Ariadne: A Hotness-Aware and Size-Adaptive Compressed Swap Technique for Fast Application Relaunch and Reduced CPU Usage on Mobile Devices

Yu Liang, Aofeng Shen, Chun Jason Xue, Riwei Pan, Haiyu Mao Nika Mansouri Ghiasi, Qingcai Jiang, Rakesh Nadig, Lei Li Rachata Ausavarungnirun, Mohammad Sadrosadati, Onur Mutlu



Available Memory on Mobile Devices is Scarce

- Memory demands of individual mobile applications increase
- Number of concurrently running applications grows
- DRAM capacity cannot be increased accordingly
 - Due to constrained power budget of mobile devices



The available memory is usually **insufficient** to support **multiple applications** running concurrently

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[Android Authority, 2025]

Memory Swap Schemes on Mobile Devices

- When available memory is **insufficient**:
 - SWAP: Swap data to secondary storage (i.e., flash memory)





Memory Swap Schemes on Mobile Devices

- When available memory is insufficient:
 - SWAP: Swap data to secondary storage (i.e., flash memory)



• ZRAM: Compress data and store it in a specific area in DRAM (i.e., Zpool)



ZRAM **increases flash memory lifetime** and **reduces read latency** since decompression latency is shorter than the data swap latency

Executive Summary

Problem: ZRAM prolongs application relaunch latency and **wastes CPU usage** because it does *not* differentiate **data hotness levels** or leverage different **compression chunk sizes** and **data locality**

<u>Goal</u>: To design a **new compressed swap scheme** for mobile devices that **reduces application relaunch latency** and **CPU usage**

Insights:

- Hot data is similar between two consecutive relaunches
- Small-size compression is fast, while large-size compression provides a better compression ratio
- There is locality in data access during application relaunch

Ariadne:

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- Leverages different compression chunk sizes based on data hotness level
- Performs speculative pre-decompression based on data locality

Key Results: Google Pixel 7 with Android 14. Compared to ZRAM, Ariadne

- Reduces application relaunch latency by 50%
- Decreases the CPU usage of compression and decompression procedures by 15%

Open source: <u>https://github.com/CMU-SAFARI/Ariadne</u>

Outline

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Problem and Motivation

New Insights into Mobile Workloads

Ariadne: Hotness-Aware and Size-Adaptive Compressed Swap

Evaluation

Conclusion



Problem 1: Delayed Application Relaunch

• Application relaunch latency in three cases



ZRAM outperforms the traditional SWAP scheme but **increases relaunch latency** 2.1× compared to NON-SWAP

Problem 2: Higher CPU Usage

• **CPU usage of memory reclaim (**i.e., freeing memory) in three cases



ZRAM has **significant CPU usage**

due to compression and decompression operations



Motivation: ZRAM Does NOT Consider Data Hotness

• We examine the fraction of hot, warm, and cold data in each part of compressed data



- Sort all compressed data in the order of compression and then divide it into ten equal parts (X-axis)
- The data in part 0 is the first to be compressed, that in part 9 is the last
- To minimize compression operations, cold data should be compressed earlier (e.g., in parts 0 and 1) and hot data later (e.g., in parts 8 and 9)

Motivation: ZRAM Does NOT Consider Data Hotness

• We examine the fraction of hot, warm, and cold data in each part of compressed data



ZRAM does *not* differentiate between **data hotness levels**, causing **frequent compression and decompression**



To design a **new compressed swap scheme** for mobile devices that reduces **application relaunch latency** and **CPU usage**

Doing so requires reducing the frequency and latency of compression and decompression



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Insight 1: Data Reuse Across Relaunches

• We examine data reuse between two consecutive relaunches of each application



70% data used in first relaunch will be reused in next relaunch

Relaunch the same application could trigger the same activities

Insight 1: Data Reuse Across Relaunches

• We examine data reuse between two consecutive relaunches of each application



Insight 1: Hot data that is used during application relaunch is usually similar between two consecutive relaunches

Key idea 1: Identify hot data based *only* on the most recent relaunch to reduce the hotness identification overhead

 We examine compression performance with different compression chunk sizes



• X-axis represents the compression chunk size



 We examine compression performance with different compression chunk sizes



Large-size compression provides a higher compression ratio



 We examine compression performance with different compression chunk sizes



Small-size compression provides short compression and decompression latencies

 We examine compression performance with different compression chunk sizes



Insight 2: Small-size compression is much faster than large-size but has lower compression ratio

Key idea 2: Use small-size on hot data for low latency and large-size on cold data for high compression ratio



Insight 3: Data Locality in Zpool

• The probability of accessing multiple consecutive pages in Zpool for each evaluated application

