

# Common access resource games with asymmetric players

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Workshop on Biodiversity and Environment: Viability and Dynamic Games Perspectives

Montreal, November 2013

# Outline of the presentation

- Motivation for heterogeneous discounting (and non-constant discounting)
- An exhaustible resource model under common access
  - The case of two-asymmetric players
  - The case of  $N$ -asymmetric players
- References

- Economic agents usually choose between profits distributed over time. To make them comparable one must discount these payments at a reference moment and the theory of the discounted utility provides one framework for evaluating such delayed payoffs.
- Preferences are time consistent if, and only if, discount functions are exponentials with a constant instantaneous time preference rate (Strotz (1956)).

$$U_t = \int_t^T e^{-\delta(s-t)} L(x(s), u(s), s) ds + e^{-\delta(T-t)} F(x(T))$$

- However, there are some situations that can not be captured by standard discounting: impatient agents for short-run decisions (hyperbolic preferences / non-constant discounting) or situations in which the relative valuation of final function increases or decreases as we approach to the end of the planning horizon.
- This last case could be, for instance, when we want to model preferences about pensions plans or the legacy that an individual will leave to her/his descendants. Here, it could be the case that a decision maker will give more importance to her/his pension plan as she/he approaches the retirement date.

# Heterogeneous (constant) discount rates

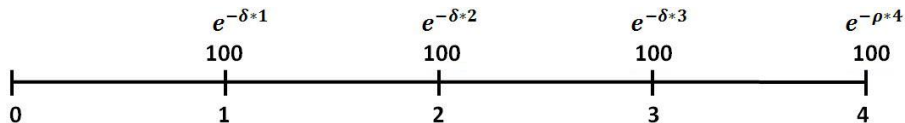
The preferences of the agent at time  $t$  take the form

$$U_t = \int_t^T e^{-\delta(s-t)} L(x(s), u(s), s) ds + e^{-\rho(T-t)} F(x(T)) ,$$

In this case, if  $\rho > \delta$ , the relative valuation of the final function  $F(x(T))$  increases as we are getting closer to  $T$ .

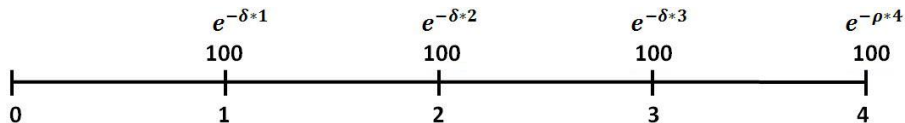
# Example

Imagine an agent who wants to value at  $t = 0$  the following payments distributed along the horizon  $[0, 4]$ .



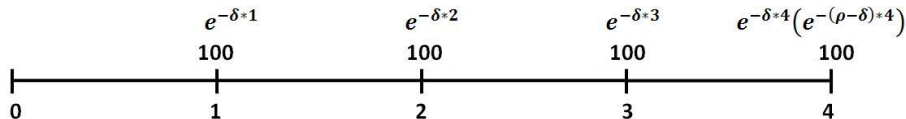
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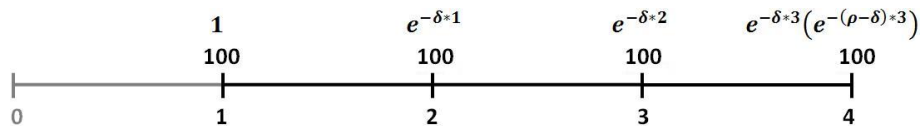
Note that we can rewrite the last discount factor term

$$e^{-\rho \cdot 4} = e^{-\delta \cdot 4} (e^{-(\rho - \delta) \cdot 4})$$



# Example

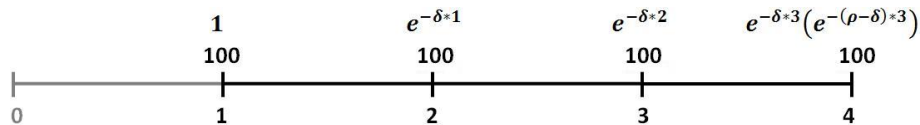
At  $t = 1$



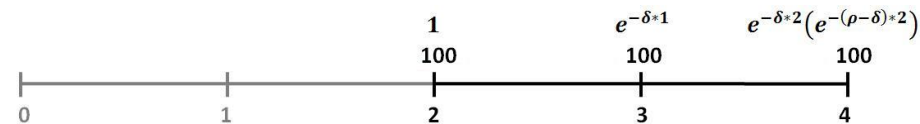


# Example

At  $t = 1$



At  $t = 2$



# Example

At  $t = 3$



Therefore, since we have assumed that  $\rho > \delta$ :

$$e^{-(\rho - \delta) * 4} < e^{-(\rho - \delta) * 3} < e^{-(\rho - \delta) * 2} < e^{-(\rho - \delta) * 1}$$

the relative value of the final function increases over time.

