

Analyzing the Problem (SAW, WP, TOPSIS)

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MADM Methods to be covered

- Elementary Methods
- Value Based Methods
 - Multi Attribute Value Theory
 - Simple Additive Weighting
 - Weighted Product
 - TOPSIS
- The Analytic Hierarchy Process (AHP)
- The Analytic Network Process (ANP)
- Outranking Methods
 - PROMETHEE methods
 - ELECTRE methods

SAW – WP – TOPSIS

- *Ranking methods: Complete preorder*
- *Performance aggregation oriented*
- *Compensatory*

SAW

Simple Additive Weighting

Weighted Average

Weighted Sum

- In the SAW, a global (total) score of an alternative is computed by adding the linear normalized performance values of the alternative w.r.t. each attribute
- Normalization 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	
a_1	0.3333	1	1	1	.7778
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

a_1 is recommended

Sensitivity Analysis

- "Conducting a sensitivity analysis on attribute i "
- When weight of that attribute is changed, the weights of other attributes are also changed:

$$w_j = (1 - w_i) (w_j^0 / (1 - w_i^0)),$$

where w^0 is the base weight (current weight used at the evaluation)

w1	w2	w3	Total
50%	40%	10%	100%
30%	56%	14%	100%
80%	16%	4%	100%

WP

Weighted Product

- In the WP, a global score of an alternative is computed by multiplying the values obtained by taking the exponents (representing the attribute weights) of the performance values of the alternative w.r.t. each attribute
- Normalization is not necessary!
- Weights become exponents;
 - a positive power for benefit attributes
 - a negative power for cost attributes
- Value (global score) of an alternative can be expressed as:

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

Example for WP

Decision Matrix and Global Scores

	Price	Comfort	Perf.	Design	V_i
<i>Norm. w</i>	.3333	.2667	.2	.2	
a_1	300	3	3	3	.3108
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

a_1 is recommended

TOPSIS

Technique for **O**rder of **P**reference by **S**imilarity to **I**deal **S**olution

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 distance) 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

No inverse rating!

Example for TOPSIS

- Weighted Normalized Decision Matrix

	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

Example for TOPSIS

- Positive-Ideal and Negative-Ideal Solutions

	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

Example for TOPSIS

- Separation from Positive-Ideal

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

	Price	Comfort	Perf.	Design		
a_1	0.1703	0.1414	0.0866	0.0824		
a^*	0.0568	0.1414	0.0866	0.0824		
$(a_1 - a^*)^2$	0.0129	0	0	0	SUM	0.0129
					SQRT	0.1135

Example for TOPSIS

- Separation from Negative-Ideal

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

	Price	Comfort	Perf.	Design		
a_1	0.1703	0.1414	0.0866	0.0824		
a^-	0.1703	0.0471	0.0577	0.0549		
$(a_1 - a^-)^2$	0.0000	0.00889	0.00084	0.00076	SUM	0.0105
					SQRT	0.1024

Example for TOPSIS

- Separations from Positive Ideal and Negative Ideal Solutions

	S^*	S^-		
a_1	0.1135	0.1024		
a_2	0.0899	0.1022		
a_3	0.0973	0.0679		
a_4	0.0787	0.0792		
a_5	0.0792	0.0787		
a_6	0.11	0.0693		
a_7	0.1024	0.1135		

Example for TOPSIS

- Similarities to Positive Ideal Solution

	S^*	S^-	C^*	
a_1	0.1135	0.1024	0.4742	
a_2	0.0899	0.1022	0.5321	
a_3	0.0973	0.0679	0.4111	
a_4	0.0787	0.0792	0.5016	
a_5	0.0792	0.0787	0.4984	
a_6	0.11	0.0693	0.3866	
a_7	0.1024	0.1135	0.5258	

