Diffusion control ranking games Part II

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A 2-player game

 \triangleright State of player 1:

$$
dX_t = \alpha(X_t, Y_t) dW_t^1, \quad X_0 = 0
$$

 \blacktriangleright State of player 2:

$$
dY_t = \beta(X_t, Y_t) dW_t^2, \quad Y_0 = 0
$$

 $\blacktriangleright \ \alpha, \beta : \mathbb{R}^2 \to [\sigma_1, \sigma_2]$ closed loop controls \blacktriangleright W^1 and W^2 are independent BM

2-player game cont'd

reward of player
$$
1 = \begin{cases} 1, & \text{if } X_T > Y_T, \\ 0, & \text{else.} \end{cases}
$$

reward of player
$$
2 = \begin{cases} 1, & \text{if } Y_T > X_T, \\ 0, & \text{else.} \end{cases}
$$

Comments:

- \blacktriangleright Zero-sum payoff
- \blacktriangleright Players can observe the opponent's state

Solving the game

The state difference $D_t := X_t - Y_t$ satisfies

$$
dD_t = (\alpha_t^2 + \beta_t^2)^{1/2} d\tilde{W}_t
$$

Target of player 1: $P(D_T > 0) \longrightarrow \text{max}!$ Target of player 2: $P(D_T > 0) \longrightarrow min!$

Irrespective of β_t :

- \bullet $\alpha_t = \sigma_2$ maximizes the diffusion rate
- \bullet $\alpha_t = \sigma_1$ minimizes the diffusion rate

Solving the game

Theorem Let

$$
\alpha^*(x,y) = \begin{cases} \sigma_1, & \text{if } x \geq y, \\ \sigma_2, & \text{if } x < y, \end{cases}
$$

and

$$
\beta^*(x,y)=\alpha^*(y,x).
$$

Then (α^*, β^*) is a saddle point (and hence a Nash equilibrium).

Rigoros proof

Isaacs equations....

What if more than 2 players?

- \triangleright management: bonus if the own company is among the best performing companies
- research competition among many $R \& D$ teams
- \triangleright Olympic games: the best three athletes receive a medal
- \blacktriangleright elections with many candidates

A large ranking game

\n- *n* players
\n- $$
X^1, \ldots, X^n
$$
 = states of player $1, \ldots, n$
\n- $dX_t^{i,a} = a_i(X_t^{i,a}, X_t^{-i,a})dW_t^i, \quad X_0^{i,a} = 0.$
\n- $a_i : \mathbb{R}^n \to [\sigma_1, \sigma_2]$
\n- W^1, \ldots, W^n independent Brownian motions
\n

A large ranking game

 \blacktriangleright empirical distribution of all states at T:

$$
\mu^{n,a} = \frac{1}{n} \sum_{i=1}^n \delta_{X^{i,a}_T}
$$

► the $\alpha*100\%$ best players receive a prize:

reward of player
$$
i = \begin{cases} 1, & \text{if } X_T^{i,a} > (1 - \alpha)\text{-quantile of } \mu^{n,a}, \\ 0, & \text{else.} \end{cases}
$$

How to find / characterize Nash equilibria?

Stochastic games with many players in general are difficult to analyze. Additional difficulty here: discontinuous reward function.

Idea: for large n the mean-field game version yields an approximate Nash equilibrium.

The mean field approximation

 \blacktriangleright 1 player

 \blacktriangleright state dynamics

$$
dX_t = a(X_t)dW_t, \quad X_0 = 0
$$

\n- a : ℝ → [σ₁, σ₂]
\n- a^{*} is an equilibrium strategy if
\n- $$
P(X_7^{2^*} > a(X_{7}^{2^*}, 1 - \alpha)) = \max P(X_7^{2^*} > a(X_{7}^{2^*}, 1 - \alpha))
$$
\n
\n

$$
P(X_T^{a^*} > q(X_T^{a^*}, 1-\alpha)) = \max_{a} P(X_T^{a} > q(X_T^{a^*}, 1-\alpha))
$$

The mean field equilibrium is a threshold strategy

Theorem

The threshold strategy with threshold

$$
b^* = \begin{cases}\n-\sigma_2 \sqrt{T} \Phi^{-1} \left(\frac{\alpha(\sigma_1 + \sigma_2)}{2\sigma_2} \right), & \text{if } \alpha \leq \frac{\sigma_2}{\sigma_1 + \sigma_2}, \\
\sigma_1 \sqrt{T} \Phi^{-1} \left(\frac{(1-\alpha)(\sigma_1 + \sigma_2)}{2\sigma_1} \right), & \text{if } \alpha > \frac{\sigma_2}{\sigma_1 + \sigma_2},\n\end{cases}
$$
\n(1)

is an equilibrium strategy for the mean field game.

The mean field equilibrium is an approximate equilibrium of the n player game

Theorem

Let $a = (a_1, \ldots, a_n)$ be the tuple of mean-field equilibrium strategies. Then there exists $\varepsilon_n \downarrow 0$ such that $a = (a_1, \ldots, a_n)$ is an ε_n -Nash equilibrium of the n-player game. We can choose $\varepsilon_n \in \mathcal{O}(n^{-\frac{1}{2}}).$

The smaller the cake...

The smaller the cake...

Figure: $\gamma := \frac{\sigma_1}{\sigma_2} = \frac{2}{3}$, e.g. $\sigma_1 = 2, \sigma_2 = 3$

$$
R(\alpha) = \frac{1}{T} \int_0^T P(X_t^{b^*} < b^*) \, dt
$$

The larger the scope of action...

The heat of the battle

Part of the players choosing the small volatility σ_1

Figure: Parameters: $\sigma_1 = 1$, $\sigma_2 = 2$, $T = 1$, $\alpha = 0.5$

Comparison of large games with the 2-player game

Conclusion

- \triangleright Closed form equilibrium for the limiting cases $n = 2$ and $n = \infty$.
- ► Games with $n > 3$ players: equilibrium strategies depend on both the relative and absolute position

Literature

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