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Creti, Anna and Villeneuve, Bertrand

Université Paris X - Nanterre, Università Commerciale L. Bocconi
Milano, Université Paris-Dauphine

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Equilibrium Storage in a Markov Economy

Anna Cretì — Università Bocconi and Université Paris-Ouest
Bertrand Villeneuve — Université Paris-Dauphine and CREST*

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Abstract

We model an economy that alternates randomly between abundance and scarcity episodes. We develop an original method to characterize in detail the structure of the Markovian competitive equilibrium. Accumulation and drainage of stocks are the main focuses. Economically appealing comparative statics results are proved. We also characterize stationary distribution of states. We extend the model to discuss price stabilization policies, injection and release costs, and limited storage capacity. Overall, the analysis delineates the notion of “flexible economy.”

Keywords. Price stabilization, strategic stocks, supply risk

JEL codes. C61, Q48, L90.

*Corresponding author: bertrand.villeneuve@ensae.fr

1 Introduction

Storage models are hard to tackle. Even the simplest specification of the production and storage technologies leads to intractable equations. Most results involve proving existence and uniqueness of the equilibrium, together with some qualitative properties (prices are monotonic in stock and in storage cost, stockouts happen with positive probability and stocks have an upper limit). As far as simulation based econometrics (GMM estimators as in Deaton and Laroque, 1992, 1996) or the illustration of a theoretical possibility are concerned (convenience yield,¹ analysis of the Samuelson effect,² backwardation³ as in Routledge *et al*, 2000), this is a suitable approach. Our aim is not to extend the existing models but rather to propose a simple case set in continuous time to facilitate the parameterization of shock persistence, and to characterize finely the behavior of the economy.

Our approach is innovative as it does not rely on fixed point methods, but directly constructs the equilibrium, assuming that agents observe the random occurrence of one of two states, the so-called abundance and scarcity periods. We focus on a Markov competitive equilibrium in which prices only depend on the current state. The relationship between the state of the economy (the endogenous stocks and the exogenous random variable) and the price is informative about behavior of the economy. We fully describe the dynamics of accumulation and drainage. Stocks are smoothly piled up in an abundance state and smoothly drained in a crisis state. We can see that the upper bound of the stocks is never reached in finite time and we can also evaluate the speed at which stocks are drained out. Besides such qualitative results, we provide comparative statics on the upper bound of the stocks and equilibrium price schedules with respect to all the parameters of the model.

Price, as functions of the state, give a logically complete picture of the equilibrium. Nevertheless, the characterization of the stationary distribution of states has an intuitive appeal as it directly informs as to where the economy is likely to be. The frequency of stockouts as well as the propensity of the economy to adjust stocks can thus be assessed. The stationary distribution is described by differential equations, which opens up the way to qualitative analysis and comparative statics. The dependency of the shape of the state density to the parameters is addressed.

Our model is motivated by two economic issues placed high on the European policy agenda. Storage as determined in response to persistent shocks is instructive for the energy policy debate about the role of gas or petroleum strategic reserves to manage supply disruptions, especially when dependency on foreign resources raises serious concerns. The existing theoretical literature on energy supply security, mostly inspired by the theory of

¹The notion of convenience yield was introduced by the economists Kaldor and Working who studied the theory of storage. In the context of commodities, the convenience yield captures the benefit from owning a commodity minus the cost of storing it. The flow of benefits from storage (the reduction in production costs) drives a wedge between the price of a commodity today and its value in the future.

²The Samuelson effect arises when, for a given commodity, forward price volatility declines with the contract horizon.

³Backwardation occurs when the price of a commodity for the actual period exceeds the price for future periods.

exhaustible resources, considers either the extraction rate of one country when foreign import, though needed to complement national production, can suddenly default,⁴ or strategic behavior of consuming countries confronting oligopolistic or cartelized supply.⁵ However useful these analyses may be for the long run, they ignore the question of how to reach any desired stock level and how to deal with uncertainty about the duration of supply disruption.

We also shed some light on the banking of CO₂ pollution permits, a financial mechanism whose application is being discussed in the context of the European Trading Scheme, i.e. a market-based approach to environmental control. A number of research works have already analyzed the role of uncertainty in emission permit markets, studying in particular the SO₂ banking mechanism allowed by the American Clean Air Act. Most results concern optimum individual strategy and not the equilibrium. This limitation notwithstanding, several aspects of risk-averse utilities' are studied.⁶ Our analysis is not focused on financial phenomena, though the model could serve that purpose; this said, we can illustrate the precautionary motive for banking emission rights, when the output market alternates between booms and busts (namely, when electricity demand is influenced by unexpected climate constraints) or there is a sudden but temporary increase in input costs.

The economic relevance of our model and its practicality are illustrated by three extensions.

First, we study the impact of a constant price policy. This apparently extreme choice is instructive for both gas and pollution permit markets because a regulatory authority could be tempted to stabilize gas or permit prices around a particular target (or path) of prices. Understanding the mechanisms of equilibrium and comparison with more interventionist policy could serve as a modest guide (or a development thereof) for market design and regulation. In contrast to the previous abundant literature on storage and price stabilization,⁷ results are clear-cut: perfect price stabilization can be reached only if the economy is

⁴This trade-off has been analyzed by several authors (for example Stiglitz, 1977, Sweeney, 1977, Hillman and Van Long, 1983, Hugues Hallet, 1984).

⁵See for instance Nichols and Zeckhauser (1977), Crawford *et al* (1984), Devarajan and Weiner (1987), Hogan (1983).

⁶On the role of banking in smoothing permit prices, see Carlson and Sholtz (1994) and Godby *et al* (1997), on its effect on control costs, see Montero (1997). The study of equilibrium in Schennach (2000) assumes that risk-neutral firms minimize their expected discounted costs. When firms anticipate the possibility of a permit stockout, the expected change in marginal abatement costs could be negative. Potential permit stockout could partially explain normal backwardation in permit prices; the same mechanism is at the core of the results in Routledge *et al* (2000).

⁷Massel (1969), generalizing previous results by Waugh (1944) and Oi (1961), considers stabilization at exactly the mean price as a decision made to eliminate price fluctuations, presumably enhancing welfare. A costless stock established by an authority achieves the objective and enhances welfare. Welfare analysis of price stabilization has been extended to encompass alternative assumptions about price expectations, risk attitudes (Newbery and Stiglitz, 1981), and nonlinearities (Turnovsky, 1974, 1976, among others). Storage in this literature is made by a public authority, which is in charge of managing a buffer stock. Helmberger and Weaver (1977) is the only model that questions the optimality of stabilization schemes. The private storage industry and arbitrage opportunities are considered, instead, in modern dynamic stochastic models

prepared to let stocks go to infinity. This simple prediction gives a partial answer to doubts as to price stabilization models raised by Williams and Wright (1991), who affirm: “[...] the possible permutations of demand curvature, disturbance structure, initial conditions, supply elasticity and so forth seem nearly infinite. [...] That is the main point: few, if any, general propositions are possible.”

In a second extension of the model, we consider the impact of non-negligible injection and release costs. This better characterizes natural gas or oil storage. Starting from the observation that the commodity is different depending on whether it lies *outside* or *inside* the reservoir, we show that the results of our analysis are unaffected by this generalization.

Limited storage capacity is also a crucial issue. For instance, gas is often stored in specific natural facilities (such as salt caverns) that are scarce. Also, market rules for emission programmes could impose the use of banking CO₂ permits up to a given threshold only. Consistently with the idea of scarcity rent, we show that in the accumulation phase, the price for storage service suddenly jumps above marginal cost when the capacity saturates. Interestingly, we find that, in contrast with the unconstrained case, the maximal stock is attained in finite time if the state of abundance is sustained.

The paper is organized as follows. Section 2 sets up the model and section 3 describes the methodology we follow to solve it. Section 4 characterizes the solution qualitatively and quantitatively and proposes comparative statics. Section 5 exposes the statistical properties of the model. Section 6 is devoted to applications and extensions of the model, while Section 7 concludes on the notion of economic flexibility. Proofs are relegated to the Appendix.

2 The model

2.1 Assumptions and parameters

The economy can be in two states, abundance (A) or scarcity (C , for crisis). A generic state is denoted σ . Time is continuous and the state changes following Bernoulli processes: the passage from A to C occurs with probability rate λ_C , and the passage from C to A with probability rate λ_A . For example, the probability that the economy switches from A to C in a time interval dt is $\lambda_C dt$, where λ_C is the publicly known parameter of this survival process. This simple Markov structure captures the fact that regimes have uncertain nature and duration that can be quantified statistically. The stationary probabilities of states A and C are, respectively, $\Pr[A] = \frac{\lambda_C}{\lambda_C + \lambda_A}$ and $\Pr[C] = \frac{\lambda_A}{\lambda_C + \lambda_A}$. The ratio λ_A/λ_C represents the relative frequencies of the two states.

Consumers and producers observe the current price and the state $\sigma (= A, C)$. Their aggregate behavior is summarized by the “excess supply functions” $\Delta_\sigma[\cdot]$ defined over \mathbb{R}_+^* , where $\Delta_\sigma[p]$ is the difference in state σ and for price p between current *primary produc-*

with i.i.d disturbances (Williams and Wright, 1991).

tion and current *final consumption*.⁸ For example, $\Delta_C[\cdot]$ incorporates a supply shock and adaptation of demand to the crisis.

