

## Part II: Boltzmann mean field games

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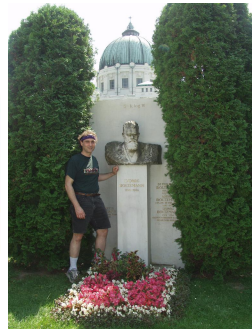
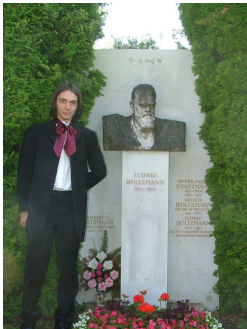
Existence of BGP solutions

## Kinetic theory

- *Kinetic theory*: was originally developed to describe the statistical evolution of a non-equilibrium many-particle system in phase space.
- Ludwig Boltzmann made significant contributions in kinetic theory by investigating the properties of dilute gases.

# Kinetic theory

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## The Boltzmann equation

The *classic Boltzmann equation* describes the evolution of the one-particle distribution function of a rarefied monatomic gas.

Let  $f = f(x, v, t)$  denote the probability to find a particle at position  $x \in \mathbb{R}^3$  with velocity  $v \in \mathbb{R}^3$  at time  $t > 0$ . Then the equation reads as:

$$\frac{\partial f}{\partial t}(x, v, t) = - \underbrace{v \cdot \nabla_x f(x, v, t)}_{\text{free particle transport}} + \underbrace{Q(f, f)(x, v, t)}_{\text{effects of binary collisions}}$$

The original Boltzmann equation was derived under the following assumptions:

- *Binary interactions*: such as in dilute gases, where interactions of more than two particles can be neglected.
- *Elastic collisions*  $\Rightarrow$  conservation of mass and momentum.
- Collisions involve only *uncorrelated particles*.

## Collisions

*Elastic binary collision:* Given two particles with velocity  $v$  and  $w$  the post-collisional velocities  $v^*$  and  $w^*$  we have

$$v^* = \frac{1}{2}(v + w + |v - w|n)$$
$$w^* = \frac{1}{2}(v + w - |v - w|n),$$

where  $n$  is the unit normal vector.

*Conservation of momentum and kinetic energy:*

$$v + w = v^* + w^*$$
$$|v|^2 + |w|^2 = |v^*|^2 + |w^*|^2$$

Collision operator in the case of hard spheres:

$$Q(f, g)(v) = \int_{\mathbb{R}^3 \times S^2} B((v - w) \cdot n)(f(v^*)g(w^*) - f(v)g(w))dwdn.$$

where  $B$  is the collision kernel.

## Fundamental properties of the collision operator

- Conservation of mass, momentum and energy

$$\int_{\mathbb{R}^3} \mathcal{Q}(f, f) \psi(v) dv = 0 \text{ for } \psi = 1, v, |v|^2.$$

- **H-Theorem:** The entropy  $-\int_{\mathbb{R}^3} f \log f dv$  is non-decreasing in time. That is

$$-\frac{d}{dt} \int_{\mathbb{R}^3} f \log f dv = - \int_{\mathbb{R}^3} \mathcal{Q}(f, f) \log(f) dv \geq 0.$$

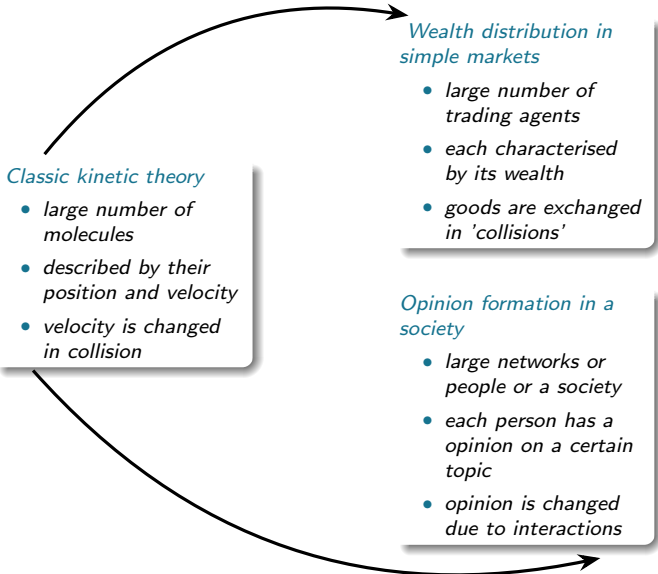
Any equilibrium distribution, which is a maximum of the entropy, has to be of Maxwellian form

