Dynamic Programming and Applications

Deterministic Dynamic Programming in Continuous Time

Lectures 3 – 4

Andreas Schaab

Outline

Part 1: Differential equations

- 1. The continuous time limit
- 2. Ordinary differential equations (ODEs)
- 3. Boundary conditions
- 4. Linear first-order ODEs
- 5. Examples of ODEs in macro
- 6. Application: solving the Solow growth model
- 7. Partial differential equations (PDEs)

Outline

Part 2: Optimization with deterministic dynamics

- 1. Neoclassical growth model in continuous time
- 2. Calculus of variations
- 3. Optimal control theory
- 4. Simple example
- 5. Hamilton-Jacobi-Bellman (HJB) equation
- 6. First-order condition for consumption
- 7. Envelope condition and Euler equation
- 8. Connection between calculus of variations / optimal control and HJBs
- 9. Boundary conditions: no-borrowing in the wealth / capital dimension

Outline

Part 3: Applications

1. XXX

Part 1: differential equations

1. Continuous time limit

Consider the two key difference equations:

$$K_{t+1} = I_t + (1 - \delta)K_t$$

and

$$\frac{1}{C_t} = \beta R_t \frac{1}{C_{t+1}}$$

- On the board: (i) generalized discrete time step Δ and (ii) continuous time limit

2. Ordinary differential equations

Consider the "discrete-time" equation

$$X_{t+\Delta t} - X_t = G(X_t, t, \Delta t)$$

• Continuous-time limit: consider the limit as $\Delta t \rightarrow 0$

$$\dot{X}_t \equiv \frac{dX}{dt} \equiv \lim_{\Delta t \to 0} \frac{X_{t+\Delta t} - X_t}{\Delta t} = \lim_{\Delta t \to 0} \frac{1}{\Delta t} G(X_t, t, \Delta t) \equiv g(X_t, t)$$

- $\dot{X}_t = g(X_t)$ is *autonomous* and dropping subscripts: $\dot{X} = g(X)$
- This is a *first-order (ordinary) differential equation*, second-order equations are:

$$\frac{d^2X_t}{dt^2} = g\left(\frac{dX_t}{dt}, X_t, t\right)$$

• We often consider ODEs in the *time dimension* but ODEs can be defined on any state space (e.g., space dimensions)

3. Boundary conditions

- Boundary conditions are critical for characterizing differential equations
- Consider an ODE on the time interval $t \in [0,1]$. We call [0,1] the *state space*. (0,1) is the *interior of the state space* and $\{0,1\}$ is the *boundary*
- The way to think about it: differential equations are defined on the interior of the state space but not on the boundary
- To characterize the function that satisfies the ODE on the interior on the full state space, we need a set of boundary conditions to also characterize the behavior on the boundary
- Heuristically: we need as many boundary conditions as the order of the differential equation

- Similar to discrete-time difference equations: forward equations have initial conditions, backward equations have terminal conditions
- For ODEs, you will often see the terminology:
 - Initial value problems specify a differential equation for X_t with some initial condition X_0
 - Terminal value problems instead specify X_T
- More broadly: We need sufficient information to characterize the function of interest along the boundary
- Types of boundary conditions: Dirichlet $(X_0 = c)$, von-Neumann $(\frac{dX_0}{dt} = c)$, reflecting boundaries, ...
- Boundary conditions are very important and can be very subtle (especially for PDEs)

4. Linear first-order ODEs

• Consider the equation:

$$\dot{X}(t) = a(t)X(t) + b(t) \tag{1}$$

- If b(t)=0, (1) is a homogeneous equation, if a(t)=a and b(t)=b we say (1) has constant coefficients
- Start with $\dot{X}(t) = aX(t)$, divide by X(t) and integrate with respect to t

$$\int \frac{\dot{X}(t)}{X(t)} dt = \int a dt$$
$$\log X(t) + c_0 = at + c_1$$
$$X(t) = Ce^{at}$$

where $C = e^{c_1 - c_0}$

• Pin down constant *C* by using the boundary condition (we need 1)

- Consider time-varying coefficient with $\dot{X}(t) = a(t)X(t)$ with initial condition $X(0) = \bar{x}$
- Dividing by X(t), integrating, and exponentiating yields

$$X(t) = Ce^{\int_0^t a(s)ds}$$

- Constant of integration again pinned down by boundary condition: $C = \bar{x}$
- Finally, for $\dot{X}(t) = aX(t) + b$, we find

$$X(t) = -\frac{b}{a} + Ce^{at}$$

after using change of variables $Y(t) = X(t) + \frac{b}{a}$

• Many results for systems of linear differential equations: $\dot{\boldsymbol{X}}(t) = \boldsymbol{A}\boldsymbol{X}(t)$

