# Discussion of "Bayesian Foundations of Constant-Gain Learning" by A. Nakov and G. Nuño

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#### Overview

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- while taking seriously recent evidence on expectation formation (Malmendier and Nagel)
- tackles aggregation issues in models with learning

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- constant probability of dying
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- key: when an agents dies, the successor inherits wealth, but not beliefs
- initial beliefs of newborn agents centered on RE beliefs
- If we want to take a model of this form to the data, we would probably have to think harder about initial beliefs - seems to be a daunting challenge

# Key Result

- ▶ Let *n* be the age of an agent
- ▶ average beliefs  $\gamma_t$  (as  $N \to \infty$ ?) can be approximated by a constant gain algorithm

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Can this be generalized to models where agents have more general PLMs?

## How good is this approximation?

► The approximation result on the previous slide only holds under the following condition:

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- ▶ It would be nice to have some error bounds
- How general is this?
- ► How well would a representative agent model with constant gain learning approximate the equilibrium dynamics?

Bayesian Foundations

#### Conclusion

- Very interesting paper
- ► As someone who uses constant gain learning, it would be great to know how general the results are
- ► The heterogenous agent model itself seems a great tool that could potentially find lots of uses