

Modeling Development Priorities for the Heterogeneous Agents Resources and toolKit

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- ▶ NOT structural changes or object representation
- ▶ UNLESS closely related to a methodological issue

The Big Methodological Issues

The same methodological/numeric issues come up in a wide array of models; details differ on case-by-case basis

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6. Dimensionality “mismatch” between states and controls

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- ▶ But what about non-concave value functions?

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- ▶ White: Got that to work in 3D as well
- ▶ Still no way to get exact location of discontinuity in 2D+... and “approximate” discount might not be that precise!

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- ▶ I got nothin’

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- ▶ Carroll: Solution asymptotes from below to PF problem
- ▶ Argument is complicated by non-concave value
- ▶ Does limiting solution exist with discrete choice?

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- ▶ Requires computing expectations at a non-optimal candidate—what EGM tries to avoid?
- ▶ Examples: Portfolio choice, NEGM, choosing house size
- ▶ Sometimes requires using value function itself
- ▶ Maybe use implicit function theorem to get a “good guess”?

Major Modeling Areas

1. Human capital acquisition / education
2. Endogenous labor supply (and demand?)
3. Housing and durable goods
4. Health and insurance

Human Capital and Education (1/2)

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- ▶ Limits difficulties from discrete choice

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- ▶ What happens to workers if a sector experiences an exogenous shock?
- ▶ Re-education is costly, lifecycle considerations
- ▶ Seneviratne (2013): Discrete choice, discrete state
- ▶ HARK version: Convexify state space, but not choice space

Labor Supply Models (1/5)

1. Intensive margin (working version by T. Magne)
2. Extensive margin (versions by MNW and P. Mogensen)
3. Job search intensity
4. Incorporating into aggregate framework

Labor Supply Models (2/5)

Model of labor supply on intensive margin:

$$u(c, \ell) = ((1 - \ell)^\alpha c)^{1-\rho} / (1 - \rho),$$

$$v_t(b_t, \theta_t) = \max_{c_t, \ell_t} u(c_t, \ell_t) + \beta \mathbb{E}_t [(\psi_{t+1} \Gamma_t)^{1-\rho} v_{t+1}(b_{t+1}, \theta_{t+1})] \quad \text{s.t.}$$

$$y_t = \ell_t \theta_t, \quad \ell_t \in [0, 1],$$

$$a_t = m_t + y_t - c_t, \quad a_t \geq \underline{a},$$

$$b_{t+1} = R / (\Gamma_t \psi_{t+1}) a_t,$$

$$\psi_{t+1} \sim F_{\psi_{t+1}}(\psi), \quad \theta_{t+1} \sim F_{\theta_{t+1}}(\theta), \quad \mathbb{E}[\psi_{t+1}] = 1.$$

Labor Supply Models (3/5)

Model of labor supply on extensive margin:

$$u(c, \ell) = c^{1-\rho}/(1-\rho) - \alpha\ell,$$

$$v_t(b_t, \theta_t, \ell_{t-1}) = \max_{c_t, \ell_t} u(c_t, \ell_t) + \beta \mathbb{E}_t [(\psi_{t+1} \Gamma_t)^{1-\rho} v_{t+1}(b_{t+1}, \theta_{t+1}, \ell_t)]$$

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Labor Supply Models (4/5)

Model of endogenous employment search:

$$u(c, s) = ((1 - s)^\alpha c)^{1-\rho} / (1 - \rho),$$

$$v_t(m_t, e_t) = \max_{c_t, s_t} u(c_t, s_t) + \beta \mathbb{E}_t [(\psi_{t+1} \Gamma_t)^{1-\rho} v_{t+1}(m_{t+1}, e_{t+1})] \quad \text{s.t.}$$

$$a_t = m_t - c_t, \quad a_t \geq \underline{a}, \quad s_t \in [0, 1],$$

$$m_{t+1} = R/(\Gamma_t^e \psi_{t+1}) a_t + \theta_t e_{t+1} + \underline{b}(1 - e_{t+1}),$$

$$\text{Prob}(e_{t+1} = 1 | e_t = 0) = s_t, \quad \text{Prob}(e_{t+1} = 0 | e_t = 1) = \mathbb{U},$$

$$\psi_{t+1} \sim F_{\psi_{t+1}}^e(\psi), \quad \theta_{t+1} \sim F_{\theta_{t+1}}(\theta), \quad \mathbb{E}[\psi_t] = 1.$$

Labor Supply Models (5/5)

Applications of Market for labor models:

- ▶ Non-trivial calculation of $L_t = \int_0^1 \ell_{it} P_{it} \theta_{it} di$ for Cobb-Douglas
- ▶ Disutility of employment search and probability of job loss depend on labor market slackness
- ▶ Can look at behavior in response to change in Social Security

Durable Goods: Housing and Health (1/4)

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- ▶ Changing level of d_t might involve fixed cost
- ▶ Generates “region of inaction” / inactivity constraint
- ▶ Grossman (1972): Health is a capital stock, can't disinvest
- ▶ Health produces longevity, reduces disutility of work

Durable Goods: Housing and Health (2/4)

General (housing) durable goods model:

$$u(c, d) = (c^\alpha, d^{1-\alpha})^{1-\rho} / (1 - \rho).$$

$$v_t(m_t, d_t) = \max_{c_t, i_t} u(c_t, d_t) + \beta \mathbb{E}_t [(\psi_{t+1} \Gamma_t)^{1-\rho} v_{t+1}(m_{t+1}, d_{t+1})] \quad \text{s.t.}$$

$$a_t = m_t - c_t - i_t, \quad a_t \geq \underline{a},$$

$$D_t = d_t + g(i_t), \quad d_{t+1} = (1 - \delta_{t+1})D_t, \quad \delta_{t+1} \sim F_\delta(\delta),$$

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Durable Goods: Housing and Health (3/4)

Variations of durable goods model require different solvers:

- ▶ Easiest case: $g(i_t)$ is convex, $i_t \in \mathbb{R}$. Every end-of-period state (a_t, D_t) associated with *some* beginning-of-period state.
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- ▶ Just ugh: $g(i_t) = i_t/\pi_t + K\mathbf{1}(i_t \neq 0)$, $i_t \geq 0$.

Durable Goods: Housing and Health (4/4)

Applications for Market with housing durable goods:

- ▶ Endogenous pricing of durable good: housing market
- ▶ Dynamics of demand for durables after an aggregate shock

Next steps for health models:

- ▶ White (2018) does health investment model with exogenous income (retired people only)
- ▶ Can add income process, interact with labor supply decision
- ▶ Health insurance and endogenous health status

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- ▶ Take my Topics in Dynamic Modeling course at UDel!