Epidemic Protocols

- Used in Bayou system from Xerox PARC
- Bayou: weakly connected replicas
  - Useful in mobile computing (mobile laptops)
  - Useful in wide area distributed databases (weak connectivity)
  - Eventually-consistent systems use similar approaches to propagate updates
- Based on theory of epidemics *(spreading infectious diseases)*
  - Upon an update, try to "infect" other replicas as quickly as possible
  - Pair-wise exchange of updates (*like pair-wise spreading of a disease)*
  - Terminology:
    - Infective store: store with an update it is willing to spread
    - Susceptible store: store that is not yet updated
- Many algorithms possible to spread updates

Spreading an Epidemic

- Anti-entropy
  - Server *P* picks a server *Q* at random and exchanges updates
  - Three possibilities: only push, only pull, both push and pull
- Rumor mongering (aka *gossiping*)
  - Upon receiving an update, *P* tries to push to *Q*
  - If *Q* already received the update, stop spreading with prob *1/k*
  - Analogous to "hot" gossip items => stop spreading if "cold"
  - Does not guarantee that all replicas receive updates
    - Chances of staying susceptible (no update received): $s = e^{-(k+1)(1-s)}$

Removing Data

- Deletion of data items is hard in epidemic protocols
- Example: server deletes data item *x*
  - No state information is preserved
    - Can't distinguish between a deleted copy and no copy!
- Solution: death certificates
  - Treat deletes as updates and spread a death certificate
    - Mark copy as deleted but don't delete
    - Need an eventual clean up
      - Clean up dormant death certificates

CAP Theorem

- Conjecture by Eric Brewer at PODC 2000 conference
  - It is impossible for a web service to provide all three guarantees:
    - **Consistency** (nodes see the same data at the same time)
    - **Availability** (node failures do not the rest of the system)
    - **Partition-tolerance** (system can tolerate message loss)
  - A distributed system can satisfy any two, but not all three, at the same time
- Conjecture was established as a theorem in 2002 (by Lynch and Gilbert)

