State & State Management

Apache Flink[®] Tuning & Troubleshooting Training



As you scale up

- You are more likely to need RocksDB, especially for incremental checkpointing
- Cleaning up stale state becomes more important
- Checkpointing becomes more likely to fail/timeout
- You are more likely to hit rate limits enforced by external services
- It becomes more worthwhile to be smart about recovery



State Backends

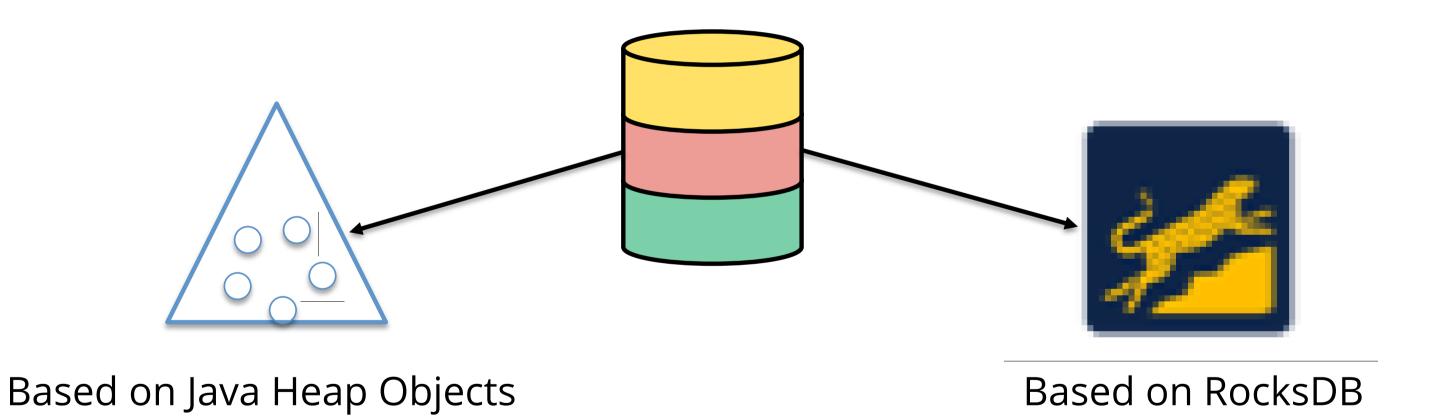
Task Manager memory layout



Task Manager memory layout



State backends





Heap-based State Backend



Recap: Heap-based state backend

- State lives as Java objects on the heap
- Organized as chained hash table, key → state
- One hash table per registered state descriptor
- Supports asynchronous state snapshots
- Data is de / serialized only during state snapshot and restore
- Highest throughput



Performance considerations

- Choose TypeSerializers with efficient copy-methods
- Flag immutability of objects where possible to avoid needing copies
- GC choice / tuning
- Scale out using multiple task managers per node



RocksDB



RocksDB keyed state backend

- State lives as serialized byte-strings in off-heap memory and on local disk
- Key / value store, organized as a log-structured merge tree (LSM tree)
- Key: serialized bytes of <keygroup, key, namespace>
- LSM naturally supports MVCC (multi-version concurrency control)
- Data is de / serialized on every read and update