5. Examples of differential equations in macro Capital accumulation:

$$\dot{K}_t = I_t - \delta K_t$$

- We can always map back and forth between DT and CT
- In discrete time with *unit* time steps, $K_{t+1} = I_t + (1 \delta)K_t$
- With arbitrary Δ time step, $K_{t+\Delta} = K_t + \Delta(I_t \delta K_t)$
- Continuous-time limit:

$$K_{t+\Delta} = K_t + \Delta(I_t - \delta K_t)$$

$$\frac{K_{t+\Delta} - K_t}{\Delta} = I_t - \delta K_t$$

$$\dot{K}_t = I_t - \delta K_t$$

- Suppose $\{I_t\}_{t>0}$ exogenously given
- Solving this *inhomogeneous equation*, we use *integrating factor*:

$$\dot{K}_t + \delta K_t = I_t$$

$$e^{\int_0^t \delta ds} \dot{K}_t + e^{\int_0^t \delta ds} \delta K_t = e^{\int_0^t \delta ds} I_t$$

• Notice that $\int_0^t \delta ds = \delta \int_0^t ds = \delta[s]_0^t = \delta(t-0) = \delta t$, so

$$e^{\delta t}\dot{K}_t + e^{\delta t}\delta K_t = e^{\delta t}I_t$$

• We have $e^{\delta t}\dot{K}_t + e^{\delta t}\delta K_t = \frac{d}{dt}(K_t e^{\delta t})$, integrating:

$$K_t e^{\delta t} = \tilde{C} + \int_0^t e^{\delta s} I_s ds$$

 $K_t = C + \int_0^t e^{-\delta(t-s)} I_s ds$

• Integrating constant solves initial condition: $C = K_0$

Wealth dynamics (*very important equation in this course*):

$$\dot{a}_t = r_t a_t + y_t - c_t$$

- r_t is the real rate of return on wealth, y_t is income, and c_t is consumption
- Structure of the equation similar to capital accumulation equation

Consumption Euler equation:

$$\frac{\dot{C}_t}{C_t} = r_t - \rho$$

- The Euler equation typically takes the form of a *backward equation* and comes with a terminal condition (C_T) or transversality condition $(\lim_{T\to\infty} C_T)$
- Stationary point only if $r_t = \rho$
- Suppose we are at $r_t = r = \rho$ and a shock is realized. $r_0 > r$ what happens? $r_0 < r$ what happens?

6. Example: Solow growth model

• As before, $Y_t = C_t + I_t$ and

$$\dot{K}_t = Y_t - C_t - \delta K_t$$

Representative firms operates neoclassical production function

$$Y_t = F(K_t, L_t, A_t)$$

- Normalize labor to $L_t = 1$ and hold TFP constant $A_t = A$
- We again assume constant savings rate: $Y_t C_t = I_t = sY_t$
- Assume Cobb-Douglas $Y_t = AK_t^{\alpha}$ so equilibrium allocation

$$\dot{K}_t = sAK_t^{\alpha} - \delta K_t$$

· Steady state is given by

$$K_{ss} = \left(\frac{sA}{\delta}\right)^{\frac{1}{1-\alpha}}$$

- Key equilibrium condition in \dot{K}_t is non-linear how to proceed?
- Let $X_t = K_t^{1-\alpha}$, then

$$\begin{aligned} \dot{X}_t &= (1 - \alpha) K_t^{-\alpha} \dot{K}_t \\ &= (1 - \alpha) K_t^{-\alpha} (sAK_t^{\alpha} - \delta K_t) \\ &= (1 - \alpha) sA - (1 - \alpha) K_t^{1-\alpha} \delta \\ &= (1 - \alpha) sA - (1 - \alpha) \delta X_t \end{aligned}$$

• Solution with initial condition X_0 (work this out):

$$X_t = X_{ss} + e^{-(1-lpha)\delta t} igg[X_0 - X_{ss} igg]$$
 , where $X_{ss} = rac{sA}{\delta}$

Transition dynamics (rate of convergence) governed by $-(1-\alpha)\delta$

7. What are partial differential equations?

- Partial differential equations (PDEs) generalize ODEs to higher-dimensional state spaces
- PDEs are at the heart of (i) continuous-time dynamic programming and (ii) heterogeneous-agent models in macro
- PDEs have long been a core tool in physics, applied math, ...
 - ⇒ increasingly used in economics

