

Optimal control and stories we tell about climate change economics

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Workshop on Modern Applications of Control Theory and Reinforcement Learning, Centrum Wiskunde & Informatica (CWI), 20-21 May 2025

Key points to be discussed in the lecture

- Formalism of the optimal control theory
- Key principles of the economics of climate change (at the global level): cost-benefit analysis, time discounting, and optimization over long time horizons
- How assumptions shape the story we tell using models, how assumptions are chosen to construct a particular story (in several examples)
- Theoretical methods to solve optimal control problems, when they can deliver an analytical solutions, may allow to craft stronger stories than numerical simulations^{*}
 *This is potentially changing with the development of ML/AI
- The specific examples presented in this talk are intended solely to illustrate general points. There is no need to fully understand each example in detail.

Optimal control: Key facts



Definition:

Optimal control is the process of determining a control (policy) for a dynamic system such that a certain performance criterion (objective function) is optimized—typically minimized cost or maximized utility—over a given time horizon.

Core Components:

- **1. State variables**: Describe the system's current condition (e.g., position, temperature).
- **2. Control variables**: Inputs or actions that influence the system (e.g., force, investment).
- **3. Dynamics**: Differential equations that govern how the state evolves over time.
- **4. Objective function**: A functional (usually integral over time) that quantifies the cost or benefit of a trajectory and control.



 $u(t) \in U$ geometric constraints on controls

Example

Fuller problem:

$$\int_0^\infty x^2(t) \mathrm{d}t \to \min$$

 $\ddot{x}(t) = u(t)$

 $x(0) = x_0, \dot{x}(0) = v_0$

 $x(\infty) = 0, \dot{x}(\infty) = 0$

 $u(t) \in [-1,1]$

$$\begin{array}{c} 0 \\ x_0 \end{array}$$

Interpretation: The goal is to steer a mechanical object (a point particle) to a given state with a minimal error

Optimal solution: A *bang-bang* control with a *countable number* of switches between +1 and -1 which occur in a progressively faster sequence (chattering phenomenon).

This solution is both non-intuitive and requries using a class of measurable functions as controls



IFAC Proceedings Volumes Volume 1, Issue 1, August 1960, Pages 520-529

Relay control systems optimized for various performance criteria

A.T. Fuller

Journal of Mathematical Sciences, Vol. 114, No. 3, 2003

OPTIMAL CHATTERING FEEDBACK CONTROL

M. I. Zelikin and V. F. Borisov



Optimal control: Key facts

Methods to solve:

1. Pontryagin's Maximum Principle:

A foundational result providing necessary conditions for optimality, it introduces the **Hamiltonian** and **adjoint variables** (costate), analogous to Lagrange multipliers in constrained optimization.

2. Dynamic Programming:

Another key method based on Bellman's Principle of Optimality, which leads to the **Hamilton–Jacobi–Bellman (HJB)** equation—a partial differential equation characterizing the value function of the control problem.

3. Numerical Methods:

Solving optimal control problems often requires discretization and numerical optimization, including:

- \circ Direct methods
- Indirect methods (solving necessary conditions)

Economics of climate change: Key postulates

At the global level:

- Anthropogenic greenhouse gas (GHG) emissions drive global warming, which in turn triggers adverse effects such as sea level rise, extreme weather events, etc. Climate outcomes are deeply uncertain and some damages may be irreversible (e.g., ice sheet collapse, biodiversity loss).
- Inter-temporal trade-offs: Climate change involves long time horizons emissions today affect the climate for centuries.
- Discounting (how we value the future relative to the present) is crucial.
 Small changes in the **social discount rate** drastically affect optimal policy (e.g., how much we invest in mitigation now).

At the national and local level:

 Climate is a global good, which makes climate action prone to free-riding because the benefits of reduced emissions are shared globally while the costs are borne locally—giving each country an incentive to undercontribute, hoping others will act instead.