$$M(\rho, u, T)(v) = \frac{\rho}{(2\pi T)^{\frac{d}{2}}} \exp\left(-\frac{|u - v|^2}{2T}\right),$$

where  $\rho$ ,  $u$  and  $T$  are the density, mean velocity and temperature of the gas

$$\rho = \int_{\mathbb{R}^3} f(v) dv, \quad u = \frac{1}{\rho} \int_{\mathbb{R}^3} v f(v) dv, \quad T = \frac{1}{3\rho} \int_{\mathbb{R}^3} |u - v|^2 f(v) dv.$$

## From molecules to agents





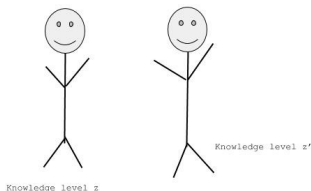
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# Knowledge diffusion and growth<sup>1</sup>

Lucas and Moll's model setup:

- Consider a continuum of individuals, which are characterised by their knowledge level  $z \in \mathbb{R}^+$ .
- Let  $s = s(z, t)$  denote the time that an individual with knowledge level  $z$  spends on learning.
- Each individual has one unit of time, which he/she can split between *producing goods* with the knowledge already obtained or *meeting others to enhance their knowledge level*.



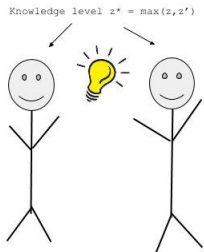
If two individuals with knowledge level  $z$  and  $z'$  meet, they exchange ideas.

<sup>1</sup>R. E. Lucas Jr and B. Moll. Knowledge growth and the allocation of time. *Journal of Political Economics*, 2014

# Knowledge diffusion and growth<sup>1</sup>

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⇒ their post-collision knowledge corresponds to

$$z^* = \max(z, z').$$

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## Knowledge diffusion and growth

Evolution of the distribution of agents  $f = f(z, t)$  with respect to their knowledge level  $z$ :

$$\partial_t f(z, t) = -\alpha(s(z, t))f(z, t) \int_z^\infty f(y, t)dy + f(z, t) \int_0^z \alpha(s(y, t))f(y, t)dy.$$

- The function  $\alpha = \alpha(s)$  is the interaction probability of an individual, which spends an  $s$ -th fraction of its time on learning (also called the learning function).  
*Possible choices:*

$$\alpha(s) = \alpha_0 s^n, \quad n \in (0, 1).$$

- Individual productivity:

$$y(t) = (1 - s(z, t))z$$

- Total earnings in an economy:

$$Y(t) = \int_0^\infty (1 - s(z, t))zf(z, t) dz.$$

## How much time should one spend on learning ?

Each individual wants to maximise its earnings by choosing the optimal fraction of learning time  $s = s(z, t)$ :

$$V(x, t') = \max_{s \in \mathcal{S}} \left[ \int_{t'}^T \int_0^\infty e^{-r(t-t')} (1 - s(z, t)) z \rho_x(z, t) dz dt \right],$$

subject to

$$\partial_t \rho_x(z, t) = -\alpha(s) \rho_x(z, t) \int_z^\infty f(y, t) dy + f(z, t) \int_0^z \alpha(s) \rho_x(y, t) dy$$

with  $\rho_x(z, t') = \delta_x$ .

Hamilton-Jacobi Bellman (HJB) equation for the value function  $V = V(z, t)$ :

$$\begin{aligned} & \partial_t V(z, t) - rV(z, t) \\ & + \max_{s \in \mathcal{S}} \left( (1 - s(z, t))z + \alpha(s) \int_z^\infty [V(y, t) - V(z, t)] f(y, t) dy \right) = 0, \end{aligned}$$

where  $\mathcal{S}$  denotes the set of admissible controls  $\mathcal{S} = \{s : \mathcal{I} \times [0, T] \rightarrow [0, 1]\}$  and  $\mathcal{I} = \mathbb{R}^+$  or  $\mathcal{I} = [0, \bar{z}]$ .