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

Example for TOPSIS

- Separations, Similarities, and Ranking

	S^*	S^-	C^*	Rank
a_1	0.1135	0.1024	0.4742	5
a_2	0.0899	0.1022	0.5321	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

a2 is recommended

Additional Example I

Fighter Aircraft Selection

			Alternatives					
Attributes		w	A1	A2	A3	A4	A5	ideal
Performance								
	Maximum speed (mach)	20%	2	2	2.5	2	1.8	2.5
	Service ceiling (1000ft)	4%	60	50	60	50	50	60
	Maximum payload (Lbs)	4%	23	20	18	20	21	23
	Ferry range (NM)	12%	1900	2000	3500	2400	2300	3500
Usability								
	Maneuverability*	9%	7	8	8	9	9	9
	Safety*	21%	8	9	7	8	8	9
Service								
	Reliability*	12%	8	7	9	8	8	9
	Maintainability*	8%	9	7	8	7	7	9
Cost								
	Acquisition cost (\$M)	6%	4.5	5	6.5	5.5	5	4.5
	Operation cost (\$1000/year)	4%	90	90	100	80	70	70
*(1: very bad; 10: very good)								

Fighter Aircraft Selection

Normalized values			Alternatives					
Attributes		w	A1	A2	A3	A4	A5	ideal
Performance								
	Maximum speed (mach)	20%	0.80	0.80	1.00	0.80	0.72	1
	Service ceiling (1000ft)	4%	1.00	0.83	1.00	0.83	0.83	1
	Maximum payload (Lbs)	4%	1.00	0.87	0.78	0.87	0.91	1
	Ferry range (NM)	12%	0.54	0.57	1.00	0.69	0.66	1
Usability								
	Maneuverability*	9%	0.78	0.89	0.89	1.00	1.00	1
	Safety*	21%	0.89	1.00	0.78	0.89	0.89	1
Service								
	Reliability*	12%	0.89	0.78	1.00	0.89	0.89	1
	Maintainability*	8%	1.00	0.78	0.89	0.78	0.78	1
Cost								
	Acquisition cost (\$M)	6%	1.00	0.90	0.69	0.82	0.90	1
	Operation cost (\$1000/year)	4%	0.78	0.78	0.70	0.88	1.00	1

Fighter Aircraft Selection

SAW

	A1	A2	A3	A4	A5	ideal
Vi	0.8396	0.8274	0.8953	0.8400	0.8323	1

WP

	A1	A2	A3	A4	A5	ideal
Vi	8.1716	8.0707	8.7689	8.2565	8.1485	9.8763
r	0.8274	0.8172	0.8879	0.8360	0.8251	1

Additional Example II

Scholar Selection

Candidates	GRE	GPA	Interview*	Recom.*	SoP*
	30%	20%	20%	15%	15%
Aynur	690	3.1	9	7	4
Batuhan	590	3.9	7	6	10
Canan	600	3.6	8	8	7
Deniz	620	3.8	7	10	6
Eda	700	2.8	10	4	6
Fatih	650	4	6	9	8
Ideal	700	4	10	10	10
*(1: very bad; 10: very good); SoP: Statement of Purpose					

Scholar Selection

TOPSIS

r_{ij}	GRE	GPA	Interview	Recom.	SoP
Aynur	0.4381	0.3555	0.4623	0.3763	0.2306
Batuhan	0.3746	0.4472	0.3596	0.3226	0.5764
Canan	0.3809	0.4128	0.4109	0.4301	0.4035
Deniz	0.3936	0.4357	0.3596	0.5376	0.3458
Eda	0.4444	0.3211	0.5137	0.2150	0.3458
Fatih	0.4127	0.4587	0.3082	0.4838	0.4611

v_{ij}	GRE	GPA	Interview	Recom.	SoP
Aynur	0.1314	0.0711	0.0925	0.0564	0.0346
Batuhan	0.1124	0.0894	0.0719	0.0484	0.0865
Canan	0.1143	0.0826	0.0822	0.0645	0.0605
Deniz	0.1181	0.0871	0.0719	0.0806	0.0519
Eda	0.1333	0.0642	0.1027	0.0323	0.0519
Fatih	0.1238	0.0917	0.0616	0.0726	0.0692
+ve ideal	0.1333	0.0917	0.1027	0.0806	0.0865
-ve ideal	0.1124	0.0642	0.0616	0.0323	0.0346

Scholar Selection

TOPSIS

	S^+	S^-	C^+	rank
Aynur	0.0617	0.0441	0.4167	6
Batuhan	0.0493	0.0608	0.5519	2
Canan	0.0424	0.0498	0.5396	4
Deniz	0.0490	0.0575	0.5399	3
Eda	0.0655	0.0493	0.4291	5
Fatih	0.0463	0.0609	0.5681	1