

# Heterogeneity and incomplete markets: a dual approach

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- Broad goal: understand heterogeneity + incomplete markets in macroeconomics.
- Prominent example: Krusell and Smith (1998), growth model with aggregate shocks and idiosyncratic labor shocks.
- When is heterogeneity important? When do we obtain approximate aggregation?
- Computational challenges of high dimensional state-spaces.
- I present a different way to look at these class of problems and begin to explore its usefulness.

# A simple example: prices

- Uncertainty tree:



- One asset with price  $(B_1, B_2)$  has span:

$$M_C = \{c \in \mathbb{R}^2 \mid c_1 = -\theta_1 B_1, c_2 = \theta_1 B_2 \text{ for some } \theta_1 \in \mathbb{R}\}$$

- No assets:

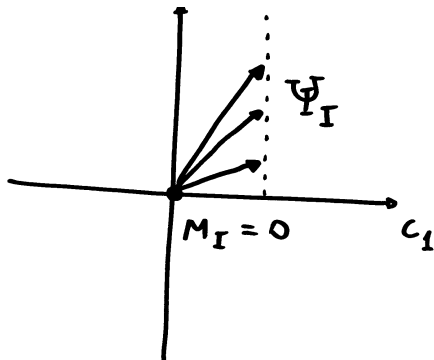
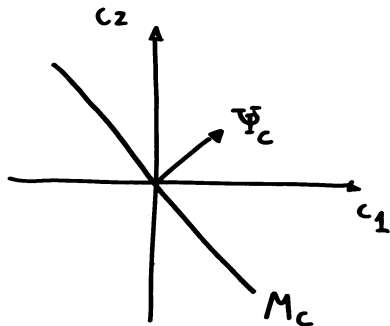
$$M_I = (0, 0).$$

- State or Arrow-Debreu prices:

$$\Psi = \{\psi \in \mathbb{R}_{++}^2 \mid \psi_1 c_1 + \psi_2 c_2 = \psi \cdot c = 0 \text{ for all } c \in M \text{ and } \psi_1 = 1\}$$

- $\Psi_C = \left(1, \frac{B_1}{B_2}\right)$  and  $\Psi_I = (1, \text{any } \psi_2 > 0)$

## A simple example: prices



# A simple example: consumers

- One consumer with endowment process  $e \in \mathbb{R}_+^2$ ,  $e \neq 0$ .
- Utility maximization program is:

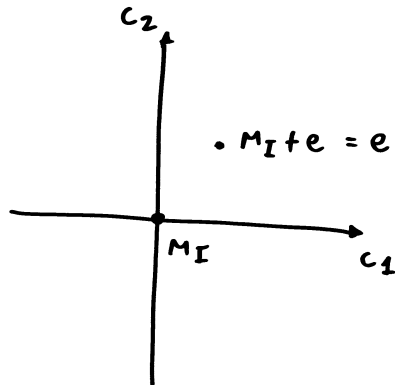
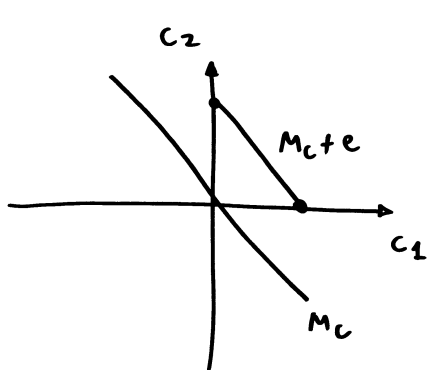
$$\begin{aligned} U^* &= \max_c U(c) = \max_{c_1, c_2} \log(c_1) + \log(c_2) \\ &\quad s.t. \\ c &\in M + e = \left\{ c \in \mathbb{R}_+^2 \mid c - e \in M \right\} \end{aligned}$$

- For complete markets, the dynamic budget constraint is

$$M_C + e = \left\{ \begin{array}{l} c_1 = -\theta_1 B_1 + e_1 \\ c_2 = \theta_1 B_2 + e_2 \\ c_1, c_2 \geq 0 \end{array} \right\} = \left\{ \begin{array}{l} \frac{(c_1 - e_1)}{B_1} + \frac{(c_2 - e_2)}{B_2} = 0 \\ c_1, c_2 \geq 0 \end{array} \right\}$$

- For incomplete markets:  $c_1 = e_1$  and  $c_2 = e_2$ .

