

# FIXED COST MODELS

Juan Herreño   Johannes Wieland

UCSD, Spring 2024

# OUTLINE

- 1 INTRODUCTION
- 2 BASELINE FC MODEL
- 3 CONTINUOUS TIME

# OUTLINE

- 1 INTRODUCTION
- 2 BASELINE FC MODEL
- 3 CONTINUOUS TIME

# FIXED COST MODELS

- Lots of behavior at the micro level is lumpy: buying houses, cars, setting prices, investment, information acquisition.
- Fixed cost models naturally give rise to such lumpiness.
- Fixed cost models are hard to solve because they are non-convex, non-differentiable, and in GE have to carry distribution around.
- But lots of interesting economics: state-dependence, size-dependence, policy targeting.
- Today we will see how to solve these models within SSJ routines.

# FIXED COST MODELS: EXAMPLES

- Does lumpy price setting behavior matter? lots from Alvarez-Lippi; Auclert, Rigato, Rognlie, Straub (2024).
- Do fixed cost frictions matter? Kahn-Thomas, Winberry (2021), Koby-Wolf (2020), Bailey-Blanco (2021).
- Lumpy durables and monetary transmission: McKay and Wieland (2021, 2022).

# OUTLINE

- 1 INTRODUCTION
- 2 BASELINE FC MODEL**
- 3 CONTINUOUS TIME

# ENVIRONMENT

- Liquid assets  $a$
- Durable stock  $d$
- Nondurable consumption  $c$
- Income  $y$
- Real rate  $r$
- Depreciation rate  $\delta$
- Fixed cost  $f$

# CONSTRAINTS

- Budget constraint if adjusting:

$$\frac{a'}{1+r} + c' + pd' = a + (1-\delta)(1-f)pd + y \equiv x_{adj}$$

- Budget constraint if not adjusting with  $d' = (1-\delta)d$

$$\frac{a'}{1+r} + c' = a + y \equiv x_{noadj}$$

- Borrowing constraint

$$a' \geq 0$$



# VALUE FUNCTIONS

- Let the value function be

$$V(y, d, a, \varepsilon) = \max\{V^{adj}(y, d, a) + \varepsilon^{adj}, V^{noadj}(y, d, a) + \varepsilon^{noadj}\}$$

- $\varepsilon^i$  are drawn from a Gumbell distribution with standard deviation  $\sigma_V \frac{\pi}{\sqrt{6}},^1$ .

Define the post-adjustment value function as

$$W(y, d', a') \equiv E_y \tilde{V}(y', d', a')$$

- Value from adjusting and not adjusting:

$$V^{adj}(y, d, a) = \max_{c' + d' + \frac{a'}{1+r} \leq x_{adj}} [\psi \ln c' + (1 - \psi) \ln d' + W(y, d', a')],$$

$$V^{noadj}(y, d, a) = \max_{c' + \frac{a'}{1+r} \leq x_{noadj}} [\psi \ln c' + (1 - \psi) \ln(1 - \delta)d + W(y, (1 - \delta)d, a')]$$

---

<sup>1</sup><https://eml.berkeley.edu/choice2/ch3.pdf>

# ADJUSTMENT PROBABILITIES AND EXPECTED VALUES

- The probability of adjustment is

$$adjust(y, d, a) = \frac{\exp\{(V^{adj}(y, d, a) - V^{noadj}(y, d, a))/\sigma_V\}}{1 + \exp\{(V^{adj}(y, d, a) - V^{noadj}(y, d, a))/\sigma_V\}}$$

- The distribution of  $V$  is

$$\begin{aligned} Prob[V \leq x] &= Prob[\varepsilon^{adj} \leq x - V^{adj}(y, d, a)] \\ &\quad \times Prob[\varepsilon^{noadj} \leq x - V^{noadj}(y, d, a)] \\ &= \exp\{-\exp[-(x - \sigma_V \log\{\exp[V^{adj}(y, d, a)/\sigma_V] \\ &\quad + \exp[V^{noadj}(y, d, a)/\sigma_V]\})/\sigma_V]\} \end{aligned}$$

- The expected value is,

$$\begin{aligned} \tilde{V}(y, d, a) &\equiv E_\varepsilon V(y, d, a, \varepsilon) \\ &= V^{noadj}(y, d, a) - \sigma_V \log noadjust(y, d, a) + \sigma_V \gamma \end{aligned}$$

where  $\gamma \approx 0.5772$  is the Euler-Mascheroni constant. If  $\sigma_V \rightarrow 0$ , then

$$\tilde{V}(y, d, a) = \max\{V^{adj}(y, d, a), V^{noadj}(y, d, a)\}$$

## NO ADJUSTMENT PROBLEM

- Break up into sequential problem of choosing  $c'$ ,  $a'$  given  $d'$ . Then choose  $d'$ .
- For given choice of  $d'$ :

$$V^{noadj}(y, n, m) = \max_{c', a'} [\psi \ln c' + (1 - \psi) \ln d' + \beta W(y', d', a')]$$

$$d' = n$$

$$a' = [m - c']$$

$$a' \geq 0$$

where  $m \geq 0$ .