# Non-constant discounting: the deterministic case

The objective of an agent at time  $t$  (*the  $t$ -agent*) is:

$$\max_{\{u(s)\}} \int_t^T \theta(s-t)L(x(s), u(s), s) ds + \theta(T-t)F(x(T)) ,$$

$$\dot{x} = f(x, u, s), \quad x(t) = x_t.$$

Exponential function with a non-increasing instantaneous discount rate  $r(s)$

$$\theta(s-t) = e^{-\int_t^s r(\tau-t)d\tau} \left( \neq \bar{\theta}(s,t) = e^{-\int_t^s r(\tau)d\tau} \right)$$

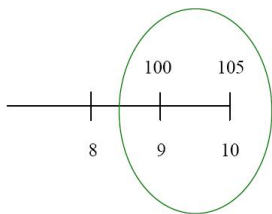
# Example

Suppose that you have to choose at  $t = 0$  between two sets of two capitals placed over a time horizon  $[0, 10]$ , the first set is placed in the short-run (**Set 1**) and the second in the long-run (**Set 2**):

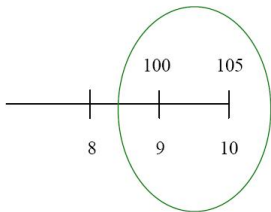


If you are more impatient for decisions in the short-run than in the long-run, then you can prefer (now) 100 EUR from Set 1 but 105 EUR from Set 2. In that case, your optimal current decision at  $t = 0$  will be to choose 100 EUR at  $t = 1$  and 105 EUR at  $t = 10$ .

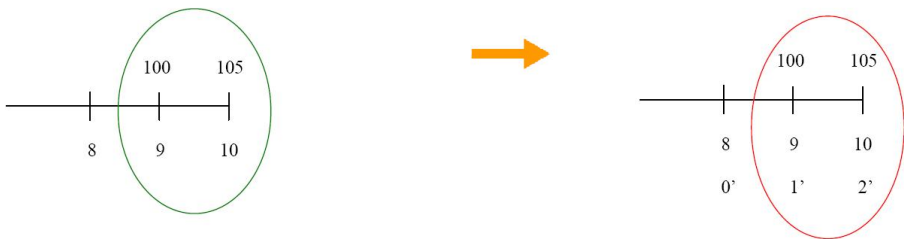
But imagine that time goes by (for instance, we are now at  $t = 8$ ) and you are offered the possibility to decide again. Now, you will prefer to receive the 100 EUR at  $t = 9$  rather than 105 EUR at  $t = 10$  (since now you are at evaluating both payments from moment  $t = 8$ ) rather than to wait a year and receive your original decision.



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Thus, our original plan is time inconsistent since, if we can re-optimize at later dates from the beginning, we will want to change our past decisions.

The relevant effect of heterogeneous discounting (and also non-constant discounting) is that preferences change with time (in a similar way than with non-constant discounting). In this sense, an agent making a decision in time  $t$  has different preferences compared with those in time  $t'$ .

Therefore, we can consider an agent who decides at different times as different agents. An agent making decisions in time  $t$  is usually called the *t-agent*.



# Deterministic case with heterogeneous constant discount rates

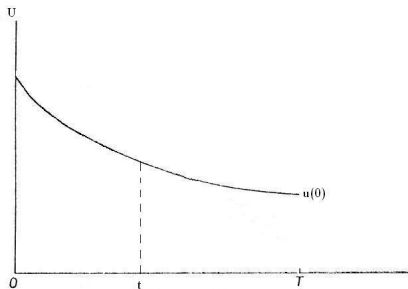
The objective of the agent at time  $t$  (*t-agent*) is:

$$\max_{\{u(s)\}} \int_t^T e^{-\delta(s-t)} L(x(s), u(s), s) ds + e^{-\rho(T-t)} F(x(T)) ,$$
$$\dot{x} = f(x, u, s), \quad x(t) = x_t.$$

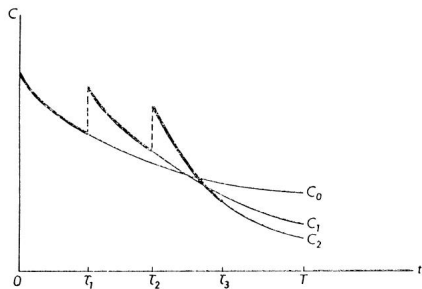
## Agents' strategies:

- Commitment solution
- Naive solution
- Sophisticated solution

**Commitment solution:** The decision maker commits himself not to change the decisions initially taken and solves a standard optimal control problem over the horizon planning  $[0, T]$ .



**Naive solution:** The decision maker takes decisions without taking into account that his preferences will change in the near future. Then, naive  $t$ -agent solves a standard optimal control problem over the horizon planning  $[t, T]$ , but at  $t + \epsilon$  he will change his decision rule by solving again a standard optimal control problem over  $[t + \epsilon, T]$ .