## A simple example: consumers



# A simple example: static budget constraint

- For a given  $\psi \in \mathbb{R}^2$ , the static budget constraint is

$$B_{\psi,e} = \left\{ c \in \mathbb{R}_+^2 \mid c_1 + \psi_2 c_2 = e_1 + \psi_2 e_2 \right\}$$

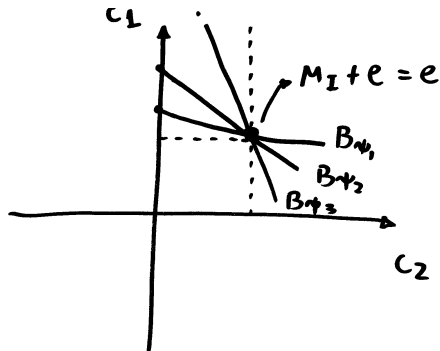
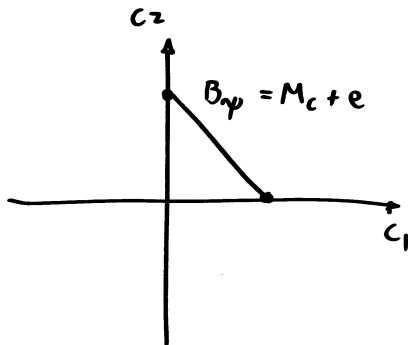
- For both complete and incomplete markets, we have

$$\bigcap_{\psi \in \Psi} B_{\psi,e} = M + e$$

- For complete markets,  $\Psi_C$  has a single element. For incomplete markets

$$\bigcap_{\psi \in \Psi} B_{\psi,e} = \bigcap_{\psi_2 > 0} \left\{ c \in \mathbb{R}_+^2 \mid c_1 - e_1 = -\psi_2 (c_2 - e_2) \right\}$$

# A simple example: static and dynamic budget constraints





# A simple example: the dual

- We can now solve one of two problems:

$$\begin{array}{ll} U^* = \max_c U(c) & U_{\Psi}^* = \max_c U(c) \\ \text{s.t. } c \in M + e & \text{s.t. } c \in \cap_{\psi \in \Psi} B_{\psi, e} \end{array} \quad \text{or}$$

- I will show we have a third way, the "dual"

$$\inf_{\psi \in \Psi} U_{\psi}^*$$

where

$$U_{\psi}^* = \max_c U(c)$$

$$\text{s.t. } c \in B_{\psi, e}$$

- Thus, we solve in two steps: first, solve the easy static problem  $U_{\psi}^*$  for given state prices. Second, minimize over all allowable ones.

# A simple example: solving the dual

- For incomplete markets

$$U_{\psi}^* = \max_{c_1, c_2} \log(c_1) + \log(c_2)$$

s.t.

$$c_1 + \psi_2 c_2 = e_1 + \psi_2 e_2$$

- The Lagrangean is  $\mathcal{L} = \log(c_1) + \log(c_2) + \lambda (e_1 + \psi_2 e_2 - c_1 - \psi_2 c_2)$  and the FOC give

$$0 = \nabla \mathcal{L}(c^*) = \nabla U(c^*) - \lambda \psi = \left( \frac{1}{c_1^*}, \frac{1}{c_2^*} \right) - \lambda (1, \psi_2)$$

# A simple example: solving the dual

- Then the dual is

$$\inf_{\psi \in \Psi} U_{\psi}^* = \inf_{\psi_2 > 0} \log \left( \frac{e_1 + \psi_2 e_2}{2} \right) + \log \left( \frac{e_1 + \psi_2 e_2}{2} \frac{1}{\psi_2} \right)$$

which indeed gives  $c = e$ .

# The dual: proof

- First problem: what is the right underlying state space  $S$ ?
- A consumption plan is a real number on each node:  $c \in \mathbb{R}^S$ , and  $S$  is "big".
- For the proof, it is enough that  $\mathbb{R}^S$  has a dot product, i.e. you can "sum" over all nodes (still, unclear how to do this sometimes).

# The dual: proof

- Since  $M + e \subseteq B_{\psi,e}$  for any  $\psi \in \Psi$ ,

$$M + e \subseteq \bigcap_{\psi \in \Psi} B_{\psi,e} \subseteq B_{\tilde{\psi},e}$$

for any  $\tilde{\psi} \in \Psi$ , so

$$U^* \leq U_{\Psi}^* \leq \inf_{\psi \in \Psi} U_{\psi}^*.$$

- If I show  $\inf_{\psi \in \Psi} U_{\psi}^* \leq U^*$ , then we're done.

# The dual: proof

- If  $c^*$  is an interior solution to  $U^*$

$$\nabla U(c^*) \cdot (c - c^*) = 0 \text{ for all } c \in M + e$$

or

$$\nabla U(c^*) \cdot c = 0 \text{ for all } c \in M$$

- This shows  $\nabla U(c^*) = \psi^* \in \Psi$ , so  $c^*$  is the solution to  $U_{\psi^*}^*$ .
- Finally

$$\inf_{\psi \in \Psi} U_{\psi}^* \leq U_{\psi^*}^* = U^*.$$

- Consumer's maximize:

$$U_i^* = \max_{c, k} E_0 \sum_{t=0}^{\infty} \beta^t \log c_t$$

s.t

$$\begin{aligned} c_t + k_{t+1} &= [r_t + 1 - \delta] k_t + w_t \varepsilon_t^i \\ c, k &\geq 0 \\ k_0^i &\text{ given.} \end{aligned}$$

# KS: prices and market clearing

- Prices are given by firm maximization:

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

$$r = Y_K$$

$$w = Y_L$$

- Market clearing:

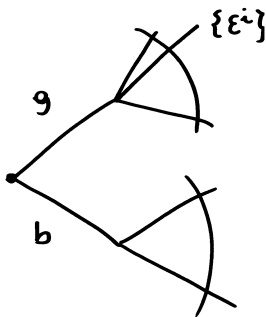
$$\int k_t^i di = K_t$$

$$\int \varepsilon_t^i di = L_t$$



# Uncertainty

- Aggregate shocks:  $z_t \in \{z_g, z_b\}$  follows a Markov structure.
- Idiosyncratic shocks:  $\varepsilon_t^i \in \{0, 1\}$  such that  $L_t = L_g$  if  $z_t = z_g$  and  $L_t = L_b$  if  $z_t = z_b$ .
- Conditional on  $z_t$ , the shocks  $\varepsilon_{it}$  are uncorrelated.
- Uncertainty tree:



# Approximate aggregation

- Krusell and Smith assume agents perceive a simple law of motion for the aggregate state:

$$\log K_{t+1} = a + b \log K_t$$

- Benefits:
  - State variables go from  $(k_{it}, \varepsilon_{it}; \Gamma_t, z_t)$  to  $(k_{it}, \varepsilon_{it}; K_t, z_t)$ , where  $\Gamma_t$  is the entire distribution of capital (and is infinite-dimensional).
  - Excellent goodness of fit for most cases: approximate aggregation.

# The dual in KS: static problem

- The static part is

$$\begin{aligned} U_{\psi}^* &= \max_c E_0 \sum_{t=0}^{\infty} \beta^t \log(c_t) \\ &\quad \text{s.t.} \\ \psi \cdot c &= k_0 + \psi \cdot e \end{aligned}$$

where  $e_t = w_t \varepsilon_t$ .

- FOC:

$$\begin{aligned} 0 &= \nabla U(c^*) - \lambda \psi \\ c_t^* &= (k_0 + \psi \cdot e) (1 - \beta) \frac{\beta^t p_t}{\psi_t} \end{aligned}$$

# The dual in KS: minimization

- The second step is to solve:  $U^* = \inf_{\psi \in \Psi} U_{\psi}^*$
- What is  $\Psi$ ? Its elements have to price capital, the only asset:

$$\begin{aligned}\psi_0 &= 1 \\ \psi_t(s^t) &= \int R_t(s^{t+1} | s^t) \psi_t(s^{t+1} | s^t) dP\end{aligned}$$

- This is simply the Euler equation  $1 = E_t [\text{stoch disc} * \text{returns}]$
- Now solve

$$\begin{aligned}U^* &= \inf_{\psi} E_0 \sum_{t=0}^{\infty} \beta^t \log \left( (1 - \beta) (k_0 + \psi \cdot e) \frac{\beta^t p_t}{\psi_t} \right) \\ &\text{s.t.} \\ \psi_t &= \int_{s^{t+1}|s^t} R_{t+1} \psi_{t+1} dP\end{aligned}$$

# The dual in KS: minimization

- FOC give dynamics of  $\psi$  state by state:

$$\frac{e_t}{(1 - \beta) (k_0 + \psi \cdot e)} - \frac{\beta^t p_t}{\psi_t} = \lambda_t - \lambda_{t-1} p_{t-1} (z_t) R_t$$

where  $\lambda_t$  is the Lagrange multiplier associated to the constraint for  $\psi_t$ .

- The state is given by  $(k_t^i, \varepsilon_t^i; \Gamma_t, z_t)$ , where  $\Gamma_t$  is the distribution of capital.
- Aggregation: how do decisions change if  $\Gamma_t$  changes, keeping  $(k_t^i, \varepsilon_t^i; K_t, z_t)$  constant?

# The dual in KS: aggregation

- For two states  $s^t$  and  $\tilde{s}^t$  that share  $s^{t-1}$  and have

$$\left(k_t^i, \varepsilon_t^i; K_t, z_t\right) = \left(\tilde{k}_t^i, \tilde{\varepsilon}_t^i; \tilde{K}_t, \tilde{z}_t\right)$$

the FOC gives:

$$c_t(s^t) + \lambda_t(s^t) = c_t(\tilde{s}^t) + \lambda_t(\tilde{s}^t).$$

- The Lagrange multiplier  $\lambda_t$  captures the effect of changing the distribution of wealth  $\Gamma_t$ , keeping aggregate capital constant.
- If  $\lambda_t = 0$ , there would be perfect aggregation.

# The dual in KS: aggregation

- If now aggregate productivity is also different  $z_t = z_g$  but  $\tilde{z}_t = z_b$

$$c_t(s^t) + \lambda_t(s^t) = c_t(\tilde{s}^t) + \lambda_t(\tilde{s}^t) + a [c_t(\tilde{s}^t) + \lambda_t(\tilde{s}^t)] + b K_t^\alpha \varepsilon_t^i$$

for two constants  $a, b$ .

- These expressions give the value of good vs. bad aggregate state, employment vs. unemployment.

# Conclusion

- A dual approach to problems with heterogeneity and incomplete markets could be useful to understand aggregation.