

An Optimal Rule of Thumb for Pollution Permits Allocation

Evangelina Dardati and Mar Reguant

ICE 08

August 2008

Goal

- Solve a dynamic strategic game with discrete states and continuous choices (similar to the one that Karl Schmedders presented)
- Introduce complementarity conditions
- Consider this dynamic strategic game as the second stage of a leader-follower game (planner in 1st state, duopoly in the 2nd)

Environment

- 2 big polluters \Rightarrow compete in a dynamic Cournot game
- There is another sector with small polluters
- Firms produce goods and need to back up emissions with pollution permits
- The 2 big polluters have market power in both the sectoral good market and the permits market
- Emissions depend on output and the emissions rate represented by $\theta_i \Rightarrow$ this is our state variable
- Firms can invest to be cleaner in the future
- Transition to next state depends on investment

Game

- Firm's production q_i
- Demand function $P(Q) = A - b(q_1 + q_2)$
- Profit function of firm i in period t is:

$$\Pi_{it} = p_t q_{it} - c_t[\theta_{it} q_{it} - z_{it}] - dx_{it}^2$$

where

c_t : cost of pollution permit

$$c_t = \gamma(q^f + \sum e_i - z^f - \sum z_i)$$

q^f : emissions of small polluters before introducing permits

z^f : permits given to small polluters

z_{it} : permits given to big polluters

x_{it} : investment

dx_{it}^2 : cost of investment

θ_{it} : efficiency

Dynamics

- States are $j = 1..S$
- $\theta_j = \Theta^{(j-1)} / (S - 1)$ for some $\Theta \in [0, 1]$
- Transition probabilities depend on investment and current state
- We constrain the transition to contiguous states
 - Prob of going to lower state (more efficient - lower emission rate)

$$\Rightarrow \frac{\theta x}{1 + \theta x + (1 - \theta)}$$
 - Prob of staying $\Rightarrow \frac{1}{1 + \theta x + (1 - \theta)}$
 - Prob of going to a higher state $\Rightarrow \frac{(1 - \theta)}{1 + \theta x + (1 - \theta)}$
- Note that investment affects mainly the probability of reducing a firm's emissions rate

Equations

- Bellman equation
- FOC with respect to quantities q_i
- $x_i \geq 0 \perp \frac{dV}{dx_i} \leq 0$
- Three equations and one complementarity per firm per state

LOG FILE WITH PATH SOLVER

Major Iteration Log

major	minor	func	grad	residual	step	type	prox	inorm	(label)
0	0	3	3	7.1624e-01		I	0.0e+00	1.0e-01	(_scon[1200])
1	1	4	4	5.9071e-03	1.0e+00	SO	0.0e+00	1.1e-03	(_scon[798])
2	1	5	5	2.3457e-06	1.0e+00	SO	0.0e+00	5.3e-07	(_scon[1199])
3	1	6	6	4.0926e-10	1.0e+00	SO	0.0e+00	2.5e-10	(_scon[285])

Major Iterations. . . . 3

Minor Iterations. . . . 3

Restarts. 0

Crash Iterations. . . . 2

Gradient Steps. 0

Function Evaluations. . 6

Gradient Evaluations. . 6

Basis Time. 0.438000

Total Time. 0.657000

Residual. 4.092554e-10

Path 4.7.01: Solution found.

5 iterations (2 for crash); 3 pivots.

6 function, 6 gradient evaluations.

Social Planner Problem

- Since there is market power in the permits market, initial allocation matters
- The planner uses the initial allocation as a policy instrument to maximize welfare
- A completely “non-parametric” rule is too highly dimensional (at least for us and by now)
- We solve for an optimal rule of thumb that depends on a single parameter ρ
- The planner sets for the optimal rule of thumb at time 0 and forever (taking into account strategic behavior of the firms)

$$z_i(\theta_i, \theta_{-i}) = \frac{\theta_{-i} + \rho}{\theta_i + \theta_{-i} + 2\rho} Z$$

where Z is the total number of permits

Social Planner Problem cont

$$\max_{\rho} W(\theta_0; \rho)$$

$$s.t. \quad W(\theta; \rho) = CS_{big}(\theta; \rho) + CS_{small}(\theta; \rho) \dots$$

$$\dots - d \sum_i x_i(\theta; \rho)^2 + \beta \sum_s Pr(\theta' | \theta, x; \rho) W(\theta'; \rho) \quad \forall i, \theta$$

