

# The Aurora Operating System

## Revisiting the Single Level Store

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### ABSTRACT

Applications on modern operating systems manage their ephemeral state in memory, and persistent state on disk. Ensuring consistency between them is a source of significant developer effort, yet still a source of significant bugs in mature applications. We present the Aurora single level store (SLS), an OS that simplifies persistence by automatically persisting all traditionally ephemeral application state. With recent storage hardware like NVMe SSDs and NVDIMMs, Aurora is able to continuously checkpoint entire applications with millisecond granularity.

Aurora is the first full POSIX single level store to handle complex applications ranging from databases to web browsers. Moreover, by providing new ways to interact with and manipulate application state, it enables applications to provide features that would otherwise be prohibitively difficult to implement. We argue that this provides strong evidence that manipulation and persistence of application state naturally belong in an operating system.

### CCS CONCEPTS

• **Software and its engineering** → **Operating systems**; • **Computer systems organization** → **Secondary storage organization**; **Reliability**; **Dependable and fault-tolerant systems and networks**; • **Information systems** → **Storage architectures**.

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### KEYWORDS

single level stores, transparent persistence, snapshots, checkpoint/restore

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## 1 INTRODUCTION

Single level storage (SLS) systems provide persistence of applications as an operating system service. Their advantage lies in removing the semantic gap between the in-memory representation and the serialized on-disk representation that uses file IO APIs. This gap often leads to increased code complexity and software bugs [13, 41]. Instead, applications solely use memory and the operating system persists this state to disk. Developers design programs as if they never crash and thus do not write code for persistence and recovery. After a crash, the SLS restores the application, including all state (i.e., CPU registers, OS state, and memory), which continues executing oblivious to the interruption.

SLSes have been impractical to build for decades for performance reasons, but this has changed with the advent of new storage technologies. Past systems suffered from a large performance gap between memory and disks in terms of bandwidth and latency. This was compounded by write-amplification due to the tracking of memory modifications at page granularity, and the overhead of CPU and OS state. Modern flash, coupled with fast PCIe Gen 4–5, has largely closed the performance gap with memory.

We introduce the Aurora Operating System, a novel non-traditional single level storage system that enables persistence and manipulation of execution state. Aurora is based on the FreeBSD kernel and is the first SLS that can run unmodified POSIX applications. Aurora provides persistence at

the granularity of process trees or containers, and supports multi-process applications with nearly all POSIX primitives. This allows for the persistence of complex applications like Firefox, a popular web browser.

Aurora differs from previous systems in several ways. EROS [45] requires application cooperation to achieve performance and its main contributions are optimizing checkpointing and swap for spinning disks. IBM's AS/400 uses runtime and compile time hooks along with application hints to achieve good performance [46].

Aurora revisits the single level store with three main contributions. First, we depend on improvements in hardware to achieve performance and functionality including new flash storage and large virtual address spaces. Second, we develop an architecture designed to support both unmodified and modified POSIX applications. Third, we expand the concept of a single level store with new primitives for the manipulating execution state to enable novel applications.

Aurora also accurately captures application state by treating all POSIX primitives (e.g., Unix domain sockets, System V shared memory, and file descriptors) as first class objects, rather than as parts of a process. This allows Aurora to handle applications composed of processes that share memory or files in arbitrary ways, without duplicating work or leaving edge cases unhandled. Using this approach, Aurora supports complex programs like the Firefox web browser, the RocksDB key value store, and the Mosh remote shell.

Aurora provides a system level service for manipulating arbitrary application state. It goes much further than traditional SLses by blurring the line between applications and data. Users can operate on running applications to persist, copy, revert, or transfer them the same way they would a file. Aurora makes state manipulation an explicit operation, which programs often need to do in an ad hoc manner by themselves. Aurora creates application checkpoints that encapsulate all information required to recreate the application, even across reboots and machines.

We argue that application persistence and manipulation of execution state naturally belong in the operating system, which enables novel applications to modern systems. Aurora allows us to solve a wide range of complex systems problems, from reducing startup times and increasing density of serverless computing, to improving debugging and simplifying database design.

## 2 BACKGROUND

Single level stores have existed for decades both in industry and academia [20, 23, 32, 45, 46]. These systems simplified software development by providing persistence, but were not POSIX compatible and not transparent. Developer effort was required for correctness and performance. They used

incremental checkpointing [51] to persist applications at regular intervals with runtime and/or application specific hooks. These systems were severely limited by the speed of storage devices at the time, e.g., the EROS research OS spent a large effort on masking spinning disk latency [45].

