

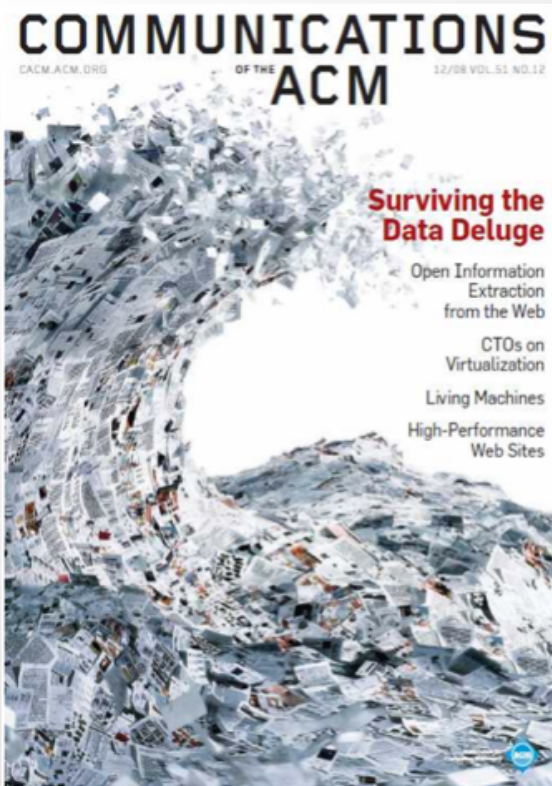
Prescriptive Analytics: Introduction

INCOSE WMA Tutorial
Kathryn Blackmond Laskey

George Mason University
Department of Systems Engineering and Operations Research

4 May 2013



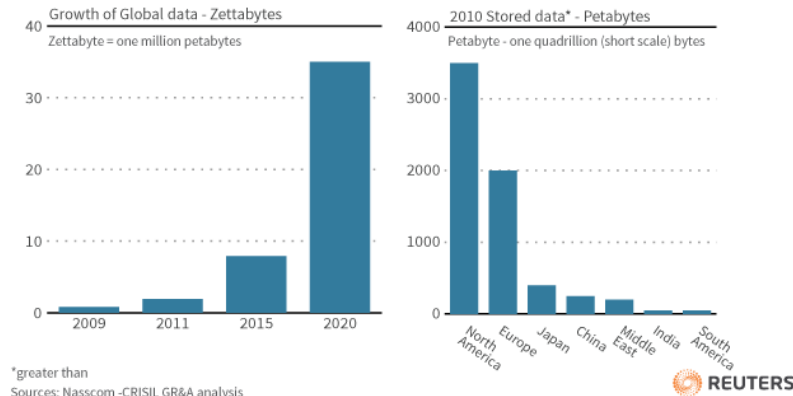


- **126 million** – The number of blogs on the Internet (as tracked by BlogPulse).
- **27.3 million** – Number of tweets on Twitter per day (November, 2009)
- **350 million** – People on Facebook.
- **50%** – Percentage of Facebook users that log in every day.
- **4 billion** – Photos hosted by Flickr (October 2009).
- **2.5 billion** – Photos uploaded each month to Facebook.
- **12.2 billion** – Videos viewed per month on YouTube in the US (November 2009).
- **924 million** – Videos viewed per month on Hulu in the US (November 2009).

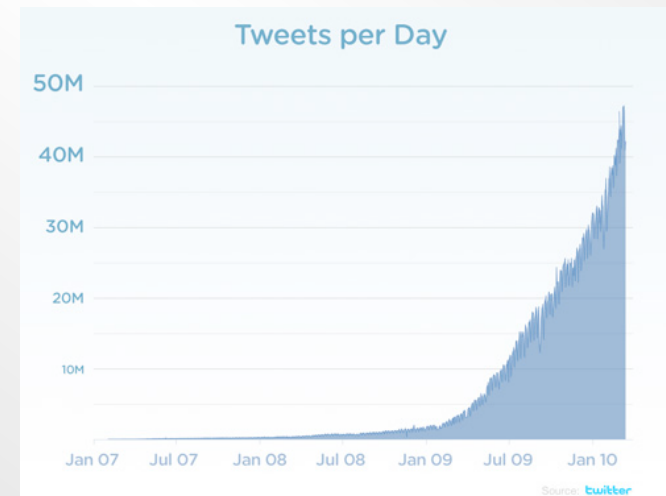
<http://royal.pingdom.com/2010/01/22/internet-2009-in-numbers/>

