

Global File Storage & Sharing System - High-Level Design (HLD)

Table of Contents

1. [Problem Statement](#)
 2. [Requirements](#)
 3. [Capacity Estimation](#)
 4. [Back-of-Envelope Calculations](#)
 5. [High-Level Architecture](#)
 6. [Core Components](#)
 7. [Data Flow](#)
 8. [Database Design](#)
 9. [Design Decisions & Trade-offs](#)
 10. [Scalability Strategy](#)
 11. [Reliability & Availability](#)
 12. [Security](#)
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Problem Statement

Design a scalable, reliable, globally distributed file storage and sharing system like Google Drive or Dropbox.

Core Functionality

- Upload/download files of any size (up to 10GB)
- Share files with other users (permissions management)
- Access files from multiple devices
- Real-time synchronization across devices
- Search files by name and content
- Version history (last 30 days)

Target Scale

- **1 Billion** total users
 - **200 Million** daily active users (DAU)
 - **10 PB** total storage
 - **99.99%** availability
 - Global distribution
-

Requirements

Functional Requirements

1. **File Operations:** Upload, download, delete, update files
2. **Folder Management:** Create folders, organize hierarchy
3. **Sharing:** Share files/folders with users or via public links
4. **Synchronization:** Real-time sync across devices
5. **Search:** Find files by name, content, metadata
6. **Version Control:** Maintain and restore file versions
7. **Permissions:** Owner, editor, viewer access levels

Non-Functional Requirements

1. **Scalability:** Handle billions of users, petabytes of data
2. **Availability:** 99.99% uptime (4 nines)
3. **Consistency:** Strong for metadata, eventual for sync
4. **Performance:**
 - Upload/Download: < 100ms latency (small files)
 - Sync latency: < 5 seconds
5. **Durability:** 99.999999999% (11 nines)
6. **Security:** Encryption at rest and in transit, access control

Capacity Estimation

Scale Assumptions

Users:

- Total Users: 1 Billion
- Daily Active Users (DAU): 200 Million (20%)
- Average Storage per User: 10 GB

Files:

- Average File Size: 2 MB
- Files per User: 5,000
- Total Files: 5 Trillion files

Read/Write Ratio: 3:1 (more downloads than uploads)

Storage Requirements

Total Storage = 1B users × 10 GB = 10 PB
With 3x replication = 30 PB

Traffic Estimation

Upload Traffic:

- Daily uploads: $200\text{M users} \times 5 \text{ files} = 1\text{B files/day}$
- Upload data: $1\text{B} \times 2 \text{ MB} = 2 \text{ PB/day}$
- Upload bandwidth: $\sim 23 \text{ GB/sec}$

Download Traffic:

- 3x uploads = $\sim 69 \text{ GB/sec}$

Total Peak Bandwidth: $\sim 100 \text{ GB/sec}$

QPS (Queries Per Second)

Total Daily Requests:

- Uploads: 1B requests
- Downloads: 3B requests
- Metadata operations: 2B requests
- Total: 6B requests/day

Average QPS: $\sim 70,000$

Peak QPS (3x): $\sim 200,000$

Memory for Caching

Metadata Cache:

- 1 KB per file metadata
- Cache 20% hot files = 1T files
- Required cache: $\sim 1 \text{ TB}$

Back-of-Envelope Calculations

Latency Numbers Every Programmer Should Know

L1 cache reference	0.5 ns		
Branch mispredict	5 ns		
L2 cache reference	7 ns		14x L1 cache
Mutex lock/unlock	25 ns		
Main memory reference	100 ns		20x L2, 200x L1
Compress 1K with Snappy	10,000 ns	10 μs	
Send 1 KB over 1 Gbps network	10,000 ns	10 μs	
Read 4 KB randomly from SSD	150,000 ns	150 μs	$\sim 1\text{GB/sec SSD}$
Read 1 MB sequentially from memory	250,000 ns	250 μs	
Round trip within datacenter	500,000 ns	500 μs	
Read 1 MB sequentially from SSD	1,000,000 ns	1000 μs	1 ms
Disk seek	10,000,000 ns	10000 μs	10 ms

```
Read 1 MB sequentially from disk 30,000,000 ns 30000 µs 30 ms
Send packet US→Europe→US          150,000,000 ns 150 ms
```

Powers of Two Table

Power	Exact Value	Approx Value	Bytes
10	1,024	~1 thousand	1 KB
16	65,536	~65 thousand	64 KB
20	1,048,576	~1 million	1 MB
30	1,073,741,824	~1 billion	1 GB
40	1,099,511,627,776	~1 trillion	1 TB
50	~1 quadrillion		1 PB

Server Capacity Estimation

Given:

- Upload bandwidth needed: 23 GB/sec
- Assume each server handles: 1 GB/sec network throughput
- Peak load factor: 3x average

Calculation:

- Average servers = 23 GB/sec ÷ 1 GB/sec = 23 servers
- Peak servers = 23 × 3 = 69 servers
- With redundancy (N+1) = ~70-80 servers

For global deployment:

- 3 regions × 80 servers = 240 servers total

Database Sizing

Metadata per file: ~1 KB

- file_id: 8 bytes
- owner_id: 8 bytes
- name: 255 bytes
- size: 8 bytes
- timestamps: 24 bytes
- path: 500 bytes
- other fields: ~200 bytes

Total: ~1 KB

For 5 Trillion files:

- Storage needed = 5T × 1 KB = 5 PB metadata
- With indexing (2x) = 10 PB
- Sharded across 1000 DB servers = 10 GB per server

Network Bandwidth Analysis

Upload Scenario:

- 200M users upload 5 files/day
- Average file size: 2 MB
- Total data: $200M \times 5 \times 2 \text{ MB} = 2 \text{ PB/day}$
- Per second: $2 \text{ PB} / 86,400 \text{ sec} \approx 23 \text{ GB/sec}$
- With peaks (3x): $\sim 70 \text{ GB/sec}$

Download Scenario (3:1 read/write ratio):

- Download traffic: $23 \times 3 = 69 \text{ GB/sec}$
- With peaks: $\sim 210 \text{ GB/sec}$

Total bandwidth requirement:

- Average: $\sim 100 \text{ GB/sec}$
- Peak: $\sim 300 \text{ GB/sec}$

Cost Estimation (Rough)

Storage Costs (AWS S3):

- $10 \text{ PB} \times \$23/\text{TB/month} = \$230,000/\text{month}$
- With replication (3x): $\sim \$700,000/\text{month}$

Bandwidth Costs:

- $2 \text{ PB upload/day} \times 30 \text{ days} = 60 \text{ PB/month}$
- $6 \text{ PB download/day} \times 30 \text{ days} = 180 \text{ PB/month}$
- At $\$0.09/\text{GB}$: $\sim \$21\text{M/month}$ for bandwidth

Database:

- $1000 \text{ instances} \times \$500/\text{month} = \$500,000/\text{month}$

Total Monthly Cost: $\sim \$22\text{M/month}$

Annual Cost: $\sim \$260\text{M/year}$

Per User Cost: $\$260\text{M} \div 1\text{B users} = \$0.26/\text{user/year}$

Availability Calculation

Target: 99.99% (Four 9s)

Downtime allowed per year:

- $365 \text{ days} \times 24 \text{ hours} \times 60 \text{ min} \times (1 - 0.9999)$
- $= 52.56 \text{ minutes/year}$
- $\approx 4.38 \text{ minutes/month}$

Component availability in sequence:

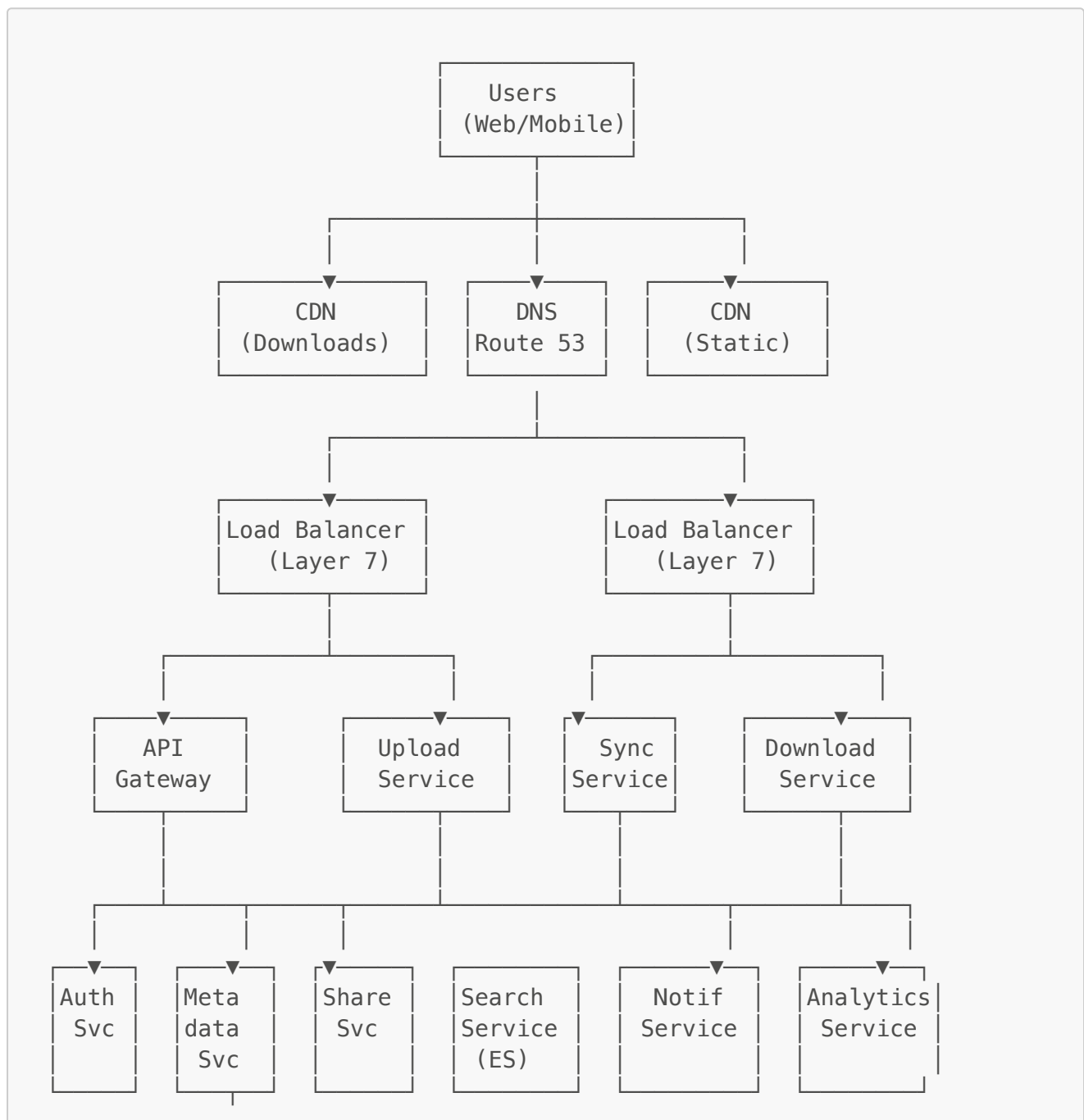
- Load Balancer: 99.99%
- Application: 99.99%
- Database: 99.99%
- Object Storage: 99.99%

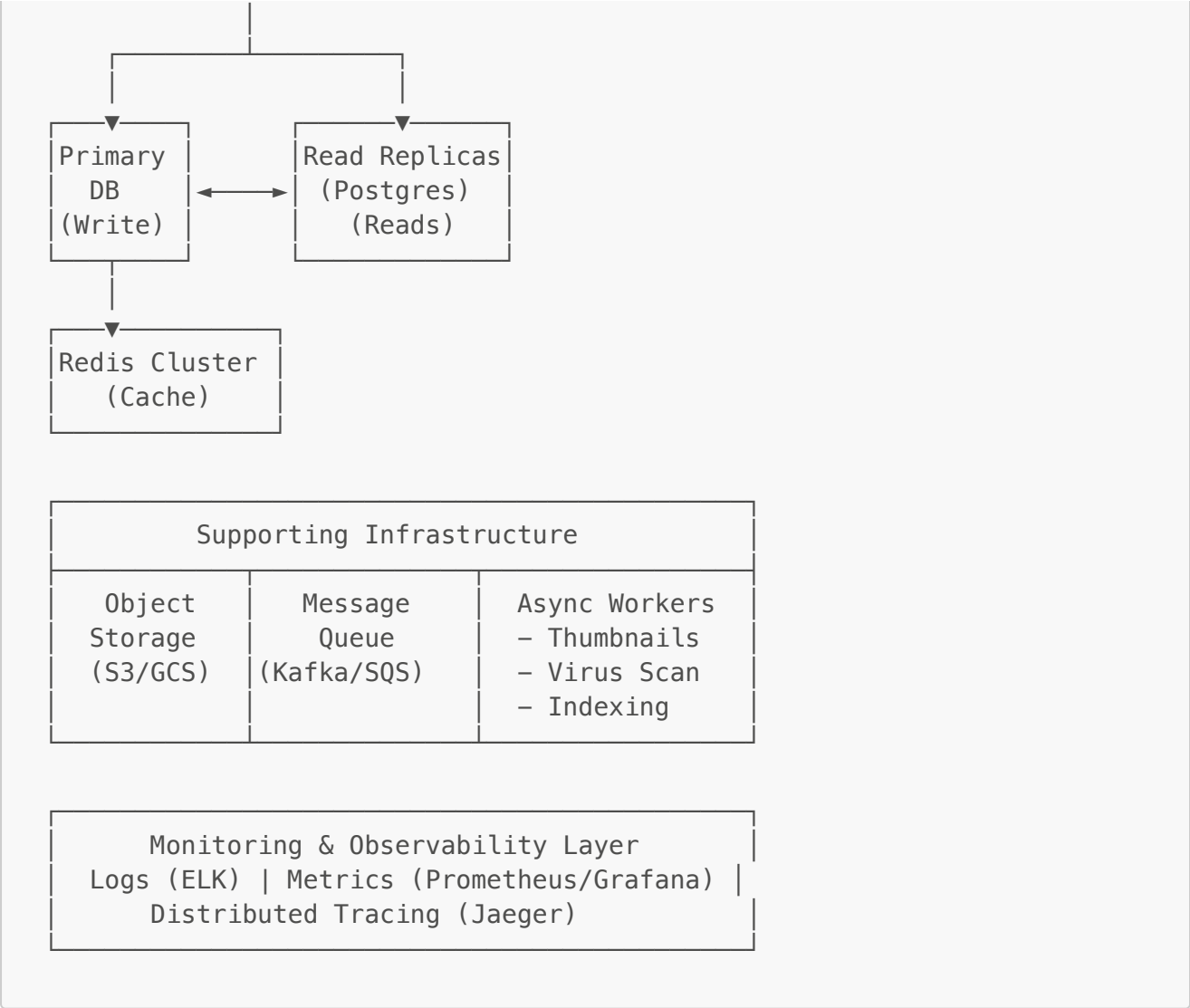
Total = $0.9999^4 = 99.96\%$

To achieve 99.99%:

- Need redundancy (active-active, failover)
- Multiple availability zones
- Health checks and auto-recovery

High-Level Architecture





Core Components Deep Dive

1. API Gateway

Purpose: Single entry point for all client requests

Responsibilities:

- Request routing to appropriate services
- Authentication and authorization (JWT tokens)
- Rate limiting (token bucket algorithm)
- Request/response transformation
- SSL termination
- API versioning

Technology Options Analysis:

Technology	Pros	Cons	Use Case
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Technology	Pros	Cons	Use Case
Kong	Plugins, extensible, open-source	Complex setup, learning curve	Recommended for flexibility
AWS API Gateway	Managed, integrated with AWS	Vendor lock-in, limited control	Quick MVP
Nginx	Lightweight, fast, proven	Manual scaling, less features	High-performance needs
Envoy	Modern, service mesh ready	Newer, smaller community	Microservices architecture

Decision: Kong API Gateway

Why Kong?

- **Plugin ecosystem:** 50+ plugins for auth, rate limiting, logging
- **Scalability:** Proven at 100K+ RPS (Mashape experience)
- **Flexibility:** Can customize with Lua plugins
- **Multi-protocol:** REST, gRPC, WebSocket support
- **Observability:** Built-in metrics and tracing

Trade-offs Accepted:

-  Best features and flexibility
-  Open-source with enterprise option
-  Active community
-  More complex than managed solution
-  Need to manage infrastructure ourselves

Configuration Strategy:

```

Rate Limiting:
├─ Free tier: 100 req/min per user
├─ Pro tier: 1,000 req/min per user
├─ Enterprise: 10,000 req/min per user
└─ Burst: 2x sustained rate for 10 seconds

Circuit Breaker:
├─ Failure threshold: 50% error rate
├─ Open duration: 30 seconds
├─ Half-open: Try 1 request
└─ Close when: 10 consecutive successes

```

2. Upload Service

Purpose: Handle file uploads efficiently at scale

Challenge: How to upload large files reliably?

