



INCOSE WMA Tutorial on Prescriptive Analytics  
George Mason University  
May 4<sup>th</sup>, 2013

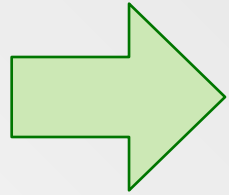
# Decision Theory in Data Analytics



Where Innovation Is Tradition



# Agenda



- ❖ Decision Theory and Prescriptive Analytics
- ❖ Deciding with Values
- ❖ Qualitative and Quantitative Assessment
- ❖ Modeling Uncertainty
- ❖ Risk Analysis

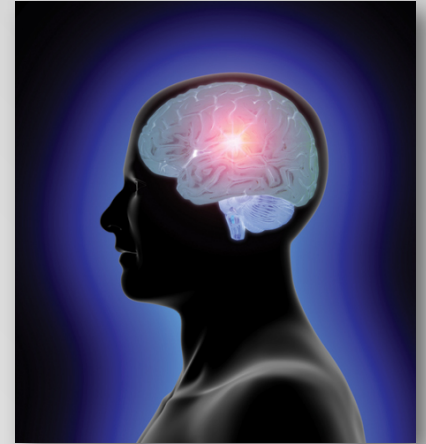
# What makes a decision hard to make?

- Competing objectives
- Uncertainty
- Differing perspectives of Decision makers
- Disagreement over what is to be accomplished
- Politics
- Lack of information
- Complexity



# What is a Decision?

- A decision is an irrevocable commitment of resources
  - Money
  - Time
  - People
- A decision **must** have
  - Alternatives to choose among
  - Outcomes for each alternative
  - A value structure within which to order / rank outcomes
  - Assumptions relating to all of the above
- A decision **may** have
  - A probability measure on possible outcomes
  - A probability measure on future conditions (states of nature)
  - An active opponent trying to defeat your decision



# The Elements of a Decision

- Objectives and means
- Alternatives to choose between
- Uncertainty in Events and Outcomes
- Consequences of the Decision



# What is Decision Analysis?

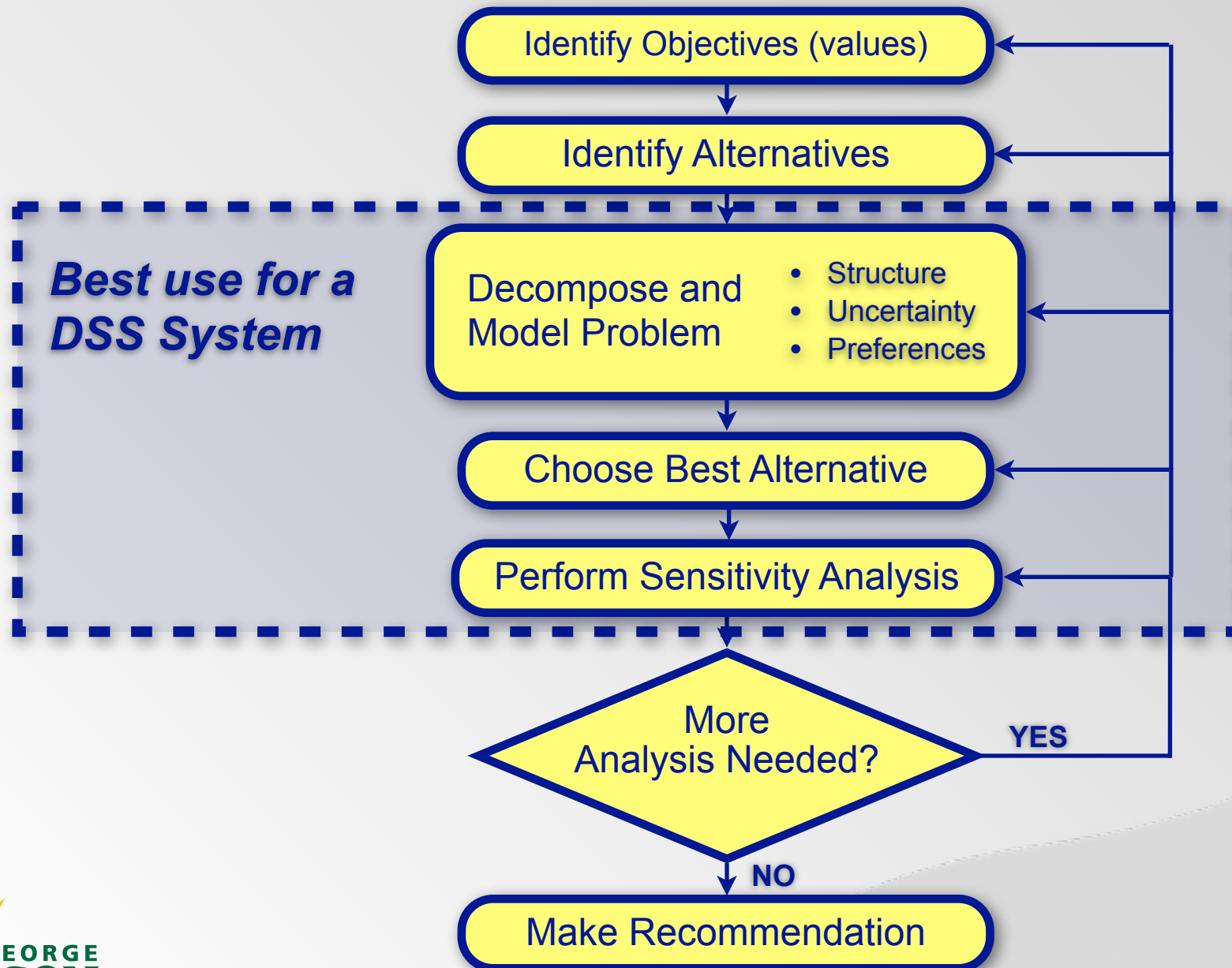
- A Method to:
  - Organize or structure complex problems for analysis
  - Deal with tradeoffs between multiple objectives
  - Identify and quantify sources of uncertainty
  - Incorporate subjective judgments



# Decision Analysis Tenets

- Quality decision making requires a *systematic process* to incorporate
  - Information, expert opinion, and preferences
- Complex decisions in large organizations involve
  - *Functional experts* (inside)
    - R&D, engineers, operations, production, finance, etc.
  - *Interested stakeholders* (outside)
    - Stockholders, government, community, etc.
- *Quantification* offers significant benefits
  - Clarifies thinking
  - Values
  - Uncertainties (Probability)
  - Consequences
  - Improves communications
  - Enables logical reasoning
- 4. *Support decision maker* judgments by providing *insights* (more about dealing with decision makers later)

# Prescriptive Analytics Process





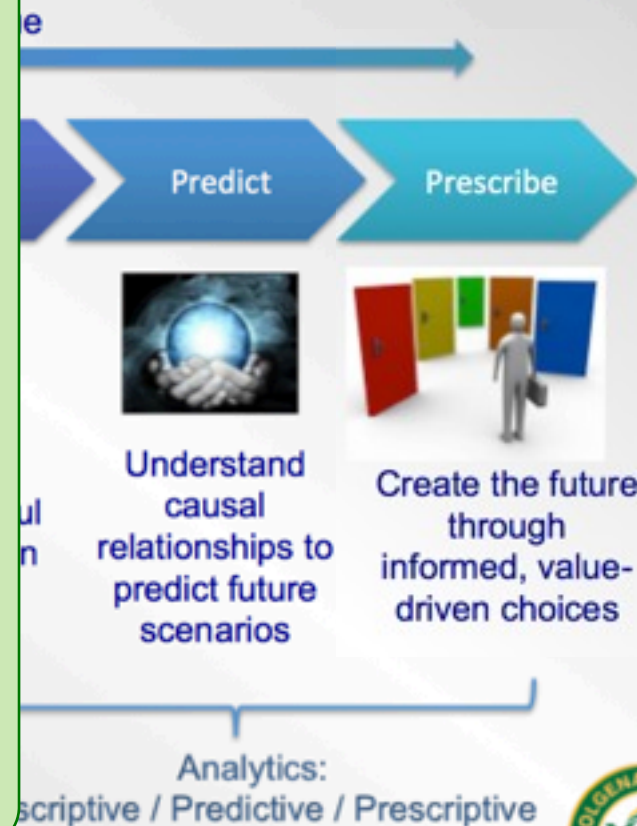
# Where Decision Theory Helps the Most?

## Analytics Value Chain

Decision Theory provides a sound framework for:

- Identifying and structuring values
- Generating better alternatives based on values
- Evaluating alternatives according to values

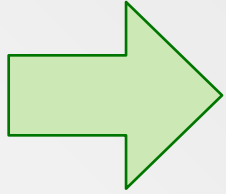
... and is thus essential to  
**Prescriptive Analytics**



