# Sophisticated Learning

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Topics in Economic Theory

# Overview

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#### Learning in Games

#### How do people get to play equilibrium?

Main question of interest in 'learning in games' (7 games with learning)

#### Goals

Provide foundations for existing equilibrium concepts.

Capture lab behaviour.

Predict adjustment dynamics transitioning to new equilibrium.

(akin to 'impulse response' in macro; uncommon but definitely worth investigating)

Select equilibria.

Algorithm to solve for equilibria.

Explain persistence of heuristics/nonequilibrium behaviour.

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#### Papers:

Kalai and Lehrer (1993, Ecta) "Rational Learning Leads to Nash Equilibria" Kalai and Lehrer (1993 Ecma) "Subjective Equilibrium in Repeated Games"

#### Model and Notation

**Stage game**: finite players  $i \in I = \{1, ..., n\}$ ; actions  $A_i$ ;  $A = \times_i A_i$ ; payoffs  $u_i : A \to \mathbb{R}$ .

**Infinite horizon repeated game**: discount  $\delta_i \in (0,1)$ ; perfect monitoring.

**Histories**:  $h^t = (a^0, \dots, a^{t-1}) \in A^t$ , with  $H^t = A^t$ ,  $H = \bigcup_{t>0} H^t$ ; empty history  $\emptyset$ .

**Behavioural strategies**:  $\sigma_i = (\sigma_{i,t})_{t \geq 0}$ ,  $\sigma_{i,t} : H^t \to \Delta(A_i)$ ; strategy profile  $\sigma = (\sigma_i)_i$ .

Outcome measure: for a fixed  $\sigma$ , let  $\mu^{\sigma}$  be the induced probability on infinite play paths  $\Omega = A^{\mathbb{N}}$ 

**Subjective beliefs**: each player *i* has a prior belief  $v_i$  over opponents' strategies  $\sigma_{-i}$ ; induces a belief over play paths  $\Pi_i$ .

**Absolute continuity (truth-compatibility)**:  $\mu^{\sigma} \ll \Pi_i$  for all *i* (no path of positive  $\mu^{\sigma}$ -probability is assigned zero by  $\Pi_i$ ).

**Objective**: players maximise expected discounted payoffs given their posteriors and choose best responses period by period.

## Bayesian Updating and Merging

**Posterior on play paths**: after  $h^t$ , player i updates  $\Pi_i(\cdot \mid h^t)$  by Bayes' rule (well-defined by absolute continuity).

**Merging (KL notion)**: posteriors become *close* on all tail events: for every  $\varepsilon > 0$ ,  $\exists T$  s.t.  $\forall s > T$ .

 $|\Pi_i(A \mid h^s) - \mu^{\sigma}(A \mid h^s)| \le \varepsilon$  for all A in a large (probability-1) class of events.

**Relation to Blackwell–Dubins**: KL's closeness  $\iff$  BD merging; KL give an elementary proof and equivalence of topologies.

## Subjective Equilibrium

#### **Definition (Subjective equilibrium)**

A history-dependent strategy profile  $\sigma$  is a **subjective equilibrium** if, along  $\mu^{\sigma}$ -almost all paths, players' posteriors about future play coincide with the truth (merging), and each  $\sigma_i$  is a best response to the posterior over  $\sigma_{-i}$ .

**Interpretation**: learning exhausted; disagreements (if any) are off path and never observed.

**Consequence**: from some finite time T, actions follow best responses to (approximately) correct forecasts of future play.

## Main Result: Rational Learning $\implies$ Nash Equilibrium

#### Theorem 1 (Kalai and Lehrer 1993)

Suppose  $\mu^{\sigma} \ll \Pi_i$  for all *i* (absolute continuity). Under Bayesian updating and optimal control of expected discounted utility:

Posteriors merge with the truth along the realised path.

From some finite time, play is  $\epsilon\text{-optimal}$  against correct forecasts.

Limit behaviour constitutes a Nash equilibrium of the repeated game.

