

# Stochastic correlated equilibrium with terminal commitment

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- 4 Extension to game and MFG

# Introduction to Correlated Equilibrium

**Correlated Equilibrium** is a generalization of Nash Equilibrium in Game Theory, introduced by Robert Aumann.

- In a Nash Equilibrium, players make independent decisions based on their own payoffs.
- In a Correlated Equilibrium, players follow a recommendation from a trusted source (a mediator), which correlates their choices.

# Mechanism of Correlated Equilibrium

- A mediator sends private signals to players.
- Each player receives a recommendation on what strategy to choose.
- Players follow the recommendation if they have no incentive to deviate given the signals.

Formally, a strategy profile is in Correlated Equilibrium if:

$$\text{for all players } i, \quad \mathbb{E} [u_i(s_i, s_{-i})] \geq \mathbb{E} [u_i(\phi_i(s_i), s_{-i})], \quad \forall \phi_i$$

where  $s_i$  is the recommended action and  $\phi_i$  is a function of deviation.

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where  $s_i$  is the recommended action and  $\phi_i$  is a function of deviation.

All Nash equilibria are Correlated equilibria. In correlated equilibria, the players are **only responding to the private signals**.

# Example of Correlated Equilibrium: Traffic Lights

## Example: Traffic Light Coordination

- Two drivers are approaching an intersection from different directions.
- If both proceed, they may crash. If one waits while the other goes, both will avoid a collision.
- The traffic light acts as a mediator, signaling "Green" to one driver and "Red" to the other.
- If each driver follows the light's recommendation, they both avoid a collision and minimize waiting time.

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- The traffic light acts as a mediator, signaling "Green" to one driver and "Red" to the other.
- If each driver follows the light's recommendation, they both avoid a collision and minimize waiting time.

Each driver responds solely to the traffic lights, **without** considering the actions or position of the other vehicle.

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# Classical Calculus of Variations: A Quick Recap

In classical calculus of variations, we seek to find the deterministic path  $x(t)$  that minimizes the action functional:

$$S[x] = \int_{t_1}^{t_2} L(x, \dot{x}, t) dt$$

where  $L(x, \dot{x}, t)$  is the Lagrangian. The path that minimizes the action satisfies the [Euler-Lagrange equation](#):

$$\frac{\partial L}{\partial x} - \frac{d}{dt} \left( \frac{\partial L}{\partial \dot{x}} \right) = 0$$

# Yasue's approach = Stochastic correlated equilibrium

**Yasue's Calculus of Variations** is an extension of classical calculus of variations to diffusion processes. The objective is to find a **signal process** in the form of diffusion:

$$dX_t = b(X_t, t)dt + \sigma dW_t,$$

from which the player does **NOT** want to **deviate** in order to minimize his action:

$$X = \arg \min_{Z_t = \phi_t(X_t), Z_T = X_T} \mathbb{E} \left[ \int_0^T L(Z_t, D_t^+ Z_t, D_t^- Z_t, t) dt \right],$$

where

$$D_t^+ Z_t := \lim_{h \rightarrow 0} \frac{1}{h} \mathbb{E}[Z_{t+h} - Z_t | X_t], \quad D_t^- Z_t := \lim_{h \rightarrow 0} \frac{1}{h} \mathbb{E}[Z_{t-h} - Z_t | X_t].$$

# Stochastic Euler-Lagrange equation

The key of Yasue's calculus of variations lies on the following [integral by parts formula](#):

$$\frac{d}{dt} \mathbb{E}[X_t Y_t] = \mathbb{E} \left[ Y_t D_t^+ X_t + X_t D_t^- Y_t \right].$$

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Then, by computing the variations of the action:

$$\mathbb{E} \left[ \int_0^T L(Z_t, D_t^+ Z_t, D_t^- Z_t, t) dt \right],$$

we simply obtain the [stochastic Euler Lagrange equation](#):

$$(\partial_1 L - D_t^- \partial_2 L - D_t^+ \partial_3 L)(X_t, D_t^+ X_t, D_t^- X_t, t) = 0$$

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# Derivation of PDEs: example of Schrödinger equation

