### PostgreSQL Distributed

**Architectures & Best Practices** 

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### Today's talk on PostgreSQL Distributed

Many distributed database talks discuss algorithms for distributed query planning, transactions, etc.

In distributed systems, trade-offs are more important than algorithms.

Vendors and even many papers rarely talk about trade-offs.

Many different PostgreSQL distributed system architectures with different trade-offs exist.

**Experiment**: Discuss PostgreSQL distributed systems architecture trade-offs by example.

### Single machine PostgreSQL

PostgreSQL on a single machine can be incredibly fast

No network latency Millions of IOPS Microsecond disk latency Low cost / fast hardware Can co-locate application server



### Single machine PostgreSQL?

PostgreSQL on a single machine comes with operational hazards

Machine/DC failure (downtime) Disk failure (data loss) System overload (difficult to scale) Disk full (downtime)



### PostgreSQL Distributed (in the cloud)

Fixing the operational hazards of single machine PostgreSQL requires a distributed set up.

The cloud enables flexible distributed set ups, with resources shared between customers for high efficiency and resiliency.



### Goals of distributed database architecture

Goal: Offer same functionality and transactional semantics as single node RDBMS, with superior properties

Mechanisms:Replication- Place copies of data on different machinesDistribution- Place partitions of data on different machinesDecentralization- Place different DBMS activities on different machines

Reality: Concessions in terms of <u>performance</u>, transactional semantics, functionality, and/or operational complexity

### PostgreSQL Distributed Layers

Distributed architectures can hook in at different layers — many are orthogonal!



Manual sharding, load balancing, write to multiple endpoints

Load balancing and sharding (e.g. pgbouncer, pgcat)

Transparent sharding (e.g. Citus, Aurora limitless), DSQL

Active-active, federation (e.g. BDR, postgres\_fdw)

DBMS-optimized cloud storage (e.g. Aurora, Neon)

Read replicas, hot standby

Cloud block storage (e.g. Amazon EBS, Azure Premium SSD)

### Practical view of Distributed PostgreSQL

Today we will cover:

- Network-attached block storage
- Read replicas
- DBMS-optimized cloud storage
- Transparent Sharding
- Active-active deployments
- Distributed key-value stores with SQL

#### Two questions:

1) What are the trade-offs?

Latency, Efficiency, Cost, Scalability, Availability, Consistency, Complexity, ...

2) For which workloads?

Lookups, analytical queries, small updates, large transforms, batch loads, ...

### The perils of latency: Synchronous protocol

Transactions are performed step-by-step on each session.



Max throughput per session = 1 / avg. response time

### The perils of latency: Connection limits

Max overall throughput: #sessions / avg.response time



Number of connections limited by app architecture

Number of processes limited by memory, contention

# Network-attached block storage



### Network-attached block storage



### Network-attached storage

#### Pros:

Higher durability (replication)

Higher uptime (replace VM, reattach)

Fast backups and replica creation (snapshots)

Disk is resizable

#### Cons:

Higher disk latency (~20µs -> ~1000µs)

Lower IOPS (~1M -> ~10k IOPS)

Crash recovery on restart takes time

Cost can be high

#### General guideline:

Always use, durability & availability are more important than performance.

# Read replicas



### Read replicas

Readable replicas can help you scale read throughput, reduce latency through cross-region replication, improve availability through auto-failover.



### Scaling read throughput

Readable replicas can help you scale read throughput (when reads are CPU or I/O bottlenecked) by load balancing queries across replicas.



### Eventual read-your-writes consistency

Read replicas can be behind on the primary, cannot always read your writes.



### No monotonic read consistency

Load-balancing across read replicas will cause you to go back-and-forth in time.



### Poor cache usage

If all replicas are equal, they all have the same stuff in cache



If working set >> memory, all replicas get bottlenecked on disk I/O.