Problem Analysis:

Single-shot upload of 5GB file:

- Network interruption → Complete failure
- 30-minute upload time
- No progress indication
- Cannot resume
- Server memory bottleneck (buffer entire file)

Solution Comparison:

Approach	Max File Size	Resume	Speed	Complexity	Choice
Single Upload	100 MB	✗	Slow	Simple	✗
Chunked Sequential	10 GB	✓	Slow	Medium	✗
Chunked Parallel	10 GB+	✓	Fast	Medium	✓ Recommended
Streaming	Unlimited	⚠	Medium	High	Special cases

Decision: Chunked Parallel Upload

Architecture:

STEP 1: Initiate Upload

Client:

```
├─ File: vacation.mp4 (500 MB)
├─ Calculate SHA-256: abc123...def
├─ POST /api/v1/files/initiate
  Body: {
    filename: "vacation.mp4",
    size: 524288000,
    checksum: "abc123...def",
    parent_folder_id: "folder_456"
  }
```

STEP 2: Deduplication Check

Upload Service:

```
├─ Query: SELECT file_id FROM files WHERE checksum = 'abc123...def'
├─ Result: NULL (file doesn't exist)
├─ Decision: Proceed with upload
```

If file existed:

- └─ Create file reference for user
- └─ No actual upload needed
- └─ Response: "Upload complete" (instant)
- └─ Savings: 500 MB bandwidth + storage

STEP 3: Generate Upload Plan

Upload Service:

- └─ Calculate chunks: $500 \text{ MB} \div 5 \text{ MB} = 100 \text{ chunks}$
- └─ Generate upload_id: "upload_789"
- └─ Create presigned S3 URLs (one per chunk):
 - └─ Chunk 1: PUT s3://bucket/uploads/upload_789/chunk_1?sig=...
 - └─ Chunk 2: PUT s3://bucket/uploads/upload_789/chunk_2?sig=...
 - └─ ... (100 URLs total)
- └─ Store in Redis:
 - └─ Key: upload:789:metadata
 - └─ Value: {status: "in_progress", chunks: 100, completed: 0}
 - └─ TTL: 24 hours
- └─ Response: {
 - upload_id: "upload_789",
 - chunk_size: 5242880,
 - chunks: 100,
 - urls: [...]

STEP 4: Upload Chunks (Parallel)

Client:

- └─ Split file into 100 chunks
- └─ Upload 10 chunks concurrently (parallel)
- └─ For each chunk:
 - └─ PUT to presigned URL
 - └─ Verify ETag matches
 - └─ Notify server: POST /api/v1/files/upload_789/chunks/1
 - └─ Update progress bar
- └─ Continue until all 100 chunks uploaded

Why 10 concurrent chunks?

- └─ Too few (1-5): Slow, underutilizes bandwidth
- └─ Optimal (8-12): Saturates bandwidth, manageable
- └─ Too many (20+): Overhead, diminishing returns

STEP 5: Track Progress

Upload Service:

- └─ Each chunk completion:
 - └─ HINCRBY upload:789:metadata completed 1
 - └─ HSET upload:789:chunks chunk_1 "completed"
- └─ WebSocket: Push progress to client
 - └─ {uploaded: 45, total: 100, percent: 45}

- └ Client shows: "Uploading... 45%"

Failure Handling:

- └ Chunk 42 fails (network timeout)
- └ Client retries chunk 42 (3 attempts)
- └ Success on retry
- └ Continue with remaining chunks

Resume from Failure:

- └ Client crashes at 60% complete
- └ Client restarts
- └ GET /api/v1/files/upload_789/status
- └ Response: {completed_chunks: [1-60], pending: [61-100]}
- └ Resume uploading chunks 61-100 only

STEP 6: Assemble File

Client:

- └ POST /api/v1/files/upload_789/complete
- Body: {checksum: "abc123...def"}

Upload Service:

- └ Verify all 100 chunks received
- └ S3 Multipart Complete API:
 - └ Combines all chunks into single file
- └ Calculate final checksum
- └ Verify matches client checksum
- └ Atomic operation (all-or-nothing)

If checksum mismatch:

- └ File corrupted during upload
- └ Delete all chunks
- └ Response: 400 Bad Request
- └ Client retries entire upload

STEP 7: Finalize & Trigger Jobs

Upload Service:

- └ Move file to permanent location:
 - └ s3://bucket/users/user_123/files/file_789
- └ Update metadata database:
 - └ INSERT INTO files (...)
 - └ UPDATE users SET storage_used += 500MB
- └ Delete temporary chunks & Redis state
- └ Publish to message queue:
 - └ Event: file_uploaded
 - └ Payload: {file_id, user_id, size, mime_type}
 - └ Consumers:
 - └ Thumbnail generator (if image/video)
 - └ Virus scanner (all files)
 - └ Content indexer (if document)

└─ Sync notifier (notify other devices)

Total Latency Breakdown:

- └─ Deduplication check: 5ms
- └─ Generate URLs: 10ms
- └─ Upload chunks (parallel): 30s (depends on bandwidth)
- └─ Assemble file: 2s
- └─ Update metadata: 50ms
- └─ Trigger jobs: 10ms

Key Optimizations:

1. Direct to S3 Upload

Without presigned URLs:

- Client → Upload Service → S3
- Upload Service buffers data
- Memory bottleneck
- Double bandwidth usage

With presigned URLs:

- Client → S3 (direct)
- Zero server load
- Single network hop
- Upload Service only coordinates

2. Chunk Size Selection: Why 5MB?

Too Small (1 MB):

- └─ 500 chunks for 500 MB file
- └─ High HTTP overhead
- └─ More S3 API calls (\$\$\$)
- └─ ❌ Not optimal

Optimal (5 MB):

- └─ 100 chunks for 500 MB file
- └─ Balanced overhead
- └─ Good for resume
- └─ ✅ **Recommended**

Too Large (50 MB):

- └─ Only 10 chunks
- └─ Less granular progress
- └─ Longer retry on failure
- └─ ❌ Not optimal

3. Deduplication Savings

Scenario: 1000 users upload same 100MB video

Without deduplication:

└─ Store: $1000 \times 100 \text{ MB} = 100 \text{ GB}$

With deduplication:

└─ Store: 100 MB (single copy)
└─ 1000 references to same file
└─ Savings: 99.9 GB (99.9%)

Real-world impact:

- Popular files (memes, movies, documents): 20–30% dedup rate
- Backups/system files: 50–70% dedup rate
- Average: 30% storage savings

3. Download Service

Purpose: Efficient file downloads

Key Features:

- **CDN Integration:** Serve files from edge locations
- **Signed URLs:** Time-limited download links (15 min)
- **Range Requests:** Support partial downloads (resume)
- **Permission Validation:** Check access before generating URL
- **Streaming:** Support for large files

4. Metadata Service

Purpose: Manage file and folder metadata at global scale

Challenge: How to store 5 trillion file metadata records?

Problem Analysis:

Single PostgreSQL instance limits:

└─ Max connections: ~10,000
└─ Max storage: ~16 TB (practical limit)
└─ QPS capacity: ~10,000 reads, ~1,000 writes
└─ Our need: 70,000 QPS, 10 PB metadata

Conclusion: Single database cannot handle the scale

Database Technology Comparison:

Database	Consistency	Querles	Scale	Ops	Choice
PostgreSQL	Strong (ACID)	Complex (JOINS)	Shard required	Mature	✅ Primary
MySQL	Strong (ACID)	Complex	Shard required	Very mature	Alternative
MongoDB	Eventual	Simple	Auto-shard	Easy	❌ Eventual consistency issue
Cassandra	Eventual	Limited	Excellent	Complex	❌ No JOINS
DynamoDB	Eventual	Limited	Excellent	Managed	❌ No JOINS for folders

Decision: PostgreSQL with Sharding

Why PostgreSQL over alternatives?

