Multiattribute Decision Models II

Decision Theory

10/28/2022

ISE 562; Dr. Smith

<u>Today</u>

- MAUT computation example from A-Z
- · Utility function assessment revisited
- Calculating K
- Uncertainty
- · Sensitivity analysis

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• Multiattribute utility functions

Let x, be the level of attribute i; $\mathbb{Q}[x]$ be the 3tribute filting front in of attribute i; $\mathbb{Q}[x]$ be the attribute is a listing of salar for intibute in the are mutually utility independent then the ownerson for the multiattribute utility function takes one of the following forms depending on the sum of the k;: $If \sum_{k=1}^{N} k_n = 1.0, \quad then :\Rightarrow U(\vec{x}) = \begin{cases} \sum_{k=1}^{N} k_k \cdot u(x) \\ \sum_{k=1}^{N} k_k \cdot u(x) \end{cases} Ed$ $U(\vec{x})$ $U(\vec{x})$ where the master scaling constant, K, is a vector much equality. $1 + K = \prod_{n=1}^{N} [1 + K \cdot k_n]$ 10/28/2022

ISE 562; Dr. Smith CtWhat is the relationship between this: $U(\vec{x}) = \frac{1}{K} \left\{ \prod_{n=1}^{N} [1 + K \cdot k_n \cdot u_n(x_n)] - 1 \right\}$ Odent this: Om $U(\vec{x}) = \left\{ \sum_{n=1}^{N} k_n \cdot u_n(x_n) \right\} ?$ Propositionship of the relation of the proposition of the proposi

ISE 562; Dr. Smith It's easier to see if we expand the multiplicative form and rearrange: $U(\bar{x}) = \sum_{i=1}^{N} k_{i}u_{i}(x_{i}) + (\text{interaction terms})$ For example, if N = 2 $U(\bar{x}) = \frac{1}{K} \left\{ \prod_{n=1}^{2} \left[[1 + K \cdot k_{n} \cdot u_{n}(x_{n})] - 1 \right] = \frac{1}{K} \left\{ [1 + Kk_{1}u_{1}(x_{1})][1 + Kk_{2}u_{2}(x_{2})] - 1 \right\}$ expanding $= \frac{1}{K} \left\{ 1 + Kk_{1}u_{1}(x_{1}) + Kk_{2}u_{2}(x_{2}) + K^{2}k_{1}k_{2}u_{1}(x_{1})u_{2}(x_{2}) - 1 \right\}$ $= \frac{1}{K} \left\{ Kk_{1}u_{1}(x_{1}) + Kk_{2}u_{2}(x_{2}) + K^{2}k_{1}k_{2}u_{1}(x_{1})u_{2}(x_{2}) \right\}$ One's cancel $= \sum_{i=1}^{2} k_{i}u_{i}(x_{i}) + \left[Kk_{1}k_{2}u_{1}(x_{1})u_{2}(x_{2}) \right]$ Rearranging $\frac{1}{10/28/2022}$ Interaction terms. Now solve for K

ISE 562; Dr. Smith Solving the master scaling constant equation for K with N=2 attributes: $1+K=\prod_{n=1}^N\left[1+K\cdot k_n\right]$ for~N=2: $1+K=[1+Kk_1][1+Kk_2]$ $1+K=1+Kk_1+Kk_2+K^2k_1k_2$ $K=\frac{1-(k_1+k_2)}{k_1k_2}$ (Note that if k_1+k_2=1, K=0)

Returning to the multiplicative model:

$$= \sum_{i=1}^{2} k_{i} u_{i}(x_{i}) + \left[K k_{1} k_{2} u_{1}(x_{1}) u_{2}(x_{2}) \right]$$

Note that if
$$\sum_{n=1}^{2} k_n = 1$$
,

then
$$K = \frac{1 - k_1 - k_2}{k_1 k_2} = 0$$

So
$$\sum_{i=1}^{2} k_i u_i(x_i) + [Kk_1 k_2 u_1(x_1) u_2(x_2)] = \sum_{i=1}^{2} k_i u_i(x_i) + [0]$$

the multiplicative reduces to the additive model!

