

Distributed Transactions: Architecture Guide

Overview

Attribute	Details
Duration	60 minutes
Level	Intermediate to Advanced
Prerequisites	Database fundamentals, microservices basics

Learning Objectives

- Understand why distributed transactions are fundamentally hard
- Compare 2PC, 3PC, Saga, and Eventual Consistency patterns
- Make informed architectural trade-offs for your use case
- Design compensation and recovery strategies
- Handle edge cases and failure scenarios

1. The Distributed Transaction Challenge

Why It's Fundamentally Hard

Monolithic (Easy):

```
Single Database
BEGIN TRANSACTION
  UPDATE accounts SET balance -= 100
  UPDATE accounts SET balance += 100
COMMIT
```

↑ ACID guaranteed by DB

Distributed (Hard):

```
Order      Payment      Inventory
DB1        DB2          DB3
```

↑ No single authority to guarantee ACID

Core Questions: - What if Payment succeeds but Inventory fails? - What if network partitions during commit? - What if a service crashes mid-transaction?

CAP Theorem: The Fundamental Trade-off

```
flowchart TD
    subgraph CAP["CAP: Pick Two"]
        direction TB
        C["C: Consistency"] --- A["A: Availability"]
        A --- P["P: Partition Tolerance"]
        P --- C
    end

    C -.- CP["CP: 2PC, Locks"]
```

```
A -.- CA["CA: Single DB"]
P -.- AP["AP: Saga"]
```

Choice	What You Get	What You Sacrifice	Example
CP	Strong consistency	Availability during partitions	2PC, distributed locks
AP	High availability	Immediate consistency	Saga, eventual consistency
CA	Both C and A	Not truly distributed	Single PostgreSQL

Reality: Network partitions WILL happen → You must choose between C and A

2. Two-Phase Commit (2PC)

How It Works

```
sequenceDiagram
    participant CO as Coordinator
    participant A as Participant A
    participant B as Participant B
    participant C as Participant C

    rect rgb(230, 240, 255)
    Note over CO,C: Phase 1: PREPARE (Voting)
    CO->>A: PREPARE
    CO->>B: PREPARE
    CO->>C: PREPARE
    A-->>CO: YES (locks held)
    B-->>CO: YES (locks held)
    C-->>CO: YES (locks held)
    end

    Note over CO: All YES → Decision: COMMIT

    rect rgb(220, 255, 220)
    Note over CO,C: Phase 2: COMMIT
    CO->>A: COMMIT
    CO->>B: COMMIT
    CO->>C: COMMIT
    A-->>CO: ACK (locks released)
    B-->>CO: ACK (locks released)
    C-->>CO: ACK (locks released)
    end
```

2PC State Machine

Coordinator States:

INIT → WAITING → COMMITTED/ABORTED

Participant States:

INIT → PREPARED → COMMITTED/ABORTED

↑

Locks held here (blocking!)

Critical Problems with 2PC

Problem	Scenario	Impact
Blocking	Participants hold locks during voting	Throughput drops, deadlock risk
Coordinator SPOF	Coordinator crashes after PREPARE	Participants stuck indefinitely
Network Partition	Can't reach all participants	Transaction hangs
Latency	2 round trips minimum	Not suitable for high-frequency

Edge Cases & Failure Scenarios

Scenario 1: Coordinator fails after sending PREPARE

Coordinator: PREPARE sent → CRASH

Participant A: PREPARED (holding locks)

Participant B: PREPARED (holding locks)

Result: Both participants BLOCKED indefinitely

Solution: Timeout + new coordinator election

But may cause inconsistency!

Scenario 2: Participant fails after voting YES

Participant A: YES → CRASH → RECOVERS

On recovery, must:

1. Check transaction log
2. Ask coordinator for decision
3. If coordinator also crashed → UNCERTAIN STATE

Scenario 3: Network partition during Phase 2

Coordinator: Sends COMMIT

Participant A: Receives COMMIT

Participant B: Network timeout

Result: A committed, B uncertain

Must retry COMMIT to B until success

When to Use 2PC

Good Fit	Poor Fit
Financial systems requiring strong consistency	High-throughput systems
Small number of participants (2-3)	Many microservices
Low-latency network (same datacenter)	Cross-region deployments
Batch processing jobs	User-facing real-time APIs