## The BMFG system

$$\partial_t f(z, t) = -\alpha(S(z, t))f(z, t) \int_z^\infty f(y, t)dy + f(z, t) \int_0^z \alpha(S(y, t))f(y, t)dy.$$

$$\partial_t V(z, t) - rV(z, t) =$$

$$- \max_{s \in \mathcal{S}} \left[ (1 - s(z, t))z - \alpha(s(z, t)) \int_z^\infty [V(y, t) - V(z, t)]f(y, t)dy \right]$$

$$S(z, t) = \arg \max_{s \in \mathcal{S}} \left[ (1 - s(z, t))z + \alpha(s(z, t)) \int_z^\infty [V(y, t) - V(z, t)]f(y, t)dy \right],$$

$$f(z, 0) = f_0(z),$$

$$V(z, T) = 0.$$

Highly nonlinear problem: Boltzmann type equation describing the evolution of individuals forward in time and a HJB equation for their optimal strategy backward in time.

## Special case $\alpha = \alpha_0$

*In this case the equations decouple and the maximum of*

$$(1 - s(z, t))z + \alpha(s) \int_z^\infty [V(y, t) - V(z, t)]f(y, t)dy$$

*is  $S(z, t) = 0$ .*

*The Boltzmann equation can be written in terms of the cdf  $F(z, t) = \int_0^z f(y, t)dy$ :*

$$\partial_t F(z, t) = -\alpha_0(1 - F(z, t))F(z, t).$$

*Then the function  $G(z, t) = 1 - F(z, t)$  satisfies the Fisher KPP equation.*

## Analysis of the Boltzmann equation <sup>2</sup>

First we consider the Boltzmann type equation for a given learning function  $\alpha = \alpha(z, t)$ :

$$\begin{aligned}\partial_t f(z, t) &= -\alpha(z, t)f(z, t) \int_z^{\bar{z}} f(y, t) dy + f(z, t) \int_0^z \alpha(y, t)f(y, t)dy, \\ f(z, 0) &= f_0(z),\end{aligned}$$

on the interval  $\mathcal{I} = [0, \bar{z}]$ , where  $f_0 \in L^\infty(\mathcal{I})$  is a given probability density.

### Theorem

Let  $\alpha = \alpha(z, t) \in L^1(\mathcal{I}) \times L^\infty([0, T])$ . Then the Boltzmann equation has a global in time solution  $f = f(z, t) \in L^1(\mathcal{I}) \times L^\infty([0, T])$ .

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<sup>2</sup>M. Burger, A. Lorz and MTW, On a Boltzmann mean-field model for knowledge growth, SIAM Appl Math 76(5), 2016

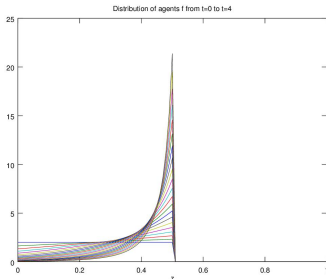


But if  $f_0$  has compact support....

### Theorem

Let  $\alpha(z, t) \geq \underline{\alpha} > 0$  and  $\bar{z} \in \text{supp}(f)$ , then

$$f(\cdot, t) \rightharpoonup^* \delta_{\bar{z}}.$$



## The Hamilton-Jacobi-Bellman equation

Consider the HJB equation for a given  $f \in C(0, T, L^1)$  on  $\mathcal{I} = \mathbb{R}^+$ :

$$\partial_t V(z, t) - rV(z, t) = - \max_{s \in \mathcal{S}} [(1 - s(z, t))z - \alpha(s(z, t))V(z, t)((1 - H) * f) \\ + \alpha(s(z, t))((1 - H) * (Vf))]$$

$$V(z, T) = 0.$$

**Assumptions:**

(A1) Let the final data  $V(\cdot, T)$  be non-negative and non-decreasing.