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# Agenda

❖ The Decision Analysis Process



❖ Deciding with Values

❖ Qualitative and Quantitative Assessment

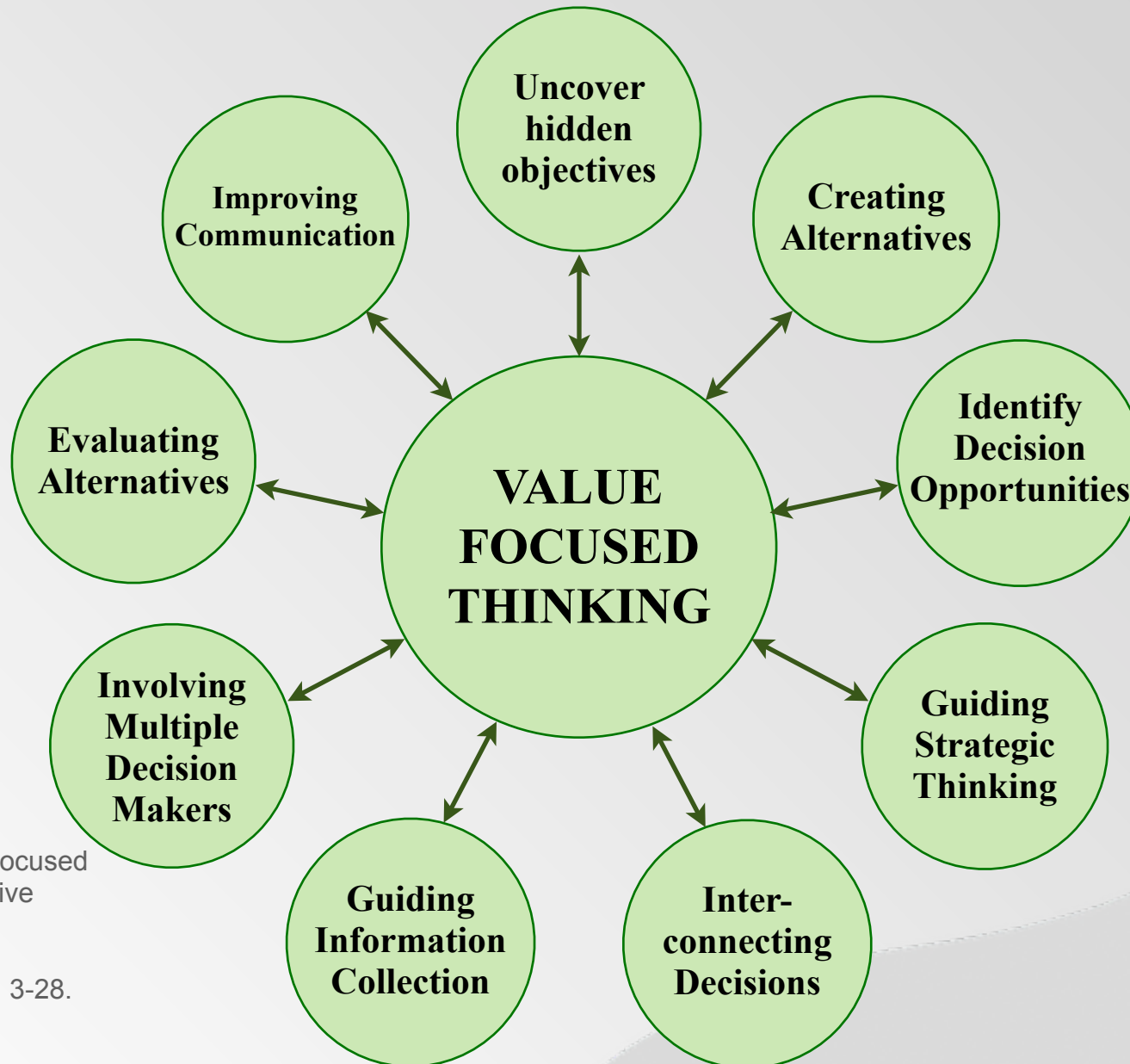
❖ Modeling Uncertainty

❖ Risk Analysis

# Value Focused Thinking

- Naïve decision making is alternative-focused
  1. Problem arises
  2. Begin problem solving
  3. List alternatives
  4. Choose alternatives
- Why is this a limited way to think?
  - Reactive, not proactive
  - Encourages incomplete analysis
  - Focuses on status quo and a few minor variations
- Value focused thinking starts with decision-maker's values
  - Identify and structure values
  - Generate better alternatives based on values
  - Evaluate alternatives according to values

# Where does thinking about values lead?



Keeney, Ralph L., Value-Focused Thinking: A Path To Creative Decisionmaking, Harvard University Press, Cambridge, MA, 1992, pp. 3-28.

# Value Modeling

## ■ Facts, values, objectives

- Facts - objectively verifiable pieces of information about events or circumstances in the world
- Values – how a stakeholder feels about the events or circumstances
- Objectives – what a stakeholder is trying to achieve

## ■ Decision maker's values should drive decision

## ■ Objectives should be consistent with values

## ■ Value model

- Explicitly represents benefits and costs of each decision alternative
- Supports rational comparison of alternatives
- Usually build qualitative value model first and move to quantitative model as required
- Helps overcome things that make decision difficult

# Identifying Values

- Interact with decision makers
- Elicit feedback throughout DSS development
- Values are stated through **objectives**
- 3 Features of objectives
  - (1) Decision Context
  - (2) Object
  - (3) Direction of Preference
- Ex: Forest products company wants to “minimize environmental impact”
  - (1) Decision Context - ***Harvesting Trees***
  - (2) Object - ***Environmental Impact***
  - (3) Direction of Preference - ***Minimum is best***

# Ways to Identify Values

(due to Professor Gregory Parnell, USMA)

- **Gold Standard:** based on an approved vision, policy, strategy, planning, or doctrine document
  - Values have been thought about, discussed, and written down
  - Use work that has already been done and approved
- **Platinum Standard:** based on interviews with decision-makers and stakeholders
  - Often difficult to get enough time with stakeholders and DMs
  - Use Affinity Diagram for group settings
- **Silver Standard:** uses data provided by stakeholder representatives
  - When real DMs and stakeholders are not available
  - Still use Affinity Diagrams for groups
- **Combined Standard:** combination of the above

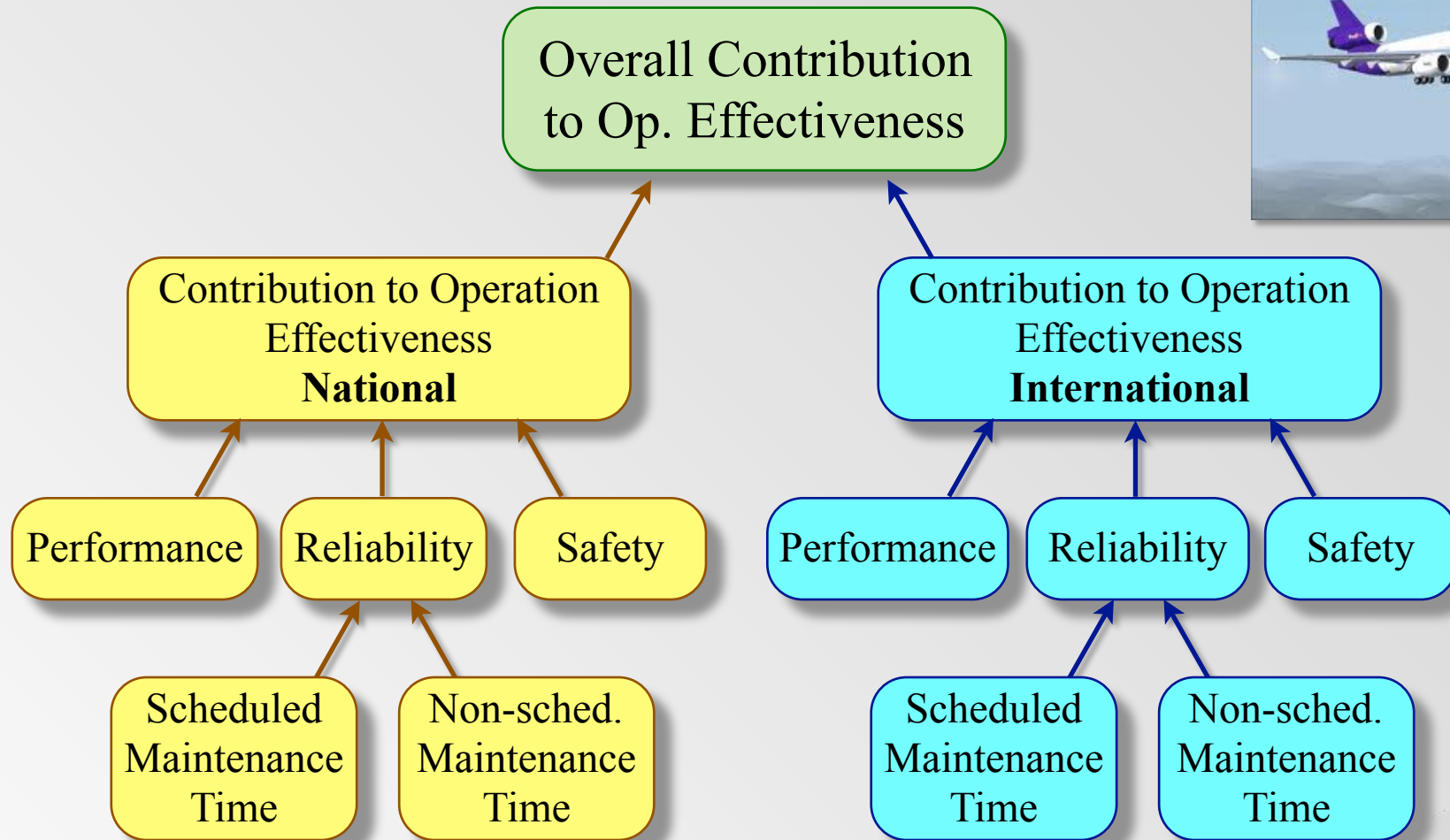
# Structuring Objectives

- Initial list has “non-objectives”
  - Alternatives
  - Constraints
  - Criteria for evaluation
- Convert “non-objectives” into objectives
- Types of objective
  - **Strategic objectives** - objectives that describe a person’s life or an organization’s purpose
  - **Fundamental objectives** -- essential to decision maker’s concerns
  - **Means objectives** -- important only because of relationship to fundamental objectives

**Strategic objectives are stable over time  
although the means to obtain these  
objectives change with time and context.**



# Value Hierarchy for Logistic Operations



**General MOEs**

**Specific MOEs**

- Complete
- Consistent
- Measurable

- Non-overlapping
- Hierarchically organized
- 7 or fewer per group

# Generating Alternatives

- No recipe – but “Values First!”
- Consider obvious alternatives (including status quo)
- Look for alternatives that satisfy subset of objectives (2 or 3 at a time)
- Try to modify alternatives to overcome shortcomings
- Combine several alternatives into a new alternative
- Focus on strategic objectives
  - Most important
  - Broader in scope
  - Will do most good

# Agenda

- ❖ The Decision Analysis Process

- ❖ Deciding with Values



- ❖ Qualitative and Quantitative Assessment

- ❖ Modeling Uncertainty

- ❖ Risk Analysis

# Qualitative Assessment

- Build a qualitative value structure and generate a list of alternatives
- This process helps to:
  - Clarify thinking of all participants
  - Improve communication among stakeholders
  - Identify the most important objectives
  - Specify ways to measure how well objectives are satisfied
- Eliminate obviously inferior alternatives
  - If an alternative is dominated (scores worse than another alternative on all attributes) we can eliminate it
  - We may be able to eliminate some non-dominated alternatives if they are clearly worse than other alternatives
- If several alternatives still remain, we need build to quantify tradeoffs among competing objectives

# Quantitative Value Function

- a.k.a. “Objective Function” or “Multi-attribute value function”
- Denoted  $v(x)$
- Assigns a number to the consequences,  $x$ , of an alternative
- Used to determine preference among alternatives
- Types of value functions
  - Ordinal value function - ranking only
  - Measurable value function – measures strength of preference
  - Utility function – measures risk attitude
- Form depends on relationships between attributes
  - Attributes are quantitative measures of how well a consequence satisfies an objective
- Convention:  $v(x^A) > v(x^B)$  iff we prefer alt A to alt B

# Structures of Value Functions

## Additive Value Function

$$v(x_1, x_2, \dots, x_n) = \sum_{i=1}^n k_i v_i(x_i)$$

where  $k_i$  is a positive scaling constant  
 $v$  and  $v_i$  are value functions scaled from 0 to 1



All combinations of attributes are  
Preferentially Independent!

