

# DECISION MAKING TECHNIQUES IN MANAGEMENT INFORMATION SYSTEMS (MIS)

## LECTURE -7- (Constructing The Decision Model, SAW, WP, TOPSIS)

### CONSTRUCTING DECISION MODEL

- Assessing performance values of alternatives w.r.t. attributes
- (If necessary) Determining relative importance of attributes
- Modeling the preference of DMs

## ASSESSING PERFORMANCE VALUES

- Determine how well each alternative achieves each attribute
- Performance value = score = rating = attribute level

## ASSESSING PERFORMANCE VALUES

- Objective evaluation
  - Natural attribute is used
  - Quantitative
  - Independent of DM
  - Attributes are readily measured in terms of some natural physical unit
    - e.g. dollars or number of people
- Subjective evaluation
  - Constructed attribute is used
  - Qualitative
  - Dependent on DM
  - No natural measuring scales exist
    - e.g. beauty or convenience

## DECISION MATRIX

- Alternative evaluations w.r.t. attributes are presented in a decision matrix
  - Entries are performance values
  - Rows represent alternatives
  - Columns represent attributes

## ATTRIBUTES

- Benefit attributes
  - Offer increasing monotonic utility. Greater the attribute value the more its preference
- Cost attributes
  - Offer decreasing monotonic utility. Greater the attribute value the less its preference
- Nonmonotonic attributes
  - Offer nonmonotonic utility. The maximum utility is located somewhere in the middle of an attribute range

## GLOBAL PERFORMANCE VALUE

- If solution method that will be utilized is performance aggregation oriented, performance values should be aggregated.
- In this case
  - Performance values are normalized to eliminate computational problems caused by differing measurement units in a decision matrix
  - Relative importance of attributes are determined

## NORMALIZATION

- Aims at obtaining comparable scales, which allow interattribute as well as intra-attribute comparisons
- Normalized performance values have dimensionless units
- The larger the normalized value becomes, the more preference it has

## NORMALIZATION METHODS

1. Distance-Based Normalization Methods
2. Proportion Based Normalization Methods (Standardization)

## DISTANCE-BASED NORMALIZATION METHODS

If we define the normalized rating as the ratio between individual and combined distance from the origin (0,0,...,0) then the comparable rating of  $x_{ij}$  is given as (Yoon and Kim, 1989):

$$r_{ij}(p) = (x_{ij} - 0) / \left\{ \sum_{k=1}^m |x_{kj} - 0|^p \right\}^{1/p}$$

*This equation is arranged for benefit attributes.*

*Cost attributes become benefit attributes by taking the inverse rating ( $1/x_{ij}$ )*

## DISTANCE-BASED NORMALIZATION METHODS

- Normalization ( $p=1$ : Manhattan distance)
- Vector Normalization ( $p=2$ : Euclidean distance)
- Linear Normalization ( $p=\infty$ : Tchebycheff dist.)

$$r_{ij}(1) = x_{ij} / \sum_{k=1}^m |x_{kj}|$$

$$r_{ij}(2) = x_{ij} / \sqrt{\sum_{k=1}^m |x_{kj}|^2}$$

$$r_{ij}(\infty) = x_{ij} / \max\{|x_{kj}|, k = 1, 2, \dots, m\} \quad (\text{BENEFIT ATTRIBUTE})$$

$$r_{ij}(\infty) = \min\{|x_{kj}|, k = 1, 2, \dots, m\} / x_{ij} \quad (\text{COST ATTRIBUTE})$$

## EXAMPLE

Variable	Value
x1	35
x2	15
x3	25
x4	30
x5	20

Obtain the normalized values by using:

- a. Manhattan distance
- b. Euclidian distance
- c. Tchebycheff distance

### EXAMPLE (NORMALIZATION-MANHATTAN DISTANCE)

Variable	Value	Manhattan
x1	35	0,28
x2	15	0,12
x3	25	0,2
x4	30	0,24
x5	20	0,16
Total	125	

