## Lecture

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#### References

#### THEORY

- Pham (2009) Continuous-time Stochastic Control maybe too finance
- Lecture notes
  - → Caldentey (????) Stochastic processes and optimal control nice lecture notes Enio uses
    them
  - → Ross (????) Stochastic Control in Continuous Time alternative to math books Fleming and Soner (2006), Øksendal (2003), Øksendal and Sulem (2007)

## THEORY (MACRO)

- Moll's website
- Bayer and Wälde (2015) recent discovery, dicuss the kind of SDE driven by a Markov chain
   Sennewald (2007) (theory paper), Walde (2008) (book on intertemporal optimization),
- Stokey (2009) book Impulse control Problem

#### Numeric

- Achdou, Han, Lasry, Lions, and Moll (2016) (mainly the numerical appendix), Moll's website (tons
  of examples and materials)
- Forsyth and Vetzal (2012) (Also has some slides) good introduction to "viscosity solutions"
- Interested? Check applications . . .
  - → HANK by Kaplan, Moll, and Violante (2016), PHACT
  - → Nuño and Moll (2017) (improved notation)
  - → Thomas and Nuño (2016) (impulse control)

# Consumption Savings Problem

#### **Problem of Household**

$$\begin{aligned} \max_{\{c_t\}_{t\geq 0}} \mathbb{E}_0 & \int_0^\infty e^{-\rho t} u(c_t) dt \\ & \text{S.t } da_t = \left\{ ra_t + z_t - c_t \right\} dt \\ & z_t \text{: is a ct markov chain on } \left\{ b, w \right\} \text{ with intensities } \lambda_1, \lambda_2 \\ & dz_t = (w-b) dq - (w-b) dQ, \quad q \sim \text{Poisson}(\lambda_1), \ Q \sim \text{Poisson}(\lambda_2) \\ & a_t \geq \underline{a} \end{aligned}$$

Individuals' consumption and saving decision is summarized by HJB equation

$$\rho v(a, z_k) = \max_{c} \left\{ u(c) + v_a(a, z) [ra + z_k - c] \right\} + \lambda_k \left[ v(a, z_{-k}) - v(a, z_k) \right]$$
(1)

Where this came from? Check Lagos lecture notes for an heuristic argument.

Theoretical results analogous to discrete time:

- Value function satisfy the HJB equation
- Verification theorems: solution of HJB  $+ \ldots \rightarrow$  value function
- Alternatively, one can show HJB has a unique "nice" solution which is the value function (viscosity solution)

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Before solving the HJB FE let's see what we can do. Analytical results from Bayer and Wälde (2015)

#### Envelope condition:

$$\rho V_{a}(a,b) = rV_{a}(a,b) + V_{aa}(a,b) \big\{ ra + b - c(a,b) \big\} + \lambda_{1} \Big[ V_{a}(a,w) - V_{a}(a,b) \Big]$$

Differential of  $V_a(a, z)$  — CVF, "Itô formula"

$$da_t = \{ra_t + z_t - c_t\}dt$$
  
 $dz_t = (w - b)dq - (w - b)dQ, \quad q \sim \text{Poisson}(\lambda_1), \quad Q \sim \text{Poisson}(\lambda_2)$ 

$$\mathrm{d} V_{\mathsf{a}}(a,b) = \underbrace{V_{\mathsf{aa}}\big\{\mathit{ra} + b - \mathit{c}(a,b)\big\}}_{\mathsf{normal term}} \mathrm{d} t + \underbrace{\left[V_{\mathsf{a}}(a,w) - V_{\mathsf{a}}(a,b)\right]}_{\mathsf{jump terms}} \mathrm{d} q_t$$

From optimization  $V_a(a,z)=u'\left(c(a,z)\right)$ . Combining both equations to get rid of  $V_{aa}$  we have

$$du'(c(a,b)) = \left\{ (\rho - r)u'(c(a,b)) - \lambda_1 u'(c(a,b)) \left[ \frac{u'(c(a,w))}{u'(c(a,b))} - 1 \right] \right\} dt + \left[ u'(c(a,w)) - u'(c(a,b)) \right] dq_t$$