1. Strong Consistency Needed:

Scenario: User deletes file while another device reads it

With Strong Consistency (PostgreSQL):

- └─ Device A: DELETE file
- └─ Transaction commits
- └─ Device B: SELECT file
- └─ File not found (correct)

With Eventual Consistency (NoSQL):

- └─ Device A: DELETE file
- └─ Device B: SELECT file (different replica)
- └─ File still exists (stale read)
- └─ Conflict! Device B might re-upload deleted file

2. Complex Queries Required:

Query: "Get all files in folder and subfolders shared with user X"

```
SELECT f.*
FROM files f
JOIN folders fol ON f.parent_folder_id = fol.folder_id
JOIN shares s ON (s.resource_id = f.file_id OR s.resource_id =
fol.folder_id)
WHERE s.shared_with_user_id = 'user_X'
AND fol.path LIKE '/Documents/%'
ORDER BY f.modified_at DESC;
```

This is trivial in SQL, very complex in NoSQL

3. ACID Transactions Required:

Operation: Move file to different folder

```
BEGIN TRANSACTION;  
  UPDATE files SET parent_folder_id = 'new_folder' WHERE file_id =  
  'file_123';  
  UPDATE users SET storage_used = storage_used WHERE user_id =  
  'user_123';  
  INSERT INTO sync_events (...);  
COMMIT;
```

If any fails, all rollback (atomic)

NoSQL databases struggle with multi-document transactions

Sharding Strategy: Why shard by user_id?


Alternatives Considered:

Option 1: Shard by file_id (Hash Sharding)

Pros:

- └─ Even distribution
- └─ No hot shards

Cons:


- └─ User's files scattered across shards
- └─ Cannot query "all files for user" on single shard
- └─ Cross-shard JOIN for folder hierarchy
- └─  Rejected

Option 2: Shard by geography

Pros:


- └─ Low latency (data near users)
- └─ Regulatory compliance (data residency)

Cons:

- └─ Users travel → need cross-region queries
- └─ Uneven distribution (US > Antarctica)
- └─  Consider for global deployment

Option 3: Shard by user_id (Range Sharding)

Pros:

- └─ All user data on same shard ✓
- └─ Single-shard queries for user operations ✓
- └─ Folder hierarchy queries efficient ✓
- └─  ****Selected****

Cons:

- └─ Celebrity users create hot shards

- Need rebalancing as users grow
- Mitigation: Virtual nodes + monitoring

Implementation: Consistent Hashing with Virtual Nodes

Hash Ring with 1000 virtual nodes:

```
Hash Ring (0 to 2^32)

VN_001 (Shard A)
  VN_334 (Shard B)
    VN_667 (Shard C)
      VN_002 (Shard A)
        ...
          VN_999 (Shard C)
```

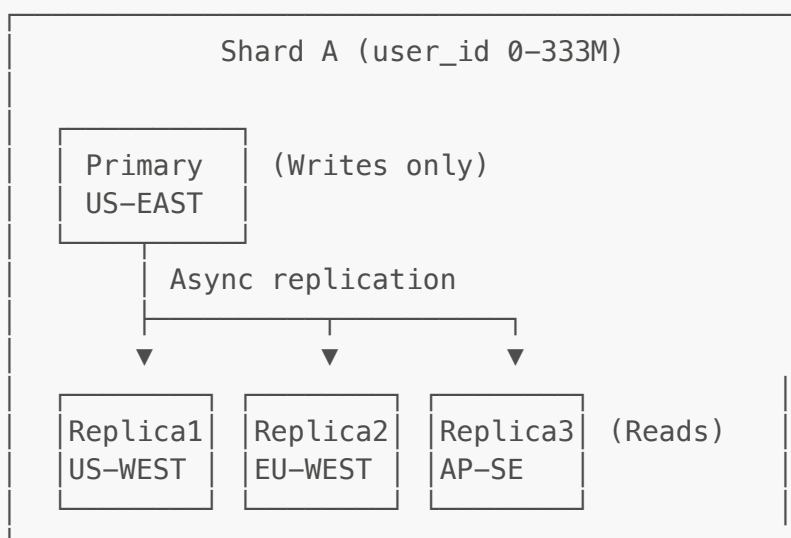
User assignment:

- $\text{hash}(\text{user_12345}) \rightarrow 445329847$
- Find next virtual node: VN_334
- VN_334 → Shard B
- Store all user_12345 data in Shard B

Benefits:

- Adding shard: Only $\sim 1/N$ data moves
- Removing shard: Evenly redistributed
- Hot users: Can assign dedicated shard

Read Replica Strategy:



Read distribution:

- 80% reads go to replicas (distributed)

- └─ 20% writes go to primary
- └─ Replication lag: <100ms (acceptable)

Failover on primary failure:

- └─ Promote Replica1 to primary (<30s)
- └─ Reconfigure replication topology
- └─ Resume operations

Caching Strategy: Multi-Layer

L1: Application Cache (In-Memory)

- └─ Store: Recently accessed file metadata
- └─ Size: 1 GB per API gateway instance
- └─ TTL: 5 minutes
- └─ Hit rate: 40%
- └─ Latency: 0.1ms

L2: Redis Cluster

- └─ Store: Hot file metadata, user quotas
- └─ Size: 100 GB total
- └─ TTL: 1 hour (metadata), 5 min (quotas)
- └─ Hit rate: 90% (of L1 misses)
- └─ Latency: 1ms

L3: Database

- └─ Store: All metadata (source of truth)
- └─ Size: 10 PB (5T files × 1KB × 2 for indexes)
- └─ Hit rate: 10% (cache misses)
- └─ Latency: 10–50ms

Combined Hit Rate = $40\% + (60\% \times 90\%) = 94\%$

Average Latency = $(0.4 \times 0.1\text{ms}) + (0.54 \times 1\text{ms}) + (0.06 \times 30\text{ms}) = 2.4\text{ms}$

Cache Invalidation Strategy:

Problem: File metadata updated, caches become stale

Solutions Compared:

Option 1: Write-Through Cache

- └─ Update cache + database simultaneously
 - └─ Pros: Always consistent
 - └─ Cons: Complex, all writes slower

Option 2: Cache Invalidation on Write

- └─ Delete cache key on update
 - └─ Pros: Simple, cache always fresh on next read
 - └─ Cons: Next read slower (cache miss)

✓ **Selected**

Option 3: TTL-based Expiration

- └ Let cache expire naturally
 - └ Pros: Zero overhead
 - └ Cons: Stale data for TTL duration
- △ Use with invalidation

Implementation:

- └ On file update:
 - └ UPDATE database
 - └ DEL cache key in Redis
 - └ Publish event to invalidate L1 cache
- └ Next read:
 - └ Cache miss
 - └ Query database
 - └ Populate cache

Quota Enforcement:

Challenge: Prevent storage quota exceeded

Naive approach (doesn't work):

GET user quota → Check limit → Upload file → Update quota

Problem: Race condition if concurrent uploads

Solution: Atomic check-and-reserve

- └ Use database transaction:
 - BEGIN;
 - SELECT storage_used, storage_quota FROM users WHERE user_id = X
 - FOR UPDATE;
 - Row locked, no other transaction can modify
 - IF storage_used + file_size <= storage_quota THEN
 - UPDATE users SET storage_used = storage_used + file_size;
 - Reserve space
 - COMMIT;
 - ELSE
 - ROLLBACK;
 - RETURN 'Quota exceeded';
 - END IF;
 - COMMIT;

Alternative: Pessimistic locking ensures atomicity

5. Synchronization Service

Purpose: Keep files in sync across devices in real-time

Challenge: How to sync files across billions of devices in real-time?

Problem Statement:

Scenario: User has 4 devices

- MacBook (editing document)
- iPhone (wants to view)
- iPad (offline)
- Work PC (wants to edit)

Requirements:

- └─ MacBook saves → iPhone sees update in <5 seconds
- └─ iPad offline → Syncs when online
- └─ Work PC edits same file → Conflict resolution
- └─ Scale: 200M users × 3 devices = 600M active connections

Synchronization Approach Comparison:

Approach	Latency	Scale	Offline	Conflicts	Bandwidth	Choice
Polling (5s)	5s avg	✓	✓	Manual	High	✗ Outdated
Long Polling	1-5s	✓✓	✓	Manual	Medium	△ Fallback
WebSocket	<1s	✓✓✓	✓	Auto	Low	✓ Primary
Server-Sent Events	<1s	✓✓	✓	Auto	Low	△ Alternative

Decision: WebSocket with Long Polling fallback

Why WebSocket?

1. Real-Time Bidirectional Communication

Polling (old way):

Every 5 seconds:

Client → Server: "Any changes?"

