

Mean Field Game Analysis of Tournaments

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Outline

- 1 Introduction
- 2 The Poissonian model (with M. Nutz)
- 3 The diffusion model (with E. Bayraktar and J. Cvitanić)
- 4 Conclusion

Motivation

- Rank-based criteria is everywhere:
 - sports, R&D, admission, ... any “competition”
 - people’s utility often depends on relative performance
- May have a mixture of absolute and relative performance criteria.
- Goal: analyze large population competition of “timing”
 - stochastic control rather than stopping problems
 - want a model with tractability
 - describe equilibrium behavior
 - tournament design
- MFG where agents interact through the ranking of the hitting times

Mean Field Games

- Introduced by Lasry–Lions and Huang–Malhamé–Caines, 2006
- Nash equilibria for $N \rightarrow \infty$ players
- Interaction through empirical distribution ν of the private states
- Typical setting: each player controls a diffusion with some reward and cost-of-effort which depend on ν
- **Coupled system**: HJB and Kolmogorov PDEs, or FBSDEs
- Only **Linear-Quadratic** control can be solved explicitly
- Cardaliaguet, Carmona, Delarue, Fouque, Lacker, Yam, ...
- **Mean field games of timing**: Carmona–Delarue–Lacker, Nutz
- Toy example of Lasry–Lions, “**When Does the Meeting Start?**”
- **Principal-agent problem**: Élie, Mastrolia & Possaimai

General strategy for finding a Nash equilibrium

① **Best-response step:**

Solve a standard stochastic control problem for a representative player, given the strategy or performance of all other players.

② **Fixed-point step:**

Find a fixed point of the best-response mapping.

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The Poissonian model

- A **continuum** of agents $i \in (I, \mathcal{I}, \mu)$; atomless
- Each agent controls intensity λ of her **independent** Poisson process
- Quadratic instantaneous cost $c\lambda_t^2 dt$
- Goal is reached if process jumps (once)

Agents...

- are **ranked** according to their **completion times**
- are paid a **rank-based reward** $R : [0, 1] \rightarrow \mathbb{R}_+$
(decreasing, piecewise Lipschitz and left-continuous at $r = 1$)
- maximize expected reward minus cost

Interaction:

- Agents observe $\rho(t) = \mu\{i : \tau^i \leq t\} =$ **proportion** of agents completed by time t , and choose **feedback control** $\lambda(\rho(t))$
- Reward $R(\rho(\tau_i))$ and cost $c(\rho(t))$ depend on rank/proportion

Solving the MFG

- The representative player's problem:

$$v(r) = \sup_{\lambda \in \Lambda} E \left[R(\rho(\tau_\lambda)) - \int_0^{\tau_\lambda} c(\rho(t)) \lambda(\rho(t))^2 dt \middle| \rho(0) = r \right]$$

where Λ = set of feedback controls (piecewise Lipschitz),

$$\rho(t) = \int_0^t \bar{\lambda}(\rho(s))(1 - \rho(s)) ds$$

- The fixed-point problem:

$$\bar{\lambda} \mapsto \rho \mapsto \text{optimal } \lambda$$

- Agent space (I, \mathcal{I}, μ) is **atomless**
- Hence this is **Nash**: no single agent influences ρ

Existence and uniqueness

Theorem

There exists a unique (a.e.) equilibrium optimal control $\lambda^ \in \Lambda$,*

$$\lambda^*(r) = \frac{R(r) - \frac{1}{2\sqrt{1-r}} \int_r^1 \frac{R(y)}{\sqrt{1-y}} dy}{2c(r)}, \quad r \in [0, 1)$$

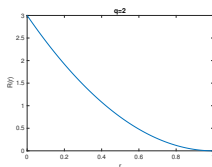
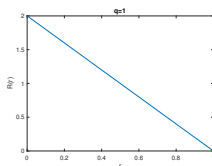
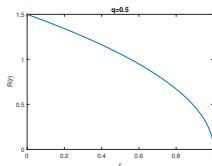
In equilibrium, the value function of any agent before completion is

$$v(r) = \frac{1}{2\sqrt{1-r}} \int_r^1 \frac{R(y)}{\sqrt{1-y}} dy, \quad r \in [0, 1)$$

