



Can You Trust Your Data? Measurement and Analysis Infrastructure Diagnosis

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Outline

The Big Picture

Measurement errors and their impact

MAID Methods

- Process Diagnosis
- Data and Information Product Quality Evaluation
- Stakeholder Feedback

Summary and Conclusion

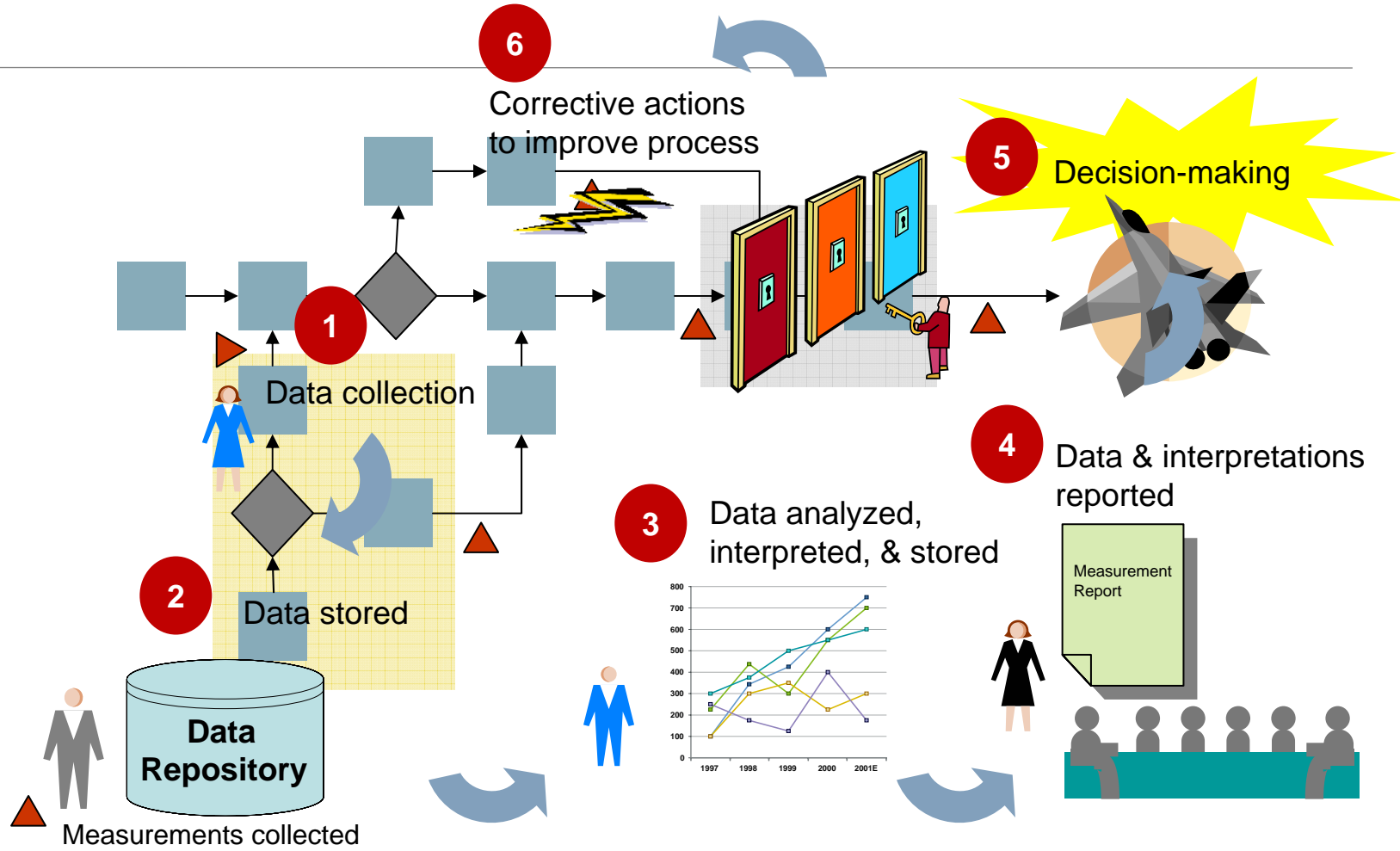


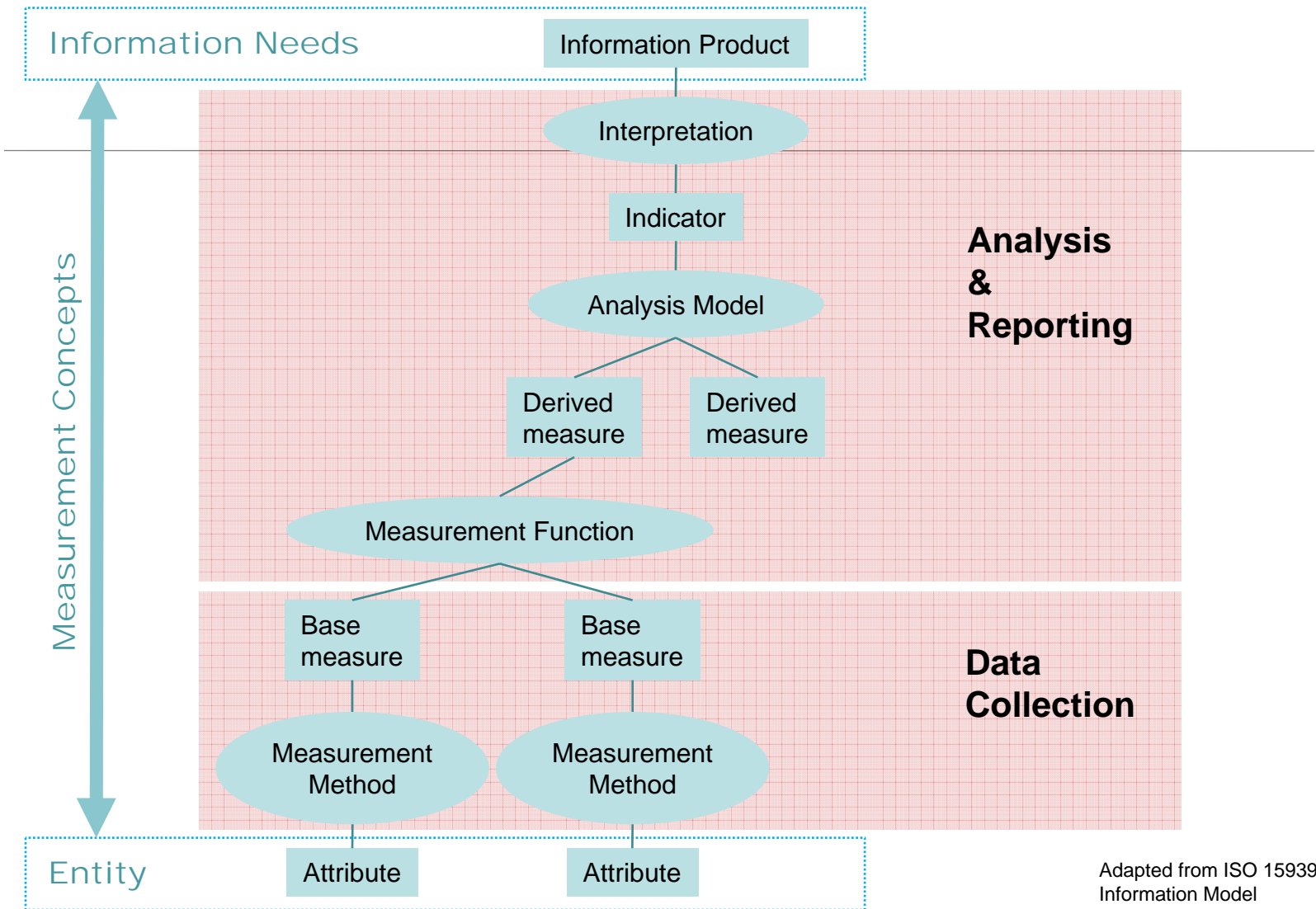
Benefit and Value of Measurement

The benefit and value of measurement comes from the **decisions and actions taken** in response to analysis of the data, not from the collection of the data.



