

# Numerical approaches for MFG

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# Outline

- 1 Mathematical framework
  - Mean field games setting
  - Available literature
- 2 General monotonic algorithms (J. Salomon, G.T.)
  - Related applications: bi-linear problems
  - Framework
  - Construction of monotonic algorithms
- 3 Technology choice modelling (A. Lachapelle, J. Salomon, G.T.)
  - The model
  - Numerical simulations
- 4 Liquidity source: heterogenous beliefs and analysis costs

# Mean field games

- Mean field games: limits of Nash equilibriums for infinite number of players (P.L.Lions & J.M.Lasry)
- equation for each player  $dX_t = \alpha dt + \sigma dW_t$ ,  $\alpha(t, x) = \text{control}$
- $m(t, x) =$  the density of players at time  $t$  and position  $x \in Q$
- evolution equation

$$\frac{\partial}{\partial t} m(t, x) - \nu \Delta m(t, x) + \text{div}(\alpha(t, x) m(t, x)) = 0,$$

$$m(0, x) = m_0(x).$$

- We consider the **optimisation setting**:  $\min_{\alpha} J(\alpha)$
- $J(\alpha) := \Psi(m(\cdot, T)) + \int_0^T \left\{ \Phi(m(t, \cdot)) + \int_Q L(x, \alpha) m(t, x) dx \right\} dt$
- $\Phi, \Psi$  can be linear, concave, ... Typical  $L : L(x, \alpha) = \frac{\alpha^2}{2}$ .

## Numerics of MFG : literature overview

- (in)finite horizon: finite-difference discretization: approximation properties, existence and uniqueness, bounds on the solutions.  
"Mean Field Games: Numerical Methods" Y. Achdou & I. Capuzzo-Dolcetta
- Y. Achdou & I. Capuzzo-Dolcetta: Newton method for the coupled direct-adjoint critical point equations (finite horizon)
- O. Gueant: study of a prototypical case: solution, stability (09), quadratic Hamiltonian (11)
- solution of the MFG equations from an optimization point of view (A. Lachapelle, J. Salomon, G. Turinici, M3AS 2010)
- Lachapelle & Wolfram (2011) (congestion modelling)

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# Optimal control of a Fokker-Plank equation (G. Carlier & J. Salomon)

Evolution equation :

$$\partial_t \rho - \epsilon^2 \Delta \rho + \operatorname{div}(v \rho) = 0 \quad (1)$$

$$\rho(x, t = 0) = \rho_0(x) \quad (2)$$

- goal: minimize w.r. to  $v$  the functional (for some given  $V(\cdot)$ ) :

$$E(v) = \int \int \rho v^2 dx dt + \int \rho(x, 1) V(x)$$

# Time dependent Schrödinger equation w. **BILINEAR** interaction (e.g. laser)

$$\begin{cases} i \frac{\partial}{\partial t} \Psi(x, t) = (H_0 - \epsilon(t)^k \mu(x)) \Psi(x, t) \\ \Psi(x, t=0) = \Psi_0(x) \end{cases} \quad (3)$$

- vectorial case (rotation control, NMR):

$$i \frac{\partial}{\partial t} \Psi(x, t) = [H_0 + (E_1(t)^2 + E_2(t)^2) \mu_1 + E_1(t)^2 \cdot E_2(t) \mu_2] \Psi(x, t).$$

$H_0 = -\Delta + V(x)$ , unbounded domain

Evolution on the unit sphere:  $\|\Psi(t)\|_{L^2} = 1, \forall t \geq 0$ .

- evaluation of the quality of a control through a objective functional to minimize

$$J(\epsilon) = -2\Re \langle \psi_{target} | \psi(\cdot, T) \rangle + \int_0^T \alpha(t) \epsilon^2(t) dt$$

$$J(\epsilon) = \|\psi_{target} - \psi(\cdot, T)\|_{L^2}^2 - 2 + \int_0^T \alpha(t) \epsilon^2(t) dt$$

$$J(\epsilon) = -\langle \Psi(T), O\Psi(T) \rangle + \int_0^T \alpha(t) \epsilon^2(t) dt$$

# General monotonic algorithms (J. Salomon, G.T.)

state  $X \in H$ , control  $v \in E$ ,  $H, E =$  Banach spaces.