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Example

System	Data	(\$	and	mpg)	

Alternative	Cost	mpg
Car A	20	16
Car B	30	25
Car C	40	32

- After collection of system data we extract the ranges:
 - Cost: 20 to 40 k\$ (less is better)
 - Mpg: 16-32 mpg (more is better)

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We need to assist the utility the interest and scaling constants for these attributes

- Cost: 20 to 40 k\$ (less is better)
 Mpg: 16-32 mpg (more steep) S://powcoder Utility function assessment (2)
- To do this we construct an interview package to meet with the DM

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Interview package outline for assessing ledis on maker value model

Introduction

- Independence check
- Attribute ranking

Attribute tradeoff scaling constants Add WeChat powder representations and the constant of the cons

Reassessment as required

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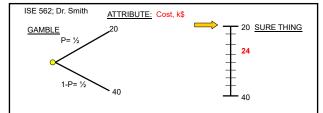
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Introduction

- Why are we here
- Purpose of study
- **Process**

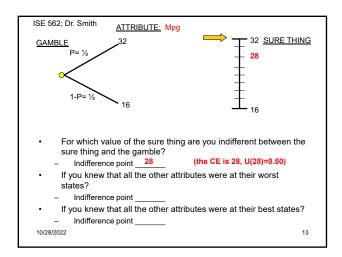
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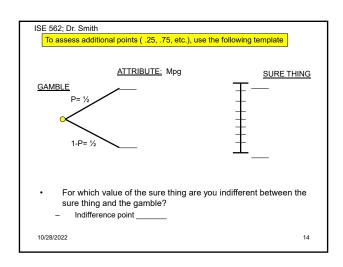
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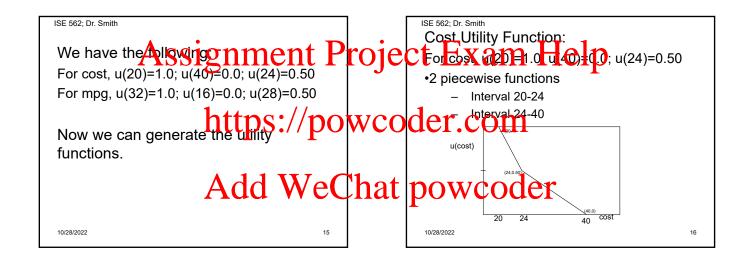


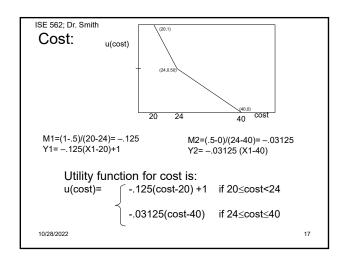
- For which value of the sure thing are you indifferent between the sure thing and the gamble?
- (the CE is 24, U(24)=0.50) Indifference point ____24
- If you knew that all the other attributes were at their worst states?
- Indifference point
- If you knew that all the other attributes were at their best states? Indifference point
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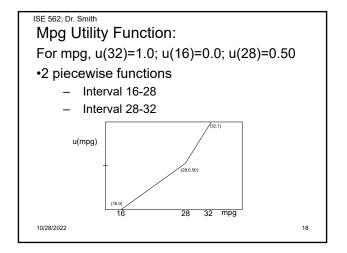
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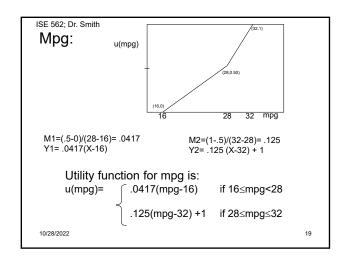