Measurement and Analysis in Action





Adapted from ISO 15939 Information Model



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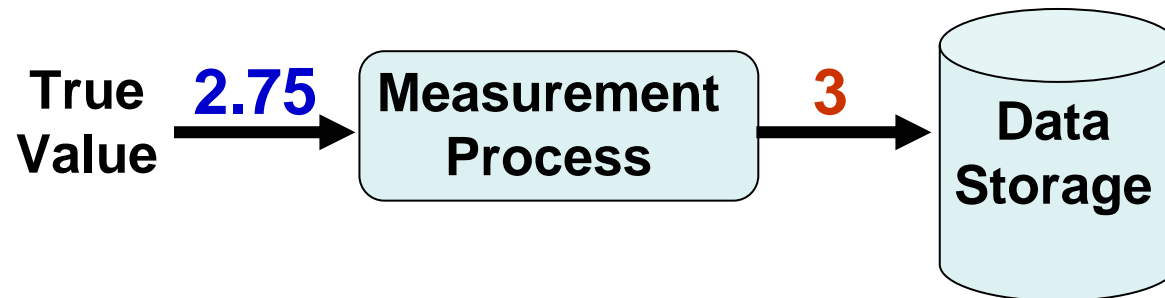
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What is Measurement Error?



Single Value: Deviation from the “true” value

- Distance is 1 mile, but your odometer measures it as 1.1 miles
- Effort really expended on a task is 2.75 hours, but it is recorded as 3

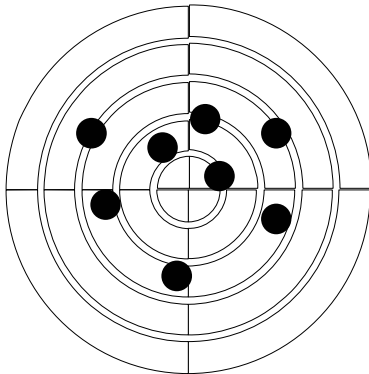
Data Set: Error introduced as a result of the measurement process used

- Not as defined, but as practiced



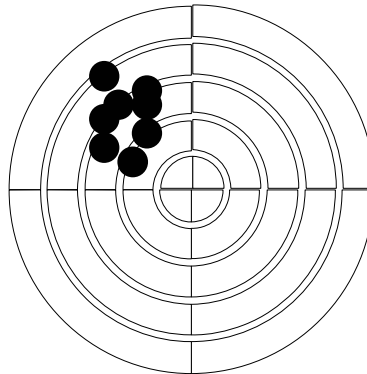
Gold Standard: Accuracy and Precision

(σ)

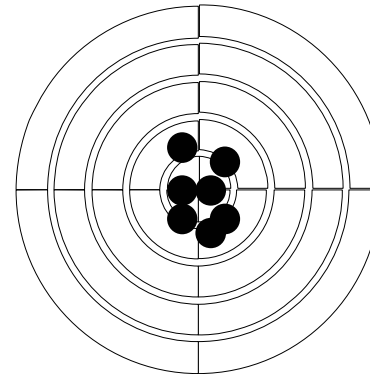


Accurate
but not precise

(μ)



Precise
but not accurate



Both accurate
and precise



Where do Measurement Errors come From₁

Data Entry Errors

- Manual data entry
- Lack of integrity checks

Differing Operational Definitions

- Project duration, defect severity or type, LOC definition, milestone completion

Not a priority for those generating or collecting data

- Complete the effort time sheet at the end of the month
- Inaccurate measurement at the source

Double Duty

- Effort data collection is for Accounting not Project Management.
 - Overtime is not tracked
 - Effort is tracked only to highest level of WBS



Where do Measurement Errors come From₂

Dysfunctional Incentives

- Rewards for high productivity measured as LoC/Hr
- Dilbert-esque scenarios

Failure to provide resources and training

- Assume data collectors all understand goals and purpose
- Arduous manual tasks instead of automation