3. Three-Phase Commit (3PC)

Improvement Over 2PC

```
sequenceDiagram
    participant C0 as Coordinator
    participant P as Participants

    rect rgb(230, 240, 255)
    Note over C0,P: Phase 1: CAN-COMMIT
    C0->>P: Can you commit?
    P-->>C0: YES/NO
    end

    rect rgb(255, 245, 220)
    Note over C0,P: Phase 2: PRE-COMMIT
    C0->>P: Prepare to commit
    P-->>C0: ACK
    Note over P: Key: Can commit on timeout!
    end

    rect rgb(220, 255, 220)
    Note over C0,P: Phase 3: DO-COMMIT
    C0->>P: Commit now
    P-->>C0: Done
    end
```

3PC vs 2PC Trade-offs

Aspect	2PC	3PC
Phases	2	3
Blocking on coordinator failure	Yes (indefinite)	No (timeout-based recovery)
Latency	Lower (2 RTT)	Higher (3 RTT)
Complexity	Medium	High
Network partition safety	Problematic	Still problematic
Practical adoption	Common (XA)	Rare

Why 3PC Isn't Widely Used

Problem: 3PC still fails under network partitions

Scenario: Network splits during PRE-COMMIT

Partition A: Coordinator + Participant 1
Partition B: Participant 2, 3

Partition A: Times out → COMMIT (has majority?)
Partition B: Times out → COMMIT (assumed safe)

Result: Both partitions may make different
decisions → INCONSISTENCY

Reality: Network partitions are common in distributed systems
→ 3PC doesn't solve the fundamental problem
→ Industry moved to Saga/Eventual Consistency instead

4. Saga Pattern

Core Concept: Local Transactions + Compensation

```
flowchart LR
    subgraph forward["Forward Flow (Happy Path)"]
        T1["T1: Create Order"] --> T2["T2: Process Payment"]
        T2 --> T3["T3: Reserve Inventory"]
        T3 --> T4["T4: Arrange Shipping"]
    end
```

```
flowchart RL
    subgraph compensation["Compensation Flow (T3 Fails)"]
        F["T3 Failed "] --> C2["C2: Refund Payment"]
        C2 --> C1["C1: Cancel Order"]
    end
```

Key Principle

Each step T_i has a compensating transaction C_i

If T_n fails:

Execute C_{n-1} , C_{n-2} , ... C_1 in reverse order

Important: Compensation Rollback

- Rollback: Undo as if never happened
- Compensation: Apply corrective action (visible in history)

Choreography vs Orchestration

```
flowchart TB
    subgraph choreo["Choreography: Event-Driven"]
        O1["Order Service"] -->|"OrderCreated"| E1["Event Bus"]
        E1 -->|"OrderCreated"| P1["Payment Service"]
        P1 -->|"PaymentDone"| E1
        E1 -->|"PaymentDone"| I1["Inventory Service"]
    end
```

```
flowchart TB
    subgraph orch["Orchestration: Central Control"]
        ORCH["Saga Orchestrator"]
        ORCH -->|"1. CreateOrder"| O2["Order Service"]
        ORCH -->|"2. ProcessPayment"| P2["Payment Service"]
        ORCH -->|"3. ReserveInventory"| I2["Inventory Service"]
    end
```

Choreography vs Orchestration Trade-offs

Aspect	Choreography	Orchestration
Coupling	Loose	Tighter
Single Point of Failure	No	Yes (orchestrator)
Visibility	Hard to track flow	Easy to monitor
Debugging	Difficult	Straightforward
Adding new steps	Modify multiple services	Modify orchestrator only
Cyclic dependencies	Risk of event loops	Not possible
Team autonomy	High	Lower

Architecture Decision Guide

Choose CHOREOGRAPHY when:

- Teams are autonomous and own their services
- Flow is simple (< 4 steps)
- Services are truly independent
- You have good distributed tracing

Choose ORCHESTRATION when:

- Flow is complex (> 4 steps)
- Business logic is centralized
- You need clear visibility/monitoring
- Compensation logic is complex
- Regulatory/audit requirements exist

Critical Edge Cases

Edge Case 1: Compensation Fails

Scenario: Payment refund fails during compensation

- T1: Order Created
- T2: Payment Processed
- T3: Inventory Reserve FAILED
- C2: Refund Payment FAILED ← What now?

Solutions:

1. Retry with exponential backoff
2. Dead letter queue for manual intervention
3. Scheduled reconciliation job
4. Alert operations team

Edge Case 2: Duplicate Execution

Scenario: Network timeout, message redelivered

Payment Service receives "ProcessPayment" twice

Without idempotency:

→ Customer charged twice!