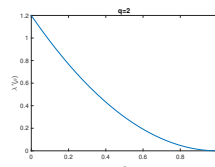
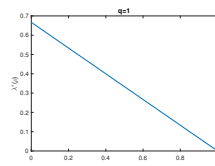
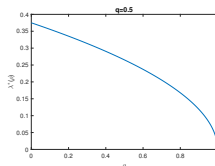
- Value of a freshly started game: $V = v(0) = \frac{1}{2} \int_0^1 \frac{R(y)}{\sqrt{1-y}} dy$.
- V is independent of c : higher cost \Rightarrow smaller optimal effort \Rightarrow state ρ is slowed down \Rightarrow same reward

Example: no cut-off

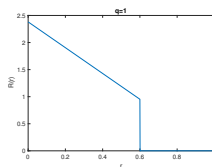
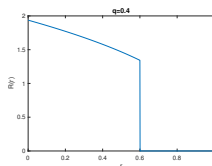
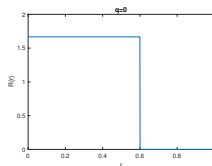
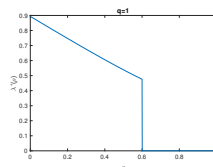
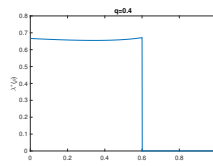
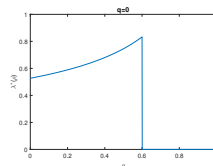
Reward R



Equilibrium effort λ^*



Example: cut-off at $\alpha = 60\%$

Reward R Equilibrium effort λ^* 

The principal's problem

- Given R , there exists a unique, deterministic equilibrium state ρ
- Time until α -fraction of the population has completed:

$$T_\alpha(R) = \inf\{t \geq 0 : \rho(t) \geq \alpha\} \in (0, \infty]$$

- Given $\alpha \in (0, 1)$ and budget $B > 0$,

$$\text{minimize } T_\alpha(R) \quad \text{subject to} \quad \int_0^1 R(r) dr \leq B$$

- What reward scheme $R^* \geq 0$ can attain $T_\alpha^* = \inf_R T_\alpha(R)$?
- A constrained calculus of variation problem

Theorem

Suppose $\frac{c(r)(1-r)}{2-r}$ is decreasing. Given a fixed reward budget $B > 0$ and $\alpha \in (0, 1)$, the optimal non-negative rank-based reward scheme to minimize $T_\alpha(R)$ is

$$R^*(r) = \frac{B}{C} \left\{ \sqrt{\frac{c(r)}{2-r}} + \frac{1}{2} \int_r^\alpha \frac{1}{1-s} \sqrt{\frac{c(s)}{2-s}} ds \right\} \mathbf{1}_{[0,\alpha]}(r),$$

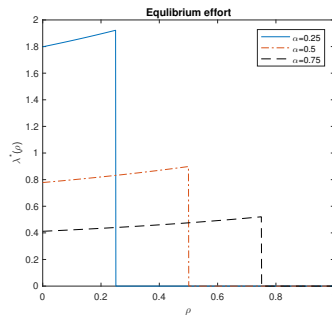
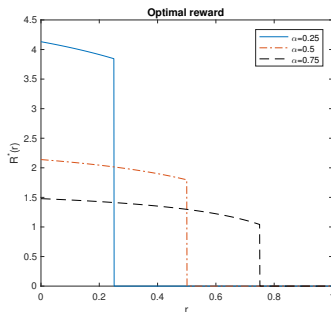
and the minimum α -completion time is

$$T_\alpha^* = \frac{4C^2}{B}, \quad \text{where} \quad C := \frac{1}{2} \int_0^\alpha \frac{\sqrt{c(r)(2-r)}}{1-r} dr.$$