Excess supply function $\Delta_\sigma[\cdot]$ is increasing and has a unique finite positive zero in R_+^* , denoted by p_σ^* ; this is the price at which the spot market would be balanced without recourse to storage. Naturally, we assume that the abundance static equilibrium price p_A^* is strictly smaller than the crisis static equilibrium price p_C^* .

$\Delta_\sigma[\cdot]$ is a *flow* in the sense that if price p is sustained for the interval dt , then the quantity that is stored or released is $\Delta_\sigma[p]dt$. Thus, if we denote the total inventories in the economy by $S \geq 0$, conservation of matter imposes the following conditions

$$\begin{cases} \frac{dS}{dt} = \Delta_\sigma[p] & \text{if } S > 0 \text{ or } \Delta_\sigma[p] > 0, \\ \frac{dS}{dt} = 0 & \text{if } S = 0 \text{ and } \Delta_\sigma[p] \leq 0. \end{cases} \quad (1)$$

Therefore, if the current price p is above p_σ^* , then the economy stores ($\Delta_\sigma[p] > 0$); if p is below p_σ^* , then the economy draws on inventories ($\Delta_\sigma[p] < 0$).

Storers are assumed to be risk-neutral price-takers with rational expectations, so that the price dynamics will be driven by arbitrage. Storage exhibits constant returns to scale. Carrying costs consist of the opportunity cost of capital (r being the interest rate) and a cost c (per unit of commodity and per unit of time).^{9,10} In equilibrium, the current price at date t equals the expected price at $t + dt$ net of the carrying costs.

2.2 Fundamental equations

We focus on Markovian equilibria where the commodity price only depend on the state variables. We characterize differentiable functions $p_C[S]$ and $p_A[S]$ such that (1) is verified as well as the arbitrage conditions: for all $S > 0$ and a time increment dt , the no-arbitrage equations

$$p_C[S] + cdt = (1 - rdt) ((1 - \lambda_A dt) \cdot p_C[S + dS] + \lambda_A dt \cdot p_A[S + dS]), \quad (2)$$

$$p_A[S] + cdt = (1 - rdt) ((1 - \lambda_C dt) \cdot p_A[S + dS] + \lambda_C dt \cdot p_C[S + dS]). \quad (3)$$

This is a first-order approximation. We let dt converge to 0 and neglect second-order terms, thus using (1), (2) and (3) become

$$\Delta_C[p_C] \cdot \frac{dp_C}{dS} = (r + \lambda_A)p_C - \lambda_A p_A + c, \quad (4)$$

$$\Delta_A[p_A] \cdot \frac{dp_A}{dS} = (r + \lambda_C)p_A - \lambda_C p_C + c. \quad (5)$$

⁸This modeling is rationalizable with agents maximizing intertemporal utility or profit, provided objectives are time separable and quasi-linear. See Chaton *et al* (2009).

⁹The assumption in Deaton and Laroque (1992, 1996) and Routledge *et al* (2000) is that a constant fraction of the stock vanishes every period. This type of cost can be included, *via* a renaming of variables, in r . Our variation is well suited to natural resources.

¹⁰A more general structure with injection and withdrawal costs and limited storage capacity is discussed in Section 6.

We search for the price functions that fully characterize the equilibrium. More precisely, we search for $p_A[\cdot]$ and $p_C[\cdot]$ defined over an interval $[0, S^*]$ where S^* , which is to be found, is the maximum stock. Besides the differential equations above, we know that $p_C[0] = p_C^*$ (when stocks are empty in a crisis, consumers just consume what is produced at the same time), and $p_A[S^*] = p_A^*$ (when inventories are maximum, nobody stores more).

Mathematically, we have a boundary value problem. In a standard problem, two differential equations with two initial conditions are easily solved qualitatively and numerically, if not in closed form.

However, here both initial conditions are on singular points:

- $\Delta_C[p_C[0]] = 0$ meaning, if we look at equation (4), that $\frac{dp_C[0]}{dS}$ is infinite;
- at S^* , both factors in the LHS of (5) are null; moreover S^* is only implicitly defined, its value being determined as the point where accumulation stops ($p_A[S^*] = p_A^*$ implies that $\Delta_A[p_A[S^*]] = 0$).

2.3 Summary of the notation

t : time;

σ : generic exogenous state; C : Scarcity/Crisis; A : Abundance;

λ_C : probability rate that state passes from A to C ;

λ_A : probability rate that state passes from C to A ;

$\Delta_\sigma[p]$: excess supply for price p in state σ (per unit of time);

p_σ^* : zero of Δ_σ ;

S : stocks at a given date;

c : marginal storage cost (per unit of time);

r : riskfree interest rate.

3 The equilibrium

This section establishes that in a Markovian equilibrium, stocks are smoothly piled up in state A and smoothly drained in state C . Moreover, stocks vary between 0 and the upper bound S^* to be determined, whereas the price is bounded below by p_A^* and above by p_C^* .

We analyze the system of equations in a phase diagram (p_C, p_A) in which S is the underlying parameter. This representation facilitates the demonstration of existence and uniqueness of the equilibrium. Most importantly for economic intuition, it enables qualitative results and comparative statics.

Readers most interested in the economic implications can skip this section.

3.1 Phase diagram

Let us define x_σ , a function of S , by differential equation $x'_\sigma = \Delta_\sigma[p_\sigma] \cdot p'_\sigma$, with $x_C[0] = 0$ and $x_A[S^*] = 0$. Using this notation in equations (4) and (5), we obtain

$$x'_C = (r + \lambda_A)p_C - \lambda_A p_A + c, \quad (6)$$

$$x'_A = -\lambda_C p_C + (r + \lambda_C)p_A + c. \quad (7)$$

We have an autonomous (S is not in the equations) system of separated variables. We draw the two-dimensional phase diagram with $p_C \leq p_C^*$ on the horizontal axis and $p_A \geq p_A^*$ on the vertical one. This quadrant is partitioned into three regions, separated by the isoclines where $x'_C = 0$ and $x'_A = 0$ are null, i.e.

$$(r + \lambda_A)p_C - \lambda_A p_A + c = 0 \text{ and } p_C \leq p_C^*, p_A \geq p_A^*, \quad (CC')$$

$$-\lambda_C p_C + (r + \lambda_C)p_A + c = 0 \text{ and } p_C \leq p_C^*, p_A \geq p_A^*. \quad (AA')$$

(CC') is never empty since by assumption $p_C^* > p_A^*$. (CC') is above (AA') in the considered quadrant (see Appendix A.1).

The phase diagram in Figure 1 indicates the shape and relative positions of the trajectories satisfying the motion equations (6) and (7). We define the lowest region, **I**, as the triangle having (AA') as a side and (p_C^*, p_A^*) as a vertex; the intermediate region, **II**, lies between the two lines, and the highest region **III** is above (CC') . In **I**, $p'_C < 0$ and $p'_A < 0$; in **II**, $p'_C > 0$ and $p'_A > 0$; in **III**, $p'_C > 0$ and $p'_A > 0$.¹¹

The intersection between (AA') and horizontal straight line $p_A = p_A^*$ is especially remarkable. We denote it as $\Omega = (\frac{r+\lambda_C}{\lambda_C}p_A^* + \frac{c}{\lambda_C}, p_A^*)$.

3.2 Characterization

Proposition 1

1. *In equilibrium, the support of S is an interval $[0, S^*]$.*
2. *Stocks are drained during scarcity episodes and accumulated during abundance episodes ($\Delta_C[p_C[S]] \leq 0$ and $\Delta_A[p_A[S]] \geq 0$ over $[0, S^*]$).*
3. *The equilibrium trajectory $\{(p_C[S], p_A[S]) | S \in [0, S^*]\}$ is in region **I**. The trajectory starts with $p_C[0] = p_C^*$ and stops at $(p_C[S^*], p_A[S^*]) = \Omega$.*
4. *There is storage in equilibrium (i.e. $S^* > 0$) if and only if*

$$(r + \lambda_C)p_A^* + c < \lambda_C p_C^*.$$

5. *The equilibrium is unique.*

¹¹This comes from the facts that in **I**, $x'_C > 0$ and $x'_A < 0$; in **II**, $x'_C > 0$ and $x'_A > 0$; in **III**, $x'_C < 0$ and $x'_A > 0$.

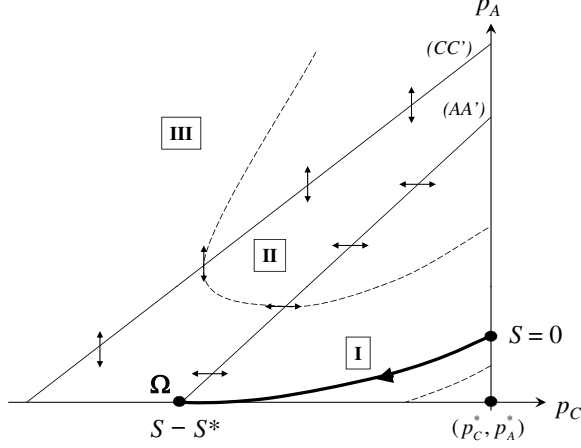


Figure 1: Phase diagram.