# Structures of Value Functions

## ■ Multiplicative Value Functions

$$1 + kv(x_1, x_2, \dots, x_n) = \prod_{i=1}^n [1 + kk_i v_i(x_i)]$$



All pairs of attributes are Weak-Difference Independent

– Example:

$$v(x_1, x_2) = k_1 v_1(x_1) + k_2 v_2(x_2) + k_3 v_1(x_1) v_2(x_2)$$

– Looks like regression equation with interactions

– More difficult to use



Bottom Line: Use Additive Independent attributes!

# Example (1 of 4)

- Problem Statement: What new aircraft system should Fedex buy to improve the effectiveness of its logistic operations?
- Procedure:
  - (0) Choose Fundamental Objective
    - Best mix of safety, reliability, and performance
  - (1) Determine important Means Objectives (multiple scenarios)
    - Contribution to safety
    - Contribution to reliability
    - Contribution to operational performance



# Example (2 of 4)

## (2) Find measures for objectives

- Safety: Events per 100K hours, Accidents per 100K hours
- Reliability: Maintenance downtime, Non-Scheduled downtime
- Operational Performance: Tons per flight, Average Ground Time

Note: Measures will be computed separately for each scenario of interest.

# Example (3 of 4)

## (3) Scale measures

- Convert each objective to scale from 0 to 1
  - 0 means “reasonable worst” score
  - 1 means “reasonable best” score
  - others measured proportionally
- Simplest transformation is linear:

$$v_{at\ trib}(x) = \frac{x - worst_{at\ trib}}{best_{at\ trib} - worst_{at\ trib}}$$

- Limitations:
  - Ignores riskiness of alternatives
  - Assumes increment is valued equally at any point on scale from worst to best
  - Can define more sophisticated rescaling functions

# Example (4 of 4)

## (4) Weight the objectives (criteria)

**Answer the question:** How do we trade increased value in one objective for lower value on the others?

- Methods:
  - Trade-off
  - Smarter
  - Swing weights \*
  - Lottery weights
- Often totally subjective
- Avoid “Importance Weights”
- Must consider range of variation of attributes

# Elicitation Techniques

- Swing Weighting
  - Emphasizes the Range of Variation of variables
- Trade-Offs
  - A Derivative of Swing Weighting in Logical Decisions
- Simple Multi Attribute Rating Technique (SMART&SMARTER)
  - An Ordinal Technique
- AHP

# Swing Weights

- Based on comparing ranges of variation of attributes
- Can be used for non-quantitative attributes
- Method:
  - Find “Worst Conceivable Alternative”
    - Lowest score on each attribute
    - May be imaginary
  - Pick attribute that gives greatest improvement when “swings” to highest level
  - Pick attribute that gives next highest increase when swung to highest level
    - by percentage - how does it compare with the first?
    - never greater than 1 since first is best
  - Repeat for rest of attributes
  - Solve for weights

$$k_2 = k_{21} w_1$$

$$k_3 = k_{31} w_1$$

⋮

$$k_n = k_{n1} w_1$$

$$\sum_{i=1}^n k_i = 1$$

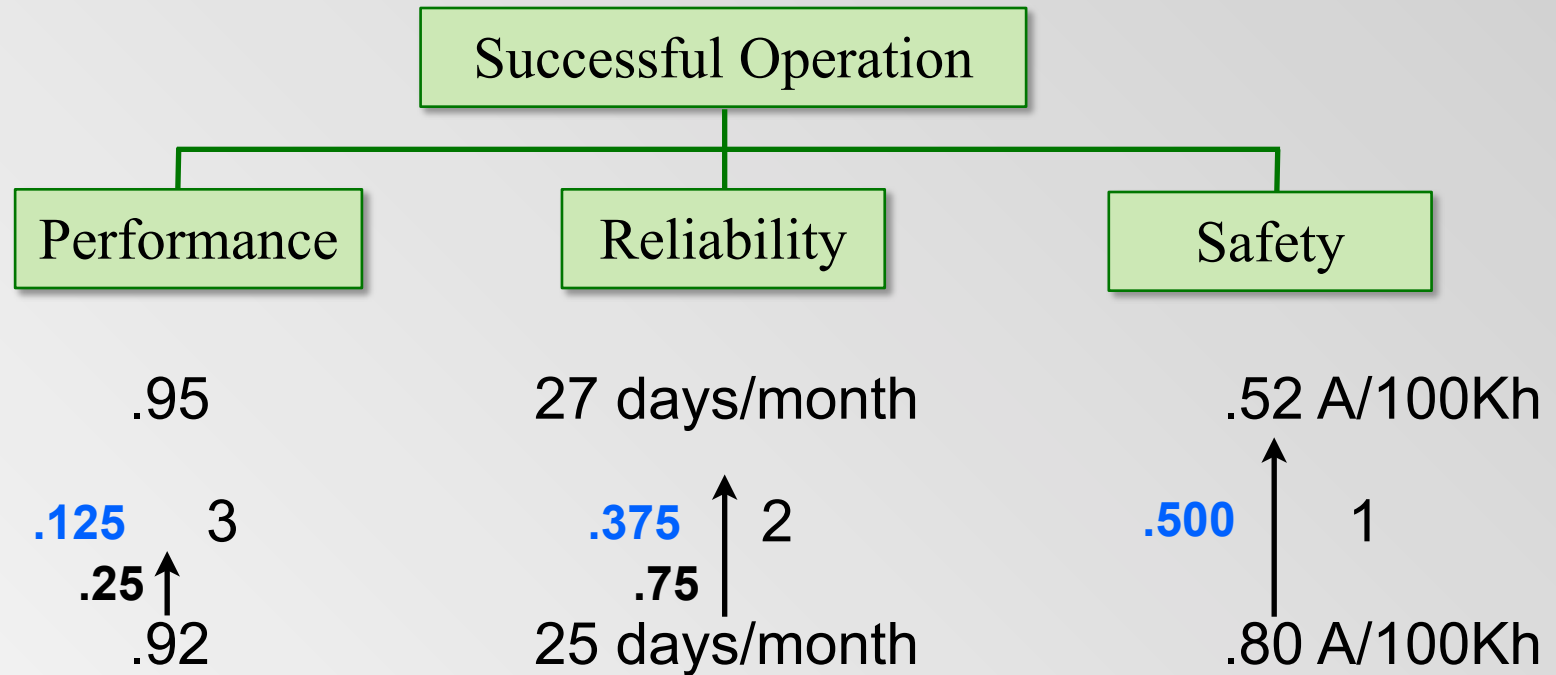
- $k_i$  is weight of  $i^{\text{th}}$  attribute
- $p_{i1}$  is percentage improvement compared with attribute 1

# Swing Weights in a Value Hierarchy

- Each level has its own additive value function with weights
  - Weights sum to 1
  - Weights correspond to relative importance of swings on each of the criteria
  - Weights assessed as in any additive value function
- Each additive value function is independent of the others

[http://c4i.gmu.edu/~pcosta/files/INCOSE\\_DTexample.xlsx](http://c4i.gmu.edu/~pcosta/files/INCOSE_DTexample.xlsx)

# Example: Swing Weight Elicitation



$$K_{\text{cost}} = (1.00) * K_{\text{cost}} = 0.5$$

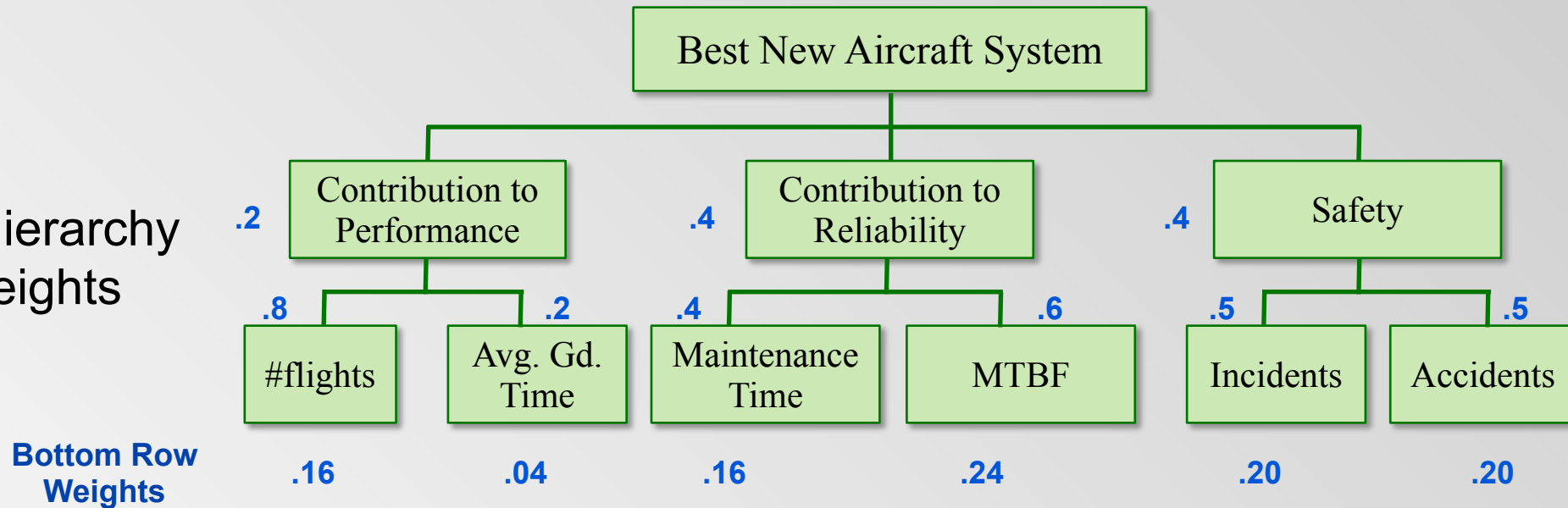
$$K_{\text{time}} = (0.75) * K_{\text{cost}} = (0.75) * (0.5) = 0.375$$