**Sophisticated solution:** The agent recognizes that he is unable to precommit his future behavior beyond the next instant and adopts a strategy of consistent planning by restricting his present behavior to his optimal future behavior.

Then, in order to obtain a time consistent strategy, we must derive the corresponding Dynamic Programming Equation. This can be done in two (essentially equivalent) different ways:

- 1 To discretize the problem, find the corresponding DPE in discrete time, and take the formal continuous time limit. (Karp (JET 2007). Hyperbolic discounting).
- 2 To follow a variational approach (Ekeland and Pirvu (2008), Hyperbolic discounting, or Marín-Solano and Patxot (OCAM 2011), Heterogeneous discounting in a deterministic setting).

# Dynamic programming equation (DPE):the deterministic case in continuous time

Let  $V(x, t)$  be the value function for the sophisticated  $t$ -agent and assume that it is continuously differentiable in  $(x, t)$ . Then  $V(x, t)$  satisfies the dynamic programming equation:

$$\begin{aligned} & \rho V(x, t) + K(x, t) - V_t(x, t) = \\ & = \max_{\{u\}} \{L(x, u, t) + V_x(x, t)f(x, u, t)\} , \\ & V(x, T) = F(x) , \end{aligned} \tag{1}$$

where

$$K(x, t) = \int_t^T e^{-\delta(s-t)} [\delta - \rho] L^*(x, s) ds .$$

and  $L^*(x, s) = L(x, u^*(x, s), s)$ .

Finally, we can now differentiate  $K(x, t)$  with respect to  $t$ , to obtain the “auxiliary DPE”

$$\begin{aligned}\delta K(x, t) - \nabla_t K(x, t) &= (\delta - \rho)U(x, \phi(x, t), t) \\ &+ \nabla_x K(x, t) \cdot f(x, \phi(x, t), t)\end{aligned}$$

together with

$$F(x, T) = 0, \tag{2}$$

so that, we can characterize the time consistent solution as the solution of the system of PDE (1)-(2).

$$J_C(u(\cdot)) = \int_t^T e^{-r_1(s-t)} L_1(x(s), u_1(s), u_2(s), s) ds \\ + \int_t^T e^{-r_2(s-t)} L_2(x(s), u_1(s), u_2(s), s) ds$$

subject to:

$$\dot{x}(s) = f(x(s), u_1(s), u_2(s), s), \quad x(t) = x_t.$$

- **Case 1:**  $L_1 = L_2 = L$  and  $r_1 = r_2 = r$
- **Case 2:**  $L_1 \neq L_2$  and  $r_1 = r_2 = r$
- **Case 3:**  $L_1 = L_2 = L$  and  $r_1 \neq r_2$
- **Case 4:**  $L_1 \neq L_2$  and  $r_1 \neq r_2$

**Cases 1 and 2** can be solved by means of PMP or HJB.

**Case 3** can be solved as a model with non-constant discounting (Karp (2007) or Marín-Solano and Navas (2009)). Note that in this case:

$$J_C(u(\cdot)) = \int_t^T \theta(s-t) L(x(s), u_1(s), u_2(s), s) ds$$

where

$$\theta(s-t) = e^{-r_1(s-t)} + e^{-r_2(s-t)} = e^{-\int_t^s \bar{r}(\tau-t) d\tau}$$

where the instantaneous time preference rate  $\bar{r}$  is a (non-constant) function of its argument:

$$\bar{r}(\tau) = -\frac{\theta'(\tau)}{\theta(\tau)} = \frac{r_1 e^{-r_1 \tau} + r_2 e^{-r_2 \tau}}{e^{-r_1 \tau} + e^{-r_2 \tau}}$$



Finally, **Case 4** can be transformed into a problem with non-homogeneous discounting:

All optimal control problem can be stated in three different (but equivalent) ways: Functional objective given

- 1 integral form (Lagrange problem)
- 2 integral and terminal value term (Bolza problem)
- 3 only terminal value term (Mayer problem)

Then, the problem

$$\begin{aligned} \max J_C(u(\cdot)) = \max & \int_t^T e^{-r_1(s-t)} L_1(x(s), u_1(s), u_2(s), s) ds \\ & + \int_t^T e^{-r_2(s-t)} L_2(x(s), u_1(s), u_2(s), s) ds \end{aligned}$$

subject to:

$$\dot{x}(s) = f(x(s), u_1(s), u_2(s), s), \quad x(t) = x_t.$$

can be transformed into

$$\max J_C(u(\cdot)) = \max \int_t^T e^{-r_1(s-t)} L_1(x(s), u_1(s), u_2(s), s) ds \\ + e^{-r_2(T-t)} Y(T)$$

subject to:

$$\dot{x}(s) = f(x(s), u_1(s), u_2(s), s), \quad x(t) = x_t. \\ \dot{Y}(s) = r_2 Y(s) + L_2(x(s), u_1(s), u_2(s), s),$$

i.e., we have rewritten the functional objective for one of the players in the Mayer form, and therefore, transformed the cooperative problem into a problem with integral and terminal value term, but with different time preferences rates.