Existing persistent-by-default designs like The Machine [31] and Twizzler [17] are not transparent and depend on special hardware. The Machine was an attempt to build a supercomputer based on memristors, while Twizzler is an OS that uses only NVDIMMs for storage. These systems break compatibility with existing systems in that they depend on the byte addressability of persistent storage. Single level stores like Aurora conversely use regular DRAM and disks, and hide the distinction between the two from the application.

Aurora makes a key observation that device bandwidth and latency has improved to rival the memory bus. Modern CPUs can provide an aggregate PCIe bandwidth up to 256 GB/s, more than that of memory [16]. New Intel 3D XPoint SSD's reduce IO latency to 10  $\mu$ s [8], within two orders of magnitude of memory. The combination of high bandwidth and low latency makes transparent persistence possible without needing byte addressability.

Popular checkpoint/restore mechanisms have been used for scientific computing to recover from failures and migrate workloads [29, 42, 47]. These systems do not checkpoint frequently enough to provide transparent persistence and the resulting checkpoints are not self contained.

Checkpointing of virtual machines (VM) has enjoyed a lot of popularity and applications. VMs package the application and all dependencies into a portable checkpoint. Live migration and incremental checkpointing have enabled distributed resource management, fault tolerance and other applications [5, 24, 36, 38].

Containers, which have less overhead than virtualization, have traditionally lacked these features. Providing the same functionality for containers enables the same distributed resource management and fault tolerance applications. Systems like CRIU [6], the standard for Linux container migration [40], piece together application state by querying the kernel through system calls and the proc file system.

While CRIU's performance is tolerable for migration, its overheads are prohibitive for other applications including transparent persistence. Even research systems that optimize memory checkpointing in CRIU have failed to reduce overheads enough [50]. Furthermore, CRIU is incredibly complex, requiring 7 years to properly add UNIX socket support [7].

Aurora is different from previous systems in two ways. First, rather than checkpointing objects exposed at the system call boundary it provides persistence throughout the OS. This includes internal kernel state, the file system, and the virtual memory subsystem. Second, Aurora treats all POSIX primitives as first class objects, persisting and replicating the

POSIX object model as seen by the kernel. Aurora’s design enables broad application support comparable to CRIU with a far simpler design.

*Serverless Computing and Scale Out:* In the serverless paradigm, developers write small stateless *functions* that run in the cloud. Invoking a function involves creating a new container or VM and starting the application, an operation that adds significant latency.

Frameworks and language based approaches have reduced invocation latencies by colocating functions [26, 52], or by restoring initialized application checkpoints. Catalyzer [25] achieves submillisecond restore times from in memory checkpoints using a combination of virtualization and kernel based techniques unique to the gVisor application kernel.

Aurora’s restore mechanism enables autoscaling and fast starts using OS containers. Aurora’s restore times from disk rival the state of the art because of lazy restores (see § 3) and cooperative warm ups (see § 4). The object store also deduplicates otherwise unrelated checkpoints on disk for higher storage density.

*Debugging and Speculation:* Record/replay systems [21, 27, 28, 33, 35, 44, 49] record non-deterministic inputs to an application to replay the complete execution. Recordings can be large and checkpointing has been a common technique for bounding the storage overhead. Aurora’s low overhead checkpointing makes record/replay practical in production, enabling developers to capture an application moments before a crash.

Limited forms of checkpoint/restore are used for application level speculation, which restores application memory to a prior state. This operation has been proposed as an OS service implemented using hardware virtualization for speculative execution and efficient backtracking [19]. Roll-backs have also been used for application security, namely intraprocess isolation [34].

*Databases:* Databases [10–12] and key-value stores [9, 14, 15] often spend a significant amount of their code managing persistence and paging of database state. The `fsync` and `msync` calls have subtle semantic issues depending on hardware and software configuration leading to data loss bugs in even mature projects like LevelDB [1–4, 41] and PostgreSQL [13]. Databases also implement memory tracking in software, which duplicates the work of the hardware MMU and the operating system.

Aurora gives developers fine-grained control over persistence behavior for modified applications. An area of ongoing work is developing a richer API for applications to communicate with the OS to further optimize paging and persistence.

