Tectonic-Shift: A Composite Storage Fabric for Large-Scale ML Training

ATC'23

Background

- > Machine learning (ML) has been very successful and widely used in industry
- The key in machine learning is model training based on massive amounts of data
 - An efficient and scalable data storage infrastructure becomes an important part of the training process



Background

> Tectonic: a prior storage fabric for Meta before this paper



Problems & Goals

- > Increasing demand for IO for training tasks
- > HDD storage nodes in Tectonic can't meet growth IO demands
 - HDDs have low storage cost overhead but high IO overhead (about 10x over storage cost)
- Goals: exploring a new file system for training tasks
 - IO efficient
 - User-transparent
 - Scalable

Main Idea

- Present a new storage fabric Tectonic-Shift for Meta's production machine learning training infrastructure
 - Add a cache layer called Shift based on Tectonic, achieving efficient IO
 - Provides the same interface for training tasks, achieving user transparency
 - Completely independent of the Tectonic and consists of a number of cache nodes named Shift Node, achieving high scalability



Data Reading Process



Design1: Cache

Uses both DRAM and SSD for data storage

- Just using DRAM tends to make the network interface card an IO bottleneck
- SSD only have about 20% the IO overhead compared to HDD



Design2: Policy

- > Shift will make additional judgments when loading data and evicting data
 - Just accept hot data into cache and reinsert part of them when eviction
- Rationality stems from 3 reasons
 - 1. SSD and DRAM have greater storage overhead (about 5x) compared to HDD
 - 2. Jobs share frequently accessed data and also access some unique data, resulting in cache pollution
 - 3. Reinsertion further reduces HDD IO and improves hardware lifetime



Design2: Policy

Using bucket to manage data

- A bucket is a set of data, cache makes one policy for a bucket
- Increase granularity to reduce management overhead

Shift designs 2 policy: history and future

- 1. History: use the historical access frequency of the data in the bucket to determine
- 2. Future: depending on each job's schedule for data reading



Workload	Jobs & Partitions Read	Description
Synchronized	$\{P_1, P_2, P_3\}_1, \{P_4, P_5, P_6\}_2, \\ \{P_1, P_2, P_3\}_3, \{P_7, P_8, P_9\}_4, \\ \{P_1, P_2, P_3\}_5$	Multi-tenant HP tuning or ex- ploratory jobs. Jobs are launched synchronously.
Pipelined	$\{P_1, P_2, P_3\}_1, \{P_2, P_3\}_2, \{P_3\}_3$	Long-running, pipelined jobs. Jobs are launched synchronously.
Sequential	$\{P_1\}_1, \{P_1\}_2, \{P_1\}_3, \{P_1\}_4, $ $\{P_2\}_5, \{P_3\}_6, \{P_1\}_7, \{P_4\}_8, $ $\{P_5\}_9$	Queued jobs that launch when training capacity is available. Jobs 1-3, 4-6, and 7-9 launch together.

Workload	Jobs & Partitions Read	
Synchronized	$\{P_1, P_2, P_3\}_1, \{P_4, P_5, P_6\}_2, \{P_1, P_2, P_2\}_2, \{P_7, P_8, P_0\}_4$	
	$\{P_1, P_2, P_3\}_5$ (P / , P / , P)	
Pipelined	$\{P_1, P_2, P_3\}_1, \{P_2, P_3\}_2, \{P_3\}_3$	
Sequential	${P_1}_1, {P_1}_2, {P_1}_3, {P_1}_4,$	
${P_2}_5, {P_3}_6, {P_1}_7, {P_4}$		
	${P_5}_9$	



(a) Synchronized workload

 All new policies is better since they do not load P7 P8 P9 for job4



(b) Pipelined workload

- Future is better since it just keeps P3 in cache
- Historic is worse since it load P2 and evict P3 when doing job1



(c) Sequential workload

 All new policies is better since they do not load P2 and P5

	Normalized Hit Rate	Normalized HDD IO
Hybrid with Reinsertion	1.03	0.82



Conclusion

- High IO requirements for training tasks cause existing data systems to become unsuitable
- The authors explore and propose a new file system Tectonic-Shift, a file system builds on Tectonic and improves IO by cache
- Uses DRAM and SSD to store cached data, realizing high IO while saving storage costs
- Introduces admission and reinsertion on top of the existing common cache structure to further improve cache utilization