#### **Proof Sketch**

**Merging**: apply KL's merging theorem to obtain posterior convergence on tail events.

**Optimality**: best responses w.r.t. posteriors  $\implies \epsilon$ -optimality w.r.t. truth for large t.

**Equilibrium**: mutual best responses along the limit set  $\implies$  Nash equilibrium of the repeated game.

#### Corollaries and Immediate Implications

#### Corollary 2 (Kalai and Lehrer 1993; BD 1962)

For  $\mu^{\sigma}$ -a.e. path, the posterior probabilities  $\Pi_i(\cdot \mid h^t)$  converge uniformly on the large class of events to  $\mu^{\sigma}(\cdot \mid h^t)$ . (Version of Blackwell–Dubins' merging.)

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#### Corollary (Incomplete information; Bayesian Nash equilibrium)

In a discounted repeated game with a finite/countable type space for payoffs, if play starts at a **Bayesian Nash equilibrium** of the incomplete-information repeated game, then eventually players play a Nash equilibrium of the realised complete-information repeated game.

Intuition: at BNE of the incomplete-information repeated game, priors imply absolute continuity on play paths; merging  $\implies$  players act as if types were known.

## Meaning and Interpretation

- What converges? Posteriors about future play; best responses to (nearly) correct forecasts ⇒ Nash play in the repeated game.
- **Role of absolute continuity**: rules out dogmatic priors that assign zero to realised events; ensures Bayes can learn from data.
- **Why repeated games?** Stationarity of opponents' *strategies* (not actions) makes learning feasible despite strategic feedback.
- **Experimentation**: endogenous via dynamic optimisation of discounted utility; no ad hoc trembles needed.

## Additional Details (Notation as in §3.1)

**Spaces**:  $\Omega = A^{\mathbb{N}}$  with product  $\sigma$ -algebra; cylinders generated by finite histories.

**Outcome law**:  $\mu^{\sigma}(\cdot) = \mathbb{P}_{\sigma}(\cdot)$  over  $\Omega$ ; filtration  $\mathcal{F}_t$  from  $H^t$ .

Beliefs on strategies  $\rightarrow$  beliefs on paths: priors  $v_i$  over  $\sigma_{-i}$  induce  $\Pi_i$  over  $\Omega$  (via mapping  $\sigma_{-i} \mapsto \mu^{(\sigma_i,\sigma_{-i})}$ ).

Absolute continuity on  $\Omega$ :  $\mu^{\sigma} \ll \Pi_i$ ; equivalently, every cylinder  $C(h^t)$  with  $\mu^{\sigma}(C(h^t)) > 0$  has  $\Pi_i(C(h^t)) > 0$ .

**Payoffs**:  $U_i(\sigma) = \mathbb{E}_{\mu^{\sigma}} \left[ \sum_{t>0} \delta_i^t u_i(a^t) \right].$ 

## Critiques (Fudenberg and Levine 1998)

- **Endogeneity of absolute continuity**: AC must hold for the *realised* play path ⇒ fixed-point flavour; as hard as equilibrium selection.
- **Grain of truth**: desirable to ensure AC *regardless* of opponents' play is impossible on uncountable history spaces; weaker classes of priors may work only in truncated/favourable settings.
- **Example (Chicken)**: plausible sets "insist n periods then yield"; symmetric beliefs  $\implies$  optimal stopping leads to paths of measure 0 under priors  $\Rightarrow$  AC fails.
- **Interpretation**: best seen as a descriptive result on eventual consensus, not as a *learning path* to equilibrium with exogenously specified priors.

#### **Takeaways**

 $\label{thm:continuity} \mbox{ Under absolute continuity, Bayesian learning merges beliefs with the truth on play paths.}$ 

Optimal control with merged beliefs  $\implies$  eventual play of a Nash equilibrium of the repeated game.

Application to *Bayesian Nash equilibria* with incomplete information: eventually play NE of the realised complete-information game.

AC is strong/endogenous; caution interpreting KL as a general path-to-equilibrium theory.

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