The overall behavior of the prices in equilibrium can be summarized as follows: the unique equilibrium trajectory is necessarily in **I**, S varies in the interval $[0, S^*)$ and $p_A[S] > p_A^*$ and $p_C[S] < p_C^*$ for $S > 0$ and $S < S^*$, and $p_A[S]$ and $p_C[S]$ decrease with respect to the level of the stock.

Figure 1 shows the shape of the equilibrium trajectory. The condition for positive storage $(r + \lambda_C)p_A^* + c < \lambda_C p_C^*$ has a precise economic meaning: crises have to be sufficiently likely and/or sufficiently marked to justify storage. If this condition were not satisfied, there would be no stocks in equilibrium: the price would alternate back and forth between p_A^* in state A and p_C^* in state C . To avoid this uninteresting case, we assume that the condition for positive storage is satisfied.

3.3 Computations

Numerically, the argument used in the proof of Proposition 1 (point 5) has an extremely useful implication: any starting point close to Ω is on a trajectory that is closer to the equilibrium as we go to the right (i.e. as S decreases) in the phase diagram. In other terms, we can control the maximum error on prices at the starting point (close to Ω) given that the system of equations is self-correcting as S decreases.

Moreover, the system of equations (6, 7) is autonomous, so leaving aside boundary conditions, we know that if $(p_C[S], p_A[S])$ follows the equilibrium trajectory for $S \in [\underline{S}, \bar{S}]$, so does $(p_C[S + \theta], p_A[S + \theta])$ for $S \in [\underline{S} - \theta, \bar{S} - \theta]$ where θ is an arbitrary real.

We can now suggest the following algorithm.

Algorithm 1 (Trajectory)

1. Fix arbitrarily the upper bound at some arbitrary value \bar{S} .
2. Choose $\varepsilon > 0$ as small as needed. Consider the trajectory through $(p_C[\bar{S}] = \frac{r+\lambda_C}{\lambda_C} p_A^* + \frac{c}{\lambda_C}, p_A[\bar{S}] = p_A^* + \varepsilon)$, a point above Ω .
3. Solve the differential equations numerically and find the stock $\underline{S} < \bar{S}$ such that $p_C[\underline{S}] = p_C^*$.
4. Shift the calculated functions p_C and p_A to the left by an amount \underline{S} ; $S_\varepsilon^* = \bar{S} - \underline{S}$ approximates the upper bound S^* .

The error is controlled by ε (a uniform bound on the error). Remark that along a trajectory,

$$\frac{dS}{dp_C} = \frac{\Delta_C[p_C]}{(r + \lambda_A)p_C - \lambda_A p_A + c} \quad (8)$$

is well defined over the range of p_C i.e. $[\frac{r+\lambda_C}{\lambda_C} p_A^* + \frac{c}{\lambda_C}, p_C^*]$ (trajectory is bounded away from (CC')). Thus we can calculate S^* as accurately as desired by computing:

$$S_\varepsilon^* = \int_{\frac{r+\lambda_C}{\lambda_C} p_A^* + \frac{c}{\lambda_C}}^{p_C^*} \frac{\Delta_C[p_C]}{(r + \lambda_A)p_C - \lambda_A p_A + c} dp_C. \quad (9)$$

4 Behavior of the economy

4.1 Comparative statics

In the absence of an explicit expression of price functions and S^* , the comparative statics relies on exploitation of the phase diagram.

Proposition 2 (Comparative statics) *For all S in the support, and for all states $\sigma = C, A$*

$$\frac{\partial p_\sigma^*[S]}{\partial c} < 0; \quad \frac{\partial p_\sigma^*[S]}{\partial r} < 0; \quad \frac{\partial p_\sigma^*[S]}{\partial \lambda_A} < 0; \quad \frac{\partial p_\sigma^*[S]}{\partial \lambda_C} > 0. \quad (10)$$

and consequently

$$\frac{\partial S^*}{\partial c} < 0; \quad \frac{\partial S^*}{\partial r} < 0; \quad \frac{\partial S^*}{\partial \lambda_A} < 0; \quad \frac{\partial S^*}{\partial \lambda_C} > 0. \quad (11)$$

The interpretations are straightforward. An increase in the unit storage costs discourages accumulation, thus at any level of the stocks, the value of the commodity is smaller. Storers will tend to pile up stocks more slowly in abundance, and to run them down faster during crisis. Also, rarer crises diminish the expected yield from storing, to the same effect. This logic has direct consequences on the comparative statics of the limit stock: the value S^* , defined as the solution to equation $p_A^*[S] = p_A^*$, must decrease if function $p_A^*[S]$ is diminished.

Linear case. The effects of varying excess supply *functions* are intricate if we do not restrict the analysis to a specific parametric family. For example, in the noteworthy case of linear excess supply functions, precise results can be found.

Proposition 3 (Linear case) *Assume that*

$$\Delta_\sigma[p_\sigma] = \beta_\sigma(p_\sigma - p_\sigma^*) \text{ with } \beta_\sigma > 0 \text{ and } p_\sigma^* > 0. \quad (12)$$

1. For all S in the support and all states $\sigma = C, A$

$$\frac{\partial p_\sigma^*[S]}{\partial p_C^*} > 0; \frac{\partial p_\sigma^*[S]}{\partial p_A^*} > 0; \frac{\partial p_\sigma^*[S]}{\partial \beta_C} > 0; \frac{\partial p_\sigma^*[S]}{\partial \beta_A} < 0. \quad (13)$$

and consequently

$$\frac{\partial S^*}{\partial p_C^*} > 0; \frac{\partial S^*}{\partial \beta_C} > 0; \frac{\partial S^*}{\partial \beta_A} < 0. \quad (14)$$

The sign of $\frac{\partial S^*}{\partial p_A^*}$ is ambiguous.

2. As a function of β_C and β_A , $p_\sigma^*[S]$ and consequently S^* are homogeneous of degree 1.

It is plain that a bigger p_C^* should increase the value of storage, hence the effect on prices and maximum stocks. In contrast, a bigger p_A^* has two effects: on the one hand, it increases the price at which stocks are built and thus prices in crisis have to increase altogether to motivate positive holding; on the other hand, the range of prices tightens, meaning that potential gains from the occurrence of a crisis could vanish at smaller values of S . This explains the ambiguity of the impact of p_A^* on S^* .

A higher parameter β_C means that a given release has a less depressing effect on the price. In other terms, the profitability of storing in view of releasing at high price when state C arises is better warranted. This gives incentives to store more. A higher parameter β_A implies that building stocks is easier, since piling up has a lesser inflationary effect on the price, hence the negative effect on the equilibrium price. The second point of the proposition illustrates that the first effect dominates when β_C and β_A are increased in the same proportion.

4.2 Approximate price functions

To better describe the behavior of the economy, we clarify the properties of the equilibrium when stocks are almost empty or close to their maximum. We see in particular how stocks are drained down and why the maximum stocks are not attained in finite time.

Draining out the stock. At $S = 0$, x'_C is finite and different from 0 (see equation 6); more precisely, $x_C[S] \underset{0}{\sim} K_C S$ with $K_C = (r + \lambda_A)p_C^* + c - \lambda_A p_A[0] > 0$. This implies (see Appendix A.5) that

$$p_C[S] - p_C^* \underset{0}{\sim} -\sqrt{\frac{2K_C}{\Delta'_C[p_C^*]}} S^{1/2} \quad (15)$$

p_C is vertically tangent at 0 (Figure 2). As a consequence, if the economy stays in crisis, the complete drainage of the stocks happens in finite time. To show this, it suffices to integrate in a neighborhood of 0 the differential equation

$$\frac{dS}{dt} = \Delta_C[p_C[S[t]]], \quad (16)$$

where the RHS can be replaced by its approximation. As long as the economy stays in crisis, starting—without loss of generality—with S_0 at date 0, the integration of equation (16) yields

$$S(t) \simeq \left(\sqrt{S_0} - \sqrt{\frac{\Delta'_C[p_C^*]K_C}{2}} t \right)^2. \quad (17)$$

Drainage exhibits smooth landing: the limit of the rate of withdrawal is zero; but drainage time is finite. It is approximately

$$T_0 \simeq \sqrt{\frac{2S_0}{\Delta'_C[p_C^*]K_C}}. \quad (18)$$

This implies that the economy is protected only twice as long when stocks are quadrupled.

The comparative statics on K_C is based on (10) in Proposition 2. We have $\frac{\partial K_C}{\partial \lambda_C} < 0$, meaning quite naturally, that a larger propensity to return to the scarcity state slows down drainage (precaution). Also, $\frac{\partial K_C}{\partial c} > 0$ and $\frac{\partial K_C}{\partial r} > 0$, meaning that higher storage costs accelerate drainage for given stocks. Remark that $\frac{\partial K_C}{\partial \lambda_A} = (p_C^* - p_A[0]) - \lambda_A \frac{\partial p_A^*[0]}{\partial \lambda_A} > 0$: a higher propensity to return to abundance also accelerates drainage (preservation value is diminished).

