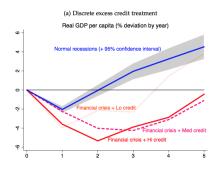
Credit Cycles, Market Liquidity and Heterogeneous Firms

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Introduction – Motivation



Importance of credit for recessions - from Jordà, Schularick, and Taylor (2013)

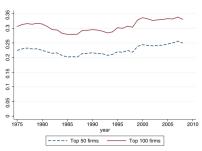


FIGURE 1.—Sum of the sales of the top 50 and 100 non-oil firms in Compustat, as a fraction of GDP. Hulten's theorem (Appendix B) motivates the use of sales rather than value added.

Granularity, top firms matter for aggregates - from Gabaix (2011)

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 - Are recessions amplified by the collapse of large companies ...
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 - *Transmission mechanisms* (does aggregation holds/distribution matters?)

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 - Transmission mechanisms (does aggregation holds/distribution matters?)
- \Rightarrow Aim of the paper:

Examine the effects of firm heterogeneity on aggregate fluctuations in presence of different dimensions of (il)liquidity

- ▶ What is the impact of firm heterogeneity on aggregate fluctuations in presence of two dimensions of (il)liquidity?
- ► Financial (il)liquidity: Two dimensions
 - Market illiquidity: Asset and capital lose value when sold (asset)
 - Funding illiquidity: Difficult to raise funds for investment (liability)
 - Consequence : Deleveraging, credit/asset-price feedback loop
- Firms heterogeneity:
 - Dispersion and *power-law* distribution
 - · Hedging both idiosyncratic & aggregate risk in incomplete market
 - Credit and Collateral constraints

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 - Credit and Collateral constraints
- \Rightarrow Aim: Study of the three dimensions together
- Propose a theoretical framework with heterogeneous firms, collateral constraints and market illiquidity

Research question – Theoretical literature

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 - Macro-Finance : both Market and funding illiquidity
 - Financial frictions and credit cycles: BGG (99), Kiyotaki Moore (JPE 1997), Brunnermeier and Sannikov (2014) and many others ...
 - Heterogeneous agents :
 - Households: Aiyagari (94), Kaplan-Moll-Violante (2017), Benhabib,
 Bisin, and Zhu (2015), Achdou, Han, Lasry, Lions, and Moll (2017)
 - Firms: Moll (2014), Winberry (2016a), Mongey and Williams (2017),
 Khan and Thomas (2013) (JPE)
 - With aggregate shocks: Krusell and Smith (1998) (JPE), Winberry (2016b), Ahn, Kaplan, Moll, Winberry, and Wolf (2018)

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- Heterogeneous entrepreneur :
 - Capital k_t subject to income shocks (CRS prod.)

$$dR_t = \bar{R} k_t dZ_t^i$$

- Z_t productivity : Z_t idiosyncratic shock : $dZ_t^i = \mu^i(z_t)dt + \sigma_t dW_t^i$
- $-z_t$ (and $\mu^i(z_t)$): jump on productivity growth: n_z states process: here, 3-states: low state, median state, high state
- Aggregate shock on average return \bar{R} (more on this later):

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 - Risk-free Bond b_t at rate r_t vs. Risky Capital k_t at rate R_t

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- Two assets portfolio choice (à la Merton 73):
 - Risk-free Bond b_t at rate r_t vs. Risky Capital k_t at rate R_t
- Incomplete market : cannot self-insure against any risk
- Collateral constraints and shocks on capital quality θ_t^c :

$$b_t \ge -\theta_t^c \, q_t \, k_t \qquad \Rightarrow \qquad k_t \le \frac{1}{(1 - \theta_t^c) q_t} \, a_t$$