Lack of priority or interest

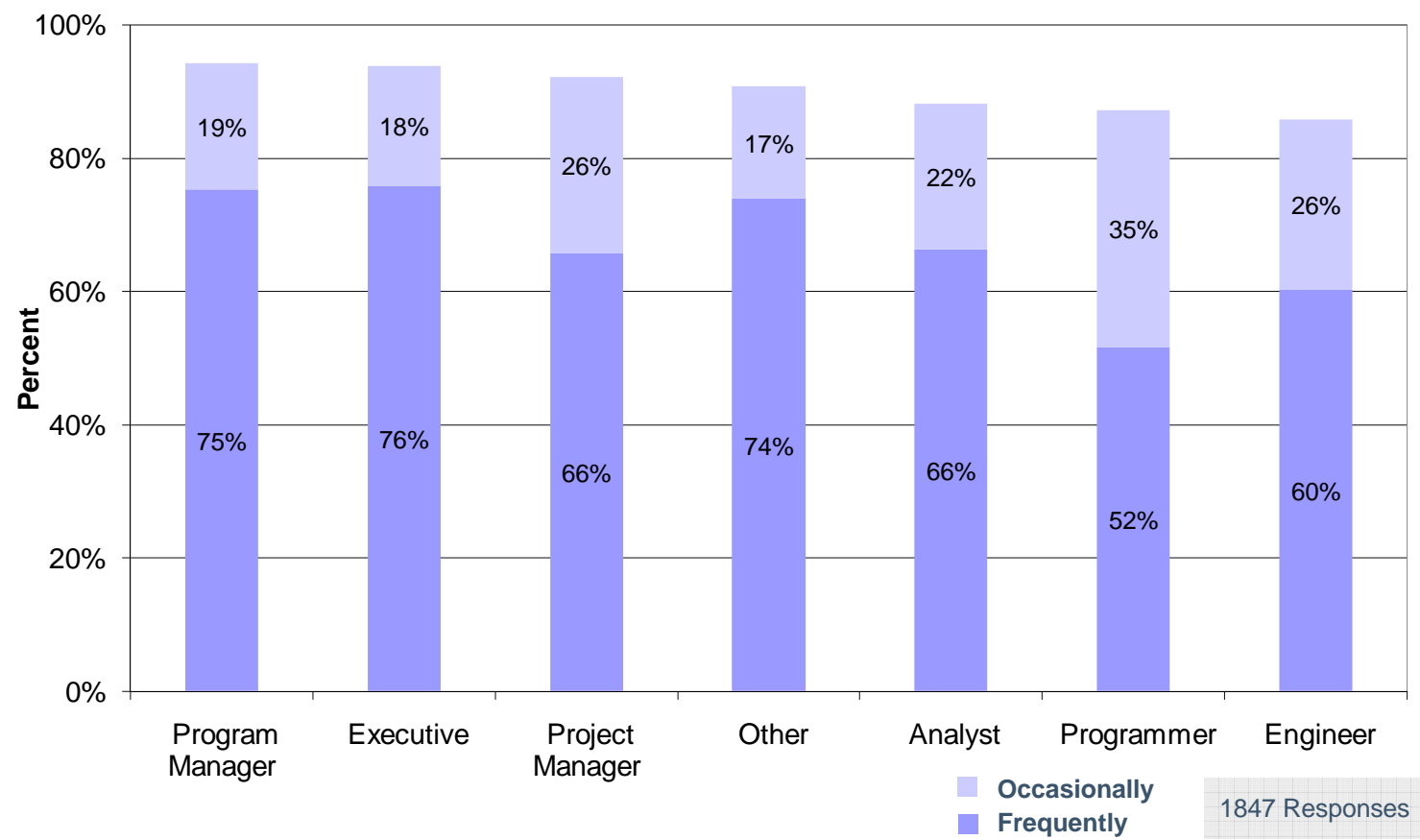
- No visible use or consequences associated with poor data collection or measurement
- No sustained management sponsorship