The corresponding equilibrium effort is

$$\lambda^*(r) = \frac{B}{2C} \frac{1}{\sqrt{(2-r)c(r)}} \mathbf{1}_{[0,\alpha]}(r)$$

Plots for $\alpha = 25\%, 50\%, 75\%$, with c constant



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The diffusion model

- Reward function: $R(t, r) = 1_{\{t \leq T\}} H(r) + 1_{\{t > T\}} 0$.
- Assume agents are **homogeneous**: (x_0, c, σ) .
- Optimization problem for a representative player:

$$\sup_a E \left[R_\mu(\tau_\mu) - \int_0^{\tau_\mu \wedge T} c a_t^2 dt \right]$$

where $R_\mu(t) = R(t, F_\mu(t)) = 1_{\{t \leq T\}} H(F_\mu(t))$,

$$dX_t = -a_t dt + \sigma dB_t, \quad X_0 = x_0,$$

$$\tau_\mu = \inf\{t \geq 0 : X_t = 0\}.$$

- Nash equilibrium: fixed points of $\mu \mapsto \mathcal{L}(\tau_\mu^*)$.

Finding the best-response

- Given μ , the best-response can be computed by solving the HJB equation for the value function:

$$v_t + \sup_{a \geq 0} \left\{ -av_x + \frac{1}{2}\sigma^2 v_{xx} - ca^2 \right\} = 0,$$

$$v(t, 0) = H(F_\mu(t)), \quad v(T, x) = 0.$$

- The nonlinear HJB equation can be linearized using the **Cole-Hopf transformation** $u = \exp(\frac{v}{2c\sigma^2})$. u has stochastic representation:

$$u(t, x) = E \left[\exp \left(\frac{R_\mu(t + \tau_{x/\sigma}^\circ)}{2c\sigma^2} \right) \right],$$

where τ_m° is the BM first passage time to level m .

- The optimal feedback control $a^* = -\sigma^2 u_x / u$.

Explicit Nash equilibrium

Proposition. The fixed point equation (for μ) is :

$$f_{\mu}(t) = \frac{u(t, 0; \mu)}{u(0, x_0; \mu)} f_{\tau_{x_0/\sigma}^{\circ}}(t).$$

Theorem ($T = \infty$)

There is a *unique* equilibrium completion time distribution μ given in terms of its quantile function $r \mapsto T_r^{\mu}$ by

$$T_r^{\mu} = F_{\tau_{x_0/\sigma}^{\circ}}^{-1} \left(\frac{\int_0^r \exp\left(-\frac{H(z)}{2c\sigma^2}\right) dz}{\int_0^1 \exp\left(-\frac{H(z)}{2c\sigma^2}\right) dz} \right).$$

Moreover, the value of the game is given by

$$V_{\infty} = -2c\sigma^2 \ln \left(\int_0^1 \exp\left(-\frac{H(z)}{2c\sigma^2}\right) dz \right).$$

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Theorem ($T < \infty$)

There is a **unique** equilibrium completion time distribution μ given in terms of its quantile function $r \mapsto T_r^\mu$ by

$$T_r^\mu = F_{\tau_{x_0/\sigma}^\circ}^{-1} \left(\frac{1 - F_{\tau_{x_0/\sigma}^\circ}(T)}{1 - F_\mu(T)} \int_0^r \exp \left(\frac{-H(z)}{2c\sigma^2} \right) dz \right), \quad r \in [0, F_\mu(T)],$$

where the equilibrium terminal completion rate $F_\mu(T) \in (0, 1)$ is the unique solution of

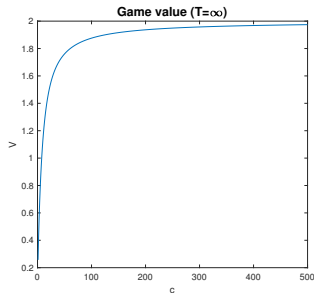
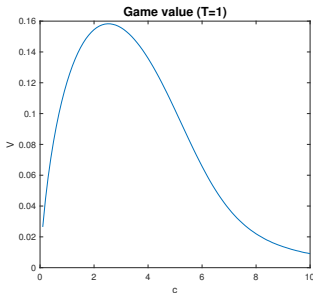
$$F_{\tau_{x_0/\sigma}^\circ}(T) = \frac{1 - F_{\tau_{x_0/\sigma}^\circ}(T)}{1 - F_\mu(T)} \int_0^{F_\mu(T)} \exp \left(\frac{-H(z)}{2c\sigma^2} \right) dz.$$