(A2) Let the interaction function satisfy:

$$\alpha : [0, 1] \rightarrow \mathbb{R}^+, \alpha \in C^\infty([0, 1]), \alpha(0) = 0, \alpha'(0) = \infty, \alpha'' < 0 \text{ and } \alpha \text{ monotone.}$$

## The full BMFG system

### Theorem

Let  $f \in C(0, T, L^1)$  be given and  $\alpha$  satisfies assumption (A2). Then there exists a unique solution  $V \in C(0, T, L^\infty)$  of the HJB equation with  $V(z, T) = 0$ . Moreover, let  $\tilde{V}$  be a solution of the HJB equation with  $\tilde{f}$ . Then there exist constants  $m$  and  $D$  (independent of  $\tilde{V}$  and  $\tilde{f}$ ) such that

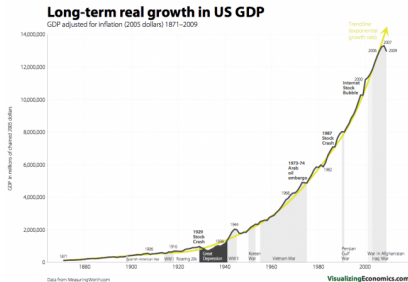
$$\|V - \tilde{V}\|_\infty \leq De^{mt} \|f - \tilde{f}\|_{C(0, T, L^1)} \|\tilde{V}\|_\infty.$$

### Theorem

Let  $f_0(z) \in L^\infty(\mathcal{I})$  be a probability density and (A1) and (A2) be satisfied. If  $\lim_{s \rightarrow 0} \frac{(\alpha')^3}{\alpha''} < \infty$ , then the fully coupled Boltzmann mean field game system on  $\mathcal{I} = \mathbb{R}^+$  has a unique local in time solution.

## Endogenous growth theory

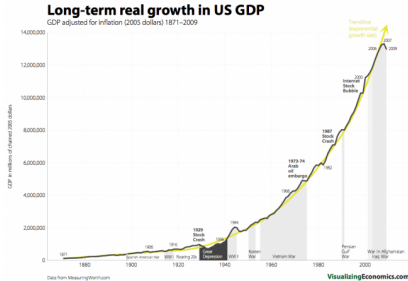
- *Economic growth describes the increase of the inflation-adjusted market value of the goods and services produced in an economy over time - commonly measured in the gross domestic product (GDP).*
- *The GDP of most developed countries has grown about two percent since World War II.*



- *Economists are interested in solutions which correspond to sustained growth - so called balanced growth path (BGP) solutions.*

# Endogenous growth theory

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Can we find BGP solutions for the BMFG system ?

## Balanced growth path solutions

Let us assume there exists a growth parameter  $\gamma \in \mathbb{R}^+$  and consider the re-scaling:

$$f(z, t) = e^{-\gamma t} \phi(ze^{-\gamma t}), \quad V(z, t) = e^{\gamma t} v(ze^{-\gamma t}) \quad \text{and} \quad s(z, t) = \sigma(ze^{-\gamma t})$$

Rescaled BMFG system in  $(v, \phi, \sigma) = (v(x), \phi(x), \sigma(x))$  with  $x = e^{-\gamma t} z$  reads as:

$$\begin{aligned} -\gamma\phi(x) - \gamma\phi'(x)x &= \phi(x) \int_0^x \alpha(\sigma(y))\phi(y) dy - \alpha(\sigma(x))\phi(x) \int_x^\infty \phi(y) dy \\ (r - \gamma)v(x) + \gamma v'(x)x &= \max_{\sigma \in \Xi} \left\{ (1 - \sigma)x + \alpha(\sigma) \int_x^\infty [v(y) - v(x)]\phi(y) dy \right\} \end{aligned}$$

where  $\Xi = \{\sigma : \mathbb{R}^+ \rightarrow [0, 1]\}$  denotes the set of admissible controls.

Re-scaling results in *exponential growth* of the overall production:

$$Y(t) = e^{\gamma t} \int_0^\infty [1 - \sigma(x)]x\phi(x)dx.$$

## Existence of BGP solutions

Does such a growth parameter  $\gamma$  exist ?