- First order conditions:

$$V_m^{noadj}(y, n, m) = \beta W_a(y', d', a') + \zeta$$

$$V_n^{noadj}(y, n, m) = \frac{(1 - \psi)}{d'} + \beta W_d(y', d', a')$$

$$\frac{\psi}{c'} = \beta W_a(y', d', a') + \zeta$$

# EGM

- We are given an initial guess  $W_a(y', d', a')$ .
- First, assume borrowing constraint is not binding and solve for  $c'$  in

$$\frac{\psi}{c'} = \beta W_a(y', d', a')$$

- Then calculate implied cash on hand

$$m = c' + a'$$

- The problem is not necessarily concave, so there may be multiple combinations of  $(c', a')$  that map into the same grid point  $m$ .
  - ▶ In this case we have to check which solution yields highest utility and discard the others.
  - ▶ The "func\_upper\_envelop.py" function performs this task.

# EGM

- Interpolate the decision rules from the  $m$  grid onto the grid for  $a$ .
- Then check if implied solution for  $a'$  violates the borrowing constraint. If so implement borrowing constrained solution  $a' = 0$  and  $c' = m$ .
- Combining the two solutions yields the policy functions

$$c^{noadj}(y, n, m), a^{noadj}(y, n, m)$$

- The no adjustment solutions use  $n = (1 - \delta)d$  and  $m = a + y$

$$c^{noadj}(y, (1 - \delta)d, m)$$

$$a^{noadj}(y, (1 - \delta)d, m)$$

$$d^{noadj} = (1 - \delta)d$$

## ADJUSTMENT PROBLEM

- Given some amount of cash on hand  $x$ :

$$V^{adj}(y, x) = \max_{d'} V^{noadj}(y, d', m')$$

$$m' = x + y - pd'$$

$$m' \geq 0$$

- FOC

$$V_x^{adj}(y, x) = V_m^{noadj}(y, d', m') + \kappa$$

$$V_n^{noadj}(y, d', m') = pV_m^{noadj}(y, d', m') + \kappa$$

- Constraint will never be binding so  $\kappa = 0$ .
- Combine with keep FOC to get

$$p \frac{\psi}{c^{noadj}(y, d', m')} = \frac{1 - \psi}{d'} + \beta W_d(y, d', a')$$

- Again must solve for upper envelope given non-concavity.

# ADJUSTMENT SOLUTION

- This yields adjustment solutions

$$c^{adj}(y, x), a^{adj}(y, x), d^{adj}(y, x)$$

and a value function

$$V^{adj}(y, x) = \psi \ln c^{adj}(y, x) + (1 - \psi) \ln d^{adj}(y, x) \\ + \beta W(y', d^{adj}(y, x), a^{adj}(y, x))$$

- We interpolate onto the existing grid using  $x = a + (1 - \delta)(1 - f)pd$  to get

$$c^{adj}(y, d, a), a^{adj}(y, d, a), d^{adj}(y, d, a)$$

and value function

$$V^{adj}(y, d, a) = \psi \ln c^{adj}(a, d, y) + (1 - \psi) \ln d^{adj}(a, d, y) + \beta W(y', d^{adj}(a,$$

# COMBINED PROBLEM

- Get policy functions by combining adjust and no-adjust solutions.

$$c'(y, d, a) = \text{adjust}(a, d, y)c^{adj}(a, d, y) + (1 - \text{adjust}(a, d, y))c^{noadj}(a, d, y)$$

$$a'(y, d, a) = \text{adjust}(a, d, y)a^{adj}(a, d, y) + (1 - \text{adjust}(a, d, y))a^{noadj}(a, d, y)$$

$$d'(y, d, a) = \text{adjust}(a, d, y)d^{adj}(a, d, y) + (1 - \text{adjust}(a, d, y))d^{noadj}(a, d, y)$$



# OUTLINE

- 1 INTRODUCTION
- 2 BASELINE FC MODEL
- 3 CONTINUOUS TIME**

# CONTINUOUS TIME

- Alternatively can formulate the problem in continuous time.
- Continuous time generally very good for handling stopping time problems, such as when to adjust.
- With discretized value function can write problem as

$$\min\{\rho v - u(v) - A(v)v, v - v^*(v)\} = 0$$

- ▶ First term is the no-adjustment utility flow.
  - ▶ Second term is the adjustment choice.
- Intuition: if  $v < v^*(v)$  then do not adjust and get flow value. If adjust  $v = v^*(v)$  since the flow value  $u(v) + A(v)v < \rho v$ .

# CONSIDERATIONS

- To solve continuous time problem see Ben Moll's code on stopping time problems.
- In principle continuous time better: do not require sufficiently large noise shocks to make value function differentiable.
- But harder to then apply discrete time sequence space methods. Even after aggregation, the distribution of states jumps after a shock so cannot use the fake news algorithm as implemented. (See McKay-Wieland 2021.)
- Possible that continuous time SSJ could be used. But need to handle the immediate jumps in the distribution.