Probability of accessing two consecutive pages together

-	Youtube	Twitter	Firefox	GoogleEarth	BangDream
2	0.86	0.81	0.69	0.77	0.61
4	0.72	0.61	0.43	0.54	0.33

Insight 3: There is data use **locality** in Zpool when decompress data during application relaunch

Key idea 3: Pre-decompress the next compressed data to hide decompression latency during application relaunch



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Ariadne: Hotness-Aware and Size-Adaptive Compressed Swap Scheme

- **Key Idea:** Reduce **the frequency and latency** of compression and decompression by
 - Leveraging different compression chunk sizes based on data hotness level
 - Performing speculative pre-decompression based on data locality

- Ariadne: Hotness-Aware and Size-Adaptive Compressed Swap Scheme
 - HotnessOrg: Quickly identifies the hotness of data based on last relaunch
 - AdaptiveComp: Uses different compression chunk sizes based on data hotness
 - Small compression for hot and warm data, leading to **fast decompression**
 - Large compression for cold data, leading to high compression ratio
 - **PreDecomp: pre-decompresses** the next compressed data

Ariadne: Hotness-Aware and Size-Adaptive Compressed Swap Scheme

• Overview







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Evaluation Methodology

- **Platform**: Google Pixel 7
 - Eight core CPU
 - 12 GB DRAM LPDDR5
 - 128 GB SSD
 - Android 14 with Linux 5.10.157

• Evaluated Schemes:

- Baseline: state-of-the-art ZRAM
 - LRU list data organization, page-size compression, no pre-decompression
- Ariadne: with different configurations
 - HotnessOrg, AdaptiveComp, PreDecomp

Open Source: <u>https://github.com/CMU-SAFARI/Ariadne</u>



Effect on Application Relaunch Latency





Effect on Application Relaunch Latency



Ariadne reduces application relaunch latency by 50% (over state-of-the-art ZRAM)



Effect on CPU Usage





Effect on CPU Usage



Ariadne reduces CPU usage of compression and decompression by 15% (over state-of-the-art ZRAM)

There is More in Our Paper

- Design details of Ariadne
- Ariadne implementation
- Evaluation methodology details
- More evaluation results
 - Compression and decompression latency
 - Compression ratio
 - Prediction accuracy and coverage
 - Sensitivity study with different parameters
- Overhead analysis
 - Small overhead in terms of computation and memory space

More in the Extended Version

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Yu Liang[§] Aofeng Shen[§] Chun Jason Xue[‡] Riwei Pan[†] Haiyu Mao[¶] Nika Mansouri Ghiasi[§] Qingcai Jiang^S Rakesh Nadig[§] Lei Li[†] Rachata Ausavarungnirun^{*} Mohammad Sadrosadati[§] Onur Mutlu[§]

> [§]ETH Zürich ^{*}MBZUAI [†]City University of Hong Kong [¶]King's College London ^SUniversity of Science and Technology of China ^{*}MangoBoost

As the memory demands of individual mobile applications continue to grow and the number of concurrently running applications increases, available memory on mobile devices is becoming increasingly scarce. When memory pressure is high, current mobile systems use a RAM-based compressed swap scheme (called ZRAM) to compress unused execution-related data (called anonymous data in Linux) in main memory. This approach avoids swapping data to secondary storage (NAND flash memory) or terminating applications, thereby achieving shorter application relaunch latency.

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their diverse needs [1–3]. To fulfill user expectations of seamless and rapid application relaunch, mobile systems preserve all execution-related data (called *anonymous data* in Linux [4]), such as stack and heap, in main memory. This practice, known as *keeping applications alive in the background* [1, 5–8], enables faster relaunches. However, it also results in significant main memory capacity requirements for each application.

As the demand for memory capacity in mobile applications grows and the number of applications running simultaneously increases, available memory is becoming an increasingly scarce



https://arxiv.org/pdf/2502.12826

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Ariadne leverages different compression chunk sizes based on data hotness level and performs speculative pre-decompression based on data locality

Evaluation results on Google Pixel 7 with Android 14 show that Ariadne reduces *both* application relaunch latency and CPU usage, compared to ZRAM



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Compression Ratio

Compression ratio under different settings





Compression Ratio

Compression ratio under different settings



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Compression and Decompression Latency

Compression and decompression latency under different settings





Prediction Accuray/Coverage

For hot data prediction





- HotnessOrg:
 - Similar to existing LRU scheme, no overhead.
- AdaptiveComp:
 - Pages compressed together may used in different time, but locality shows impact is minimal.
- PreDecomp:
 - May predict data wrongly, so only predict one page based on locality.



Implication on Application Relaunch Latency





Implication on Application Relaunch Latency