# An exhaustible resource model under common access: the case of two-asymmetric players

Consider the following model of a common property non-renewable resource extraction where the objective for the coalition is to maximize

$$\int_0^T \ln(c_1(s)) e^{-r_1 s} ds + \int_0^T \ln(c_2(s)) e^{-r_2 s} ds$$

subject to

$$\dot{x}(t) = -c_1(t) - c_2(t), \quad x(0) = x_0, \quad x(T) = 0.$$

**Precommitment (at  $t = 0$ ) solution (PMP):**

$$c_m^0(s) = \frac{e^{-r_m s}}{\sum_{i=1}^2 \frac{1 - e^{-r_i T}}{r_i}} x_0 = \frac{e^{-r_m s}}{\sum_{i=1}^2 \frac{e^{-r_i s} - e^{-r_i T}}{r_i}} x_s,$$

The **precommitment solution**, which is optimal from the viewpoint of the 0-coalition (we can associate it with the existence of some binding agreement), is not longer optimal if players in the coalition can recalculate the cooperative solution at some instant  $t \in (0, T]$ . Note that the maximum of

$$\int_t^T \ln(c_1(s)) e^{-r_1(s-t)} ds + \int_t^T \ln(c_2(s)) e^{-r_2(s-t)} ds,$$

subject to

$$\dot{x}(s) = -c_1(s) - c_2(s), \quad x(t) = x_t, \quad x(T) = 0$$

is given by

$$c_m^t(s) = \frac{e^{-r_m(s-t)}}{\sum_{i=1}^2 \frac{1-e^{-r_i(T-t)}}{r_i}} x_t, \quad s \in [t, T].$$

This solution differs from that calculated at  $t = 0$ . For instance,  $c_1^t(t) = c_2^t(t)$ , whereas  $c_1^0(t) \neq c_2^0(t)$  for every  $t > 0$ .

In general, if players in the coalition can continuously re-calculate the “cooperative” solution, they will follow what we call the (time inconsistent) **naive** decision rule  $c_m^N(t)$ . In this case  $c_m^t(s)$  is followed only at the time  $s = t$  at which the agents of the  $t$ -coalition have calculated the extraction rate, so that the actual extraction rate becomes

$$c_m^N(t) = c_m^t(t) = \frac{1}{\sum_{i=1}^2 \frac{1 - e^{-r_i(T-t)}}{r_i}} x_t.$$

► Note that the precommitment and naive solutions do not coincide unless  $r_1 = r_2$ . In fact,  $c_1^P(t) \neq c_2^P(t)$ , for every  $t \in (0, T]$ , whereas  $c_1^N(t) = c_2^N(t)$  for every  $t \in [0, T]$ .

# How to obtain a time-consistent solution?

In order to determine a time-consistent equilibrium, we first reformulate our problem by rewriting the payoff of player 2 in the Mayer form. The objective functional becomes now

$$\int_t^T e^{-r_1(s-t)} \ln(c_1(s)) ds + e^{-r_2(T-t)} y(T)$$

subject to

$$\dot{x}(s) = -c_1(s) - c_2(s), \quad \dot{y}(s) = r_2 y(s) + \ln(c_2(s))$$

with  $x(T) = 0$ .

We can now use the DPE introduced above:

$$\begin{aligned} & r_2 W(x, y, t) + K(x, y, t) - W_t(x, y, t) \\ & = \max_{\{c_1, c_2\}} \{ \ln c_1 + W_x(x, y, t)(-c_1 - c_2) + W_y(x, y, t)(r_2 y + \ln(c_2)) \}, \end{aligned}$$

where  $K(x, y, t) = (r_1 - r_2) \int_t^T e^{-r_1(s-t)} \ln(c_1^*, s) ds$ .

For this particular problem, the solution obtained for the naive coalition is a time-consistent policy. This feature, also arising in non-constant discounting models (see Pollak (1968) and Marín-Solano and Navas (2009)), is a consequence of using logarithmic utility functions, and it no longer holds when more general utility functions are considered.

# An exhaustible resource model under common access: the case of $N$ -asymmetric players

Next, we extend the two-player case analyzed above. Then, consider the case of  $N$  players who decide to form a coalition seeking for a time-consistent solution maximizing

$$J(c(\cdot)) = \sum_{m=1}^N \lambda_m \int_t^T e^{-r_m(s-t)} U^m(x(s), c(s), s) ds \quad (3)$$

subject to

$$\dot{x}(s) = f(x(s), c(s), s), \quad x(t) = x_t. \quad (4)$$

Note that in this case we cannot use the above approach of transforming the Lagrange problem into a Bolza problem!