Missing data is reported as a valid value

- Can't distinguish 0 from missing when performing calculations



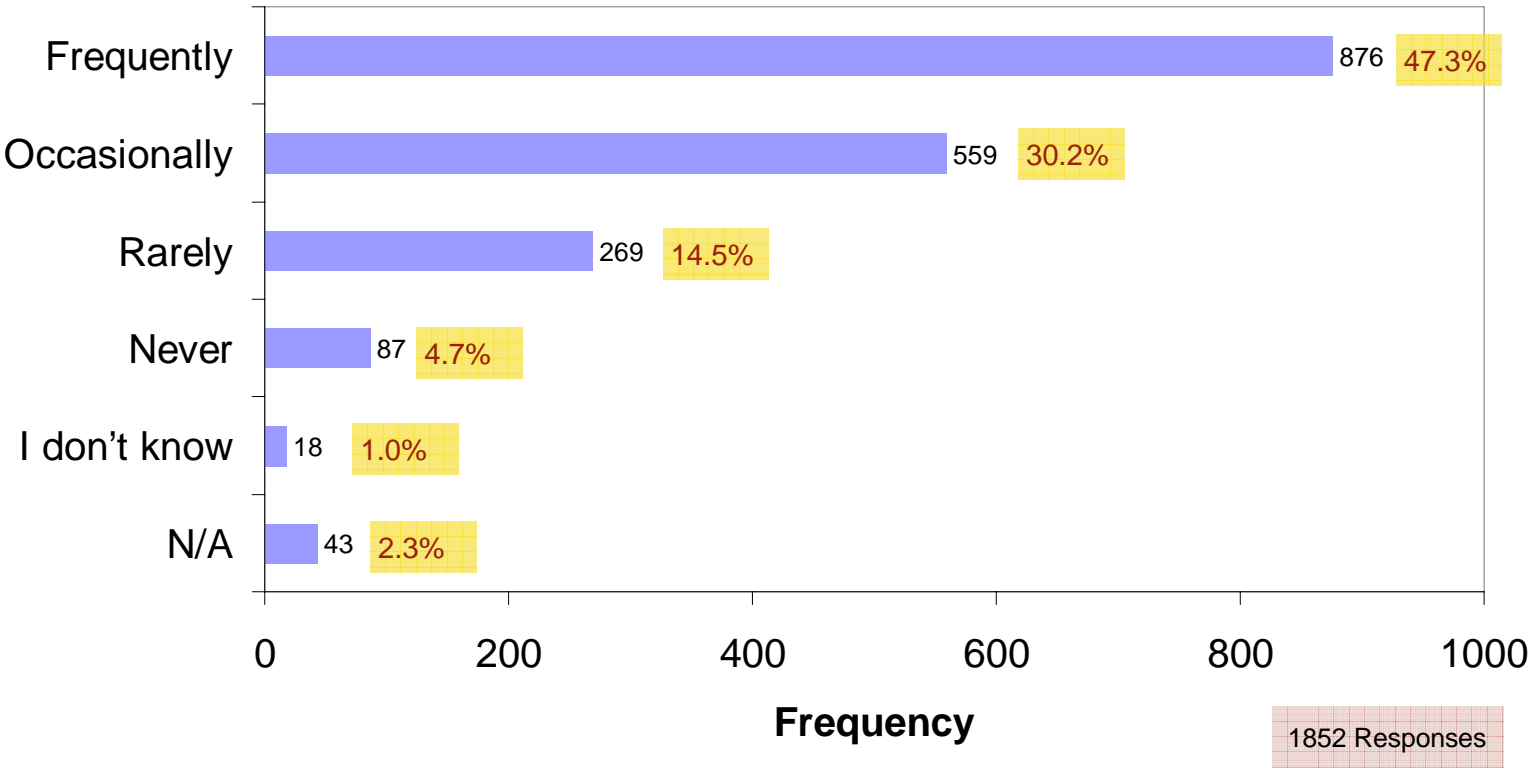
Purpose for Measuring is Understood



Source: CMU/SEI-2006-TR-009



Are Documented Processes Used?



Source: CMU/SEI-2006-TR-009



Cost of Poor Data Quality to an Enterprise – Typical Issues and Impacts

Typical Issues

- Inaccurate data [1-5% of data fields are erred]
- Inconsistencies across databases
- Unavailable data necessary for certain operations or decisions

Typical Impacts

Operational

- Lowered customer satisfaction
- Increased cost
- Lowered employee satisfaction

Tactical

- Poorer decision making & decisions take longer
- More difficult to implement data warehouses
- More difficult to engineer
- Increased organizational mistrust

Strategic

- More difficult to set strategy
- More difficult to execute strategy
- Contribute to issues of data ownership
- Compromise ability to align organization
- Divert management attention

Source: Redman, 1998



Impacts of Poor Data Quality

Inability to

- manage the quality and performance of software or application development
- Estimate and plan realistically

Ineffective

- process change instead of process improvement
- and inefficient testing causing issues with time to market, field quality and development costs

Products that are painful and costly to use within real-life usage profiles

Bad Information leading to Bad Decisions



The Need for a Measurement and Analysis Infrastructure Diagnostic

Quality of data is important

- Basis for decision making and action
- Erroneous data can be dangerous or harmful
- Need to return value for expense

Cannot go back and correct data once it is collected – opportunity/information lost

Keep from collecting the wrong type of data

Need to get the quality information to decision makers in an appropriate form at the right time



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MAID Objectives

Compare an organization's current measurement and analysis activities against a defined set of criteria

- Are we doing the right things in terms of measurement and analysis?
- How well are we doing those things?
- How good is our data?
- How good is the information we generate?
- Are we providing value to the organization and stakeholders?