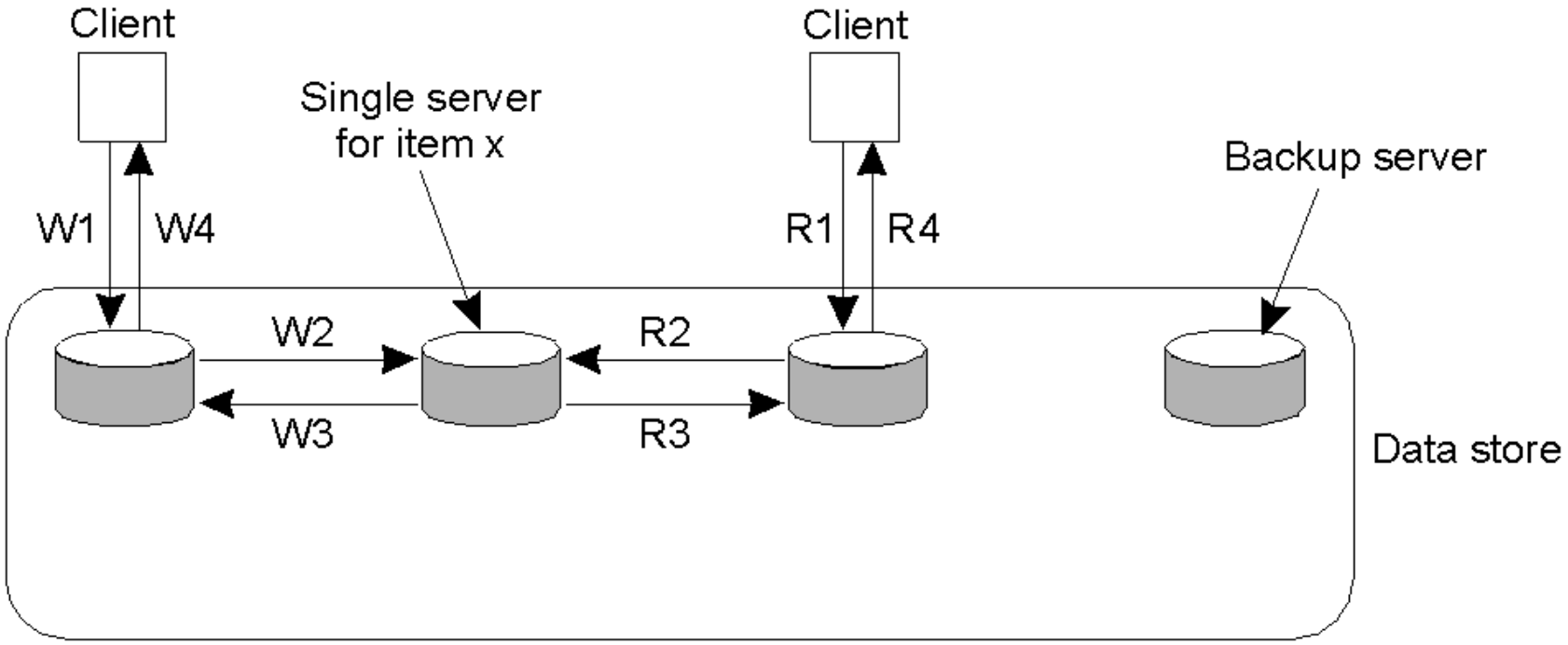
CAP Theorem Examples

- Consistency + Availability
  - Cannot ensure availability with message loss
  - Eg: Single database, distributed database
- Consistency + Partition tolerance
  - Trivial: system does nothing (not available)
  - Many CA systems become unavailable with partitions but they are *always* Consistent: they transition from CA to CP but they never have all three CAP properties at the same time
- Availability + Partition tolerance
  - Allow writes even if consistency cannot be achieved
  - E.g.: Cassandra, Web caching, DNS

Implementation Issues

- Two techniques to implement consistency models
  - Primary-based protocols
    - Assume one single primary replica for each data item
    - Primary applies writes first and then propagates updates
  - Replicated write protocols
    - No primary is assumed for a data item
    - Writes can take place at any replica