In this talk, we focus on the global level. An economic perspective is only one possible perspective for tackling climate change



Economics of climate change: Key postulates

 Cost-benefit analysis: At the global level, the economics of climate change views the challenge of reducing GHG emissions as a problem of optimizing social welfare over a time horizon long enough to capture the full impacts (damages) of climate change.

$$\int_0^T e^{-rt} L(t) u(c(t)) dt \to \max$$

- Exponential discounting e^{-rt} ensures time-consistency of optimal solutions
- u(·) stands for the utility of consumption (a concave function ensuring a decreasing return to scale of any additional unit of consumption)
- $c(\cdot)$ is consumption of a 'representative agent'
- $L(\cdot)$ is the population size
- Cost-benefit analysis equalizes the marginal benefit and the marginal cost

$$B(x) - C(x) \rightarrow \max \Rightarrow B'(x) = C'(x)$$





DICE: A flagship model of the economics of climate change



An optimal control problem with six dynamic equations and two controls (consumption rate and abatement rate)

Yale University

EliScholar – A Digital Platform for Scholarly Publishing at Yale

Cowles Foundation Discussion Papers

Cowles Foundation

2-1-1992

The 'Dice' Model: Background and Structure of a Dynamic Integrated Climate-Economy Model of the Economics of Global Warming

William D. Nordhaus

DICE: A flagship model of the economics of climate change



Projected CO₂ Emissions in Different Scenarios Global industrial emissions of CO₂ (Gt / year) 80 Baseline 70 60 50 40 Optimal 30 20 Optimal with low Optimal with maximum temperature increase of 2.5 °C 10 discount rate (Stern discounting) 0 2010 2025 2040 2055 2070 2085 2100 Source: W. D. Nordhaus, NBER Working Paper No. 22933 (2016)

Temperature Change in Different Scenarios





Three examples of IIASA research



Example 1: Reachable sets in DICE model

Within-reach sets

 The Brundtland Report's definition requires that sustainable development must meet "the needs of the present without compromising the ability of future generations to meet their own needs" (1987). Economic dimension

- A within-reach set (reachable set, attainability domain) in year Y comprises all system states, to which the system can move as a result of the application of various feasible controls (policies) between today and year Y.
- After a certain control has been applied between today and year X, the within-reach set typically shrinks.



dimension

Within-reach sets in the DICE model

Attainability Domain at the Year 2010

1000 1200 1400 160 M (billion tons CO2 equivalent, carbon weight)

Attainability Domain at the Year 2100

M (billion tons CO2 equivalent, carbon weight)

Attainability Domain at the Year 2000

1000 1200 1400 16 M (billion tons CO2 equivalent, carbon weight)

Attainability Domain at the Year 2050

M (billion tons CO2 equivalent, carbon weight)

g 1000

n 1200

g 1000

DICE version from: Nordhaus, W.D., Boyer, J., Warming the World — Economic Models of Global Warming, MIT Press, 2001

8 1000



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Interim Report

IR-05-049

Attainability Analysis of the DICE Model

Alexey Smirnov (asmirnov@cs.msu.su)

Κ

Within-reach sets in the DICE model



Application of business-asusual policies leads to shrinkage of the within-reach sets IIASA

• Work in progress

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DICE version from: Nordhaus, W. Projections and Uncertainties about Climate Change in an Era of Minimal Climate Policies. *Am. Econ. J. Econ. Policy* 10, 333–360 (2018).



Example 2: Two agents in DICE

Two-agent version of the DICE model

- Work in progress.
- The model differentiates between capital-owning and wage-dependent households (i.e., two representative households instead of one as in the original DICE).
- Motivation: Empirical evidence suggests that households at different income levels perceive their responsibilities for funding climate change mitigation differently. This observation becomes even more relevant, given the widening of income gap between wealthier and poorer segments of society.
- Capital-owning households possess all capital and receive wages. They invest in capital growth a part of their income and consume the remaining part. Income of wage-dependent households relies on wages only and they consume their full income.

Socially optimal solution (social planner's problem)

Solution optimal for capital owners

 $\int_0^T e^{-rt} L_{CO}(t) u(c_{CO}(t)) dt \to \max$

population of capital owners

population of wage-dependent households

 $\int_0^{\infty} e^{-rt} \left(L_{CO}(t) u(c_{CO}(t)) + L_{WD}(t) u(c_{WD}(t)) \right) dt \to \max$



Two-agent version of the DICE model



RA: regular DICE 2h: social planner's problem CO: capital owners' problem

q: share of capital owners in the economy (between 0 and 1)

- Economic variables are insensitive to capital dispersion in the social planner's problem.
- In the capital owners' problem, higher capital dispersion (q → 1) leads to higher saving rate and economic growth.
- In the social planner's problem, optimal abatement rate is higher with higher capital concentration and low warming is optimal. This is because of their low weight in the social welfare function.
- In the capital owners' problem, optimal abatement is lower than that in the DICE model and temperatures are higher due to free riding.