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Applying "Itô lemma" to get consumption over time

$$dc(a,b) = \frac{u'(c(a,b))}{-u''(c(a,b))} \left\{ r - \rho - \lambda_1 \left[ 1 - \frac{u'(c(a,w))}{u'(c(a,b))} \right] \right\} dt + \left[ c(a,w) - c(a,b) \right] dq_t \quad (2)$$

$$dc(a, w) = \frac{u'(c(a, w))}{-u''(c(a, w))} \left\{ r - \rho + \underbrace{\lambda_2 \left[ \frac{u'(c(a, b))}{u'(c(a, w))} - 1 \right]}_{\text{prec. savings}} \right\} dt + \underbrace{\left[ c(a, b) - c(a, w) \right] dQ_t}_{\text{jumps}}$$
(3)

neoclassical growth model 
$$\dot{c}(t) = \frac{u'(c)}{-u''(c)} \Big(r-\rho\Big)$$

Looking at period between jumps. What the signs tell us?

**Proposition.** Consider the case  $0 < r \le \rho$ . Define the threshold level  $a_w^*$  by

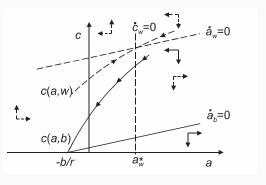
$$\frac{u'\left(c(a_{w}^{*},b)\right)}{u'\left(c(a_{w}^{*},w)\right)} = 1 + \frac{\rho - r}{\lambda_{2}} \tag{4}$$

Then (i) Consumption of employed workers is increasing on  $[\underline{a}, a_w^*]$  and decreasing  $a > a_w^*$ ; (ii) consumption of unemployed workers always decrease

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Properties of this system can be illustrated in the usual phase diagram

#### Policies



## Change

- Results help build some intuition on the problem. Look at Bayer and Wälde (2015) for much more...
- Now we change the approach.
   Instead of looking at households' saving behavior in terms of a differential equation for its consumption policy function, we will focus on the HJB equation for the value function and how to solve it numerically.
- draw heavily on Moll's notes

#### Problem of Household

$$\begin{aligned} \max_{\{c_t\}_{t\geq 0}} \mathbb{E}_0 & \int_0^\infty e^{-\rho t} u(c_t) dt \\ & \text{S.t } da_t = \left\{ ra_t + z_t - c_t \right\} dt \\ & z_t \text{ is a ct markov chain on } \{b, w\} \text{ with intensities } \lambda_1, \lambda_2 \\ & dz_t = (w-b) dq_\mu - (w-b) dq_s, \quad q_\mu \sim \mathsf{Poisson}(\lambda_1), \ q_s \sim \mathsf{Poisson}(\lambda_2) \\ & a_t \geq \underline{a} \end{aligned}$$

Individuals' value function must satisfy HJB equation<sup>1</sup>

$$\rho v_k(a) = \max_{c} \left\{ u(c) + v'_k(a)[ra + z_k - c] \right\} + \lambda_k \left[ v_{-k}(a) - v_k(a) \right]$$
 (5)

Borrowing constraint shows only as state constraint boundary condition

$$u'(c_i(\underline{\mathbf{a}})) = v_i'(\underline{\mathbf{a}}) \ge u'(r\underline{\mathbf{a}} + z_i) \tag{6}$$

which ensures  $s_i(\underline{a}) = r\underline{a} + z_i - c_i(\underline{a}) \ge 0$  so that the borrowing constraint is <u>never violated</u>.

<sup>&</sup>lt;sup>1</sup>change notation

#### Continuous × Discrete time

Consider the first-order condition for consumption

$$u'(c) = \partial_a v(a, z) \tag{7}$$

$$u'(c) \ge \beta \int \partial_{a} v(a', z') dF(z'|z), \quad a' = z + (1+r)a - c$$
 (8)

Continuous time advantages:

- 1. "today" = "tomorrow" foc is static
- 2. HJB is not stochastic evolution of stochastic process is captured by additive terms
- 3. foc always holds with equality

Borrowing constraint shows only as state constraint boundary condition

$$u'(c_i(\underline{a},t)) = \partial_a v_i(\underline{a},t) \ge u'(r_t \underline{a} + z_i)$$
(9)

which ensures  $c_i(\underline{a},t) \leq r_t \underline{a} + z_i$  so that the borrowing constraint is never violated.