Remote-Write Protocols

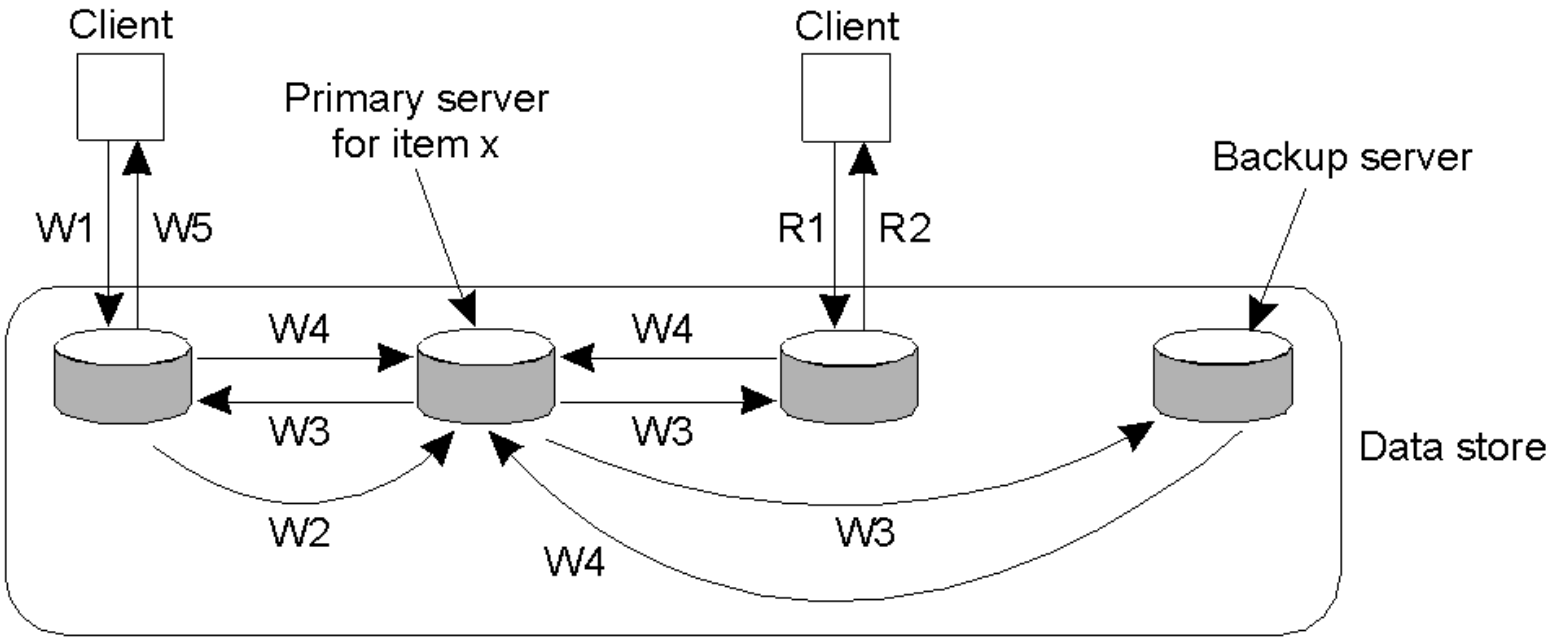


W1. Write request
W2. Forward request to server for x
W3. Acknowledge write completed
W4. Acknowledge write completed

R1. Read request
R2. Forward request to server for x
R3. Return response
R4. Return response

- **Traditionally used in client-server systems (no replication)**

Remote-Write Protocols (2)