$$K_{\text{perf}} = (0.25) * K_{\text{cost}} = (0.25) * (0.5) = 0.125$$

$$1 = 2.0 * K_{\text{cost}} \longrightarrow K_{\text{cost}} = 0.5$$

# Example Problem – Choose the Best Aircraft

Value Hierarchy  
With Weights



Simulation  
Results

Alternative Aircraft System	# flights per Op. Month	Avg. Ground Time (m)	Maintenance Time (days)	MTBF (days)	# of Incidents /100Kh	# of Accidents / 100Kh
1	385	26	8	85	22	0.52
2	420	32	4	92	28	0.73
3	350	38	10	120	18	0.80

### Scaled Scores for Flights Per Month of Operation

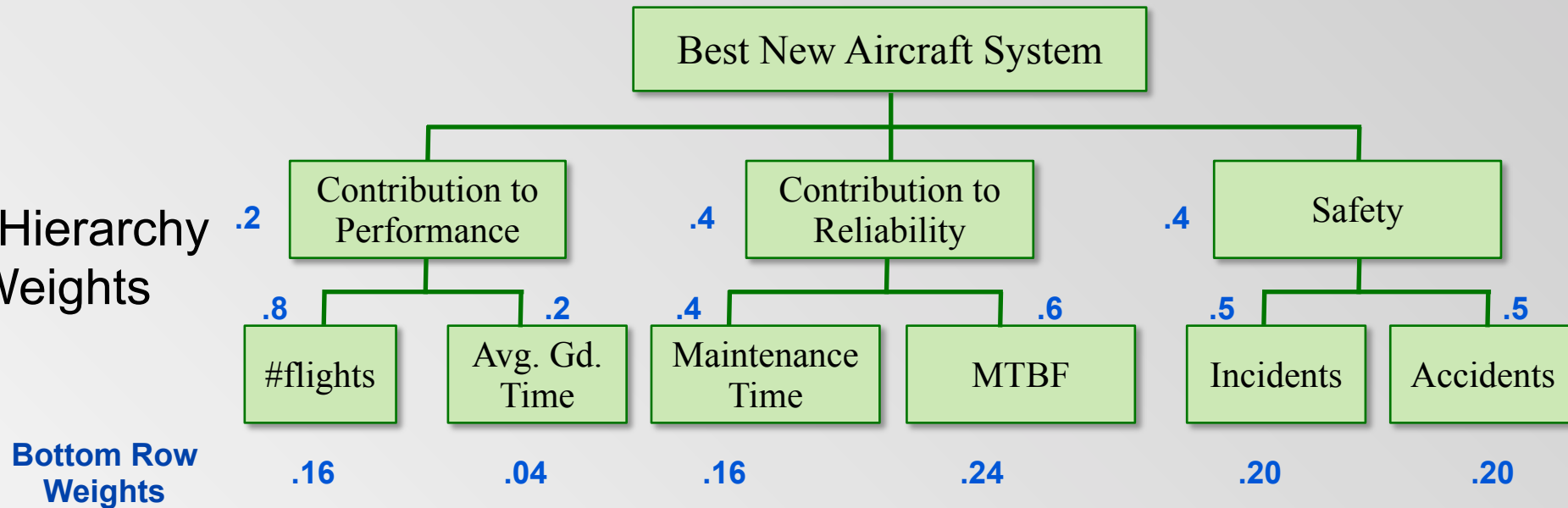
$S_{f2} = 0$  (since it is the worst)  $S_{f3} = 1$  (since it is the most preferred)

$S_{f1} = (x-worst)/(best-worst) = (385 - 420)/(350 - 420) = 0.5$



# Example Problem – Choose the Best Aircraft

Value Hierarchy  
With Weights



Simulation  
Results

Alternative Aircraft System	# flights per Op. Month	Avg. Ground Time (m)	Maintenance Time (days)	MTBF (days)	# of Incidents /100Kh	# of Accidents / 100Kh
1	0.5	1	0.33	0	0.6	1
2	1	0.5	1	0.2	0	0.25
3	0	0	0	1	1	0

$$S1 = (.16)(0.5) + (.04)(1) + (.16)(.33) + (.24)(0) + (.2)(.6) + (.2)(1) = 0.49$$

$$S2 = .44$$

$$S3 = .44$$

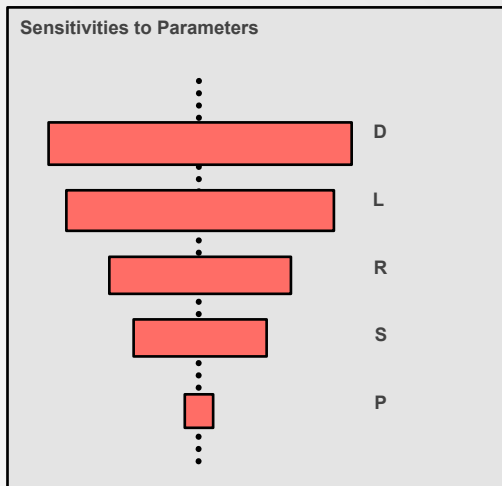
# Sensitivity Analysis

- Sensitivity analysis means varying the inputs to a model to see how the results change
- Sensitivity analysis is a very important component of exploratory use of models
  - model is not regarded as “correct”
  - sensitivity analysis helps user explore implications of alternate assumptions
  - human computer interface for sensitivity analysis is difficult to design well
- In many models we need to make assumptions we cannot test
  - Sensitivity analysis examines dependence of results on these assumptions

# Visualizing Sensitivity Analysis Results

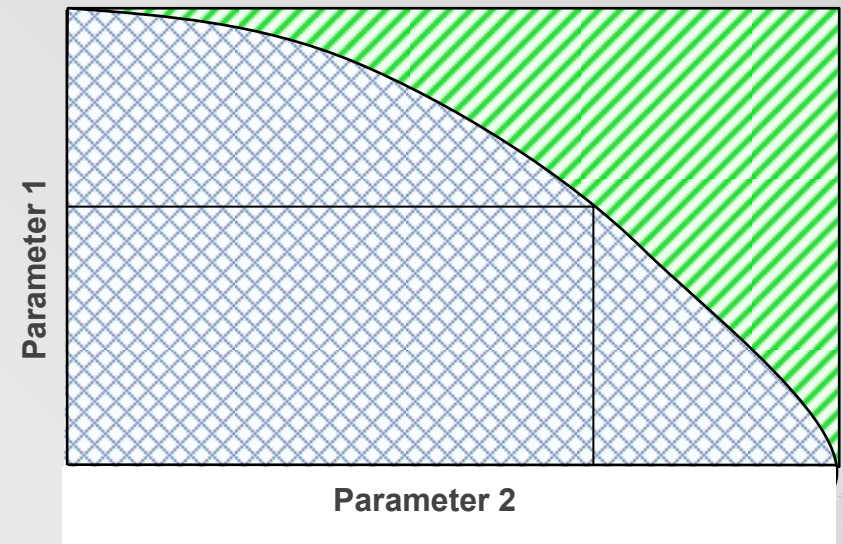
## ■ Tornado Diagram

- Visualizes result of varying a set of parameter through specified ranges on an output of interest



## ■ Strategy Region Graph

- Visualizes changes in optimal strategy as 2 parameters are varied through a range

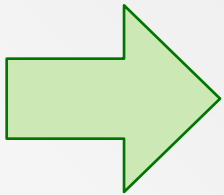


# Things to Remember

- Attributes must be independent for additive model
- Consider the Range of Variation of the attributes when eliciting weights
- Show bottom row (effective) weights to decision maker to ensure they reflect his or her preferences
- Examine results to understand why model makes the recommendations it does
- Do sensitivity analysis!

