

# Generating Realistic Wear Distributions for SSDs

**Ziyang Jiao, Bryan S. Kim**

Syracuse University

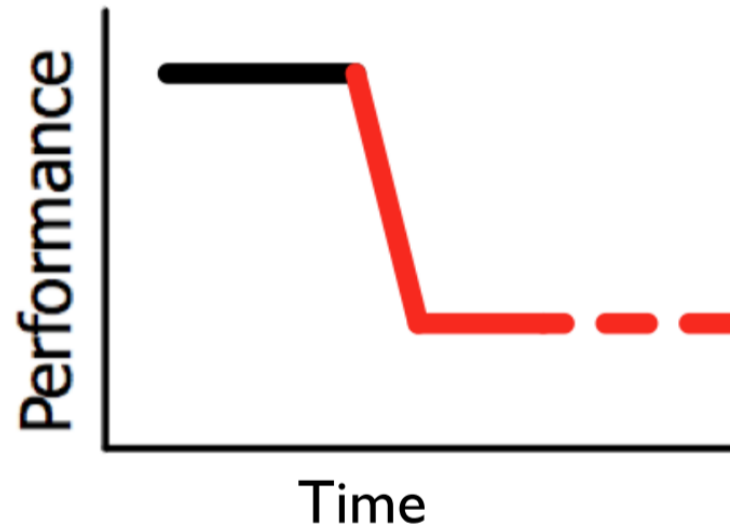


# In a nutshell

- The rise of SSDs
- SSD development platforms
  - FlashSim - SIMUL 2009
  - SSDSim - ICS 2011
  - FTLSim - SYSTOR 2012
  - FEMU - FAST 2018
  - MQSim - FAST 2018
  - Amber - MICRO 2018

# In a nutshell

- The rise of SSDs
- SSD development platforms
- The fail-slow symptom of SSDs



- Haryadi S. Gunawi et al, “Fail-Slow at Scale: Evidence of Hardware Performance Faults in Large Production Systems”, FAST 2018

# Current art

## *Preconditioning:*

The process of writing data to the device to prepare it for steady state measurement.

- <https://www.snia.org/sites/default/files/technical-work/pts/release/SNIA-SSS-PTS-Enterprise-v1.1.pdf>

# Current art

## *Preconditioning:*

The process of writing data to the device to prepare it for steady state measurement.

**Expensive:**

**✘ Resources**

**✘ Time**

- <https://www.snia.org/sites/default/files/technical-work/pts/release/SNIA-SSS-PTS-Enterprise-v1.1.pdf>

