after training is complete.



The same training methods apply to the assessment of uncertain ranges for quantities like the duration of project, the impact of a major data breach, etc.

		Interval
Group	Subject	% Correct (target 90%)
Harvard MBAs	General Trivia	40%
Chemical Co. Employees	General Industry	50%
Chemical Co. Employees	Company-Specific	48%
Computer Co. Managers	General Business	17%
Computer Co. Managers	Company-Specific	36%
AIE Seminar (before training)	General Trivia & IT	35%-50%
AIE Seminar (after training)	General Trivia & IT	~90%

90% Confidence

Doing The Math With Monte Carlo



Published in International Journal of Forecasting, 10 (1994), 495-906

Judgmental Decomposition: When Does It Work?

Donald G. MacGregor Decision Research, Eugene, OR

Society of Petroleum Engineers (2000)

The Application of Probabilistic and Qualitative Methods to Asset Management Decision Making

G. S. Simpson, F. E. Lamb, J. H. Finch, and N. C. Dinnie

SSCAG/SCAF/EACE Joint International Conference (2008)

An Assessment of the Inherent Optimism in Early Conceptual Designs and Its Effect on Cost and Schedule Growth

D. Bearden, C. Freaner, R. Bitten, and D. Emmons

Abstract

<u>A</u> In

> When missions experience cost growth, cost estimators are often criticized for underestimating the cost of missions in the early conceptual design stage. The final spacecraft and instrument payload configuration at launch, however, can be significantly different as the project evolves, thereby leading to cost "growth" as



The Need for Simulation in Project Plans

- A project plan even just moderately more complicated requires a simulation for even the most basic risk analysis.
- Furthermore, the simulation must include inputs like benefits and discrete/conditional events to support project decisions comprehensively.

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2 🍤	Estimate market and make more exact	3 d	2/18/2014	2/20/2014	\$1,440.00	★★★ ☆☆	1(50.0 %); Ka ren Martin [50.0 %); Mary Williams	
3 🐓	Design and order final package	5 d	2/21/2014	2/27/2014	\$3,600.00	★★★ ☆☆	E Karen Martin; Susan White; Mary Williams	
4 🍫	Create press releases	5 d	2/21/2014	2/27/2014	\$3,000.00	★★★ ☆☆	Karen Nartin; Susan White [50.0 %]; Mary Williams	
5 🍫	Create product specification materials	7 d	2/21/2014	3/3/2014	\$3,360.00	★★★ ☆☆	Karen Martin; Susan White	
6 🐓	Create marketing presentations	5 d	2/21/2014	2/27/2014	\$1,800.00	★★★ ☆☆	Karen Martin [50/0 %]; Mary Williams	
7 🍫	Transmit product launch details to internal	5 d	3/4/2014	3/10/2014	\$2,400.00	★★★ ☆☆	Susan White; Mary Williams	
8 🍫	Create sales, local, and product support	5 d	3/11/2014	3/17/2014	\$2,400.00	* ★★☆☆	Karen Martin	n; Mary Willia
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10 🍫	Update launch plan based on forecast	3 d	2/25/2014	2/27/2014	\$1,440.00	★★★ ☆☆	10 Karen Martin; Susan White	
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12 🍫	Complete and test product	15 d	2/25/2014	3/17/2014	\$12,000.00	***	Andrew And	Jerson; Barb
13 🐤	Production	12 d	2/25/2014	3/12/2014	\$12,592.00	****		
14 🤣	Create product prototypes	12 d	2/25/2014	3/12/2014	\$10,512.00	☆☆☆☆	Mark Robinson; Robert Wilson; Donna	a Hall [50.0 '
15 🍫	Prepare distribution channel	5 d	2/25/2014	3/3/2014	\$2,080.00	***	Donna Hall [50.0 %]; Paul King; John Brown	
16 🐤	Sales	3 d	2/28/2014	3/4/2014	\$1,728.00	****		
17 🍫	Establish sales channels	3 d	2/28/2014	3/4/2014	\$1,728.00	****	Michael Smith; William Jones; Robert Moore	
18 🍫	Product Support	4 d	3/5/2014	3/10/2014	\$1,808.00	****		
19 🍫	Establish product maintenance	4 d	3/5/2014	3/10/2014	\$1,808.00	***	Nancy Garcia; David Harris [50.0 %]; Patricia Jon	nes
20 🍫	Local Service	7 d	3/4/2014	3/12/2014	\$4,480.00	***		
21 🍫	Establish local service organizations	7 d	3/4/2014	3/12/2014	\$4,480.00	****	Charles Lawis; Helen Clark; Laura Rod	Jriguez; Lind
22 🍫	Prepare for Production	6 d	3/11/2014	3/19/2014	\$1,680.00	***		
23 🍫	Introduce changes control	5 d	3/11/2014	3/17/2014	\$1,400.00	***	Richard Mille	er
24 🍫	Finalize maintenance policy	1 d	3/18/2014	3/18/2014	\$280.00	****	Richard	d Miller
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Aggregating Uncertainty in Project Plans