### Read scaling trade-offs

#### **Pros:**

Read throughput scales linearly

Low latency stale reads if read replica is closer than primary

Lower load on primary

#### Cons:

Eventual read-your-writes consistency

No monotonic read consistency

Poor cache usage

#### **General guideline**:

Consider at >100k reads/sec or heavy CPU bottleneck, but avoid for dependent transactions and large working sets.

# DBMS-optimized storage

Like Aurora, Neon, AlloyDB



### DBMS-optimized storage

Cloud storage that can perform background page writes autonomously, which saves on write I/O from primary. Also optimized for other DBMS needs (e.g. read replicas).



### DBMS-optimized storage trade-offs

#### Pros:

Potential performance benefits by avoiding page writes from primary

No long crash recovery

Replicas can reuse storage, incl. hot standby

Less rigid than network-attached storage implementations (faster reattach, branching, ...)

#### Cons:

Write latency is high by default

High cost / pricing

PostgreSQL is not designed for it

#### General guideline:

Consider using for complex workloads, but measure whether price-performance under load is better than a bigger machine.

# Transparent sharding

Like Citus



### Transparent sharding

Distribute tables by a shard key and/or replicate tables across multiple (primary) nodes.

Queries & transactions are transparently routed / parallelized.



Tables can be co-located to enable local joins, foreign keys, etc. by the shard key.

### Single shard queries for operational workloads

Scale capacity for handling a high rate of single shard key queries:



Per-statement latency can be a bottleneck!

### Data loading in sharded system

Pipelining through COPY can make data loading a lot more efficient and scalable



### Compute-heavy queries

Compute-heavy queries (shard key joins, json, vector, ...) get the most relative benefit



select compute\_stuff(...) from users join items using (user\_id) where user\_id = 123 ...

### Multi-shard queries for analytical workloads

Parallel multi-shard queries can quickly answer analytical queries across shard keys:



select country, count(\*) from items, users where ... group by 1 order by 2 desc limit 10;

### Multi-shard queries for operational workloads

Multi-shard queries add significant overhead for simple non-shard-key queries



### Multi-shard queries for analytical workloads

Snapshot isolation is a challenge (involves trade-offs):



### Sharding trade-offs

#### Pros:

Scale throughput for reads & writes (CPU & IOPS)Scale memory for large working setsParallelize analytical queries, batch operations

#### Cons:

High read and write latency

Data model decisions have high impact on performance

Snapshot isolation concessions

#### General guideline:

Use for multi-tenant apps, otherwise use for large working set (>100GB) or compute heavy queries.

### Active-active

Like BDR, pgactive, pgEdge, ...



### Active-active / n-way replication

Accept writes from any node, use logical replication to asynchronously exchange and consolidate writes.



### Active-active / n-way replication

All nodes can survive network partitions by accepting writes locally, but no linear history (C<u>AP</u>).



### Active-active trade-offs

#### Pros:

Very high read and write availability

Low read and write latency

*Read* throughput scales linearly

#### Cons:

Eventual read-your-writes consistency

No monotonic read consistency

No linear history (updates might conflict after commit)

#### General guideline:

Consider only for simple data models (e.g. queues) and only if you really need the benefits.

## Distributed SQL

Like Yugabyte, CockroachDB, Spanner



### Distributed key-value storage with SQL (DSQL)

Tables are stored on distributed key-value stores, shards replicated using Paxos/Raft. Distributed transactions with snapshot isolation via global timestamps (HLC or TrueTime).



### Distributed key-value storage trade-offs

#### Pros:

Good read and write availability (shard-level failover)

Single table, single key operations scale well

No additional data modelling steps or snapshot isolation concessions

#### Cons:

Many internal operations incur high latency No local joins in current implementations Less mature and optimized than PostgreSQL **General guideline**: Just use PostgreSQL ;)

but for simple apps, the availability benefits can be useful

### Conclusion

PostgreSQL can be distributed at different layers.

Each architecture can introduce severe trade-offs. Almost nothing comes for free..

Keep asking:

What do I really want?

Which architecture achieves that?

What are the trade-offs?

What can my application tolerate? (can I change it?)



## Questions?

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