# DPE for the $N$ player case (I)

In order to solve the  $N$  player case, we discretize (3-4):

$$\max_{\{c_1, \dots, c_n\}} V_j = \sum_{m=1}^N V_j^m = \sum_{i=0}^{n-j-1} \sum_{m=1}^N \lambda_m e^{-r_m(i\epsilon)} U^m(x_{(i+j)}, c_{(i+j)}, (i+j)\epsilon)\epsilon$$

subject to  $x_{i+1} = x_i + f(x_i, c_i, i\epsilon)\epsilon$ ,  $i = j, \dots, n-1$ ,  $x_j$  given .

In this case, the can obtain the following dynamic programming algorithm:

$$V_j^*(x_j, j\epsilon) = \max_{\{c_j\}} \left\{ \sum_{m=1}^N \lambda_m U^m(x_j, c_j, j\epsilon)\epsilon \right. \\ \left. + \sum_{k=1}^{n-j-1} \sum_{m=1}^N \lambda_m (1 - e^{r_m\epsilon}) e^{-r_m k\epsilon} \bar{U}_{(j+k)}^m(x_{(j+k)}, (j+k)\epsilon)\epsilon \right. \\ \left. + V_{(j+1)}^*(x_{(j+1)}, (j+1)\epsilon) \right\},$$

## DPE for the $N$ player case (II)

with  $x_{(j+1)} = x_j + f(x_j, c_j, j)\epsilon$ ,  $j = 0, \dots, n - 1$ , and  $V_n^* = 0$ .

We then define the value function for Problem 3-4 as the solution to the DPE obtained by taking the formal continuous time limit when  $\epsilon \rightarrow 0$  of the DPE obtained from the discrete approximation to the problem, assuming that the limit exists and that the solution is of class  $C^1$  in all their arguments. Proceeding in this way, it can be easily proved that:

If  $W^m(x, t)$ ,  $m = 1, \dots, N$ , is a set of continuously differentiable functions in  $(x, t)$  satisfying the DPE

$$\sum_{m=1}^N r_m W^m(x, t) - \sum_{m=1}^N \nabla_t W^m(x, t) = \max_{\{c\}} \left\{ \sum_{m=1}^N \lambda_m U^m(x, c, t) + \sum_{m=1}^N \nabla_x W^m(x, t) \cdot f(x, c, t) \right\} \quad (5)$$

with  $W^m(x, T) = 0$ , for every  $m = 1, \dots, N$ , and

$$W^m(x, t) = \lambda_m \int_t^T e^{-r_m(s-t)} U(x(s), \phi(x(s), s), s) ds, \quad (6)$$

where,  $c^*(t) = \phi(x(t), t)$  is the maximizer of the right hand term in Equation (5), then  $W(x, t) = \sum_{m=1}^N W^m(x, t)$  is the value function of the whole coalition, the decision rule  $c^* = \phi(x, t)$  is the (time-consistent) MPE, and  $W^m(x, t)$ , for  $m = 1, \dots, N$ , is the value function of player  $m$  in the cooperative problem (3-4).

## Remark

Note that, throughout the equilibrium rule  $c^* = \phi(x, t)$ , for every player  $m$ ,  $W^m(x, t)$  is a solution to the partial differential equation

$$\begin{aligned} & r_m W^m(x, t) - \nabla_t W^m(x, t) \\ &= \lambda_m U^m(x, \phi(x, t), t) + \nabla_x W^m(x, t) \cdot f(x, \phi(x, t), t), \end{aligned} \quad (7)$$

for  $m = 1, \dots, N$ , with  $W^m(x, T) = 0$ . Hence, we can compute the value function by first determining the decision rule solving the right hand term in Eq. (5) as a function of  $\nabla_x W^m(x, t)$ ,  $m = 1, \dots, N$ , and then substituting the decision rule into the system of  $N$  partial differential equations (7).

# An exhaustible resource model under common access: the case of $N$ -asymmetric players

Now we can extend the results for the non-renewable resource model in Section 2 to the general case of  $N$  asymmetric players. If  $\lambda_1 = \dots = \lambda_N = 1$ , we must solve

$$\max_{\{c_1, \dots, c_n\}} \sum_{m=1}^N \int_t^T e^{-r_m(s-t)} \frac{(c_m(s))^{1-\sigma_m} - 1}{1 - \sigma_m} ds \quad (8)$$

subject to

$$\dot{x}(s) = - \sum_{m=1}^N c_m(s), \quad x(t) = x_t, \quad x(T) = 0. \quad (9)$$

## Precommitment and naive solutions

For  $m = 1, \dots, N$ , the precommitment and naive solutions are:

$$c_m^P(t) = \frac{e^{-\gamma_m t}}{\sum_{i=1}^N \frac{1}{\gamma_i} (e^{-\gamma_i t} - e^{-\gamma_i T})} x_t$$

and

$$c_m^N(t) = \frac{1}{\sum_{i=1}^N \frac{1}{\gamma_i} (1 - e^{-\gamma_i (T-t)})} x_t ,$$

respectively, where  $\gamma_m = \frac{r_m}{\sigma_m}$ .