Big data growth

Big data market is estimated to grow 45% annually to reach \$25 billion by 2015



Reuters.graphic/Catherine Trevehan 05/10/12



A Useful Distinction

- Big Data
 - Collection, management and processing of large quantities of complex structured and unstructured data
- Data Analytics
 - Extracting value from data by using formal analysis to:
 - Find patterns
 - Understand meaning of patterns
 - Make predictions
 - Recommend decisions



Driving Forces

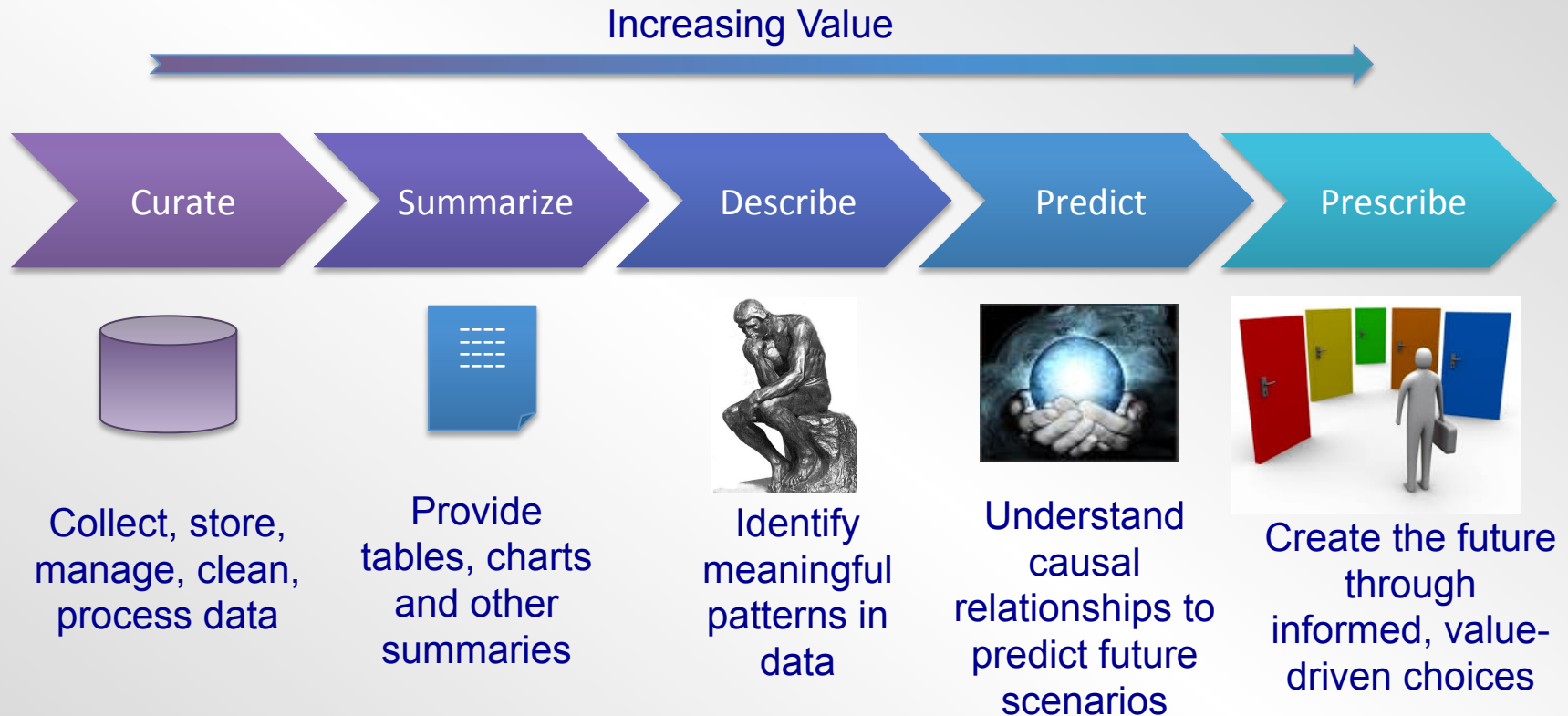
- Unprecedented data availability
- Intense competition
- Culture of constant improvement
- Changing customer demographic
- Expanding customer expectations



Why Analytics?