### EXAMPLE (VECTOR NORMALIZATION-EUCLIDIAN DISTANCE)

Variable	Value	x2	Euclidian
x1	35	1225	0,60
x2	15	225	0,26
x3	25	625	0,43
x4	30	900	0,52
x5	20	400	0,34
Total	125	3375	

## EXAMPLE (LINEAR NORMALIZATION-TCHEBYCHEFF DISTANCE)

Variable	Value	Linear
x1	35	1,00
x2	15	0,43
x3	25	0,71
x4	30	0,86
x5	20	0,57
Total	<b>125</b>	

## PROPORTION-BASED NORMALIZATION METHODS

The proportion of difference between performance value of the alternative and the worst performance value to difference between the best and the worst performance values  
(Bana E Costa, 1988; Kirkwood, 1997)

$$r_{ij} = (x_{ij} - x_j^-) / (x_j^* - x_j^-) \text{ benefit attribute}$$

$$r_{ij} = (x_j^- - x_{ij}) / (x_j^- - x_j^*) \text{ cost attribute}$$

where \* represents the best and – represents the worst  
(best: max. perf. value for benefit; min. perf. value for cost or ideal value determined by DM for that attribute)



## EXAMPLE

Variable	Value
x1	35
x2	15
x3	25
x4	30
x5	20

Obtain the normalized values by using proportion based normalization.

## EXAMPLE (PROPORTION BASED NORMALIZATION)

Variable	Value	Proportion
x1	35	1
x2	15	0
x3	25	0,5
x4	30	0,75
x5	20	0,25
Total	<b>125</b>	

## TRANSFORMATION OF NONMONOTONIC ATTRIBUTES TO MONOTONIC

- exp( $-z^2/2$ ) exponential function is utilized for transformation

where  $z = (x_{ij} - x_j^0) / \sigma_j$

$x_j^0$  is the most favorable performance value w.r.t. attribute  $j$

$\sigma_j$  is the standard deviation of performance values w.r.t. attribute  $j$

### EXAMPLE

- The temperature values for different locations are provided below. If the best condition is 25<sup>0</sup> celcius. Normalize the related values.

Location	Temperature
x1	35
x2	15
x3	25
x4	30
x5	20

## EXAMPLE

Variable	Value	z	$\exp(-z^2/2)$
x1	35	1,26	0,45
x2	15	-1,26	0,45
x3	25	0,00	1,00
x4	30	0,63	0,82
x5	20	-0,63	0,82
std	7,905694		

## ATTRIBUTE WEIGHTING

- Most methods translate the relative importance of attributes into numbers which are often called as “weights” (Vincke, 1992)
- Methods utilized for assignment of weights can be classified in two groups (Huylenbroeck, 1995; Munda 1993; Al-Kloub *et al.*, 1997; Kleindorfer *et al.*, 1993; Yoon and Hwang, 1995):
  - Direct Determination
  - Indirect Determination

## WEIGHT ASSIGNMENT METHODS

- Direct Determination
  - Rating, Point allocation, Categorization
  - Ranking
  - Swing
  - Ratio (Eigenvector prioritization)
  - ...
- Indirect Determination
  - Centrality
  - Regression – Conjoint analysis
  - Interactive

## RATING

- Rating
  - Each attribute's importance is rated on a scale (e.g. 0-100)
- Point allocation
  - Allocate a specific amount of points (e.g. 100) among attributes in proportion of their importance
- Categorization
  - Assign attributes to different categories of importance, each carrying a different weight.

## RANKING

- We assign 1 to most important attribute, and n to the least important. The cardinal weights can be obtained from one of the following formulas:
  - Rank reciprocal weights
 
$$w_j = (1/r_j) / \left( \sum_{k=1}^n 1/r_k \right)$$
  - Rank sum weights
 
$$w_j = (n - r_j + 1) / \left( \sum_{k=1}^n (n - r_k + 1) \right)$$

*where  $r_j$  is the rank of the  $j$ th attribute*

If two attributes are tied for the  $n$ th and  $(n+1)$ th place, the number  $(2n+1)/2$  is assigned to both of them

## SWING

- DM considers a hypothetical alternative where attributes are all at their worst value.
- S/he is then asked which of the attributes would most prefer to swing from its worst value up to its best value
- S/he is then asked which attribute s/he would swing up second and so on...
- After ranking the attributes in this manner, DM is asked to give the most important attribute a weight of 100, and then assign weights to the other attributes in proportion to the importance of their ranges.