## Existence of BGP solutions

The initial commutative distribution function  $F(z, 0) = \int_0^z f_0(z) dz$  has a *Pareto tail*, if there exist constants  $k, \theta \in \mathbb{R}^+$  such that

$$\lim_{z \rightarrow \infty} \frac{1 - F(z, 0)}{z^{-1/\theta}} = k. \quad (P)$$

### Lemma

Let (P) be satisfied. Then  $F = F(z, t)$  has a Pareto tail with the same decay rate  $\theta$  for all times  $t \in [0, T]$ .

### Theorem

Let (P) be satisfied and  $\alpha = \alpha_0$ . then there exists a unique BGP solution  $(\Phi, v, 0)$  and a scaling constant  $\gamma$  given by

$$\gamma = \alpha_0 \theta \int_{\mathcal{I}} f_0(z) dz, \quad \Phi(x) = \frac{1}{1 + kx^{-1/\theta}} \text{ with } \Phi(x) = \int_0^x \phi(y) dy.$$



## Existence of BGP solutions

*Degenerate solution:*

$$\begin{aligned}\gamma = 0, \quad v = \frac{x}{r} \text{ and } S \equiv 0 &\Rightarrow \Phi(x) = 1 \text{ for } x > 0 \\ &\Rightarrow \phi(x) = \delta_0\end{aligned}$$

*Challenge for the analysis and numerics: construct a solution  $\Phi$  with a strictly positive Pareto tail  $k > 0$ .*

Variable transformation:

$$\zeta := x^{-1/\theta} \text{ and } K(\zeta) := \frac{1 - \Phi(x)}{\gamma\zeta},$$

where  $\theta$  and  $k$  denote the Pareto indices. We solve the correspondingly transformed equation with an initial condition at  $\zeta = 0$  (determined by the Pareto tail condition).

## Existence of BGP solutions

### Theorem

*Let  $r > \theta\alpha(1)$  and  $\tilde{k} > 0$ , then the BGP system has a non-trivial solution satisfying the Pareto-tail condition with  $k = \frac{\gamma}{\theta}\tilde{k}$ .*

**Idea of proof:** Fixed point argument

- Solve equations for  $(\Phi, \gamma)$  given  $(v, S)$ .
- Solve equations for  $(v, S)$  and given  $(\Phi, \gamma)$ .

## Diffusion and knowledge growth<sup>3</sup>

*Achdou et al. postulate that diffusion*

- *enhances growth in the case of a Pareto tail*
- *and leads to exponential growth also for compactly supported initial values.*

*Special case  $\alpha = \alpha_0$ :*

- *The Fisher KPP equation (with diffusion) admits travelling wave solutions*

$$G(z, t) = \Phi(z - \gamma t)$$

*with a minimal wave speed  $\gamma = 2\sqrt{\nu\alpha_0}$ .*

- *Travelling waves correspond to BGP solutions (in logarithmic variables).*

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<sup>3</sup>Y. Achdou, F.J. Buera, J.-M. Lasry, P.-L. Lions and B. Moll, PDE models in macroeconomics, *Phil. Trans. Roy. Soc. A*, 372, 2014.

## Diffusion and knowledge growth

Let the knowledge of each agent evolve by a **geometric Brownian motion** (independent of the time spent on learning), that is

$$Z_t = \exp(\sqrt{2\nu}W_t)$$

where  $W_t$  is a Wiener process, independently for each agent.

*Then the corresponding Boltzmann mean field game system with diffusion reads as:*

$$\begin{aligned} \partial_t f(z, t) - \nu \partial_{zz}(z^2 f(z, t)) + \nu \partial_z(zf(z, t)) &= f(z, t) \int_0^z \alpha(S(y, t)) f(y, t) dy \\ &\quad - \alpha(S(z, t)) f(z, t) \int_z^\infty f(y, t) dy, \\ \partial_t V(z, t) + \nu z^2 \partial_{zz} V(z, t) + \nu z \partial_z V(z, t) - rV(z, t) &= \\ &\quad - \max_{s \in \mathcal{S}} \left[ (1-s)z + \alpha(s) \int_z^\infty [V(y, t) - V(z, t)] f(y, t) dy \right]. \end{aligned}$$