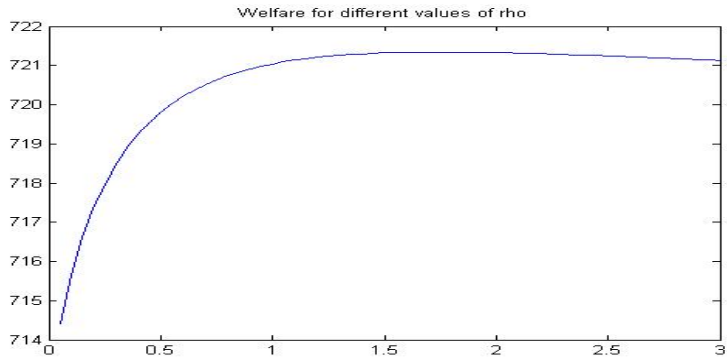
$$V(\theta; \rho) = \Pi(\theta; \rho) + \delta \sum_s Pr(\theta' | \theta, x; \rho) V(\theta'; \rho) \quad \forall i, \theta$$

$$\frac{dV}{dq} = 0 \quad \forall i, \theta$$

$$x \geq 0 \perp \frac{dV}{dx} \leq 0 \quad \forall i, \theta$$

Some computational issues

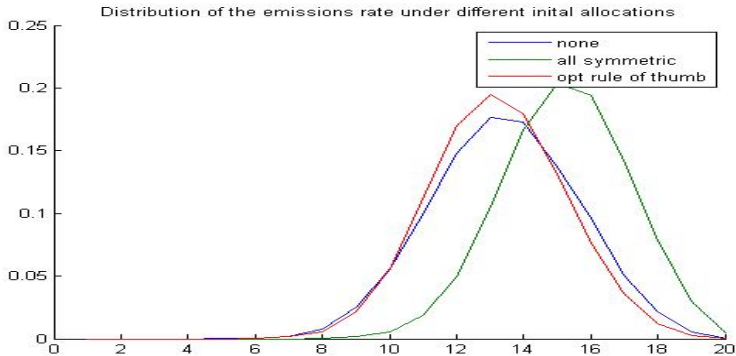
- Given that PATH was very efficient in the second stage, we thought of using the solver in this context as well
- However, the PATH solver does not allow for an objective function, so we need to move to KNITRO
- Luckily, KNITRO solves the optimization very rapidly and converges to our optimal rule of thumb
- We check SOC conditions
- We do a grid to evaluate our function and it seems well behaved. The maximum coincides with the solver solution.
- A starting value helps. We solve for the model for a fixed ρ using PATH in less than one second and then use it as an initial guess to solve the MPEC

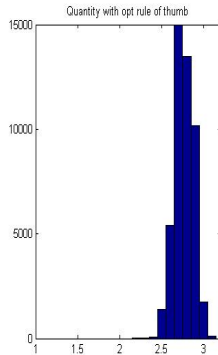
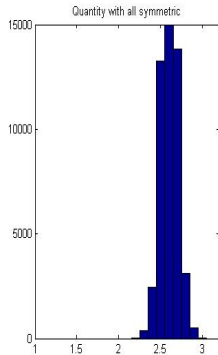
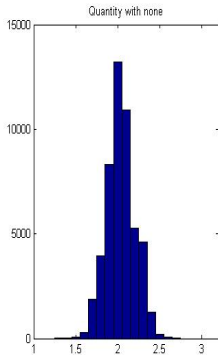


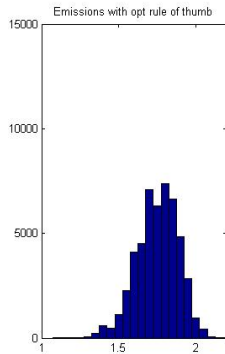
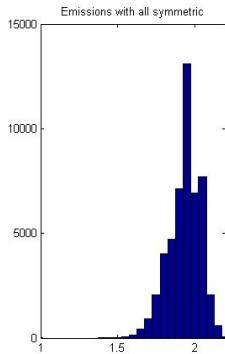
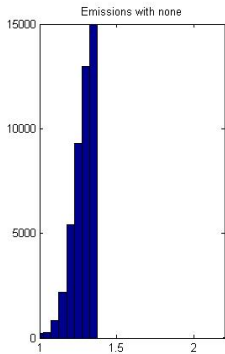
Our Experiment

We consider three different hypothetical cases:

1. The social planner gives no permits for free
2. The social planner distributes all permits to the oligopolists symmetrically
3. The social planner follows the optimal rule of thumb







A more flexible rule of thumb

- We consider two possibilities to make our rule of thumb more flexible:
 1. Make the formula of the rule of thumb more non-parametric adding higher order terms (normalize c_1 to 1) - simulation with linear and squared terms solves well with KNITRO.

$$z_i(\theta) = \frac{c_0 + c_1\theta_{-i} + c_2\theta_i + c_3\theta_{-i}^2 + c_4\theta_i^2}{2c_0 + (c_1 + c_2)(\theta_i + \theta_{-i}) + (c_3 + c_4)(\theta_i^2 + \theta_{-i}^2)} Z$$

2. Express the complete game as a complementarity problem by deriving the FOC conditions of the planner - taking policy functions into account. Solve using PATH.
- The polynomial approach works, and results suggest that other rules of thumb might perform better. To add more terms we should move to more adequate approximation functions
 - We have tried to push the second approach, but we did not quite get there

Conclusions

- We have presented a dynamic game with complementarities within a leader-follower framework
- The dynamic game with complementarities worked efficiently with the PATH solver
- The leader-follower game did not work with the PATH solver but worked with KNITRO
- The fact that the solvers have different purposes highlights the importance of understanding the details (or looking for optimization people)
- In our application, we solved for an optimal rule of thumb to allocate pollution permits that performed well
- It could be interesting to see if we can allow for a more flexible rule that accommodates more parameters