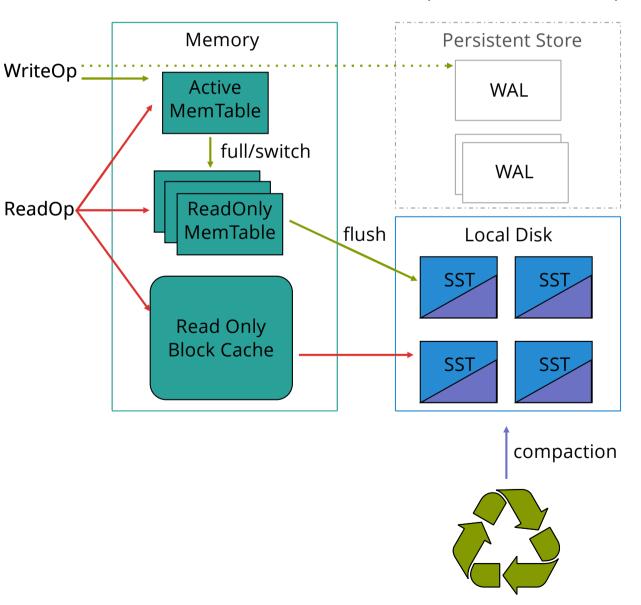


High level architecture

- Keys and values: arbitrary byte streams, limited to at most 2³¹ bytes
- Organization
 - memtable: new writes are inserted into this in-memory structure
 - logfile: new writes are optionally also written to this sequentially written WAL file
 - not used by Flink
 - SST file: full memtables are flushed to an SST file (sorted by key for fast lookups)
 - SST = static sorted table
 - immmutable: once written, never modified

In Flink:

- WAL disabled
- persistence via checkpoints





RocksDB uses log-structured merge-trees

Observations

- Sequential disk access is faster than randomly accessing RAM
- Sequential disk access is much, much faster than random disk access
- Simple, append-only logs take advantage of this, but can't provide efficient key-based access

SST files

- Each contains a small, chronological, sorted subset of changes
- Since they are immutable, duplicate entries are created as records are updated or removed
- Reads start in the memtables, then the sstfiles are checked, in order from newest to oldest
- sstfiles are compacted and merged

Optimizations

Indexes and Bloom filters can be used to reduce the effort to locate keys within sstfiles



RocksDB resource consumption

- One RocksDB instance per operator subtask
- One column family per registered state
- SST files, MemTable's, block caches, compaction threads per column family
- state.backend.rocksdb.block.cache-size
 - amount of cache for data blocks (8MB)
- state.backend.rocksdb.writebuffer.size
 - max size of a MemTable (4MB)
- state.backend.rocksdb.writebuffer.count
 - maximum number of MemTable's allowed in memory before flushing to SST file (2)



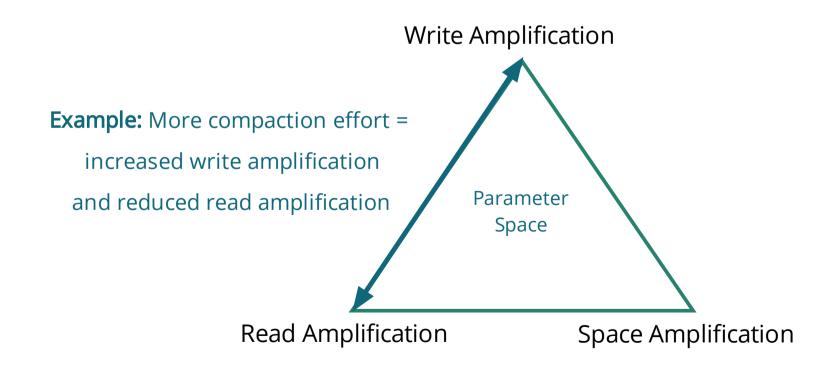
RocksDB resource consumption(2)

- One RocksDB instance per operator subtask
- One column family per registered state
- SST files, MemTable's, block caches, compaction threads per column family
- Indexes and Bloom filters
 - optional, via ConfigurableOptionsFactory
 - see PredefinedOptions#SPINNING_DISK_OPTIMIZED_HIGH_MEM for an example
- Table Cache
 - caches open file descriptors to SST files
 - default: unlimited
- Shared compaction threads, via ConfigurableOptionsFactory
 - DBOptions#setEnv(Env.getDefault()) see FLINK-10198



RocksDB performance tuning

- Write amplification: bytes written to storage / bytes written to the DB
- Read amplification: disk reads / query
- Space amplification: size of database files / data size





Predefined options

- These profiles are predefined collections of RocksDB settings
- Select one of these profiles via state.backend.rocksdb.predefinedoptions
 - DEFAULT
 - SPINNING_DISK_OPTIMIZED
 - SPINNING_DISK_OPTIMIZED_HIGH_MEM
 - FLASH_SSD_OPTIMIZED
- Look at the PredefinedOptions class to see what these do
- See the docs if you want to do your own fine-grained tuning



RocksDB metrics

- Some RocksDB native metrics can be forwarded to Flink's metric reporter(s)
- Can provide valuable insight for tuning
- Enabling native metrics can degrade performance, so proceed with caution
- See the docs for details