Moreover, the value of the game is given by

$$V = 2c\sigma^2 \ln \left(\frac{1 - F_{\tau_{x_0/\sigma}^\circ}(T)}{1 - F_\mu(T)} \right).$$

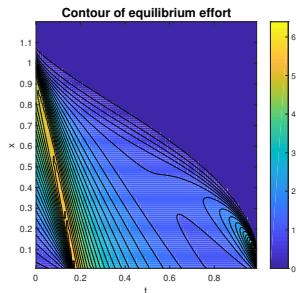
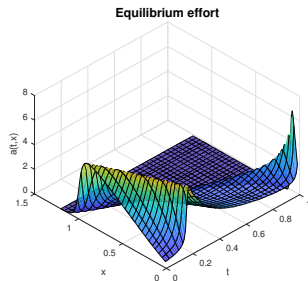
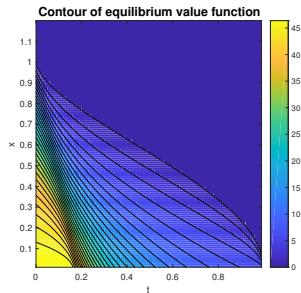
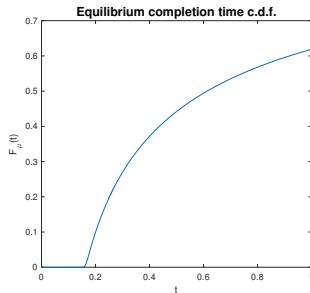
Dependence on the cost parameter

- (One-stage) Poissonian: game value is independent of c .
- Diffusion: game value is increasing in c when $T = \infty$, and non-monotone in c when $T < \infty$.

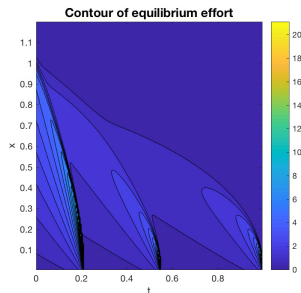
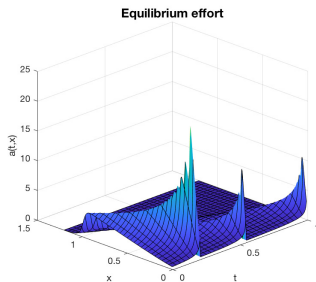
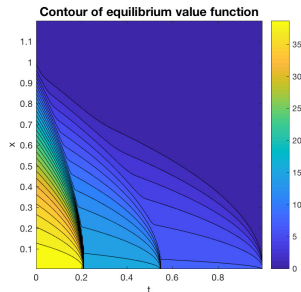
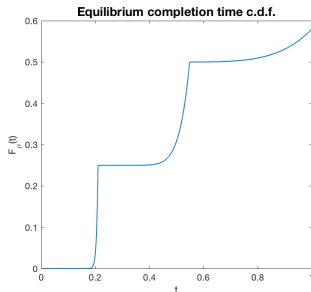


$$x_0 = 1, \sigma = 0.25, H(r) = 6(1 - r)^2.$$

- Efficiency does not necessarily make people happier.



$$x_0 = c = 1, \sigma = 0.25, H(r) = 6(1-r)^2.$$



$$x_0 = c = 1, \sigma = 0.25, H = 5 \cdot 1_{[0,0.25)} + 2 \cdot 1_{[0.25,0.5)} + 1_{[0.5,1]}.$$

