

Stationary social learning in a changing environment

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- Social learning literature:
 - two sources of information: private and social learning,
 - permanent state
- Changing state
 - natural assumption in many settings
 - rare and rapid political transitions: Arab Spring, 1830 liberal revolutions, carnival of Solidarity August 1980 - December 1981
- Why changing state matters?
 - Grossman-Stiglitz paradox makes stopping learning (i.e., informational cascade) not possible.

Introduction

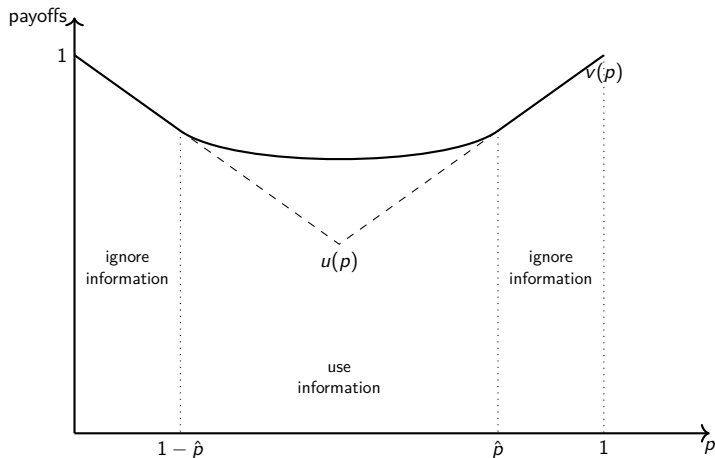
Questions and results

Question	Result
learning efficiency + welfare	no asymptotic learning, even with slowly changing state
is more social learning better?	it can be worse
behavior and beliefs	uniformity under slowly changing state, rare, rapid transitions

- Most striking results are when state is (very) persistent, but not permanent.
- Related lit: Moscarini Ottaviani Smith (98), Dasaratha Golub Hak (20), Kabos Meyer (21)

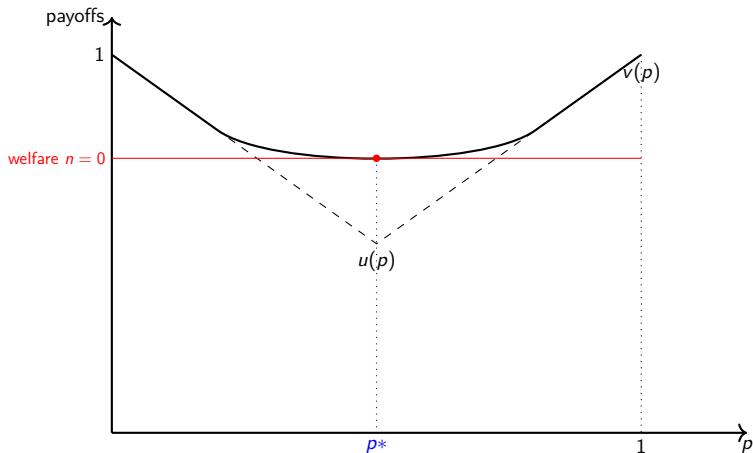
- Markov-changing state $\theta_t \in \{0, 1\}$.
 - $\lambda = P(\theta_{t+1} \neq \theta | \theta_t = \theta)$,
 - neither independent ($\lambda = \frac{1}{2}$), nor permanent ($\lambda = 0$),
 - slow transitions ($\lambda \rightarrow 0$)
- Social learning: in each period, continuum of short-lived agents
 - random sample of n actions from the previous period(s),
 - private signal at cost $c \geq 0$,
 - action $a \in \{0, 1\}$,
 - utility $u(a, \theta_t) = \mathbf{1}(a = \theta_t)$.
- Stationary equilibrium $\mu \in \Delta(X \times \{0, 1\})$, where
 - $x \in X = [0, 1]$ if the fraction of population playing 1.

- Assumption: private signal is
 - either costly $c > 0$, or
 - free but with bounded precision,
- p - private belief about the state
 - $u(p)$ - expected payoff from optimal action
 - $v(p) \geq u(p)$ - expected payoff from optimally using information and then taking action



- Assumption implies that $v(p) = u(p)$ for extreme beliefs.

