ECE 473/573 Cloud Computing and Cloud Native Systems Lecture 08 Transaction Log

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September 10, 2025

Outline

Transaction Log

Implementing a Transaction Log File

Reading Assignment

- ► This lecture: 5
- Next lecture: 4

Outline

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Implementing a Transaction Log File

Services as Finite State Machines

- Computations can be modeled as finite state machines (FSMs)
- Networked services like microservices
 - React to requests received via the network.
 - Update internal data structures and objects as needed.
 - Generate responses to be sent via the network.
- Services as FSMs
 - State: data model stored in data structures and objects.
 - Initial state: initial values of variables and objects.
 - Input: requests
 - Output: responses
 - State transitions: function and method calls

Persisting Resource State

- Objective: allow applications and services to start from where they were, after being shutdown.
 - In particular unexpected shutdown due to faults and failures.
- Delegate to another service that will be able to handle persistence.
 - E.g. a database service that supports the data model.
 - A good choice in practice but doesn't answer the fundamental problem.
- ► Make use of persistent storage devices
 - E.g. hard drives and SSDs where only binary blocks are supported.
 - A more fundamental problem we need to study today.

Persisting Resource State as Binary Blocks

- Option 1: Direct State Storage
 - Encode data structures and objects into a binary format that can be decoded later.
 - Intuitive but require efforts to design algorithms for individual data structures and objects.
- ► Option 2: Transaction Log
 - Store all requests as binary data in the order of their arrival.
 - Compute state from the initial state and the stored requests.
 - ► To encode requests is usually simple since they are just names of functions and methods plus their arguments.

Performance Considerations

- Storage devices are slow.
 - Maximum throughput can only be achieved by sequential reads and writes – storage devices are able to optimize for such cases.
 - Random accesses are limited by latency, resulting in much smaller available throughput.
- Direct State Storage
 - Random access to the binary data is required to avoid encoding and saving the whole state every time there is an update.
 - Need to reduce random accesses not easy.
- Transaction Log
 - ► To store requests as they arrive only requires sequential writes.
 - ▶ To compute the state requires only sequential reads.
 - Nevertheless, to store all requests may require a lot of storage, and to read and process them may require a lot of time.

Scalability Considerations

- Size and throughput of storage services can be improved by horizontal scaling.
 - Replication improves read throughput by making data available from multiple servers.
 - Sharding improves write throughput by partitioning data into different servers.
- Sharding is usually not quite difficult.
- ► For replication,
 - Direct State Storage
 - Too costly to replicate the whole state frequently.
 - ► How to only replicate updates?
 - ► Transaction Log
 - Replicate requests by forwarding them to other servers.
 - Each server can then compute the state by themselves.

Resilience Considerations

- Possible faults and failures.
 - Hardware failure causing loss of data.
 - Power failure in the middle of saving binary data.
- Replication helps to resolve issues of loss of data.
 - ▶ But replication won't help if it corrupts data.
- ► For power failures,
 - Direct State Storage
 - ▶ If there is a power failure when updating the binary data, then it is very difficult to tell what data is changed.
 - ▶ This may lead to data corruption that cannot be repaired.
 - ► Transaction Log
 - Storing new requests only requires to append data and will not overwrite existing data for past requests.
 - If there is power failure, either the new request is stored successfully or there is some extra data at the end that can be detected and removed without much efforts – data corruption can be avoided.

Discussions

- Transaction log provides better scalability and resilience.
- Transaction log helps troubleshooting.
 - ▶ Making it possible to reproduce all system transactions.
- Restarting a service using transaction log may take more time than that using direct state storage.
 - Need time to read and process all past requests to compute the current state.
- Practical solutions combine the two options to make trade-offs.
 - As we will discuss for distributed database systems.

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Transaction Log File for Key-Value Store

- ► To support two operations Put, Delete.
 - ▶ There is no need to record Get as it doesn't change the state.
- ▶ File format
 - Each request is encoded into a line.
 - Each line contains four fields delimited by tabs.
 - Sequence number: monotonically increasing to represent the order of arrival.
 - Event type: PUT or DELETE
 - Key
 - Value: for PUT only.
- Additional considerations.
 - ► Key/Value cannot contain tabs or newline characters.
 - A line at the end of the file without a newline character indicating a corrupted line that should be removed.