- Consider a function $u(x_1, x_2, ..., x_n)$ where $x_1, ..., x_n$ are coordinates in \mathbb{R}^n
- Partial derivatives of $u(\cdot)$

$$\frac{\partial u}{\partial x_i} \equiv \partial_{x_i} u$$
 and $\frac{\partial^2 u}{\partial x_i \partial x_j} = \partial_{x_i x_j} u$

• A PDE is an equation in u and its partial derivatives — fully generally:

$$0 = G(u, \partial_{x_1}u, \ldots, \partial_{x_n}u, \partial_{x_1x_1}u, \ldots)$$

- The *order* of the PDE, is the order of the highest partial derivative
- Examples from physics
 - Heat equation: $\partial_t u = \partial_{xx} u$ (second-order, linear, homogeneous)
 - Wave equation: $\partial_{tt}u = \partial_{xx}u$ (second-order, linear, homogeneous)
 - Transport equation: $\partial_t u = \partial_x u$ (first-order, linear, homogeneous)
- Income distribution "solves heat equation", wealth dynamics "solve transport equations", dynamic programming often transport + heat

Part 2: optimization with deterministic dynamics

1. Neoclassical growth model in continuous time

• The lifetime value of the representative household is

$$v(k_0) = \max_{\{c_t\}_{t \ge 0}} \int_0^\infty e^{-\rho t} u(c_t) dt$$

subject to

$$\dot{k}_t = F(k_t) - \delta k_t - c_t$$
 k_0 given ,

where $\dot{x}_t = \frac{d}{dt}x_t$, ρ is the discount rate, c_t is the rate of consumption, $u(\cdot)$ is instantaneous utility flow, and \dot{k}_t is the rate of (net) capital accumulation

- · No uncertainty for now
- This is the sequence problem in continuous time

2. Calculus of variations

- Resources:
 - LeVeque: Finite Difference Methods for Ordinary and Partial Differential Equations
 - Kamien and Schwartz: Dynamic Optimization
 - Gelfand and Fomin: Calculus of Variations
- This dynamic optimization problem is associated with the Lagrangian

$$L = \int_0^\infty e^{-\rho t} \left[u(c_t) + \mu_t \left(F(k_t) - \delta k_t - c_t - \dot{k}_t \right) \right] dt$$

- μ_t is the Lagrange multiplier on the capital accumulation ODE
- What do we do with k_t ??

· Integrate by parts:

$$\int_0^\infty e^{-\rho t} \mu_t \dot{k}_t dt = e^{-\rho t} \mu_t k_t \Big|_0^\infty - \int_0^\infty \frac{d}{dt} \left(e^{-\rho t} \mu_t \right) k_t dt$$
$$= -\mu_0 k_0 + \int_0^\infty e^{-\rho t} \rho \mu_t k_t dt - \int_0^\infty e^{-\rho t} \dot{\mu}_t k_t dt$$

Plugging into Lagrangian:

$$L = \int_0^\infty e^{-\rho t} \left[u(c_t) + \mu_t \left(F(k_t) - \delta k_t - c_t \right) - \rho \mu_t k_t + \dot{\mu}_t k_t \right] dt + \mu_0 k_0$$

- · What have we accomplished?
- Notice $\mu_0 k_0$, this is crucial. What's intuition?

$$L = \int_0^\infty e^{-\rho t} \left[u(c_t) + \mu_t \left(F(k_t) - \delta k_t - c_t \right) - \rho \mu_t k_t + \dot{\mu}_t k_t \right] dt + \mu_0 k_0$$

- The planner optimizes over paths $\{c_t\}$ and $\{k_t\}$
- At an optimum, there cannot be any small perturbation in these paths that the planner finds preferable
- Let $\{c_t\}$ and $\{k_t\}$ be *candidate* optimal paths. Consider $\hat{c}_t = c_t + \alpha h_t^c$ and $\hat{k}_t = k_t + \alpha h_t^k$ for arbitrary functions h_t^c and h_t^k

$$L(\alpha) = \int_0^\infty e^{-\rho t} \left[u(c_t + \alpha h_t^c) + \mu_t \left(F(k_t + \alpha h_t^k) - \delta k_t - \delta \alpha h_t^k - c_t - \alpha h_t^c \right) \right.$$
$$\left. - \rho \mu_t (k_t + \alpha h_t^k) + \dot{\mu}_t (k_t + \alpha h_t^k) \right] dt + \mu_0 (k_0 + \alpha h_0^k)$$