Timers and RocksDB

- By default, timers are stored on the heap
- If you don't have many timers, leave them on the heap
- For timers in RocksDB:
 - state.backend.rocksdb.timer-service.factory: rocksdb
- Note: If you use RocksDB for keyed state and have timers on the heap, the timers will be snapshotted during the synchronous portion of the snapshot.
 - see FLINK-10026



Incremental checkpointing

- Only the RocksDB state backend offers incremental checkpointing
- Flink observes the SST files created/deleted since the last checkpoint
- These changes are (carefully) mirrored into Flink's checkpoint store
- Expected trade-off: faster* checkpoints, slower recovery
- Creates write amplification (upload compacted SST files to eventually prune checkpoint history)
 - Sum of all increments can be larger than the full state size
- No rebuild required: simply re-open the RocksDB backend from the SST files
- SST files are snappy-compressed by default



General Performance Considerations

- Use efficient TypeSerializer's and serialization formats
- Decompose user code objects
 - ValueState<List<Integer>> → ListState<Integer>
 - ValueState<Map<Integer, Integer>> → MapState<Integer, Integer>
- Flatten POJOs / avoid deep objects
 - Reduces object overheads and avoids following references
- Use the correct configuration for your hardware setup
- Consider enabling RocksDB native metrics to profile your applications
- File Systems
 - Working directory on fast storage, ideally local SSD. Could even be memory.
 - EBS performance can be problematic. Shares network connection.



Expiring State



Be careful about indefinite state retention

It's easy to either implicitly or explicitly require Flink maintain unbounded state, e.g.:

```
DataStream<Tuple2<Long, Float>> maximums = input
    .keyBy(0)
    .maxBy(1);
```

To clean up expired state:

- Use a ProcessFunction to hold the state, and register a Timer
- Use State Time-To-Live (next slides!)
- If using Flink SQL, configure an idle state retention time



State TTL: cleanup

- Expired values are removed when they are read, e.g., by calling ValueState.value()
- You can avoid restoring expired state from full state snapshots via

```
StateTtlConfig ttlConfig = StateTtlConfig
    .newBuilder(Time.seconds(1))
    .cleanupFullSnapshot()
    .build();
```



State TTL: generic cleanup in the background

```
StateTtlConfig = StateTtlConfig
    .newBuilder(Time.seconds(1))
    .cleanupInBackground()
    .build();
```

This will activate the default background cleanup (if any exists) for the current state backend.



State TTL: incremental cleanup

Heap-based state backend only

```
StateTtlConfig = StateTtlConfig
    .newBuilder(Time.seconds(1))
    .cleanupIncrementally(10, true)
    .build();
```

- 10: the number of state entries to check per state access (default: 5)
- true: whether to also trigger cleanup on a per-record basis (default: false)
- Cleanup depends on records being accessed or processed
- Incremental cleanup increases processing latency



State TTL: clean-up during RocksDB compaction

```
StateTtlConfig ttlConfig = StateTtlConfig
    .newBuilder(Time.seconds(1))
    .cleanupInRocksdbCompactFilter(1000)
    .build();
```

- 1000: RocksDB will query Flink for the current timestamp after checking each batch of 1000 entries
 - Smaller values will lead to more timely expirations and slower compactions
- This feature requires enabling RocksDB compaction filtering
 - state.backend.rocksdb.ttl.compaction.filter.enabled
- This feature slows down RocksDB compaction



State TTL: caveats

- Using state TTL increases the consumption of state storage
- Specifying TTL in event time is not supported
- Adding or removing a StateTtlConfig to an existing state descriptor is not supported
 - will cause a StateMigrationException



Working with Checkpoints

(Especially at Large Scale)