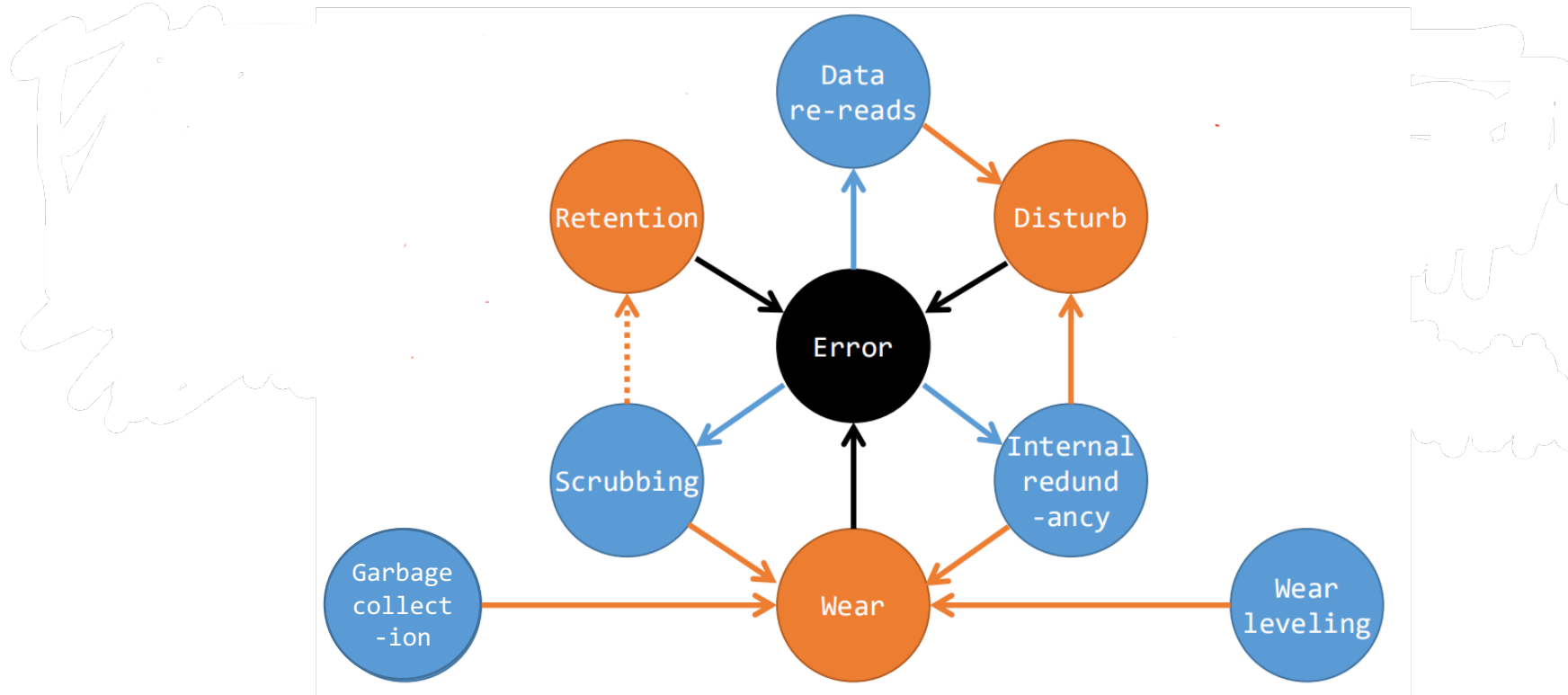
# Our approach

Fast-Forwardable SSD:

a machine learning-based SSD aging framework that  
generates representative future wear-out states

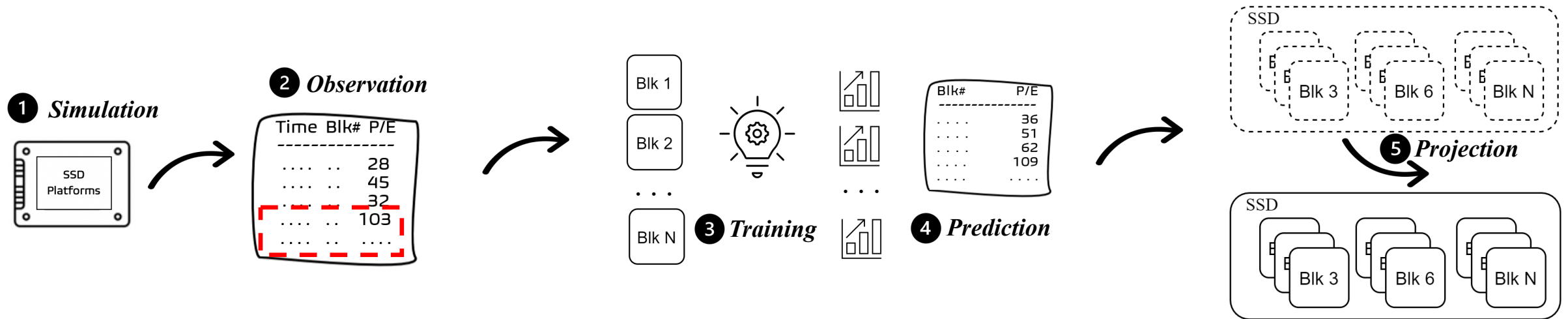
# Challenges

- The cost of machine learning models → **Efficiency**
- The internal intricacy of SSDs → **Accuracy**