- Reduce to one state-variable (asset/networth) $a_t = b_t + q_t k_t$
- Optimization problem (without agg. risk)

$$\max_{\{c_t, k_t\}_{t_0}^{\infty}} \mathbb{E}_{t_0} \int_{t_0}^{\infty} e^{-\rho_t t} u(c_t) dt$$

$$da_t = \left[r_t a_t - c_t + q_t k_t \left(\bar{R} \mu^i(z_t) - r_t \right) \right] dt + q_t k_t \bar{R} \sigma_t^i dW_t^i \qquad (P)$$

$$0 \le k_t \le a_t / (1 - \theta_t^c) q_t \qquad a_t \ge 0$$

- Two states variables : assets a_t (size) and productivity z_t
- Portfolio : capital k_t /bonds b_t determined endogenously
- Typical heterogenous agents problem (Mean Field Game)
 - (i) Optimal choices given by the Hamilton Jacobi Bellman (HJB)
 - (ii) Law of motion of the distribution by Kolmogorov Forward (KF)

- Problem analogous to :
 - Bewley-Huggett-Aiyagari, with collateral constraint :
 - generates high mass of firm m(a, z) on the left tail of the distribution
 - Merton portfolio choice for large firms :
 - generates power law $m(a, z) \sim a^{-\zeta 1}$ on right tail of distribution

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- Coupling: strategic complementarity/pecuniary externality
 - Risk-free interest r_t determined by market clearing

$$\int_{\mathcal{A}\times\mathcal{Z}} b_t(a,z) \, m(t,da,dz) = \underline{B} = 0$$
with $b_t(a,z) = a - q_t k_t(a,z)$

- New element : Market liquidity
 - Endogenous asset price q_t

A model of heterogeneous firms

Market liquidity: asset pricing

Market liquidity: asset pricing

- **Determination** of asset price q_t : different solutions
 - Incomplete market : no hedging by firms
 - One price for capital that has different return for different firms :

$$\mathbb{E}^{m}[R_{t}] = \int_{\mathcal{A}\times\mathcal{Z}} \bar{R} \, k_{t}(a,z) \mu^{i}(z) \, m(t,da,dz)$$

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$$dq_t = q_t \, \mu_t^e \, dt + q_t \, \sigma_t^{c,e} \, \varepsilon_t^c \, dN_t^c$$

Market liquidity mechanism:
 aggregate fluctuations in q_t exacerbate exogenous shocks

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- Market liquidity mechanism:
 aggregate fluctuations in q_t exacerbate exogenous shocks
- **Determination** of q_t : No arbitrage :
 - Between risk-free rate r_t and a "market portfolio" of all firms :

$$q_t = \mathbb{E}\Big[\int_t^\infty e^{-r_{\tau}(\tau-t)} \mathbb{E}^m[R_{\tau}] d au\Big]$$

$$dq_t = \Big(q_t r_t - \mathbb{E}^m[R_{\tau}]\Big) dt \qquad \text{without aggr. risk yet}$$

• Aggregate shocks : Jump-drift processes (jump dN^c) of size ε

- ► Treated as unexpected (zero probability) events
- ► Stationary equilibrium without aggregate shocks :

$$\sigma_R^c = \sigma_\sigma^c = \sigma_\theta^c = \sigma_\rho^c = 0$$

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- ► Change in average return of all firms

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- RBC-style exogenous supply shock
- Here, affect the return of all entrepreneurs to capital

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- RBC-style exogenous supply shock
- Here, affect the return of all entrepreneurs to capital
- Change in patience/discount factor

$$d\rho_t = \eta(\mu^{\rho} - \rho_t)dt + \sigma_{\rho}^c \,\varepsilon_t^c dN_t^c$$

- NK-style aggr. demand shock (e.g. Eggertsson and Krugman (2012))
- Here, affect the patience of experts and saving motive
- Treated as unexpected (zero probability) events
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└ Aggregate risk

Aggregate shocks – 2

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Credit Cycles, Market Liquidity & Heterogeneous firms

- Aggregate shocks: Jump-drift processes (jump dN^c) of size ε
- Capital quality shocks (experts)

$$d\theta_t^c = \eta(\mu^{\theta} - \theta_t^c)dt + \sigma_{\theta}^c \varepsilon_t^c dN_t^c$$

- Driver of the Great Recession according to Khan and Thomas (2013) and Jermann and Quadrini (2012)
- Implies deleveraging: large drop in investment by constrained firms

