# Price Dynamics and Repeated games 

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## CIRM

Innovative Research in Mathematical Finance
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- The price process should be a CMMV (Continuous Martingale of Maximal Variation)


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The Market as a 2 player game:
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- The message $m$ can be identified with $L(m)$. $\mu=$ law of $L(m)$.


## The game $\Gamma_{n}(\mu)$ :

- Stage 0:

Nature chooses $L \sim \mu$
$P 1$ is informed of $L$ not $P 2$.
P1 and P2 know $\mu$.

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$I, J=P 1$ 's and P2's action spaces.
$T: I \times J \rightarrow \mathbb{R}^{2}$.
If choices $=(i, j), T(i, j)=\left(A_{i j}, B_{i j}\right)$ where $A_{i j}$ and $B_{i j}$ are the numbers of $R$ and $N$ shares that $P 2$ gives to $P 1$.


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- At stage q: P1 and P2 chose simultaneously $\left(i_{q}, j_{q}\right)$.
$\left(i_{q}, j_{q}\right)$ is then publicly announced.
$y_{q}=\left(y_{q}^{R}, y_{q}^{N}\right)=$ P1's portfolios after $q$
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For player 1: $\sigma=\left(\sigma_{1}, \ldots, \sigma_{n}\right)$, where
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A cautious P1 will play his max-min strategy
=Equilibrium strategy in the 0-sum game where a risk neutral P2 aims to maximize the liquidation value of his final portfolio.

## Natural exchange mechanism

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A trading mechanism $\langle I, J, T\rangle$ is natural if

- Numéraire scale invariance
- Invariance with respect to the riskless part of the risky asset.
- Existence of the value
- Positive value of information.
- Continuity of the value


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- $\Rightarrow \forall \alpha>0, \forall X: V_{1}([\alpha \cdot X])=\alpha \cdot V_{1}([X])$
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- $\exists p \in[1,2[, \exists A$ s. th. $\forall$ v.a. $X, Y$ :
$\left|V_{1}([X])-V_{1}([Y])\right| \leq A\|X-Y\|_{L^{p}}$


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- It also appears if player 2 is risk averse (De Meyer- Fournier


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Black and Scholes model is a CMMV.

$$
\begin{aligned}
& \Pi_{t}=\Pi_{0} \exp \left(\sigma B_{t}-\frac{\sigma^{2}}{2} t\right) \\
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- In a market with risk aversion, the actualized price process $\Pi$ is not a martingale!
- There exists a unique equivalent probability measure $Q$ under which the actualized price process is a CMMV.
- Is this conjecture in accordance with real data?

Black and Scholes model is a CMMV.

$$
\begin{aligned}
& \Pi_{t}=\Pi_{0} \exp \left(\sigma B_{t}-\frac{\sigma^{2}}{2} t\right) \\
& d \Pi_{t}=\Pi_{t} \sigma d B_{t}
\end{aligned}
$$

- This conjecture is the basic assumption of the CMMV pricing model


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## Example: European Call on CAC40



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- Metaphore


## Sunset over a foggy sea...



## Sunset over a foggy sea... <br> ...with a drunk captain



## Theorem

Let $\Pi^{1}$ and $\Pi^{2}$ be two distinct CMMV (i.e. with distinct $f$ ),
Let $\epsilon>0$,
Let $\nu^{i}$ denote the probability measure induced by $\Pi^{i}$ on $\mathcal{C}[0, \epsilon]$
then $\nu^{1}$ and $\nu^{2}$ are mutually singular.

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- $\partial_{K} C_{T, 0}^{K}=E_{Z}\left[\mathbb{1}_{f_{T}(Z)>K}\right]$
$\rightarrow 1-\partial_{K} C_{T, 0}^{K}=F_{\mathcal{N}}\left(f_{T}^{i n v}(K)\right)$


## Thank you!

## Happy Birthday Yuri!