Two-agent version of the DICE model



RA: regular DICE 2h: social planner's problem CO: capital owners' problem

q: share of capital owners in the economy (between 0 and 1)

- Capital owners face intertemporal 'unfairness' because optimal mitigation requires earlier generations to reduce their consumption, while later generations benefit from significantly higher consumption levels compared to a no-mitigation scenario.
- Wage-dependent households do not face intertemporal 'unfairness' and with higher dispersion of capital enjoy higher consumption gains when capital owners optimize their abatement.



Example 3: Two sectors in an environmental economics model



Journal of Mathematical Economics Volume 98, January 2022, 102554



Optimal transition to greener production in a pro-environmental society

Sergey Orlov Ӓ 🖾, Elena Rovenskaya 🖾

Two-sector model

- $K_G(\cdot)$ and $K_B(\cdot)$ stand for the 'green' and 'brown' capital stocks
- $Y_B(t) = A_B K_B(t)$ and $Y_G(t) = A_G K_G(t)$ are outputs of the 'green' and 'brown' sectors (AK production function)
- Brown capital is more productive than the green capital: $A_B > A_G$
- The total output then becomes

$$Y(t) = A_G K_G(t) + A_B K_B(t)$$

The social planner distributes total output Y(·) between consumption C(·), investment I_G(·) in the green sector, and investment I_B(·) in the brown sector:

$$Y(t) = C(t) + I_G(t) + I_B(t)$$

• We obtain the following dynamic system

 $\dot{K}_{G}(t) = I_{G}(t) - \delta_{G}K_{G}(t), \ K_{G}(0) = K_{G0}$

 $\dot{K}_B(t) = I_B(t) - \delta_B K_B(t), \ K_B(0) = K_{B0}$

where $K_{G0} > 0$ and $K_{B0} > 0$ determine the initial respective capital stocks; parameters $\delta_G > 0$ and $\delta_B > 0$ are the respective

depreciation rates.

Two-sector model

• We assume the following social planner's instantaneous utility function

 $U(t) = \ln C(t) + \omega \ln Y_G(t)$

where

 $\ln C(\cdot)$ is the utility of consumption

 $\ln Y_G(\cdot)$ is an 'amenity value' of green production (e.g., a political will). In this model there are no damages

hence environmental quality is an externality

 $\omega \geq 0$ is a weight coefficient

• We consider the model over an infinite time horizon and introduce the total utility as

$$J = \int_0^\infty e^{-rt} U(t) \, \mathrm{d}t$$

where r > 0 is a discount rate.



Two-sector model

Optimal solution (a theorem with a rigorous proof using Pontryagin's maximum principle)



Notations:

• Let $D = D_B - D_G$ be the difference between augmented productivities of the brown and green capitals respectively; here $D_B = A_B - \delta_B > 0$ and $D_G = A_G - \delta_G > 0$

• Let
$$q = \left(\frac{D}{\omega r} + \frac{D-r}{r}\right)^{-1}$$

Assumptions:

(A1) The augmented productivity of the brown sector exceeds the augmented productivity of the green sector at least by the discount factor: D > r;

(A2) The productivity of the green sector is greater than the discount factor: $A_G > r$.



- The optimal control theory relies on the description of dynamical systems using differential equations with time-varied controls. Because of the time dimension, optimal controls can be non-intuitive even in rather simple models.
- Theoretical methods to solve optimal control problems exist but their application is limited to 'stylized' models.
- Stylized models are a powerful basis to inform stories as they focus on key processes and omit many details (allowing to "see the forests for the trees").
- While climate change mitigation is a very complex matter, scientists have developed numerous rather stylized cost-benefit models to illustrate and analyze key tradeoffs involved. Some of these models have been used to inform real-world policies, e.g., the value of carbon tax.
- When developing new models, it is essential to aim for the simplest model necessary to address the research question—no simpler.

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Thank you for your attention! Questions? Comments?

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