Replenishing. The upper bound S^* corresponds, as we noticed in subsection 3.2, to a singular point. The calculation of an approximate solution requires several steps. We show in the Appendix A.5 that $x_A[S] - x_A[S^*] \underset{S^*}{\sim} K_A^2 (S - S^*)^2$, where K_A is a non-negative real number. This implies that p_A has a negative finite non-null derivative at S^* (Figure 2):

$$p_A[S] - p_A^* \underset{S^*}{\sim} K_A (S - S^*). \quad (19)$$

Even if the economy stays in a state of abundance, the upper bound S^* is never reached in finite time. The reasoning reminds us of Zeno's classical paradox, Achilles and the Tortoise. As p_A covers half its difference with the limit p_A^* , the variation rate of the stock per unit of time, namely Δ_A , is approximately halved (linear approximation of excess

demand at p_A^*), meaning that the convergence speed dS/dt is approximately halved. This implies that, whatever the proximity of the target, the duration to cover half the distance to the target is approximately constant, thus the target is never attained.

Example. Using the algorithm of subsection 3.3, we solve numerically the system with the parameters in Table 1. The time unit could be the year. We find approximately $S^* \simeq 9.5$. See Figure 2.

Table 1: Parameter values				
Financial and physical costs	$r = .1$	$c = .1$		
Linear excess supply	$\beta_C = 1$	$p_C^* = 5$	$\beta_C = 5$	$p_A^* = 1$
Rates of jumps	$\lambda_C = 1$	$\lambda_A = 1$		

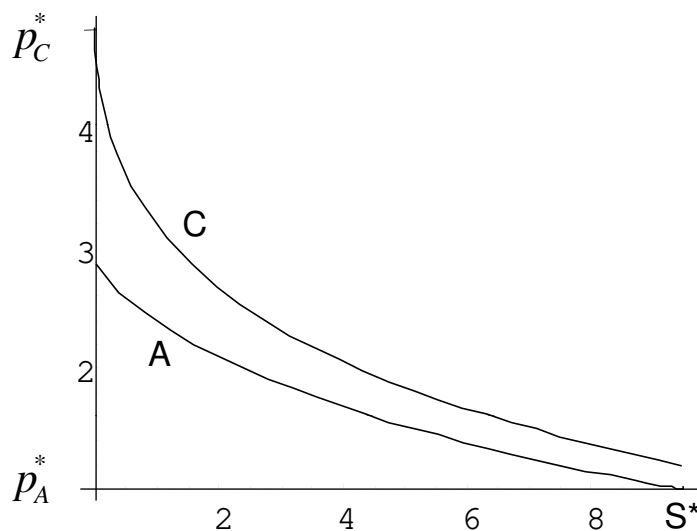


Figure 2: Price functions.

5 Stock statistics

A state is described by the stock S and the conjuncture (C or A). All states $S \in [0, S']$ with $S' < S^*$ are crossed in finite time with probability 1: starting from an arbitrary state, it is easy to find a history (or a set of histories) that leads to another arbitrary state in finite time, this history being associated with positive probability (it suffices to have a long enough accumulation followed by a long enough drainage, or the other way around). Since the equilibrium is Markovian, this property guarantees that there is a

unique stationary distribution. This section is essentially devoted to the analysis of this stationary distribution.

5.1 Dynamics

The statistical evolution of the system over time when we observe perfectly the state at a given date is covered in Appendix A.6.

A feature of the model is that in the long run initial information loses relevance. Interior states (i.e. values of S different from 0 and S^*) are just crossed as accumulation or drainage goes on; boundary states, if reached, remain in force until a downward or upward jump occurs. Thus, in the long run, $S = 0$ and $S = S^*$ are associated with probabilities whereas values in between are associated with densities.

Densities. Assume that, for interior values of the stock $S \in (0, S^*)$, a density $f_\sigma[S, t]$ (with $\sigma = C, A$) represents the information we have on the system. Take $\sigma = C$ to fix ideas. Choose S_1 and S_2 ($0 < S_1 < S_2 < S^*$) two levels of the stocks. By definition

$$\Pr[C, S \in [S_1, S_2], t] = \int_{S_1}^{S_2} f_C[S, t] dS. \quad (20)$$

This gives

$$\begin{aligned} \frac{d \Pr[C, S \in [S_1, S_2], t]}{dt} &= f_C[S_1, t] \cdot \Delta_C[p_C[S_1]] - f_C[S_2, t] \cdot \Delta_C[p_C[S_2]] \\ &\quad + \lambda_C \int_{S_1}^{S_2} f_A[S, t] dS - \lambda_A \int_{S_1}^{S_2} f_C[S, t] dS, \end{aligned} \quad (21)$$

where the first two terms represent the endogenous evolution of the stocks if the economy remains in crisis, and the third and fourth terms represent the exogenous jumps in and out of the segment due to state changes. Figure 3 illustrates this probability balance.

To find the dynamics of the density, we make S_2 converge toward S_1 to get

$$\frac{df_C[S, t]}{dt} = -\frac{d}{dS} (f_C[S, t] \cdot \Delta_C[p_C[S]]) + \lambda_C f_A[S, t] - \lambda_A f_C[S, t]. \quad (22)$$

Similarly

$$\frac{df_A[S, t]}{dt} = -\frac{d}{dS} (f_A[S, t] \cdot \Delta_A[p_A[S]]) + \lambda_A f_C[S, t] - \lambda_C f_A[S, t]. \quad (23)$$

Probabilities. States $S = 0$ or S^* are associated with probabilities. We have¹²

$$\frac{d \Pr[C, 0, t]}{dt} = -\lambda_A \Pr[C, 0] - \lim_{S \rightarrow 0} (f_C[S, t] \cdot \Delta_C[p_C[S]]), \quad (24)$$

¹²This expression can be derived from (21) with $S_1 = 0$ and by letting S_2 converge to 0.

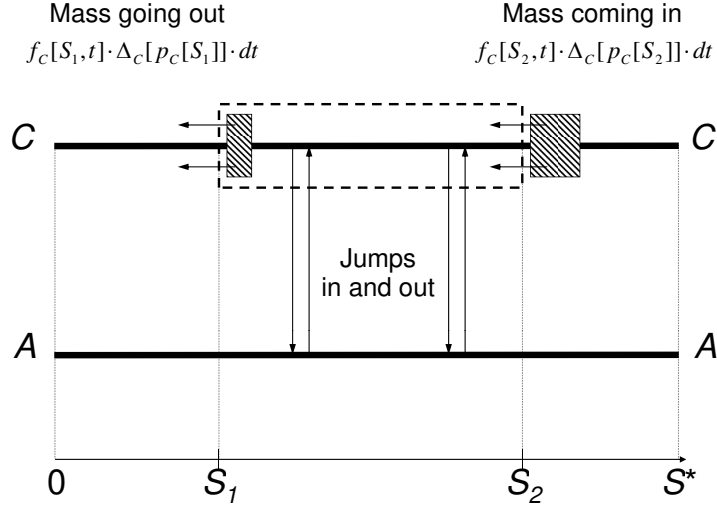


Figure 3: Probability balance between t and $t + dt$.

where the first-term represents jumps out (jumps in are negligible since $\Pr[A, 0, t] = 0$: due to accumulation, this state is left as soon as attained), and the second term represents the depletion of the last remaining stock.

Similarly,

$$\frac{d\Pr[A, S^*, t]}{dt} = -\lambda_C \Pr[A, S^*] + \lim_{S \rightarrow S^*} (f_A[S, t] \cdot \Delta_A[p_A[S]]). \quad (25)$$

5.2 Stationary distribution

The study of stationary distribution can use directly the preceding analysis. We denote the stationary densities by $f_\sigma^*[S]$ for all $S \in (0, S^*)$. Define $\phi_C[S] \equiv f_C^*[S] \cdot \Delta_C[p_C[S]]$ and $\phi_A[S] \equiv f_A^*[S] \cdot \Delta_A[p_A[S]]$ (density flows). Dropping the time-dependency factor, and replacing the rates of variation of the stocks by their equilibrium values, equations (22,23) become the system of *ordinary* differential equations

$$\frac{d\phi_A}{dS} = \lambda_A f_C^* - \lambda_C f_A^*, \quad (26)$$

$$-\frac{d\phi_C}{dS} = \lambda_A f_C^* - \lambda_C f_A^*. \quad (27)$$

We also have from (24,25)

$$\Pr[0] = \frac{1}{\lambda_A} \lim_{S \rightarrow 0} \phi_C, \quad (28)$$

$$\Pr[S^*] = \frac{1}{\lambda_C} \lim_{S \rightarrow S^*} \phi_A. \quad (29)$$

Remark that

$$-\phi_C[S] = \phi_A[S]. \quad (30)$$

Indeed, consider the open system $\{(C, s), (A, s) | s \in [0, S]\}$, this equation states that, in a stationary distribution, density flows in (at S for state C) equal flows out (at S for state A). Jumps do not matter since they happen within the system.

Equations (26) and (27) collapse to:

$$\frac{d\phi_A}{dS} = - \left(\frac{\lambda_A}{\Delta_C[p_C[S]]} + \frac{\lambda_C}{\Delta_A[p_A[S]]} \right) \phi_A. \quad (31)$$

This first order ordinary differential equation is well defined for $S \in]0, S^*[$ and can be solved numerically. The Cauchy-Lipschitz theorem is applicable.