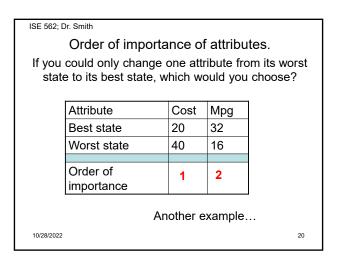


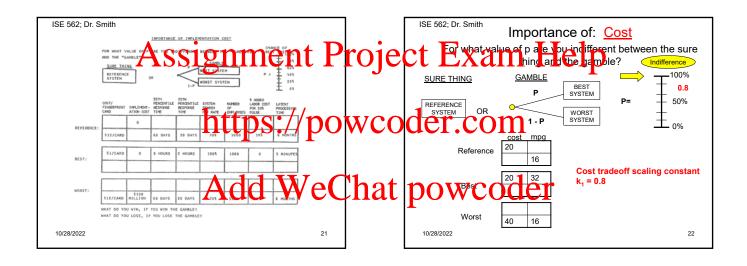


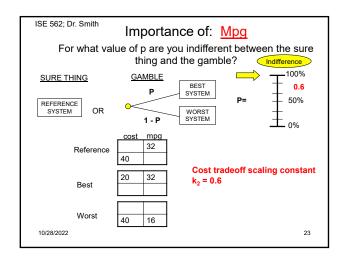


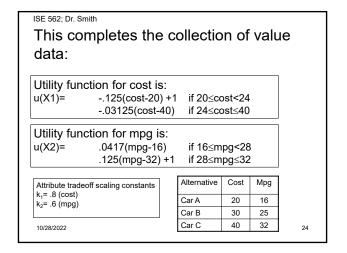












We check which MAU model applies:

 Σk_i = 1.4 \neq 1, so use multiplicative model

$$U(\vec{x}) = \frac{1}{K} \left\{ \prod_{n=1}^{N} \left[1 + K \cdot k_n \cdot u_n(x_n) \right] - 1 \right\}$$

where the master scaling constant, K,

is solved from the equation: $1 + K = \prod_{n=1}^{N} [1 + K] k_n$

First, we need the master scaling constant, K

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Solving for K, the master scaling constant:

we have
$$1 + K = \prod_{n=1}^{N} [1 + K \cdot k_n]$$

or, for this example, n = 2

$$1 + K = [1 + Kk_1][1 + Kk_2]$$

The equation to solve for K:

$$1 + K = [1 + .8K][1 + .6K]$$

$$1 + K = 1 + 1.4K + .48K^2$$

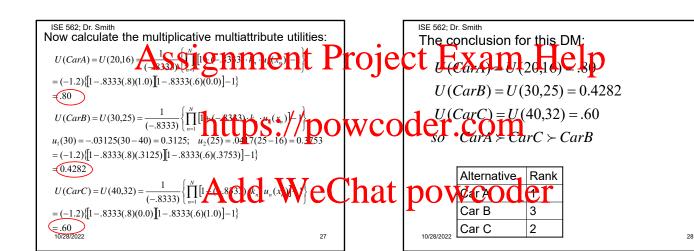
$$.48K^2 + .4K = 0$$

$$K(.48K + .4) = 0$$

K = -.8333

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- •What if the sum of the scaling constants were = 1.0?
 - Use the <u>additive</u> multiattribute utility function
 - For example, suppose we had obtained k₁=.7 and k₂=.3...
 - Note: no master scaling constant

$$1+K = [1+.7K][1+.3K]$$
$$1+K = 1+K+.21K^{2}$$

$$.21K^2 = 0$$

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K = 0

ISE 562; Dr. Smith Calculate the additive multiattribute utilities:

$$U(CarA) = U(20,16) = \sum_{n=1}^{\infty} k_n \cdot u_n(x_n) = .7u_1(x_1) + .3u_2(x_2)$$

 $=.7u_1(20) + .3u_2(16) = .7(1) + .3(0)$

=.70

$$U(CarB) = U(30,25) = \sum_{n=1}^{2} k_n \cdot u_n(x_n) = .7u_1(x_1) + .3u_2(x_2)$$

 $u_1(30) = -.03125(30 - 40) = 0.3125;$ $u_2(25) = .0417(25 - 16) = 0.3753$

 $=.7u_1(30) + .3u_2(25) = .7(.3125) + .3(.3753)$

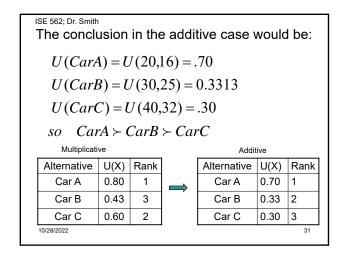
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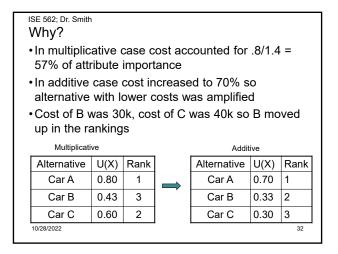
$$U(CarC) = U(40,32) = \sum_{n=1}^{\infty} k_n \cdot u_n(x_n) = .7u_1(x_1) + .3u_2(x_2)$$