Make recommendations for improvement

- How can identified gaps or weaknesses be addressed?
- How can we prepare for achieving higher maturity?
 - Many mistakes made in establishing M&A at ML2 and 3 that do not create a good foundation for ML4 and 5



Methods Overview

The MAID approach includes

- a thorough review of measurement-based planning documents, processes/procedures, analysis results, and management reports
- an evaluation of project and organizational data with respect to specified quality criteria
- a series of individual and group interviews with personnel who
 - collect measurement data
 - analyze, interpret and report the measurement info
 - use the reported data to make decisions
- a briefing and detailed report describing the strengths and weaknesses of the measurement program



Criteria for Evaluation: Measurement Planning Criteria₁

Measurement Objectives and Alignment

- business and project objectives
- prioritized information needs and how they link to the business, organizational, regulatory, product and/or project objectives
- necessary organizational and/or software process changes to implement the measurement plan
- criteria for the evaluation of the measurement process and quality assurance activities
- schedule and responsibilities for the implementation of measurement plan including pilots and organizational unit wide implementation
- The measurement plan is aligned and synchronized with the project plan and other organizational plans

Adapted from ISO 15939.



Measurement Planning Criteria₂

Measurement Process

- definition of the measures and how they relate to the information needs
- responsibility for data collection and sources of data
- schedule for data collection (e.g., at the end of each inspection, monthly)
- tools and procedures for data collection
- data storage
- requirements for data validation and verification procedures
- confidentiality constraints on the data and information products, and actions/precautions necessary to ensure confidentiality
- procedures for configuration management of data, measurement experience base, and data definitions
- data analysis plan including frequency of analysis and reporting

Adapted from ISO 15939.



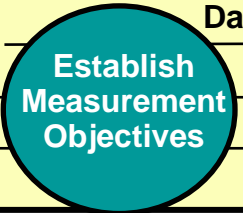
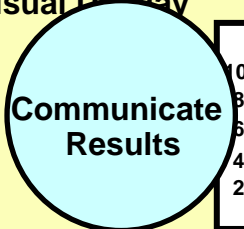
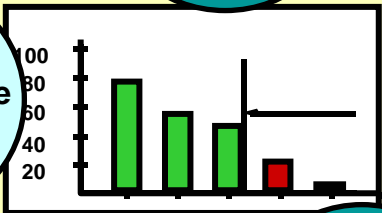
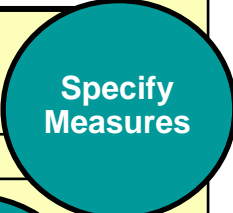


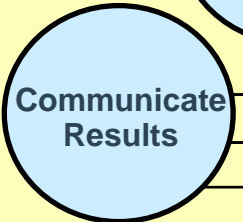
Criteria for Evaluation: Measurement Processes and Procedures


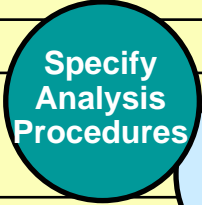
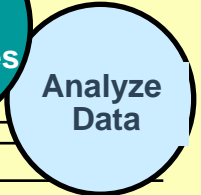
Measurement Process Evaluation

- Availability and accessibility of the measurement process and related procedures
- Defined responsibility for performance
- Expected outputs
- Interfaces to other processes
 - Data collection is integrated into work processes
 - Use of analysis results is incorporated into decision making
- Resources for implementation provided and appropriate
- Training and help are available



Documenting Measurement Objectives, Indicators, and Measures

Indicator Name/Title	_____	Date	_____
Objective	_____		_____
Questions	_____		_____
Visual Display	_____		_____
			_____
	Perspective		_____
Input(s)	_____		_____
Data Elements	_____		_____
Definitions	_____		_____
Data Collection	_____		_____
How	_____		_____
When/How Often	_____		_____
By Whom	_____		_____
Form(s)	_____		_____
Data Reporting	_____		_____
Responsibility for Reporting	_____		_____
By/To Whom	_____		_____
How Often	_____		_____

Data Storage	_____		_____
Where	_____		_____
How	_____		_____
Security	_____	_____	_____
Algorithm	_____		_____
Assumptions	_____		_____
Interpretation	_____		_____
Probing Questions	_____		_____
Analysis	_____		_____
Evolution	_____		_____
Feedback Guidelines	_____	_____	_____
X-reference	_____	_____	_____



Criteria for Evaluation: Data Definitions

Data Definitions (meta data)

- Completeness of definitions
 - Lack of ambiguity
 - Clear definition of the entity and attribute to be measures
 - Definition of the context under which the data are to be collected
- Understanding of definitions among practitioners and managers
- Validity of operationalized measures as compared to conceptualized measure (e.g., size as SLOC vs. FP)



