

Introduction to Belief Functions

Decision analysis

Thierry Denœux

Summer 2022



Outline

- 1 Decision-making under complete ignorance
- 2 Decision-making with probabilities
- 3 Decision-making with belief functions



Example of decision problem under uncertainty

Act (Purchase)	Good Economic Conditions	Poor Economic Conditions
Apartment building	50,000	30,000
Office building	100,000	-40,000
Warehouse	30,000	10,000



Formal framework

Acts, outcomes, states of nature

- A decision problem can be seen as a situation in which a **decision-maker (DM)** has to choose a course of action (an **act**) in some set $\mathcal{F} = \{f_1, \dots, f_n\}$
- An act may have different **consequences** (outcomes), depending on the **state of nature**
- Denoting by $\Omega = \{\omega_1, \dots, \omega_r\}$ the set of states of nature and by \mathcal{C} the set of consequences (or outcomes), an act can be formalized as a **mapping f from Ω to \mathcal{C}**
- In this lecture, the three sets Ω , \mathcal{C} and \mathcal{F} will be assumed to be finite



Formal framework

Utilities

- The desirability of the consequences can often be modeled by a numerical **utility function** $u : \mathcal{C} \rightarrow \mathbb{R}$, which assigns a numerical value to each consequence
- The higher this value, the more desirable is the consequence for the DM
- In some problems, the consequences can be evaluated in terms of monetary value. The utilities can then be defined as the payoffs, or a function thereof
- If the actions are indexed by i and the states of nature by j , we will denote by u_{ij} the quantity $u[f_i(\omega_j)]$
- The $n \times r$ matrix $U = (u_{ij})$ will be called a **payoff or utility matrix**



Payoff matrix

Act (Purchase)	Good Economic Conditions (ω_1)	Poor Economic Conditions (ω_2)
Apartment building (f_1)	50,000	30,000
Office building (f_2)	100,000	-40,000
Warehouse (f_3)	30,000	10,000



Formal framework

Preferences

- If the true state of nature ω is known, the desirability of an act f can be deduced from that of its consequence $f(\omega)$
- Typically, the state of nature is unknown. Based on partial information, it is usually assumed that the DM can express **preferences among acts**, which may be represented mathematically by a **preference relation** \succsim on \mathcal{F}
- This relation is interpreted as follows: given two acts f and g , $f \succsim g$ means that f is found by the DM to be **at least as desirable** as g
- We also define
 - The **strict preference relation** as $f \succ g$ iff $f \succsim g$ and $\text{not}(g \succsim f)$ (meaning that f is strictly more desirable than g) and
 - The **indifference relation** $f \sim g$ iff $f \succsim g$ and $g \succsim f$ (meaning that f and g are equally desirable)



Decision problems

- The **decision problem** can be formalized as building a preference relation among acts, from a utility matrix and some description of uncertainty, and finding the maximal elements of this relation
- Depending on the nature of the available information, different decision problems arise:
 - 1 Decision-making under ignorance
 - 2 Decision-making with probabilities
 - 3 Decision-making with belief functions



Outline

- 1 Decision-making under complete ignorance
- 2 Decision-making with probabilities
- 3 Decision-making with belief functions



Problem and non-domination principle

- We assume that the DM is **totally ignorant of the state of nature**: all the information given to the DM is the utility matrix U
- A act f_i is said to be **dominated** by f_k if the outcomes of f_k are at least as desirable as those of f_i for all states, and strictly more desirable for at least one state

$$\forall j, u_{kj} \geq u_{ij} \text{ and } \exists j, u_{kj} > u_{ij}$$

- **Non-domination principle**: an act cannot be chosen if it is dominated by another one



Example of a dominated act

Act (Purchase)	Good Economic Conditions (ω_1)	Poor Economic Conditions (ω_2)
Apartment building (f_1)	50,000	30,000
Office building (f_2)	100,000	-40,000
Warehouse (f_3)	30,000	10,000



Criteria for rational choice

- After all dominated acts have been removed, there remains the problem of ordering them by desirability, and of finding the **set of most desirable acts**
- Several criteria of “rational choice” have been proposed to derive a preference relation over acts, including:

1 **Laplace criterion**

$$f_i \succeq f_k \text{ iff } \frac{1}{r} \sum_j u_{ij} \geq \frac{1}{r} \sum_j u_{kj}.$$

2 **Maximax criterion**

$$f_i \succeq f_k \text{ iff } \max_j u_{ij} \geq \max_j u_{kj}.$$

3 **Maximin (Wald) criterion**

$$f_i \succeq f_k \text{ iff } \min_j u_{ij} \geq \min_j u_{kj}.$$



Example

Act	ω_1	ω_2	ave	max	min
Apartment (f_1)	50,000	30,000	40,000	50,000	30,000
Office (f_2)	100,000	-40,000	30,000	100,000	-40,000



Hurwicz criterion

- Hurwicz criterion: $f_i \succsim f_k$ iff

$$\alpha \min_j u_{ij} + (1 - \alpha) \max_j u_{ij} \geq \alpha \min_j u_{kj} + (1 - \alpha) \max_j u_{kj}$$

where α is a parameter in $[0, 1]$, called the **pessimism index**

- Boils down to
 - the maximax criterion if $\alpha = 0$
 - the maximin criterion if $\alpha = 1$
- α describes the DM's **attitude toward ambiguity**.
- Formal justification given by Arrow and Hurwicz (1972).



