## UTA Method

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#### Abstract

This report contains the summary of the chapter **UTA Methods** from the Book: Multiple Criteria Decision Analysis. This document was made during my internship at LAMSADE in the summer of 2017.

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## Introduction

In 1982, E. Jacquet-Lagrze and J. Siskos proposed a decision mehtod called UTA. This method is proposed by the Multi-Attribute Utility Theory (MAUT) that build a utility function based on the DM¹ preferences.

Assuming that the decision is given, the UTA method will find the rational basis for the decision being made. Or how can we use the DM's preference leading to the exact same or "similar" decision.

The UTA method is used to solve a multi-criteria problem. It build a utility function based on the preferences of the DM and it consist in solving a linear program (LP).

UTA methods adpot the aggregation-disaggregation principles. The aggregation is where the model is known and the preferences are unknown, while the disaggregation is where the model is based on a given preferences

 $<sup>^{1}</sup>$  Decision Maker

## Method

#### 2.1 Principles and Notation

#### 2.1.1 Weighted form

A weighted form of the UTA Method has the following form:

$$u(g) = \sum_{i=1}^{n} p_i u_i(g_i)$$
 (2.1)

subject to normalization constraints:

$$\begin{cases}
\sum_{i=1}^{n} p_i = 1 \\
u_i(g_{i*}) = 0, u_i(g_{i*}) = 1, \forall i = 1, 2, ..., n
\end{cases}$$
(2.2)

where  $u_i, i = 1, 2, ..., n$  are non decreasing real valued functions, and  $p_i$  is the weight of  $u_i$ .

#### 2.1.2 Unweighted form

An unweighted form of the UTA Method has the following form:

$$u(g) = \sum_{i=1}^{n} u_i(g_i)$$
 (2.3)

subject to normalization constraints:

$$\begin{cases}
\sum_{i=1}^{n} u_i(g_{i*}) = 1 \\
u_i(g_{i*}) = 0, \forall i = 1, 2, ..., n
\end{cases}$$
(2.4)

#### 2.2 Development

The value of each alternative  $a \in A_R$  may be written as:

$$u'[g(a)] = \sum_{i=1}^{n} u_i[g_i(a)] + \sigma(a) \forall a \in A_R$$
 (2.5)

where  $\sigma(a)$  is a potential error relative to u'[g(a)].

We use the linear interpolation to estimate the marginal value function. For each criteria, the interval  $[g_{i*}, g_i^*]$  is cut into  $(\alpha_i - 1)$  equals interval, and the end points  $g_i^j$  are given by the formula:

$$g_i^j = g_{i*} + \frac{j-1}{\alpha_i - 1} (g_i^* - g_{i*}) \forall j = 1, 2, ..., \alpha_i$$
 (2.6)

The marginal value of an action a is calculated by a linear interpolation

$$u_i[g_i(a)] = u_i(g_i^j) + \frac{g_i(a) - g_i^j}{g_i^{j+1} - g_i^j} [u_i(g_i^{j+1}) - u_i(g_i^j)]$$
(2.7)

The set of reference action  $A_R = a_1, a_2, ..., a_m$  is arranged where  $a_1$  is the best action and  $a_m$  is the worst action. Which means:

$$\begin{cases}
\Delta(a_k, a_{k+1}) \ge \delta \\
\Delta(a_k, a_{k+1}) = 0
\end{cases}$$
(2.8)

where  $\Delta(a_k, a_{k+1}) = u'[g(a_k)] - u'[g(a_{k+1})]$  and  $\delta$  is a small positive number.

The marginal value  $u_i(g_i)$  must satisfy these constraints:

$$u_i(g_i^{j+1}) - u_i(g_i^{j}) \ge s_i \forall j = 1, 2, ..., \alpha_i - 1, i = 1, 2, ..., n$$
 (2.9)

where  $s_i \geq 0$  and being indifference for each criteria.