Server → Client: "No" (99% of the time)

Waste: 99% of requests return "no changes"

Latency: Average 2.5 seconds to detect change

WebSocket (new way):

One persistent connection

Server pushes when change occurs

Efficiency: 100× less network traffic

Latency: Milliseconds to detect change

2. Scalability Analysis

Polling approach:

- 200M users × 12 polls/min = 2.4B requests/min = 40M RPS
- Each poll: Network roundtrip, server processing
- Cost: Extremely high server load

WebSocket approach:

- 200M persistent connections
- Each connection idle most of the time (epoll)
- One server handles 100K connections (C10K problem solved)
- Need: 2,000 servers for connections
(Much better than 400,000 servers for polling)

3. Battery Efficiency (Mobile)

Polling on mobile:

- Wake up every 5 seconds
- Establish TCP connection
- Send HTTP request
- Battery drain: Significant

WebSocket:

- Single persistent connection
- Server pushes data
- No repeated wake-ups
- Battery drain: Minimal

For Mobile: Use native push notifications (APNs/FCM)

Architecture in Detail:

SYNC SCENARIO: User edits file on MacBook

Step 1: Detect Local Change (MacBook)

File System Watcher (inotify/FSEvents) |
— Monitors: ~/Drive directory
— Detects: file.txt modified
— Debounce: Wait 1s for more changes

↓

Step 2: Calculate Delta

Sync Client

- Get file's last sync version
- Calculate diff (rsync algorithm)
- Compression: gzip the diff
- Result: 500 KB → 50 KB (90% smaller)

↓

Step 3: Upload Change

```
POST /api/v1/sync/upload
Body: {
  file_id: "file_123",
  device_id: "device_mac",
  version: 42,
  delta: <compressed_binary>,
  checksum: "xyz789"
}
```

↓

Step 4: Store & Version

```
Upload Service
- Store delta in S3
- Update metadata:
  UPDATE files
  SET version = 43,
      modified_at = NOW()
  WHERE file_id = 'file_123'
```

↓

Step 5: Publish Sync Event

```
Message Queue (Kafka)
Topic: file_changes
Event: {
  type: "file_modified",
  file_id: "file_123",
  user_id: "user_456",
  version: 43,
  device_id: "device_mac",
  timestamp: 1640995200
}
```

↓

Step 6: Sync Service Processes Event

```
Sync Service Consumer
- Read from Kafka
- Query: Get all devices for user_456
- Filter: Exclude device_mac (originator)
- Result: [device_iphone, device_ipad,
           device_workpc]
```

↓

Step 7: Notify Devices

```
Device iPhone (WebSocket - Active)
- Connection alive
- Push: {file_id, version, delta_url}
- Latency: 50ms
```

```
Device iPad (Offline)
- No active connection
- Store in pending_sync queue
- Deliver when device comes online
```

```
Device WorkPC (WebSocket - Active)
- Currently editing same file!
- Conflict detected!
- Handle conflict resolution
```

↓

Step 8: Apply Changes

```
iPhone Sync Client
- Download delta from URL
- Apply patch to local file
- Verify checksum
- Update local version to 43
- Notify user: "file.txt updated"
```

Conflict Resolution: The Hard Problem

CONFLICT SCENARIO:

Time 10:00:00 – Both devices start with file version 42

Time 10:00:30 – MacBook saves "Hello World" (version 43)

Time 10:00:31 – WorkPC saves "Goodbye World" (version 43)

Both think they're at version 43!

```
Conflict Detection Algorithm
```

Server receives MacBook update:

```
├─ Current server version: 42
├─ MacBook claims base version: 42 ✓
├─ No conflict
└─ Accept, increment to version 43
```

```
Server receives WorkPC update (1 second later):
├─ Current server version: 43 (MacBook just updated)
├─ WorkPC claims base version: 42 x
├─ CONFLICT DETECTED!
└─ Need resolution strategy
```

Conflict Resolution Strategies Compared:

Strategy 1: Last-Write-Wins (LWW)

```
├─ Compare timestamps
├─ MacBook: 10:00:30
├─ WorkPC: 10:00:31 → Winner
├─ WorkPC version becomes version 44
├─ MacBook version discarded
└─ Problem: Data loss! (MacBook's edits lost)
```

Strategy 2: Merge Attempts (Operational Transform)

```
├─ Analyze both edits
├─ Try to merge intelligently
├─ "Hello World" + "Goodbye World" = ???
└─ Problem: Very complex, doesn't always work
```

Strategy 3: Last-Write-Wins + Conflict Copy

```
├─ WorkPC version becomes version 44 (latest timestamp)
├─ Save MacBook version as conflict copy:
│   "file (conflicted copy from MacBook 10:00:30).txt"
├─ Notify user on both devices
└─ User manually resolves
    Pros: No data loss, simple, always works
    Cons: Manual intervention needed
```

Implementation of LWW + Conflict Copy:

Conflict Resolution Flow

Server detects conflict:

```
├─ Base version: 42
├─ Current version: 43 (MacBook)
├─ Incoming version: 43 (WorkPC claims 42 base)
└─ Timestamp comparison:
    ├─ MacBook: 10:00:30
    └─ WorkPC: 10:00:31 (winner)
```

Resolution:

```
├─ Accept WorkPC as version 44
└─ Create conflict file for MacBook:
```

```

├─ Name: "file (MacBook 10:00:30 conflicted).txt"
├─ Content: MacBook's version
├─ Metadata: marked as conflict
├─ INSERT INTO files (conflict file)
├─ Publish sync events:
├─   └─ MacBook: Download conflict file
├─   └─ WorkPC: Download winning version
├─ Notify both users:
├─   "Conflicted copy created"

```

User Experience:

MacBook sees:

```

├─ file.txt (WorkPC version)
├─ file (MacBook 10:00:30 conflicted).txt (their version)
├─ Notification: "Conflict detected, review changes"

```

WorkPC sees:

```

├─ file.txt (their version, now official)
├─ file (MacBook 10:00:30 conflicted).txt (other version)
├─ Notification: "Another device edited simultaneously"

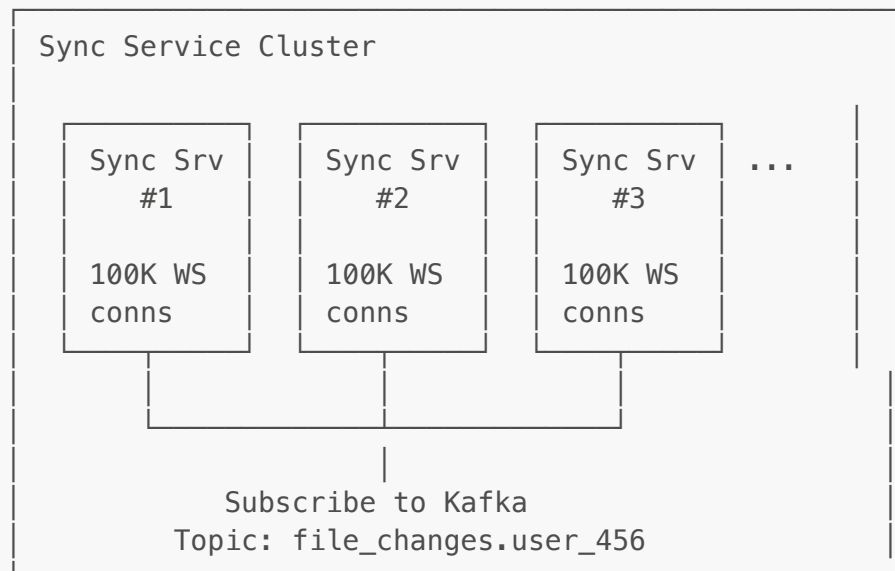
```

WebSocket Connection Management:

Challenge: Maintain 600M WebSocket connections

Solution: Connection Pooling + Load Distribution

Architecture:



Distribution Strategy:

```

├─ Each server: 100,000 connections
├─ Total servers: 600M ÷ 100K = 6,000 servers
├─ Load balancer: Consistent hashing by user_id
├─ Benefit: All user's devices on same server
├─   (Easier to broadcast changes)

```


Connection stickiness:

- └─ user_456 → Always routes to Sync Server #42
- └─ All user_456's devices on same server
- └─ Broadcast to 3-5 devices instead of searching 6000 servers

Cost optimization:

- └─ WebSocket keep-alive: 30 seconds
- └─ Idle connections use ~4 KB memory
- └─ 100K connections = 400 MB memory
- └─ Acceptable on modern servers (128 GB RAM)

Fallback to Long Polling:

Why need fallback?