- Note that in the naive case the extraction rates of all agents coincide.

## Time consistent solutions

Now, we have to solve:

$$\begin{aligned} & \sum_{m=1}^N r_m W^m(x, t) - \sum_{m=1}^N \frac{\partial W^m(x, t)}{\partial t} \\ &= \max_{c_1, \dots, c_N} \left\{ \sum_{m=1}^N \frac{c_m(s)^{1-\sigma_m} - 1}{1 - \sigma_m} + \left( \sum_{m=1}^N \frac{\partial W^m(x, t)}{\partial x} \right) \left( - \sum_{m=1}^n c_m(s) \right) \right\}. \end{aligned}$$

where  $c_m^S(t) = \left( \sum_{j=1}^N \frac{\partial W^j(x, t)}{\partial x} \right)^{-\frac{1}{\sigma_m}}$ , for  $m = 1, \dots, N$ .

► The extraction rates of agents  $m$  and  $m'$  coincide ( $c_m^S = c_{m'}^S$ ) if, and only if,  $\sigma_m = \sigma_{m'}$ . Thus, if there are two players  $m$  and  $m'$  such that  $\sigma_m \neq \sigma_{m'}$  (hence  $c_m^S \neq c_{m'}^S$ ), the naive solution is always time-inconsistent.

In order to compute the actual decision rule we can solve the **system of  $N$  coupled partial differential equations**:

$$\begin{aligned}
 r_m W^m(x, t) - \frac{\partial W^m(x, t)}{\partial t} \\
 &= \frac{1}{1 - \sigma_m} \left[ \left( \sum_{j=1}^N \frac{\partial W^j(x, t)}{\partial x} \right)^{\frac{\sigma_m - 1}{\sigma_m}} - 1 \right] \\
 &\quad - \frac{\partial W^m(x, t)}{\partial x} \sum_{j=1}^N \left( \sum_{i=1}^N \frac{\partial W^i(x, t)}{\partial x} \right)^{-\frac{1}{\sigma_j}},
 \end{aligned}$$

for  $m = 1, \dots, N$



In the particular case that  $\sigma_1 = \dots = \sigma_N = \sigma$ , the above system simplifies to

$$r_m W^m(x, t) - \frac{\partial W^m(x, t)}{\partial t} = \frac{1}{1-\sigma} \left[ \left( \sum_{j=1}^N \frac{\partial W^j(x, t)}{\partial x} \right)^{1-\frac{1}{\sigma}} - 1 \right] - N \frac{\partial W^m(x, t)}{\partial x} \left( \sum_{i=1}^N \frac{\partial W^i(x, t)}{\partial x} \right)^{-\frac{1}{\sigma}},$$

$m = 1, \dots, N.$

By guessing  $W^m(x, t) = A^m(t) \frac{x^{1-\sigma}-1}{1-\sigma} + B^m(t)$ ,  $m = 1, \dots, N$ , with  $A^m(t) > 0$  for every  $t \in [0, T)$ , and substituting in the system of DPE, we find that the functions  $A^m(t)$  are the solution to the following system of ordinary differential equations

$$\dot{A}^m - r_m A^m = N(1-\sigma) A^m \left( \sum_{j=1}^N A^j \right)^{-\frac{1}{\sigma}} - \left( \sum_{j=1}^N A^j \right)^{1-\frac{1}{\sigma}},$$

for  $j = 1, \dots, N.$

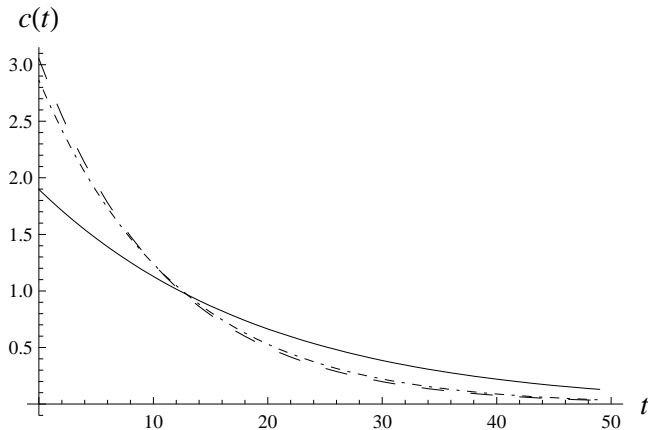
## Some results

- ▶ For the case of logarithmic utility functions ( $\sigma = 1$ ), the naive solution also is time consistent in the case of  $N$  asymmetric players. On the contrary, if  $\sigma \neq 1$ , the naive solution is not longer time consistent.
- ▶ In the time-consistent solution, the extraction rates of two agents coincide if, and only if, they have the same marginal elasticity  $\sigma$ .
- ▶ If  $U^m(c_m) = U(c_m)$ , i.e., all the agents have the same utility function (in the isoelastic case,  $\sigma_1 = \dots = \sigma_N = \sigma$ ), along the equilibrium rule all players extract the resource at the same rate and problem becomes equivalent to the problem of a representative agent using the discount function  $\sum_{m=1}^N e^{-r_m(s-t)}$  (this is the case of non-constant discounting!). On the contrary, if there two agents  $m$  and  $m'$  with different marginal utilities ( $\sigma_m \neq \sigma_{m'}$ ), the problem cannot be simplified to a non-constant discounting problem.