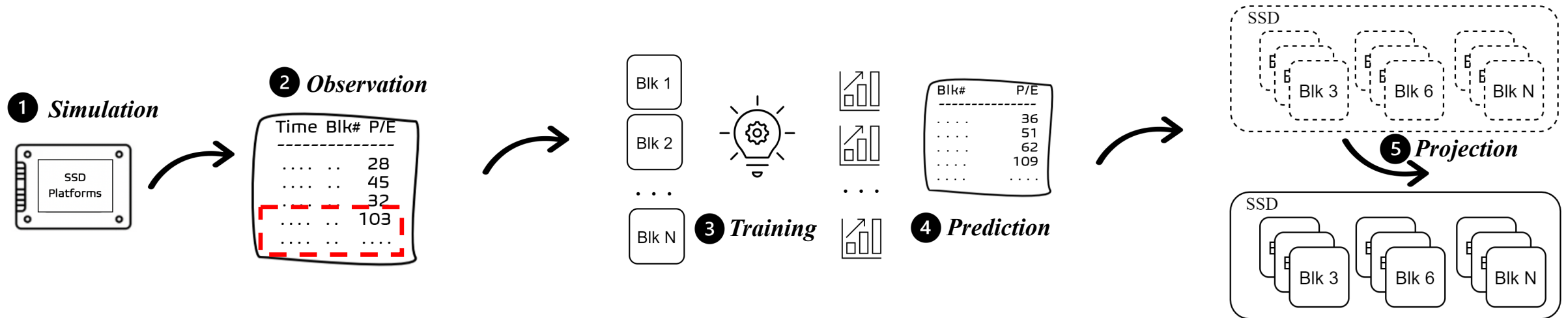
- Bryan S. Kim et al, "CPR for SSDs", HotOS 2019

