

Decision Methods and Models

Master's Degree in Computer Science

Roberto Cordone

DI - Università degli Studi di Milano

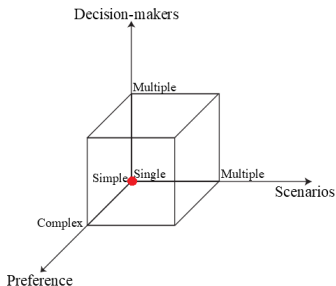


- Schedule: **Thursday 16.30 - 18.30 in Aula Magna (CS department)**
Friday 12.30 - 14.30 in classroom 301
- Office hours: **on appointment**
- E-mail: **roberto.cordone@unimi.it**
- Web page: **<https://homes.di.unimi.it/cordone/courses/2024-mmd/2024-mmd.html>**
- Ariel site: **<https://myariel.unimi.it/course/view.php?id=4467>**

Structured models of preference

We assume

- a **preference relation** Π with a **consistent utility function** $u(f)$
- a **certain environment**: $|\Omega| = 1 \Rightarrow f(x, \bar{\omega})$ reduces to $f(x)$
- a **single decision-maker**: $|D| = 1 \Rightarrow \Pi_d$ reduces to Π



We know the preference Π , not the utility function $u(f)$

The general process to find it is complex and error-prone

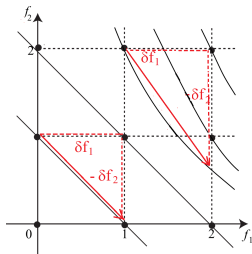
For additive functions, it is much simpler, but is $u(f)$ additive?

- for $p \geq 3$, mutual preferential independence \Leftrightarrow additivity
- for $p = 2$, mutual preferential independence \Leftarrow additivity

The problem lies in how the indifference curves behave in F

Marginal rate of substitution (MRS)

The behaviour of a curve is described by the relation between f_1 and f_2 :
from f , vary f_1 and update f_2 so as to remain on the indifference curve



Marginal rate of substitution (MRS) of f_1 with f_2 in f

$$\lambda_{12}(f) = \lim_{\delta f_1 \rightarrow 0} - \frac{\delta f_2(f, \delta f_1)}{\delta f_1}$$

with $\delta f_2(f, \delta f_1)$ such that $\begin{bmatrix} f_1 \\ f_2 \end{bmatrix} \sim \begin{bmatrix} f_1 + \delta f_1 \\ f_2 + \delta f_2(f, \delta f_1) \end{bmatrix}$

- the value depends on f , not on δf_1 , thanks to the limit operation
- the minus sign is used to obtain positive rates on decreasing curves (a frequent case: e.g., when both indicators are costs or benefits)

Equivalent expressions of the MRS (1)

Let the indifference curve be represented as a regular parametric line:

$$\begin{cases} f_1 = f_1(\alpha) \\ f_2 = f_2(\alpha) \end{cases}$$

with two functions of parameter α (continuous up to the first derivative)

The MRS can be expressed as

$$\lambda_{12}(f) = \lim_{\delta\alpha \rightarrow 0} - \frac{(f_2(\alpha + \delta\alpha) - f_2(\alpha))}{(f_1(\alpha + \delta\alpha) - f_1(\alpha))} = \frac{\lim_{\delta\alpha \rightarrow 0} - \frac{(f_2(\alpha + \delta\alpha) - f_2(\alpha))}{\delta\alpha}}{\lim_{\delta\alpha \rightarrow 0} \frac{(f_1(\alpha + \delta\alpha) - f_1(\alpha))}{\delta\alpha}} = - \frac{\frac{df_2}{d\alpha}}{\frac{df_1}{d\alpha}}$$

This form is useful to prove **reciprocity**: $\lambda_{12}(f) = \frac{1}{\lambda_{21}(f)}$

Example (Cobb-Douglas):

$$\begin{cases} f_1 = \frac{1}{\sqrt{\alpha}} \\ f_2 = \sqrt[3]{\alpha} \end{cases} \Rightarrow \lambda_{12}(f) = - \frac{\frac{1}{3}\alpha^{-2/3}}{-\frac{1}{2}\alpha^{-3/2}} = \frac{2}{3} \frac{f_2}{f_1}$$

(multiply both numerator and denominator by α)

Equivalent expressions of the MRS (2)

A second expression shows the **relation between the MRS and $u(f)$**

By changing α , we move on the indifference curve, but $u(f(\alpha)) = c$

$$\frac{du(f(\alpha))}{d\alpha} = 0 \Rightarrow \frac{\partial u}{\partial f_1} \frac{df_1}{d\alpha} + \frac{\partial u}{\partial f_2} \frac{df_2}{d\alpha} = 0 \Rightarrow -\frac{\frac{df_2}{d\alpha}}{\frac{df_1}{d\alpha}} = \lambda_{12}(f) = \frac{\frac{\partial u}{\partial f_1}}{\frac{\partial u}{\partial f_2}}$$