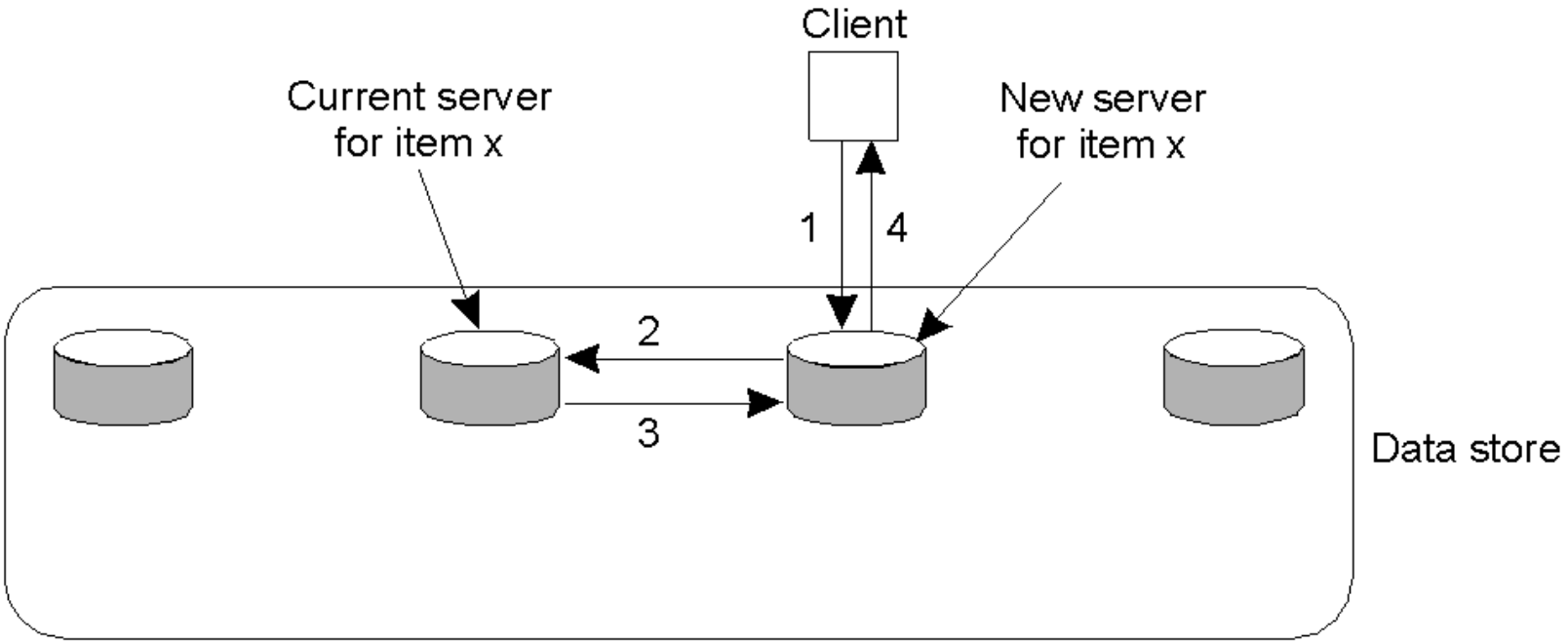


W1. Write request
W2. Forward request to primary
W3. Tell backups to update
W4. Acknowledge update
W5. Acknowledge write completed

R1. Read request
R2. Response to read

- Primary-backup protocol
  - Allow local reads, sent writes to primary
  - Block on write until all replicas are notified
  - Implements sequential consistency

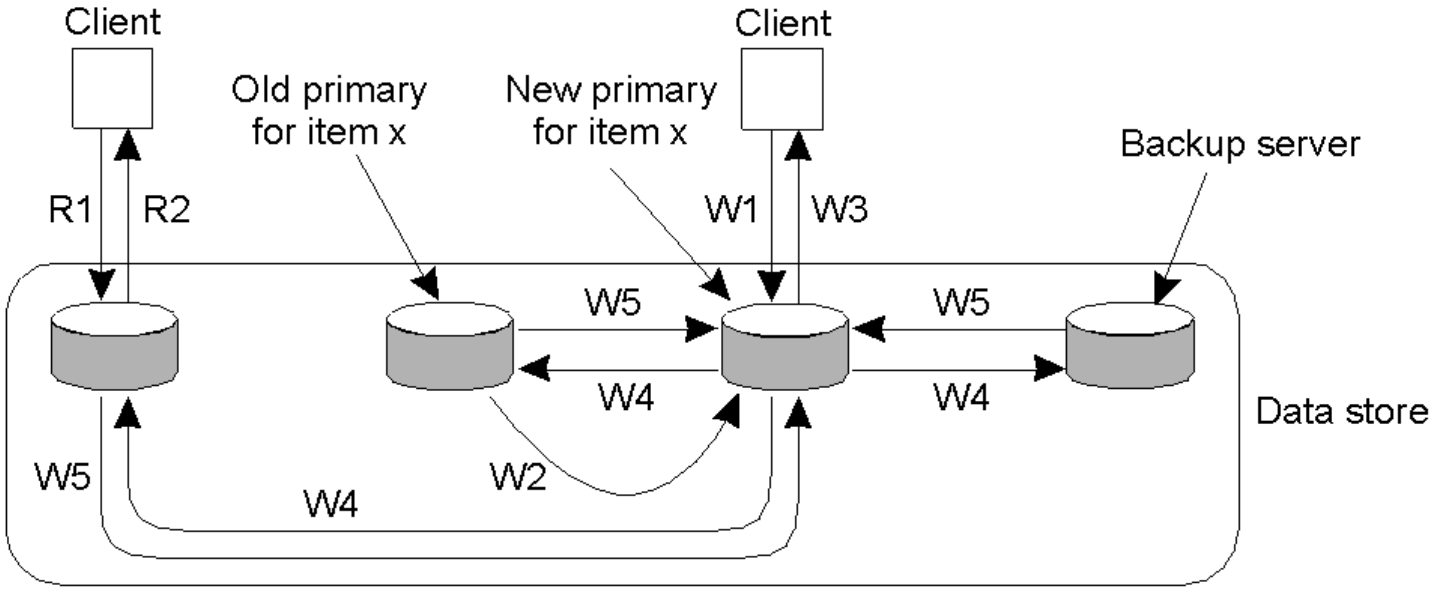
Local-Write Protocols (1)