# System overview





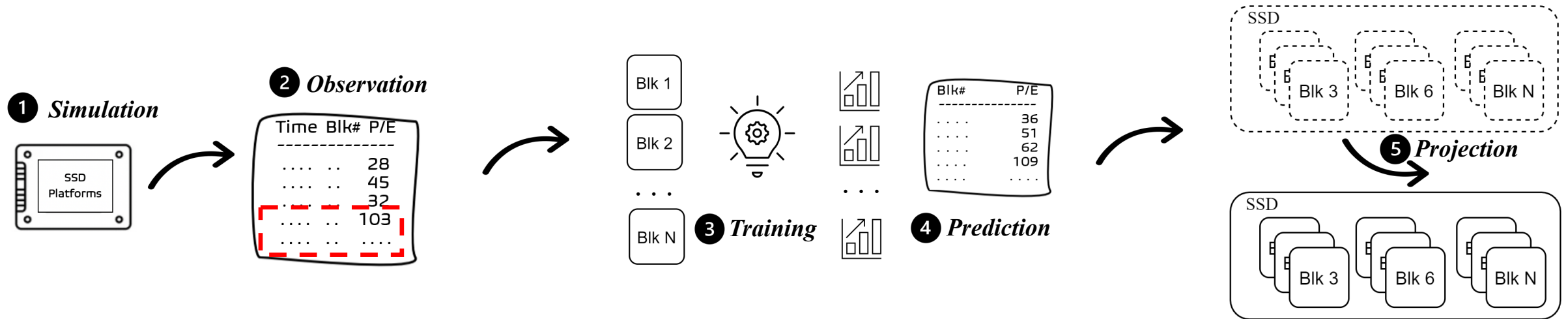
# System overview



## Simulation Stage

- Observe SSD's internal activities
- Output log periodically

# System overview



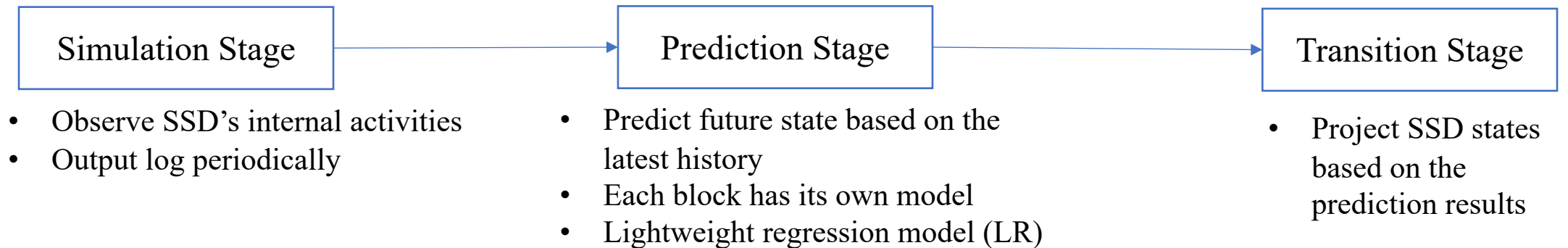
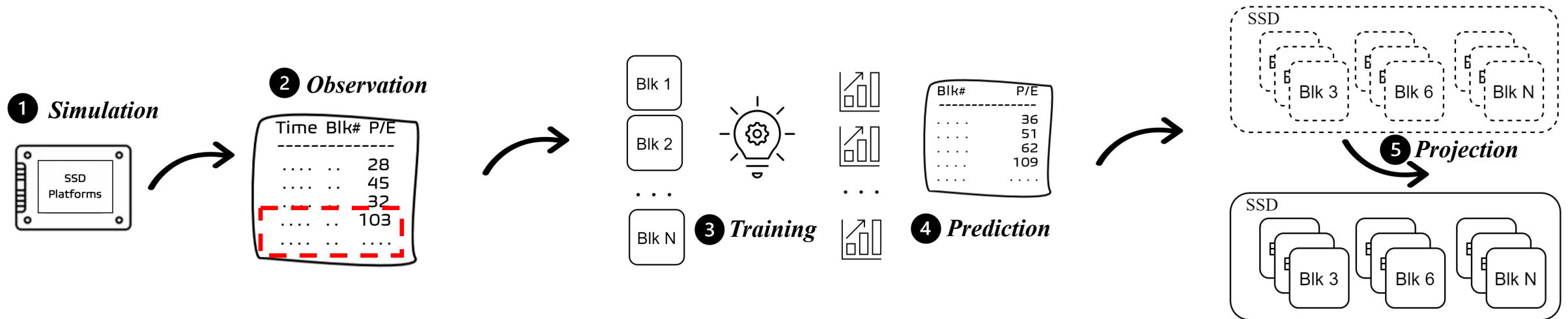
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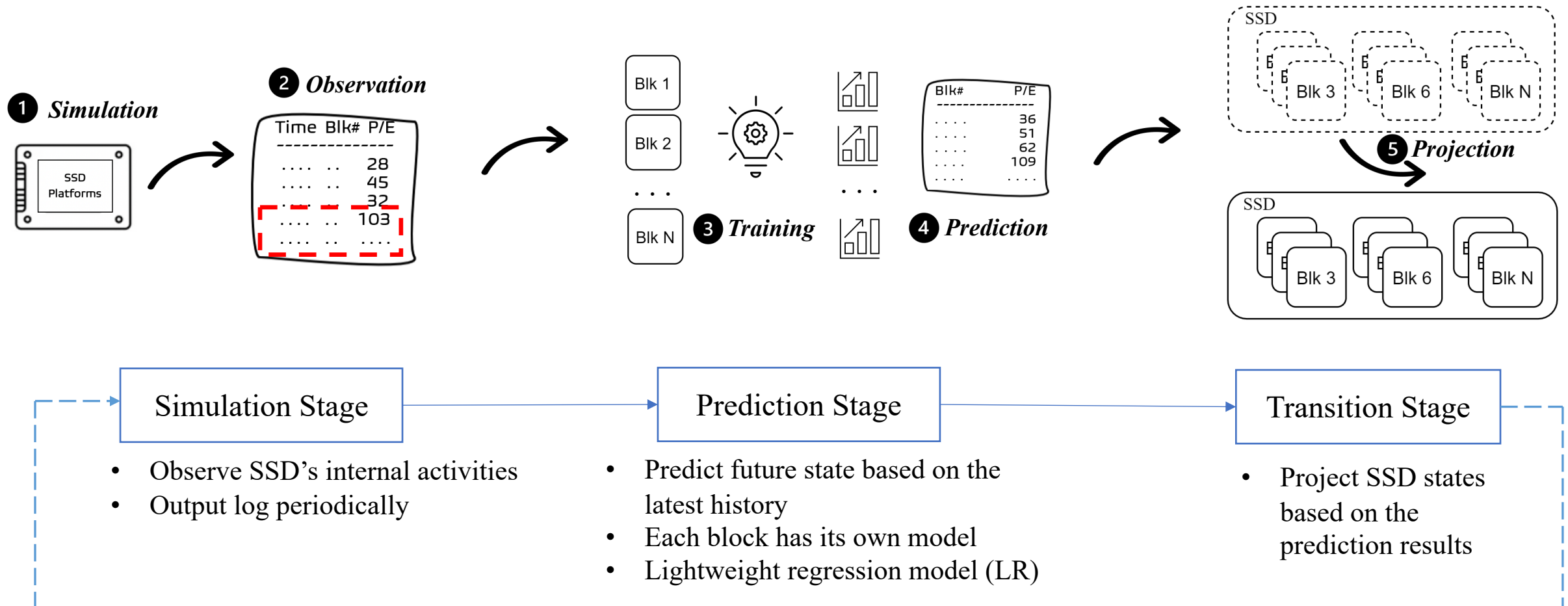
## Prediction Stage