# Numeric solution HJB

We can write our HJB

$$\rho v_k(a) = \max_c \left\{ u(c) + v_k'(a)[ra + z_k - c] \right\} + \lambda_k \left[ v_{-k}(a) - v_k(a) \right]$$

in a PDE notation

$$0 = F(x, V, DV, D^2V)$$

$$\tag{10}$$

where  $\mathbf{x} := (\mathbf{x}, \tau)$ . How do we proceed to solve it??

→ Finite difference methods: replace derivatives by differences. Simple right?

Suppose we define a grid  $\{x_0,x_1,\ldots,x_i,\ldots\}$  and a set of timesteps  $\{i\Delta:i\in\mathbb{N}\}$  Let  $V_i^n\approx V(x_i,\tau_n)$  be the approximate value of the solution at node  $x_i$  time  $\tau^n:=T-t$ . Then we can write a general **discretization** of the HJB equation at node  $(x_i,\tau^{n+1})$ 

$$0 = S_i^{n+1} \Big( (\Delta, \Delta x), V_i^{n+1}, \{ V_j^m \}_{m \neq n+1, j \neq i} \Big)$$
 (11)

## **Sufficient Conditions Convergence**

Condition (Monotonicity) . — The numerical scheme (11) is monotone if

$$S_i^{n+1}(\cdot, V_i^{n+1}, \{Y_i^m\}) \le S_i^{n+1}(\cdot, V_i^{n+1}, \{Z_i^m\})$$

for all  $Y \geq Z$ .

Condition (Stability) .— The numerical scheme (11) is stable if for every  $\tilde{\Delta}>0$  it has a solution which is uniformly bounded independently of  $\tilde{\Delta}$ .

Condition (Consistency) .— The numerical scheme (11) is consistent if for every smooth function  $\phi$  with bounded derivatives we have

$$S_i^{n+1}(\tilde{\Delta}, \phi(\boldsymbol{x}_i^{n+1}), \{\phi(\boldsymbol{x}_j^m)\}) \to F(\boldsymbol{x}, \phi, D\phi, D^2\phi)$$

as 
$$\tilde{\Delta} o 0$$
 and  $oldsymbol{x}_i^{n+1} o oldsymbol{x}$ .

## **Sufficient Conditions Convergence**

**Theorem** Barles and Souganidis (1990). If the numerical scheme S (11) satisfies monotonicity, stability and consistency conditions, then its solution converges locally uniformly to the unique viscosity solution of (10).

- $\bullet$  Convergence here is about  $\tilde{\Delta} \to 0$
- For given  $\tilde{\Delta}$ , we have a system of I non-linear equations for each timestep that we must solve somehow. Theorem guarantees that the solution  $\{V_i^\tau\}$  of this system converges to the "viscosity solution" of the original PDE as  $\tilde{\Delta} \to 0$
- "viscosity solution" of the HJB is the the value function

Recall our time-dependent HJB equation as

$$\partial_{\tau} v_k(a,\tau) + \rho v_k(a,\tau) - \sup_{c \in \Gamma_k(a)} \left\{ u(c) + \mathcal{D}^c_{\tau_{-\tau}} v_k(a,\tau) \right\} = 0 \tag{12}$$

where

$$\mathcal{D}_t^c \phi_k(\mathbf{a}) = \partial_{\mathbf{a}} \phi_k(\mathbf{a}) [r_t \mathbf{a} + z_k - c] + \lambda_k \Big[ \phi_{-k}(\mathbf{a}) - \phi_k(\mathbf{a}) \Big]$$

Define a grid  $\{a_1, a_2, \ldots, a_i, \ldots\}$  and let  $v_k^n = \left(v_k(a_1, \tau^n), \ldots, v_k(a_i, \tau^n), \ldots\right)'$ . Discretizing this equation requires deciding upon

· which fd approximation to use: forward/backward differencing

$$\partial_a v_k(a) \approx \frac{v_{k,i+1} - v_{k,i}}{a_{i+1} - a_i}, \quad \partial_a v_k(a) \approx \frac{v_{k,i} - v_{k,i-1}}{a_i - a_{i-1}},$$

• Implicit/explicit timestepping

Let  $\mathscr{D}^c$  be the discrete form of the differential operator  $\mathcal{D}^c$ , so that

$$\left(\mathscr{D}^{c}v\right)_{k,i} = \alpha_{k,i}(c)v_{k,i-1} + \beta_{k,i}(c)v_{k,i+1} - \left(\alpha_{k,i}(c) + \beta_{k,i}(c) + \lambda_{i}\right)v_{k,i} + \lambda_{i}v_{-k,i}$$

and the discretization

$$\frac{v_{k,i}^{n+1} - v_{k,i}^{n}}{\Delta} - \sup_{c \in \Gamma_{k,i}} \left\{ u(c) + (\mathscr{D}^{c} v^{n(+1)})_{k,i} \right\} = 0$$
 (13)

where discretization can use forward, backward or central discretization. If

$$\alpha_{k,i}^{n+1} \ge 0, \ \beta_{k,i}^{n+1} \ge 0$$

we say that (13) is positive coefficient discretization.