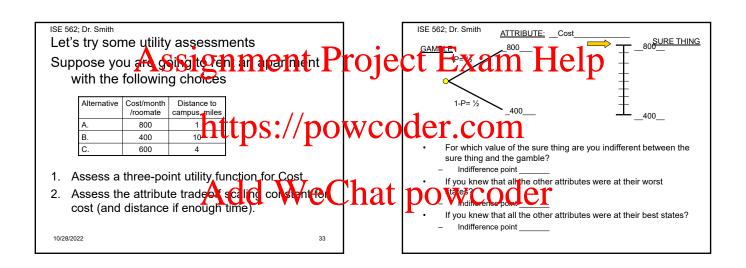
 $=.7u_1(40) + .3u_2(32) = .7(0) + .3(1)$

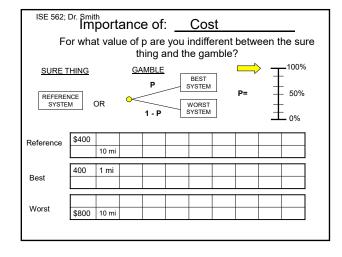
=.30

30









•By the way, we have a problem when n>4

- Can you see it?

- Hint: we need Isaac Newton's help

From earlier example, k₁=.8, k₂=.6; check which MAU model applies: $\Sigma k_i = 1.4 \neq 1$, use multiplicative

$$U(\vec{x}) = \frac{1}{K} \left\{ \prod_{n=1}^{N} \left[1 + K \cdot k_n \cdot u_n(x_n) \right] - 1 \right\}$$

where the master scaling constant, K,

is solved from the equation : $1 + (K) = \prod_{n=1}^{\infty} [1 + (K)k_n]$

First, we need the master scaling constant, K

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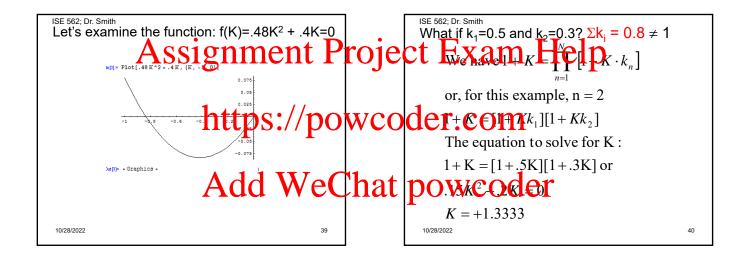
Solving for K, the master scaling constant: We have $1 + K = \prod_{n=1}^{N} [1 + K \cdot k_n]$ or, for this example, n = 2 $1 + K = [1 + Kk_1][1 + Kk_2]$ The equation to solve for K:

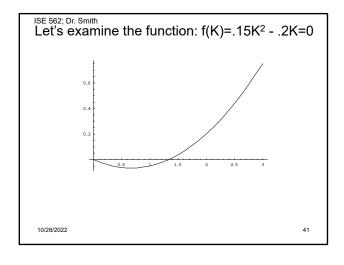
$$.48K^2 + .4K = 0$$

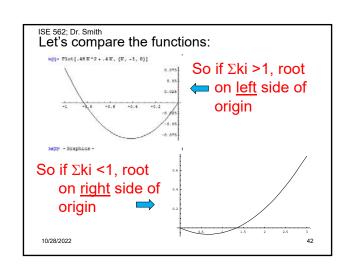
$$K = -.8333$$

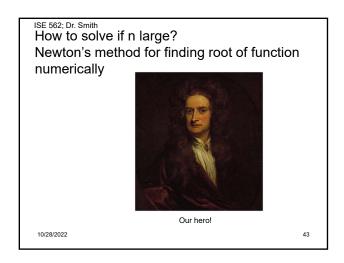
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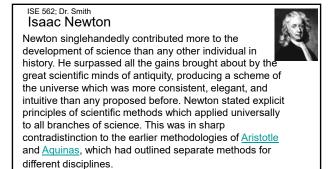
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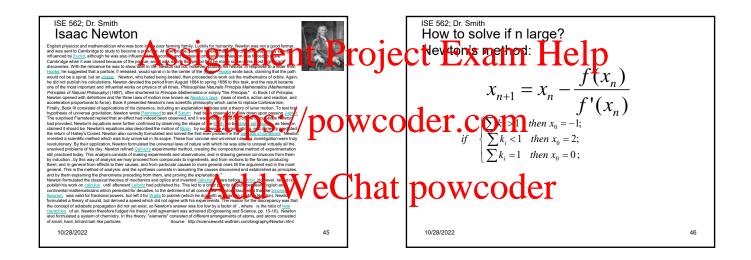


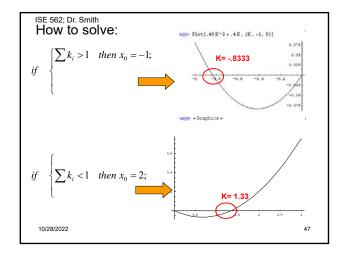


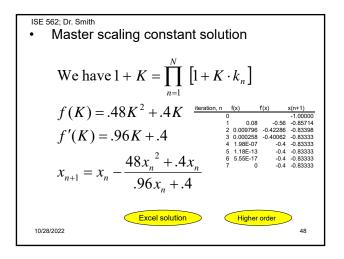


Source: http://scienceworld.wolfram.com/biography/Newton.html

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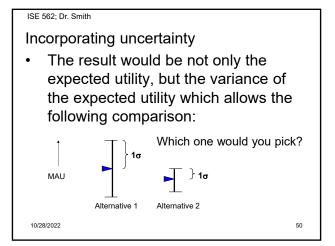
Incorporating uncertainty:

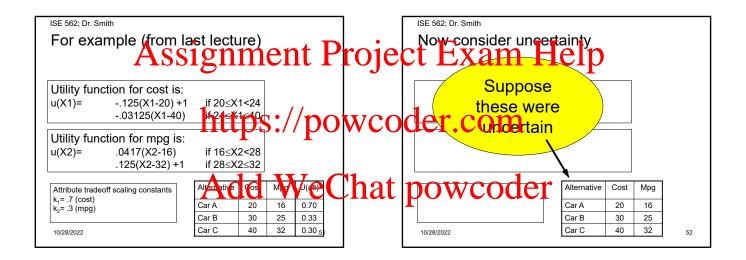
$$U(\vec{x}) = \frac{1}{K} \left\{ \prod_{n=1}^{N} \left[1 + K \cdot k_n \cdot u_n(x_n) \right] - 1 \right\}$$

If the attribute states are represented by probability distributions, we perform a transformation of random variables from the pdf's of the x_n 's to the resulting U(X) using Monte Carlo simulation.

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ISE 562; Dr. Smith For example, for Car B: Alternative Cost Mpg Car A 20 16 25 Car C Suppose cost is a uniform pdf that ranges from 28 to 32 $f(c) = \frac{1}{4} \quad 28 \le c \le 32$ Suppose the pdf of mpg is triangular: $f(c) = \frac{1}{192}m \quad 20 \le m \le 28$ 10/28/2022

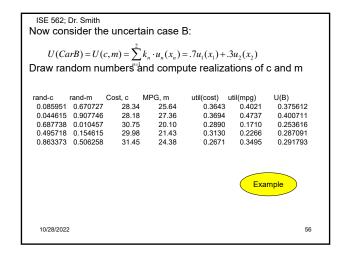
ISE 562; Dr. Smith To simulate cost with Monte Carlo simulation Solve CDF to obtain realization function $f(c) = \frac{1}{4}$ $28 \le c \le 32$ $F(c) = \int_{28}^{c} \frac{1}{4} d\tau = \frac{1}{4} (c - 28)$ $r_i = \frac{1}{4}(c_i - 28)$ $c_i = 4r_i + 28 \quad 0 \le r_i \le 1$ 10/28/2022 54