Some project planning methods estimate durations based on "expected" time values. This is a classic "Flaw of Averages" problem (Sam Savage, Stanford).

Project Components



Days Duration

- Eight project tasks all have to be completed before going to the next phase of the project.
- The duration of each task is an average of 45 days but could be within +/- 15 days of that.
- What is the chance the start of the next phase will be after 45 days?
- Answer: 99.6%

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The Value of Information & The Measurement Inversion

$$EVI = \sum_{i=1}^{k} p(r_i) \max\left[\sum_{j=1}^{z} V_{1,j} p(\Theta_j | r_i), \sum_{j=1}^{z} V_{2,j} p(\Theta_j | r_i), \dots \sum_{j=1}^{z} V_{l,j} p(\Theta_j | r_i), \right] - EV *$$

Or more simply: The chance of being wrong times the cost of being wrong.

The economic value of measuring a variable is usually inversely proportional to the measurement effort.

HDR has observed a "Measurement Inversion" in nearly every industry, profession and type of decision model we've every made.

The cure for starts with knowing which variables are the highest information value.





The Measurement Inversion

Examples of Measurement Inversions from Information Technology Projects

Lowest Information Value



Initial cost

- Long-term costs
- Cost saving benefit other than labor productivity
- Labor productivity
- Revenue enhancement
- Technology adoption rate
- Project completion

Most Measured



Least Measured



Information Value vs. Information Cost



- EVPI Expected Value of Perfect Information
- ECI Expected Cost of Information
- EVI Expected Value of Information

If you know almost *nothing*, almost *anything* will tell you *something*.



CONCEPT of Measurement	The definition of measurement itself is widely misunderstood.
OBJECT of Measurement	The thing being measured is not well defined.
METHOD of Measurement	Many procedures of empirical observation are misunderstood.



- 1. Why do you care? (What decision could depend on the outcome of this measurement?)
- 2. What do you see when you see more of it? (Describe it in terms of observable consequences, then units of measure.)
- 3. How much do you know about it now?
- 4. At what point will the value make a difference?
- 5. How much is additional information worth?

If you can answer the first three, you can usually compute the last two.



- We can think through three categories of decisions. Your specific decision will depend on where you are in the project.
- Your initial business case is not "throw away." It should stay with the project for its lifespan.

	Initial Decision		When	Examples of Decisions
IV	lodel for Projec Approval		Before	Should I engage in this specific effort?
	Keep the model for following stages		During	Should I change course including: stop the project, reduce the features, change the managers, etc.
		7	After	Do I need to do more? (Also informs future project approval decisions.)



The Object of Measurement

A General Procedure for Measurement

AIE quantifies and then optimizes decisions by focusing measurements where it matters most.





- When the inputs to a decision model are uncertain, the output should be uncertain this is what simulations are for.
- Is this a "good" distribution or a "bad" one? How would you know?





What Measuring Risk Looks Like

The Loss Exceedance Curve

What if we could measure risk more like an actuary? For example, "The probability of losing more than \$10 million due to security incidents in 2016 is 16%."

What if we could prioritize security investments based on a "Return on Mitigation"?