1. Read or write request
2. Forward request to current server for x
3. Move item x to client's server
4. Return result of operation on client's server

- Primary-based local-write protocol in which a single copy is migrated between processes.

Local-Write Protocols (2)



W1. Write request
W2. Move item x to new primary
W3. Acknowledge write completed
W4. Tell backups to update
W5. Acknowledge update

R1. Read request
R2. Response to read

- Primary-backup protocol in which the primary migrates to the process wanting to perform an update

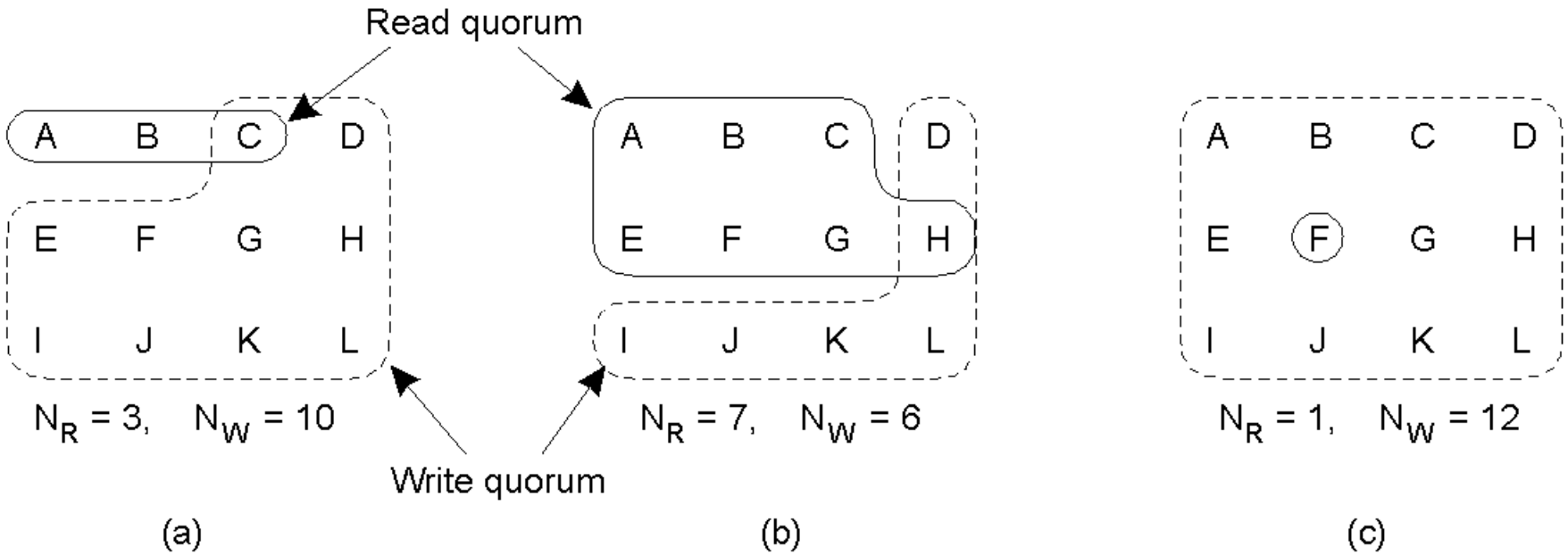
Active Replication

- No primary, any replica is allowed to update
  - Relax the assumption of one primary
- Support for operations, not only updates
  - Example: x = x+1
  - Output depends on current state + operation
- Consistency is more complex to achieve
  - Need to serialize updates
  - Q: Possible approaches to do that?