#### Why do we care?

We care because a *positive coefficient discretization* is also *monotone*. To see it check that

$$S_{k,i}^{n+1} \Big( \tilde{\Delta}, v_{k,i}^{n+1}, v_{k,i+1}^{n(+1)}, v_{k,i-1}^{n(+1)}, v_{k,i}^{n}, v_{-k,i}^{n(+1)} \Big)$$

is a nonincreasing function of the neighbor nodes  $\{v_{\ell,j}^m\}$ . Check a example!

## **Upwind scheme**

In order to ensure a *positive coefficient discretization* our choice of central/forward/backward differencing will depend, in general, on the control c. A useful rule for this problem is to use the so-called *upwind scheme*.

IDEA: Use forward difference whenever drift is positive, and use backward whenever it is negative.

Suppose that we have the value of consumption  $c_{k,i}$  at a particular node. Let  $s_{k,i} = ra_i + z_k - c_{k,i}$ . In this case, the derivatives are approximated

$$\dots \frac{v_{k,i+1} - v_{k,i}}{a_{i+1} - a_i} s_{k,i}^+ + \frac{v_{k,i} - v_{k,i-1}}{a_i - a_{i-1}} s_{k,i}^- + \dots$$

which in terms of our  $\alpha, \beta$ 

$$\alpha_{k,i}^{up} = -\frac{s_{k,i}^{-}}{a_{i} - a_{i-1}} \ge 0, \quad \beta_{k,i}^{up} = \frac{s_{k,i}^{+}}{a_{i+1} - a_{i}} \ge 0$$

But we don't know  $c_{k,i}!!!$  HJB equation is highly nonlinear, so we need an iterative method to solve it.

## Implicit timestepping

Start with a vector  $v^n$  and update  $v^{n+1}$  according to

$$\frac{v_{k,i}^{n+1} - v_{k,i}^{n}}{\Delta} + \rho v_{k,i}^{n+1} = u(c_{k,i}^{n}) + \frac{v_{k,i+1}^{n+1} - v_{k,i}^{n+1}}{a_{i+1} - a_{i}} \left[ s_{k,i}^{F,n} \right]^{+} + \frac{v_{k,i}^{n+1} - v_{k,i-1}^{n+1}}{a_{i} - a_{i-1}} \left[ s_{k,i}^{B,n} \right]^{-} + \lambda_{k} \left[ v_{-k,i}^{n+1} - v_{k,i}^{n+1} \right]$$

$$(14)$$

- Compute the policy from the foc  $\left(u'(c_{k,i}^n) = \partial_a v_{k,i}^n\right)$  for the backward AND forward derivative of the value function.
- Define  $s_{k,i}^{B,n} = ra_i + z_k c_{k,i}^{B,n}, \ s_{k,i}^{F,n} = ra_i + z_k c_{k,i}^{F,n}$ . Set

$$c_{k,i}^{n} = \mathbb{1}\left\{s_{k,i}^{B,n} \leq 0\right\} \times c_{k,i}^{B,n} + \mathbb{1}\left\{s_{k,i}^{F,n} \geq 0\right\} \times c_{k,i}^{F,n} + \mathbb{1}\left\{s_{k,i}^{F,n} \leq 0 \leq s_{k,i}^{B,n}\right\} \times (\textit{ra}_{i} + \textit{z}_{k})$$

• Collecting terms with the same subscripts on the right-hand side

$$\frac{v_{k,i}^{n+1} - v_{k,i}^n}{\Lambda} + \rho v_{k,i}^{n+1} = u(c_{k,i}^n) + \alpha_{k,i} v_{k,i-1}^{n+1} + \beta_{k,i} v_{k,i+1}^{n+1} - \left(\alpha_{k,i} + \beta_{k,i} + \lambda_k\right) v_{k,i}^{n+1} + \lambda_i v_{k,i}^{n+1}$$
(15)

where

$$\alpha_{k,i}^{up} = -\frac{\left[s_{k,i}^{B,n}\right]^{-}}{a_{i} - a_{i-1}} \ge 0, \quad \beta_{k,i}^{up} = \frac{\left[s_{k,i}^{F,n}\right]^{+}}{a_{i+1} - a_{i}} \ge 0$$

• Equation (15) is just a system of linear equations!!