Criteria for Evaluation: Data Collection

Data collection

- Implementation of data collection is consistent with data definitions and procedures
- Reliability of data collection (actual behavior of collectors)
- Reliability of instrumentation (manual/automated)
- Training in data collection methods
- Ease/cost of collecting data
- Storage
 - Raw or summarized
 - Period of retention
 - Ease of retrieval



Criteria for Evaluation: Data

Quality

- Data integrity and consistency
- Amount of missing data
 - Performance variables
 - Contextual variables
- Accuracy of collected data is assessed
- Validity of collected data is assessed
- Timeliness of collected data
- The precision and reliability (repeatability and reproducibility) of collected data are known
- Measurements are traceable to their source (meta data collected)

Audits of Collected Data



Criteria for Evaluation: Data Analysis

Data analysis

- Data used for analysis vs. data collected but not used
- Appropriateness of analytical techniques used
 - For data type
 - For hypothesis or model
- Analyses performed vs. reporting requirements
- Data checks performed
- Assumptions made explicit

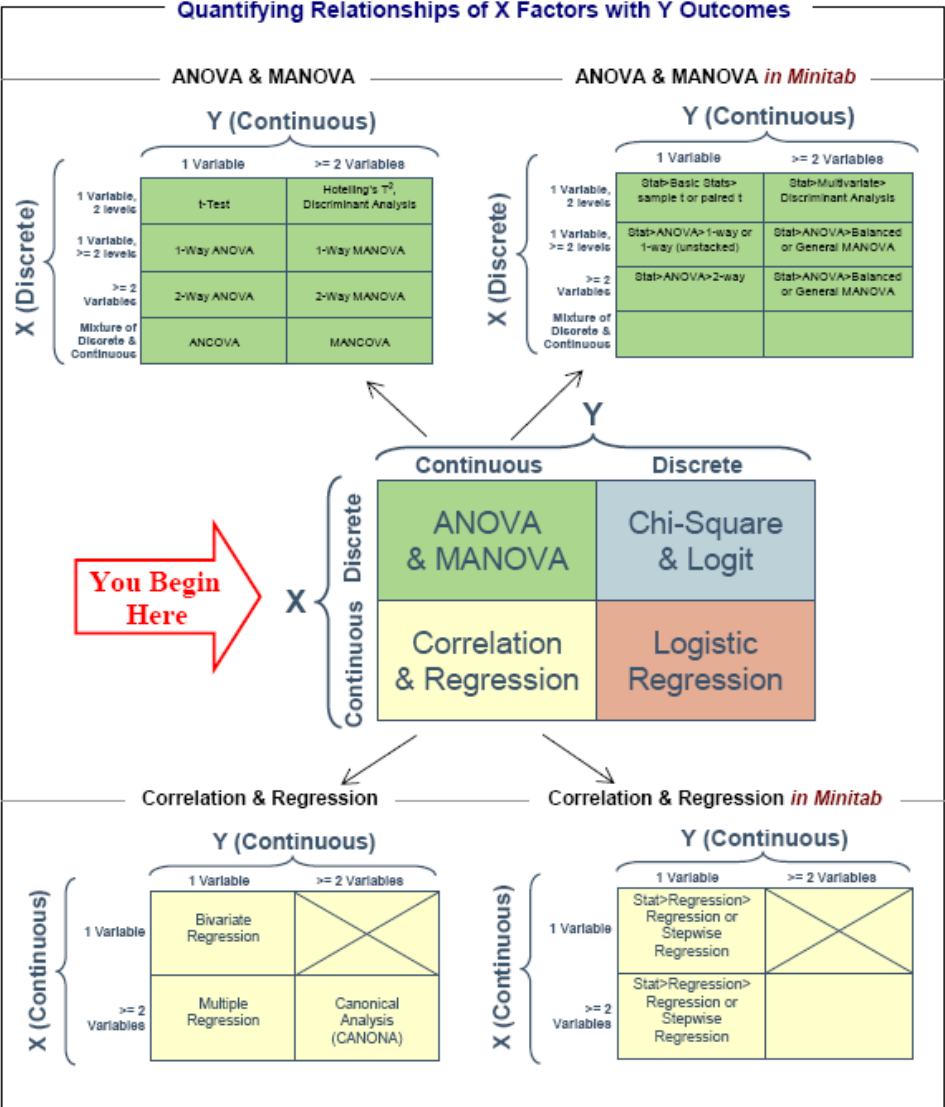


Appropriate Analysis: Types of Hypothesis Tests

Data Type	Interval or Ratio (Parametric Tests)		Ordinal (Non-Parametric Tests)		Nominal	Proportion
	Mean	Variance	Median	Variance / Fit	Similarity	Similarity
# Samples (Data groups) 1 Sample	1-sample t test	1-sample Chi-Square test	1 sample Wilcoxon Signed Ranks test	Kolmogorov-Smirnov Goodness of Fit test	>2 cells Chi-Square Binomial Sign Test =2 cells	1 Proportions test
2 Samples	<i>Independent</i> 2-sample t test Paired t test <i>Paired</i>	<i>Normal</i> F test Levene test <i>Not Normal</i>	<i>Independent</i> Mann Whitney U test Wilcoxon matched <i>Paired</i>	= Medians Siegel-Tukey test Moses test ≠ Medians	Fisher Exact test (1-way ANOVA); Chi-Square test	2 Proportions test
