# SOLVING HETEROGENEOUS AGENT MODELS WITH AGGREGATE UNCERTAINTY

Solution Methods for Macroeconomic Models

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# SOLUTION METHODS FOR MACROECONOMIC MODELS

- Monday Tuesday: Solving models with "representative agents"
  - Linearization in theory and practice: Dynare
  - · Non-linear solutions methods: value function iteration, projection
  - Analyzing models: parameterization/estimation, simulation/IRFs
- Wednesday Thursday: Solving models with "heterogeneous agents"
  - Models without aggregate uncertainty: basic algorithm
  - Models with aggregate uncertainty: key issues and alternatives
- Friday: "Final assignment"
  - Solve/estimate model with heterogeneous firms and aggregate uncertainty

#### OVERVIEW FOR TODAY

Extend Ayiagari model to include aggregate uncertainty

- 1. extend Aiyagari model with aggregate uncertainty
- 2. Krusell-Smith algorithm
- 3. practical issues (simulation, accuracy)
- 4. one alternative

AIYAGARI MODEL WITH AGGREGATE

UNCERTAINTY

# EXTEND AIYAGARI MODEL

# Assume presence of aggregate uncertainty

enters production function of representative firm

$$Y_t = Z_t K_t^{\alpha} L_t^{1-\alpha}$$

 $\cdot$  otherwise there is no difference from model before

# **NEW INDIVIDUAL PROBLEM**

The individual problem is now

$$\max_{\{c_{i,t}, k_{i,t+1}\}_{t=0}^{\infty}} \mathbb{E}_t \sum_{t=0}^{\infty} \beta^t \ln(c_{i,t})$$
s.t.
$$c_{i,t} + k_{i,t+1} = r_t k_{i,t} + w_t \epsilon_{i,t} + (1 - \delta) k_{i,t}$$

$$k_{i,t+1} \ge 0$$

how is this different from before?!

# **NEW INDIVIDUAL PROBLEM**

• 
$$r_t = Z_t \alpha K_t^{\alpha - 1}$$
 and  $w_t = Z_t (1 - \alpha) K_t^{\alpha}$ 

Moreover, what do agents really care about?

- not just  $r_t$  and  $w_t$
- but also all future values of *r* and *w*!

#### WHAT DO AGENTS CARE ABOUT?

# The need to forecast prices necessitates

- the need for information for forecasting K
- in general, we need to forecast the joint distribution
  - of capital holdings and idiosyncratic productivity levels
- now this is really a tough problem
  - · the distribution is infinite-dimensional!

#### BEFORE MOVING ON

# The above discussion makes clear

- · that partial equilibrium is much easier
- · even with aggregate uncertainty

# The individual's problem is easy to solve if

- you know the path of  $r_t$  and  $w_t$ !
- the above hints at how to solve the GE model

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# MAIN IDEA OF KRUSELL-SMITH ALGORITHM

# Principle is similar to that in Aiyagari model

- guess (evolution of) prices
- individual problem solved for given (evolution of) prices
- · simulate economy, aggregate, check implied price paths

# The difficulty is that to forecast *K*

- · we really need to forecast the entire joint distribution
  - · of capital holdings and idiosyncratic productivity

# FORECASTING THE JOINT DISTRIBUTION

Let  $\mathbb{F}_t$  be the beginning-of-period joint distribution

of capital holdings and idiosyncratic productivity

Its evolution can then be described as

$$\mathbb{F}_{t+1} = \mu(Z_t, \mathbb{F}_t)$$

- given today's joint distribution
- and today's aggregate shock
- can forecast tomorrows joint distribution
  - recall that idiosyncratic shocks are exogenous

# FORECASTING THE JOINT DISTRIBUTION

However, this is still a crazy difficult problem!

· distribution is infite-dimensional

Krusell and Smith (1998) propose to instead

- approximate the distribution by
- focusing on a limited set of characteristics

AN ITERATIVE PROCEDURE

# Instead of figuring out $\mu(.,.)$

· assume an approximating "aggregate law of motion"

$$m_{t+1} = \overline{\mu}(Z_t, m_t; \psi_{\overline{\mu}})$$

- $m_t$  is a set of moments of the joint distribution
  - you need to make a stand on  $m_t$  beforehand
  - · we'll discuss such choices later