Algorithm 2 (Stationary distribution)

1. Calculate equilibrium prices $p_C[S]$ and $p_A[S]$.
2. Fix arbitrarily $\phi[S]$ as an initial condition for some $S \in (0, S^*)$.
3. Solve numerically the differential equation (31) over $]0, S^*[$.
4. Calculate conditional densities f_A^* and f_C^* .
5. Calculate the integrals over $]0, S^*[$ of f_A^* and f_C^* .
6. Remark that $\Pr[S^*] = 0$. Use step 5 to calculate the residual $\Pr[0]$ using the facts that $\Pr[C] = \lambda_A/(\lambda_C + \lambda_A)$ and $\Pr[A] = \lambda_C/(\lambda_C + \lambda_A)$.
7. Normalize f_A^* and f_C^* so that the total probability mass equals 1.

Step 3 must be analyzed in detail. Indeed, $\Delta_C[p_C[0]] = \Delta_C[p_C^*] = 0$ and $\Delta_A[p_A[S^*]] = \Delta_A[p_A^*] = 0$, meaning that ϕ might diverge in such a way that normalization is impossible (integrals at step 5 could diverge). In fact, we check in Appendix A.7, that

$$\int_0^{S^*} f_C^*[S] dS < \infty \text{ and } \int_0^{S^*} f_A^*[S] dS < \infty. \quad (32)$$

Remark also that the integrals in step 5 can be calculated as accurately as needed, implying that the residual for $\Pr[0]$ in step 6 can be computed with the same accuracy.¹³

The numerical analysis gives instructive results on the overall behavior of the economy. How frequent are stockouts, i.e. how much is $\Pr[S = 0]$ compared to $\Pr[C](= \frac{\lambda_C}{\lambda_C + \lambda_A})$? Is the economy often close to having maximum stocks or is S^* a practically unapproachable limit? The last question can be addressed theoretically by characterizing the shape of the

¹³The alternative method would be to use property (28); however, the numerical accuracy of the limit of ϕ_C at 0 is not sufficient to recommend this procedure.

density of the stationary equilibrium around S^* . Here we can identify which are the critical parameters that determine the regime of the economy.

Figure 4 shows the stationary densities for the parameters in Table 1. We find $\Pr[0] = .1$, $\Pr[S^*] = 0$. In fact $\lim_{S \rightarrow 0} f_C = +\infty$, but f_C is approximately proportional at 0 to $1/\sqrt{S}$, meaning that the probability of C remains finite (see equation 28). This high density around 0 comes from the fact that the rate of consumption of the stocks decreases steeply as S approaches zero. The high density on the left of S^* is explained by the fact that accumulation slows down as the stock approaches S^* (see equations (17,18) about drainage speed and time).

In contrast to 0, S^* is never attained, as we mentioned in Subsection 4.2. Nevertheless, as Proposition 4 shows, the probability mass can be quite concentrated, under precise circumstances, in the neighborhood of the maximum.

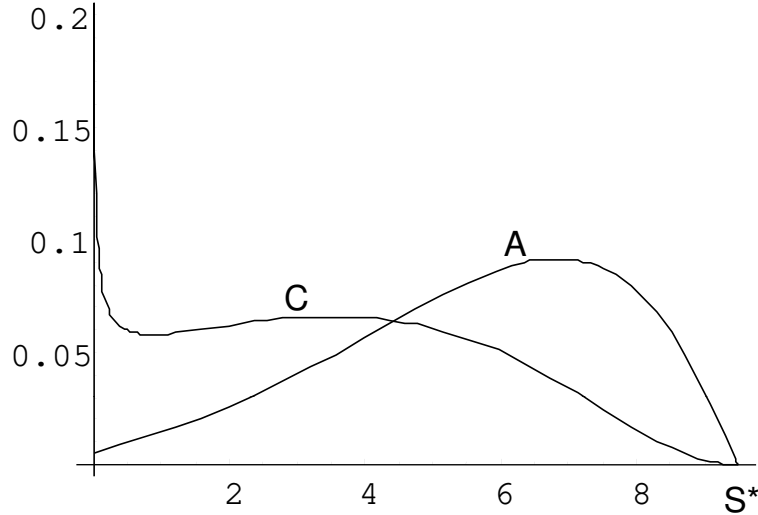


Figure 4: Densities.

Proposition 4 *Let*

$$K_{S^*} = \frac{2\lambda_C}{\sqrt{(r+\lambda_C)^2 + 4\Delta'_A[p_A^*]\lambda_C M - r - \lambda_C}} \quad (33)$$

$$\text{with } M = \frac{[(r+\lambda_A)(r+\lambda_C) - \lambda_A\lambda_C]p_A^* + (r+\lambda_A+\lambda_C)c}{-\lambda_C\Delta_C[\frac{r+\lambda_C}{\lambda_C}p_A^* + \frac{c}{\lambda_C}]} > 0. \quad (34)$$

At S^* , f_C is of the order of $(S^* - S)^{K_{S^*}}$ and f_A is of the order of $(S^* - S)^{K_{S^*}-1}$.

Consequently,

1. If $K_{S^*} < 1$: f_A increases and diverges as $S \rightarrow S^*$. Though the maximum is never attained, any neighborhood of S^* has a positive probability.
2. If $1 < K_{S^*} < 2$: f_A converges to 0 at S^* with vertical negative slope. The system is close to the maximum with a positive probability.
3. If $K_{S^*} > 2$: f_A converges to 0 at S^* with a null slope. The economy is almost surely far from the upper bound.

Given the discontinuous nature of the comparative statics, singular cases with either $K_{S^*} = 1$ or $K_{S^*} = 2$ would require higher order approximations than the one used in Appendix A.8 to be described.

The understanding of the conditions above is relatively complex since all the fundamental parameters play a role. In particular, no simple comparative statics with respect to r or λ_C emerge. In contrast, the effects of c , λ_A are obvious

$$\frac{\partial K_{S^*}}{\partial c} < 0; \quad \frac{\partial K_{S^*}}{\partial \lambda_A} < 0. \quad (35)$$

In the linear case, where in particular $\Delta'_A[p_A^*] = \beta_A$, we have

$$\frac{\partial K_{S^*}}{\partial \beta_C} > 0; \quad \frac{\partial K_{S^*}}{\partial \beta_A} < 0; \quad \frac{\partial K_{S^*}}{\partial p_C^*} < 0; \quad \frac{\partial K_{S^*}}{\partial p_A^*} < 0. \quad (36)$$

The comparative statics on K_{S^*} , together with the ones on S^* exposed in the comments of Proposition 2, outline a notion of flexibility: the higher K_{S^*} , the less flexible the economy is. Excess supply functions measure the response of prices to given variations in stocks. Small maximum stocks correspond to *flexible* economies for which large storage would be useless, and accordingly the economy has, statistically, enough time to approach this modest target during abundance period. On the contrary, large maximum stocks mean that the economy will seize (almost) any opportunity to accumulate, which happens in economies where building stocks is a costly process. Accordingly, it is very likely that the random alternation between abundance and scarcity episodes will keep the economy far from the bliss point.

6 Applications and extensions

In this Section, we extend the model by assessing the impact of three kinds of constraints: politically imposed bounds on prices, non-negligible injection and release costs and limited storage capacity.

Analyzing the impact of a constant price on the dynamic system allows a comparison of the results with those proposed by the abundant literature on stabilization. Following on from this, we show that non-negligible injection and release fees can be modelled as parallel shifts in the functions $p_A[S]$ and $p_C[S]$. The main results of our analysis are unaffected by

this generalization. Finally, assuming that storage capacity is exogenously constrained, we show that in the accumulation phase, the maximal stock is attained in finite time. Moreover, the price for storage service suddenly jumps above marginal cost when capacity saturates.

6.1 Stabilization, storage and persistent crises

Assume that a central authority imposes a constant price p^* . A price below p_A^* would not be sustainable in the long run (stock will be drained out shortly). A price above p_C^* would cause never ending accumulation, which would be uneconomical. So, the relevant policies consider $p_A^* < p^* < p_C^*$. Remark that if we preclude rationing, the policy is not strictly applicable since the price must turn to p_C^* when stocks are empty in state C . With rationing, the price may remain formally at p^* , but the marginal shadow value of the commodity would be p_C^* anyway.

To summarize the effect of the policy, the simplest approach is to search for stationary distribution. We can solve (31), i.e.

$$\frac{d\phi_A}{dS} = - \left(\frac{\lambda_A}{\Delta_C} + \frac{\lambda_C}{\Delta_A} \right) \phi_A \text{ for all } S > 0, \quad (37)$$

where $\frac{\lambda_A}{\Delta_C} + \frac{\lambda_C}{\Delta_A}$ here is a constant (with a constant price, $\Delta_C < 0$ and $\Delta_A > 0$ are constant). Define \bar{p}^* as the solution to the equation $\lambda_A \Delta_A [p] + \lambda_C \Delta_C [p] = 0$.

If $p^* < \bar{p}^*$, then $\frac{\lambda_A}{\Delta_C} + \frac{\lambda_C}{\Delta_A} > 0 \Leftrightarrow \lambda_A \Delta_A + \lambda_C \Delta_C < 0$: on average, the economy draws on the stocks. This implies that ϕ_A is decreasing and the density is a decreasing exponential: lower stocks are more likely. The distribution has an unbounded support, empty stocks in crisis is an event of positive probability during which the price is p_C^* .