# Agenda

- ❖ The Decision Analysis Process
- ❖ Deciding with Values
- ❖ Qualitative and Quantitative Assessment
- ❖ Modeling Uncertainty
- ❖ Risk Analysis



# Dealing with Uncertainty

- Value Models: Multi-attribute Utility
- Decision Trees
  - A structured representation for options and outcomes
  - A computational architecture for solving for expected utility
  - Best with “asymmetric” problems (different actions lead to qualitatively different worlds)
- Influence Diagrams
  - A structured representation for options, outcomes and values
  - A computational architecture for solving for expected utility
  - Best with “symmetric” problems (different actions lead to worlds with qualitatively similar structure)
- Decision analysis software:
  - [INFORMS software review page \(http://www.informs.org/Apply-Operations-Research-and-Analytics/Software-Reviews\)](http://www.informs.org/Apply-Operations-Research-and-Analytics/Software-Reviews)

# Example: Patient Treatment

*A patient is suspected of having a disease. Treated patients recover quickly from the illness, but the treatment has unpleasant side effects. Untreated patients suffer a long and difficult illness but eventually recover.*

## Goals:

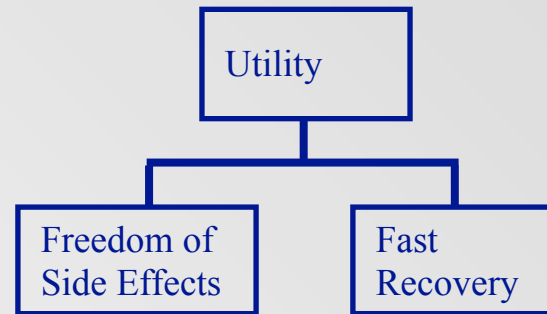
- Freedom from side effects
- Recovery

## Options:

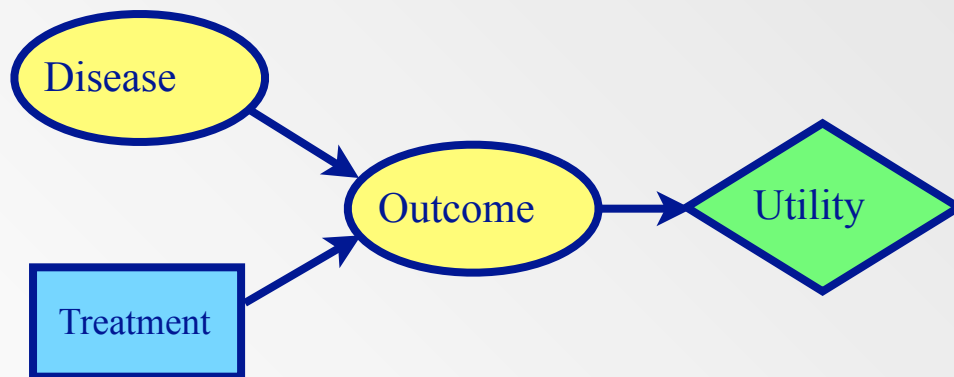
- Treat
- Don't treat

## Outcomes:

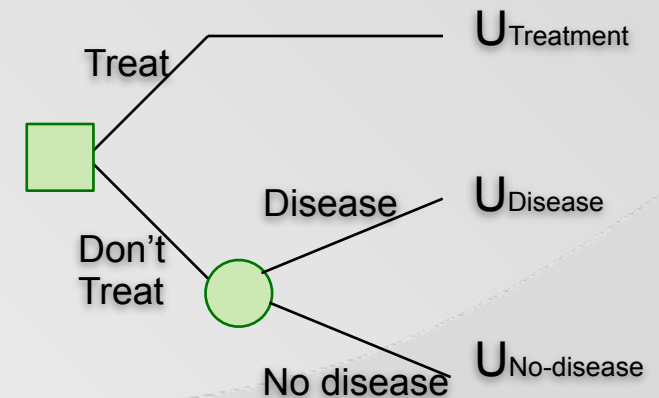
- Sick/Well
- Side Effects / No Side Effects



Multi-Attribute Hierarchy



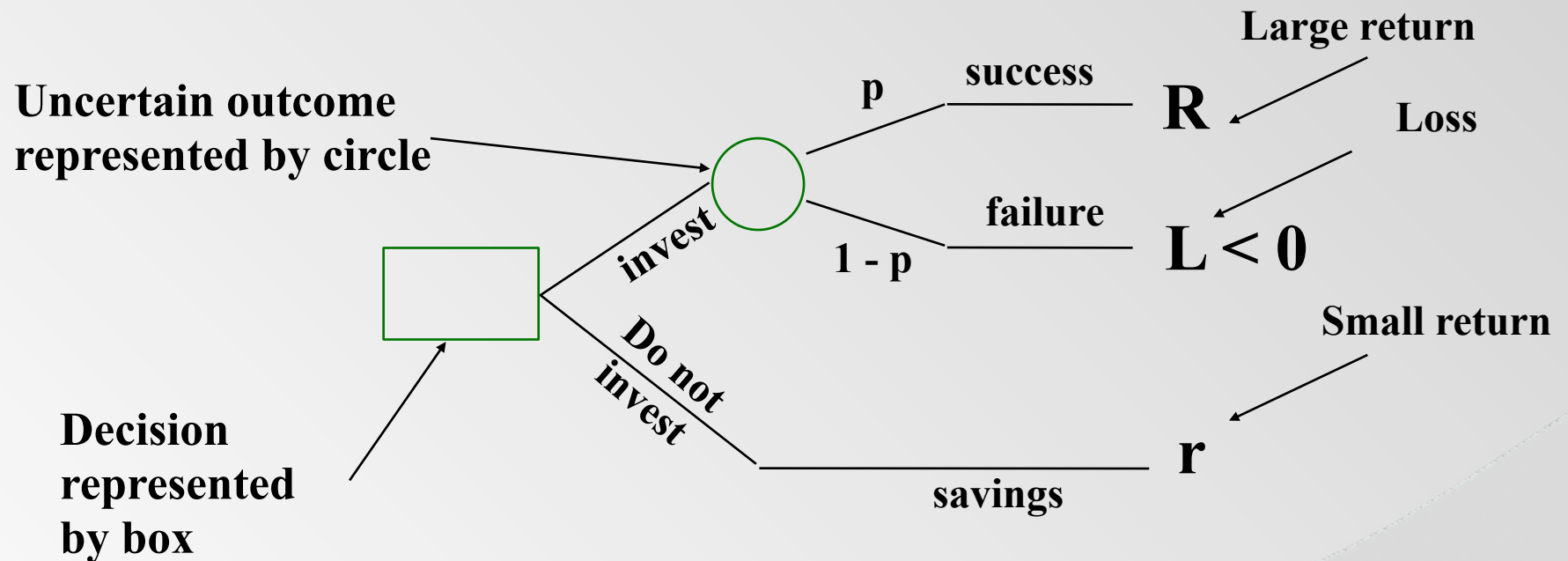
Influence Diagram



Decision Tree

# Decision Trees

- Probabilities replace the weights
  - Account for uncertainty
  - Used to evaluate expected values
- Example - Venture Capital Problem





# Decision Trees (continued)

## ■ Expected return on investment:

- If investment is made  $E(I) = pR + (1 - p)L$
- If investment not made  $E(N) = r$

## ■ Decision:

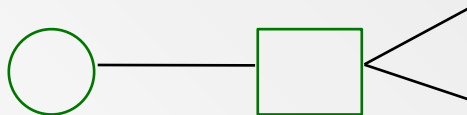
- Invest if  $pR + (1 - p)L > r$
- Don't invest if  $r > pR + (1 - p)L$

What would you do if  
 $r = pR + (1 - p)L$ ?

## ■ Decision Trees evaluated left to right



**decision must be made before uncertain event takes place**



**decision is conditional on the known outcome of the uncertain event**

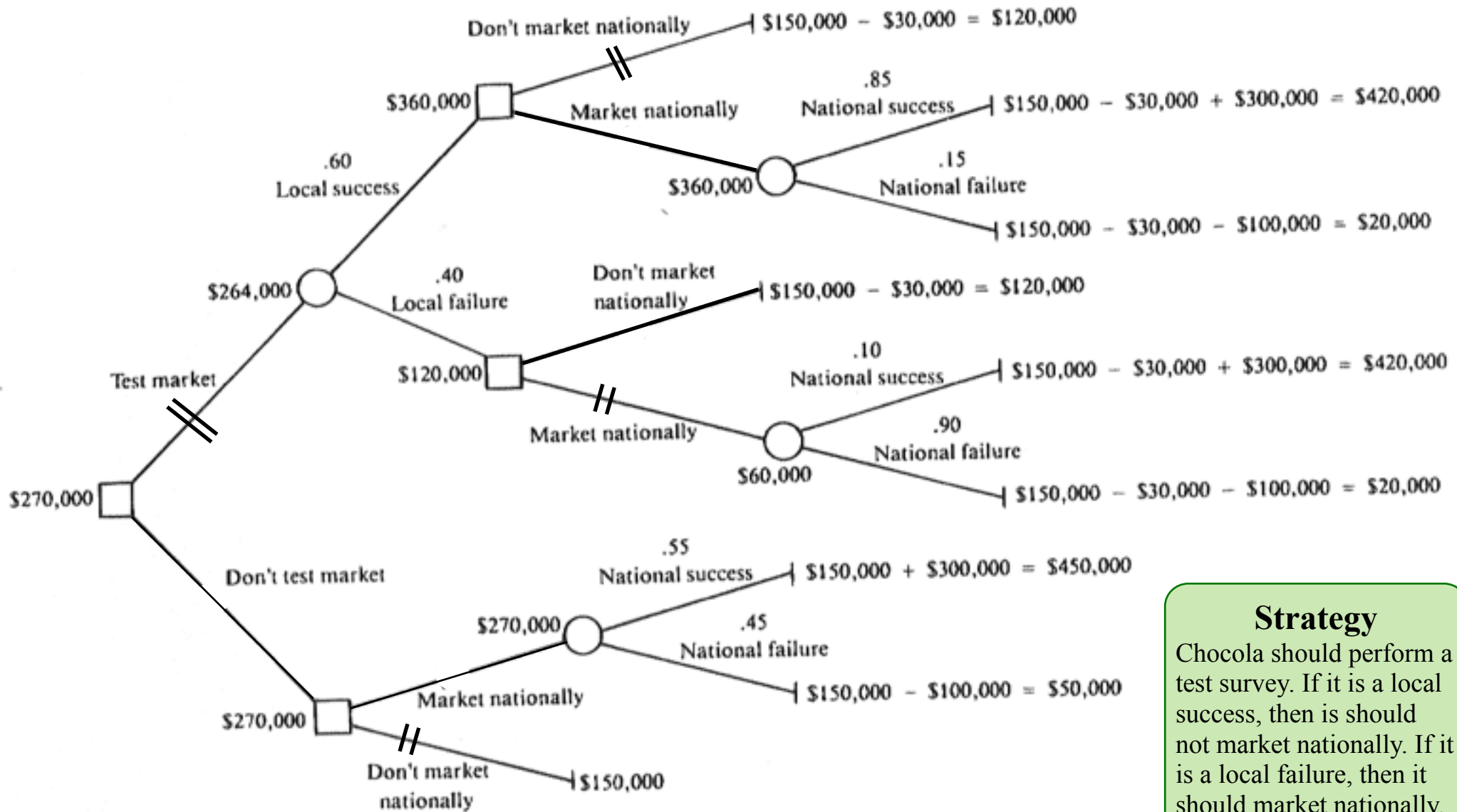
# Chocola Decision Tree

- Colaco currently have \$150,000 and wants to decide whether to market a new chocolate-flavored soda, Chocola.
- It believes that there is a chance of 55% of Chocola being a national success.
  - If it is, then Colaco's asset position will increase by \$300,000.
  - If Chocola is a national failure, Colaco's asset position will decrease by \$100,000.
- What should the company do?