Snapshots, Checkpoints, and Savepoints

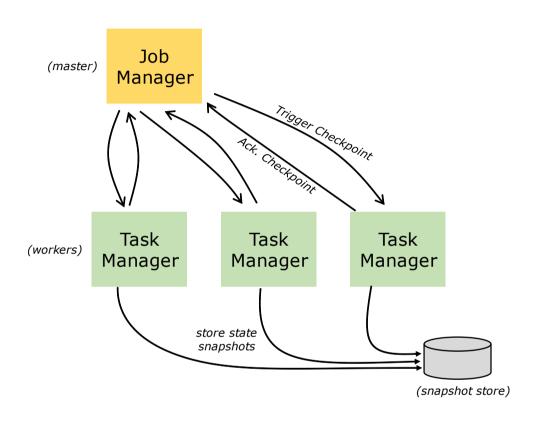
- Snapshot is the generic term; checkpoints and savepoints are snapshots
- Broadly speaking:
 - Checkpoints are managed by Flink, intended for
 - failure recovery
 - Savepoints are user-triggered and are retained indefinitely, intended for
 - rescaling
 - upgrades

Current differences:

	user-controlled	incremental	rescaling	unified format
checkpoints	Maybe	Yes	Maybe	No
savepoints	Yes	No	Yes	Yes



Checkpoints require global coordination



- At scale, the JM eventually becomes a bottleneck
- May need to increase checkpoint timeout and/or increase slots/TM and reduce # of TMs



Checkpoint configuration, basics

- Exactly-once vs at-least-once (i.e., barrier alignment on/off)
- Incremental vs full (incremental only w/ RocksDB)
- Checkpoint interval
 - Has a big impact on recovery time
 - Also has a big impact on latency for committing transactions



Checkpoint configuration, continued

- Number of retained checkpoints (default: 1)
- env.getCheckpointConfig().setCheckpointTimeout(n) (default: 10 minutes)
- Compression (default: off)
 - RocksDB SST files are already compressed (used in incremental snapshots)
- Fail/continue on checkpoint errors (default: fail)
- Number of concurrent checkpoints (default 1)
- Minimum pause between checkpoints

StreamExecutionEnvironment.getCheckpointConfig().setMinPauseBetweenCheckpoints(milliseconds)



External (retained) checkpoints

```
CheckpointConfig config = env.getCheckpointConfig();
config.enableExternalizedCheckpoints(ExternalizedCheckpointCleanup.RETAIN_ON_CANCELLATION);
```

- By default, checkpoints are deleted when a program is cancelled.
- Example above will retain them instead
 - Cheap / incremental "savepoint" but in internal state backend format
- Location is specified in
 - config state.checkpoints.dir: hdfs://checkpoints/
 - code env.setStateBackend(new RocksDBStateBackend("hdfs:///checkpointsdata/"))
- You can resume from a retained checkpoint (just as with a savepoint)

```
$ bin/flink run -s :checkpointMetaDataPath [:runArgs]
```



Why checkpoints might fail (timeout)

- Timeout is simply too short
 - very large cluster
 - very large state (non-incremental checkpoints)
 - high rate of state change (incremental checkpoints)
 - checkpoint storage and/or network are slow or unavailable
- Backpressure is preventing checkpoint barriers from advancing fast enough
 - many possible causes



Scenario 1

- Computing an event time join on multiple sources
- One source is lagging far behind the other(s)
- Leading to excessive buffering in Flink state, and large checkpoints



Scenario 2

- Many timers firing simultaneously
- Operator isn't reading from its inputs while triggering on Timer calls
- Checkpoint barriers delayed



Scenario 3

- An operator makes synchronous calls to an external service
 - e.g., a flatMap that calls a REST API
- Everything is fine under normal conditions
- But when the external service is under heavy load, it responds more slowly
- Leading to intermittent backpressure and checkpoint failures



Checkpoint Counts

Triggered: 71 In Progress: 1 Completed: 70 Failed: 0 Restored: 0

Latest Completed Checkpoint ID: 70 | Completion Time: 16:59:12 | End to End Duration: 11ms | State Size: 9.55 KB

Checkpoint Detail: Path: <checkpoint-not-externally-addressable> Discarded: true

Operators: Name **Acknowledged Latest Acknowledgment End to End Duration State Size Buffered During Alignment** 261 B 0 B Source: test data 1/1 (100%) 16:59:12 8ms + sawTooth 4/4 (100%) 16:59:12 9ms 1.02 KB 0 B + assignKey(temp) 4/4 (100%) 16:59:12 9ms 0 B 0 B + sineWave 4/4 (100%) 16:59:12 0 B 0 B 9ms + assignKey(pressure 16:59:12 0 B 0 B 4/4 (100%) 10ms + 0 B squareWave 4/4 (100%) 16:59:12 0 B + 9ms assignKey(door) 0 B 0 B 4/4 (100%) 16:59:12 10ms + 0 B + Sink: sensors-sink 4/4 (100%) 16:59:12 10ms 0 B window 4/4 (100%) 16:59:12 11ms 8.28 KB 0 B