- 1. guess a value for  $\psi_{\overline{\mu}}$ 
  - $\cdot$  implies values for  $K_t^D$  and thus  $r_t$  and  $w_t$
- 2. solve individual problem with given aggregate law of motion
- 3. simulate economy and calculate moments of joint distribution
- 4. estimate  $\hat{\psi}_{\overline{\mu}}$  implied by simulation
- 5. compare to previous guess
  - if  $\hat{\psi}_{\overline{\mu}} = \psi_{\overline{\mu}} \to \operatorname{stop}$
  - · if  $\hat{\psi}_{\overline{\mu}} \neq \psi_{\overline{\mu}} \to \text{update}$  and go to 2

#### WHAT TO KEEP IN MIND

# Algorithm based on idea of "approximate aggregation"

- evolution of prices described well using
  - exogenous shocks
  - a limited number of moments of current distribution
- does not mean that
  - behavior aggregates to rep-agent model
  - or that individual variables behave as aggregates!

#### TAKING STOCK

# Krusell-Smith algorithm

- $\boldsymbol{\cdot}$  same principle as when solving model without aggregate uncertainty
- key difficulty lies in having to forecast evolution of prices
- need to track entire joint distribution of state variables

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# PRACTICAL ISSUES

- 1. guess aggregate law of motion  $(\psi_{\overline{\mu}})$ 
  - $\cdot$  implies values for  $K_t^D$  and thus  $r_t$  and  $w_t$
- 2. solve individual problem with given aggregate law of motion
- 3. simulate economy and calculate moments of joint distribution
- 4. estimate  $\hat{\psi}_{\overline{\mu}}$  implied by simulation
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#### **CHOICES TO BE MADE**

# There are several choices you must make

- how to solve individual problem?
- · which moments to pick (and of what)?
- how to update the aggregate law of motion?
- how to simulate the economy?
- when to stop iterating?
- how to check for accuracy?
- non-trivial market clearing and imposing equilibrium?

# PRACTICAL ISSUES

INDIVIDUAL PROBLEM

# HOW TO SOLVE INDIVIDUAL PROBLEM

You can use your favorite solution method, but

· number of state variables has increased

# HOW TO SOLVE INDIVIDUAL PROBLEM

What trade-offs are you facing?

# PRACTICAL ISSUES

AGGREGATE LAW OF MOTION

#### WHICH MOMENTS TO PICK?

# Krusell-Smith used only mean of capital

- there is no rule about this  $\rightarrow$  depends on application
- · in their setup, means give sufficiently accurate results
  - if policy rules are exactly linear in levels
  - only mean necessary for computing next period's mean
    - · distribution of wealth doesn't matter
  - in their setup, policy rules close to linear in levels

#### WHICH MOMENTS TO PICK?

# Essentially trial and error

- use "bottom-up" approach
  - start with means
  - solve model (i.e. until aggregate law of motion converges)
  - check accuracy
  - if you're lucky, you're done
  - if not, adjust (number of) moments

# WHICH MOMENTS TO PICK?

"Non-traditional" moments can be informative

· e.g. mass of agents around important cutoff

Past values of aggregate shocks

- easier to implement
- often quite informative about distribution
- · intuition?

# AGGREGATE LAW OF MOTION FOR WHAT?

Above we've used moments of joint distribution

 $\cdot$  this implied path for  $K_t^D$  and in turn  $w_t$  and  $r_t$ 

But it is really prices that agents care about

- specify aggregate law of motion directly for prices
- $\boldsymbol{\cdot}$  but, still need laws of motion for moments of interest!

#### HOW TO UPDATE COEFFICIENTS?

Let  $\psi^q_{\overline{\mu}}$  be the coefficient guess in the qth iteration

- $\cdot$  after solving and simulating the model economy
- $\cdot$  obtain a time-series of the moments of interest  $m_t$
- · use this time-series to estimate coefficients
  - regress  $m_{t+1}$  on  $Z_t$  and  $m_t$
  - obtain  $\hat{\psi}_{\overline{\mu}}$
- · update coefficient guess according to

$$\psi_{\overline{\mu}}^{q+1} = \lambda \hat{\psi}_{\overline{\mu}} + (1 - \lambda)\psi_{\overline{\mu}}^{q}$$

- choosing  $\lambda$  is again tricky
- strike compromise between speed and convergence

# PRACTICAL ISSUES

SIMULATION

#### TWO WAYS HOW TO SIMULATE

# Given individual policy rules

- simulate a large cross-section of agents
  - · use Monte-Carlo integration to get moments
- use a grid method not requiring stochastic simulation
  - not introducing any sampling noise