3+ Samples	ANOVA (1 & 2 way ANOVA; Balanced ANOVA; GLM) MANOVA (General & Balanced)	<i>Normal</i> Bartlett test Levene test <i>Not Normal</i>	<i>Independent</i> Kruskal-Wallis 1-way ANOVA Friedman 2-way ANOVA <i>Paired</i>	Van der Waerden Normal scores test	Chi-Square test	ANOM (Analysis of Means)



Analysis Evaluation: Appropriate Modeling



Criteria for Evaluation: Reporting

Reporting

- Evidence of use of the information
- Timing of reports produced
- Coverage of information needs
 - Per CMMI
 - Per Stakeholders
- Inclusion of definitions, contextual information, assumptions and interpretation guidance



Criteria for Evaluation: Stakeholder Satisfaction

Stakeholder Satisfaction

- Survey of stakeholders regarding the costs and benefits realized in relation to the measurement system
- What could be improved
 - Timeliness
 - Efficiency
 - Defect containment
 - Customer satisfaction
 - Process compliance

Adapted from ISO 15939.



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Summary

Measurement and analysis is a process

- It needs to be supported to be institutionalized and effective
- Some measurement error and diminished utility will result from choice of measurement infrastructure elements, procedures and instrumentation

Measurement Infrastructure Diagnostic:

- Characterizes performance of measurement system
- Identifies improvement opportunities for:
 - Measurement processes and data quality

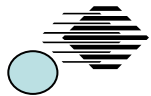
Good information from high quality measures and analyses to support decision making



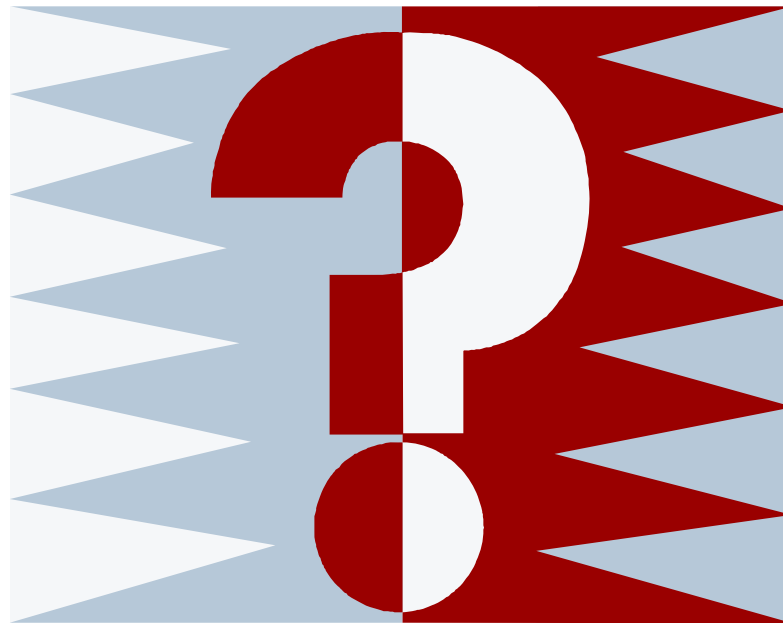


In God We
Trust,
All Others Bring
GoodData.

[Attributed to W. Edwards Deming, father of quality revolution]



Questions?



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