WebSocket failures:

- └─ Corporate firewalls block WS
- └─ Proxy servers don't support WS
- └─ Old browsers (IE 9)
- └─ Mobile apps occasionally

Long Polling mechanism:

- | |
|--|
| 1. Client: GET /sync/poll?cursor=123
Server: Hold connection (60s) |
| 2. If change occurs → Return immediately
If 60s timeout → Return "no changes" |
| 3. Client immediately reconnects
(Keeps connection quasi-persistent) |

Comparison:

- └─ WebSocket: 1 connection, push messages
- └─ Long Poll: Reconnect every 60s, pull messages
- └─ WebSocket preferred, Long Poll acceptable fallback

Sync Protocol Design:

Delta Sync vs Full Sync:


Small change (edited 1 line in 10 MB document):

Full Sync:

- └─ Download entire 10 MB file
- └─ Bandwidth: 10 MB

- └─ Time: 10 seconds

Delta Sync:

- └─ Download only changed bytes
- └─ Bandwidth: 5 KB (0.05%)
- └─ Time: 50ms (200× faster)
- └─  ****Use for files < 100 MB****

Algorithm: rsync-like binary diff

- └─ Divide file into blocks (4 KB each)
- └─ Calculate rolling checksum per block
- └─ Compare checksums
- └─ Transfer only changed blocks
- └─ Reconstruct file from blocks + deltas

When to use Full Sync:

- └─ New file (no previous version)
- └─ File completely rewritten
- └─ Delta larger than original (rare)
- └─ First sync for device

Sync State Management:

Per-Device State Tracking:

Redis storage:

Key: sync:state:user_456:device_iphone

Value: {

```
  last_sync_timestamp: 1640995200,  
  last_event_id: "event_789",  
  pending_downloads: ["file_123", "file_456"],  
  pending_uploads: [],  
  sync_status: "synced"
```

}

Sync cursor:

- └─ Tracks last successfully processed event
- └─ On reconnect: Send all events after cursor
- └─ Ensures no events missed
- └─ Even if device offline for days

Offline Support:

iPad goes offline for 6 hours:

During offline:

- └─ Local changes queued (SQLite queue)
- └─ 10 files edited

- 5 files created
- Queue stored persistently

When online:

1. Reconnect to Sync Service
GET /sync/changes?cursor=last_event

↓

2. Download Remote Changes
 - 15 files changed by other devices
 - Download deltas
 - Apply to local files

↓

3. Upload Local Changes
 - 10 edited files → Check conflicts
 - 5 new files → Upload
 - 2 conflicts detected → Create copies

↓

4. Resolve Conflicts
 - file.txt has conflict
 - Create: file (iPad conflicted).txt
 - Notify user

Conflict Detection:

- Remote version: 44 (edited at 10:05)
- Local version: 43 (edited at 10:03 while offline)
- Both diverged from version 42
- CONFLICT! (cannot auto-merge)

Connection Recovery:

Challenge: WebSocket connections drop frequently

Causes:

- Network switches (WiFi → Cellular)
- Network congestion
- Server restarts
- Load balancer timeout
- Laptop sleep/wake

Recovery Strategy:

Detection:

- Client: Heartbeat every 30s

- └─ Server: Heartbeat every 30s
- └─ No heartbeat for 60s → Connection dead
- └─ Reconnect immediately

Reconnection:

- └─ Exponential backoff: 1s, 2s, 4s, 8s, 16s, 30s (max)
- └─ Include cursor in reconnect request
- └─ Server sends missed events
- └─ Resume normal operation

State preservation:

- └─ Server: Keep connection state for 5 minutes after disconnect
- └─ Client: Keep pending changes in queue
- └─ No data loss even with frequent reconnects

Bandwidth Optimization:

Smart Sync Features:

1. Selective Sync:

User marks folders:

- └─ "Always available offline" → Full sync
- └─ "Available on demand" → Metadata only
- └─ "Online only" → No local copy

Bandwidth savings: 50–70% for typical user

2. Sync Scheduling:

Mobile devices:

- └─ WiFi: Sync immediately
- └─ Cellular: Sync metadata only
- └─ Low battery: Pause sync
- └─ Charging + WiFi: Full sync

Battery & data plan friendly

3. Compression:

- └─ Text files: gzip (70% savings)
- └─ Images: Already compressed (skip)
- └─ Videos: Already compressed (skip)
- └─ Applied automatically per file type

4. Throttling:

- └─ Detect rapid changes (save every second)
- └─ Debounce: Wait 5 seconds for more changes
- └─ Batch upload: Send one delta with all changes
- └─ Reduces sync events by 80%

6. Sharing Service

Purpose: Manage file/folder sharing

Features:

- **User Sharing:** Share with specific users
- **Public Links:** Generate shareable URLs
- **Permission Levels:** Owner, Editor, Commenter, Viewer
- **Expiration:** Time-limited shares
- **Access Logs:** Track who accessed what

Security:

- Cryptographically secure tokens
 - Store hashed tokens (bcrypt)
 - Password protection for public links
 - Audit trail for all access
-

7. Search Service

Purpose: Fast file search across content and metadata

Technology: Elasticsearch

Search Capabilities:

- Full-text search in document content
- Metadata search (name, type, owner, date)
- Filters and facets
- Relevance ranking

Indexing Strategy:

- Asynchronous indexing via background workers
 - Text extraction from PDFs, Office docs
 - OCR for images (if enabled)
 - Real-time index updates
-

8. Storage Layer (Object Storage)

Purpose: Store actual file data

Technology: AWS S3 / Google Cloud Storage

Features:

- **Versioning:** Keep multiple versions of files
- **Lifecycle Policies:** Auto-tier to cold storage

- **Cross-Region Replication:** For disaster recovery
- **Encryption:** At rest (AES-256)

Storage Classes:

- **Hot:** Frequently accessed (S3 Standard)
 - **Warm:** Infrequent access (S3 IA)
 - **Cold:** Archive (S3 Glacier)
-

9. Message Queue

Purpose: Asynchronous communication between services

Technology: Apache Kafka, RabbitMQ, AWS SQS

Event Types:

- File upload complete → Trigger virus scan, thumbnail generation
- File modified → Trigger sync notification
- File shared → Send notification
- File accessed → Update analytics

Benefits:

- Decouple services
 - Handle traffic spikes
 - Guarantee delivery
 - Replay capability
-

10. Async Workers

Purpose: Process background jobs

Worker Types:

1. **Thumbnail Generator:** Create image/video previews
2. **Virus Scanner:** Scan files with ClamAV
3. **Content Indexer:** Extract text, update search index
4. **Analytics Processor:** Track usage metrics
5. **Notification Sender:** Email/push notifications

Processing:

- Pull jobs from message queue
 - Process independently
 - Retry with exponential backoff
 - Dead letter queue for failed jobs
-

11. Notification Service

Purpose: Real-time notifications to users

Channels:

- **WebSocket/SSE:** Real-time web notifications
- **Push Notifications:** Mobile (FCM/APNS)
- **Email:** Via SendGrid/SES
- **In-App:** Notification center

Events:

- File shared with you
 - Comment on your file
 - Upload complete
 - Storage quota warning
-

12. Cache Layer (Redis)

Purpose: Reduce database load, improve performance

Cached Data:

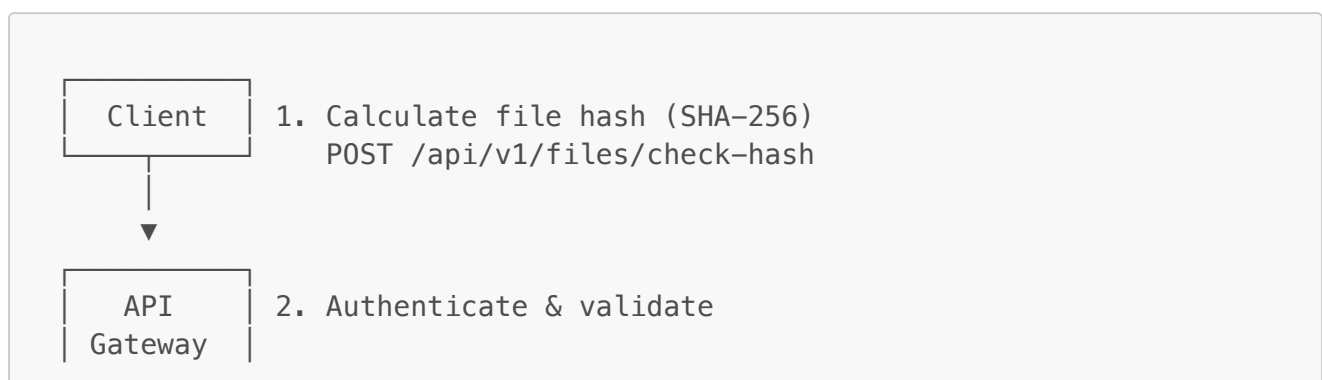
- User sessions (7 day TTL)
- File metadata (1 hour TTL)
- User quotas (5 min TTL)
- Recent files list (1 hour TTL)
- Sync cursors (24 hour TTL)