- Competitive differentiator
 - Individualized customer experience
 - Faster and better service
 - Cost effective processes
- Increasingly pervasive
 - Do you remember life without your computer?
Your cell phone? The web?
 - Soon it will be inconceivable to do business without analytics

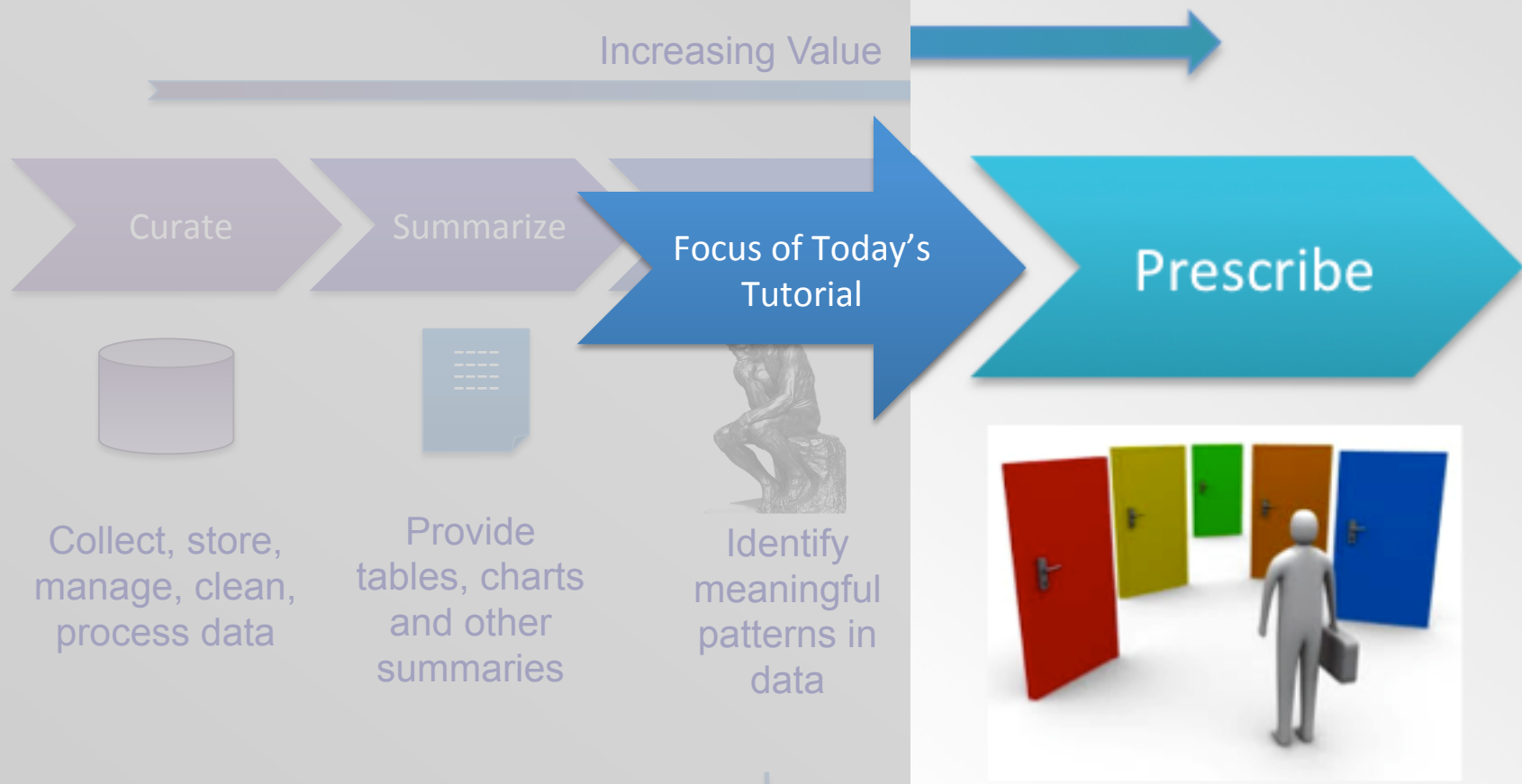
Analytics Value Chain



Analytics:
Descriptive / Predictive / Prescriptive



Analytics Value Chain

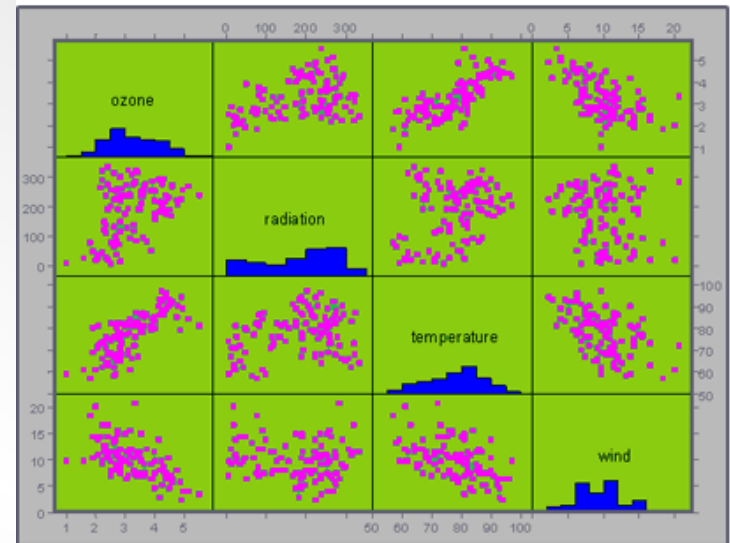


Analytics:
Descriptive / Predictive / Prescriptive



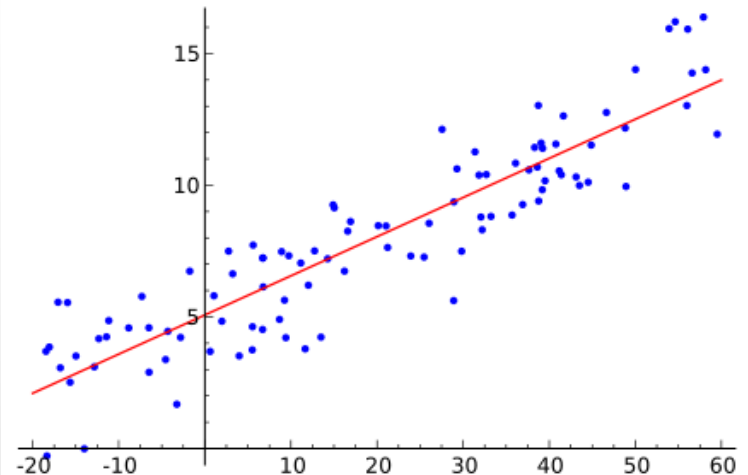
Descriptive Analytics

- Identify patterns in data
- Common categories of analytical methods
 - Data visualization methods
 - Data transformation
 - Outlier detection
 - Data mining methods
 - Clustering
 - Exploratory identification of trends and patterns



Predictive Analytics

- Analyze data to make predictions
 - Identify and formally verify patterns in data
 - Understand cause and effect relationships
 - Extrapolate patterns to future
- Common categories of analytical methods
 - Regression analysis
 - Time series models
 - Machine learning methods



Prescriptive Analytics

- Suggest actions to decision maker based on descriptive and predictive analytics
 - Frame the problem
 - Identify candidate actions
 - Predict consequences of actions
 - Assess value of consequences
 - Suggest highest value actions
- Common categories of analytical methods
 - Optimization
 - Simulation



Analytics Supports Better Decisions

- Example: FedEx

- Package scans and active sensors in high-value packages provide data on shipments
- Events are analyzed in real time and shipments rerouted in case of problems
- Analytics are used to identify and respond to customer service issues



- Example: Petroleum industry

- Analytics improves problem diagnosis, improves maintenance and repair policies, reduces cost and prevents catastrophic failures
- Analytics helped ConocoPhillips predict ice floe movement and extend the drilling season by weeks