Outline

- 1 Decision-making under complete ignorance
- 2 Decision-making with probabilities**
- 3 Decision-making with belief functions



Lottery

- Let us now consider the situation where uncertainty about the state of nature is **quantified by a probability distribution** π on Ω .
- These probabilities can be objective (**decision under risk**) or subjective.
- An act $f : \Omega \rightarrow \mathcal{C}$ induces a probability measure p on the set \mathcal{C} of consequences (assumed to be finite), called a **lottery**:

$$\forall c \in \mathcal{C}, \quad p(c) = \sum_{\{\omega: f(\omega)=c\}} \pi(\omega).$$



Maximum Expected Utility principle

- Given a utility function $u : \mathcal{C} \rightarrow \mathbb{R}$, the **expected utility** for a lottery p is

$$\mathbb{E}_p(u) = \sum_{c \in \mathcal{C}} u(c)p(c).$$

- Maximum Expected Utility (MEU) principle:** a lottery p_i is more desirable than a lottery p_k if it has a higher expected utility:

$$p_i \succeq p_k \Leftrightarrow \mathbb{E}_{p_i}(u) \geq \mathbb{E}_{p_k}(u).$$

- The MEU principle was first axiomatized by von Neumann and Morgenstern (1944).



Example

Act	ω_1	ω_2
Apartment (f_1)	50,000	30,000
Office (f_2)	100,000	-40,000

- Assume that there is 60% chance that the economic situation will be poor (ω_2).
- Act f_1 induces the lottery p_1 such that $p_1(50,000) = 0.4$ and $p_1(30,000) = 0.6$. Act f_2 induces the lottery p_2 such that $p_2(100,000) = 0.4$ and $p_2(-40,000) = 0.6$.
- The expected utilities are

$$\mathbb{E}_{p_1}(u) = 50,000 \times 0.4 + 30,000 \times 0.6 = 38,000$$

$$\mathbb{E}_{p_2}(u) = 100,000 \times 0.4 - 40,000 \times 0.6 = 16,000$$

- Act f_1 is thus more desirable according to the maximum expected utility criterion.



Outline

- 1 Decision-making under complete ignorance
- 2 Decision-making with probabilities
- 3 Decision-making with belief functions**



How belief functions come into the picture

Belief functions become components of a decision problem in any of the following two situations (or both)

- 1 The decision maker's subjective beliefs concerning the state of nature are described by a belief function Bel^Ω on Ω
- 2 The DM is not able to precisely describe the outcomes of some acts under each state of nature



Case 1: uncertainty described by a belief function

- Let m^Ω be a mass function on Ω
- Any act $f : \Omega \rightarrow \mathcal{C}$ carries m^Ω to the set \mathcal{C} of consequences, yielding a mass function $m_f^\mathcal{C}$, which quantifies the DM's beliefs about the outcome of act f
- Each mass $m^\Omega(A)$ is transferred to $f(A)$

$$m_f^\mathcal{C}(B) = \sum_{\{A \subseteq \Omega : f(A) = B\}} m^\Omega(A)$$

for any $B \subseteq \mathcal{C}$

- $m_f^\mathcal{C}$ is a **credibilistic lottery** corresponding to act f



Case 2: partial knowledge of outcomes

- In that case, an act may formally be represented by a **multi-valued mapping** $f : \Omega \rightarrow 2^{\mathcal{C}}$, assigning a set of possible consequences $f(\omega) \subseteq \mathcal{C}$ to each state of nature ω
- Given a probability measure P on Ω , f then induces the following mass function $m_f^{\mathcal{C}}$ on \mathcal{C} ,

$$m_f^{\mathcal{C}}(B) = \sum_{\{\omega \in \Omega: f(\omega)=B\}} p(\omega)$$

for all $B \subseteq \mathcal{C}$



Example

- Let $\Omega = \{\omega_1, \omega_2, \omega_3\}$ and m^Ω the following mass function

$$\begin{aligned} m^\Omega(\{\omega_1, \omega_2\}) &= 0.3, & m^\Omega(\{\omega_2, \omega_3\}) &= 0.2 \\ m^\Omega(\{\omega_3\}) &= 0.4, & m^\Omega(\Omega) &= 0.1 \end{aligned}$$

- Let $\mathcal{C} = \{c_1, c_2, c_3\}$ and f the act

$$f(\omega_1) = \{c_1\}, \quad f(\omega_2) = \{c_1, c_2\}, \quad f(\omega_3) = \{c_2, c_3\}$$