## Implicit Timestepping

Equation (15) can be written in matrix notation as

$$\frac{1}{\Delta}(v^{n+1} - v^n) + \rho v^{n+1} = u(c^n) + \mathbf{A}^n v^{n+1}$$

where the sparse matrix A looks like

entries of row i

$$\begin{bmatrix} \underline{\alpha_{k,i}} & \underline{-(\alpha_{k,i} + \beta_{k,i} + \lambda_k)} & \underline{\beta_{k,i}} \\ \text{inflow } i-1 & \underline{\text{outflow}} & \underline{\text{inflow } i+1} \end{bmatrix} \begin{bmatrix} v_{k,i-1} \\ v_{k,i} \\ v_{k,i+1} \end{bmatrix}$$

# **Computing the Distribution**

#### **Distributions**

- We now know how to solve the Household consumption/savings problem
- But interesting questions require dealing with distributions
- Denote by  $g_i(a, t)$  i = 1, 2 the joint density of income  $z_i$  and welath a.
- The evolution of the density given a fixed initial distribution  $g_i(a,0)$  is described by the Kolmogorov forward equation
  - time dependent

$$\frac{\partial}{\partial t}g(a,t) = -\frac{\partial}{\partial a}\Big[s_k(a,t)g_k(a,t)\Big] - \lambda_k g_k(a,t) + \lambda_{-k}g_{-k}(a,t)$$
 (16)

stationary

$$0 = -\frac{\mathrm{d}}{\mathrm{d}a} \left[ s_k(a) g_k(a) \right] - \lambda_k g_k(a) + \lambda_{-k} g_{-k}(a) \tag{17}$$

Consider the stationary KFE

$$0 = -\frac{\mathrm{d}}{\mathrm{d}a} \left[ s(a, z_k) g(a, z_k) \right] - \lambda_k g(a, z_k) + \lambda_{-k} g(a, z_k)$$

with the following discretization

$$0 = -\frac{\left(s_{k,i}^{F}\right)^{+}g_{k,i} - \left(s_{k,i-1}^{F}\right)^{+}g_{k,i-1}}{\Delta a} - \frac{\left(s_{k,i+1}^{B}\right)^{-}g_{k,i+1} - \left(s_{k,i}^{B}\right)^{-}g_{k,i}}{\Delta a} - \lambda_{k}g_{k,i} + \lambda_{-k}g_{-k,i} \quad (18)$$

Collecting terms with the same subscripts on the right-hand side

$$0 = \underbrace{\frac{\left(s_{k,i-1}^F\right)}{\Delta_{\mathbf{a}}}}_{\beta_{k,i-1}} g_{k,i-1} + \underbrace{\left(\frac{\left(s_{k,i}^B\right)}{\Delta_{\mathbf{a}}} - \frac{\left(s_{k,i}^F\right)}{\Delta_{\mathbf{a}}} - \lambda_k\right)}_{\gamma_{k,i}} g_{k,i} + \underbrace{\left(-\frac{\left(s_{k,i+1}^B\right)}{\Delta_{\mathbf{a}}}\right)}_{\alpha_{k,i+1}} g_{k,i+1} + \lambda_{-k} g_{-k,i}$$

which in matrix notation reads  $\mathbf{A}^T g = \mathbf{0}$ .

This makes sense: the operation is exactly the same as that used for finding the stationary distribution of a discrete Poisson process (continuous-time Markov chain). The matrix A captures the evolution of the stochastic process over a very short interval — it is our discretized *infinitesimal generator* of our state — and to find the stationary distribution, one solves the eigenvalue problem  $\mathbf{A}^Tg=\mathbf{0}$ .