If $p^* > \bar{p}^*$, then $\frac{\lambda_A}{\Delta_C} + \frac{\lambda_C}{\Delta_A} < 0 \Leftrightarrow \lambda_A \Delta_A + \lambda_C \Delta_C > 0$: on average, the economy piles up stocks. This implies that ϕ_A is increasing unboundedly with respect to S . Higher stocks being increasingly likely, normalization is impossible; in other words there is no well defined stationary distribution. Stocks diverge to infinity with probability one and stabilization, in this sense, succeeds.

The case $p^* = \bar{p}^*$ is intriguing. The economy has no tendency to pile up nor to drain out stocks. All positive levels of the stocks are equally likely (the stationary density is flat), meaning that the behavior of the system in the long run is unpredictable.

Stabilization should not be understood in the narrow sense of averaging the price that would be observed in the absence of storage capabilities. Remark indeed that \bar{p}^* , which is the critical threshold, could be higher or lower than the average no-storage price $\frac{\lambda_A}{\lambda_A + \lambda_C} p_A^* + \frac{\lambda_C}{\lambda_A + \lambda_C} p_C^*$. This depends on the sensitivity of excess supply functions to price variations.

The conclusion is straightforward: perfect price stabilization can be reached only if the economy is prepared to let stocks go to infinity. The analysis above is easily extended to the case of limited storage capacity. Any upper bound on stocks leaves positive probability on empty stocks. In that case, the probabilities of full storages and stockouts depends on the policy p^* chosen.

6.2 Injection and release costs

Denote unit injection cost by i and unit release cost by s . Assume that in each state $\sigma = A, C$, and for any stocks level S , there are markets for the gas *outside* and *inside* the reservoir, the prices being respectively $p_\sigma[S]$ and $p_\sigma^I[S]$. The (competitive) market equilibrium between outside and inside gases implies that, whenever $S > 0$,

$$p_A[S] + i = p_A^I[S] \text{ and } p_C[S] = p_C^I[S] + s. \quad (38)$$

The structure of the system of equations is preserved, with p_σ^I replacing p_σ . No arbitrage conditions read

$$\Delta_C[p_C^I + s] \cdot \frac{dp_C^I}{dS} = (r + \lambda_A)p_C^I - \lambda_A p_A^I + c, \quad (39)$$

$$\Delta_A[p_A^I - i] \cdot \frac{dp_A^I}{dS} = (r + \lambda_C)p_A^I - \lambda_C p_C^I + c. \quad (40)$$

Remark that the excess supply functions are shifted, thus boundary conditions are

$$p_C^I[0] = p_C^* - s, \quad (41)$$

$$p_A^I[S^*] = p_A^* + i. \quad (42)$$

The range of p_σ^I is narrower than that of p_σ : the minimum is higher, the maximum is lower. As a result, the condition ensuring that there *is* storage in equilibrium is more restrictive than the one in Proposition 1 (Point 4), i.e. in the linear case

$$p_C^* - s > \left(\frac{r + \lambda}{\lambda_C} \right) (p_A^* + i) + \frac{c}{\lambda_C}. \quad (43)$$

The phase diagram enables us to show that S^* is decreasing with respect to the cost parameters s and i . The rest of the comparative statics is identical.

6.3 Limited storage capacity

If the total storage capacity \bar{S} exceeds S^* , then the unconstrained trajectory remains sustainable; else, rational storers anticipate that boundary conditions are modified.

As long as some capacity is vacant, then storage price per unit of gas (per unit of time) remains equal to marginal cost c ; the system of equations is exactly the same as the one without any constraint, so the equilibrium is described by a trajectory in region **I** of the phase diagram. Trajectories below the unconstrained equilibrium start on the vertical axis at a given price for $S = 0$ and stop on the horizontal axis on the right of Ω for a maximum stock which is smaller than S^* . There is a unique trajectory such that this maximum stock equals exactly \bar{S} . It describes the unique equilibrium with limited storage capacity. See for example the dashed trajectory below the bold one in Figure 1.

In the accumulation phase, the price for storage service suddenly jumps above marginal cost when capacity is saturated. We denote it by π_A .¹⁴ Given that $p_A[\bar{S}] = p_A^*$, the no-arbitrage argument in state $(A, S = \bar{S})$ can be expressed

$$\lambda_C(p_C[\bar{S}] - p_A^*) = rp_A^* + \pi_A. \quad (44)$$

The LHS measures the potential profit from holding stocks and the RHS the cost. Given that $p_C[\bar{S}] > \frac{r+\lambda_C}{\lambda_C}p_A^* + \frac{c}{\lambda_C}$ (the terminal point is on the right of Ω in Figure 1), we have $\pi_A > c$.

In contrast to the unconstrained case, the maximal stock is attained in finite time if abundance lasts long enough. This explains that the jump in the price of storage services (a discontinuity) is consistent with a continuous price function $p_A[S]$ (continuity is necessary for no-arbitrage): before the capacity is full, the price $p_A[S]$ decrease steadily; storers incur non-negligible capital losses if the state does not change; this depreciation term does not converge to zero as the maximal stocks are reached; this term is relayed by cost $\pi_A > 0$ when the constraint becomes binding.

7 Conclusion

Our model has fully described the behavior of a Markov economy in which storage dynamics are determined by random occurrence of crises. Overall, we have proposed the quite appealing notion of “flexible economy”. We have proved that in equilibrium, a more flexible economy (i.e. better able to absorb shocks via production and consumption changes), is less keen to build up large stocks and is much more likely (in terms of probability) to hold maximum stocks. If the reluctance to build large stocks is intuitive, since overall, the value of stocks (or the convenience yield) decreases when an economy can promptly react to a shock, release dynamics are less intuitive. We show that flexible economies go fast towards maximum stocks and just stay there until a shock leads to fast drainage, while inflexible economies incur permanent movement of their stocks, and over a wider interval. This relationship between flexibility and maximum stocks is a result of interest. On this ground, it could be argued that security of supply policies for energy or banking rules for emission rights, which are never neutral with respect to the market equilibrium, should not be set equally across European states, inasmuch as their capabilities to respond to shocks is heterogeneous.

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¹⁴In case of crisis, the stock immediately starts being used so that state $(C, S = \bar{S})$ does not last. This implies that π_C , the price of storage services for congestion during the crisis, has no measurable economic effect.

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A Appendix

A.1 Phase diagram

Compared to (AA') , (CC') cuts the horizontal axis with a smaller value of p_C . It suffices to inject $p_A = p_A^*$ in both equations and to check that

$$\frac{\lambda_A}{r + \lambda_A} p_A^* - \frac{c}{r + \lambda_A} < \frac{r + \lambda_C}{\lambda_C} p_A^* + \frac{c}{\lambda_C}. \quad (45)$$

This is obvious.

To show that (CC') cuts the vertical axis with a larger value of p_A than (AA') , we inject $p_C = p_C^*$ into both equations and check that the following equation remains true (obvious)

$$\frac{r + \lambda_A}{\lambda_A} p_C^* + \frac{c}{\lambda_A} > \frac{\lambda_C}{r + \lambda_C} p_C^* - \frac{c}{r + \lambda_C}. \quad (46)$$

A.2 Proof of Proposition 1

1. and 2. Price functions are continuous and the stocks vary continuously with respect to time, thus the support of stocks is necessarily an interval.

Assume that over some interval J of stocks, $\Delta_C[p_C[S]] \geq 0$ and $\Delta_A[p_A[S]] \geq 0$ (accumulation only) or $\Delta_C[p_C[S]] \leq 0$ and $\Delta_A[p_A[S]] \leq 0$ (drainage only). Clearly, interval J can be traversed once at most in history. Consequently, interval J cannot be part of the equilibrium support of the stocks.

If over some interval, $\Delta_C[p_C[S]] \geq 0$ (i.e. $p_C[S] \geq p_C^*$) and $\Delta_A[p_A[S]] \leq 0$ (i.e. $p_A[S] \leq p_A^*$), accumulation during crises is motivated only by the fact that the price is expected to rise if the economy stays in crisis (an episode of abundance causes a negative shock on the price). This means that $dp_C^*[S]/dS > 0$. In consequence, the price in state C is unbounded as well as the support of S , since any length for an episode of scarcity has positive probability to be exceeded. Such a bubble is unreasonable since it would imply unbounded storage capacity (a transversality condition would translate this common sense remark into mathematical language). These arguments show that $\Delta_C[p_C[S]] \leq 0$ (i.e. $p_C[S] \leq p_C^*$) and $\Delta_A[p_A[S]] \geq 0$ (i.e. $p_A[S] \geq p_A^*$).

Remark now that if the lower bound of the support were strictly positive, this would mean that a certain quantity of the commodity would be permanently frozen. Denote this level by S_{\min} . Necessarily, $p_C[S_{\min}] = p_C^*$, meaning that if the economy stays in crisis for a long time (a possible event), storers have to pay storage costs without compensation in potential price movements. Consequently, support is bounded below by 0 ($S_{\min} = 0$).

3. If $p_C[0] < p_C^*$, final demand would exceed production for $S = 0$, which is impossible with empty storages. This proves that the trajectory starts on the vertical axis ($p_C[0] = p_C^*$).