#### SIMULATING A CROSS-SECTION

- very intuitive and simple to implement
- however, quite computationally costly
  - need large cross-section
  - need long cross-section

#### **GRID METHOD**

Policy rules  $k_{i,t+1} = k(k_{i,t}, \epsilon_{i,t}, S_t)$  are given

• where  $S_t$  is aggregate state (including tracked moments)

Create fine grid of nodes  $(\kappa_{i,j})$  for joint distribution

- $\cdot$   $p_{i,j,t}$  is the beginning-of-period mass of agents with
  - $\cdot$   $k_{i,t}$  capital holdings and
  - $\cdot$   $\epsilon_{j,t}$  level of idiosyncratic productivity

### **GRID METHOD**

## Assume an initial joint distribution

- · first, focus on capital choice
  - 1. for each node, figure out end-of-period capital choice

$$k_{i,j,t+1} = k(k_{i,t}, \epsilon_{j,t}, S_t)$$

- 2. assign beginning-of-period mass  $(p_{i,j,t})$  to nodes
  - · issue: no mass in between nodes
  - split mass proportionally between neighboring nodes

### **GRID METHOD**

Split beginning-of-period mass proportionally

$$\omega_{i,j,t} = \frac{k_{i,j,t+1} - \kappa_{i-1,j}}{\kappa_{i,j} - \kappa_{i-1,j}}$$

$$\overline{p}_{i-1,j,t} = \overline{p}_{i-1,j,t} + p_{i,j,t} (1 - \omega_{i,j,t})$$

$$\overline{p}_{i,j,t} = \overline{p}_{i,j,t} + p_{i,j,t} \omega_{i,j,t}$$

- $\overline{p}_{i,j,t}$  is end-of-period mass
- careful at end points
- · careful, one grid point can "be filled" by many others!

### **GRID METHOD**

- second, focus on idiosyncratic productivity
  - · use transition law to figure out
  - beginning-of-period joint distribution in next period
  - 1. for each node, figure out next-periods mean productivity value

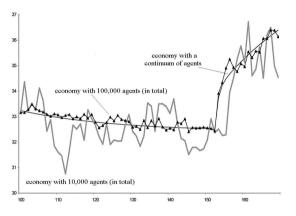
$$\epsilon_{j,t+1} = \rho \epsilon_{j,t}$$

- 2. assign mass end-of-period mass  $(\overline{p}_{i,j,t})$  to nodes
  - according to distribution of idiosyncratic shocks, e.g.  $N(0,\sigma_{\epsilon}^2)$
  - again split mass proportionally between nodes

### TRADE-OFFS WITH THE ABOVE

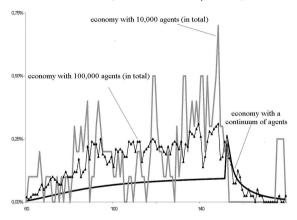
- Monte-Carlo simulation
  - can be computationally costly
  - introduces sampling noise
- · Grid method
  - constructing the grid is not easy
    - may need many nodes
    - sometimes need precession in parts of state-space

## WHEN IS SIMULATION CHOICE IMPORTANT (ALGAN ET AL., 2010)?



Notes: This graph plots the simulated aggregate capital stock of the unemployed using either a finite number (10,000) or a continuum of agents. It displays a subset of the observations shown in Figure 1.

## WHEN IS SIMULATION CHOICE IMPORTANT (ALGAN ET AL., 2010)?



Notes: This graph plots the simulated fraction of unemployed agent at the borrowing constraint using either a finite number (10,000) or a continuum of agents.

## PRACTICAL ISSUES

**ACCURACY** 

### WHEN TO STOP?

## When to stop what?

- when to stop iterating on aggregate law of motion?
- when to stop looking for accurate law of motion?
  - i.e. even if it has converged!

### **UPDATING PROCEDURE**

Let  $\psi^q_{\overline{\mu}}$  be the coefficient guess in the qth iteration

- after solving and simulating the model economy
- · obtain a time-series of the moments of interest  $m_t$
- · use this time-series to estimate coefficients
  - regress  $m_{t+1}$  on  $Z_t$  and  $m_t$
  - $\cdot$  obtain  $\hat{\psi}_{\overline{\mu}}$
- update coefficient guess according to

$$\psi_{\overline{\mu}}^{q+1} = \lambda \hat{\psi}_{\overline{\mu}} + (1 - \lambda)\psi_{\overline{\mu}}^{q}$$