# Chocola Tree: Value of Information

- Before deciding whether to market Chocola, Colaco can perform a market survey at a cost of \$30,000.
- There is a 60% chance that the study will yield favorable results (i.e. a local success), and 40% otherwise (i.e. a local failure).
  - If a local success is observed, there is an 85% that Chocola will be a national success.
  - If a local failure is observed, there is a 90% chance that Chocola will be a national failure.
- Should the company convey the market survey?
- What is the value of the information?

# The Chocolate Decision Tree



**Strategy**  
 Chocolate should perform a test survey. If it is a local success, then it should not market nationally. If it is a local failure, then it should market nationally.

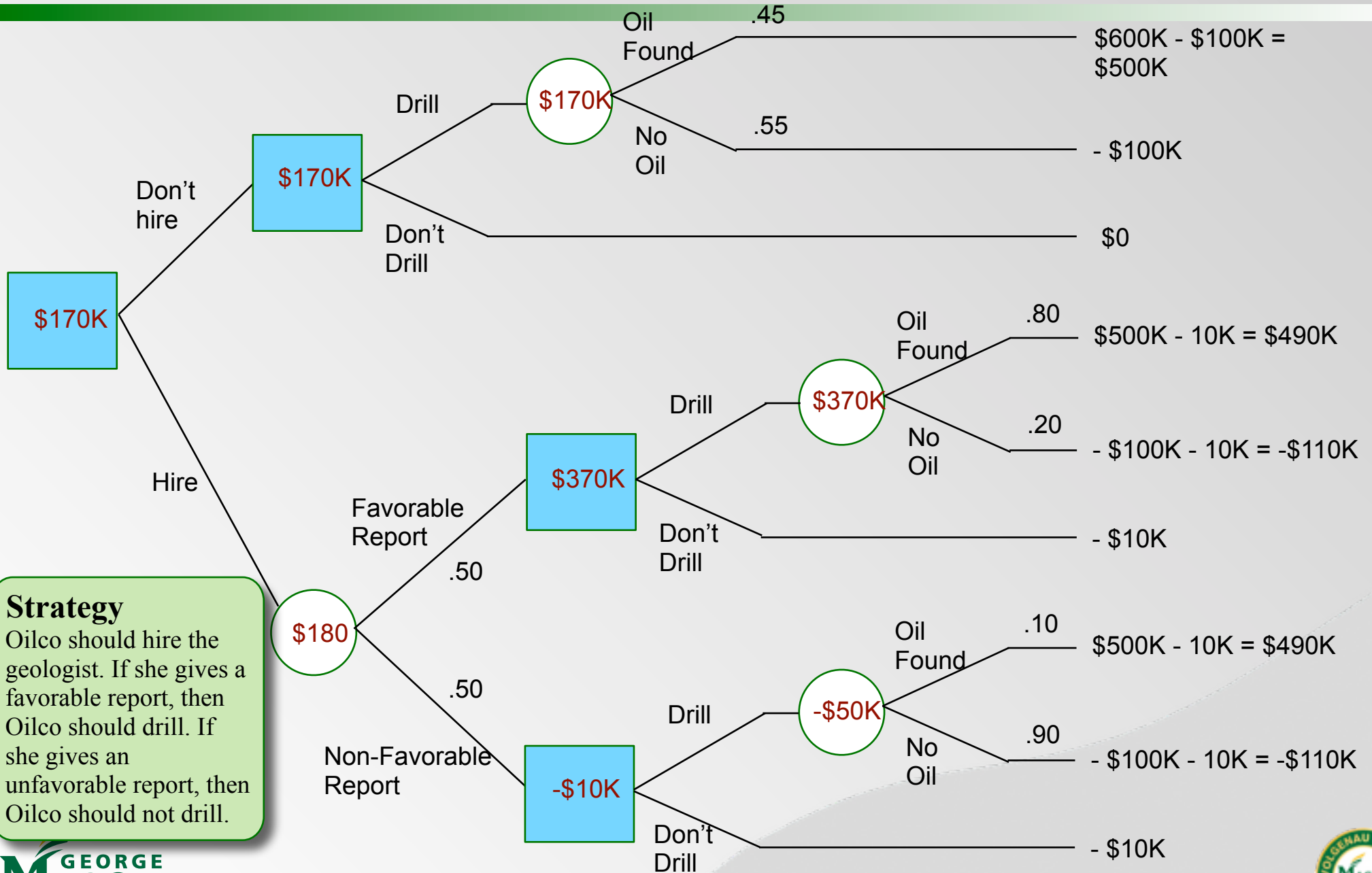
# How Much the Test Should Cost?

- If no test is done, then the Expected Value for the decision tree is 270,000.
  - Let's call it EVWOI, for Expected Value With Original Information.
- If Colaco does a test, then the Expected Value for the decision tree becomes 294,000.
  - Let's call it EVWSI, for Expected Value With Sample Information.
- The difference is the Expected Value of Sample Information (EVSI), which is the difference between the EVWOI and the EVWSI.
  - Since it is \$24,000 and the test costs \$30,000, then Colaco should not perform the test.

# Class Exercise: Oilco DT

- Oilco must determine whether or not to drill in the South China Sea.
  - It costs \$100,000 and if oil is found the value is estimated to be \$600,000.
  - At present, Oilco believes there is a 45% chance that the field contains oil.
- Before drilling, Oilco can hire (for \$10,000) a geologist to obtain more information about the likelihood that the field contains oil.
- There is a 50% chance that the geologist will issue a favorable report and a 50% chance of an unfavorable report.
  - Given a favorable report, there is an 80% chance that the field contains oil.
  - Given an unfavorable report, there is a 10% chance that the field contains oil.
- Determine Oilco's optimal course of action.
- What is the EVSI?

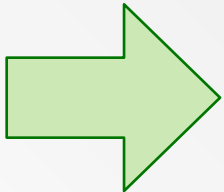
# Oilco Decision Tree



**Strategy**  
 Oilco should hire the geologist. If she gives a favorable report, then Oilco should drill. If she gives an unfavorable report, then Oilco should not drill.

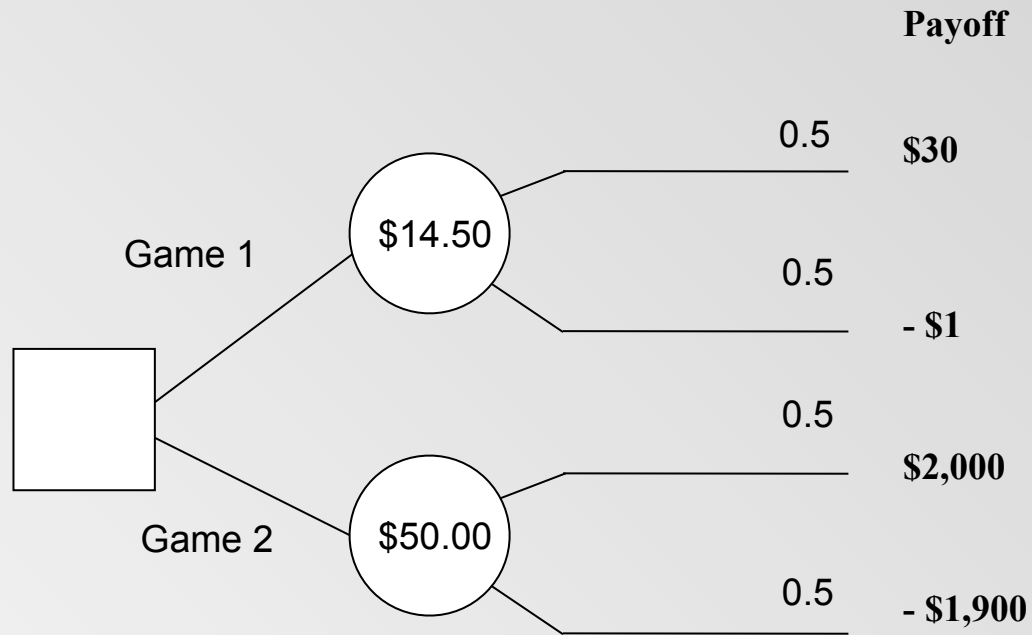
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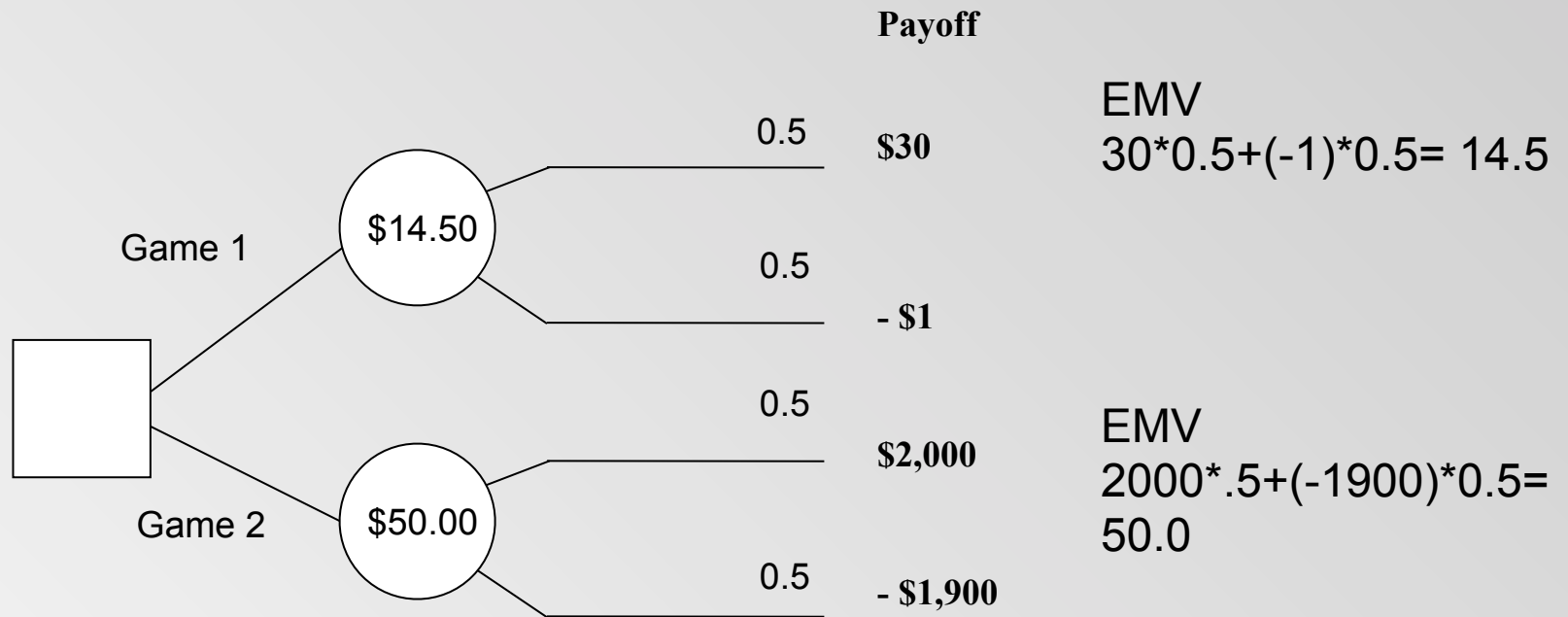