Solution: Idempotency keys

1. Each request has unique idempotency_key
2. Store processed keys in database

3. Check before processing, skip if exists

Edge Case 3: Out-of-Order Events

Scenario: Events arrive out of order

Expected: OrderCreated → PaymentDone → Shipped

Actual: PaymentDone → OrderCreated → Shipped

Solutions:

1. Event versioning/sequencing
2. State machine validation
3. Buffer and reorder
4. Reject and retry later

Edge Case 4: Long-Running Transactions

Scenario: Shipping takes 3 days

Problem: Can't hold resources for days

Solutions:

1. Reservation pattern (soft lock with expiry)
2. Split into sub-sagas
3. State machine with timeout transitions
4. Async notification when complete

Saga State Machine Design

```
stateDiagram-v2
    [*] --> STARTED
    STARTED --> ORDER_CREATED: createOrder()
    ORDER_CREATED --> PAYMENT_PROCESSED: processPayment()
    PAYMENT_PROCESSED --> INVENTORY_RESERVED: reserveInventory()
    INVENTORY_RESERVED --> COMPLETED: success

    ORDER_CREATED --> COMPENSATING: failure
    PAYMENT_PROCESSED --> COMPENSATING: failure
    INVENTORY_RESERVED --> COMPENSATING: failure

    COMPENSATING --> COMPENSATED: all compensations done
    COMPENSATING --> COMPENSATION_FAILED: compensation fails

    COMPLETED --> [*]
    COMPENSATED --> [*]
    COMPENSATION_FAILED --> MANUAL_INTERVENTION
```

5. Eventual Consistency & Outbox Pattern

The Dual Write Problem

```
flowchart LR
    S["Service"] -->|"1. Write DB "| DB[(Database)]
    S -->|"2. Publish Event "| MQ[Message Broker]

    style MQ stroke:#ff0000,stroke-width:2px
```

Problem: Two separate systems, no atomic guarantee

Failure scenarios:

1. DB write succeeds, event publish fails
→ Data saved but other services never notified
2. Event published, DB write fails
→ Other services act on non-existent data
3. Service crashes between the two operations
→ Inconsistent state

Transactional Outbox Pattern

```
flowchart TB
    S["Service"] --> TX
    subgraph TX["Single Database Transaction"]
        W1["1. Write business data"]
        W2["2. Write event to outbox table"]
    end
    TX --> DB[(Database)]
    RELAY["Message Relay<br/>(Polling or CDC)"] -->|"Read outbox"| DB
    RELAY -->|"Publish"| KAFKA["Message Broker"]
```

Outbox Table Design

```
CREATE TABLE outbox_events (
    id                UUID PRIMARY KEY,
    aggregate_type    VARCHAR(255),    -- e.g., 'Order'
    aggregate_id      VARCHAR(255),    -- e.g., order_id
    event_type        VARCHAR(255),    -- e.g., 'OrderCreated'
    payload           JSONB,           -- event data
    created_at        TIMESTAMP,
    published_at      TIMESTAMP NULL,  -- NULL = not yet published
    INDEX idx_unpublished (published_at) WHERE published_at IS NULL
);
```

Message Relay Strategies

Strategy	Pros	Cons
Polling	Simple, no extra infrastructure	Latency, DB load
CDC (Debezium)	Real-time, low DB load	Complex setup
Transaction log tailing	Very efficient	DB-specific

CDC vs Polling Trade-offs

Polling:

Latency: 1-5 seconds (configurable)
 DB Load: Constant queries
 Complexity: Low
 Ordering: Must handle carefully
 Best for: Simple setups, low volume

CDC (Change Data Capture):

Latency: Milliseconds
 DB Load: Minimal (reads transaction log)
 Complexity: High (Kafka Connect, Debezium)
 Ordering: Guaranteed by log position
 Best for: High volume, low latency requirements

Idempotent Consumer Pattern

Problem: Message broker may deliver same message twice
(at-least-once delivery)

Solution: Track processed message IDs

```

Consumer receives message
↓
Check: Is message_id in processed_messages table?
↓
YES → Skip (already processed)
NO  → Process + Insert message_id + Commit

All in single transaction!
  