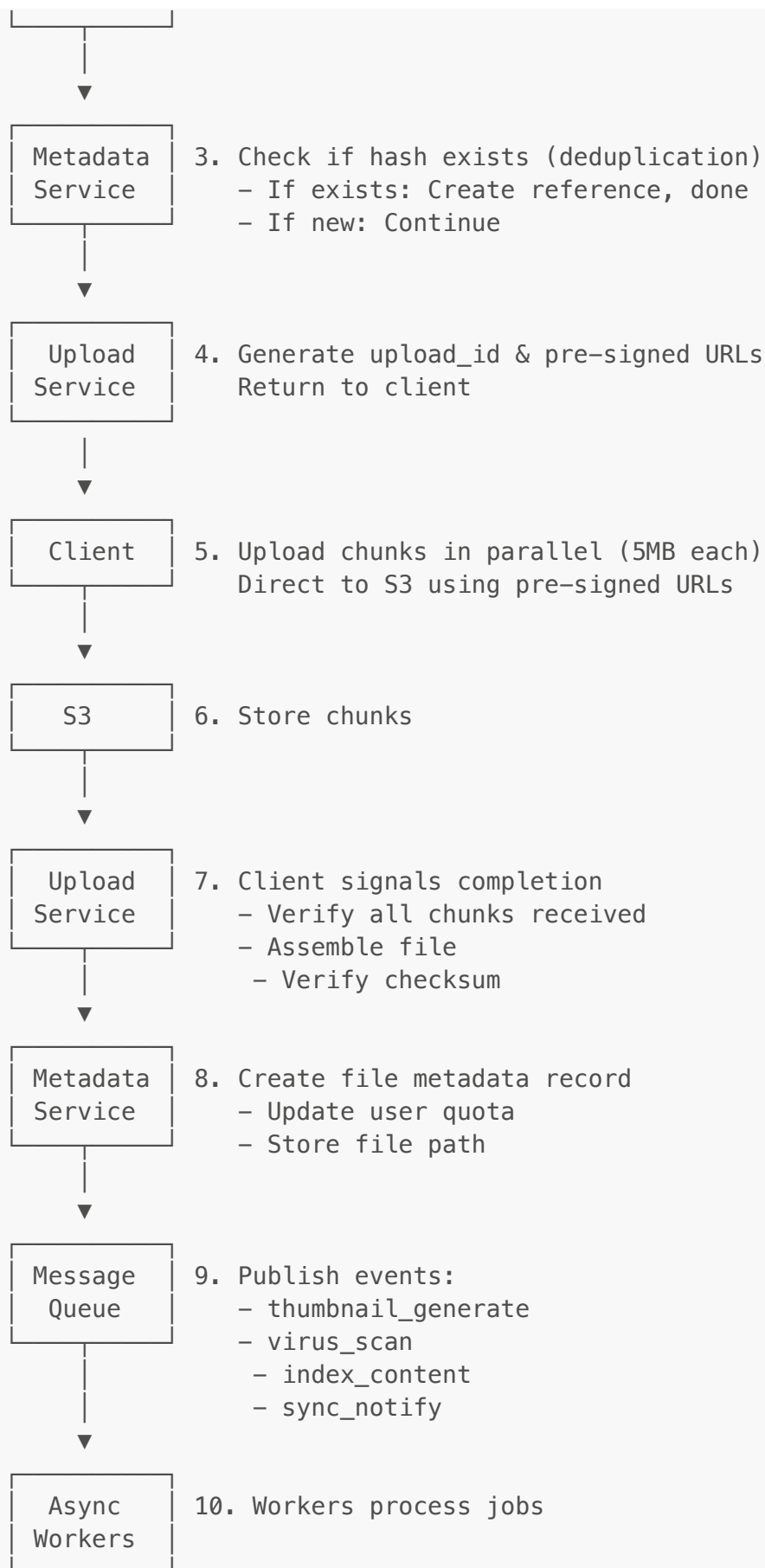
Strategy:

- Cache-aside pattern
 - Write-through for critical data
 - Cache invalidation on updates
-

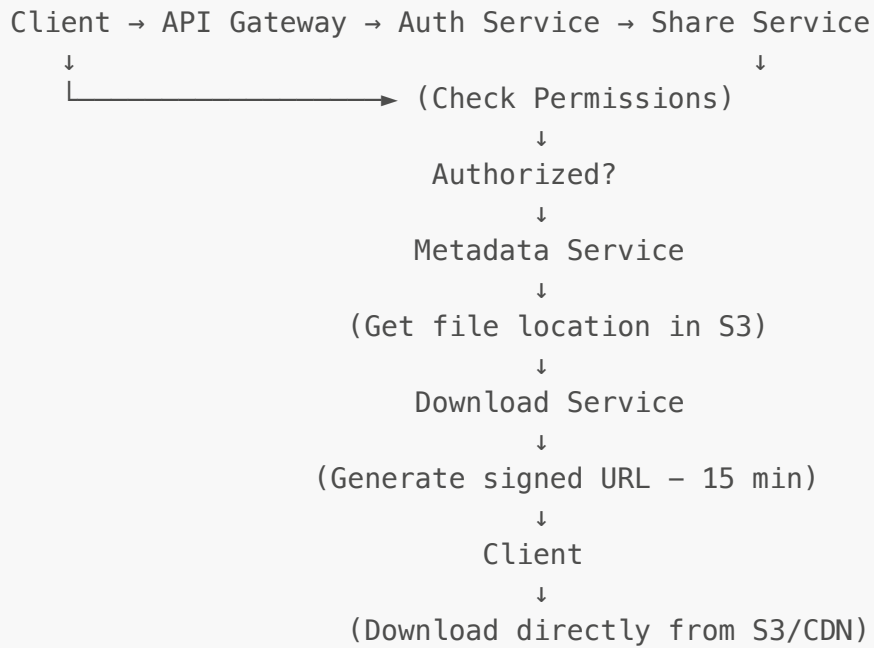
Data Flow

1. File Upload Flow

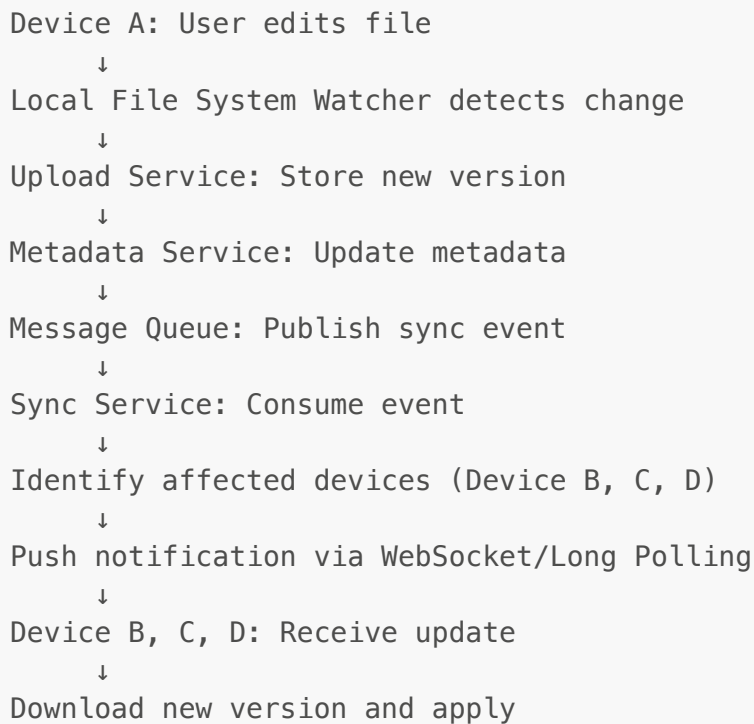




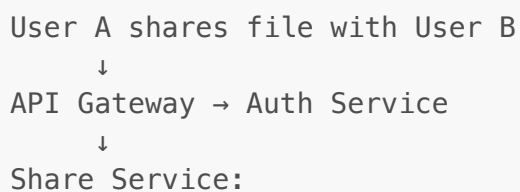
2. File Download Flow



3. Real-Time Sync Flow



4. File Sharing Flow



- Validate User A owns file
- Create share record in DB
- Generate share token

↓

Message Queue: Publish notification event

↓

Notification Service:

- Send email to User B
- Send push notification
- Create in-app notification

↓

User B receives notification

- Click link → Download Service
- Validate share token
- Generate signed URL
- Download file

Database Design

Metadata Database (PostgreSQL)

Core Entities:

1. **Users:** user_id, email, storage_quota, storage_used
2. **Files:** file_id, owner_id, name, size, checksum, parent_folder_id, version
3. **Folders:** folder_id, owner_id, name, parent_folder_id, path
4. **File_Versions:** version_id, file_id, version_number, storage_path
5. **Shares:** share_id, resource_id, owner_id, shared_with_user_id, permission
6. **Devices:** device_id, user_id, device_name, last_sync_at
7. **Sync_Events:** event_id, user_id, resource_id, event_type, timestamp

Key Indexes:

- owner_id + parent_folder_id (files/folders)
- checksum (for deduplication)
- share_token (for public links)
- user_id + timestamp (for sync events)

Sharding Strategy:

- Shard by user_id
- Consistent hashing
- Virtual nodes for better distribution

Document Store (MongoDB/DynamoDB)

Collections:

1. **File_Chunks**: Track chunks during upload
 2. **Activity_Logs**: User actions audit trail
 3. **Analytics_Events**: Usage metrics
-

Search Index (Elasticsearch)

Document Structure:

- file_id, name, content (extracted text)
 - owner, tags, created_at, modified_at
 - path, mime_type, size
-

Cache (Redis)

Key Patterns:

- session:{session_id} → User session
 - file:metadata:{file_id} → File metadata
 - user:quota:{user_id} → Storage usage
 - user:recent:{user_id} → Recent files list
 - sync:cursor:{user_id}:{device_id} → Sync state
-

Design Decisions & Trade-offs





1. Metadata: SQL vs NoSQL

Decision: PostgreSQL (SQL)

Reasoning:

- Strong consistency needed for ownership, permissions
- Complex queries (folder hierarchy, shared files)
- ACID transactions required
- Mature ecosystem

Trade-offs:

-  Strong consistency
 -  Relational integrity
 -  Complex queries
 -  Harder horizontal scaling (needs sharding)
-





2. Storage: Object Storage (S3/GCS)

Decision: Object Storage

Reasoning:

- Designed for petabyte scale
- Built-in 11 9's durability
- Cost-effective
- Versioning support
- Global CDN integration

Trade-offs:

-  Infinite scalability
 -  High durability
 -  Low cost per GB
 -  Higher latency than block storage
-




3. Consistency Model: Hybrid

Decision: Strong for metadata, Eventual for sync

Reasoning:

- Metadata (ownership, permissions) must be consistent
- Sync can tolerate eventual consistency
- Best balance of consistency and availability

Trade-offs:

-  Guarantees where needed
 -  High availability
 -  More complex implementation
-





4. Upload: Chunked Multipart

Decision: 5MB chunks, parallel upload

Reasoning:

- Support large files (>5GB)
- Faster (parallel upload)
- Resumable after failures
- Better progress tracking

Trade-offs:

-  Fast for large files
 -  Resumable
 -  Better UX
 -  More complex
-





5. Deduplication: Client-side hash

Decision: Client calculates hash, server verifies

Reasoning:

- Save bandwidth (don't upload duplicates)
- Save storage (single copy)
- Fast for users

Trade-offs:

-  30% bandwidth savings
 -  Storage savings
 -  Privacy concerns (hash reveals content)
 -  Reference counting complexity
-





6. Sync: WebSocket + Push Notifications

Decision: WebSocket for web/desktop, FCM/APNS for mobile

Reasoning:

- Real-time updates (low latency)
- Single persistent connection
- Battery efficient on mobile

Trade-offs:

-  Real-time (milliseconds)
 -  Efficient
 -  Connection management complexity
 -  Firewall issues
-





7. Sharding: By user_id

Decision: Shard metadata by user_id

Reasoning:

- User's data co-located
- No cross-shard queries for user operations
- Simple to implement

Trade-offs:

-  Query efficiency
 -  Data locality
 -  Hot shards (power users)
 -  Rebalancing complexity
-






8. Caching: Multi-layer (CDN + Redis + DB)

Decision: 3-tier caching

Reasoning:

- CDN for downloads (edge caching)
- Redis for metadata (application caching)
- DB buffer for query results

Trade-offs:

-  80%+ cache hit rate
 -  Reduced DB load
 -  Lower latency
 -  Cache invalidation complexity
 -  Additional cost
-

9. Conflict Resolution: Last-Write-Wins

Decision: LWW with conflict copies




Reasoning:

- Simple to implement
- Works for 99% of cases
- Preserves all data

Flow:

1. Detect conflict
2. Latest timestamp wins
3. Create conflict copy
4. User reviews

Trade-offs:

-  Simple
 -  No data loss
 -  User intervention needed
-

10. Database Replication: Primary-Replica





Decision: 1 Primary + 3 Read Replicas

Reasoning:

- Scale reads (most operations)
- High availability

- Geographic distribution

Trade-offs:

-  Read scalability (3x)
 -  HA
 -  Replication lag
 -  Failover complexity
-

Scalability Strategy

Horizontal Scaling

Application Layer:

- Stateless services behind load balancers
- Auto-scaling based on CPU/memory
- Can add/remove instances dynamically

Database Layer:

- Shard by user_id
- Add more shards as needed
- Read replicas for read scaling
- Eventually migrate to distributed SQL (CockroachDB, YugabyteDB)

Cache Layer:

- Redis cluster with consistent hashing
- Add nodes to handle more load
- Partition hot keys

Storage Layer:

- Object storage scales automatically
- No manual intervention needed

Vertical Scaling (When Needed)

- Upgrade database instance types
- More memory for cache
- Faster disks for database

Geographic Distribution

- Multi-region deployment
 - Route users to nearest region
 - Cross-region replication for DR
-

Reliability & Availability

High Availability Strategies

1. Redundancy:

- Multiple availability zones
- Load balancers in active-active
- Database primary-replica setup
- 3x replication for object storage

2. Failover:

- Automatic failover for database
- Health checks for all services
- Circuit breakers to prevent cascade failures

3. Graceful Degradation:

- Serve stale cache if DB down
- Disable features if dependencies fail
- Show user-friendly error messages

Disaster Recovery

Backup Strategy:

- Database: Continuous WAL archiving, daily full backups
- Object Storage: Cross-region replication, versioning
- Metadata: Daily exports to S3

Recovery:

- RPO (Recovery Point Objective): < 1 hour
- RTO (Recovery Time Objective): < 4 hours

Security Considerations

1. Authentication & Authorization

- JWT tokens with short expiry (15 min)
- Refresh tokens for session management
- OAuth 2.0 for third-party integrations
- Multi-factor authentication (TOTP, SMS)

2. Encryption

- **At Rest:** AES-256 encryption in S3
- **In Transit:** TLS 1.3 for all connections
- **End-to-End** (optional): Client-side encryption

3. Access Control

- Role-based permissions (Owner, Editor, Viewer)
- Least privilege principle
- Regular permission audits

4. Network Security

- VPC with private subnets
- Security groups and NACLs
- DDoS protection (CloudFlare, AWS Shield)
- Rate limiting at API Gateway

5. Data Protection

- Regular backups
- Data lifecycle policies
- Compliance (GDPR, HIPAA)
- Audit logs for all access

Summary & Key Points

Architecture Highlights

1. **Microservices:** Separate concerns (upload, download, metadata, sync, search)
2. **Horizontal Scalability:** Shard databases, object storage, read replicas
3. **High Availability:** Multi-region, replication, failover
4. **Strong Security:** Encryption, access control, audit logging
5. **Optimized Performance:** Caching, CDN, chunking, async processing

Key Design Principles

- **Scalability First:** Built to handle billions of users
- **Reliability:** Multiple layers of redundancy
- **Performance:** Caching and CDN for low latency
- **Security:** Defense in depth
- **Cost Efficiency:** Lifecycle policies, deduplication

Technology Choices

- **Metadata:** PostgreSQL (sharded)
- **Storage:** AWS S3 / GCS
- **Cache:** Redis Cluster
- **Message Queue:** Kafka / SQS
- **Search:** Elasticsearch
- **CDN:** CloudFront / Cloudflare

This design can handle billions of users and petabytes of data while maintaining 99.99% availability and strong security! 🚀