[Pass slide to Transition dynamics]

Consider the time dependent

$$\frac{\partial}{\partial t}g(a,t) = -\frac{\partial}{\partial a}\Big[s_k(a,t)g_k(a,t)\Big] - \lambda_k g_k(a,t) + \lambda_{-k}g_{-k}(a,t)$$

Given an initial condition, the KFE can be easily solve through a implicit method

$$\frac{g^{n+1} - g^n}{\Delta} = \left(\mathbf{A}^{n(+1)}\right)^T g^{n+1} \tag{19}$$

In the case wiht a dirac mass on the boundary constraint

$$M_1(t) + \int_0^a g_1(x,t)dx = G_1(a,t), \qquad \int_0^a g_2(x,t)dx = G_2(a,t)$$
 (20)

and we may write

$$\frac{\partial}{\partial t}g_1(a,t) = -\frac{\partial}{\partial a}\Big[s_1(a,t)g_1(a,t)\Big] - \lambda_1 g_1(a,t) + \lambda_2 g_2(a,t)$$
(21)

$$\frac{\partial t}{\partial t}g_{2}(a,t) = -\frac{\partial}{\partial a}\left[s_{2}(a,t)g_{2}(a,t)\right] - \lambda_{2}g_{2}(a,t) + \lambda_{1}g_{1}(a,t) + \underbrace{\lambda_{1}M_{1}\mathbb{I}_{\{a=0\}}}_{\text{flow? mass?}} + \underbrace{\lambda_{1}M_{1}\delta_{0}(a)}_{\text{flow? mass?}}$$
(22)

$$\frac{\partial}{\partial t} M_1(t) = -\lim_{\epsilon \to 0} s_1(\underline{a} + \epsilon, t) g_1(\underline{a} + \epsilon, t) - \lambda_1 M_1$$
(23)

Integrating the KFE between  $\underline{\mathbf{a}} + \epsilon$  and  $\infty$ 

$$\frac{\partial}{\partial t} \int_{\underline{\underline{a}}+\epsilon} g_1(a,t) da = s_1(\underline{\underline{a}}+\epsilon,t)g_1(\underline{\underline{a}}+\epsilon,t) - \lambda_1 \int_{\underline{\underline{a}}+\epsilon} g_1(a,t) da + \lambda_2 \int_{\underline{\underline{a}}+\epsilon} g_2(a,t) da$$

Using the definition on distributions and taking the limit as  $\epsilon o 0$ 

$$-\frac{\partial}{\partial t}\Big(M_1(t)-G_1(t)\Big)=\lim_{\epsilon\to 0}s_1(\underline{a}+\epsilon,t)g_1(\underline{a}+\epsilon,t)+\lambda_1\Big(M_1(t)-G_1(t)\Big)+\lambda_2G_2(t)$$

but note that  $\frac{\partial}{\partial t}G_1(t)=-\lambda_1G_1(t)+\lambda_2G_2(t)$ . That leave us with (23).

# Stationary Equilibrium + Transion Dynamics

# Experiment

# **Equilibrium (stationary)**

$$\rho v_k(a) = \max_{c} \left\{ u(c) + \partial_a v_k(a) \right\} + \lambda_k \left[ v_\ell(a) - v_k(a) \right]$$
 [HJB]

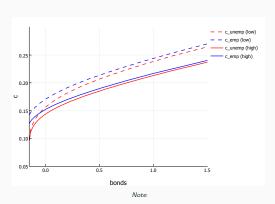
$$\begin{split} 0 &= \frac{\mathrm{d}}{\mathrm{d}a} \big[ s_k(a) g_k(a) \big] - \lambda_k g_k(a) + \lambda_\ell g_\ell(a) \\ 1 &= \int_{\underline{a}}^{\infty} \Big( g_1(a) + g_2(a) \Big) da \end{split}$$
 [KFE]

$$0=\int_{\underline{a}}^{\infty} a \Big(g_1(a)+g_2(a)\Big) da$$
 [Equil]

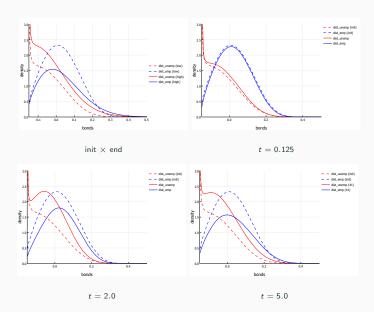
## **Experiment**

Suppose an increase in the unemployment risk  $\lambda_2$ . What would you expect has to happen to the interest rate? Distributional effect of more unemployed in equilibrium makes me converge to an higher interest rate!