Transaction Logger

```
type TransactionLogger interface {
  WriteDelete(key string)
  WritePut(key, value string)
  Err() <-chan error
  ReadEvents() (<-chan Event, <-chan error)
  Run()
}</pre>
```

- An interface to support transaction log.
- ▶ WriteDelete and WritePut record requests.
- ▶ ReadEvents reads past requests when the service restarts.
 - Communication through channels: <-chan Event is a channel of Events where past requests can be read out.
 - ► Reduce memory usage by <u>not</u> reading and storing all past requests at the same time.
- Run the logger in its own threads with channels.
 - Avoid racing conditions from multiple RESTful requests without using locks.

Implementing File Based Transaction Logger

```
type FileTransactionLogger struct {
   events chan<- Event // Write-only channel for sending events
   errors <-chan error // Read-only channel for receiving errors
   lastSequence uint64 // The last used event sequence number
   file *os.File // The location of the transaction log
}
func NewFileTransactionLogger(filename string) (TransactionLogger, error) {
   file, err := os.OpenFile(filename, os.O_RDWR|os.O_APPEND|os.O_CREATE, 0755)
   ...
   return &FileTransactionLogger{file: file}, nil
}</pre>
```

- Implement FileTransactionLogger to store requests in file
 - Lowercase members are private.
 - Members not explictly initialized are set to nil or 0.
 - Need to implement the 5 methods from the TransactionLogger interface.
- We will omit error handling to focus on functionalities when necessary.

Read Past Requests

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```
func (1 *FileTransactionLogger) ReadEvents() (<-chan Event, <-chan error) {</pre>
  scanner := bufio.NewScanner(1.file) // Create a Scanner for 1.file
  outEvent := make(chan Event) // An unbuffered Event channel
  outError := make(chan error, 1) // A buffered error channel
  go func() {
    var e Event
    defer close(outEvent) // Close the channels when the
    defer close(outError) // goroutine ends
    for scanner.Scan() {
      line := scanner.Text()
      if err := fmt.Sscanf(line, "%d\t%d\t%s\t%s",
        &e.Sequence, &e.EventType, &e.Key, &e.Value); err != nil {
        outError <- fmt.Errorf("input parse error: %w", err)</pre>
        return
      1.lastSequence = e.Sequence // Update last used sequence #
      outEvent <- e // Send the event along, block if channel is full
  1()
  return outEvent, outError
```

► Send event e to channel outEvent by outEvent <- e.

Write Requests to File

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```
func (1 *FileTransactionLogger) WritePut(key, value string) {
  1.events <- Event{EventType: EventPut, Key: key, Value: value}</pre>
func (1 *FileTransactionLogger) WriteDelete(key string) {
  1.events <- Event{EventType: EventDelete, Key: key}</pre>
func (1 *FileTransactionLogger) Run() {
  1.events = make(chan Event, 16) // Make an events channel
  1.errors = make(chan error, 1) // Make an errors channel
  go func() { // start a goroutine that runs in a single thread
    for e := range l.events { // Retrieve the next Event
      1.lastSequence++ // Increment sequence number
      _, err := fmt.Fprintf(l.file, "%d\t%d\t%s\t%s\n",
        1.lastSequence, e.EventType, e.Key, e.Value)
```

- ► Thread confinement: multiple threads may call WritePut and WriteDelete but only a single thread will handle them.
 - Synchronization via a channel without a lock.

Initialization

```
var logger TransactionLogger
func initializeTransactionLog() error {
  logger, err := NewFileTransactionLogger("transaction.log")
  events, errors := logger.ReadEvents()
  e, ok := Event{}, true
  for ok && err == nil {
    select { // use select to read from multiple channels
    case err, ok = <-errors: // Retrieve any errors</pre>
    case e. ok = <-events:
      switch e.EventType {
      case EventDelete: // Got a DELETE event!
        err = Delete(e.Key)
      case EventPut: // Got a PUT event!
        err = Put(e.Key, e.Value)
  logger.Run()
  return err
func main() {
  err := initializeTransactionLog()
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```

Recording PUT Requests

```
func keyValuePutHandler(w http.ResponseWriter, r *http.Request) {
  vars := mux.Vars(r)
  key := vars["key"]
  value, err := ioutil.ReadAll(r.Body)
  defer r.Body.Close()
  ...
  err = Put(key, string(value))
  ...
  logger.WritePut(key, string(value))
  w.WriteHeader(http.StatusCreated)
  log.Printf("PUT key=%s value=%s\n", key, string(value))
}
```

▶ DELETE is recorded in a similar way.

Summary

Use transaction logs to store states indirectly for better scalability and resilience.