• What about *boundary conditions*? At t = 0, capital stock is fixed (k_0 given) while consumption is free. So must have: $h_0^k = 0$ while h_0^c is free

Necessary condition for optimality: $\frac{d}{d\alpha}L(0) = 0$

$$L(\alpha) = \int_0^\infty e^{-\rho t} \left[u(c_t + \alpha h_t^c) + \mu_t \left(F(k_t + \alpha h_t^k) - \delta k_t - \delta \alpha h_t^k - c_t - \alpha h_t^c \right) - \rho \mu_t (k_t + \alpha h_t^k) + \dot{\mu}_t (k_t + \alpha h_t^k) \right] dt + \mu_0 (k_0 + \alpha h_0^k)$$

Work this out yourselves (many times, in many applications!)

$$\frac{d}{d\alpha}L(0) = \int_0^\infty e^{-\rho t} \left[u'(c_t)h_t^c + \mu_t \left(F'(k_t)h_t^k - \delta h_t^k - h_t^c \right) - \rho \mu_t h_t^k + \dot{\mu}_t h_t^k \right] dt + \mu_0 h_0^k$$

where $h_0^k = 0$ because k_0 is fixed

Group terms:

$$\frac{d}{d\alpha}L(0) = \int_0^\infty e^{-\rho t} \left[\left(u'(c_t) - \mu_t \right) h_t^c + \left(\mu_t \left(F'(k_t) - \delta \right) - \rho \mu_t + \dot{\mu}_t \right) h_t^k \right] dt$$

Fundamental Theorem of the Calculus of Variations: Since h_t^c and h_t^k were arbitrary, we must have *pointwise*

$$0 = u'(c_t) - \mu_t$$

$$0 = \mu_t \Big(F'(k_t) - \delta \Big) - \rho \mu_t + \dot{\mu}_t$$

Proposition. (Euler equation for marginal utility)

$$\frac{\dot{\mu}_t}{\mu_t} = \frac{\dot{u}_{c,t}}{u_{c,t}} = \rho - F'(k_t) + \delta = \rho - r_t$$

- We have now solved the neoclassical growth model in continuous time. Its solution is given by a system of two ODEs.
- Suppose $u(c) = \log(c)$ and $F(k) = k^{\alpha}$, then:

$$\frac{\dot{c}_t}{c_t} = \alpha k_t^{\alpha - 1} - \delta - \rho
\dot{k}_t = k_t^{\alpha} - \delta k_t - c_t$$

with k_0 given

- Derive the consumption Euler equation yourselves!
- What are the boundary conditions? (Always ask about BCs!)
 - Initial condition on capital: k_0 given
 - Terminal condition on consumption : $\lim_{T \to \infty} c_T = c_{ss}$

3. Optimal control theory

- · Optimal control theory emerged from the calculus of variations
- Applies to dynamic optimization problems in continuous time that feature (ordinary) differential equations as constraints
- Again the neoclassical growth model:

$$v(k_0) = \max_{\{c_t\}_{t \ge 0}} \int_0^\infty e^{-\rho t} u(c_t) dt$$

subject to

$$\dot{k}_t = F(k_t) - \delta k_t - c_t$$
, k_0 given

- Three new terms:
 - State variable: k_t
 - Control variable: c_t
 - Hamiltonian: $H(c_t, k_t, \mu_t) = u(c_t) + \mu_t [F(k_t) \delta k_t c_t]$

- With Hamiltonian in hand, *copy-paste* formula that we can always use:
 - Optimality condition: $\frac{\partial}{\partial c}H=0$
 - Multiplier condition: $\rho \mu_t \dot{\mu}_t = \frac{\partial}{\partial k} H$
 - State condition: $\dot{k}_t = \frac{\partial}{\partial \mu} H$
- This gives us the same equations that we derived using calc of variations:

$$u'(c_t) = \mu_t$$

$$\rho \mu_t - \dot{\mu}_t = \mu_t (F'(k_t) - \delta)$$

$$\dot{k}_t = F(k_t) - \delta k_t - c_t$$

• We again get system of Euler equation and capital accumulation:

$$\dot{c}_t = \frac{u'(c_t)}{u''(c_t)} \Big(\rho - F'(k_t) + \delta \Big)$$

$$\dot{k}_t = F(k_t) - \delta k_t - c_t$$

4. Simple example [skip]

- Credit: Kamien-Schwartz p. 129
- Simple problem: not much intuition, but illustrates mechanics

$$\max \int_0^1 (x+u)dt$$

subject to $\dot{x} = 1 - u^2$ and initial condition $x_0 = 1$

- Step 1: form Hamiltonian $H(t, x, u, \lambda) = x + u + \lambda(1 u^2)$
- Step 2: necessary conditions (note: no discounting here)

$$0 = H_u = 1 - 2\lambda u$$
$$-\dot{\lambda} = H_x = 1$$

and terminal condition $\lambda_1 = 0$ (because u_1 is *free*)