	Expected Loss/Yr	Cost of Control	Control Effectiveness	Return on Control	Action
DB Access	\$24.7M	\$800K	95%	2,832%	Mitigate
Physical Access	\$2.5M	\$300K	99%	727%	Mitigate
Data in Transit	\$2.3M	\$600K	95%	267%	Mitigate
Network Access Control	\$2.3M	\$400K	30%	74%	Mitigate
File Access	\$969K	\$600K	90%	45%	Monitor
Web Vulnerabilities	\$409K	\$800K	95%	-51%	Track
System Configuration	\$113K	\$500K	100%	-77%	Track





The Method of Measurement

Why Does Our Risk Tolerance Change?





Quantifying Risk Tolerance

Monetizing risk and risk reduction is a basis for computing "Return on Mitigation"







The Object of Measurement

Expert Inconsistency in Estimates & Risks

- Adjusting for risk causes some previously-acceptable projects to be rejected.
- Also, some low return but low risk projects would now be acceptable.
- More projects with "intangible" benefits are now economically justified.
- The net result: A completely reshuffled deck of IT project approvals.



(i.e. 50% = project makes up half the work in the entire portfolio)



The Object of Measurement

One Decision During a Project: Managing Scope

- One reason for "scope creep" may be that the true cost of adding additional features to software in development is greatly underestimated.
- If costs are computed at all, they usually consider only initial development.



Actual Case: Cost of Adding Feature Which Extends Delivery by One Month

(Avg. Proportions in Simulation Shown in Pie Chart)


Measurement Misconceptions

CONCEPT of Measurement	The definition of measurement itself is widely misunderstood.
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Testing Measurement Intuition

A Sample of 5

- You randomly sample 5 people out of a company of 10,000 people and they spend an amount of time in a specific activity as shown by the data points below.
- Can this be a statistically significant estimate of time spent in this activity?
- Is it possible to estimate the chance the median time spent per person per day is between 15 and 40 minutes?





Intuitions About Samples Are Wrong

- There are widely held misconceptions about probabilities and statistics especially if they vaguely remember some college stats.
- These misconceptions lead many experts to believe they lack data for assessing uncertainties or they need some ideal amount before anything can be inferred.

"Our thesis is that people have strong intuitions about random sampling...these intuitions are wrong in fundamental respects...[and] are shared by naive subjects and by trained scientists" Amos Tversky and Daniel Kahneman, Psychological Bulletin, 1971





The "Math-less" Statistics Table

Approximate 90% Confidence Interval		
Sample Size	N th largest & smallest sample value	
5	1 st	
8	2 nd	
11	3 rd	
13	4 th	
16	5 th	
18	6 th	
21	7 th	
23	8 th	

- Simple Measurement Takeaway This table makes estimating a 90% confidence interval of a population median easy.
- <u>The Rule of Five:</u> There is a 93.75% chance that the median of any population is between the smallest and largest values in a random sample of five.
- This table expands on the Rule of Five. If you take 16 random samples of something, the 5th largest and 5th smallest values of that sample set approximate a 90% confidence interval.



How Much Samples Can Tell Us

The graph below shows the average of relative reduction in uncertainty as sample sizes increase by showing the 90% CI getting narrower and narrower with each sample according to the student-t method.



With a few samples, there is still high uncertainty but...

... each new sample reduces uncertainty a lot and the first few samples reduce uncertainty the most when initial uncertainty is high.

As number of samples increases, the 90 % CI get much narrower, but each new sample reduces uncertainty only slightly and beyond about 30 samples you need to quadruple the sample size to cut the error in half.



Here are a few key things I've learned measuring the "immeasurable"

- It's been measured before.
- You have more data than you think and you need less data than you think.
- Your probably need *different* data than you think.
- Decision makers can always define observable consequences.
- Decision makers understand it just fine if explained well.
- The best investment in most portfolios was better measurements of investments.



The Distribution of Canceled Projects vs. Duration

- Duration alone is the single best predictor of project cancellation.
 (Duration is also affected by several other factors normally associated with project failure – sponsorship, proper project management, quality and avoiding rework, etc.)
- Duration seems to follow a "Poisson" distribution – as if "cancellation events" are evenly distributed through time and longer projects are more likely to contain one of these events.