- $\partial_t X + A(t, v(t))X = B(t, v(t))$
- $\min_v J(v)$ ,  $J(v) := \int_0^T F(t, v(t), X(t)) dt + G(X(T))$ .
- $F, G: C^1 +$  **concavity** with respect to  $X$  (not  $v$ !)

$$\forall X, X' \in H, G(X') - G(X) \leq \langle \nabla_X G(X), X' - X \rangle$$

$\forall t \in \mathbb{R}, \forall v \in E, \forall X, X' \in H :$

$$F(t, v, X') - F(t, v, X) \leq \langle \nabla_X F(t, v, X), X' - X \rangle.$$



## Direct-adjoint equations and first lemma

$$\partial_t X + A(t, v(t))X = B(t, v(t))$$

$$X(0) = X_0$$

$$\partial_t Y_v - A^*(t, v(t))Y_v + \nabla_X F(t, v(t), X_v(t)) = 0$$

$$Y_v(T) = \nabla_X G(X_v(T)).$$

### Lemma

*Suppose that  $A, B, F$  are differentiable everywhere in  $v \in E$ , then there exists  $\Delta(\cdot, \cdot; t, X, Y) \in C^0(E^2, E)$  such that, for all  $v, v' \in E$*

$$J(v') - J(v) \leq \int_0^T \Delta(v', v; t, X_{v'}, Y_v) \cdot_E (v' - v) dt \quad (4)$$

# Well-posedness

$$J(v') - J(v) \leq \int_0^T \Delta(v', v; t, X_{v'}, Y_v) \cdot E(v' - v) dt \quad (5)$$

Remark: useful factorisation because can test at each step if  $J$  goes the right way; also can choose  $v'(t^*) = v(t^*)$  if pb.

Remark:  $\Delta(v', v; t, X, Y)$  has an explicit formula once the problem is given; also note the dependence on  $Y_v$  any not  $Y_{v'}$ .

## Lemma

*Under hypothesis on  $A, B, F, G, \theta > 0$*

$$\Delta(v', v; t, X, Y) = -\theta(v' - v) \quad (6)$$

*has an unique solution  $v' = \mathcal{V}_\theta(t, v, X, Y) \in E$ .*

# Well-posedness

Theorem (J. Salomon, G.T. Int J Contr, 84(3), 551, 2011)

*Under hypothesis ...*

- *the following eq. has a solution:*

$$\partial_t X_{v'}(t) + A(t, v')X_{v'}(t) = B(t, v') \quad (7)$$

$$v'(t) = \mathcal{V}_\theta(t, v(t), X_{v'}(t), Y_v(t)) \quad (8)$$

$$X_{v'}(0) = X_0 \quad (9)$$

- $\exists (\theta_k)_{k \in \mathbb{N}}$  such that  $v^{k+1}(t) = \mathcal{V}_{\theta_k}(t, v^k(t), X_{v^{k+1}}(t), Y_{v^k}(t))$
- $J(v^{k+1}) - J(v^k) \leq -\theta_k \|v^{k+1} - v^k\|_{L^2([0, T])}^2$ ;
- if  $v^{k+1}(t) = v^k(t) : \nabla_v J(v^k) = 0$ .

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# The Model : framework

- large economy: **continuum** of consumer agents
- time period:  $[0, T]$
- any household owns exactly one house and cannot move to another one until  $T$

# The Model : the agents

- **arbitrage** between insulation and heating. A generic player (agent) has an insulation level  $x \in [0, 1]$  ( $x = 0$ : no insulation,  $x = 1$ : maximal insulation)
- controlled process of the agent:  $dX_t = \sigma dW_t + v_t dt + dN_t(X_t)$  where  $v$  is the **control** parameter (insulation effort), the noise level  $\sigma$  is given.
- note that  $X_t$  is a diffusion process with reflexion, in the above equality,  $dN_t(X_t)$  has the form  $\chi_{\{0,1\}}(X_t) \vec{n} d\xi_t$  ( $\xi$  = local time at the boundary  $\{0, 1\} = \partial[0, 1]$  cf. Freidlin)
- initial density:  $X_0 \sim m_0(dx)$

# The Model : the costs

An agent of the economy solves a minimization problem composed of several terms:

- *Insulation acquisition cost*:  $h(v) := \frac{v^2}{2}$
- *Insulation maintenance cost*:  $g(t, x, m) := \frac{c_0 x}{c_1 + c_2 m(t, x)}$  increasing in  $x$  decreasing in  $m$  : **economy of scale, positive externality**. The agents should do the same choice, stay together. The higher is the number of players having chosen an insulation level, the lower are the related costs.
- *Heating cost*:  $f(t, x) := p(t)(1 - 0,8x)$  where  $p(t)$  is the unit heating cost (unit price of energy, say)

# The model - The minimization problem and MFG (1)

- Define the aggregate state cost:

$$\Phi(m) := \int_0^1 \left( p(t)(1 - 0,8x) + \frac{c_0 x}{c_1 + c_2 m(t, x)} \right) m(t, x) dx$$

and  $V = \Phi'$ .

- In the model, the agents have **rational expectations**, i.e they see  $m$  as given; we can write the individual agent's problem:

$$\left\{ \begin{array}{l} \inf_{v \text{ adm}} \mathbb{E} \left[ \int_0^T h(v(t, X_t^x)) + V[m](X_t^x) dt \right] \\ dX_t = v_t dt + \sigma dW_t + dN_t(X_t), X_0 = x \end{array} \right.$$



## The model - The minimization problem and MFG (2)

- We already know that it is linked with the optimal control problem:

$$\begin{cases} \inf_{v \text{ adm}} \int_0^T \int_0^1 h(v(t, x)) + \Phi(m_t)(t) dt \\ \partial_t m - \frac{\sigma^2}{2} \Delta m + \operatorname{div}(vm) = 0, \quad m|_{t=0} = m_0(\cdot), \\ m'(\cdot, 0) = m'(\cdot, 1) = 0 \end{cases}$$

- Finally, if  $\nu := \frac{\sigma^2}{2}$ , a **Mean field equilibrium** (Nash equilibrium with an infinite number of players) corresponds to a solution of the following system:

$$\begin{cases} \partial_t m - \nu \Delta m + \operatorname{div}(vm) = 0, \quad m|_{t=0} = m_0 \\ \nabla u = v \\ \partial_t u + \nu \Delta u + v \cdot \nabla u - \frac{u^2}{2} = \Phi'(m), \quad v|_{t=T} = 0 \end{cases} \quad (10)$$

# The model - externality & scale effect

The MFG framework is interesting to describe a situation which lives between two economical ideas: **positive externality** and **economy of scale**

- **positive externality**: positive impact on any agent utility NOT INVOLVED in a choice of an insulation level by a player
- **economy of scale**: economies of scale are the cost advantages that a firm obtains due to expansion (unit costs decrease)

## Criticism of the model:

- **stylised** from the "industrial" point of view
- not realistic (heating price, maintenance...)
- **transition effect** (continuous time, continuous space)
- **atomised** agent (her/his action has no influence on the global density, micro-macro approach)
- non-cooperative equilibrium with rational expectations

# Numerical simulations

- Optimization method: **Monotonic algorithm**

$$\left\{ \begin{array}{l} \partial_t m^{k+1} - \nu \Delta m^{k+1} + \operatorname{div}(v^{k+1} m^{k+1}) = 0, \quad m^{k+1}(x, 0) = m_0 \\ v^{k+1} = \frac{(\theta-1/2)v^k - \nabla u^k}{(\theta+1/2)} \\ \partial_t u^{k+1} + \nu \Delta u^{k+1} + v^{k+1} \cdot \nabla u^{k+1} - \frac{(u^{k+1})^2}{2} = \Phi'(m^{k+1}), \quad v^{k+1}(T) = 0 \end{array} \right. \quad (11)$$

- Discretization of the PDEs: **Godunov scheme** (to preserve the positivity of the density  $m$ )

- The costs:

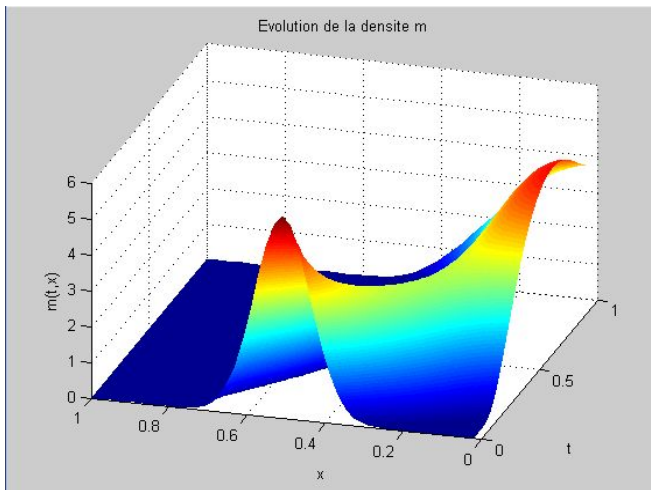
heating:  $f(t, x) = p(t)(1 - 0,8x)$

insulation:  $g(t, x, m) = \frac{x}{0.1+m(t,x)}$

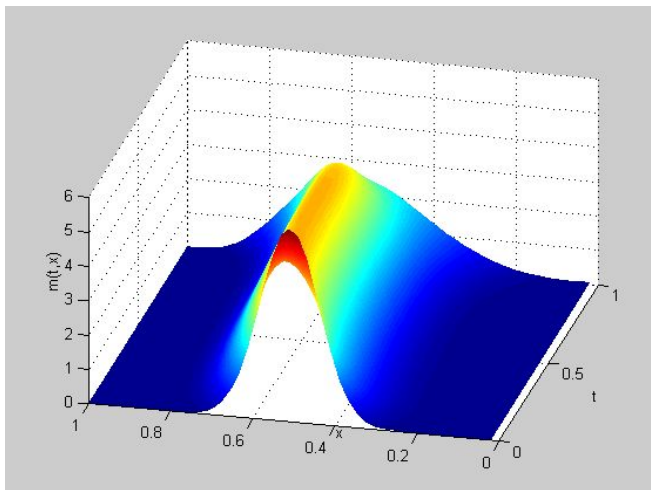
- *1st example*:  $p(t)$  constant / same choices
- *2d example*:  $p(t)$  reaching a peak (non constant) / multiplicity of equilibria

## Numerical results - First case

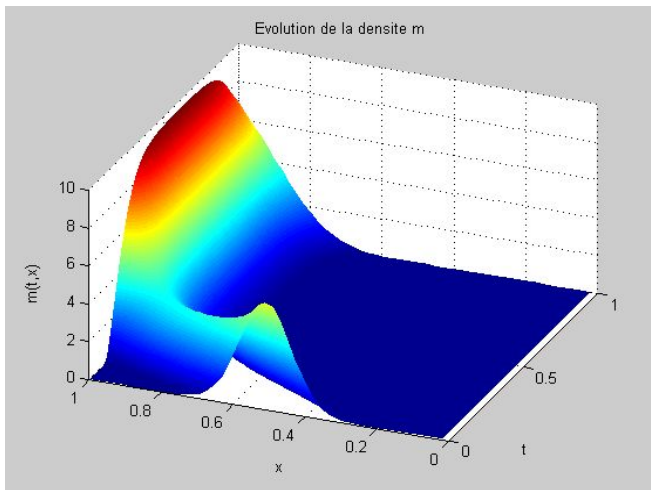
- the initial density of the householders is a gaussian centered in  $\frac{1}{2}$
- the time period and the noise are respectively  $T = 1$  and  $\nu = 0.07$
- the **energy price is constant** ( $p(t) \equiv 0, 3.2$  and  $10$ )



**Figure:** Numerical results :  $p(t) \equiv 0$ . Since the cost of energy is null all agents choose to heat their house, move to this choice together.



**Figure:** Numerical results :  $p(t) \equiv 3.2$ . Cost of energy is intermediary, agents keep their status.

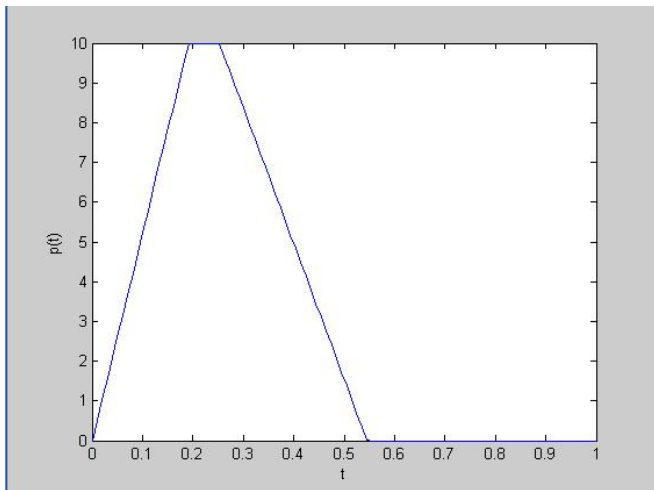


**Figure:** Numerical results :  $p(t) \equiv 10$ . Cost of energy is high, agents choose to better insulate, all have the same behavior.

