

# Dynamic Programming: Theory and Applications

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# Outline

- 1 Asset Pricing
  - Model
  - Estimation

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## 2 Cake Eating

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## Model

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- Price an asset (tree) with stochastic dividends (fruit)
- Recursive Asset Pricing:

$$q(d) = d + \beta E_{d'|d} q(d') \quad \forall d \in D \quad (1)$$

- here:
  - $d$  is the current dividend (state)
  - $q(d)$  is the unknown asset price (value function)
  - $\beta \in [0, 1)$
  - stationarity
  - $d$  follows a first-order Markov process by assumption
  - $D$  is the set of possible realizations of  $d$  .. often discrete
- goal is to find  $q(d)$  solving (1)

## Solving the functional equation

- Assume two states:  $d \in (d_L, d_H)$
- transition matrix is  $\Pi$ :  $\pi_{ij}$  is the probability of going from state  $i$  to  $j$
- Solution solves two equations:

$$\begin{aligned} q(d_L) &= d_L + \beta[\pi_{LL}q(d_L) + (1 - \pi_{LL})q(d_H)] \\ q(d_H) &= d_H + \beta[\pi_{HH}q(d_H) + (1 - \pi_{HH})q(d_L)] \end{aligned} \quad (2)$$

- Easy: solve by substitution

## Guess/Verify

- assume  $d_{t+1} = \rho d_t + \varepsilon_{t+1}$  so  $E_{d_{t+1}|d_t} d_{t+1} = \rho d_t$
- $|\rho| < 1$
- Guess/Verify:  $q(d) = \xi d$

- 

$$\xi d = d + \beta \xi E_{d'|d} d'$$

for all  $d$

- 

$$\xi d = d + \beta \xi \rho d$$

for all  $d$

- 

$$\xi = \frac{1}{1 - \beta \rho}$$

- fails if  $d_{t+1} = \mu_d + \rho d_t + \varepsilon_{t+1}$
- guess/verify only works for special problems



## Value Function Iteration

- Overkill for this problem, essential for others
- iteration indexed by  $i$
- initial guess,  $q^0(d) \quad \forall d \in D$
- $D$  is discrete, transition matrix  $\Pi$
- iterate

$$q^{i+1}(d) = d + \beta E_{d'|d} q^i(d') \quad \forall d \in D \quad (3)$$

- will converge to solution of the functional equation regardless of  $q^0(d)$  if discounting and monotonicity (Blackwell)

## Estimating Parameters

- Asset Pricing Model

$$q(d) = u(d) + \beta E_{d'|d} q(d') \quad \forall d \in D \quad (4)$$

- Data:  $(q_t, d_t)$  for  $t = 1, 2, 3 \dots T$
- Want to estimate:  $\Theta = (\Pi, \beta, u(\cdot))$
- How can we do that?
- Estimate  $\Pi$  directly from observations on dividends ...
  - count transitions
  - use Tauchen Procedure to create a discrete representation of the AR(1) process
- Estimate  $(\beta, u(\cdot))$  using: GMM, SMM, non-linear least squares

## SMM example

- estimate  $\beta$ , assume  $u(d) = d$  and  $\Pi$  known
- compute a data moment: eg. mean of  $q$ , denoted  $\mu^d$
- estimation loop
  - solve (4) for a given  $\beta$  to get  $q_{\Theta}(d)$
  - simulate a data set using  $\Pi$  and  $q_{\Theta}(d)$
  - calculate the mean of  $q$  from the simulated data,  $\mu^s(\beta)$
  - find  $\beta$  to minimize  $(\mu^s(\beta) - \mu^d)^2$
- Exercise: Graph simulated and data moments as a function of  $\beta$

## Extensions

- estimate more parameters
- overidentified model:
  - more moments than parameters
  - test fit of theory as well as estimate parameters
- Indirect Inference
  - obtain moments from a regression:  $q_t = \alpha_0 + \alpha_1 d_t + \alpha_2 d_t^2$
  - match these regression coefficients in the simulated data
- GMM
  - if there was a choice
  - obtain ex post error from:  $\varepsilon_{t+1}(\Theta) = q_t - u(d_t) - \beta q_{t+1}$
  - theory implies  $E_t \varepsilon_{t+1}(\Theta^*) = 0$
  - use sample analogue to estimate  $\Theta$

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## Dynamic Discrete Choice

$$V(W, \varepsilon) = \max(V^E(W, \varepsilon), V^N(W, \varepsilon)) \quad (5)$$

$$V^E(W, \varepsilon) = \varepsilon u(W) \quad (6)$$

and

$$V^N(W, \varepsilon) = \beta E_{\varepsilon'|\varepsilon} V(\rho W, \varepsilon'). \quad (7)$$

for all  $(W, \varepsilon)$

## Policy Function

- $Z(W, \varepsilon) \in \{0, 1\}$ , probability of eating the cake
- Leading Example
  - Assume  $\varepsilon$  is iid,  $\rho = 1$ .
  - $Z(W, \varepsilon) = 1$  iff  $\varepsilon \geq \varepsilon^*$  for all  $W$ .
  - usual "reservation wage" structure

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## Estimation

- Estimate  $\Theta \equiv (\beta, \Pi, u(\cdot))$
- $\varepsilon^*(\Theta)$ : actions depend on parameters
- What can we identify here? Not Much!



## Two State Example

- $\varepsilon \in \{\varepsilon_L, \varepsilon_H\}$
- $Z(W, \varepsilon_H) = 1$  for sure
- $Z(W, \varepsilon_L) = ???$
- create program
  - to illustrate VFI approach for this example
  - study policy function and identification

# Continuous Choice

$$V(W, \varepsilon) = \max_{W'} \varepsilon u(W - W') + \beta E_{\varepsilon'|\varepsilon} V(W', \varepsilon') \quad \forall (W, \varepsilon) \quad (8)$$

- $\varepsilon$  is a taste shock: exogenous state
- $W$  is the size of the cake: endogenous state
- control is future cake (or could use consumption)
- policy function:  $W' = \phi(W, \varepsilon)$
- characterize solution through an Euler equation

## Conditions for Optimality

- FOC

$$\varepsilon u'(W - W') = \beta E_{\varepsilon'|\varepsilon} V_1(W', \varepsilon') \quad (9)$$

- Envelope Condition

$$V_1(W, \varepsilon) = \varepsilon u'(W - W') \quad (10)$$

- Euler Equation

$$\varepsilon u'(W - W') = \beta E_{\varepsilon'|\varepsilon} \varepsilon' u'(W' - W'') \quad (11)$$

- Condition for ex-ante optimality: no errors anticipated

## Estimation Approach

- estimate  $\Theta = (\beta, u(\cdot))$
- Ex-post Error in Time Domain:

$$\zeta_{t+1}(\Theta) = \frac{\beta \varepsilon_{t+1} u'(c_{t+1})}{\varepsilon_t u'(c_t)} - 1 \quad (12)$$

- Orthogonality:  $E_t \zeta_{t+1}(\Theta^*) = 0$
- estimate  $\Theta$  from sample analogue ala Hansen-Singleton
- under conditions for doing this in the next lecture