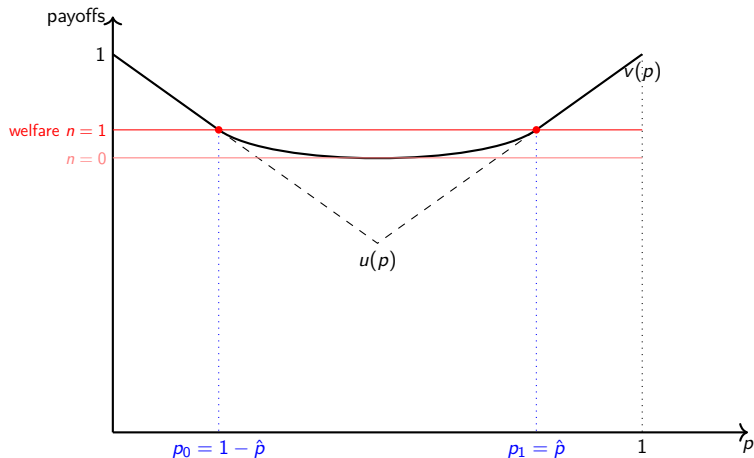
Welfare $n = 0$



- each generation is identical and can only learn from private signal

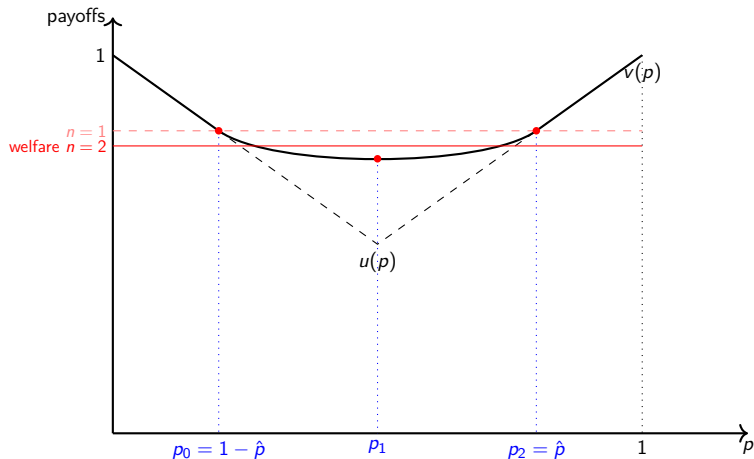
- Permanent state ($\lambda = 0$)
 - $n = 0$: no social learning
 - $n = 1$: some social learning but herding (Banerjee 92, Bikchandani et al 92)
 - $n = 2$: asymptotic full learning (Banerjee Fudenberg 2004)
- If state is permanent, social learning helps!

Welfare $n = 1$



- $p_0, p_1 \in [1 - \hat{p}, \hat{p}]$ and if $\lambda \leq \lambda^*$, then $p_1 = \hat{p} = 1 - p_0$.
- p_k - belief after observing k agents (out of n sample) playing 1.

Welfare $n = 2$



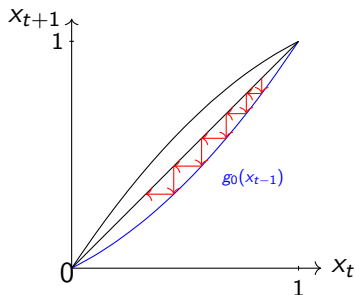
- **Theorem:** $p_0, p_1, p_2 \in [1 - \hat{p}, \hat{p}]$ and if $\lambda \leq \lambda^*$, then $p_2 = \hat{p} = 1 - p_0$.

Welfare $n = 2$

Proof sketch

- Suppose $p_2 > \hat{p}$: non-confused agents *don't* buy information.
 - ϕ_θ the probability that an agent with interim beliefs $p_1 = \frac{1}{2}$ (who thus acquires info) plays action 1 in state θ .
 - x_t the fraction of agents playing action 1 at date t

$$x_{t+1} = x_t^2 + 2x_t(1 - x_t)\phi_{\theta_{t+1}} =: g_{\theta_t}(x_t).$$



Welfare $n = 2$

Proof sketch

$$x_{t+1} = x_t^2 + 2x_t(1 - x_t)\phi_{\theta_{t+1}} =: g_{\theta_t}(x_t).$$

- Sooner or later, x_t will be close to 0.
- Around 0, geometric random walk with Markov drift.

$$x_{t+1} \simeq (2\phi_{\theta_{t+1}}) x_t$$

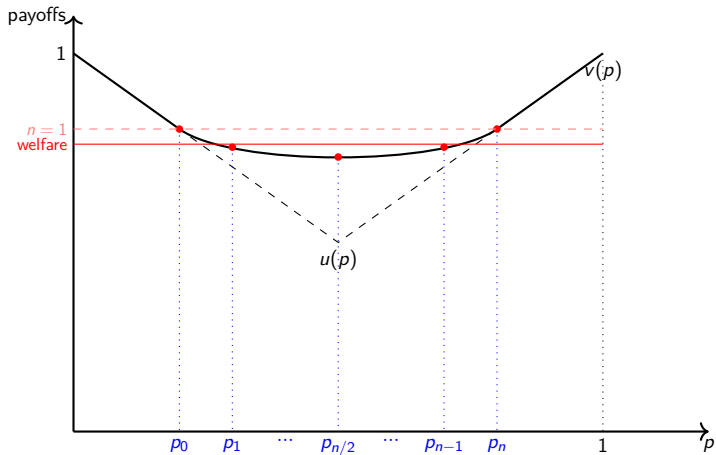
- since $(2\phi_0)(2\phi_1) = 4\phi_0(1 - \phi_0) < 1$, the drift is negative.
- ⇒ $\lim x_t \in \{0, 1\}$, a.s.
- ⇒ Samples are **uninformative**