- To compute $m_f^{\mathcal{C}}$, we transfer the masses as follows

$$m^\Omega(\{\omega_1, \omega_2\}) = 0.3 \rightarrow f(\omega_1) \cup f(\omega_2) = \{c_1, c_2\}$$

$$m^\Omega(\{\omega_2, \omega_3\}) = 0.2 \rightarrow f(\omega_2) \cup f(\omega_3) = \{c_1, c_2, c_3\}$$

$$m^\Omega(\{\omega_3\}) = 0.4 \rightarrow f(\omega_3) = \{c_2, c_3\}$$

$$m^\Omega(\Omega) = 0.1 \rightarrow f(\omega_1) \cup f(\omega_2) \cup f(\omega_3) = \{c_1, c_2, c_3\}$$

- Finally, we obtain the following mass function on \mathcal{C}

$$m^{\mathcal{C}}(\{c_1, c_2\}) = 0.3, \quad m^{\mathcal{C}}(\{c_2, c_3\}) = 0.4, \quad m^{\mathcal{C}}(\mathcal{C}) = 0.3$$



Decision problem

- In the two situations considered above, we can assign to each act f a **credibilistic lottery**, defined as a mass function on \mathcal{C}
- Given a utility function u on \mathcal{C} , we then need to **extend the MEU model**
- Several such extensions will now be reviewed



Upper and lower expectations

- Let m be a mass function on \mathcal{C} , and u a utility function $\mathcal{C} \rightarrow \mathbb{R}$
- The **lower and upper expectations** of u are defined, respectively, as the averages of the minima and the maxima of u within each focal set of m

$$\underline{\mathbb{E}}_m(u) = \sum_{A \subseteq \mathcal{C}} m(A) \min_{c \in A} u(c)$$

$$\overline{\mathbb{E}}_m(u) = \sum_{A \subseteq \mathcal{C}} m(A) \max_{c \in A} u(c)$$

- It is clear that $\underline{\mathbb{E}}_m(u) \leq \overline{\mathbb{E}}_m(u)$, with the inequality becoming an equality when m is Bayesian, in which case the lower and upper expectations collapse to the usual expectation
- If $m = m_A$ is logical with focal set A , then $\underline{\mathbb{E}}_m(u)$ and $\overline{\mathbb{E}}_m(u)$ are, respectively, the minimum and the maximum of u in A



Corresponding decision criteria

- Having defined the notions of lower and upper expectations, we can define two preference relations among credibilistic lotteries as

$$m_1 \succcurlyeq m_2 \text{ iff } \underline{\mathbb{E}}_{m_1}(u) \geq \underline{\mathbb{E}}_{m_2}(u)$$

and

$$m_1 \succbar m_2 \text{ iff } \bar{\mathbb{E}}_{m_1}(u) \geq \bar{\mathbb{E}}_{m_2}(u)$$

- Relation \succcurlyeq corresponds to a **pessimistic (or conservative)** attitude of the DM. When m is logical, it corresponds to the **maximin criterion**
- Symmetrically, \succbar corresponds to an **optimistic attitude** and extends the **maximax criterion**
- Both criteria boil down to the MEU criterion when m is Bayesian



Generalized Hurwicz criterion

- The **Hurwicz criterion** can be generalized as

$$\begin{aligned}\mathbb{E}_{m,\alpha}(u) &= \sum_{A \subseteq C} m(A) \left(\alpha \min_{c \in A} u(c) + (1 - \alpha) \max_{c \in A} u(c) \right) \\ &= \alpha \underline{\mathbb{E}}_m(u) + (1 - \alpha) \overline{\mathbb{E}}(u)\end{aligned}$$

where $\alpha \in [0, 1]$ is a **pessimism index**

- This criterion was first introduced and justified axiomatically by Jaffray (1988)



Transferable belief model

- A completely different approach to decision-making with belief function was advocated by Smets, as part of the **Transferable Belief Model**
- Smets defended a two-level mental model
 - 1 A **credal level**, where an agent's beliefs are represented by belief functions, and
 - 2 A **pignistic level**, where decisions are made by maximizing the EU with respect to a probability measure derived from a belief function
- The rationale for introducing probabilities at the decision level is the avoidance of **Dutch books**
- Smets argued that the belief-probability transformation T should be **linear**, i.e., it should verify

$$T(\alpha m_1 + (1 - \alpha)m_2) = \alpha T(m_1) + (1 - \alpha)T(m_2),$$

for any mass functions m_1 and m_2 and for any $\alpha \in [0, 1]$



Pignistic transformation

- The only linear belief-probability transformation T is the **pignistic transformation**, with $p_m = T(m)$ given by

$$p_m(c) = \sum_{\{A \subseteq \mathcal{C} : c \in A\}} \frac{m(A)}{|A|}, \quad \forall c \in \mathcal{C}$$

- The expected utility w.r.t. the pignistic probability is

$$\mathbb{E}_p(u) = \sum_{c \in \mathcal{C}} p_m(c) u(c) = \sum_{A \subseteq \mathcal{C}} m(A) \left(\frac{1}{|A|} \sum_{c \in A} u(c) \right)$$

- The maximum pignistic expected utility criterion thus extends the **Laplace criterion**



Summary

non-probabilized		belief functions	probabilized
maximin	\longleftrightarrow	lower expectation	
maximax	\longleftrightarrow	upper expectation	
Laplace	\longleftrightarrow	pignistic expectation	expected utility
Hurwicz	\longleftrightarrow	generalized Hurwicz	

