

Principles of Software Construction: Objects, Design and Concurrency

Data Consistency

15-214 toad

Spring 2013

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Administrivia

- Homework 6 due tonight
 - With late days, as late as Saturday at 11:59 p.m.
 - Office hours:
 - Christian: Th 4:30 6 p.m. and by appt.
 - Alex: Th 6 10 p.m.
 - Charlie: F 1:30 3 p.m. and by appt.
- Final exam Monday, May 13th 5:30 8:30 p.m.
 - Review session TBD

Key topics from Tuesday



The course roadmap

- Object-oriented programming
- Software design
- Complex systems
- Process and tools to manage software complexity



Today: Data consistency and concurrency control

- A formal definition of consistency
- Introduction to transactions
- Introduction to concurrency control
- Distributed concurrency control
 - Two-phase commit

An aside: Double-entry bookkeeping

 A style of accounting where every event consists of two separate entries: a credit and a debit

```
void transfer(Account fromAcct, Account toAcct, int val) {
    fromAccount.debit(val);
    toAccount.credit(val);
static final Account BANK LIABILITIES = ...;
void deposit(Account toAcct, int val) {
    transfer(BANK LIABILITIES, toAcct, val);
boolean withdraw(Account fromAcct, int val) {
    if (fromAcct.getBalance() < val) return false;</pre>
    transfer(fromAcct, BANK LIABILITIES, val);
    return true;
```

Some properties of double-entry bookkeeping

- Redundancy!
- Sum of all accounts is static
 - Can be 0

Data consistency of an application

- Suppose $\mathcal D$ is the database for some application and ϕ is a function from database states to $\{true, false\}$
 - We call φ an *integrity constraint* for the application if $\varphi(\mathcal{D})$ is true if the state \mathcal{D} is "good"
 - We say a database state $\mathcal D$ is consistent if $\phi(\mathcal D)$ is true for all integrity constraints ϕ
 - We say ${\mathcal D}$ is inconsistent if $\phi({\mathcal D})$ is false for any integrity constraint ϕ

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 - We say ${\mathcal D}$ is inconsistent if $\phi({\mathcal D})$ is false for any integrity constraint ϕ
- E.g., for a bank using double-entry bookkeeping one possible integrity constraint is:

```
def IsConsistent(D):
    If ( sum(all account balances in D) == 0 ):
        Return True
    Else:
        Return False
```

Database transactions

- A transaction is an atomic sequence of read and write operations (along with any computational steps) that takes a database from one state to another
 - "Atomic" ~ indivisible
- Transactions always terminate with either:
 - Commit: complete transaction's changes successfully
 - Abort: undo any partial work of the transaction

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```
boolean withdraw(Account fromAcct, int val) {
    begin_transaction();
    if (fromAcct.getBalance() < val) {
        abort_transaction();
        return false;
    }
    transfer(fromAcct, BANK_LIABILITIES, val);
    commit_transaction();
    return true;</pre>
```

A functional view of transactions

- A transaction \mathcal{T} is a function that takes the database from one state \mathcal{D} to another state $\mathcal{T}(\mathcal{D})$
- In a correct application, if \mathcal{D} is consistent then $\mathcal{T}(\mathcal{D})$ is consistent for all transactions \mathcal{T}

A functional view of transactions

- A transaction \mathcal{T} is a function that takes the database from one state \mathcal{D} to another state $\mathcal{T}(\mathcal{D})$
- In a correct application, if \mathcal{D} is consistent then $\mathcal{T}(\mathcal{D})$ is consistent for all transactions \mathcal{T}
 - E.g., in a correct application any serial execution of multiple transactions takes the database from one consistent state to another consistent state

Database transactions in practice

- The application requests commit or abort, but the database may arbitrarily abort any transaction
 - Application can restart an aborted transaction
- Transaction ACID properties:

Atomicity: All or nothing

Consistency: Application-dependent as before

Isolation: Each transaction runs as if alone

Durability: Database will not abort or undo work of

a transaction after it confirms the commit

Concurrent transactions and serializability

 For good performance, database interleaves operations of concurrent transactions

Concurrent transactions and serializability

- For good performance, database interleaves operations of concurrent transactions
- Problems to avoid:
 - Lost updates
 - Another transaction overwrites your update, based on old data
 - Inconsistent retrievals
 - Reading partial writes by another transaction
 - Reading writes by another transaction that subsequently aborts
- A schedule of transaction operations is serializable if it is equivalent to some serial ordering of the transactions
 - a.k.a. linearizable

Concurrency control for a centralized database

- Two-phase locking (2PL)
 - Phase 1: acquire locks
 - Phase 2: release locks
- E.g.,
 - Lock an object before reading or writing it
 - Don't release any locks until commit or abort

Concurrency control for a distributed database

- Distributed two-phase locking
 - Phase 1: acquire locks
 - Phase 2: release locks
- E.g.,
 - Lock all copies of an object before reading or writing it
 - Don't release any locks until commit or abort
- Two new problems:
 - Distributed deadlocks are possible
 - All participants must agree on whether each transaction commits or aborts

Two-phase commit (2PC)

• Two roles:

Coordinator: for each transaction there is a unique server

coordinating the 2PC protocol

Participants: any server storing data locked by the

transaction

• Two phases:

Phase 1: Voting (or Prepare) phase

Phase 2: Commit phase

Failure model:

- Unreliable network:
 - Messages may be delayed or lost
- Unreliable servers with reliable storage:
 - Servers may crash or temporarily fail
 - Will eventually recover persistently-stored state

The 2PC voting phase

- ullet Coordinator sends canCommit? (\mathcal{T}) message to each participant
 - Messages re-sent as needed
- Each participant replies yes or no
 - May not change vote after voting
 - Must log vote to persistent storage
 - If vote is yes:
 - Objects must be strictly locked to prevent new conflicts
 - Must log any information needed to successfully commit
- Coordinator collects replies from participants

The 2PC commit phase

- If participants unanimously voted yes
 - Coordinator logs commit(T) message to persistent storage
 - Coordinator sends doCommit(T) message to all participants
 - Participants confirm, messages re-sent as needed
- If any participant votes no
 - Coordinator sends doAbort(T) message to all participants
 - Participants confirm, messages re-sent as needed

2PC time sequence of events

"done"

Coordinator: Participants: "prepared" canCommit? "prepared" (persistently) yes "uncertain" (objects still "committed" locked) (persistently) doCommit confirmed "committed"

Problems with two-phase commit?



Problems with two-phase commit?

- Failure assumptions are too strong
 - Real servers can fail permanently
 - Persistent storage can fail permanently
- Temporary failures can arbitrarily delay a commit
- Poor performance
 - Many round-trip messages

The CAP theorem for distributed systems

- For any distributed system you want...
 - Consistency
 - Availability
 - tolerance of network Partitions
- ...but you can support at most two of the three