- For $n > 2$, we need some assumptions:
 - persistent (but not permanent) state: $\lambda \rightarrow 0$
 - precise signals: $n^2 \phi_1 (1 - \phi_1) < 1$,
 - example: perfect signals
 - regular equilibrium: $p_{n-1} \geq \frac{1}{2}$.

Theorem

Under the above assumptions, in equilibrium, $\lim_{\lambda \rightarrow 0} p_n^\lambda = \hat{p} = 1 - \lim p_0^\lambda$.

Welfare $n \geq 2$



- Let $\mu_\lambda \in \Delta(X \times \{0, 1\})$ be the stationary equilibrium.

Theorem

If $n \geq 2$, then, for each $\varepsilon > 0$

$$\lim_{\lambda \rightarrow 0} \mu_\lambda \{\varepsilon \leq x \leq 1 - \varepsilon\} = 0.$$

- uniform behavior, most of the time,
- together with previous result, uniform beliefs

Behavior: Consensus

Proof sketch

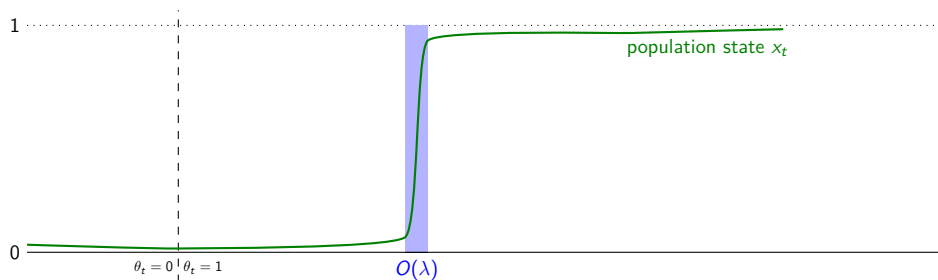
- Each agent observes random sample (a_1, \dots, a_n)
 - a_1 is λ -close to optimal action,
 - a_2 cannot add too much information if λ is small,
- $\Rightarrow a_1 = a_2$, most likely.

Theorem

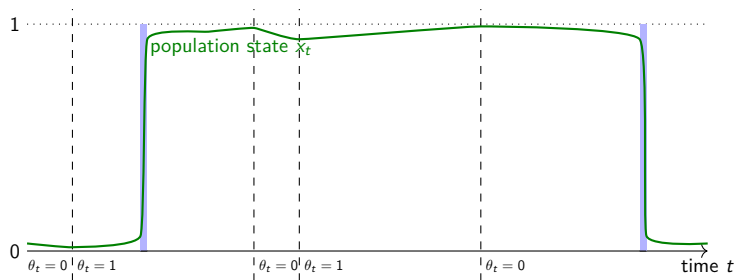
There exists a constant $K < \infty$ such that

$$\int x(1-x) d\mu^\lambda(x, \theta) \leq K\lambda.$$

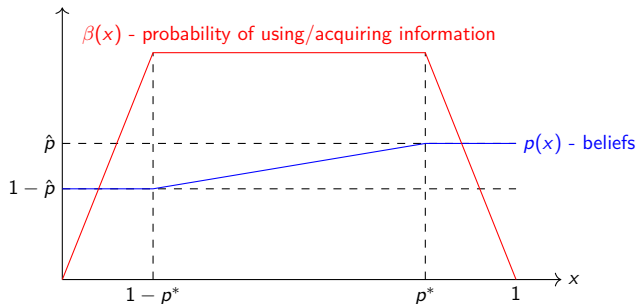
Behavior: Rapid transitions



Behavior: Rapid transitions



- Continuum time version \leftarrow we can compute equilibria.
- Binary signals



- Suppose $n \geq 2$, and
 - $A = [0, 1]$ and $u(a, \theta) = -(a - \theta)^2$,
 - perfect signals
- With permanent state ($\lambda = 0$) \rightarrow asymptotic learning and welfare = 1.
- With persistent state ($\lambda > 0$) \rightarrow there are stationary equilibria with welfare < 1 .

- Social learning with changing state
- Even when state is very persistent (but not permanent):
 - no asymptotic learning, uniformly bounded welfare
- Additionally, when state is persistent
 - behavior and beliefs exhibit consensus,
 - rare and rapid transitions.