## Knowledge diffusion

Assuming the existence of the scaling parameter  $\gamma$  for a balanced growth path we rewrite the system in the known BGP variables  $(\phi, \sigma, v)$

$$\begin{aligned} -\gamma\phi(x) - \gamma x\phi'(x) - \nu(x^2\phi(x))'' + \nu(x\phi(x))' = \\ \phi(x) \int_0^x \alpha(\sigma(y))\phi(y)dy - \alpha(\sigma(x))\phi(x) \int_x^\infty \phi(y)dy \\ (r - \gamma)v(x) + \gamma xv'(x) - \nu x^2 v''(x) - \nu xv'(x) = \\ - \max_{\sigma \in \Sigma} \left[ (1 - \sigma)x + \alpha(\sigma) \int_x^\infty [v(y) - v(x)]\phi(y)dy \right]. \end{aligned}$$

Achdou et al. <sup>4</sup> postulated the existence of BGP solutions to this system with a rescaling parameter  $\gamma$  given by

$$\gamma = 2\sqrt{\nu \int_0^\infty \alpha(\sigma(y))\phi(y)dy}.$$

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<sup>4</sup>Y. Achdou, F.J. Buera, J.-M. Lasry, P.-L. Lions and B. Moll, PDE models in macroeconomics, *Phil. Trans. Roy. Soc. A*, 372, 2014.

## This model is quite simplistic....

... since meetings between individuals are completely **asymmetric**. Individuals can only increase their knowledge through active search, the 'smarter' individual gains nothing in the meeting.

*Symmetric meetings: if an individual with knowledge level  $y$  initiated the meeting, the one with the higher knowledge level  $z$  may learn with a probability  $\beta$ . This gives:*

$$\begin{aligned} \frac{\partial f}{\partial t} = & -f(z, t) \int_z^\infty [\alpha(s(z, t)) + \beta\alpha(s(y, t))]f(y, t)dy \\ & + f(z, t) \int_0^z [\alpha(s(y, t)) + \beta\alpha(s(z, t))]f(y, t)dy. \end{aligned}$$

## Limits to learning

If two individuals meet, the one with the lower knowledge level  $z$  adopts the higher knowledge level  $y$  with a certain probability  $k(\frac{y}{z})$ . Then

$$\begin{aligned}\partial_t f(z, t) = & f(z, t) \int_0^z \alpha(s(y, t)) f(y, t) k\left(\frac{z}{y}\right) dy \\ & - \alpha(s(z, t)) f(z, t) \int_z^\infty f(y, t) k\left(\frac{y}{z}\right) dy.\end{aligned}$$

Possible choice for  $k$ :

$$k(x) = \delta + (1 - \delta)x^{-\kappa} \quad \text{where } \kappa > 0.$$

*Alternative interpretation of  $k$ : interaction probability depends on the distance between knowledge levels.*

## Exogenous knowledge shocks

In the case of a constant interaction rate  $\alpha = \alpha_0$  the CDF  $F = F(z, t)$  evolves according to

$$\partial_t F(z, t) = -\alpha(1 - F(z, t))F(z, t).$$

Then

$$\lim_{t \rightarrow \infty} F(z, t) = \frac{1}{1 + kx^{-\frac{1}{\theta}}}$$

*Exogenous knowledge shock:* undiscovered ideas modelled by a CDF  $G = G(z)$

$$\partial_t F(z, t) = -\alpha(1 - F(z, t))F(z, t) - \beta(1 - G(z))F(z, t)$$



## Asymptotic behaviour

Depends on the 'tails' of  $F$  and  $G$ :

- If neither  $F(z, 0)$  nor  $G(z)$  has a Pareto tail there will be no growth in the long run.
- If  $F(z, 0)$  has a fatter tail than  $G(z)$  then the BGP has a growth rate of  $\gamma = \alpha\theta$  (external ideas do not influence the asymptotic behaviour).
- If  $G(z)$  has a fatter tail (denoted by  $\frac{1}{\xi}$ ) the BGP path grows at a rate  $\gamma = \alpha\xi$  and the asymptotic distribution satisfies

$$\lim_{t \rightarrow \infty} F(z, t) = \frac{1}{1 + \frac{\beta}{\alpha} m x^{-\frac{1}{\xi}}}$$

where  $m > 0$ .