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- E.g. Bloom (2009) and more recently Mongey and Williams (2017)
- Induces "wait-and-see" effect (and reduction investment) but more dispersion in firms: ambiguous effects on the distribution of firms
- Treated as unexpected (zero probability) events
- Stationary equilibrium without aggregate shocks :

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► Stationary MFG system without aggregate shocks :

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 - Hamilton-Jacobi-Bellman equation : \Rightarrow pin down the value v(t, a, z) and choices of firms

$$-\partial_{t}v(t,a,z) + \rho v(t,a,z) = \max_{\substack{c \geq 0, \ a \geq 0 \\ 0 \leq k_{t} \leq a/(1-\theta')a}} u(c) + \partial_{a}v(t,a,z) \left[ra - c + (\bar{R}\mu(z) - r) \, q \, k \right]$$
 (HJB)

$$+ \Delta_{a}v(t,a,z) \left(q_{t}k\bar{R} \, \sigma_{t}^{i} \right)^{2} / 2 + \sum_{z'}^{n_{z}} \lambda_{z'} \left(v(t,a,z') - v(t,a,z) \right)$$

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- Kolmogorov Forward : ⇒ pin down the measure of agents m(t, a, z)

$$\partial_{t} m(t,a,z) = -\partial_{a} \left[m(t,a,z) \left(r_{t} a - c_{t}^{\star}(a,z) + \left(\bar{R} \mu(z) - r \right) q_{t} k_{t}^{\star}(a,z) \right] \right]$$

$$+ \partial_{a}^{2} \left[m(t,a,z) \left(q_{t} k_{t}^{\star}(a,z) \bar{R} \sigma^{i} \right)^{2} / 2 \right] - \lambda_{z} m(t,a,z) + \sum_{z' \neq z}^{n_{z}} \lambda_{z'} m(t,a,z')$$

$$\forall (t,a,z) \in [t_{0},\infty] \times \mathcal{A} \times \mathcal{Z} = [t_{0},\infty] \times [0,\infty) \times \{z_{1} \dots z_{n}\}$$

- A competitive industry stationary equilibrium (without aggregate shocks) is a set of prices $\{r_t, q_t\}_t$ such that :
 - The firms value and optimal policies c_t, k_t, b_t are given by the solution of HJB
 - The firms distribution $m_t(da, dz)$ is given by the solution of KF
 - Bond market clears $\int_{A \vee Z} b_t(a, z) m_t(da, dz) = \underline{B}$
 - Asset market is priced by no arbitrage $r_t dt = \frac{dq_t + \mathbb{E}[R_t]}{q_t}$
- Extensions :
 - Alternative way of pricing the asset q_t
 - General equilibrium : Household consuming and pricing the asset
 - Government policy: stimulus and subsidy to demand
 - Entry and exit : deleveraging + default

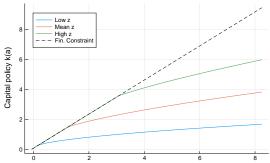
- ► Characterization of value/policies functions $c^*(a,z)$, $s^*(a,z)$ & $k^*(a,z)$
 - **Result 1 :** For each productivity $z \in \mathcal{Z}$, there is an interval over size (asset) such that all firms are constrained & have the same policy :

$$a \in [0, \gamma(\theta^c, q, z)]$$
 \Rightarrow $k(a, z) = k(a) = \frac{a}{(1 - \theta^c)q}$

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Capital choice at the left tail and financial constraint

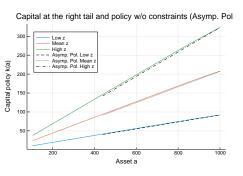


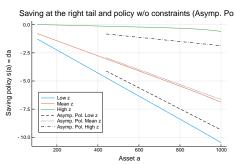
- Policy functions : c(a, z), s(a, z) and k(a, z) at the upper tail : linearity in state
 - **Prop. 1:** In stat. eq., with CRRA = γ , and no jump in level of productivity growth, i.e. $\lambda_z = 0 \ \forall z$, firms consumption, saving and capital investment policies are asymptotically linear in wealth a, as wealth grows large: $a \to \infty$:

$$\begin{split} c(a,z) &\sim \left(\frac{\rho - (1-\gamma)r}{\gamma} - \frac{1}{2}\frac{(\bar{R}\mu(z) - r)^2}{\bar{R}^2\sigma^2}\frac{1-\gamma}{\gamma^2}\right)a\\ s(a,z) &\sim \left(\frac{r-\rho}{\gamma} + \frac{1+\gamma}{2\gamma}\frac{(\bar{R}\mu(z) - r)^2}{\gamma\,\bar{R}^2\sigma^2}\right)a\\ k(a,z) &\sim \frac{\bar{R}\mu(z) - r}{\gamma\,q\,\bar{R}^2\sigma^{i\,2}}\,a \end{split}$$

with
$$f(x) \sim g(x)$$
 as $x \to \infty$ notation for $\lim_{x \to \infty} \frac{f(x)}{g(x)} = 1$.

Characterization of decisions k(a, z) and s(a, z) as wealth grows large : $a \to \infty$: linearity as a function of size :





Results: Stationary distribution of firms

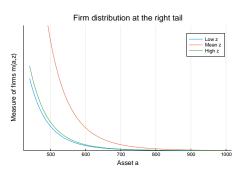
- Skewed firm distribution : mass highly concentrated on the left tail
 - Dual result of constrained capital policy k^*
- Characterization of the right tail of the firm distribution :
- ▶ Pareto distribution : $1 G(a) \sim \xi a^{-\zeta}$ with tail exponent $\zeta(z)$
 - Smoothing effect of precautionary saving. Case without jump z:

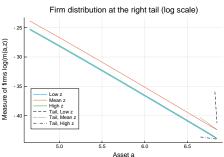
$$\zeta(z) = \gamma \left(\frac{2\bar{R}^2 \sigma^2 (\rho - r)}{(\bar{R}\mu(z) - r)^2} - 1 \right)$$

- ▶ Share of top firm $1/\zeta$ is
 - decreasing in volatility σ , risk aversion γ and time preference ρ
 - Increasing in interest rate r_t and excess return $\bar{R}\mu(z) r_t$.

Result : Stationary equilibrium

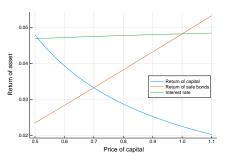
Firm distribution : Pareto : $1 - G(a) \sim \xi a^{-\zeta}$



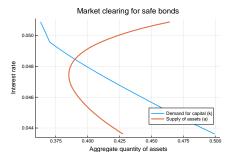


Equilibrium:

 Asset market (and market illiquidity) and bond market (and precautionary saving)



No arbitrage for asset (capital) : $\mathbb{E}[R_t]$ (blue) vs. $q_t r_t$ (red)

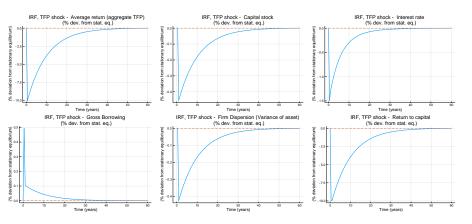


Market clearing for bonds b_t , due to interest r_t

Aggregate shocks:

- Three experiments: Unanticipated aggregate shock:
 - 1. Standard recession : \bar{R}_t drop of 10%
 - 2. Credit crunch : θ_t^c : drop of 15%
 - 3. Uncertainty shock : σ^2 : rise of 15%
- Revealed in the first year
- Generalized Ornstein Uhlenbeck process (\approx AR(1) with $\rho = 0.9$)
- ⇒ For now : abstract from market illiquidity :
 - Partial equilibrium in q_t but general equilibrium in r_t