# Risk - Introduction



Which game will you play?  
Which game is risky?

# Risk - Introduction



Which game will you play?  
Which game is risky?

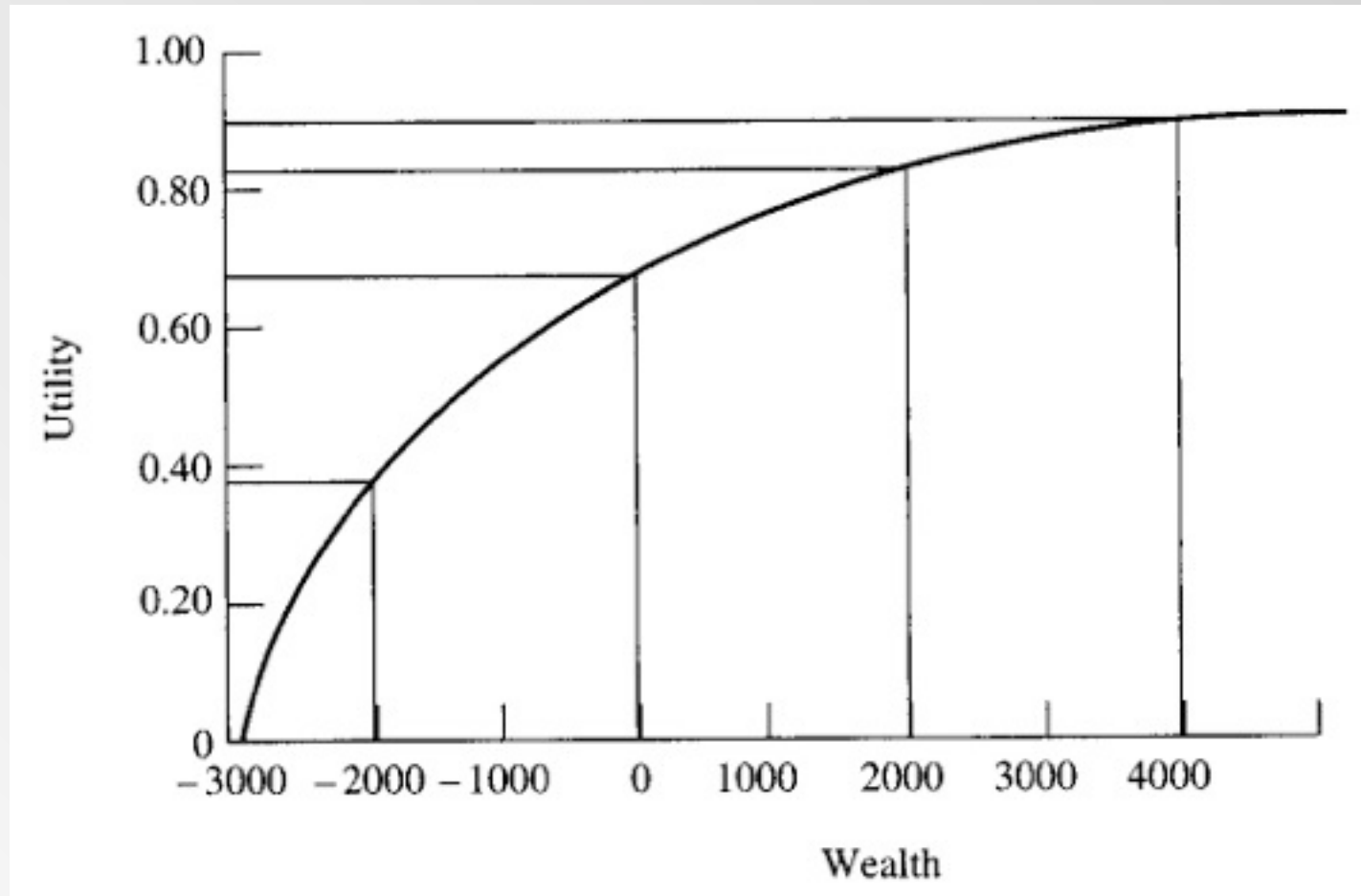
Going by expected monetary value (EMV) or the additive value function Game 2 has Higher EMV but also higher risk

**CONCLUSION: EMV alone is not enough for decision making. Risk is very important too**

# What is an Utility function?

- A way to translate dollars into “utility units”
- It should help choose between alternatives by maximizing the expected utility
- Typical shapes of utility function include log, and exponential

# Risk-Averse Utility Function

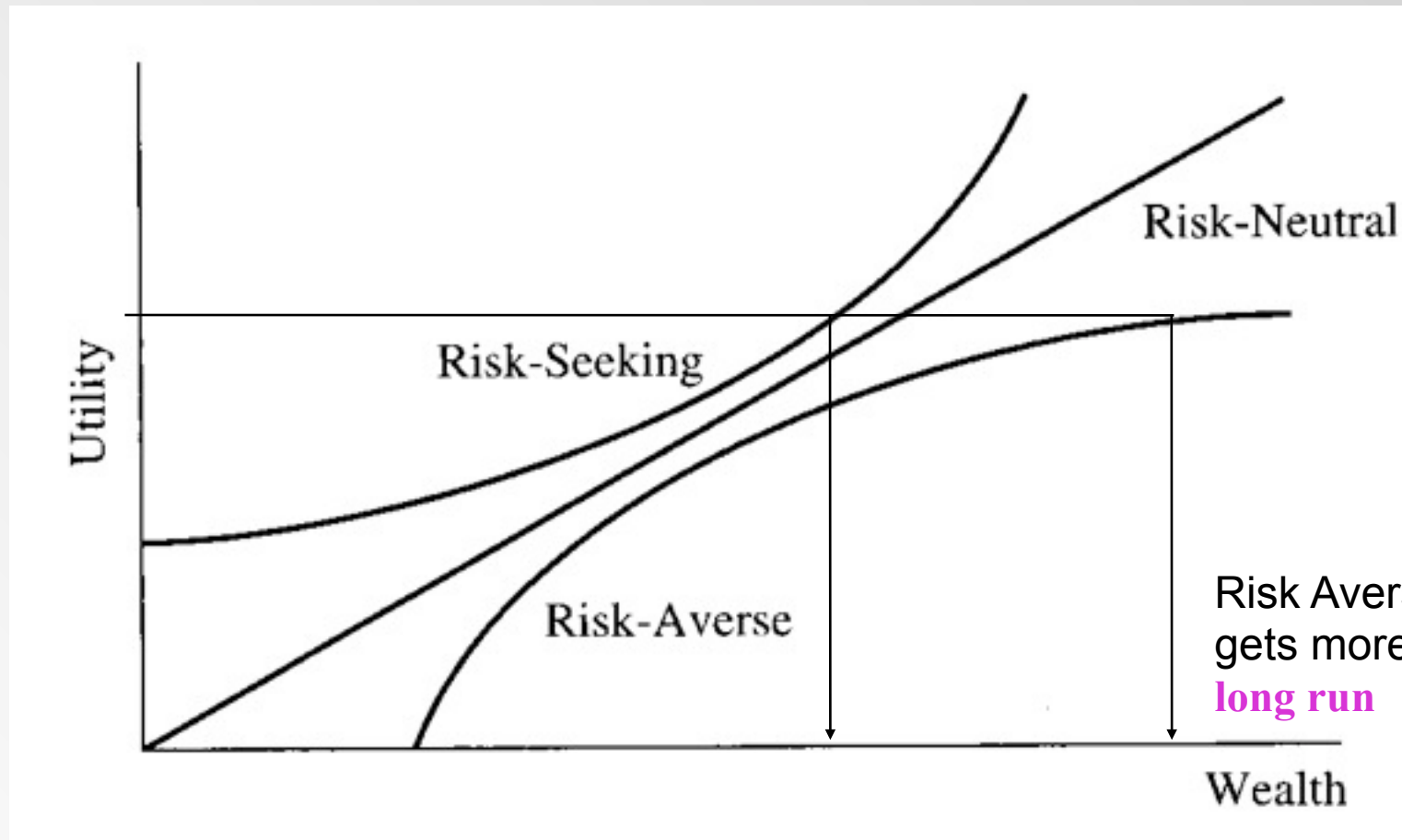


Note the Concave curve - this denotes Risk Averse - typical for most people

# Risk averse person

- Imagine that you are gambling and you hit this situation:
  - Win \$500 with prob 0.5 or lose \$500 with prob 0.5
- A risk-seeking person will play the game but a risk averse person will try to trade in the gamble (try to leave the game) for a small penalty (example: pay \$100 and quit).
- The EMV of the game is \$0 and a risk averse person will trade in the gamble for an amount that is always less than the EMV value.
  - In this case  $-\$100 < \$0$