Evolution of Decision Support



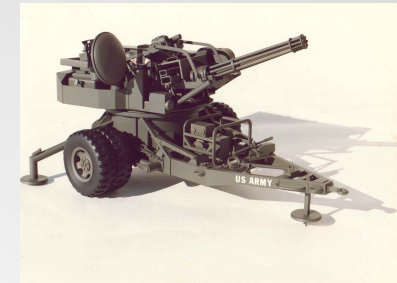
Decisions and Models

- A model is a representation of a system that can be used to answer questions about the system
 - “All decisions are based on models... and all models are wrong.” – John D. Stearman
 - “All models are wrong but some are useful” – George Box
- Models are constructed from:
 - Past data on the system
 - Past data related to the system
 - Judgment of subject matter experts
 - Judgment of experienced model builders



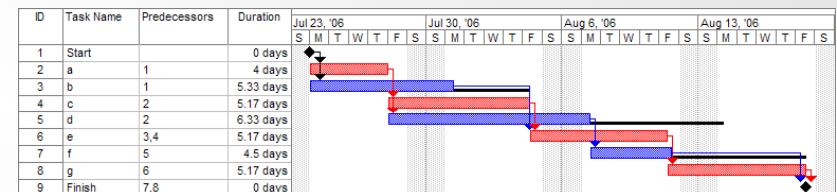
Models in Engineering

- Engineers design a system by:
 - Building a model to represent the system they want to design
 - Manipulating the model
 - Using behavior of the model to
 - Predict behavior of the system
 - Evaluate and compare alternative design options



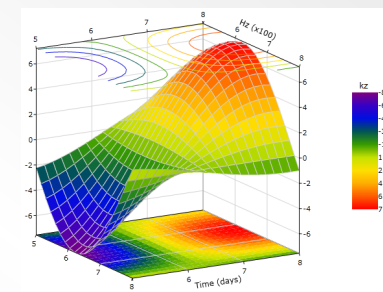
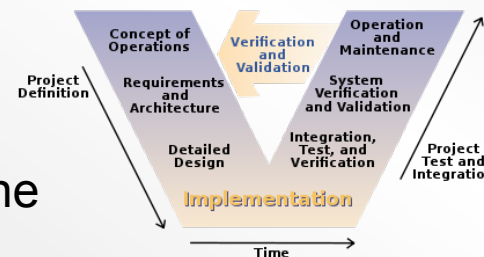
- Types of representation

- Physical
- Mathematical
- Computer
- Verbal



- Examples

- Wind tunnel and model airplane
- CAD model of a bridge
- Computer simulation of traffic flows on highway network
- Linear program model for inventory planning
- Bayesian spam filter



Human Decision Makers and Models

- People are good at:
 - Identifying what objectives are important
 - Identifying what features are relevant
 - Identifying relationships
 - Generating options
- People need support to:
 - Bump out of pre-conceived ideas and established conventions
 - Integrate large numbers of factors
 - Combine numerical and statistical information with judgment
 - Perform tedious bookkeeping
 - Coordinate among multiple actors
- Effective models produce understandable rationale for recommendations