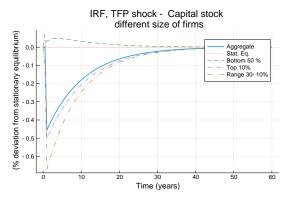
Supply side shock: "Standard" recession



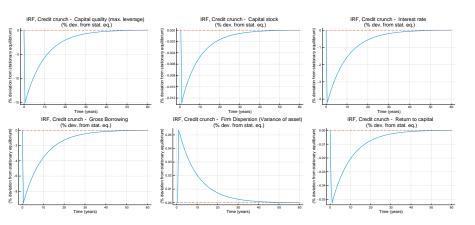
IRF – TFP shock to aggregate return \bar{R}_t

Supply side shock: "Standard" recession

- Decomposition by firms size :
 - Bottom firm behavior : constrained and smoother.
 - Top firms: change their exposure to capital risk: reduction in precautionary saving and income effect



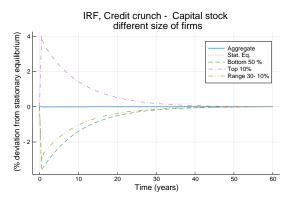
Capital quality shock: Credit crunch



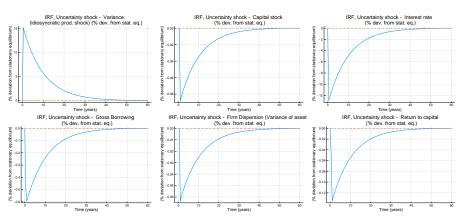
IRF – Credit crunch : shock θ_t^c

Capital quality shock: Credit crunch

- Decomposition by firms size :
 - Bottom firm behavior : deleveraging and large drop in capital
 - Top firms : benefit from the general equilibrium effect to invest



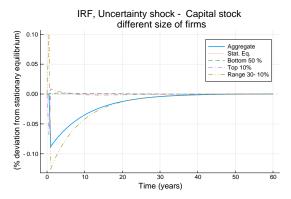
Uncertainty shock: Rise in volatility



IRF – Uncertainty shock : shock on σ_t

Uncertainty shock: Rise in volatility

- Decomposition by firms size :
 - Median/large firms reduce exposure while top firms benefit from equilibrium effect (interest): opposite behaviors on impact!
 - Bottom firm behavior : constrained and barely affected.



Conclusion and Future research

- Question :
 - What is the influence of firms heterogeneity on credit cycles?
 - \rightarrow Standard recessions driven by the median and larger firms
 - → Effects of credit crunch partly offset/smoothed by the hedging of largest firms (top 10%)
 - → Uncertainty shocks or TFP shock do not impact the bottom constrained firm
- Bridging macro-finance and heterogeneous agents literatures
 - Propose a theoretical framework with heterogeneous firms, collateral constraints, and market illiquidity, in presence of aggregate risk
 - Non-linearity of shocks + failure of approximate aggregation
 - Does market (il)liquidity matters? (next step on this project!)

Future research – Empirical evidence

- Empirical relevance (next step on this project!)
 - Micro-level Firms heterogeneity
 - Cross-section data from Compustat
 - Largest firms : reacting to aggregate shocks + power law $\zeta \approx 1.0$
 - ⇒ Structural estimation (SMM)
 - What determine the firms distribution?
 - Preference? or structure/variance of risks?
- Application to Boom and Bust dynamics
 - Great Recession: From credit crunch to economic downturn
- Aggregate risk : more elaborate treatment :
 - Exploiting MIT shocks (BKM, Auclert et al.),
 Tree structure of aggreg. shocks (Bourany-Achdou, cf. appendix)

Conclusion

- ► This paper examine the influence of firm heterogeneity in presence of market illiquidity and financial frictions
- It revisits the transmission and amplification mechanisms of different aggregate shocks that could explain deep recessions

Conclusion

- ► This paper examine the influence of firm heterogeneity in presence of market illiquidity and financial frictions
- ► It revisits the transmission and amplification mechanisms of different aggregate shocks that could explain deep recessions
- ► THANK YOU FOR YOUR FEEDBACKS!