```

6. Pattern Comparison & Selection Guide

Comprehensive Comparison

Aspect	2PC	3PC	Saga	Eventual Consistency
Consistency	Strong	Strong	Eventual	Eventual
Isolation	Full	Full	None	None
Availability	Low	Medium	High	High
Latency	High	Higher	Medium	Low
Scalability	Poor	Poor	Good	Excellent
Complexity	Medium	High	High	Medium

Aspect	2PC	3PC	Saga	Eventual Consistency
Recovery	Automatic	Timeout-based	Compensation	Retry + Idempotency

Decision Matrix

flowchart TB

```

    START["Need distributed transaction?"] --> Q1{"Strong consistency<br/>required?"}

    Q1 -->|"Yes"| Q2{"Can tolerate<br/>blocking?"}
    Q1 -->|"No"| Q3{"Complex multi-step<br/>workflow?"}

    Q2 -->|"Yes"| TWO_PC["2PC<br/>(XA Transactions)"]
    Q2 -->|"No"| CONSIDER["Consider relaxing<br/>consistency requirements"]

    Q3 -->|"Yes"| SAGA["Saga Pattern"]
    Q3 -->|"No"| OUTBOX["Outbox + Eventual<br/>Consistency"]

    SAGA --> Q4{"Need visibility<br/>& control?"}
    Q4 -->|"Yes"| ORCH["Orchestration"]
    Q4 -->|"No"| CHOREO["Choreography"]

```

Industry Use Cases

Company/Domain	Pattern	Reason
Banking (transfers)	2PC or Saga with strict compensation	Regulatory, money involved
E-commerce (orders)	Saga (orchestration)	Complex flow, need visibility
Social media (posts)	Eventual consistency	High scale, consistency less critical
Ride-sharing (booking)	Saga (choreography)	Real-time, multiple services
Inventory systems	Saga + reservation pattern	Prevent overselling

7. Production Considerations

Monitoring & Observability

Essential Metrics:

- Saga completion rate
- Compensation frequency
- Average saga duration
- Failed/stuck sagas count
- Outbox lag (unpublished events)
- Message processing latency

Essential Logs:

- Saga state transitions
- Compensation triggers

- Retry attempts
- Timeout events

Distributed Tracing:

- Correlation ID across all services
- Span for each saga step
- Parent-child relationship for compensation

Failure Recovery Strategies

Strategy 1: Automatic Retry

- Exponential backoff: 1s, 2s, 4s, 8s...
- Max retries: 3-5 typically
- Circuit breaker after threshold
- Alert on repeated failures

Strategy 2: Dead Letter Queue

- Move failed messages to DLQ
- Manual inspection and replay
- Audit trail preserved
- No blocking of other messages

Strategy 3: Scheduled Reconciliation

- Periodic job compares expected vs actual state
- Fixes inconsistencies automatically
- Reports discrepancies
- Last resort safety net

Testing Distributed Transactions

Test Categories:

- Happy path (all services succeed)
- Single service failure
- Multiple service failures
- Network partition simulation
- Timeout scenarios
- Duplicate message handling
- Out-of-order message handling
- Compensation failure scenarios

Tools:

- Chaos engineering (Chaos Monkey, Litmus)
- Network fault injection (Toxiproxy)
- Contract testing (Pact)
- Integration test containers

8. Key Takeaways

#	Takeaway
1	2PC provides strong consistency but blocks and doesn't scale
2	3PC reduces blocking but still fails under network partitions
3	Sagas trade isolation for availability — design for compensation

#	Takeaway
4	Choreography is loosely coupled but hard to debug
5	Orchestration centralizes logic but creates SPOF
6	Outbox pattern solves dual-write — use CDC for production
7	Design for idempotency — messages will be delivered multiple times
8	Compensation Rollback — it's a corrective action, not undo
9	Monitor saga health — stuck sagas indicate systemic issues
10	Choose based on requirements — not all systems need strong consistency

9. Practical Exercise

Design Challenge

Design a distributed transaction strategy for a ride-sharing booking:

Flow: 1. User requests ride 2. Find available driver 3. Reserve driver (can't accept other rides) 4. Process payment authorization 5. Confirm booking

Requirements: - Driver reservation expires after 30 seconds - Payment failures should release driver - Handle driver cancellation after booking - Support concurrent booking attempts for same driver

Discussion Questions: 1. Choreography or orchestration? Why? 2. How do you handle “driver reserved but payment timeout”? 3. What if compensation (release driver) fails? 4. How do you prevent double-booking a driver? 5. How do you show booking status to user in real-time?