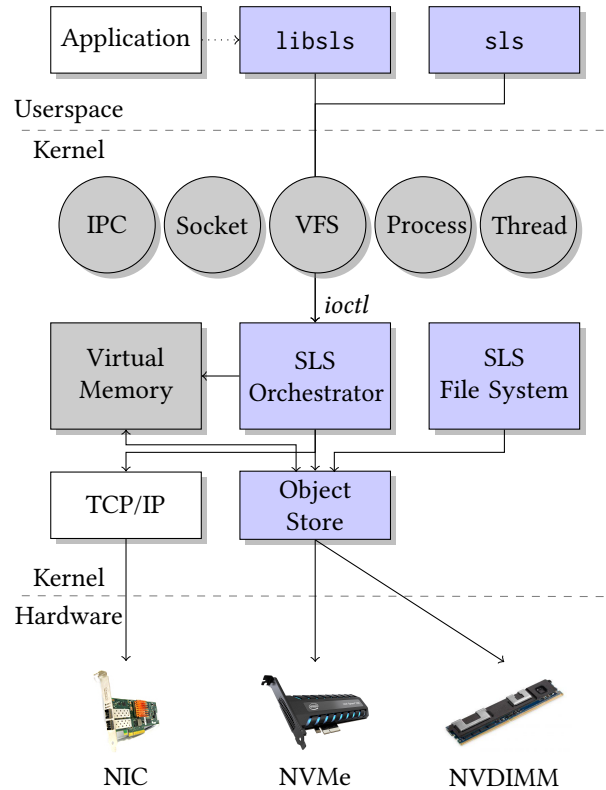


Figure 1: Basic system diagram

### 3 THE AURORA OPERATING SYSTEM

Figure 1 shows the architecture of the Aurora persistent operating system. Aurora has three components: the SLS orchestrator, the object store, and a custom file system. Each POSIX object in the operating system (e.g., socket pairs, POSIX shared memory or System V message queues) contains code that continuously serializes and stores the state in the object store. Each object is serialized independently, and contains enough user and kernel state to recreate the object on restore.

The SLS orchestrator maps kernel objects to the on-disk store and manages the checkpoint and resume operations. Aurora provides persistence for individual processes, process trees or containers. The orchestrator provides serialization barriers across the entire OS to provide consistent application-wide checkpoints. All processes are momentarily paused and remaining unflushed state is copied into memory buffers or tracked using copy-on-write (COW). These updates are flushed asynchronously to disk. In our current prototype this occurs up to 100× per second with modest overhead. The orchestrator also manages restores by recreating all POSIX objects and resuming applications.

Aurora uses an optimized and custom COW mechanism for its applications. The standard COW scheme used by fork breaks shared memory semantics by forcing each process within an application to create their own private copy on a write. The OS normally prohibits marking shared pages as COW to avoid this problem. Aurora instead modifies FreeBSD’s Mach derived VM subsystem [43] to create a new page shared between all processes on a copy-on-write fault, while Aurora flushes the original page.

Aurora optimizes checkpoints by only flushing dirtied data on every checkpoint. It implements *incremental checkpointing* using our custom COW mechanism that tracks writes between successive checkpoints. It thus never flushes the same page twice for shared memory or COW memory regions. On restore, Aurora faithfully reproduces the entire memory hierarchy to preserve page deduplication.

The object store simplifies synchronizing memory and file system checkpoints. The snapshot operations of popular file systems are too slow to keep up with the SLS orchestrator. The object store does hundreds of checkpoints per second using a lower overhead COW layout than that of WAFL [30] and ZFS [18] that snapshot less frequently. The COW layout enables in-place garbage collection without needing to rewrite incremental checkpoints.

Applications are placed into a *persistence group* attached to one or more backing devices. Locally persistent applications are backed by NVMe flash or NVDIMMs when available. Applications can also be remotely persisted through the network backend to another host. For debugging and speculative execution applications can use a local memory backend to store ephemeral checkpoints. Aurora allows for attaching multiple backends at the same time, e.g., sending an application’s incremental checkpoints to both a local disk and a remote machine for replication.

The Aurora file system provides a file API into the object store that enables functionality and fixes limitations in POSIX file systems. Users can create zero copy snapshots and clones of a container including process and file system state. The file system must handle special edge cases that are not normally considered in a typical POSIX file system.

An example of an edge case is unlinked but open files (i.e., anonymous files). In POSIX file systems, these files would be reclaimed after a crash, preventing application restoration. We solve this by maintaining an on-disk *open reference count* storing the number of persistent virtual file system vnodes.