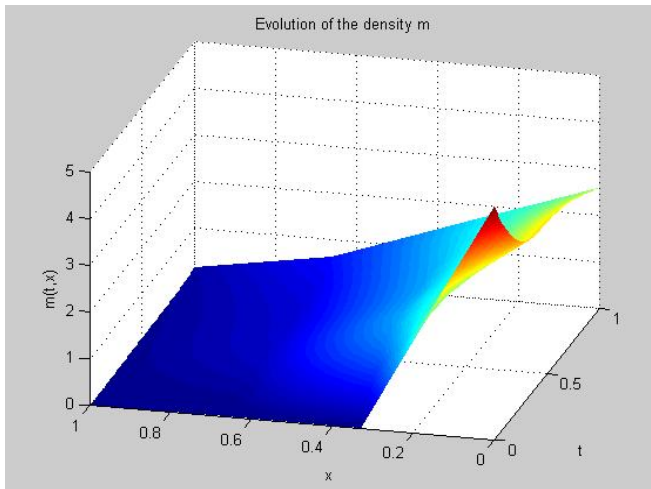


## Numerical results - Second case

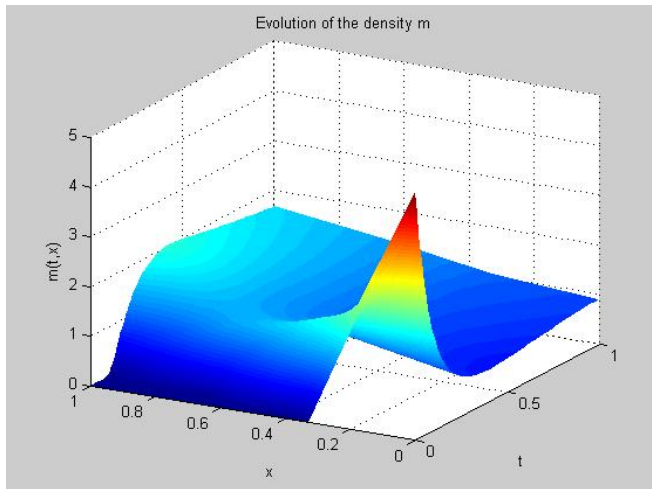
- the initial density of the agents is an approximation of a Dirac in 0.1 (*i.e* agents are not equipped in insulation material)
- the energy price is **not a constant parameter**, we look at the following case: the price first **reaches a peak** and then decreases to its initial level.



**Figure:** Numerical results -  $p(t)$ . Question: In such a case, can we find two Mean Field equilibria, the first related to the expectation of a higher insulation level, the second to the expectation of heating ?



**Figure:** Numerical results - One of the two equilibria: the energy consumption equilibrium. Agents expect that everybody will keep a low insulation level so there are no gains in insulating.



**Figure:** Numerical results - One of the two equilibria: the insulation equilibrium. Agents expect that everybody will better insulate, which makes insulating attractive.

# Multiplicity of equilibria - Incentive policy

- we found an **insulation-equilibrium** and an **energy consumption-equilibrium**
- from the ecological point of view: the best is the insulation-equilibrium
- **incentive public policies** could steer towards the "best" equilibrium (from a certain point of view) when the solution is not unique.

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# Liquidity from heterogenous beliefs and analysis costs (joint work with Min Shen, Université Paris Dauphine)

- Why do agents trade ? **Here: heterogenous beliefs and expectations**
- Liquidity : many definitions (bid/ask spread, rapidity to recover price after shock, max volume traded at same price etc). **Here: trading volume.**
- Several approaches: limit order book modeling and optimal order submission (Avellaneda et al. 2008) Heterogenous beliefs: asset pricing (working paper by Emilio Osambela), short sale constraints (Gallmeyer and Hollifield 2008) etc.,
  - **Specific investigation of this work: question on analysis time/cost**