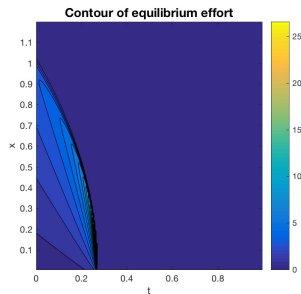
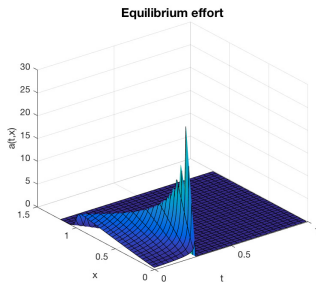
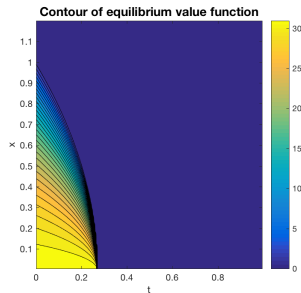
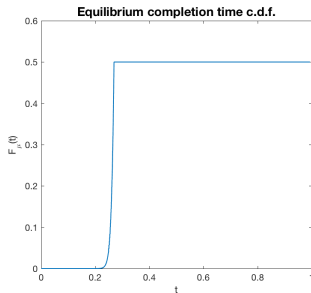
Minimizing the α -quantile

Objective: Given limited reward budget K , what (non-negative) reward scheme minimizes the time it takes $\alpha \in (0, 1)$ fraction of the population to complete their projects in equilibrium?

Theorem: regardless of the tournament horizon, the unique (up to a.e. equivalence) optimal reward scheme is the uniform scheme with cutoff rank α :

$$H^*(r) = \frac{K}{\alpha} 1_{[0, \alpha]}(r).$$

- cf. Poissonian competition.



$$x_0 = c = 1, \sigma = 0.25, K = 2, \alpha = 0.5, T = 1.$$

Maximizing profit

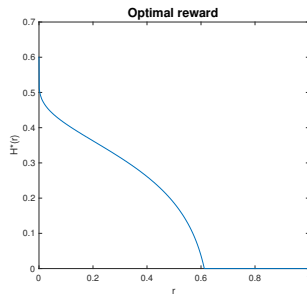
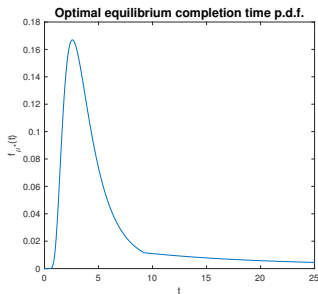
Objective: Suppose each project completed at time t generates a profit $g(t)$ for the principal. What reward scheme maximizes the aggregate net profit $E[g(\tau) - R_\mu(\tau)]$, $\tau \sim \mu$ for the principal?

Theorem: Suppose $T = \infty$. Given a non-constant decreasing profit function $g \in C_b(\mathbb{R}_+)$, an optimal reward scheme $H^* \geq 0$ is given by

$$H^*(r) = g(F_{\mu^*}^{-1}(r) \wedge t_b^*) - g(t_b^*),$$

where $t_b^* = \operatorname{argmax}$ of a one-dimension static optimization problem, and f_{μ^*} has an explicit formula involving t_b^* .

- In equilibrium, the agents receive $g(t \wedge t_b^*) - g(t_b^*)$ for finishing at time t . It is optimal for the principal to align his interest with that of the agents.
- Optimizing over the set of feasible μ .



$g(t) = x_0 e^{-0.1t}$, $x_0 = c = 1$, $\sigma = 0.25$. The participation reward constraint starts to be binding at rank 0.61.

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Conclusion

- Two tractable mean field models of competition
- Unique equilibrium in semi-closed form
 - Poissonian: game value is **independent** of c
 - Diffusion: game value is **increasing** in c (when $T = \infty$)
- Principal-agent problem explicitly solvable via calculus of variation
 - Poissonian: quantile-minimizing scheme is **not uniform**
 - Diffusion: quantile-minimizing scheme is **uniform**
- **Open questions:** multi-stage Poissonian model, heterogeneous agents, time-consistent formulation of the principal's problem, un-fixed pie, multiple tournaments, ...

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Thank you for your attention!