

MULTI-CRITERIA DECISION MAKING FOR REAL ESTATE SELECTION – A PROBLEM STRUCTURING FRAMEWORK AND REAL ESTATE DECISION MODEL

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ABSTRACT

The real estate selection process is a typical Multi-Criteria Decision Making (MCDM) problem. This paper presents the methodology adopted to identify relevant decision problem parameters for luxury real estate selection in Majorca. The fieldwork was translated into a generic MCDM problem structuring process and a modified Evidential Reasoning (ER) approach was used for decision analysis.

INTRODUCTION

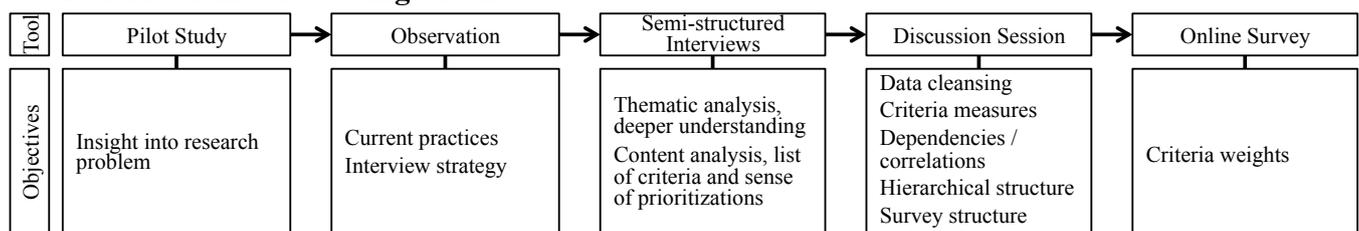
At present, a homebuyer’s selection decision is greatly reliant on intuition and is frequently simplified using rules of thumb to deal with high complexity. This complexity results from a large number of correlated and dependent criteria, and is further intensified by scarcity of available information. Accordingly, to provide a robust analysis of alternatives, the research first investigates current practices to enhance problem understanding, proposes a generic MCDM problem structuring process to contribute to limited literature in this area [1] and employs a modified ER approach to model the decision problem.

FIELDWORK AND GENERIC MCDM PROBLEM STRUCTURING FRAMEWORK

Data for this study was collected in Majorca due to its well-defined geographical space. A multi-stage mixed methods approach was designed with each collection method having a unique purpose to capture certain data and support subsequent phases. A pilot study with five estate agents provided a first impression of the problem and target audience, i.e. luxury homebuyers. Then, observations helped capture interactions between these two stakeholders, offering invaluable insights into current practices. Subsequently, 28 semi-structured interviews highlighted 214 non-repeating criteria. A discussion session with three property experts reduced this number to 144 relevant but potentially dependent and correlated factors, whilst also specifying associated measures and organising remaining criteria into a hierarchy. In the final stage, 255 online surveys provided more representative data and elicited weight distributions. From the study, valuable data collection and analysis procedures emerged. Figure 1 presents a generic MCDM problem structuring process that can be followed to define a decision problem and subsequently choose an adequate decision-making model, adjust one or develop a tailored method.

FIGURE 1

MCDM Problem Structuring Framework



EVIDENTIAL REASONING RULE

The ER rule, based on the Dempster-Shafer evidence theory, is a general probabilistic reasoning process facilitating the combination of multiple independent pieces of evidence, whilst acknowledging evidence weight and reliability [3] and forming a prior-free inference process [2].

First, a frame of discernment (Θ) is defined of mutually exclusive and collectively exhaustive propositions $\{\theta_1, \dots, \theta_N\}$ [2]. Basic probabilities are assigned to any subset of propositions, profiling each piece of evidence by a belief distribution (BD) on the power set of Θ [3]. The power set of the frame of discernment $P(\Theta)$ entails 2^N subsets of Θ [3].

$$P(\Theta) = \{\emptyset, \{\theta_1\}, \dots, \{\theta_N\}, \{\theta_1, \theta_2\}, \dots, \{\theta_1, \theta_N\}, \dots, \{\theta_1, \dots, \theta_{N-1}\}, \Theta\} \quad (1)$$

where \emptyset represents the empty set. A piece of evidence e_j profiled using a BD is defined by:

$$e_j = \left\{ (\theta, p_{\theta,j}), \forall \theta \subseteq \Theta, \sum_{\theta \subseteq \Theta} p_{\theta,j} = 1 \right\} \quad (2)$$

where $p_{\theta,j}$ is the degree of belief to which e_j supports proposition θ , being any element of $P(\Theta)$ except \emptyset [3] [2]. Following the section above, there are three key elements that are used when combining multiple pieces of evidence, the BD, weight and reliability [3] [2]. First, to attain the reasoning process in the ER rule, a weighted BD with reliability needs to be computed [3]:

$$m_j = \{(\theta, \tilde{m}_{\theta,j}), \forall \theta \subseteq \Theta; (P(\Theta), \tilde{m}_{P(\Theta),j})\} \quad (3)$$

where $\tilde{m}_{\theta,j}$ is the degree of support for proposition θ from e_j with reliability r_j and weight w_j :

$$\tilde{m}_{\theta,j} = \begin{cases} 0 & \theta = \emptyset \\ c_{rw,j} m_{\theta,j} & \theta \subseteq \Theta, \theta \neq \emptyset \\ c_{rw,j} (1 - r_j) & \theta = P(\Theta) \end{cases} \quad (4)$$

A normalisation factor $c_{rw,j}$ is used to ensure $\sum_{\theta \subseteq \Theta} \tilde{m}_{\theta,j} + \tilde{m}_{P(\Theta),j} = 1$, where $c_{rw,j} = 1/(1 + w_j - r_j)$. $m_{\theta,j}$ is the probability of proposition θ multiplied by the weight ($p_{\theta,j} w_j$). Note, only if pieces of evidence are acquired from different information sources, w_j is not equivalent to r_j , i.e. $w_j \neq r_j$.

Given the weighted BD with reliability, combined support for proposition θ using two pieces of evidence e_i and e_j can be computed using the ER rule [3]:

$$p_{\theta,e(2)} = \begin{cases} 0 & \theta = \emptyset \\ \frac{\hat{m}_{\theta,e(2)}}{\sum_{D \subseteq \Theta} \hat{m}_{D,e(2)}} & \theta \subseteq \Theta, \theta \neq \emptyset \end{cases} \quad (5)$$

$$\hat{m}_{\theta,e(2)} = [(1 - w_j) m_{\theta,i} + (1 - w_i) m_{\theta,j}] + [m_{\theta,i} m_{\theta,j}] \quad (6)$$

The first part of equation (6) symbolises the bounded sum of individual support on proposition θ from evidence e_i and e_j , whilst $[m_{\theta,i} m_{\theta,j}]$ represents the orthogonal sum of collective support from both pieces measuring the level of all intersected support on proposition θ [3] [2].

A modified version of the ER rule capable of handling dependencies amongst evaluation criteria was applied to the identified real estate selection problem.

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