## Heterogeneous beliefs and liquidity: the model

- One security of "true" value  $V$ .
  - each agent has each own estimation  $V \cdot A$  with  $A$  random variable with density  $\rho(A)$ , mean=1.
    - The estimation is uncertain with precision  $B(A)$  ( $B$  can be e.g.  $V^2/\sigma^2(A)$  ( $\sigma^2(A)$  = variance of the estimate)).
      - Precision can be improved paying  $f(B)$  and/or waiting for the estimation to converge or new data to be revealed.
    - Each agent has an utility function  $U(\text{mean}(\theta, B), \text{variance}(\theta, B))$  (equivalent: expected utility framework for normal variable). Linear situation  $U(x, y) = x - \lambda y$ . Here  $V \cdot \theta(A)$  = size of the position of agent at  $A$ 
      - Price  $Vp^A$  maximizes liquidity and equals offer and demand.
- Note:  $p^A$  is not necessarily equal to 1 even if the mean  $\mathbb{E}(A) = 1$ .



# Heterogeneous beliefs and liquidity: theoretical results

Technical framework: Mean Field Games by Lasry & Lions; Nash equilibrium

$$\text{mean}(\theta, B) = V\theta(A - p^A) - f(B); \text{variance}(\theta, B) = \theta^2 V^2 / B.$$

Theorem (M Shen, G.T. 2011)

*Under assumptions on functions  $f$  and  $U$  the equilibrium exists. Offer and demand functions are monotone with respect to  $p^A$ .*

Theorem (M Shen, G.T. 2011)

*Under assumptions on functions  $f$  and  $U$  if  $\rho$  is symmetric around  $p^1$  then (liquidity is maximal for  $p = p^1$  i.e.)  $p^A = p^1$ .*

# Heterogenous beliefs and liquidity: linear case $U = x - \lambda y$

## Theorem (M Shen, G.T. 2011)

*For the linear case the equilibrium relative price is:*

$$P^A = \frac{\int_0^\infty AB(A)\rho(A)dA}{\int_0^\infty B(A)\rho(A)dA}. \quad (12)$$

*The relative accuracy  $B(A)$  cost is*

$$B = (f')^{-1} \left( \frac{(A - P^A)^2}{2\lambda} \right). \quad (13)$$

# Heterogenous beliefs and liquidity: linear case $U = x - \lambda y$

The relative market price  $P^A$  is solution to the equation:

$$\frac{1}{2V\lambda} \int_0^\infty (A - P^A)(f')^{-1} \left( \frac{(A - P^A)^2}{2\lambda} \right) \rho(A) dA = 0 \quad (14)$$

The trading volume  $TV_f$  is

$$TV_f = \frac{P^A}{2\lambda} \int_0^\infty (A - P^A)_+(f')^{-1} \left( \frac{(A - P^A)^2}{2\lambda} \right) \rho(A) dA \quad (15)$$

## Theorem (anti-monotony of trading volume)

*Let  $f, g$  be two information cost functions such that  $g'(b) \geq f'(b)$  for any  $b \in \mathbb{R}_+$ . Then the trading volume satisfies  $TV_f > TV_g$ .*

Application: for constant total cost  $\int f(B)\rho(A)$  which is the greatest volume : is volume brought by best paid analysts ?

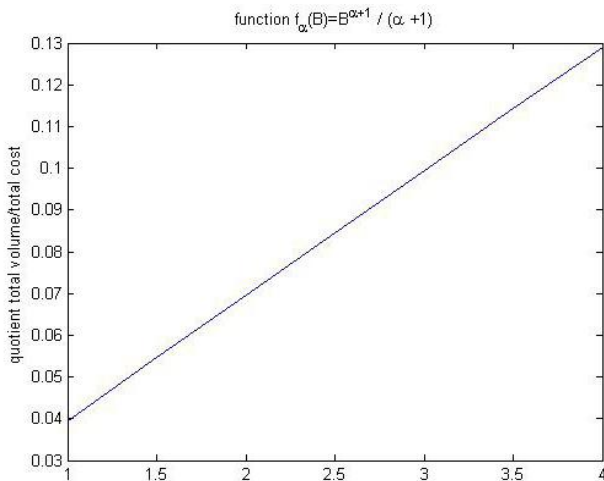


Figure: Quotient of the total volume over total cost for functions

$$f(B) = \frac{B^{\alpha+1}}{\alpha+1}$$