- Predict future state based on the latest history
- Each block has its own model
- Lightweight regression model (LR)

# System overview



# System overview



# Prior work

## Simulation acceleration with ML:

- DEVS

DEVS execution acceleration with machine learning.

*SpringSim 2016*: <https://dl.acm.org/doi/10.5555/2975389.2975399>

- Consider multiple model candidates

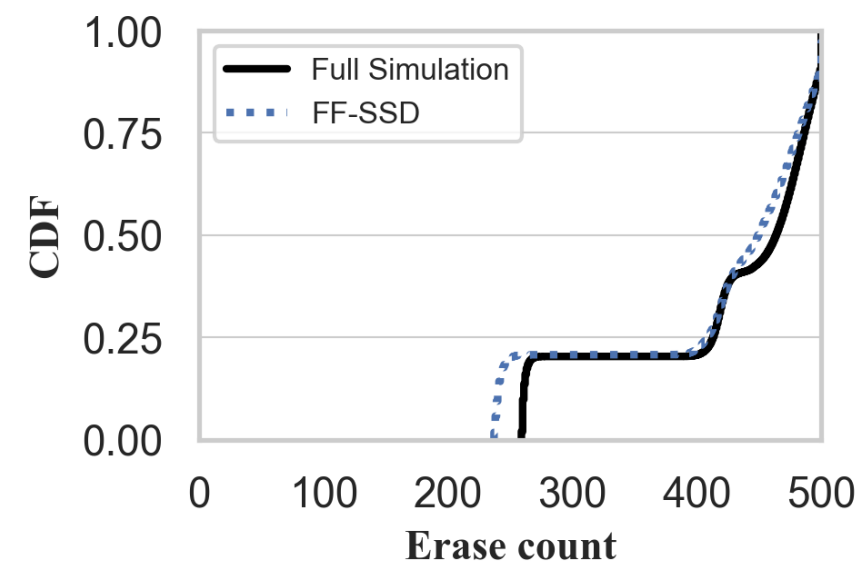
- CML

Using continuous statistical machine learning to enable high-speed performance prediction in hybrid instruction-/cycle-accurate instruction set simulators.

