



# Non-Determinism and the Lawlessness of Machine Learning Code

A. Feder Cooper, Jonathan Frankle, and Chris De Sa

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Machine learning (ML) code is **non**-deterministic our paper gets into this in detail

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- 3. Clarifying the effects of non-determinism by reasoning about distributions over outcomes
- 4. Breaking the cyberlaw "code is law" frame with non-deterministic ML code

#### Non-determinism

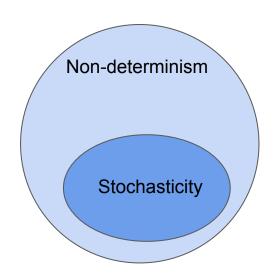
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#### **Stochasticity**

A property of non-deterministic processes whose outcomes can be reasoned about using probability theory.

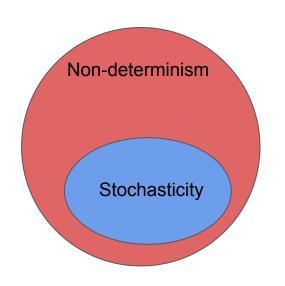


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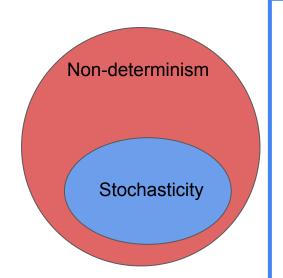


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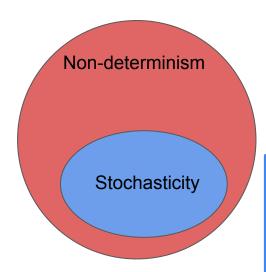
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Existing popular ML frameworks (e.g., PyTorch)



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**Existing legal literature** 

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What about **non**-stochastic non-determinism?

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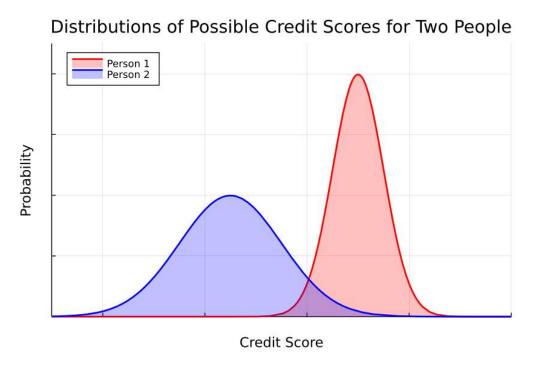
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#### Taking a distributional approach

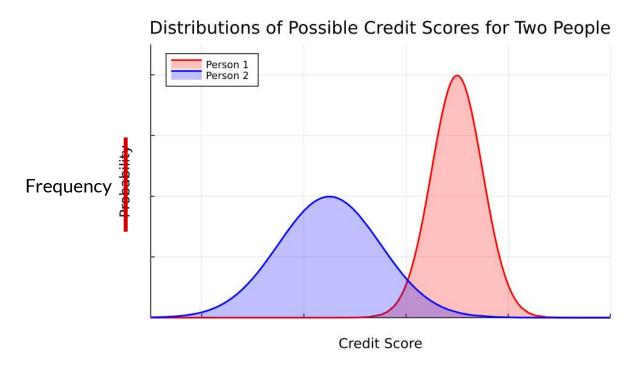
- 1. Clarifies prior scholarship on stochasticity
- 2. Accounts for other types of non-determinism

### Ex. 1: Distributions over Individual Outcomes



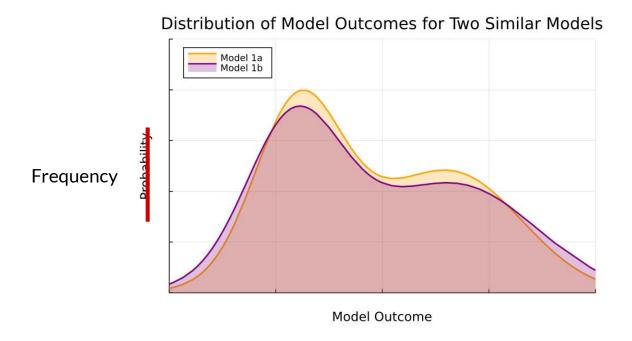
A distributional view of stochasticity

### Ex. 1: Distributions over Individual Outcomes



A distributional view of non-stochastic non-determinism

### Ex. 2: Distributions over Models



A distributional view of non-stochastic non-determinism

# Takeaways from these examples

Reasoning about distributions over outcomes clarifies the importance of non-determinism

Non-stochastic non-determinism also has important normative effects on ML outcomes

### Non-deterministic code is lawless

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**Non-stochastic non-determinism** can cause **contradictions** and is **unpredictable** 

# Takeaways and other work

The law can take care to reason about **distributions over outcomes**, to expose the effects of different types of non-determinism

ML can develop more precise tools to reason about different types of arbitrariness and to mitigate its impacts





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Thank you!