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- Credit Cycles, Market Liquidity & Heterogeneous firms
- Appendix : Algorithm and mathematical problems
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Appendix: Mathematical problems and Algorithms

- ► Heterogeneous agents (HA) :
 - \rightarrow Usually no analytical solutions :
 - → Numerics : Value Fct. Iteration / finite difference for PDEs
 - For stationary equilibrium / MFG (mean field games)

Appendix : Algorithm and mathematical problems

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- HA models with aggregate shocks
 - → Infinite dimensional problem (Master equation) or Stochastic system of PDEs (SPDE)
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- HA models with aggregate shocks, and non-trivial distribution and non-linear dynamics
 - \rightarrow ...?
 - Achdou-Bourany (2018)
 - Ongoing work by other researchers

Appendix : Algorithm and mathematical problems

Appendix: Tree structure for aggregate shocks, Achdou-Bourany

- ► Achdou-Bourany (2018)
 - Master thesis under supervision of Y. Achdou

Appendix: Tree structure for aggregate shocks, Achdou-Bourany

- Achdou-Bourany (2018)
 - Master thesis under supervision of Y. Achdou
- Main idea: approximate the process for the agg. shock Z_t by a finite number of "simple" shocks:
 - Every ΔT (deterministic times), Z_t jumps stochastically to one of the K outcomes
 - Repeat this: a finite M number of "waves" of uncertainty
 - This way, you can build a tree of K^M paths of Z_t with deterministic branches separated by stochastic shocks
 - Taking $\Delta T \rightarrow 0$, you can approximate any process (e.g. Donsker's theorem for Brownian motion)
 - Need to link the branches together in an appropriate way

Appendix : Algorithm and mathematical problems

Appendix: Tree structure for aggregate shocks:

Achdou-Bourany

- Grafting branches :
 - On each branch (between each shock), compute the evolution of the system: HJB and KF: v(a,z_i,ž) and g(a,z_i,ž)
- ► To account for future and past shocks?
 - ⇒ use boundary conditions of the PDEs!

Appendix: Tree structure for aggregate shocks:

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 - On each branch (between each shock), compute the evolution of the system: HJB and KF: v(a,z_i,\(\tilde{z}\)) and g(a,z_i,\(\tilde{z}\))
- ► To account for future and past shocks?
 - ⇒ use boundary conditions of the PDEs!
 - t_m^- time before revelation of the shock $(Z_{t_m^-} = Z_m)$
 - t_m^+ : time when shocks hits ($Z_{t_m^+} = Z_{m+1}$ take K values)

$$v(a,z_j,Z_m) = \sum_{k|Z_{m+1}=Z_k}^{m} \mathbb{P}(Z_{m+1}|Z_m) v(a,z_j,Z_{m+1})$$

$$g(a,z_j,Z_m)=g(a,z_j,Z_{m+1})$$

- Agents are forward looking, form expectations over the different future branches (paths of Z_t)
- Continuity of $g(\cdot)$ in time t

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$$v(a,z_{j},Z_{m}) = \sum_{k|Z_{m+1}=Z_{k}}^{\infty} \mathbb{P}(Z_{m+1}|Z_{m}) v(a,z_{j},Z_{m+1})$$

$$g(a,z_j,Z_m)=g(a,z_j,Z_{m+1})$$

- Agents are forward looking, form expectations over the different future branches (paths of Z_t)
- Continuity of $g(\cdot)$ in time t
- ► Loop to find eq. fixed point on the entire tree (all branches!)

Appendix : Algorithm and mathematical problems

Appendix: Tree structure for aggregate shocks: Achdou-Bourany

- This type of algorithm :
 - Global: no linearization, simulate the entire set of shocks histories
 - Non-linear : shock of different variance/sign with different effects : break the certainty equivalence!
 - Keep track of the movement of the entire distribution
 - > ... But stark assumption of the structure of uncertainty (finite sequence of deterministic time of revelation of shocks)
 - > ... and slow (exponentially growing complexity)
- Simulation of the model above. Need to compare with :
 - Krusell Smith: distribution of agents matters beyond first moment
 - Reiter/Perturbation methods: credit crunch and market illiquidity feedback loops create non-linear dynamics
 - BKM and Auclert et al : anticipation/hedging/pricing of aggregate risk changes transmission of shocks!
- ► This comparison project will be implemented very soon...