• Step 3: manipulate necessary conditions:

$$\lambda = 1 - t$$
$$u = \frac{1}{2\lambda}$$

and therefore: $u = \frac{1}{2}(1-t)$

• Finally: solve for all paths (control, state, multiplier)

$$x_t = t - \frac{1}{4}(1 - t) + \frac{5}{4}$$

 $\lambda_t = 1 - t$
 $u_t = \frac{1}{2}(1 - t)$

5. Hamilton-Jacobi-Bellman equation

· Recall the neoclassical growth model in continuous time

$$v(k_0) = \max_{\{c_t\}_{t \ge 0}} \int_0^\infty e^{-\rho t} u(c_t) dt$$

subject to

$$\dot{k}_t = F(k_t) - \delta k_t - c_t \ k_0$$
 given ,

where $\dot{x}_t = \frac{d}{dt}x_t$, ρ is the discount rate, c_t is the rate of consumption, $u(\cdot)$ is instantaneous utility flow, and \dot{k}_t is the rate of (net) capital accumulation

- · No uncertainty for now
- This is the infinite-horizon sequence problem, $t \in [0, \infty)$
- A function $v(\cdot)$ that solves this problem is a solution to the neoclassical growth model

- We will now work towards a recursive representation (good reference: Stokey textbook)
- The discrete-time Bellman equation would be

$$v(k_t) = \max_{c} \left\{ u(c)\Delta t + \frac{1}{1 + \rho \Delta t} v(k_{t+\Delta}) \right\}$$

where $eta = rac{1}{1 +
ho \Delta t}$

$$egin{aligned} (1 +
ho \Delta t) v(k_t) &= \max_c \left\{ (1 +
ho \Delta t) u(c) \Delta t + v(k_{t+\Delta})
ight\} \
ho \Delta t v(k_t) &= \max_c \left\{ u(c) \Delta t + v(k_{t+\Delta}) - v(k_t)
ight\} \end{aligned}$$

$$ho v(k_t) = \max_c \left\{ u(c) + rac{v(k_{t+\Delta}) - v(k_t)}{\Delta t}
ight\}$$

• Finally: take limit $\Delta t \to 0$ and drop t subscripts

• Next: multiply by $1 + \rho \Delta t$ and note that $(\Delta t)^2 \approx 0$

$$\rho v(k) = \max_{c} \left\{ u(c) + dv \right\}$$

- We want to express dv in terms of $v'(\cdot)$ and dk
- Different ways to think about this: chain rule, Ito's lemma (though no uncertainty here), generator
- Recall generator of (stochastic) process dk_t : For any $f(\cdot)$

$$\mathcal{A}f(k_t) = \lim_{\Delta t \to 0} \mathbb{E}_t \frac{f(k_{t+\Delta t}) - f(k_t)}{\Delta t}$$

• For simple ODE (no uncertainty) $dk = (F(k) - \delta k - c)dt$, we have

$$\mathcal{A}f(k) = (F(k) - \delta k - c)f'(k)$$

• Therefore, we arrive at the **Hamilton-Jacobi-Bellman equation**:

$$\rho v(k) = \max_{c} \left\{ u(c) + \left(F(k) - \delta k - c \right) v'(k) \right\}$$

• Notice: We conjectured a stationary value function (what does this mean?)

6. First-order condition for consumption

HJB still has "max" operator:

$$\rho v(k) = \max_{c} \left\{ u(c) + \left(F(k) - \delta k - c \right) v'(k) \right\}$$

- To get rid of this, we have to resolve optimal consumption choice
- First-order condition:

$$u'(c(k)) = v'(k)$$

- This defines the consumption policy function
- We can now plug back in, obtaining an ODE in v'(k)

$$\rho v(k) = u(c(k)) + \left(F(k) - \delta k - c(k)\right)v'(k)$$

 Why is this a "stationary" value function and ODE? What would a time-dependent ODE look like? When would we get one?

