Price Dynamics and Repeated games

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CIRM

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Price Dynamics and Repeated games - p. 1/16

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- The price process should be a CMMV (Continuous Martingale of Maximal Variation)

General idea of DM (2010)

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• The message m can be identified with L(m). μ =law of L(m).

• Stage 0: Nature chooses $L \sim \mu$ P1 is informed of L not P2. P1 and P2 know μ .

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• Using a general trading mechanism $\langle I, J, T \rangle$: I, J=P1's and P2's action spaces. $T: I \times J \rightarrow \mathbb{R}^2$. If choices= $(i, j), T(i, j) = (A_{ij}, B_{ij})$ where A_{ij} and B_{ij} are the numbers of R and N shares that P2 gives to P1.

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- At stage q: P1 and P2 chose simultaneously (i_q, j_q) . (i_q, j_q) is then publicly announced. $y_q = (y_q^R, y_q^N) =$ P1's portfolios after q $y_q = y_{q-1} + T(i_q, j_q)$ and $y_0 = (0, 0)$.

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 =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

A trading mechanism $\langle I, J, T
angle$ 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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 - $\forall \mu \in \Delta^2$, $\Gamma_n(\mu)$ has an equilibrium.
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 - $\exists p \in [1, 2[, \exists A \text{ s. th. } \forall \text{ v.a. } X, Y]$ $|V_1([X]) - V_1([Y])| \le A ||X - Y||_{L^p}$

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 - It also appears if player 2 is risk averse (De Meyer-Fournier Price Dynamics and Repeated games - p. 7/16

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

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$$C_{T,t}^{K} = g_{T,t,\Pi_{t}}(K)$$

Example: European Call on CAC40



The CMMV pricing model

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 \rightarrow We just have to know f_T .

Theoretically, just by observing Π_t during a small interval of time

•
$$f_t(B_t) = \Pi_t$$

$$= E_Q[\Pi_T | \mathcal{F}_t]$$

$$= E_Q[f_T(B_t + (B_T - B_t))|\mathcal{F}_t]$$

$$= E_Z[f_T(B_t + \sqrt{T - t} Z)]$$

ightarrow We just have to know f_T .

• Metaphore





Sunset over a foggy sea... ...with a drunk captain



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

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 $\rightarrow 1 - \partial_K C_{T,0}^K = F_\mathcal{N}(f_T^{inv}(K))$



Happy Birthday Yuri!