		End to End Duration			State Size		Checkpoint Duration (Sync)		Checkpoint Duration (Async)		Alignment Buffered		Alignment Duration			
Mi	nimum	10ms		2.0)1 KE	}	Oms			0ms		0 B		2m	S	
Av	erage	10ms		2.0	D7 KE	}	Oms			Oms		0 B		3m	S	
Ma	aximum	11ms		2.	17 KE	3	Oms			Oms		0 B		4m	S	
I D	Acknow Time	<i>l</i> ledgement		E2E Duratio	n	\$	State Size	\$	Checkpoint Durat	tion	Checkpoint Duratio (Async)	n 💠	Align Buffered	\$	Align Duration	\$
1	16:59:1	2		10ms			2.17 KB		Oms		Oms		0 B		2ms	
2	16:59:1	2		10ms			2.01 KB		Oms		Oms		0 B		3ms	
3	16:59:1	2		11ms			2.01 KB		Oms		Oms		0 B		4ms	
4	16:59:1	2		10ms			2.09 KB		Oms		0ms		0 B		3ms	

11ms

16:59:12

4/4 (100%)

window

8.28 KB

0 B

Latest Failed Checkpoint

ID: 9 **Failure Time:** 15:36:08 **Cause:** The job has failed.

Checkpoint Detail: Path: -Discarded: -Failure Message: The job has failed.

perators:									
	Name	Acknowledged	Latest Acknowledgment	End to End Duration	State Size	Buffered During Alignment			
+	Source: test data	1/1 (100%)	15:36:08	3ms	261 B	0 B			
+	sawTooth	1/1 (100%)	15:36:08	12ms	260 B	0 B			
+	assignKey(temp)	1/1 (100%)	15:36:08	12ms	0 B	0 B			
+	sineWave	1/1 (100%)	15:36:08	11ms	0 B	0 B			
+	assignKey(pressure	1/1 (100%)	15:36:08	13ms	0 B	0 B			
+	squareWave	1/1 (100%)	15:36:08	11ms	0 B	0 B			
+	assignKey(door)	1/1 (100%)	15:36:08	15ms	0 B	0 B			
+	Sink: sensors-sink	0/1 (0%)	n/a	n/a	0 B	0 B			
+	window	1/1 (100%)	15:36:08	16ms	2.37 KB	0 B			
+	Sink: summed-sensors-sink	0/1 (0%)	n/a	n/a	0 B	0 B			

Diagnosing checkpoint timeouts

Look for clues in the logs and/or web UI:

- end_to_end_duration synchronous_duration asynchronous_duration
 - this is how long it look each subtask to begin checkpointing
 - high values typically indicate constant backpressure
 - constant backpressure → application is under provisioned
- amount of data buffered during alignment and alignment duration
 - ideally these are small; large values indicate trouble
- unbalanced / asymmetric behavior
 - do some subtasks have much more state or signs of backpressure than others?



Heavy alignments

- Heavy alignment means different load on different paths
- E.g., could be caused by
 - skewed window emission (one node with some hot keys)
 - GC stall of one subtask
- Might want to set min-time-between-checkpoints if problem is severe
- Unaligned checkpoints (planned for Flink 1.10) should resolve this issue
 - see FLIP-76



Reducing the likelihood of checkpoint failures

- Don't implement operators that can block (e.g., don't do synchronous i/o)
- Reduce latency
- Upgrade to Flink 1.10 when it arrives
 - FLIP-27 and FLINK-10886 will support event time alignment between sources
 - FLIP-76: unaligned checkpoints will allow checkpoint barriers to overtake in-flight data, and make the in-flight data part of the checkpoint
 - goal is to make checkpointing immune to backpressure



Avoid DDOSing other systems

Scenario:

- Job with 5000 stateful subtasks
- Checkpoint interval: 1 second
- State size: KBs per subtask
- Problem: S3 will block off connections after exceeding 1000s of requests/sec
- Solution: reduce FS stress for small state
 - by increasing the value of state.backend.fs.memory-threshold (default: 1KB)
 - state chunks smaller than this value are stored inline in the checkpoint metadata file

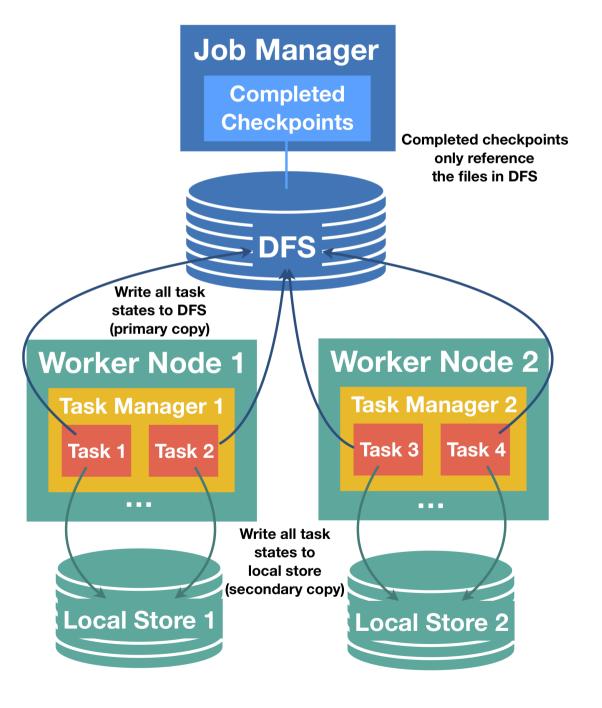


Configuring Recovery



Local recovery: setup and cost

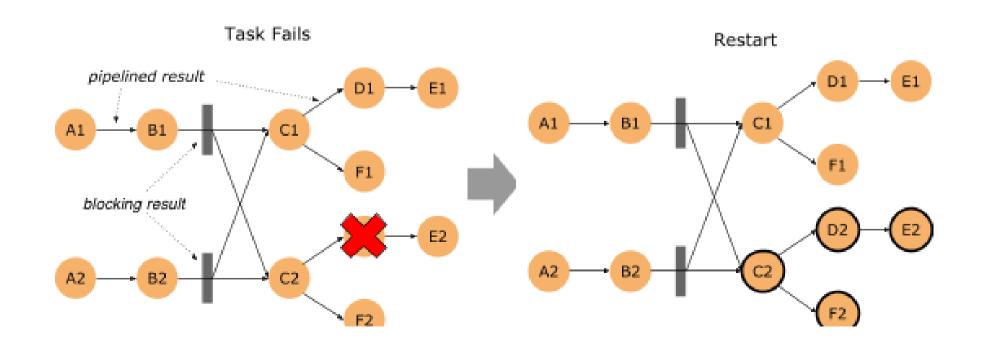
- How to configure:
 - state.backend.local-recovery
 - disabled by default
 - taskmanager.state.local.root-dirs
 - where the files used for local recovery are stored
- Impact:
 - MemoryStateBackend: not supported
 - FsStateBackend: duplicate writes + storage
 - RocksDBStateBackend:
 - incremental checkpoints: no cost
 - full checkpoints: duplicate writes + storage





Fine-grained recovery (batch jobs)

- Rather than cancelling all tasks and restarting the whole job, recovery can be limited to only those tasks in the same *failover region*
- Regions are disjoint sets of tasks connected via pipelined data exchanges
- Set jobmanager.execution.failover-strategy to "region" (default in 1.9)
- Set ExecutionMode in ExecutionConfig to BATCH





Fine-grained recovery (streaming jobs)

- Set jobmanager.execution.failover-strategy to "region" (default in 1.9)
- Only applies to embarrassingly parallel jobs
 - no keyBy()
 - no rebalance

See FLIP-1: Fine Grained Recovery from Task Failures for details.



Wrap-up; we've looked at

- A high-level overview of how RocksDB is organized, and how to optimize/tune it
- Expiring state with state TTL
- How to approach checkpoint failures
- Tuning recovery