Integrating swap with Aurora optimizes restores by lazily paging application memory. Aurora restores the minimal application state, including the OS and CPU state, with memory effectively swapped out. Applications fault in their working set during execution. Aurora uses the clock page replacement algorithm [22] to optimize restore by eagerly paging in the

Command	Description
sls persist	Add an application to a persistence group
sls attach	Attach a persistence group to a backend
sls detach	Detach a persistence group from a backend
sls checkpoint	Checkpoint an application
sls restore	Restore an application from an image
sls ps	List applications in Aurora
sls send	Send an application to a remote
sls recv	Receive an application from a remote

**Table 1: A subset of the command line interface available to users.**

Function	Description
sls_checkpoint()	Create an image
sls_restore()	Restore a checkpoint
sls_rollback()	Roll back state to last checkpoint
sls_ntflush()	Non-temporal flush (outside checkpoint)
sls_barrier()	Wait for a checkpoint to be flushed
sls_mctl()	Include/exclude memory regions
sls_fdctl()	Enable/disable external consistency

**Table 2: A subset of the API available to developers from the Aurora library.**

hottest pages to avoid excessive page faults. When pages are swapped out due to memory pressure they are incorporated into the subsequent checkpoint.

### 3.1 Command Line Interface

To help illustrate Aurora we show the usage of a subset of the command line interface as shown in Table 1. Users begin by enabling transparent persistence of an application using the `persist` command. This registers all OS state that is part of the application with the orchestrator. By default the application is persisted 100× per second. The host and each container have their own persistence group. The `attach` and `detach` commands allow the user to register backends with persistence groups.

Users can manually create named checkpoints with the `checkpoint` command and view all application checkpoints in Aurora using `ps`. Users can restore to a point in time or resume execution of a persistence group after a system crash using `restore`. Aurora uses free space on-disk to provide a short execution history as incremental checkpoints.

Users can easily share or migrate applications using the `send` and `recv` commands to serialize a checkpoint state or continually feed incremental checkpoints to a remote host. Flags to these commands allow the user to pipe a single checkpoint to a file to give to another user, live migrate the application, or provide fault tolerance.

### 3.2 Aurora API

Table 2 shows the Aurora API that custom applications use to control and optimize persistence. Applications can manually initiate checkpoints, restore, or roll back checkpoint state.

Applications control persistence using `sls_barrier` as a persistence barrier for a given thread and `sls_ntflush` to initiate a low latency flush of an append-only log to a storage medium. These applications require custom code during restore to repair data structures based on the log.

Applications can also control the behavior of memory and other resources. Using `sls_mctl`, applications can selectively include/exclude memory regions and hint to Aurora the best policy for lazy restore. Developers can control the external consistency per file descriptor using `sls_fdctl`.

External consistency [39] is enforced when a communication spans a persistence group boundary (including remote hosts). Any data transmitted on a file descriptor are buffered until the corresponding checkpoint is persisted on disk to prevent other machines from seeing state that could be lost in a crash. If the remote application can handle observing such state, the developer can disable external consistency to improve latency.

## 4 APPLICATIONS OF AURORA

In this section we show how Aurora’s abstractions enable very different applications including serverless computing and scale out, debugging and speculation, and databases.

*Serverless Computing and Scale Out.* Aurora can be used to optimize serverless warm starts using its lazy restore, combined with its ability to distribute and scale function runtimes. Serverless function warm starts are similarly implemented by restoring a container with the function already initialized. Scaling out amounts to repeatedly restoring an already checkpointed application.

Aurora’s COW design maximizes function density in persistent storage by deduplicating shared runtime memory between different functions. The object store represents each function as a small delta over the runtime container’s checkpoint. All functions share this data, allowing machines to potentially hold billions of functions.

Similar functions share unmodified pages, increasing function density and memory efficiency. This sharing causes instances to *warm each other up*: an instance faulting a page into memory shares it with the rest using COW. Recent research [48] shows that the working set of many workloads is almost identical between invocations, leading Aurora to eliminate major faults for popular functions.

*Debugging and Speculation.* Aurora can also provide improvements to application debugging. Aurora creates periodic checkpoints of a running application that can later be

inspected with a debugger or executed. We can use this to build a type of time travel debugger [37] or, since new incremental checkpoints leave old ones intact, to bisect the history to find violations of invariants. Repeatedly restoring from the same image can uncover nondeterministic failures that do not manifest on every execution. We regularly used this while developing Aurora itself.

Aurora integrates with record/replay systems to bound record log size by only keeping the records since the last checkpoint. On a failure, the application is rolled back to this checkpoint and replays the remaining log. Developers can thus witness the last seconds before a crash on a production machine with a very small disk and CPU overhead compared to standalone RR.

Aurora’s rollback primitive allows apps to implement speculative execution for increased performance. For example, a client sending data to a server can execute assuming that the server received it, saving a round trip’s worth of time. If the transfer ends up failing, the client rolls back to before it sent the data. Aurora notifies the client of the rollback, allowing it to try a more conservative code path.