### WHEN TO STOP UPDATING?

### Choice of $\lambda$

- · strike compromise between speed and chance of convergence
- must specify a measure of (update) distance
  - · e.g. max-abs-difference between updates

$$e^q = \max(\operatorname{abs}(\psi_{\overline{\mu}}^q - \psi_{\overline{\mu}}^{q-1}))$$

· or often more conveniently in percentage terms

$$e^q = \max( \text{abs}((\psi_{\overline{\mu}}^q - \psi_{\overline{\mu}}^{q-1})/\psi_{\overline{\mu}}^{q-1}))$$

• stop when  $e^q$  falls below a certain threshold (e.g. 0.01%)

## WHEN TO STOP UPDATING?

What does convergence of the aggregate law of motion mean?

### **ACCURACY CHECKS**

What dimensions of inaccuracy are there in our setup?

### AFTER WE STOP UPDATING LAW OF MOTION

### We have

- individual policy rules
- initial joint distribution
- a simulation method
- a candidate aggregate law of motion

$$m_{t+1} = \overline{\mu}(Z_t, m_t; \psi_{\overline{\mu}}) + u_{t+1} \tag{1}$$

- the above needs to be chekced for accuracy
- true law of motion has  $u_{t+1} = 0 \ \forall t$

### POPULAR WAY TO CHECK FOR ACCURACY

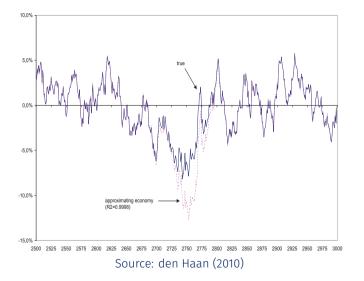
## Use only individual policy rules

- $\cdot$  simulate economy and obtain a time-series for  $m_t$
- use these time-series to run regression (1)
- evaluate goodness of fit by looking at R<sup>2</sup>
  - $\cdot$  and sometimes standard error of regression

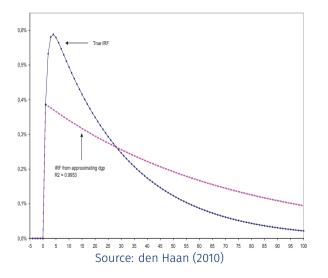
### **ISSUES WITH POPULAR ACCURACY TESTS**

- $R^2$  and  $\hat{\sigma}_u$  are averages
  - · can mask potentially large, counteracting, differences
- not clear what is a low R<sup>2</sup>
  - · check out den Haan's work for fun examples!
- main problem is conceptual!
  - each period, the truth is used to update the approximation!
  - $\cdot$  i.e. we're cutting the law of motion too much slack

## GREAT R<sup>2</sup> BUT POOR PERFORMANCE



## GREAT R<sup>2</sup> BUT POOR PERFORMANCE



## BETTER ACCURACY TESTS (DEN HAAN, 2010)

Generate two sequences of desired moments  $m_t$ 

And compare the two sequences

- accuracy plot!
- compute max(abs) difference and see where it occurs
  - important part of the state-space?
- what is the average/minimum error?
- · look at IRFs under the two laws of motion etc.

# PRACTICAL ISSUES

IMPOSING EQUILIBRIUM

## APPROACHING EQUILIBRIUM?

We've seen that the iterative algorithm approaches to (the) equilibrium

• is this always going to be the case?

Imagine the same economy, but with bonds in zero net supply

• how to solve for the aggregate bond price  $q_t$ ?

### SOLVING FOR THE BOND PRICE

We could try to use an aggregate law of motion for  $q_t$ 

- 1. guess coefficients of  $q_t = q(S_t; \psi_q)$
- 2. solve individual problem
- 3. simulate economy
- 4. update coefficients  $\psi_q$  accordingly

So where's the problem?!

### **IMPOSING EQUILIBRIUM**

There is nothing ensuring zero net supply across iterations!

· departures from equilibrium are likely to accumulate!

Instead of  $q_t$  we can approximate something else to impose equilibrium

- $\cdot d(s_{i,t}) = b'(s_{i,t}) + q_t$ 
  - $q_t$  is aggregate bond price
  - $b'(s_{i,t})$  is individual bond choice

How does this help?!

- $q_t = \sum_i d(s_{i,t})$  in each period
- · i.e. are imposed to clear!