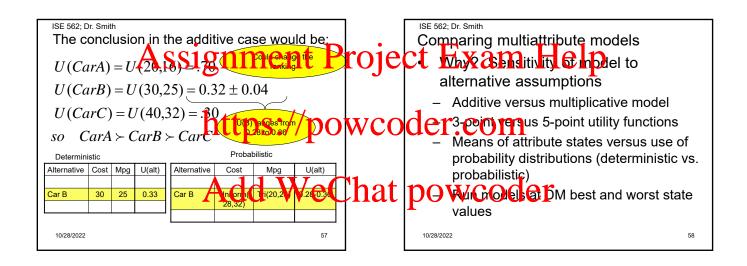
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To simulate mpg with Monte Carlo simulation Solve CDF to obtain realization function

$$f(m) = \frac{1}{192}m \quad 20 \le m \le 28$$

$$F(m) = \int_{20}^{m} \frac{1}{192} \cdot \tau \cdot d\tau = \frac{1}{384}(m^2 - 400)$$
so
$$r_i = \frac{1}{384}(m_i^2 - 400)$$
or
$$m_i = \sqrt{384r_i + 400} \quad 0 \le r_i \le 1$$





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Comparing multiattribute models

- What are we looking for?
 - Consistency in rankings
 - Robustness of conclusions
 - Indicators of reversals (from system data or value data?)
 - Impact of uncertainty
 - Top 3 and bottom 3 alternatives preserved?
 - Primary tradeoff attributes

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Comparing multiattribute models

- · Additive vs. Multiplicative
 - Since $\Sigma k_n \neq 1$ majority of time, compute

$$k'_i = \frac{k_i}{\sum_{i=1}^{N} k_j}$$
 for i = 1,2,..., N

 Then Σk_n = 1 and recompute with additive model:

$$If \sum_{k=1}^{N} k_n = 1.0, \quad then :\Rightarrow U(\vec{x}) = \left\{ \sum_{n=1}^{N} k_n \cdot u_n(x_n) \right\}$$
 Eq 1

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Comparing multiattribute models

- 3-point vs 5-point utility functions
 - Case of precision of value model vs. ranking outcomes
 - If severe risk aversion or risk seeking indicated 3-point may not reveal.
 - Must assess 5-point utility functions during interviews to obtain data. Then use CE for u()=.50 for 3-point case.

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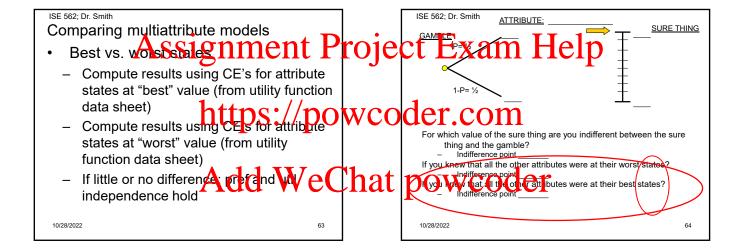
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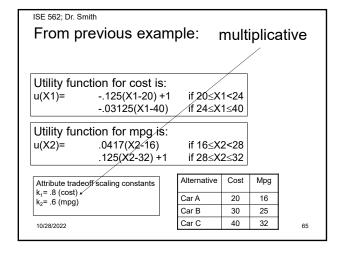
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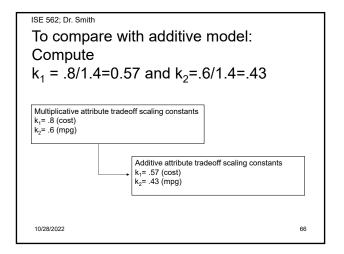
Comparing multiattribute models

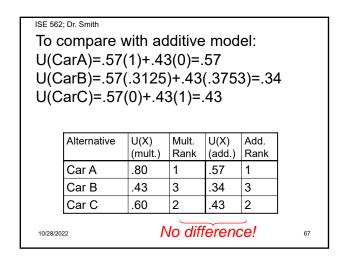
- Deterministic vs probabilistic
 - Compute means of attribute state pdf's and run analysis once
 - Compute rankings with and without uncertainty
 - If dramatically different, technical and preferential risk likely to be high.

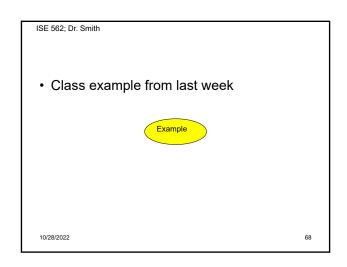
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