- If they have the same Pareto tail then the asymptotic distribution satisfies

$$\lim_{t \rightarrow \infty} F(z, t) = \frac{1}{1 + [k + \frac{\beta}{\alpha} m] x^{-\frac{1}{\theta}}}$$

with  $m > 0$ .

## The time-dependent solver

The solver is based on a fixed point scheme:

- 1 Given  $f_0$  and  $S^k$  solve

$$\frac{1}{\tau}(f_i^{k+1} - f_i^k) - \frac{\nu}{h^2}(z_{i+\frac{1}{2}}^2 f_{i+1}^{k+1} - (z_{i+\frac{1}{2}}^2 + z_{i-\frac{1}{2}}^2) f_i^{k+1} + z_{i-\frac{1}{2}}^2 f_{i-1}^{k+1}) + \frac{\nu}{h}(z_{i+\frac{1}{2}} f_i^{k+1} - z_{i-\frac{1}{2}} f_{i-1}^{k+1}) = g_1(f^k, S^k),$$

for every time  $t^k = k\tau$ ,  $k > 1$ , using a trapezoidal rule to approximate the integrals in  $g_1$ .

- 2 Update the maximizer  $S^k$ .
- 3 Given the evolution of the density  $f^k$  and the maximizer  $S^k$  solve the HJB equation

$$\frac{1}{\tau}(V_i^{k+1} - V^k) + \frac{\nu}{h^2} z_i^2 (V_{i+1}^k - 2V_i^k + V_{i-1}^k) + \frac{\nu}{h} z_i (V_i^k - V_{i-1}^k) - rV_i^k = g_2(S^{k+1}, f^{k+1}, V^{k+1}),$$

backward in time using a trapezoidal rule to approximate  $g_2$ .

- 4 Go to step (1) until convergence.

## The BGP solver

The BGP solver is also based on a fixed point scheme:

- 1 Given  $\phi^{n+1}$ ,  $\gamma^n$  and  $\sigma^n$  solve

$$(r - \gamma^n)v_i^{n+1} + \frac{(\gamma^n - \nu)}{h}x_i(v_i^{n+1} - v_{i-1}^{n+1}) - \frac{\nu x_i^2}{h^2}(v_{i+1}^{n+1} - 2v_i^{n+1} + v_{i-1}^{n+1}) \\ = -q_2(\phi_{n+1}, v^n, \sigma^n)$$

using the trapezoidal rule to approximate the right hand side  $q_2$ .

- 2 Compute the maximum  $\sigma^{n+1}$  and update the growth parameter  $\gamma^{n+1}$  via

$$\gamma^{n+1} = 2(\nu \int_{\mathcal{I}} \alpha(\sigma^{n+1}(y))\phi^{n+1}(y)dy)^{\frac{1}{2}}.$$

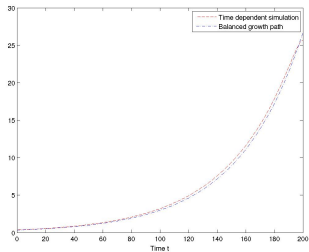
- 3 Given  $v^n$ ,  $\sigma^n$  and  $\gamma^n$  solve

$$-(\gamma^n - \nu)\phi_i^{n+1} - \frac{(\gamma^n - \nu)}{h}x_i(\phi_{i+1}^{n+1} - \phi_i^{n+1}) - (\Xi_i - \alpha(\sigma_i^n)(1 - \Phi_i))\phi_i^{n+1} \\ - \frac{\nu}{h^2}(x_{i+\frac{1}{2}}^2\phi_{i+1}^{n+1} - (x_{i+\frac{1}{2}}^2 + x_{i-\frac{1}{2}}^2)\phi_i^{n+1} + x_{i-\frac{1}{2}}^2\phi_{i-1}^{n+1}) = 0,$$

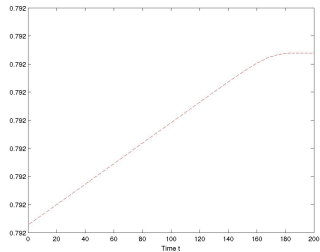
subject to the constraint  $(\phi_1^{n+1} + \phi_2^{n+1} + \dots + \frac{1}{2}\phi_N^{n+1})h = 1$  (normalisation  $\int_0^{\bar{z}} \phi(y)dy = 1$  plus  $\phi_0^{n+1} = 0$ ).

