# The Evolution of Belief Ambiguity during the Process of High-School Choice by Pamela Giustinelli and Nicola Pavoni 

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- What are their costs?
- What assumptions do we need to make progress?


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- Characteristics of $S$ perfectly known (probability of success at $S$
- Two children, Alice $(A)$ and Beth $(B)$
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- Set of states of nature: $\Omega_{1} \times \Omega_{2}$
- $\Omega_{1}=\{$ Both pass, Both fail, Only Alice passes, Only Beth passes $\}$
- $\Omega_{2}=$
$\{$ Lots of math, Little math $\} \times\{$ Ancient greek offered, not offered $\} \times$ \{Will be stuck on drawing homework every Sunday morning, not stuck\}


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$\{$ Lots of math, Little math $\} \times\{$ Ancient greek offered, not offered $\} \times$ \{Will be stuck on drawing homework every Sunday morning, not stuck\}
- Alice and Beth ex ante identical:
- Same prior $\mu_{0}$ or set of priors $M_{0}$
- Probability of success is the same conditional on any $\omega_{2} \in \Omega_{2}$.


## Alice and Beth as Bayesians

- Observe $A$ and $B$ 's posterior beliefs at 3 stages, $\mu_{i j}, i=1,2,3$, $j=A, B$
- Evolution of beliefs dictated by learning about $\omega_{2} \in \Omega_{2}$
- Learning may be idiosyncratic, beliefs may be different...
- ... but they should converge if $\omega_{2}$ becomes known.


## Alice and Beth meet Gilboa and Schmeidler (or Epstein and Schneider)

- $A$ and $B$ have a range of beliefs about success given each $\omega_{2}$.
- $A$ and $B$ have a range of beliefs over which $\omega_{2}$ is true.
- Updating: Bayesian belief by belief.
- Belief range should converge as $\omega_{2}$ becomes known.
- Convergence might be messy


## Example of Messy Convergence

- Alice and Beth have $90 \%$ chance of passing if Greek is not part of curriculum
- With Greek, they have no idea (support $[0,1]$ )
- Prior: $50 \%$ that Greek is offered.
- Prior range: [45\%, 95\%]
- Posterior range: $90 \%$ or $[0,1]$


## A Way to Make Progress

- Assume that all uncertainty is about learnable characteristics $\left(\omega_{2}\right)$
- or, follow alternative approach to updating (Hansen and Sargent)
- Then range of beliefs will shrink with learning
- Will also converge across $A$ and $B$ in the limit


## What can I identify?

- Suppose I have panel with short time dimension, many ex ante identical people with i.i.d. learning process
- Individual learning does not converge, but cross-section informative of true state
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- For each student, observe belief, choice
- $\Longrightarrow$ Infer preferences
- $\Longrightarrow$ Infer measure of people that made wrong choice


## Identification under ambiguity

- Cannot learn true probability in general
- Can get bounds that are tighter than individual students'
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- Might also quantify role of forgetfulness (assuming that it is forgetfulness)


## Problem: People are Different

- Try matching over observable characteristics
- Impose monotonicity restrictions (better GPA makes certain schools more desirable)