### TAKING STOCK

Challenges when solving heterogeneous agent models with aggregate uncertainty

- solving individual problem
- which moments to track
- how to update guesses
- how to simulate economy
- · when to stop iterating
- how to check accuracy

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**ALTERNATIVE SOLUTION METHOD** 

### MAIN IDEA

## Krusell-Smith algorithm is computationally costly

- simulation step takes time
- iterative updating takes time
- accuracy checks and moment selection take time

## Why do we need to do all the above?

- need to know the joint distribution of capital and productivity
  - distribution of infinitely-lived agents
  - hit by persistent idiosyncratic shocks
- · can we get around the above?

# ALTERNATIVE SOLUTION METHOD

MAIN IDEA

### **BASIC IDEA**

## Agent heterogeneity?

• what if we had (many) ex-ante heterogeneous agents (1)?

## Infinitely-lived agents?

• what if we consider an OLG framework instead (*T*)?

## What have these chagnes given us?

- ex-ante heterogeneity: finite number of types of agents
- · finite life-time: finite number of periods to track each type
- $\cdot \to T \times I$  types of agents

### **BASIC IDEA**

This is great, because  $T \times I$  is finite!

• moreover, the  $T \times I$  types describe the *entire* distribution!

What about the curse of dimensionality?

 $\cdot \to \mathsf{do}$  perturbation

No need for iterative procedure

- entire distribution described by transitions between types
  - i.e. aging of agents

# ALTERNATIVE SOLUTION METHOD

**DETAILS** 

### **DETAILS**

## Specific application: firm dynamics model

- there are *I types* of firms
  - productivity/demand heterogeneity (different long-run sizes)
- firms live for (at most) A periods
  - life-cycle heterogeneity within firm types

### **DETAILS CONT.**

Each age-type has a different value function!

$$V_{i,a}(S_t) = \pi_{i,a}(S_t) + \beta(1 - \delta_a) \mathbb{E} V_{i,a+1}(S_{t+1})$$

With

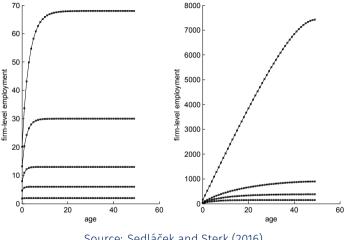
- $\delta$  is age-specific death rate
- $S_t$  being the aggregate state
  - · aggregate shocks, but also entire distribution of firms!

Keeping track of entire distribution is easy!

$$\omega_{i,a} = (1 - \delta)\omega_{i,a-1} \quad a \in (1,A]$$

 $\omega_{i,0}$  = free entry condition

### RESULTING ECONOMY HAS LOTS OF HETEROGENEITY



Source: Sedláček and Sterk (2016)

### RESULTING ECONOMY HAS LOTS OF HETEROGENEITY

**Table 1:** Employment share distribution by size and age

|                |       | data   |       |       | model  |       |
|----------------|-------|--------|-------|-------|--------|-------|
|                | small | medium | large | small | medium | large |
| 0 to 5 years   | 50    | 41     | 8     | 52    | 40     | 8     |
| 6 to 10 years  | 36    | 46     | 18    | 36    | 48     | 16    |
| 11 to 15 years | 29    | 45     | 26    | 27    | 46     | 27    |
| 16 to 20 years | 24    | 42     | 34    | 22    | 44     | 34    |
| 21 to 25 years | 18    | 39     | 43    | 18    | 39     | 43    |

Notes: Employment shares in percentages of small (1-19 employees), medium-sized (20-499) and large (500 and over) firms, by age. Data (averages) and models (steady states).

# **IMPLEMENTATION**

**ALTERNATIVE SOLUTION METHOD** 

### QUANTITATIVE IMPLEMENTATION

- · results in more than 900 state variables
  - · include the entire joint distribution of
  - firm employment and their masses
- solved with first-order perturbation
  - along steady state growth path
- solution takes several seconds
  - · fast enough that we can estimate parts of the model

### IMPLEMENTATION IN DYNARE

## Macro-language in Dynare

- · can define loops over certain criteria (e.g. types)
- · ideal for heterogeneous-agent setups
  - structure of first-order conditions the same across types
  - they differ in (some) parameter values
- other examples of its use include
  - multi-country models

TAKING STOCK

**ALTERNATIVE SOLUTION METHOD** 

### **TAKING STOCK**

## Solving models with ex-ante heterogeneity

- can be computationally easier (if linearizing)
- · contains the entire distribution of state-variables
- no need to revert to iterative procedures like Krusell-Smith
- not applicable to some models + question of "true" heterogeneity

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