# Different Risk Attitudes

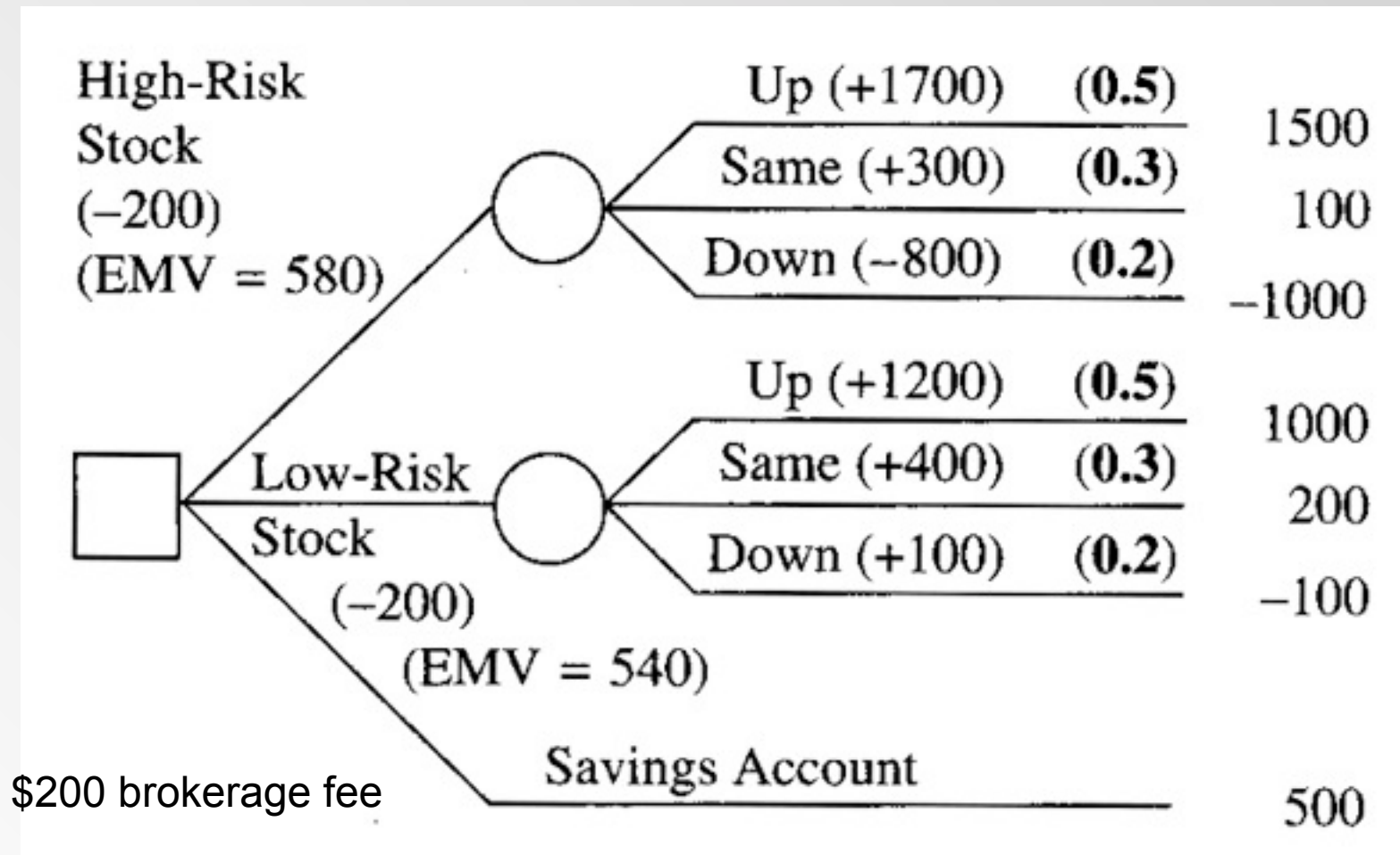


Ignores risk  
Uses EMV

Risk Averse person  
gets more wealth **in the long run**

## Different Risk Attitudes

# Investing in the Stock Market



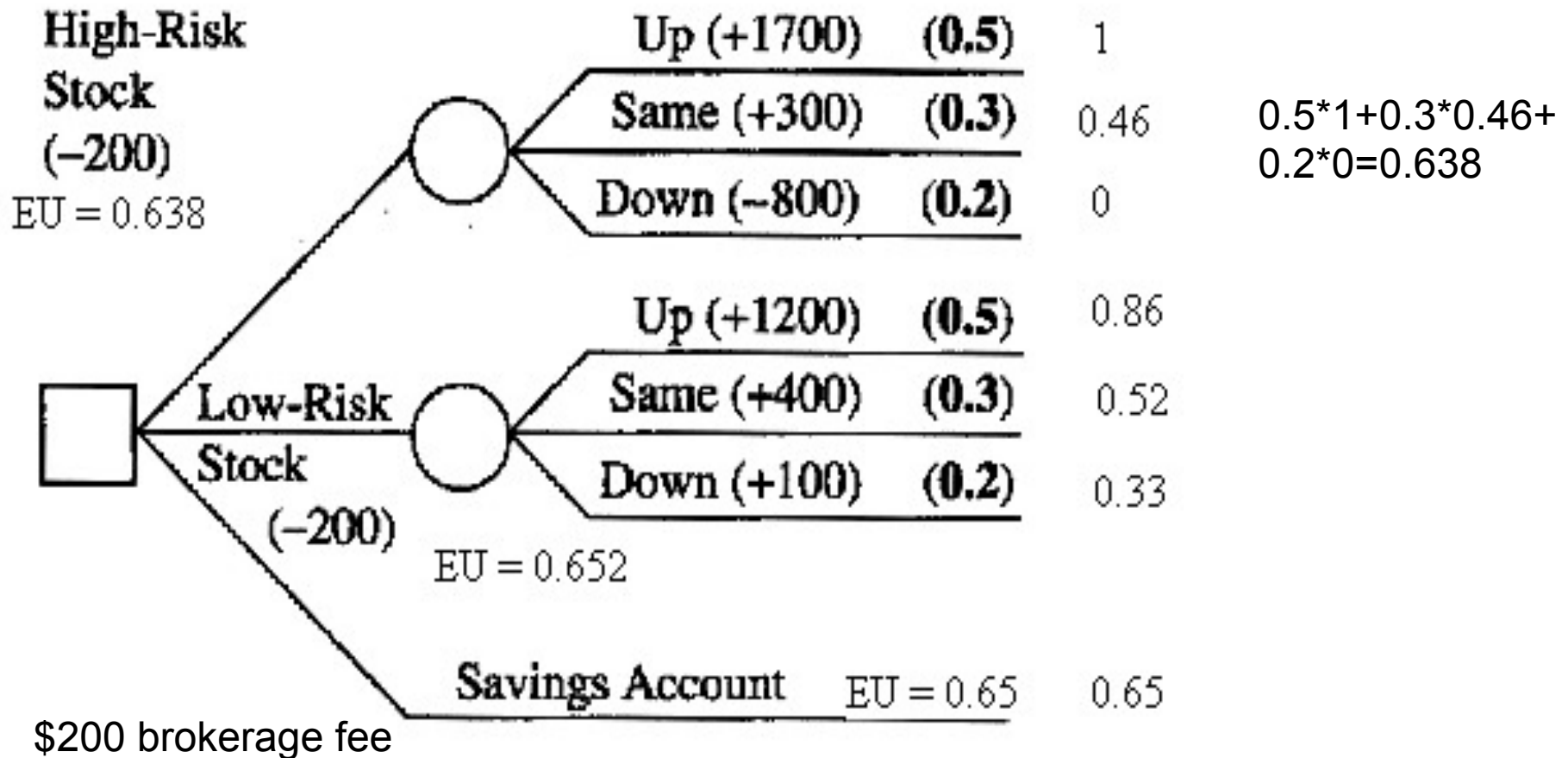
Solution to the decision tree is to invest in high-risk stock. Here risk is not incorporated

# Utility function for investment

Dollar Value	Utility Value
1500	1.00
1000	0.86
500	0.65
200	0.52
100	0.46
-100	0.33
-1000	0.00



# Investing in the Stock Market



Solution to the decision tree is to invest in low-risk stock. Here risk is incorporated via utility function

# Certainty Equivalents and Utility

- Suppose you face the following gamble:
  - Win \$2000 with probability .5
  - Lose \$20 with probability .5
- The expected monetary value for this game is:
  - $\$2000 * .5 + (-\$20) * .5 = \$990$
- After some discussion, you decide to sell your position to an interested friend for \$300.
  - That is, you “gave up” \$690 ( $\$990 - \$300$ ) just to avoid the risk involved in the gambling

# In Other Words...

- A Certainty Equivalent is the amount of money you think is equal to a situation that involves risk.
- The Expected Monetary Value - EMV - is the expected value (in dollars) of the risky proposition
- A Risk Premium is defined as:
  - Risk Premium = EMV – Certainty Equivalent
- The Expected Utility (EU) of a risky proposition is equal to the expected value of the risks in terms of utilities, and  $EU(\text{Risk}) = \text{Utility}(\text{Certainty Equivalent})$

# Some Interesting Facts...

- For risk-averse individuals (risk premium is positive), the horizontal EU line reaches the concave curve before the vertical EMV line.
- For risk-neutral individuals (risk premium is zero), the lines intercept at the linear 45 degree curve.
- For risk-seeking individuals (risk premium is negative), the horizontal EU line reaches the concave curve before the vertical EMV line.

# Agenda

- ❖ The Decision Analysis Process
- ❖ Deciding with Values
- ❖ Qualitative and Quantitative Assessment  
Text
- ❖ Modeling Uncertainty
- ❖ Risk Analysis