#### Consumption Policy



#### Transition Dynamics



Note:

Why do I care??

## **Example**

- looks nice, but why should I pay a cost if I can do discrete time?
- · It was my original tought, but recently...
- Consumption/savings problem + direct search labor market

$$\begin{split} V^{u}(a) &= \max_{c,\,a'} u(c) + \beta \mathcal{R}^{u}(a') \\ \text{S.t. } c &+ \frac{a'}{1+c} = b+a, \quad a' \geq \underline{a} \end{split}$$

euler equation for asset holdings

$$u'(c(\cdot)) \geq \beta(1+r) \left\{ \left(1 - p(\theta(a', \tilde{w}))\right) V_a^u(a') + p(\theta(a', \tilde{w})) V_a^e(a', \tilde{w}) + p'(\theta(a', \tilde{w}) \frac{\partial \theta(a', \tilde{w})}{\partial a} \left[ V^e(a', \tilde{w}) - V^u(a', \tilde{w}) \right] \right\}$$

- EGM for consumption/savings, VFI for labor choice.
- Finding the equilibrium requires iterating over a lot of stuff

$$V(\cdot) \hookrightarrow \hat{\theta}(\cdot) \hookrightarrow \mathcal{J}(\cdot) \hookrightarrow \theta(\cdot)$$

• couldn't do it = ( ...

## **Example**

How does this look in continuous time, today is tomorrow

$$\rho V^{u}(a) = \max_{c} \left\{ u(c) + V_{a}^{u}(b + ra - c) \right\} + \underbrace{\lambda_{u}}_{\text{rate of search}} \max_{\tilde{w}} \left\{ p(\theta(a, \tilde{w})) \left[ V^{e}(a, w) - V^{u}(a) \right] \right\}$$
(24)

This is wayyyyy simpler and importantly it doesn't seem I am throwing away anything
of the economics

## This is a title 10pt

Here is some content on the slide

$$f(x) = ax^2 + bx + c$$

Some more content

## References

- Achdou, Y., J. Han, J.-M. Lasry, P.-L. Lions, and B. Moll (2016): "Heterogeneous Agent Models in Continuous Time," .
- Barles, G., and P. Souganidis (1990): Convergence of Approximation Schemes for Fully Nonlinear Second Order Equations.
- BAYER, C., AND K. WÄLDE (2015): "The Dynamics of Distributions in Continuous-Time Stochastic Models," .
- CALDENTEY, R. (????): Stochastic Control.
- FLEMING, W., AND H. SONER (2006): Controlled Markov Processes and Viscosity Solutions, Stochastic Modelling and Applied Probability. Springer New York.
- FORSYTH, P. A., AND K. R. VETZAL (2012): Numerical Methods for Nonlinear PDEs in Financepp. 503–528. Springer Berlin Heidelberg, Berlin, Heidelberg.
- KAPLAN, G., B. MOLL, AND G. L. VIOLANTE (2016): "Monetary Policy According to HANK," Working Papers 1602, Council on Economic Policies.
- Nuño, G., AND B. MOLL (2017): "Social Optima in Economies with Heterogeneous Agents," Discussion paper.

- ØKSENDAL, B. (2003): Stochastic Differential Equations: An Introduction with Applications, Hochschultext / Universitext. Springer.
- ØKSENDAL, B., AND A. SULEM (2007): Applied Stochastic Control of Jump Diffusions. Springer Berlin Heidelberg.
- Pham, H. (2009): Continuous-time Stochastic Control and Optimization with Financial Applications. Springer Publishing Company, Incorporated, 1st edn.
- Ross, K. (????): Stochastic Control in Continuous Time.
- SENNEWALD, K. (2007): "Controlled stochastic differential equations under Poisson uncertainty and with unbounded utility," *Journal of Economic Dynamics and Control*, 31(4), 1106 – 1131.
- STOKEY, N. L. (2009): The Economics of Inaction: Stochastic Control Models with Fixed Costs. Princeton University Press.
- THOMAS, C., AND G. NUÑO (2016): "Monetary Policy and Sovereign Debt Vulnerability," 2016 Meeting Papers 329, Society for Economic Dynamics.
- WALDE, K. (2008): Applied Intertemporal Optimization, no. econ1 in Books. Business School - Economics, University of Glasgow.