For the action  $L(X, D^+X, D^-X) = V(X) + \frac{1}{4}|D^+X|^2 + \frac{1}{4}|D^-X|^2$ , the stochastic Euler Lagrange equation states:

$$\nabla V = \frac{1}{2}D_t^- D_t^+ X_t + \frac{1}{2}D_t^+ D_t^- X_t, \quad \text{Nelson's Newton's law.}$$

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To find the PDEs governing the correlated equilibrium, we need to note

$$D_t^+ f(t, X_t) = \mathcal{L}f(t, X_t) = (\partial_t f + \frac{1}{2}\sigma^2 \Delta^2 f + b \cdot \nabla f)(t, X_t),$$

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Assuming  $\exists R, S$  s.t.  $R = \frac{1}{2} \ln m$ ,  $\nabla S = b$ , then  $\Psi := e^{R+iS}$  satisfies the Schrödinger equation:

$$-i\partial_t \Psi = \frac{1}{2}\Delta \Psi - V\Psi, \quad \text{Nelson's stochastic mechanics (1965).}$$



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# Without backward kinetic energy

Consider the action  $L(X, D^+X, D^-X) = V(X) + \frac{1}{2}|D^+X|^2$  and  $\sigma = 1$ .  
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Denote by  $\eta := \log m$ . We find a system of forward PDEs.

# Wellposedness and verification

Borrowing the elements from Hao Xing and Gordan Zitković (AOP '18) and Joe Jackson (ECP '23), we can prove that the system of PDEs admits a **unique classical solution** on a **torus**, provided that  $b_0$ ,  $\eta_0$  are both **bounded** and  $\nabla V$  is Lipschitz.



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Given a classical solution of the system on  $\mathbb{R}^d$ , by further assuming that  $V$  is **convex**, we can verify that as the signal process

$$dX_t = b(t, X_t)dt + dW_t$$

is a stochastic correlated equilibrium.

# A possible long-time behavior on torus

Recall the system of parabolic equations:

$$\begin{aligned}
 -\partial_t \eta + \frac{1}{2} \Delta \eta - b \cdot \nabla \eta + \frac{1}{2} |\nabla \eta|^2 - \nabla \cdot b &= 0, & \eta(0, \cdot) &= \eta_0 \\
 -\partial_t b + \frac{1}{2} \Delta b + \nabla \eta \cdot \nabla b - b \nabla b + \nabla V &= 0.
 \end{aligned}$$

We realize that there is a stationary solution such that  $\nabla \eta^* = 2b^*$  and

$$\frac{1}{2} \Delta \eta^* + \frac{1}{4} |\nabla \eta^*|^2 + 2V = \lambda.$$

It is possible that  $(b_t, \eta_t) \rightarrow (b^*, \eta^*)$  as  $t \rightarrow \infty$ .

No stationary state on  $\mathbb{R}^d$ 

Suppose there is a stationary pair  $(m = e^\eta, b)$ , and necessarily we have  $\nabla \cdot ((b - \frac{1}{2} \nabla \log m)m) = 0$ . Using the equations, we can obtain

$$\begin{aligned}
 d \int \frac{1}{2} |b_t|^2 m_t &= \int (b \partial_t b + b^2 \nabla b - \frac{1}{2} b \nabla b \nabla \ln m) m \\
 &= \int (b \nabla V - \frac{1}{2} |\nabla b|^2) m \\
 &= - \int V \nabla \cdot (bm) - \int \frac{1}{2} |\nabla b|^2 m \\
 &= -\frac{1}{2} \int (\Delta V + |\nabla b|^2) m < 0.
 \end{aligned}$$

# 1-d Linear-quadratic case

Assume  $V(x) = D_0x^2$ , and initial conditions of  $\eta$  and  $b$  are quadratic and linear. By considering the solution in the form:

$$\begin{aligned}\eta(t, x) &= -A(t)x^2 + a(t)x + c(t), \\ b(t, x) &= B(t)x + \alpha(t).\end{aligned}$$

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Then we may obtain the system of Riccati equations:

$$\begin{aligned}\dot{A}(t) &= -2A(t)B(t) - 2A(t)^2, \\ \dot{B}(t) &= -2B(t)A(t) - B(t)^2 + 2D_0.\end{aligned}$$