7. Envelope condition and Euler equation

- · We now derive the Euler equation in continuous time
- We start with the HJB envelope condition. Differentiating in k:

$$\rho v'(k) = u'(c(k))c'(k) + \left(F'(k) - \delta - c'(k)\right)v'(k) + \left(F(k) - \delta k - c(k)\right)v''(k)$$

$$\rho v'(k) = \left(\underbrace{F'(k) - \delta}_{\text{interest rate }r}\right)v'(k) + \left(F(k) - \delta k - c(k)\right)v''(k)$$

$$(\rho - r)v'(k) = \left(F(k) - \delta k - c(k)\right)v''(k)$$

• Next, we characterize process dv'(k). Using Ito's lemma (even though no uncertainty):

$$dv'(k) = v''(k)dk$$

= $v''(k)(F(k) - \delta k - c(k))dt$
= $(\rho - r)v'(k)dt$.

- Recall first-order condition u'(c(k)) = v'(k).
- The Euler equation for marginal utility is given by

$$\frac{du'(c)}{u'(c)} = (\rho - r)dt.$$

• To go from marginal utility to consumption, we use CRRA utility: $u(c) = \frac{1}{1-\gamma}c^{1-\gamma}$. $u'(c) = c^{-\gamma}$ is a function of *process* c, so by Ito's lemma:

$$du'(c) = -\gamma c^{-\gamma - 1} dc$$
$$= -\gamma u'(c) \frac{dc}{c}$$

Plugging in yields Euler equation for consumption in continuous time:

$$\frac{dc}{c} = \frac{r - \rho}{\gamma} dt$$

or (you'll often see this notation when no uncertainty): $\frac{\dot{c}}{c}=\frac{r-\rho}{\gamma}$

Connection between calculus of variations and HJB:

- What is the connection between costate / multiplier μ_t and marginal value of wealth V'(k)?
- What is the connection between multiplier equation and envelope condition?

9. Boundary conditions

- This is really important: everything we have done so far is only valid in the interior of the state space
- What's the state space of a model?
- For the neoclassical growth model without uncertainty, state space is $k \in [0, \infty)$, or

$$\mathcal{X} = \left\{ k \mid k \in [0, \bar{k}] \right\}$$

where we impose an upper boundary \bar{k}

- This is like the domain of the function v(k) that will be valid
- We say $\partial \mathcal{X}=\{0,\bar{k}\}$ is the **boundary** of the state space and $\mathcal{X}\setminus\partial\mathcal{X}=(0,\bar{k})$ is the **interior**
- As is the case for all differential equations, the HJB holds on the interior and we need boundary conditions to characterize v(k) along the boundary

- What kind of differential equation is the HJB in this model?
- So how many boundary conditions do we need?
- In terms of the economics, what is the correct boundary condition? I.e., what is the correct economic behavior at the boundary $k \in \{0, \bar{k}\}$?
- Nice intuition: 2 boundary inequalities do same job as 1 boundary equality
- We want households to not leave the state space, so we impose that they do not dissave / borrow as $k\to 0$ and save as $k\to \bar k$
- This implies: (why?)

by the boundary conditions

$$u'(c(0)) \ge v'(0)$$

$$u'(c(\bar{k})) < v'(\bar{k})$$

• If households ever hit the boundaries (in the neoclassical growth model, this doesn't really happen), then consumption behavior is no longer determined by the Euler equations but rather

Part 3: Applications

XX

5. Example: tax competition

- Two countries, $i \in \{A, B\}$, setting corporate tax rates τ_t^i on firms operating / headquartered in country i
- Mass of multinational firms j, with μ_t denoting % in country A at time t
- Firms relocate activity / headquarters at rate θ towards low-tax country:

$$d\mu_t = \theta \mu_t (\tau_t^B - \tau_t^A)^{\gamma} dt$$

- Country A maximizes tax revenue: $\max \int_0^\infty e^{-\rho t} \tau_t^A \mu_t dt$. Countries compete over taxes $\{\tau_{it}\}$
- Dynamic Nash: country A sets τ_t^A as best response taking τ_t^B as given
- Recursive representation: the only state variable is μ_t

$$\rho V^{A}(\mu) = \max_{\tau^{A}} \left\{ \tau^{A} \mu + \theta \mu \left(\tau^{B}(\mu) - \tau^{A} \right)^{\gamma} \partial_{\mu} V^{A}(\mu) \right\}$$

Best response strategies: $0 = \mu + \gamma \theta \mu (\tau^B(\mu) - \tau^A)^{\gamma-1} V_\mu^A(\mu)$