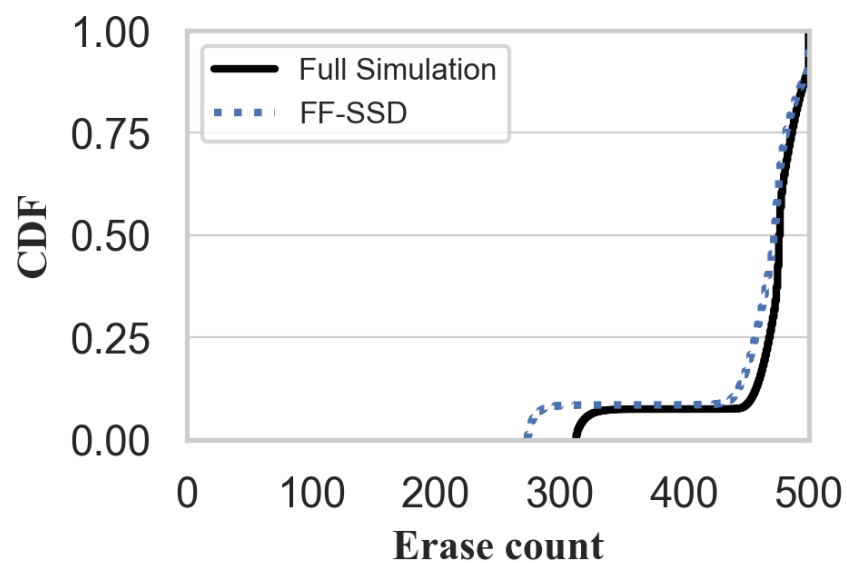
*CODES+ISSS 2009*: <https://dl.acm.org/doi/10.1145/1629435.1629478>