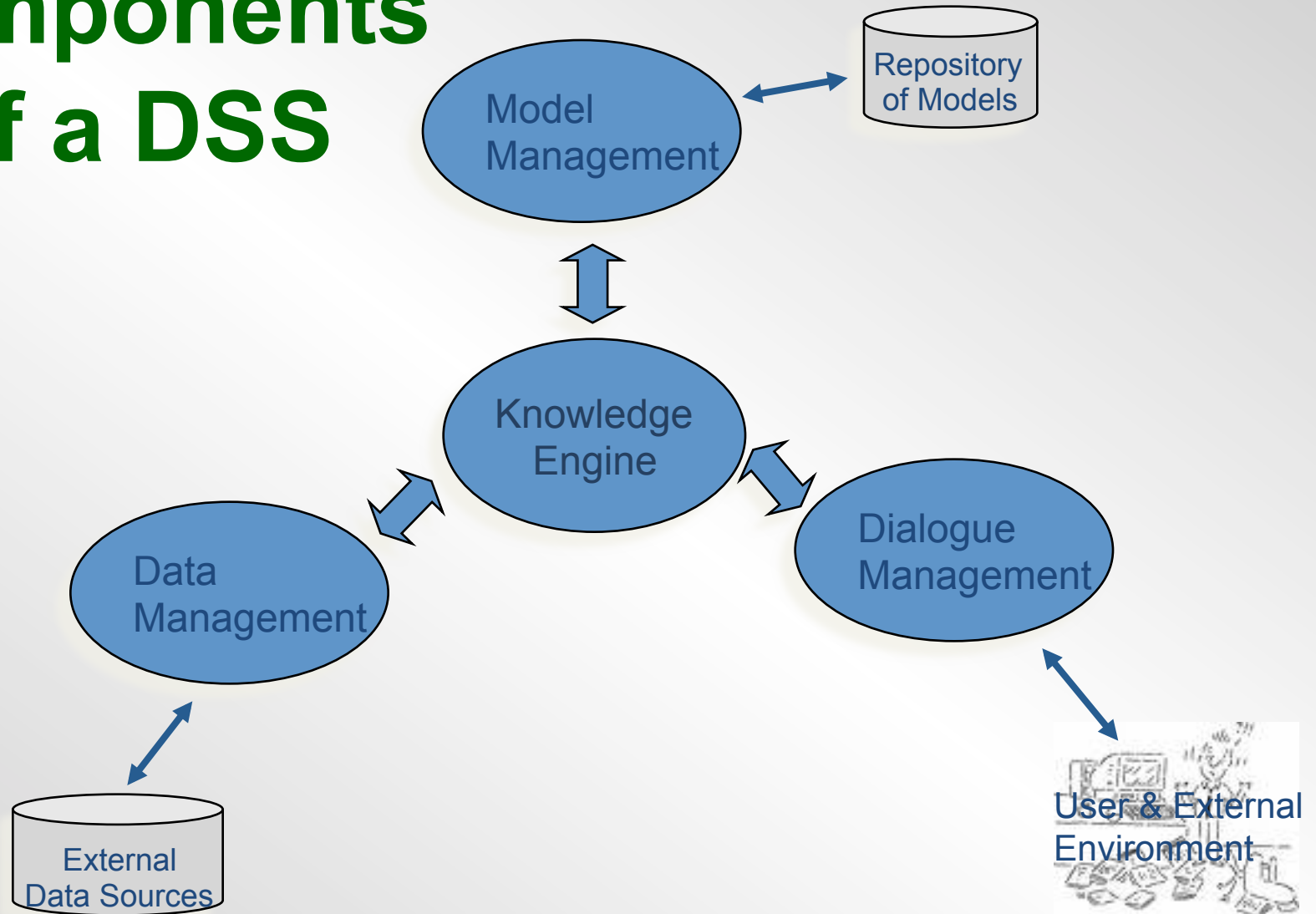
Cognitive Tools

(von Winterfeldt and Edwards)

- We would never start a construction project without tools
- We should not make important decisions without effective *cognitive tools*
- Cognitive tools can assist with:
 - Problem structuring
 - Elicitation of human judgmental inputs
 - Organizing and displaying relevant data
 - Aggregating inputs to produce
 - Predictions of outcomes for options suggested by decision maker
 - Recommendations of options for decision maker to consider
 - Understanding strengths and weaknesses of candidate solutions
 - Selecting a solution
 - Justifying the selected solution
 - Implementing the selected solution
- Prescriptive analytics gives us cognitive tools



Functional Components of a DSS



Decision Support Trends

- IT is increasingly pervasive
- Computer hardware is increasingly smaller and more powerful
- Systems are increasingly interconnected
- The Web is interwoven into all aspects of life
- Demand for usable, flexible, powerful decision support will continue to grow
- Decision support is increasingly embedded into consumer and business products
- User expectations are exploding



Data Analytics at GMU

- Graduate certificate program starts Fall 2013
 - Broad overview of value chain for Big Data Analytics
 - Framework for methodologies to organize, visualize, analyze, and generate value from data
 - Interdisciplinary
- MS program awaits state approval
- Prescriptive analytics
 - Course in certificate program
 - Track in MS program



Schedule

- 9:00 Introduction
- 9:30 Modeling Decision Problems
- 10:15 Break
- 10:30 Model-Based Systems Engineering and Prescriptive Simulation
- 11:15 Optimization Methods for Prescriptive Analytics
- 12:15 Conclusion



