

Discussion:  
Will Artificial Intelligence Replace Computational  
Economists Any Time Soon?  
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# Motivation

- ▶ Great paper:

- ▶ Important research question:

- ▶ Can a generic AI algorithm replace the need for computational economists to write model-specific code?
    - ▶ Answer: No!

It's complicated ... My lesson learned: Efficient use will need experts to choose best implementation even if a lot is standardized.

We can probably massively augment our computational capacity.

- ▶ Particular contributions:

- ▶ Elegant, clearly written “cookbook” of machine learning applied to economics
    - ▶ Presentation of a generic, non-supervised deep-learning algorithm with one static objective function

# Summary 1/2

- ▶ What's in the cookbook?
  - ▶ Elegant formulation of commonly faced dynamic stochastic optimization problem as a single optimization problem, with two possible objective functions:
    - ▶ life-time reward function
    - ▶ total sum of squared residuals from model equations, e.g. Euler (relevant if there is more than a single agent)
  - ▶ Solution technique: deep learning which arises naturally from decision rules nested over time
    - ▶ employs stochastic gradient method
    - ▶ approximates policy function
    - ▶ assume full knowledge of the model
    - ▶ does not need wide data availability due to simulations
  - ▶ Specific recipe: Application to consumption-savings problem

## Summary 2/2

- ▶ Advantages/disadvantages of the approach discussed:
  - ▶ The math is simple!
  - ▶ Linear, not quadratic growth of the number of parameters with dimensionality
  - ▶ Approximates kinked functions naturally
  - ▶ Represents complicated structures due to deep-learning
  - ▶ Technology-intensive: ideally, use Google Cloud TPU, Tensorflow or Pytorch software – a high fixed cost, but a technological improvement (graph representation)

# Discussion Overview

Next:

- ▶ A layperson's questions
- ▶ 2 comments

# Comment 1: Comparison to the literature, in particular reinforcement-learning

- ▶ How does your paper compare to Duarte (2018), both formally and in terms of content?
- ▶ In terms of content:
  - ▶ Your approach assumes that the model is known while reinforcement learning does not and is model-free
  - ▶ Online learning in reinforcement learning leads to some tradeoffs
  - ▶ Your formulation proposes offline learning in a single optimization problem setting
  - ▶ Can you flesh out 2 cases where either approach does clearly better? The reader would like some guidance what to look for when making algorithmic choices.

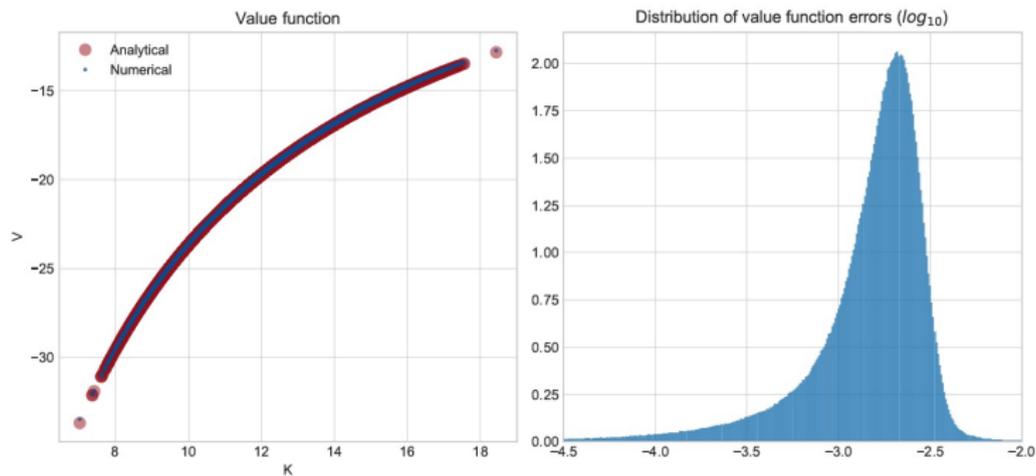
## Comment 1: Comparison to the literature, in particular reinforcement-learning

- ▶ How does your paper compare to Duarte (2018), both formally and in terms of content?
- ▶ Formally:
  - ▶ Duarte explicitly compares reinforcement learning approximate solutions to exact numerical benchmarks
  - ▶ Overall conclusion: can handle up to 10 dimensions well
  - ▶ High-dimension application still missing in draft... Can you be more specific about its features?

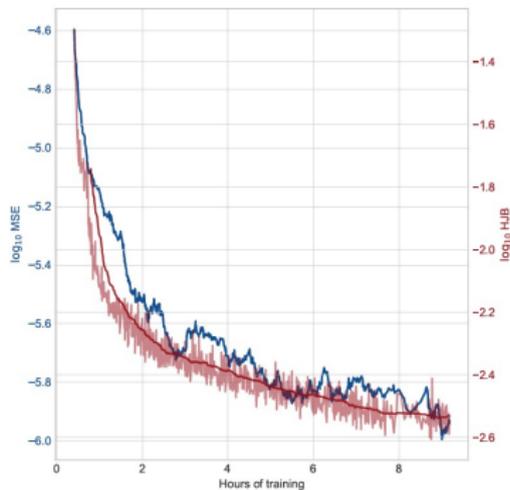
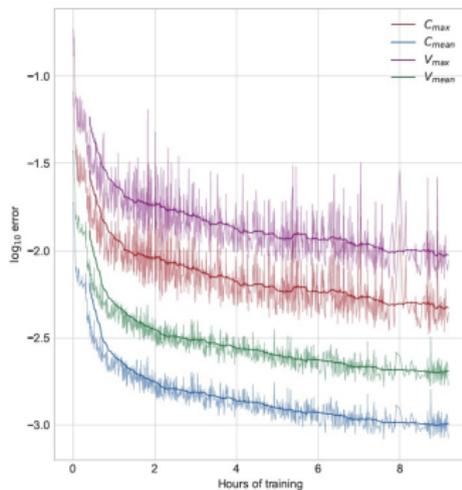
## Comment 2: Careful comparisons of approaches?

- ▶ The reader would like some tables with more evaluations of your algorithm, and comparisons to other algorithms
  - ▶ Explicitly compare exact and approximate solutions where possible
  - ▶ Large-scale capacities? (e.g. multiproduct pricing problem as in Alvarez)
  - ▶ Performance in terms of time, and costs (e.g. Google Cloud)?
  - ▶ Can you present more of a cookbook, perhaps with some stylized “recipes”? When does a particular algorithm do well? What are some clearcut criteria - is it ultimately research \$? (time yourselves) What about carbon emissions?
  - ▶ Given complexity of attention allocation, some clear discussion of pros and cons of algorithms in the context of specific examples might be more useful

## Comment 2: Careful comparisons of approaches?



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# Conclusion

- ▶ Very nice paper - great exposition of generic AI algorithm.
- ▶ Can become a useful guide as an AI cookbook.