- Continuously incorporate the latest data to update model

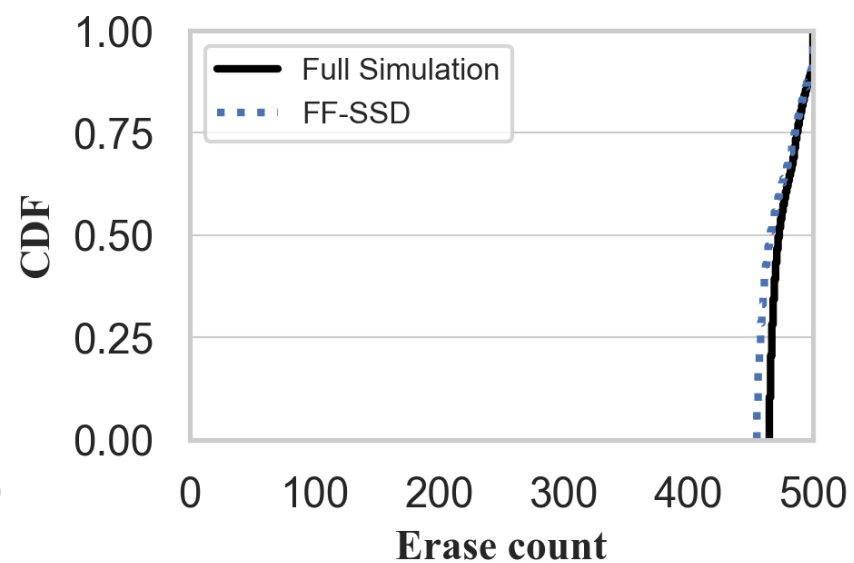
# Evaluations



Windows Build Server



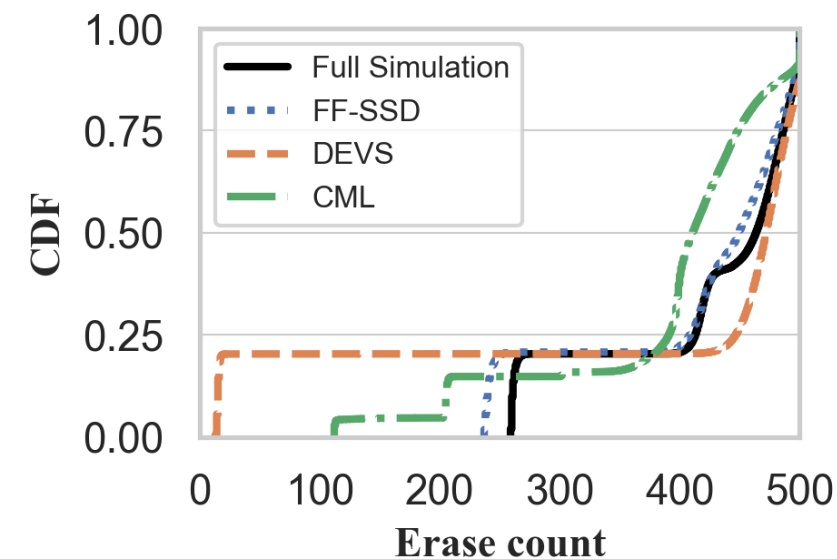
Developer Tools Release



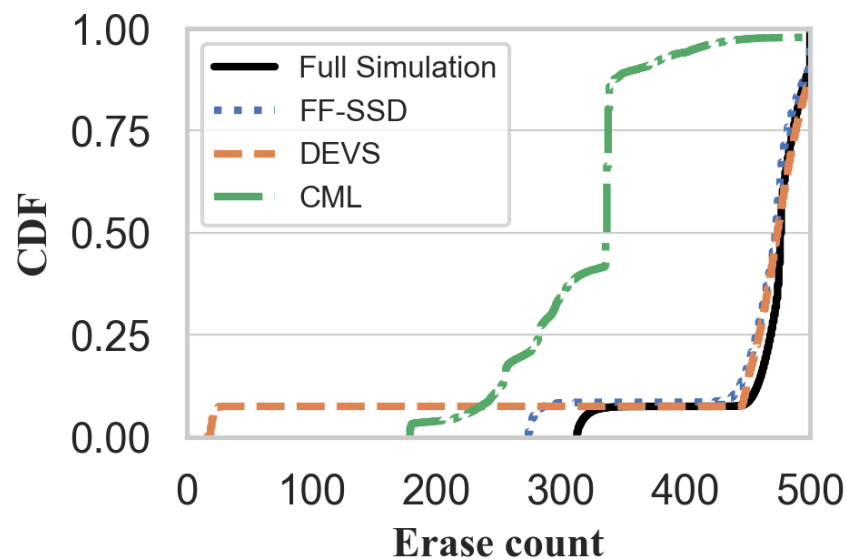
Virtual Desktop Infra

Evaluation under real-world workloads. With one-third workload saving compared to full simulations, FF-SSD generates the final states of SSD with high accuracy.

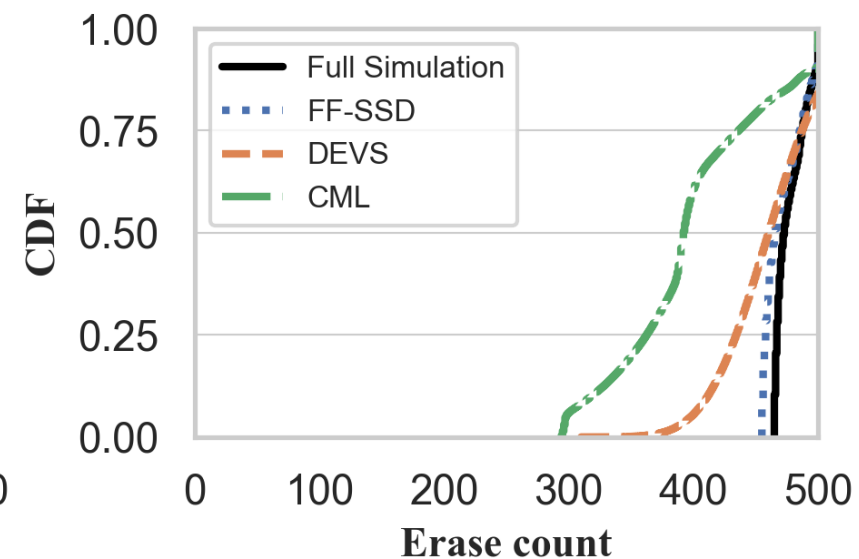
# Evaluations



Windows Build Server



Developer Tools Release



Virtual Desktop Infra

Evaluation under real-world workloads. With one-third workload saving compared to full simulations, FF-SSD generates the final states of SSD with high accuracy.

# Future work

- Analyze *FF-SSD* over various workloads
- Expand *FF-SSD* to support different platforms and SSD models
  - Platforms:
    - Amber
    - FEMU
    - FTLSim
    - ...
- Improve both prediction accuracy and aging efficiency further



*Thank you!*

*Ziyang Jiao*  
*[zjiao04@syr.edu](mailto:zjiao04@syr.edu)*