The marginal value functions are estimated by means of the Linear Programm with (2.3), (2.4), (2.8), (2.9) as constraints, and an objective function depending on the  $\sigma(a)$ :

$$[min]F = \sum_{a \in A_B} \sigma(a)$$

subject to:

$$\begin{cases}
\Delta(a_k, a_{k+1}) \ge \delta & \text{or } \Delta(a_k, a_{k+1}) = 0 \\
u_i(g_i^{j+1}) - u_i(g_i^{j}) \ge 0 & \forall i \text{ and } j \\
\sum_{i=1}^n u_i(g_i^*) = 1
\end{cases}$$

$$(2.10)$$

$$u_i(g_i^*) = 0 \quad u_i(g_i^{j}) \ge 0 \quad \sigma(a) \ge 0 \quad \forall a \in A_R \quad \forall i \text{ and } j$$

The analysis found by this LP is a post-optimality analysis.

#### 2.3 UTASTAR

The UTASTAR is an improved version of the UTA.

#### 2.3.1 Principles and Notations

In UTA we used a single error  $\sigma(a)$  in UTASTAR we use a double positive error function :

$$u'[g(a)] = \sum_{i=1}^{n} u_i[g_i(a)] - \sigma^+(a) + \sigma^-(a) \quad \forall a \in A_R$$
 (2.11)

where  $\sigma^+(a)$  and  $\sigma^-(a)$  are the underestimation and overstimation of the error function.

So, the UTASTAR algorithm has the following steps:

1. Find the value of reference actions  $u[g(a_k)]$  in terms of marginal values  $u_i(g_i)$  and in terms of variables  $w_{ij}$ 

$$\begin{cases} u_i(g_i^1) = 0 & \forall i = 1, 2, ..., n \\ u_i(g_i^j) = \sum_{t=1}^{j-1} w_{it} & \forall i = 1, 2, ..., n \quad and \quad j = 2, 3, ..., \alpha_i - 1 \end{cases}$$
 (2.12)

2. Write for each pair of consecutive action by including the errors function  $\sigma^+(a)$  and  $\sigma^-(a)$ 

$$\Delta(a_k, a_{k+1}) = u[g(a_k)] - \sigma^+(a_k) + \sigma^-(a_k) - u[g(a_{k+1})] + \sigma^+(a_{k+1}) - \sigma^-(a_{k+1})$$
(2.13)

3. Solve the Linear Program

$$[min]z = \sum_{k=1}^{m} [\sigma^{+}(a_k) + \sigma^{-}(a_k)]$$

subject to:

$$\begin{cases}
\Delta(a_k, a_{k+1}) \ge \delta & \text{or } \Delta(a_k, a_{k+1}) = 0 \\
\sum_{i=1}^n \sum_{j=1}^{\alpha_i - 1} w_{ij} = 1 \\
w_{ij} \ge 0, \quad \sigma^+(a_k) \ge 0, \quad \sigma^-(a_k) \ge 0, \quad \forall i, j \text{ and } k
\end{cases} \tag{2.14}$$

4. Test the exitence of the optimal solution of the LP. You can find the mean additive value fuction of the optimal solutions which maximise the

objective functions (in case of multiple optimal solutions) :

$$u_i(g_i^*) = \sum_{j=1}^{\alpha_i - 1} w_{ij} \quad \forall i = 1, 2, ..., n$$
 (2.15)

With the additional constraint of (2.14), we add the following constaint:

$$\sum_{k=1}^{m} [\sigma^{+}(\alpha_k) + \sigma^{-}(\alpha_k)] \ge z^* + \varepsilon \tag{2.16}$$

where  $z^*$  is the optimal value of the LP found in step 3 and  $\varepsilon$  is a very small positive value

#### 2.3.2 Comparison

The UTASTAR has been considered a better algorithm than UTA. Better result were found using the UTASTAR algorithm.

Variants

# Applications and UTA-Based DSS

## Conclusion

The UTA method build a utility function based on the preferences of the DM and it consist in solving a linear program (LP) to solve a multi-criteria problem.

This method will elaborate a model of preferences which is as similiar as possible to the DM's preferences.

# **Bibliography**

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