We consider as a baseline case the following:

- $N = 3$
- Time preference rates:  $r_1 = 0.03$ ,  $r_2 = 0.06$  and  $r_3 = 0.09$ .
- Initial stock of the resource of  $x_0 = 100$
- Time horizon from  $t_0 = 0$  to  $T = 50$  periods.
- Utilities from consumption are assumed to be of the iso-elastic type with equal intertemporal elasticity of substitution ( $1/\sigma$ ) for all three players in the coalition.

Next two Figures show the individual extraction rate for every agent in the coalition under the assumption of cooperation for the **naive** (dot dashed line) and the **sophisticated** solutions (dashed line), with  $\sigma = 0.6$  (Figure 1) and  $\sigma = 2$  (Figure 2). In both graphs, the solid line shows the extraction rate for logarithmic utilities.



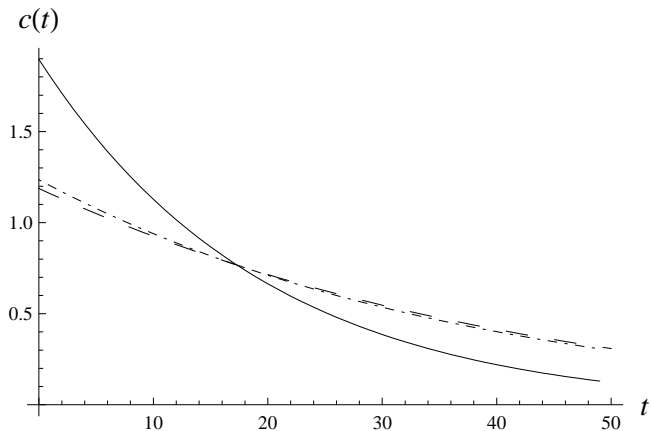
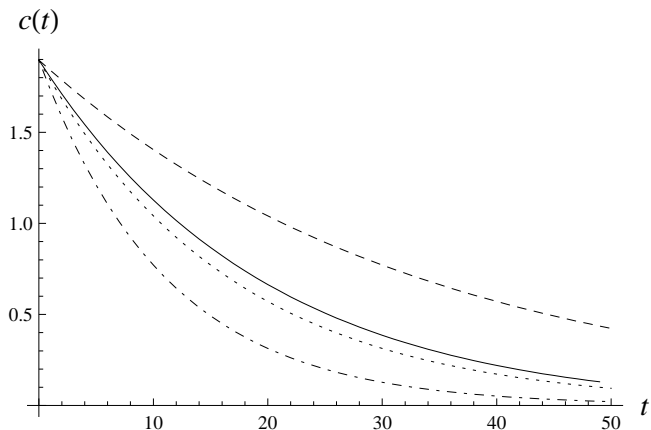


Figure: Extraction rates for naive and sophisticated agents ( $\sigma = 2$ ) and logarithmic case.

## Some comments

- Unless  $\sigma = 1$  (logarithmic utilities), the time-consistent and naive solutions do not coincide, as expected. For  $\sigma = 0.6$ , the time-consistent agents' extraction rate is higher at initial periods compared with naive agents, this behavior being reversed for  $\sigma = 2$ .
- It can be observed that the equilibrium appears to be more sensitive to the value of  $\sigma$  than to the behavior (naive or time-consistent) of the  $t$ -coalitions. In addition, higher values of  $\sigma$  lead agents to smooth their extraction rate path along the time horizon.

## Precommitment vs. sophisticated solution



**Figure:** Extraction rates for sophisticated agents in the coalition (solid line) and individual extraction rates under precommitment at  $t = 0$  (dashed, dotted and dot dashed lines correspond to players 1, 2 and 3, respectively). Logarithmic utility.

# An extension: infinite planning horizon

Let's consider now the problem of

$$\max_{\{c_1, \dots, c_n\}} \sum_{m=1}^N \int_t^{\infty} e^{-r_m(s-t)} \frac{(c_m(s))^{1-\sigma_m} - 1}{1 - \sigma_m} ds ,$$

subject to

$$\dot{x}(s) = g(x) - \sum_{m=1}^N c_m(s) , \quad x(t) = x_t ,$$

where  $c_m(t)$  is the harvest rate of agent  $m$ , for  $m = 1, \dots, N$ , and  $g(x)$  is the natural growth function of the resource stock  $x$ .