You Have More Data Than You Think

You have a lot of data in the organization. Even samples of some of this data can be revealing indicators of communication, involvement, etc.

- Calendars
- Working group discussion threads
- Project deliverables
- Frequency and cost of errors/rework
- Utilization of various systems
- Turnover
- Milestones on schedule



What Project Management Should Really Manage

Project completion is consistently a high value measurement, followed by adoption and benefits. Therefore:

- <u>Manage cancellation by managing duration</u> Project management should be about managing duration to avoid uncontrolled cancelation events, and ensuring the utility of the outcome.
- <u>Manage user adoption and benefits</u> User involvement will generally increase project benefits, which is also a responsibility of the project manager.
- <u>Fail early</u> Someone, if not project managers, needs to be in the position of asking, "Do we still need this?" based on changing external factors and evolving knowledge of the project difficulty.



Using Risk Analysis to Improve Projects

If the Risk is significant (it usually is), consider doing the following:

- Reduce the size and functionality of the proposed system focus on fewer high-return features.
- Define "Independently Justifiable Phases" (IJP) and quicker, iterative development methods (Agile, Lean, etc.).
- Wait until specific uncertainties in the environment subside e.g. major mergers, reengineering, etc.
- Wait to tackle big projects until proper skills are developed and methods are in place.
- "Off the Shelf" looks better when risk is considered.
- Invest more on a proper economic analysis of the largest project investments this should reduce uncertainty about critical quantities.
- Include deferred benefits in any estimate of scope creep costs.



Value of Quantitative Analysis for Projects

- The cost of analysis routinely comes in below 1% and has always been under 2% of the investment size including initial training.
- Considering the risk of bad project approval decisions, this would be entirely appropriate (and some types of projects exceed this).
- Quantitative analysis is not necessarily more time consuming that some qualitative methods. (One of the reasons this analysis is efficient is we conduct a Value of Information Analysis we only measure what is economically justified).
- Using the standard information value calculation for the value of AIE analysis, quantitative analysis
 itself was the best investment of all the IT investments we analyzed very conservative measures of
 payoffs put \$20 to every \$1 spent on AIE.



What to Do Next

Things you can do now or very soon:

- Drop the use of "scores" and "matrices" define actual observables in units of measure and quantify risk with probabilities.
- Identify the specific decisions you are trying to support.
- Build a decision model/business case for the project and keep it for the life cycle of the project and what it builds.
- Get calibrated so you can quantify your uncertainty.

Things to strive toward (the effort is easily justified for large projects or even small but frequent projects):

- Learn to model the uncertainty of a decision in a simulation evolve the model with more detail over time.
- Learn to compute the value of information.
- Learn a few more simple statistical methods especially models for what experts usually estimate.



Doug Hubbard Hubbard Decision Research dwhubbard@hubbardresearch.com www.hubbardresearch.com

Measure What Matters. Make Better Decisions.



Supplementary Material

Hubbard Decision Research 2 South 410 Canterbury Ct Glen Ellyn, Illinois 60137 www.hubbardresearch.com



- Tsai C., Klayman J., Hastie R. "Effects of amount of information on judgment accuracy and confidence" *Org. Behavior and Human Decision Processes,* Vol. 107, No. 2, 2008, pp 97-105.
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- Williams M. Dennis A., Stam A., Aronson J. "The impact of DSS use and information load on errors and decision quality" *European Journal of Operational Research*, Vol. 176, No. 1, 2007, pp 468-81.
- Knutson et. al. "Nucleus accumbens activation mediates the influence of reward cues on financial risk taking" NeuroReport, 26 March 2008
 Volume 19 Issue 5 pp 509-513.
- A small study presented at Cognitive Neuroscience Society meeting in 2009 by a grad student at U. of Michigan showed that simply being briefly exposed to smiling faces makes people more risk tolerant in betting games.
- Risk preferences show a strong correlation to testosterone levels which change daily (Sapienza, Zingales, Maestripieri, 2009).
- Recalling past events that involved fear and anger change the perception of risk (Lerner, Keltner, 2001).



Measuring and Removing Inconsistency

Methods that statistically "smooth" estimates of experts show reduced error in several studies for many different kinds of problems.



Reduction in Errors