Lecture 3 of the

MLArchSys Seminar

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Outline of Today's Lecture

Systems	ML <i>for</i> Systems	Machine
Software		Learning

Today's Paper:

Kleio: A Hybrid Memory Page Scheduler with Machine Intelligence

Thaleia Dimitra Doudali Georgia Institute of Technology thdoudali@gatech.edu

Lecture Outline:

Sergey Blagodurov Advanced Micro Devices, Inc. Sergey.Blagodurov@amd.com

Hybrid Memories

Hybrid Memory Management

How to Integrate ML?

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LSTMs

for Hybrid Memory Management (HMem Management)

System Design

Evaluation

Lessons Learned

2/36

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LSTMs

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Hyl	orid Memory Management	
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Big Data hits the Memory Wall







Processing speed grows faster than memory and data transfer speed.

Need for larger and faster memory configurations.

New Memory Technologies

Characteristic	Technology	Vendors	
Low Latency	MRAM		
Uniform Latency	DRAM		
High Bandwidth	HBM	HBM2E HBM2	
Persistent / Non Volatile	PMEM / NVM		

Application Performance	200 PF	
Number of Nodes	4,608	
Node performance	42 TF	
Memory per Node	512 GB DDR4 + 96 GB HBM2	
NV memory per Node	1600 GB	
Total System Memory	>10 PB DDR4 + HBM2 + Non-volatile	
Processors	2 IBM POWER9™ 9,216 CPUs 6 NVIDIA Volta™ 27,648 GPUs	
File System	250 PB, 2.5 TB/s, GPFS™	
Power Consumption	13 MW	
Interconnect	Mellanox EDR 100G InfiniBand	
Operating System	Red Hat Enterprise Linux (RHEL) version 7.4	

Example Configuration of a Supercomputer.

We are in the era of Hybrid Memory Systems. A mix of different technologies at different speeds / capacities / costs.

Hybrid Memory Configurations

In today's paper we assume a hybrid memory system with **DRAM** and **NVM** (Non Volatile Memory).

The NVM actual product was released by Intel in 2019, after the paper was published.

So, the paper had to assume various possible configurations, e.g., capacity ratios.

Intel Optane is packaged together with DRAM.



Source: memverge.com

Guess what? The product got discontinued in 2022..

Intel kills off Optane Memory, writes off \$559 million inventory

In a terrible quarter for the chip giant

July 29, 2022 By: Sebastian Moss 💭 1 Comment

But this is not the end for hybrid memory configurations. It is a mix of **any** number of **fast vs. slow**, **small vs. big** memories.



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OS-level Memory Management



Physical Address Space

Hybrid Memory Page Allocation



If the memory allocation doesn't change throughout time, then we get no use out of the *fast* memory.

Hybrid Memory Page Scheduling



Page Scheduling as a Prediction Problem



Existing Predictors

OS uses a very simple **history**-based predictor, to minize operational overheads.

Predicts for all pages that page hotness at period p_N = page hotness at period p_{N+1}



Need something more clever, to close this big gap.. How effective would Machine Learning be?

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Lecture Outline:







LSTMs

The Vision: Integrate ML in Page Sceduling



Toward Realizing the Vision: Questions to Ask





Replace the Page Scheduler with a Reinforcement Learning (RL) agent.

Learn the Action: Learn from moving pages across hybrid memory. Learn from mistakes (e.g., cold pages in DRAM).

Why it is not a good fit:

- Exponential Action Space = 2^N , when moving N pages across 2 memories.
- Need to re-train if configuration of hybrid memory changes.
 - Number of memory units.
 - Difference in access speeds / capacities.



Learning Memory Access Patterns

Don't learn the action.. Learn the memory access pattern?



Can we do something similar?

Where to Use ML? (2)



Learning the Page Access Sequence

Learn which pages will be accessed in the next period of time, given a window of history.



Learning the Page Hotness

Learn **how hot** a page will be in the next period of time, given a window of history.

Flip the way you look at a problem!



Page Access Hotness per Period



No. models	Overheads	Accuracy
Many per app. 1 per page.	Smaller models. Parallel Training. Faster to train and fine-tune.	High, because hotness value is smaller, depending on the period length.
0		

Challenge: 1 model per page, means millions of models..

Do all Pages Need ML?



Probably the pages that are part of a pattern, need ML-based management.

For the rest, the simple history page scheduler works well (page hotness at period p_N = page hotness at period p_{N+1}).



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Use ML for subset of pages, and the existing *history* scheduler for the rest.

Proposed ML Integration



Apply ML on a small page subset.



Foundations for practical use of ML.

Carefully select pages for ML.



Application performance boost.

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Lecture Outline:





for Hybrid Memory Management (HMem Management)

Greek Trivia: According to the ancient Greek mythology, Kleio was the muse of history, daughter of Mnemosyne, goddess of memory.





Which Pages to Select?

Need to create a metric to select pages.







Select a **small subset** of pages in the **order** that brings the desired **performance** level.

Page Selector Design

Page Selector calculates internally the perfomance curve, using a performance estimate analytical model.



It is not a lightweight process, but necessary to deliver the desired application performance levels.

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Lecture Outline:







Effect on Application Performance



For half of the applications, Kleio reaches **95%** the possible performance levels!

Effect on Quality of Page Placement



For the pages managed with ML.

Kleio reduces more than 80% of the page misplacements, due to the improved **page movement decisions**, via the more **accurate** page hotness **predictions**.

LSTM Prediction Accuracy



e.g. MAE = 50 means that the RNN predicted on average 50 more accesses per scheduling epoch per page.



High prediction error **does not impact** application performance, when it does not affect the quality of page movement decision.

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for Hybrid Memory Management (HMem Management)

LSTMs

Lecture Outline:



Lessons Learned

1. Understand what would be the benefit from using ML.



Perfect Predictions

2. Learn the Behavior, not the Action, for the use case of hybrid memory management.



Replacing the Page Scheduler with an RL agent would not be practical, nor scalable.

Lessons Learned

3. Look at the same problem from a different angle.





Learn which pages will be accessed next.

Learn **how hot** a page will be next.





4. High prediction error of page hotness **does not impact** application performance, when it does not affect the quality of page movement decision.

Lessons Learned

5. Use existing solutions to the best of their ability, and deploy ML where necessary.



Apply ML on a small page subset.



Foundations for practical use of ML.

Carefully select pages for ML.



Application performance boost.



Can we reduce the number of LSTMs via clustering?

Coeus: Clustering (A)like Patterns for Practical Machine Intelligent Hybrid Memory Management. [CCGrid 2022]



Can we accelerate the page selection process?

Cronus: Computer Vision-based Machine Intelligent Hybrid Memory Management. [MEMSYS 2022]

Report Due April 4 at 18.00

Answer / expand upon these 4 questions:

- 1. What problem is the paper addressing and why is it important?
- 2. How do they approach to solve the problem?
- 3. What are the main evaluation results?
- 4. What are 2 things you will remember from this paper?

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Teaching

🖬 Spring 2023

MLArchSys Seminar Series.

At the MUSS and EMSE Master Programs of the School of Computer Science at Universidad Politécnica de Madrid. *P* MUSS Link *P* EMSE Link Seminar 1: Introduction to Maching Learning for Computer Architecture and Systems. *Research* Sides *Paper* Seminar 2: Maching Learning for Cache Prefetching. *Research* Slides *Paper* Seminar 3: Maching Learning for Hybrid Memory Management. *Research* Slides *Paper*