Assume that there is an $\varepsilon > 0$ such that p_A stays larger than $p_A^* + \varepsilon$ as S increases. Given the phase diagram, p_A reaches its minimum on the straight line (AA') , if the trajectory starts in I. In any case, as S

increases, the trajectory enters in II, then in III, implying that p_A overshoots $\frac{r+\lambda_A}{\lambda_A}p_C^* + \frac{c}{\lambda_A}$ (intersection between (CC') and vertical axis). In other terms, for any large enough S , p_A overshoots p_C^* while p_C remains below p_C^* . Thus, as S continues to grow, x'_A/x'_C remains finite whether or not p_A tends towards infinity. We conclude that any trajectory in III reaches the vertical axis $p_C = p_C^*$ for a finite value of the stock. At this point, drainage stops in state C , but accumulation continues in state A ($x_A > 0$), meaning that the stock is never going to be used in the future, which is uneconomical (see point 1 in this proof).

We conclude from the contradiction that accumulation must stop in state A for some stocks S^* ; thus necessarily $p_A[S^*] = p_A^*$. Remark that this can only happen in I on the horizontal axis at point Ω .

4. From (AA') , we have

$$(r + \lambda_C)p_A[S^*] - \lambda_C p_C[S^*] + c = 0 \quad (47)$$

i.e.

$$p_C[S^*] = \frac{r + \lambda_C}{\lambda_C} p_A^* + \frac{c}{\lambda_C} \quad (48)$$

This corresponds to $\Omega = (\frac{r+\lambda_C}{\lambda_C}p_A^* + \frac{c}{\lambda_C}, p_A^*)$. For this intersection to exist, i.e. for I not to be empty, a sufficient and necessary condition is

$$(r + \lambda_C)p_A^* + c \leq \lambda_C p_C^*. \quad (49)$$

5. We show that there is a unique trajectory passing through Ω . The Cauchy-Lipschitz Theorem cannot be applied since the system is singular at Ω . We use the following argument: choose any starting point in the interior of I, denoted by (p_C^0, p_A^0) ; it is necessarily nonsingular. The trajectory passing through this point is unique (Cauchy-Lipschitz). Consider the point $(p_C^0, p_A^0 + \varepsilon)$ where ε is some small real. Straightforward calculations show that the slope of the trajectory passing through $(p_C^0, p_A^0 + \varepsilon)$, which is positive, decreases as ε increases. One can directly reason on

$$dp_A/dp_C = \frac{\Delta_C[p_C]}{\Delta_A[p_A]} \cdot \frac{(r + \lambda_C)p_A - \lambda_C p_C + c}{(r + \lambda_A)p_C - \lambda_A p_A + c}. \quad (50)$$

This means that trajectories move apart as S increases, i.e. as they approach Ω . The consequence is that there cannot be multiple trajectories through Ω . This proves existence and uniqueness.

A.3 Proof of Proposition 2

We first determine how trajectories move in the phase diagram as parameters change. Rewrite the system of ODE in compact form as

$$p'_C = P_C(p_C, p_A, c, r, \lambda_A, \lambda_C) \text{ or simply } P_C \text{ (} > 0 \text{ in region I),} \quad (51)$$

$$p'_A = P_A(p_C, p_A, c, r, \lambda_A, \lambda_C) \text{ or simply } P_A \text{ (} < 0 \text{ in region I).} \quad (52)$$

Note that $\frac{\partial P_C}{\partial c} = 1/\Delta_C[p_C] < 0$ and $\frac{\partial P_A}{\partial c} = 1/\Delta_A[p_A] > 0$, thus $p'_A/p'_C = P_A/P_C$ decreases as c increases (all trajectories in I are flatter). Similar observations prove that all trajectories in I are also flatter when r increases, when λ_A increases and when λ_C decreases.

We can now position equilibrium trajectories as parameters change. Increasing c or r , or decreasing λ_C , move Ω to the right; increasing λ_A has no effect on Ω . In all cases, the equilibrium trajectory moves below the former one: to each p_C is associated a smaller p_A .

Remark that $\frac{dS}{dp_C} = 1/P_C < 0$, thus

$$S = - \int_{p_C[S]}^{p_C^*} \frac{dp_C}{P_C} \quad (\text{summation along the equilibrium trajectory}). \quad (53)$$

Since Ω goes to the right as c increases, the range of p_C becomes smaller; it remains to be verified that $1/P_C$, as a function of p_C , is also smaller. For example, along the equilibrium trajectories, for a fixed p_C

$$\frac{dP_C}{dc} = \underbrace{\frac{1}{\Delta_C[p_C]}}_{-} + \underbrace{\frac{\partial p_A}{\partial c}}_{-} \times \underbrace{\frac{\partial P_C}{\partial p_A}}_{+} < 0. \quad (54)$$

(P_C grows in absolute value and thus $1/P_C$ decreases in absolute value.) This proves that as c increases, a given price is associated with a smaller S . Similar reasonings can be applied to the other parameters to prove the claims.

A.4 Proof of Proposition 3

We have

$$P_C = \frac{(r + \lambda_A)p_C - \lambda_A p_A + c}{\beta_C(p_C - p_C^*)}, \quad (55)$$

$$P_A = \frac{(r + \lambda_C)p_A - \lambda_C p_C + c}{\beta_A(p_A - p_A^*)}. \quad (56)$$

Clearly, trajectories in I are steeper with a higher β_C or a smaller β_A . Remark that the frontier of I (Ω in particular) is unchanged in this comparative statics. Remark also (this concerns point 2) that a proportional increase of β_C and β_A does not change the trajectories (but a given point corresponds to a different S). The type of reasoning used in the proof of Proposition 2 can now be applied to show the claims.

The comparative statics with respect to p_C^* and p_A^* require further precautions. In the former, remark that trajectories are steeper with a higher p_C^* ($p_C < p_C^*$) and that I is extended to the right (trajectories are simply going further to the right). These two effects concur to increase the price for given stocks. In the latter, trajectories are flatter with a (say) smaller p_A^* but Ω moves along down (AA'). The first effect decreases prices, hence point 1, but the second could lead to a higher S^* (a smaller function is integrated over a longer interval, since the range of p_C increases, see equation 53).

A.5 Equivalent expressions for prices

On the right of 0. We have

$$x_C[S] = \int_{p_C^*}^{p_C[S]} \Delta_C[p] dp, \quad (57)$$

thus, writing first-order approximation on both sides we get

$$K_C S + o(S) = \frac{1}{2} \Delta'_C[p_C^*] (p_C[S] - p_C^*)^2 + o(p_C[S] - p_C^*)^2, \quad (58)$$

which yields

$$p_C[S] - p_C^* \underset{0}{\sim} -\sqrt{\frac{2K_C}{\Delta'_C[p_C^*]} S^{1/2}}. \quad (59)$$

On the left of S^* . Given that, according to (6), $p'_C[S^*] \neq 0$, we can approximate $p_C[S]$ around S^* with $p_C[S^*] + p'_C[S^*](S - S^*) + o(S - S^*)$. We denote $p'_C[S^*]$ by $-M$, with

$$M = -\frac{(r + \lambda_A)(r + \lambda_C) - \lambda_A \lambda_C p_A^* + (r + \lambda_A + \lambda_C)c}{\lambda_C \Delta_C[\frac{r + \lambda_C}{\lambda_C} p_A^* + \frac{c}{\lambda_C}]} > 0. \quad (60)$$

Given that

$$x_A[S] = \int_{p_A[S]}^{p_A^*} \Delta_A[p] dp, \quad (61)$$

we can calculate that $p_A[S] - p_A^* + o(p_A[S] - p_A^*) = \sqrt{\frac{2}{\Delta'_A[p_A^*]} x_A^{1/2}[S]}$, or equivalently $p_A[S] - p_A^* = \sqrt{\frac{2}{\Delta'_A[p_A^*]} x_A^{1/2}[S]} + o(x_A^{1/2}[S])$.

We plug these two equivalent expressions into (7), which yields

$$x'_A = (r + \lambda_C) \sqrt{\frac{2}{\Delta'_A[p_A^*]}} x_A^{1/2} + \lambda_C M(S - S^*) + o(S - S^*) + o(x_A^{1/2}), \quad (62)$$

Consider now the ODE

$$y' = (r + \lambda_C) \sqrt{\frac{2}{\Delta'_A[p_A^*]}} y^{1/2} + \lambda_C M(S - S^*) \text{ with } y[S^*] = 0. \quad (63)$$

The unique solution to (63) is $K_A^2(S^* - S)^2$ with

$$K_A = \frac{\sqrt{(r + \lambda_C)^2 + 4\Delta'_A[p_A^*]\lambda_C M - r - \lambda_C}}{2\Delta'_A[p_A^*]}. \quad (64)$$

We show now that this exact solution of approximate ODE (63) is an approximation of the solution to ODE (62).