The MRS measures how much $u(f)$ depends on f_1 with respect to how much it depends on f_2

Example (Cobb-Douglas):

$$u(f) = f_1^2 f_2^3 \Rightarrow \lambda_{12}(f) = \frac{2f_1 f_2^3}{3f_1^2 f_2^2} = \frac{2}{3} \frac{f_2}{f_1}$$

Equivalent expressions of the *MRS* (3)

The third expression shows the **geometric relation** between the *MRS* and the shape of indifference curves (if invertible!)

$$\begin{cases} f_1 = f_1(\alpha) \\ f_2 = f_2(\alpha) \Leftrightarrow \alpha = \alpha(f_2) \end{cases} \Rightarrow f_1 = f_1(\alpha(f_2))$$

which implies that

$$\lambda_{12}(f) = -\frac{df_2}{d\alpha} \frac{d\alpha}{df_1} = -\frac{df_2}{df_1}$$

The *MRS* is the **negative slope of the tangent to the indifference curve**

Example (Cobb-Douglas):

$$u(f) = f_1^2 f_2^3 = c \Rightarrow f_2 = \sqrt[3]{\frac{c}{f_1^2}}$$

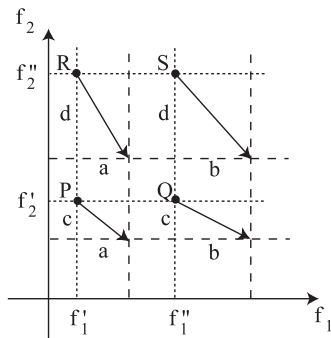
implies that

$$\lambda_{12}(f) = -\sqrt[3]{c} \left(-\frac{2}{3}\right) f_1^{-5/3} = \frac{2}{3} \frac{f_2}{f_1}$$

Corresponding trade-off condition

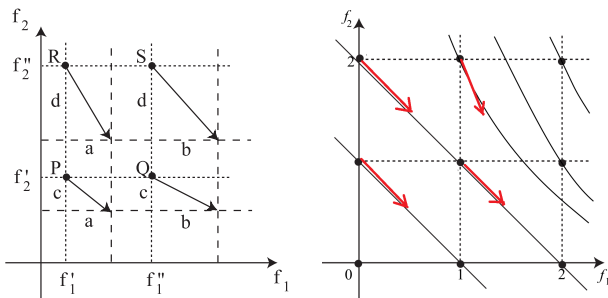
Some indifference maps enjoy the **corresponding trade-off condition**

$$\lambda_{12}(f'_1, f'_2) \cdot \lambda_{12}(f''_1, f''_2) = \lambda_{12}(f'_1, f''_2) \cdot \lambda_{12}(f''_1, f'_2)$$



This is a **global property**: it **relates far away impacts**

Corresponding trade-off condition



The corresponding trade-off condition is easier to interpret if rewritten as

$$\lambda_{12}(f_1', f_2') / \lambda_{12}(f_1', f_2'') = \lambda_{12}(f_1'', f_2') / \lambda_{12}(f_1'', f_2'')$$

$$\lambda_{12}(f_1', f_2') / \lambda_{12}(f_1'', f_2') = \lambda_{12}(f_1', f_2'') / \lambda_{12}(f_1'', f_2'')$$

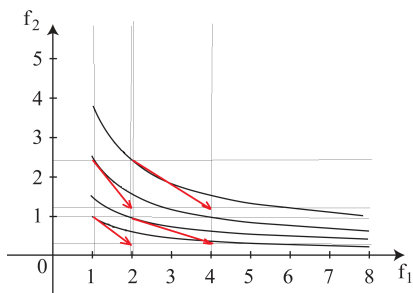
The ratio of λ between points with the same abscissa (ordinate) does not depend on the abscissa (ordinate)

In other words, even if the MRS is nonuniform, it changes by the same factor moving between the same coordinates

The example on the right violates it in $(0, 1)$, $(1, 1)$, $(0, 2)$, $(1, 2)$

Corresponding trade-off condition: positive example

On the contrary, the condition holds for the Cobb-Douglas example



In fact

$$\frac{2 f_2'}{3 f_1'} \cdot \frac{2 f_2''}{3 f_1''} = \frac{2 f_2'}{3 f_1''} \cdot \frac{2 f_2''}{3 f_1'}$$

Corresponding trade-off condition and additivity

Theorem

A preference relation Π admits an additive utility function $u(f)$ if and only if it enjoys both

- 1 mutual preferential independence
- 2 the corresponding trade-off condition

The “only-if” part is easy to prove

We needed the “if” part to close the gap in order to assume additivity:

- when $p \geq 3$, check mutual preferential independence for $p - 1$ indicator pairs
- when $p = 2$, check
 - independence for the single indicators
 - the corresponding trade-off condition

Building an additive utility function

$$u(f) = \sum_{l=1}^p u_l(f_l)$$

This expression assumes the same measure unit for all u_l functions

In practice, experts in different fields use different units. Therefore:

- 1 adopt normalised utilities: pure numbers $\tilde{u}_l(f_l)$ instead of $u_l(f_l)$
- 2 introduce weights w_l to combine them

$$u(f) = \sum_{l=1}^p w_l \tilde{u}_l(f_l)$$

Intuitively, we are splitting the task in two

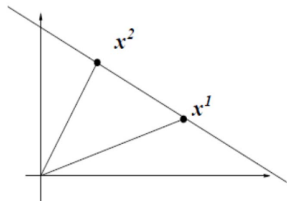
- 1 “rescale” indicators into utilities, removing all nonlinearities
- 2 “combine” heterogeneous utilities into a single one

Normalised additive utility function

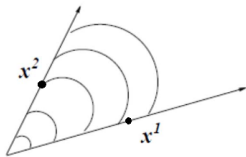
The normalised utility expression is more than a linear combination

It is a **convex combination**

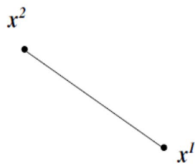
- **conic**: all coefficients are nonnegative ($w_l \geq 0$ for all $l \in P$)
- **affine**: the coefficients have unitary sum ($\sum_{l=1}^P w_l = 1$)



Affine combination



Conical combination



Convex combination

Building a normalised utility: the bisection method

Build the normalised utility $\tilde{u}_C(C)$ for the daily calorie intake C

Given $F_C = [0, 10000]$, interview an expert about a specific individual

- 1 ask the expert for the **worst values in F_C** (for example, $C_1^\dagger \leq 1000$ and $C_2^\dagger \geq 6000$) and **set $\tilde{u}_C(C) = 0$ for such values**
- 2 ask the expert for the **best values in F_C** (say, $2200 \leq C^\circ \leq 2600$) and **set $\tilde{u}_T(22) = 1$ for such values**
- 3 ask the expert for **values of exactly intermediate utility** between T^\dagger and T° (for example, $C = 1800$ and 3000) and **set $\tilde{u}_C(C) = 1/2$**
- 4 go on, asking for **values of intermediate utility between the fixed ones** and **set \tilde{u}_T accordingly**
- 5 **guess an interpolating function**

Costs and benefits proportional to the indicator f_I are easy to normalise

$$\tilde{u}_I(f_I) = \frac{f_I - \min_{x \in X} f_I(x)}{\max_{x \in X} f_I(x) - \min_{x \in X} f_I(x)}$$

What if $\min_x f_I(x)$ or $\max_x f_I(x)$ are unknown or hard to compute?

Computing the weights

As in the general case, find $p - 1$ independent pairs of indifferent impacts

- the equations are linear in w

$$f \sim f' \Leftrightarrow \tilde{u}(f) = \tilde{u}(f') \Leftrightarrow \sum_{l=1}^p \tilde{u}_l(f_l) w_l = \sum_{l=1}^p \tilde{u}_l(f'_l) w_l$$

- the normalisation condition imposes convexity

$$\sum_{l=1}^p w_l = 1$$

The process works correctly in the ideal case

- Problem: if indifference is imprecise, $(p - 1)$ pairs give wrong weights
- Solution: build a complete pairwise comparison of all indicators and analyse its consistency

Pairwise comparison matrix

Select a pair of indifferent impacts (f, f') with

- different values for f_l and f_m
- identical values for all other indicators: $f_n = f'_n$ for all $n \in P \setminus \{l, m\}$

The equation reduces to

$$w_l \tilde{u}_l(f_l) + w_m \tilde{u}_m(f_m) = w_l \tilde{u}_l(f'_l) + w_m \tilde{u}_m(f'_m)$$

that simply becomes

$$\tilde{\lambda}_{lm} = \frac{w_l}{w_m} = - \frac{\tilde{u}_m(f'_m) - \tilde{u}_m(f_m)}{\tilde{u}_l(f'_l) - \tilde{u}_l(f_l)}$$

The **pairwise comparison matrix** contains all the weight ratios

$$\tilde{\Lambda} = \left\{ \frac{w_l}{w_m} \right\}$$

expressing the **relative weights between the single normalised utilities**, that is, indicators (once nonlinearities and units of measures are removed)

Properties of the pairwise comparison matrix

A correct pairwise comparison matrix $\tilde{\Lambda}$ enjoys the following properties:

- 1 positivity: $\tilde{\lambda}_{lm} > 0$ for all $l, m \in P$
- 2 reciprocity: $\tilde{\lambda}_{lm} = \frac{1}{\tilde{\lambda}_{ml}}$ for all $l, m \in P$
- 3 consistency: $\tilde{\lambda}_{ln} = \tilde{\lambda}_{lm} \tilde{\lambda}_{mn}$ for all $l, m, n \in P$

Check these properties on $\tilde{\Lambda}$ to be sure that $u(f)$ makes sense

We shall discuss what to do when they are not satisfied