- 4 Go to (1) until convergence.

# Simulations



(a) Transient vs. BGP



(b) Linear growth

**Figure:** Evolution of the production function  $Y = Y(t)$  in time for different choices of  $n$  and  $\theta$

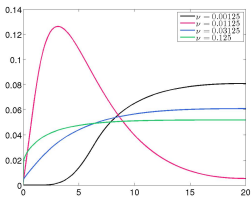
## Numerical simulations

- *Simulations of the time-dependent problem as well as the BGP system are performed iteratively.*
- *We solve the systems on a bounded domain with no-flux boundary conditions.*
- *To exclude degenerate BGP solutions we set*

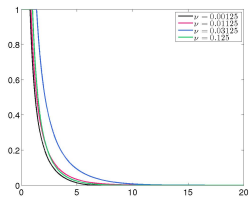
$$\phi_0 = 0.$$

- *We use a finite difference discretization in space and approximate the integrals using the trapezoidal rule.*

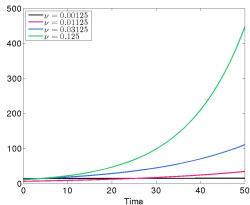
## BGP simulations with knowledge diffusion



(a) Agent distribution for different values of  $\nu$ .



(b) Fraction of time  $\sigma$  devoted to learning for different values of  $\nu$ .



(c) Production function  $Y$ .

## This model is quite simplistic....

... since meetings between individuals are completely *asymmetric*. Individuals can only increase their knowledge through active search, the 'smarter' individual gains nothing in the meeting.

*Symmetric meetings:* if an individual with knowledge level  $y$  initiated the meeting, the one with the higher knowledge level  $z$  may learn with a probability  $\beta$ . This gives:

$$\begin{aligned}\frac{\partial f}{\partial t} = & -f(z, t) \int_z^\infty [\alpha(s(z, t)) + \beta\alpha(s(y, t))]f(y, t)dy \\ & + f(z, t) \int_0^z [\alpha(s(y, t)) + \beta\alpha(s(z, t))]f(y, t)dy.\end{aligned}$$

## Limits to learning

If two individuals meet, the one with the lower knowledge level  $z$  adopts the higher knowledge level  $y$  with a certain probability  $k(\frac{y}{z})$ . Then

$$\begin{aligned}\partial_t f(z, t) = & f(z, t) \int_0^z \alpha(s(y, t)) f(y, t) k\left(\frac{z}{y}\right) dy \\ & - \alpha(s(z, t)) f(z, t) \int_z^\infty f(y, t) k\left(\frac{y}{z}\right) dy.\end{aligned}$$

Possible choice for  $k$ :

$$k(x) = \delta + (1 - \delta)x^{-\kappa} \quad \text{where } \kappa > 0.$$

*Alternative interpretation of  $k$ : interaction probability depends on the distance between knowledge levels.*



To finish....

References:

- Y. Achdou, F.J. Buera, J.-M. Lasry, P.-L. Lions and B. Moll, PDE models in macroeconomics, *Phil. Trans. Roy. Soc. A*, 372, 2014.
- R. E. Lucas Jr and B. Moll. Knowledge growth and the allocation of time. *Journal of Political Economics*, 2014.
- M. Burger, A. Lorz and M.T. Wolfram, On a Boltzmann model for knowledge growth. *SIAM Applied Math* 76(5), 2016.
- M. Burger, A. Lorz and M.T. Wolfram, Balanced growth path solutions of a Boltzmann mean field game model for knowledge growth, *KRM* 10(1), 2017.

To finish....

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*Thanks for the attention and have a great time at rest of the school !*