Consider the residual $o(S - S^*) + o(x_A^{1/2}[S])$ in the ODE (62). For all $\varepsilon > 0$, there is a left neighborhood of S^* , denoted V_ε , in which the absolute value of the residual is smaller than $\varepsilon \times (S^* - S)$ and $\varepsilon \times (x_A^{1/2}[S])$. Consider the ODE

$$y' = \left[(r + \lambda_C) \sqrt{\frac{2}{\Delta'_A[p_A^*]}} + \varepsilon \right] y^{1/2} + (\lambda_C M - \varepsilon)(S - S^*) \text{ with } y[S^*] = 0. \quad (65)$$

The solution to this equation is smaller than x_A on V_ε : indeed, both x'_A and y' are negative, but if $y > x_A$ for some S in V_ε , it remains so for any larger stock because $y' > x'_A$. This is in contradiction with the fact that $y[S^*] = x_A[S^*]$. In other terms,

$$x_A[S] \geq \left[\frac{\sqrt{(r + \lambda_C + \sqrt{\frac{\beta_A}{2}}\varepsilon)^2 + 4\beta_A(\lambda_C M - \varepsilon) - r - \lambda_C - \sqrt{\frac{\beta_A}{2}}\varepsilon}}{2\beta_A} \right]^2 (S^* - S)^2. \quad (66)$$

A similar reasoning shows that

$$x_A[S] \leq \left[\frac{\sqrt{(r + \lambda_C - \sqrt{\frac{\beta_A}{2}}\varepsilon)^2 + 4\beta_A(\lambda_C M + \varepsilon) - r - \lambda_C + \sqrt{\frac{\beta_A}{2}}\varepsilon}}{2\beta_A} \right]^2 (S^* - S)^2. \quad (67)$$

These two inequalities give the approximation of x_A at S^* .

A.6 Statistical evolution from a known state

To fix ideas, denote the observed state at, say, date $t = 0$, by conjuncture A and stocks S_0 . Conditionally on staying in state A , the probability mass (initially 1) is attached to a unique level of the stocks; but this mass itself is eroded by potential jumps. Other states (except 0 and S^*) are associated with a density, as we shall develop now.

The mass point $S_{\max}[t]$ (where the index \max recalls that the support of the distribution is bounded above by this time dependent value) evolves following

$$\frac{dS_{\max}}{dt} = \Delta_A[p_A[S_{\max}]] > 0 \text{ with } S_{\max}[0] = S_0, \quad (68)$$

whereas the lower bound of the distribution follows

$$\frac{dS_{\min}}{dt} = \Delta_C[p_C[S_{\min}]] < 0 \text{ until 0 is reached, with } S_{\max}[0] = S_0. \quad (69)$$

Both values move deterministically. Clearly,

$$\Pr[A, S_{\max}[t], t] = e^{-\lambda_C t}, \quad (70)$$

which is the probability that no crisis has ever happened since date 0.

States inside the interval $[S_{\min}[t], S_{\max}[t]]$ are associated with a positive density; outside the interval, density is null. States $(C, 0, t)$ and $(A, S_{\max}[t], t)$ are mass points (the former only after some time has elapsed, so that S_{\min} has reached 0).

To fully characterize the statistical evolution from a known state, we prove the following preliminary lemma that explains how a pointwise probability becomes a density.

Lemma 1

$$\lim_{S \rightarrow S_{\max}[t]} f_C[S, t] = -\frac{\lambda_C e^{-\lambda_C t}}{\Delta_C [p_C[S_{\max}[t]]] + \Delta_A [p_A[S_{\max}[t]]]}, \quad (71)$$

Proof. Consider states $\{(C, S) | S \in [S_{\max}[t] - dS, S_{\max}[t]]\}$ with dS small. To calculate the probability of this event at date t , we have to trace back histories that lead to these states.

Denote dt and dS_A the unique time interval and stock variation such that a jump from $(A, S_{\max}[t] - dS_A)$ to $(C, S_{\max}[t] - dS_A)$ at date $t - dt$ followed by drainage between date $t - dt$ and t has led to state $(C, S_{\max}[t] - dS)$.

Clearly, all the jumps happening from the current mass point between dates $t - dt$ and t (and only these), give states in $[S_{\max}[t] - dS, S_{\max}[t]]$ at date t .¹⁵ So a first-order approximation of the probability we seek is the probability that a jump from $\{(A, S) | S \in [S_{\max}[t] - dS_A, S_{\max}[t]]\}$ to C has taken place between date $t - dt$ and t . This probability is $\lambda_C e^{-\lambda_C t} dt$, thus

$$\Pr\{(C, S) | S \in [S_{\max}[t] - dS, S_{\max}[t]]\} = \frac{\lambda_C \exp[-\lambda_C t] dt}{dS}. \quad (72)$$

We have to calculate the relationship between dS_C , dS_A , and dt . From the stock dynamics we have

$$dt = \frac{dS - dS_A}{\Delta_C [p_C[S_{\max}[t]]]} = \frac{dS_A}{\Delta_A [p_A[S_{\max}[t]]]}. \quad (73)$$

The middle term is the time needed, given drainage speed, to pass from stocks $S_{\max}[t] - dS_A$ to stocks $S_{\max}[t] - dS$. The RHS is the time needed, given accumulation speed, to pass from $S_{\max}[t] - dS_A$ to $S_{\max}[t]$. After straightforward substitutions, we let dS go to 0 and prove the result. ■

We are now equipped with a complete set of equations to describe the dynamics. Indeed, (71) and (22,23) can be used to describe statistically the system over time. The exercise is numerically demanding.

A.7 Proof of convergence of Algorithm 2

Remark that the ODE commanding ϕ_A can be written

$$\frac{\phi'_A}{\phi_A} = -\left(\frac{\lambda_A}{\Delta_C} + \frac{\lambda_C}{\Delta_A}\right). \quad (74)$$

On the right of $S = 0$, $\Delta_C \rightarrow 0$ so the RHS of (74) is equivalent to $-\frac{\lambda_A}{\Delta_C}$, i.e., using (15), to $\frac{K_0}{\sqrt{S}}$ where K_0 is a nonnegative real

$$K_0 = \frac{\lambda_A}{\Delta'_C [p_C^*] \sqrt{K_C}}. \quad (75)$$

¹⁵Given the small size of the interval, we can neglect histories involving *two* jumps, so relevant histories can only consist of one jump from state A and some drainage.

Thus $\lim_{S \rightarrow 0} \phi_A$ is finite and strictly positive. Indeed, for all $\varepsilon > 0$, there exists η such that for all $S \leq \eta$,

$$(1 - \varepsilon) \frac{K_0}{\sqrt{S}} \leq \frac{\phi'_A}{\phi_A} \leq (1 + \varepsilon) \frac{K_0}{\sqrt{S}}. \quad (76)$$

Take S_1 and S_2 both smaller than η with $S_1 \leq S_2$ and integrate the inequality above between these two reals. We find

$$2(1 - \varepsilon)K_0(\sqrt{S_2} - \sqrt{S_1}) \leq \ln \frac{\phi_A[S_2]}{\phi_A[S_1]} \leq 2(1 + \varepsilon)K_0(\sqrt{S_2} - \sqrt{S_1}). \quad (77)$$

This proves that ϕ_A is bounded away from 0 (fix S_2 and let S_1 converge to 0). Given that ϕ_A is also monotonic (increasing) in a neighborhood of 0, the limit that we denote by $\phi_A[0]$ exists and is nonnegative.

So, at 0, f_A is finite and nonnegative whereas $f_C \sim_0 \frac{K_{f_C}}{\sqrt{S}}$ where K_{f_C} is some nonnegative real. This implies that, though the density f_C diverges at 0, its integral is well defined.

A.8 Proof of Proposition 4

On the left of S^* , $\Delta_A \rightarrow 0$ so the RHS of (74) is equivalent to $-\frac{\lambda_C}{\Delta_A}$, i.e. $\frac{K_{S^*}}{S^* - S^*}$ where K_{S^*} is a nonnegative real with

$$K_{S^*} = \frac{\lambda_C}{\beta_A K_A}. \quad (78)$$

For all $\varepsilon > 0$, there exists η such that for all $S \geq S^* - \eta$,

$$(1 - \varepsilon) \frac{K_{S^*}}{S^* - S} \leq -\frac{\phi'_A}{\phi_A} \leq (1 + \varepsilon) \frac{K_{S^*}}{S^* - S}. \quad (79)$$

Take S_1 and S_2 both larger than $S^* - \eta$ with $S_1 \leq S_2$ and integrate the inequality between these two real numbers. We find

$$-(1 - \varepsilon)K_{S^*} \ln \frac{S^* - S_2}{S^* - S_1} \leq -\ln \frac{\phi_A[S_2]}{\phi_A[S_1]} \leq -(1 + \varepsilon)K_{S^*} \ln \frac{S^* - S_2}{S^* - S_1}, \quad (80)$$

i.e.

$$\left[\frac{S^* - S_2}{S^* - S_1} \right]^{(1+\varepsilon)K_{S^*}} \leq \frac{\phi_A[S_2]}{\phi_A[S_1]} \leq \left[\frac{S^* - S_2}{S^* - S_1} \right]^{(1-\varepsilon)K_{S^*}}. \quad (81)$$

This implies that $\lim_{S \rightarrow S^*} \phi_A = 0$, from which we can conclude that $\Pr[S^*] = 0$.

We can now derive a tight condition on the shape of the density function f_A around the upper bound S^* . Indeed, given that $\phi_A = f_A \cdot \Delta_A$,

$$f_A[S] \text{ is proportional to } (S^* - S)^{K_{S^*} - 1}. \quad (82)$$

Equation (82) together with $\phi_C = -\phi_A$ proves Proposition 4.