## DECISION MATRIX

- Alternative evaluations w.r.t. attributes are presented in a decision matrix
  - Entries are performance values
  - Rows represent alternatives
  - Columns represent attributes

## SAW

- Simple Additive Weighting – Weighted Average – Weighted Sum (Yoon & Hwang, 1995; Vincke, 1992...)
- A global (total) score in the SAW is obtained by adding contributions from each attribute.
- A common numerical scaling system such as normalization (instead of single dimensional value functions) is required to permit addition among attribute values.
- Value (global score) of an alternative can be expressed as:

$$V(a_i) = V_i = \sum_{j=1}^n w_j r_{ij}$$

## EXAMPLE FOR SAW

Normalized (Linear) Decision Matrix and Global Scores

	Price	Comfort	Perf.	Design	$V_i$
<i>Norm. w</i>	0.3333	0.2667	0.2	0.2	
<b><math>a_1</math></b>	0.3333	1	1	1	<b>.7778</b>
$a_2$	0.4	1	0.6667	1	.7334
$a_3$	0.4	0.6667	1	1	.7111
$a_4$	0.5	0.6667	1	0.6667	.6778
$a_5$	0.5	0.6667	0.6667	1	.6778
$a_6$	0.5	0.3333	1	1	.6555
$a_7$	1	0.3333	0.6667	0.6667	.6889

## EXAMPLE FOR SAW

$$0.7778 = (0.3333 \cdot 0.3333) + (0.2667 \cdot 1) + (0.2 \cdot 1) + (0.2 \cdot 1)$$

	Price	Comfort	Perf.	Design	$V_i$
<i>Norm. w</i>	0.3333	0.2667	0.2	0.2	
<b><math>a_1</math></b>	0.3333	1	1	1	<b>.7778</b>
$a_2$	0.4	1	0.6667	1	.7334
$a_3$	0.4	0.6667	1	1	.7111
$a_4$	0.5	0.6667	1	0.6667	.6778
$a_5$	0.5	0.6667	0.6667	1	.6778
$a_6$	0.5	0.3333	1	1	.6555
$a_7$	1	0.3333	0.6667	0.6667	.6889

## WP

- Weighted Product (Yoon & Hwang, 1995)
- Normalization is not necessary!
- When WP is used weights become exponents associated with each attribute value;
  - a positive power for benefit attributes
  - a negative power for cost attributes
- Because of the exponent property, this method requires that all ratings be greater than 1. When an attribute has fractional ratings, all ratings in that attribute are multiplied by  $10^m$  to meet this requirement

$$V_i = \prod_j (x_{ij})^{w_j}$$

## EXAMPLE FOR WP

Quantitative Decision Matrix and Global

Scores	Price	Comfort	Perf.	Design	$V_i$
<i>Norm. w</i>	.3333	.2667	.2	.2	
<b><math>a_1</math></b>	300	3	3	3	<b>.3108</b>
$a_2$	250	3	2	3	.3045
$a_3$	250	2	3	3	.2964
$a_4$	200	2	3	2	.2944
$a_5$	200	2	2	3	.2944
$a_6$	200	1	3	3	.2654
$a_7$	100	1	2	2	.2843



## EXAMPLE FOR WP

$$0.3108 = 300^{-0.3333} * 3^{0.2667} * 3^{0.2} * 3^{0.2}$$

	Price	Comfort	Perf.	Design	$V_i$
<i>Norm. w</i>	.3333	.2667	.2	.2	
<b><math>a_1</math></b>	300	3	3	3	<b>.3108</b>
$a_2$	250	3	2	3	.3045
$a_3$	250	2	3	3	.2964
$a_4$	200	2	3	2	.2944
$a_5$	200	2	2	3	.2944
$a_6$	200	1	3	3	.2654
$a_7$	100	1	2	2	.2843