Quorum-Based Protocols

- Use voting to request/acquire permissions from replicas
- Consider a file replicated on *N* servers
  - $N_R+N_W > N$ $\quad\quad$ $N_W > N/2$
- Update: contact N_W servers and get them to agree to do update (associate version number with file)
- Read: contact N_R and obtain version number
  - Read latest version

Gifford's Quorum-Based Protocol



Replica Management

- Replica server placement
  - Web: geographically skewed request patterns
  - Where to place a proxy?
    - K-clusters algorithm
- Permanent replicas versus temporary
  - Mirroring: all replicas mirror the same content
  - Proxy server: on demand replication
- Server-initiated versus client-initiated

Content Distribution

- CDN: network of proxy servers
- Caching:
  - Update versus invalidate
  - Push versus pull-based approaches
  - Stateful versus stateless
- Q: Web caching: what semantics to provide?

Issue	Push-based	Pull-based
State at server	List of client replicas and caches	None
Messages sent	Update (and possibly fetch update later)	Poll and update
Response time at client	Immediate (or fetch-update time)	Fetch-update time

Final Thoughts

- Replication and caching improve performance in distributed systems
- Consistency of replicated data is crucial
- Many consistency semantics (models) possible
  - Need to pick appropriate model depending on the application
  - Example: web caching: weak consistency is OK since humans are tolerant to stale information (can reload browser)
  - Implementation overheads and complexity grows if stronger guarantees are desired

Fault Tolerance

- Single machine systems
  - Failures are all or nothing
    - OS crash, disk failures
- Distributed systems: multiple independent nodes
  - Partial failures are also possible (some nodes fail)
- *Question:* Can we automatically recover from partial failures?
  - Important issue since probability of failure grows with number of independent components (nodes) in the systems
  - Prob(failure) = Prob(Any one server fails)=$1-(1-p)^N$
    - p: probability that one server fails in a time interval

A Perspective

- Computing systems are not very reliable
  - OS crashes frequently (Windows), buggy software, unreliable hardware, software/hardware incompatibilities
  - Growing popularity of Internet/World Wide Web
    - "Novice" users
    - Need to build more reliable/dependable systems
  - Example: what is your TV (or car) broke down every day?
    - Users don't want to "restart" TV or fix it (by opening it up)
- Need to make computing systems more reliable
  - Important for online banking, e-commerce, online trading, webmail…

Basic Concepts

- Need to build *dependable* systems
- Requirements for dependable systems
  - Availability: system should be available for use at any given time
    - Can be expressed as probability that a request is successful
    - 99.999 % availability (five 9s) => very small down times
  - Reliability: system should run continuously without failure
  - Safety: component failures should not result in an overall catastrophic failure
    - Example: computing systems controlling an airplane, nuclear reactor
  - Maintainability: a failed system should be easy to repair

Basic Concepts (contd)