*Databases.* Aurora aims to provide a clean memory management interface between database applications and the OS. Databases communicate application and workload specific memory management policies to Aurora, and use checkpointing to trigger data transfers to and from storage. Aurora allows applications to checkpoint data without associated execution state, providing an explicit persistence primitive that does not suffer from the semantic complexities of file and memory syncing.

Aurora’s APIs provide a drop in replacement for common persistence mechanisms found in key value stores. We use Aurora’s persistent log (`sls_ntflush`), manual checkpoints (`sls_checkpoint`) and barriers (`sls_barrier`), to replace existing persistence mechanisms in RocksDB that uses a log structured merge tree and Redis that uses fork for checkpoints with a write ahead log. In the case of Redis our initial port is already faster with less code.

## 5 PRELIMINARY RESULTS

Aurora is a fully working system that we continue to develop, comprising ~19k SLOC on top of FreeBSD 12.1. We have been testing the system on a dual 2.1 GHz Intel Xeon Silver 4116 CPU (Skylake-SP) with 96 GiB of RAM, four Intel Optane 900P NVMe drives and an Intel X722 10 GbT network interface card. We continue to explore how to improve APIs for different use cases and improve performance by optimizing various system components.

We have evaluated Aurora’s fitness for the use cases we outlined by measuring the checkpoint and restore times of two workloads: Redis, a key value store, with a working set

Checkpoint	Full	Incremental
Metadata copy	267.9 $\mu s$	239.7 $\mu s$
Lazy data copy	5145.9 $\mu s$	711.1 $\mu s$
Application stop time	5413.8 $\mu s$	950.8 $\mu s$

**Table 3: Stop time breakdown when checkpointing a Redis instance with a 2 GiB working set. Aurora flushes data in the background concurrently with application execution.**

of 2 GiB, and a smaller hello world application. Redis represents heavier applications that use Aurora for persistence or debugging purposes, and the hello world app represents serverless functions.

We checkpointed Redis both by copying its entire address space, and by using incremental checkpointing. The results in Table 3 show that, while the cost of grabbing metadata is the same in both cases, lazily copying the address space using COW is 7 $\times$  faster with incremental checkpointing, for a total stop time below 1 ms. In neither case does Redis stop to wait for the data to reach storage, due to Aurora’s external consistency semantics.

Application overhead includes the stop time for each checkpoint and the cost of servicing COW faults while the application runs. Most of the stop time is spent applying COW tracking through page table manipulations. Checkpointing frequency is bounded by the speed with which Aurora can flush incremental checkpoints to disk.

Next, we restored the same Redis instance after bringing the image to memory, as we would do when using Aurora for debugging. According to the results in Table 4, restoring takes around 755  $\mu s$ , two thirds of which are spent recreating the address space. No memory is copied, since Aurora uses COW semantics to share pages between the image and the running application.

Finally, we measured the potential improvement in serverless function startup time by restoring a small workload both from memory and disk. Table 4 includes the results, which show that the only extra latency when restoring from backing storage is due to bringing in the function metadata from the object store. In both cases, restore times are well under a millisecond. Restoring metadata state for disk restores is slightly faster, because reading in the checkpoint implicitly restores some application state.

Restore Backend	Redis Memory	Serverless Memory	Serverless Disk
Object Store Read	N/A	N/A	322.7 $\mu s$
Memory state	494.4 $\mu s$	144.6 $\mu s$	122.6 $\mu s$
Metadata state	261.1 $\mu s$	240.4 $\mu s$	206.9 $\mu s$
Total latency	755.5 $\mu s$	454.4 $\mu s$	652.2 $\mu s$

**Table 4: Restore time breakdown for a Redis instance with a 2 GiB working set from an in-memory image, and a serverless workload from an in-memory image and from disk. Reading data from the object store implicitly restores some state, causing lower memory and metadata restore times.**

## 6 CONCLUSION

We present Aurora, the first modern single level store, and the first to provide a complete POSIX interface. Aurora’s performance characteristics show that, due to recent storage hardware advances, SLSes are once again a practical approach to OS design. SLSes provide both a simple and elegant solution for application and data persistence, and a flexible tool for manipulating execution state in new and interesting ways.

We have applied Aurora to a wide variety of domains, including serverless computing and scale out, debugging and speculation, and databases. In these domains, the combination of simple but efficient persistence and extremely fast checkpoints and restores allows us to provide both performance benefits and new functionality while reducing application complexity. Many of its benefits are available to applications without any modifications at all; additional benefits are available through a simple but powerful API.

We believe that Aurora will enable creating new and complex applications with a fraction of the effort they would otherwise need, and will pave the way for new research into application/OS co-design.

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