In the case that both utility functions and state equation are autonomous, we concentrate in the case of state dependent value functions where the DPE is:

$$\sum_{m=1}^N r_m W^m = \max_{c_1, \dots, c_N} \left\{ \sum_{m=1}^N \frac{c_m^{1-\sigma_m} - 1}{1 - \sigma_m} + \left( \sum_{j=1}^N W_x^j \right) \left( g(x) - \sum_{m=1}^N c_m \right) \right\},$$

hence

$$c_m^* = \phi_m(x) = \left( \sum_{j=1}^N W_x^j \right)^{-\frac{1}{\sigma_m}}.$$

Note that:

- 1 Therefore,  $c_m^* = c_{m'}^*$  if, and only if,  $\sigma_m = \sigma_{m'}$
- 2 In this model, in general, along the equilibrium rule, marginal utilities coincide, i.e.,  $U'(c_m^*) = U'(c_{m'}^*) = \sum_{j=1}^N W_x^j$ , for all  $m \neq m'$ .

As in the finite horizon case, now we also have the set of DPEs

$$r_m W^m = \frac{(\phi_m(x))^{1-\sigma_m} - 1}{1 - \sigma_m} + W_x^m \left( g(x) - \sum_{j=1}^N (\phi_j(x)) \right),$$

for all  $m = 1, \dots, N$ , where  $\phi_m(x)$  are given by the expression above.

Next, let us restrict our attention to the case of linear decision rules.

► Since  $(c_i^*)^{-\sigma_i} = (c_j^*)^{-\sigma_j}$ , for all  $i, j = 1, \dots, N$ , if  $c_m^* = \phi_m(x) = \alpha_m x$  then  $(\alpha_i x)^{-\sigma_i} = (\alpha_j x)^{-\sigma_j}$ . Therefore, no linear decision rules exist unless  $\sigma_i = \sigma_j$ , for all  $i, j$ .

► For  $\sigma_i = \sigma_j = \sigma$ , then  $\alpha_i = \alpha_j$  and the DPE becomes  $\sum_{m=1}^N r_m W^m = \frac{N}{1-\sigma} (\alpha^{1-\sigma} x^{1-\sigma} - 1) + \alpha^{-\sigma} x^{-\sigma} (g(x) - N\alpha x)$ . This equation has a solution if  $g(x) = ax$ . In this case, we obtain

$$\sum_{m=1}^N r_m W^m(x) = \left[ \frac{N\sigma}{1-\sigma} \alpha^{1-\sigma} + a\alpha^{-\sigma} \right] x^{1-\sigma} - \frac{N}{1-\sigma},$$

together with  $\sum_{m=1}^N W_x^m(x) = \alpha^{-\sigma} x^{-\sigma}$ .

If we try  $W^m(x) = A^m \frac{x^{1-\sigma}-1}{1-\sigma} + B^m$ , by simplifying we obtain that  $A^m$ ,  $B^m$  and  $\alpha$  are obtained by solving the equation system

$$[r_m - (1 - \sigma)(a - N\alpha)] A^m = \alpha^{1-\sigma},$$

$$r_m A^m - (1 - \sigma)r_m B^m = 1 \quad \text{and} \quad \sum_{m=1}^N A^m = \alpha^{-\sigma}.$$

► In the case of logarithmic utilities (corresponding to the limit  $\sigma = 1$ ), by trying  $W^m(x) = A^m \ln x + B^m$ , we can reproduce the calculations to obtain  $A^m = \frac{1}{r_m}$  and  $\alpha = \frac{1}{\sum_{m=1}^N \frac{1}{r_m}}$ . If  $r_1 = \dots = r_N = r$  then

# Main results:

- 1 In the infinite case problem, the extraction rates of two agents are equal if, and only if, they have the same marginal elasticity (equal  $\sigma$ ). Note that agents with different discount rates harvest the resource at equal rates. This solution is different from that obtained in a noncooperative setting, or from that obtained by applying the PMP (the precommitment solution).
- 2 Since  $c_i^{\sigma_i} = c_j^{\sigma_j}$ , for every  $i, j = 1, \dots, N$ , extraction / harvesting rates are higher for agents with a higher intertemporal elasticity of substitution (lower value of the parameter  $\sigma$ ) when  $c_i, c_j > 1$ . This property is reversed when  $c_i, c_j < 1$ . Note that this property is independent on the use of different discount rates (although discount rates affect to the value of extraction / harvesting rates).

- 1 If there are two players with different marginal elasticities, **no linear decision rules exist**. This property is independent on the use of different discount rates. As a consequence, in the case of different marginal elasticities, it becomes very difficult to derive analytic solutions.
- 2 If the natural growth function is linear and all the agents have the same marginal elasticity  $\sigma$ , then the decision rules  $c_m = \alpha x$  and the value functions  $W^m(x) = A^m \frac{x^{1-\sigma} - 1}{1-\sigma} + B^m$ ,  $m = 1, \dots, N$  solve our infinite time horizon problem.

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