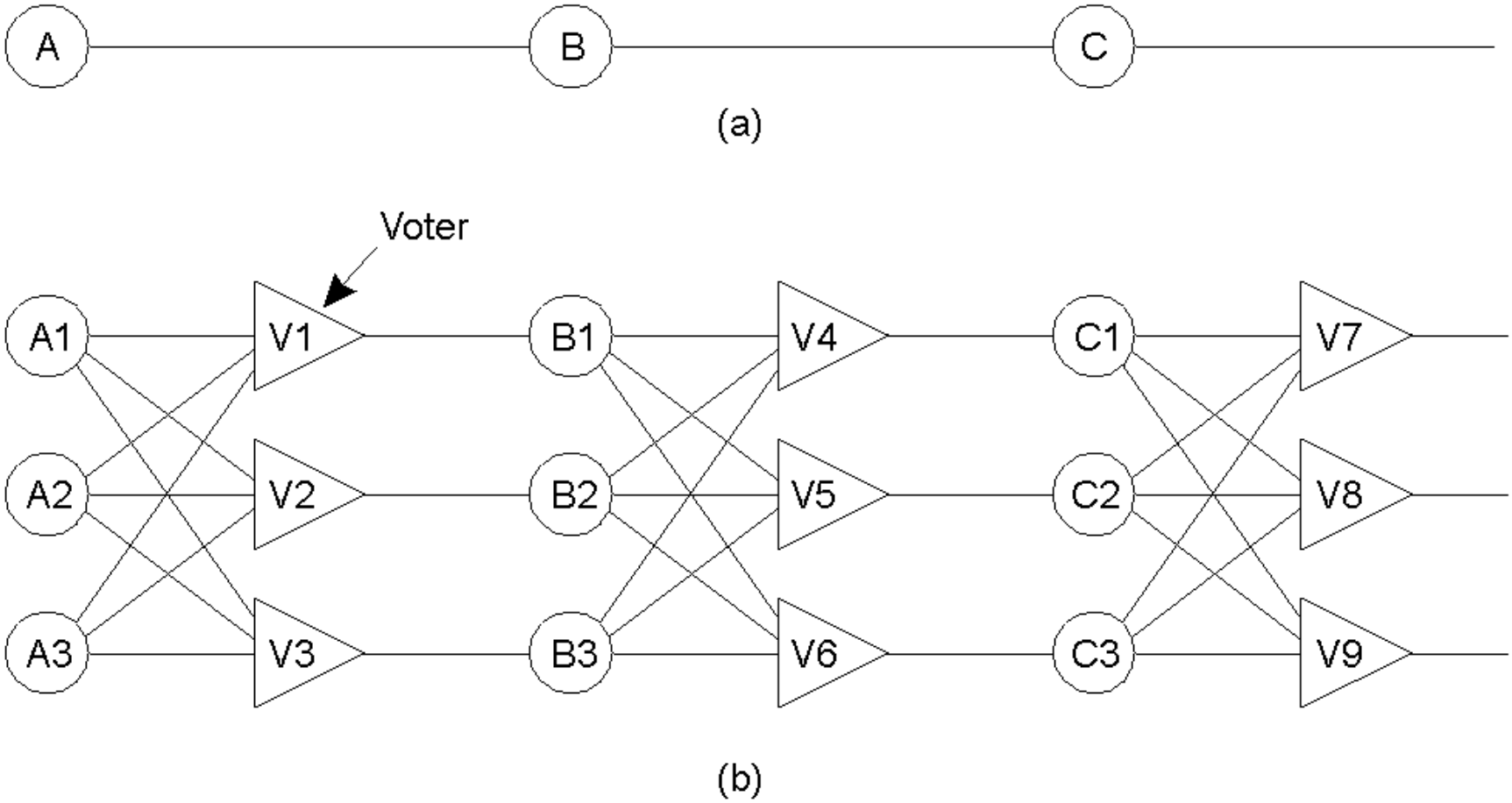
- Fault tolerance: system should provide services despite faults
  - Transient faults
  - Intermittent faults
  - Permanent faults

Failure Models

Type of failure	Description
Crash failure	A server halts, but is working correctly until it halts
Omission failure *Receive omission* *Send omission*	A server fails to respond to incoming requests A server fails to receive incoming messages A server fails to send messages
Timing failure	A server's response lies outside the specified time interval
Response failure *Value failure* *State transition failure*	The server's response is incorrect The value of the response is wrong The server deviates from the correct flow of control
Arbitrary failure	A server may produce arbitrary responses at arbitrary times

- Different types of failures
  - Server seen as black box component + interface

Failure Masking by Redundancy



- Triple modular redundancy (TMR)
  - 1 out of 3 replicas can have value errors