## TOPSIS

- Technique for Order Preference by Similarity to Ideal Solution (Yoon & Hwang, 1995; Hwang & Lin, 1987)
- Concept:
  - Chosen alternative should have the shortest distance from the positive ideal solution and the longest distance from the negative ideal solution
- Steps:
  - Calculate normalized ratings
  - Calculate weighted normalized ratings
  - Identify positive-ideal and negative-ideal solutions
  - Calculate separation measures
  - Calculate similarities to positive-ideal solution
  - Rank preference order

## STEPS

- Calculate normalized ratings
  - Vector normalization (Euclidean) is used
  - Do not take the inverse rating for cost attributes!
- Calculate weighted normalized ratings

- $v_{ij} = w_j * r_{ij}$

- Identify positive-ideal and negative-ideal solutions

$$a^+ = \{v_1^+, v_2^+, \dots, v_j^+, \dots, v_n^+\} = \left\{ \left( \max_i v_{ij} \mid j \in J_1 \right), \left( \min_i v_{ij} \mid j \in J_2 \right) \mid i = 1, \dots, m \right\}$$

$$a^- = \{v_1^-, v_2^-, \dots, v_j^-, \dots, v_n^-\} = \left\{ \left( \min_i v_{ij} \mid j \in J_1 \right), \left( \max_i v_{ij} \mid j \in J_2 \right) \mid i = 1, \dots, m \right\}$$

where  $J_1$  is a set of benefit attributes and  $J_2$  is a set of cost attributes

## STEPS

- Calculate separation measures
  - Euclidean distance (separation) of each alternative from the ideal solutions are measured:

$$S_i^+ = \sqrt{\sum_j (v_{ij} - v_j^+)^2} \quad S_i^- = \sqrt{\sum_j (v_{ij} - v_j^-)^2}$$

- Calculate similarities to positive-ideal solution

$$C_i^+ = S_i^- / (S_i^+ + S_i^-)$$

- Rank preference order
  - Rank the alternatives according to similarities in descending order.
  - Recommend the alternative with the maximum similarity

## EXAMPLE FOR TOPSIS

### ○ Normalized (Vector) Decision Matrix

	Price	Comfort	Perf.	Design
<i>Norm. w</i>	0.3333	0.2667	0.2	0.2
$a_1$	0.5108	0.5303	0.433	0.4121
$a_2$	0.4256	0.5303	0.2887	0.4121
$a_3$	0.4256	0.3536	0.433	0.4121
$a_4$	0.3405	0.3536	0.433	0.2747
$a_5$	0.3405	0.3536	0.2887	0.4121
$a_6$	0.3405	0.1768	0.433	0.4121
$a_7$	0.1703	0.1768	0.2887	0.2747

## WEIGHTED NORMALIZED RATINGS & POSITIVE-NEGATIVE IDEAL

	Price	Comfort	Perf.	Design
$a_1$	0.1703	0.1414	0.0866	0.0824
$a_2$	0.1419	0.1414	0.0577	0.0824
$a_3$	0.1419	0.0943	0.0866	0.0824
$a_4$	0.1135	0.0943	0.0866	0.0549
$a_5$	0.1135	0.0943	0.0577	0.0824
$a_6$	0.1135	0.0471	0.0866	0.0824
$a_7$	0.0568	0.0471	0.0577	0.0549
$a^*$	.0568	.1414	.0866	.0824
$a^-$	.1703	.0471	.0577	.0549

## SEPARATION MEASURES & SIMILARITIES TO POSITIVE IDEAL SOLUTION

	$S^*$	$S^-$	$C^*$	Rank
$a_1$	0.1135	0.1024	0.4742	5
$a_2$	0.0899	0.1022	<b>0.5321</b>	1
$a_3$	0.0973	0.0679	0.4111	6
$a_4$	0.0787	0.0792	0.5016	3
$a_5$	0.0792	0.0787	0.4984	4
$a_6$	0.11	0.0693	0.3866	7
$a_7$	0.1024	0.1135	0.5258	2

## REFERENCES

- Lecture notes of "Prof. Dr. Y. İlker Topçu",  
<http://web.itu.edu.tr/topcuil/>