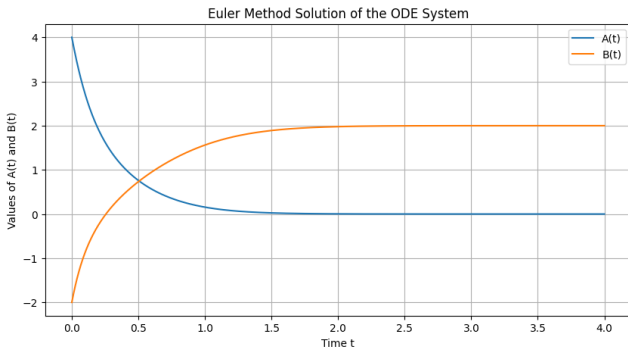
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## Some remarks/mysteries...

In this definition of stochastic correlated equilibrium, we need to highlight:

- $\eta_0 = \log m_0$ : The initial distribution  $m_0$  cannot be a Dirac measure, that is, the initial signal is inherently random. This randomness could arise from uncertainty in measuring the player's initial state;

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- The signal process is of the form of diffusion. This is either a choice or a constraint of the social planner (to hedge a risk for example);



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- The terminal commitment  $Z_T = X_T$  could arise from the verification of the social planner. Since the solution to the PDE system does not depend on  $T$ , the verification time  $T$  does not need to be determined at the beginning.

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- The terminal commitment  $Z_T = X_T$  could arise from the verification of the social planner. Since the solution to the PDE system does not depend on  $T$ , the verification time  $T$  does not need to be determined at the beginning.
- Note that the randomness of  $X_t$  consists of two components: the initial randomness and the randomness introduced by  $W$ . If we restart the game at time  $T$ , the randomness associated with  $W$  should be filtered out, as its realization has already been observed (analogous to a quantum collapse);

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# Stochastic correlated equilibrium for $N$ players

In a game with  $N$  players, each player receive a personal signal process  $X^i$  in the form:

$$dX_t^i = b(X_t^i, t)dt + \sigma dW_t^i.$$

At the correlated equilibrium the signal processes satisfy:

$$X^i = \arg \min_{Z_t^i = \phi_t(X_t), Z_T^i = X_T^i} \mathbb{E} \left[ \int_0^T L(\mu_t^N, Z_t^i, D_t^+ Z_t^i, t) dt \right], \quad \mu_t^N := \frac{1}{N} \sum_{j \neq i}^N \delta_{X_t^j}.$$

Here we shall modify the definition:

$$D_t^+ Z_t^i := \lim_{h \rightarrow 0} \frac{1}{h} \mathbb{E}[Z_{t+h}^i - Z_t^i | X_t].$$

# Intuition of mean field limit

For  $L(\mu, X^i, D^+X^i) = V(\mu, X^i) + \frac{1}{2}|D^+X^i|^2$ , the corresponding stochastic Euler-Lagrange equation for fixed  $\mu$  reads:

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Recall that

$$\begin{aligned} D_t^- f(t, X_t^i) &= (\mathcal{L}_{x^i} f - \sigma^2 \nabla_{x^i} \ln m \cdot \nabla f)(t, X_t^i), & m_t &= \text{Law}(X_t) \\ &\approx (\mathcal{L}_{x^i} f - \sigma^2 \nabla_{x^i} \ln m^i \cdot \nabla f)(t, X_t^i), & m_t &= \text{Law}(X_t^i). \end{aligned}$$

# Mean field equilibrium: a system of three equations

As we see, the Euler-Lagrange equation evolves the triple  $(X^i, m^i, \mu)$ . Therefore, the mean field equilibrium is governed by the system of the three equations ( $\eta = \log m^i, \xi = \log \mu$ ):

$$-\partial_t \xi + \frac{1}{2} \Delta \xi - b \cdot \nabla \xi + \frac{1}{2} |\nabla \xi|^2 - \nabla \cdot b = 0, \quad \xi(0, \cdot) = \xi_0$$

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Note that the initial distribution of the population ( $\mu_0$ ) can be different from that of the individual ( $m_0^i$ ).

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Note that the initial distribution of the population ( $\mu_0$ ) can be different from that of the individual ( $m_0^i$ ).

Very different from the correlated mean field game studied by Campi, Fisher, Elie, Lauriere...



Thank you for your attention!