PA152: Efficient Use of DB 12. Replication and High Availability

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Credits

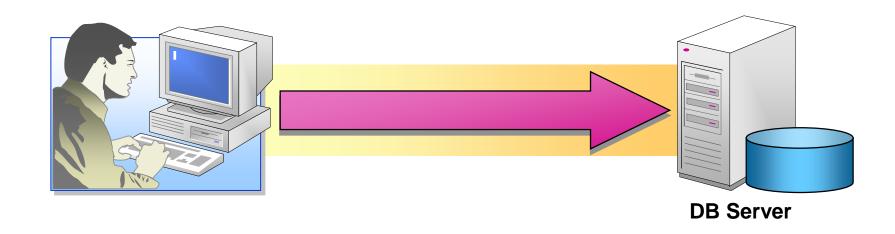
- This presentation is based on:
 - ☐ Microsoft MSDN library
 - □ Course NoSQL databases and Big Data management
 - Irena Holubová
 - Charles University, Prague
 - http://www.ksi.mff.cuni.cz/~holubova/NDBI040/
 - □ PostgreSQL documentation
 - http://www.postgresql.org/docs/9.3/static/highavailability.html

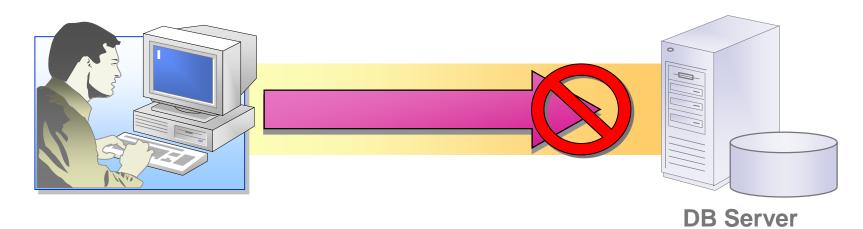


Contents

- Availability
- Data distribution & Replication
- High availability
- Failover
- Recommendations

Availability





Source: Microsoft



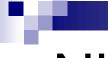
Determining Availability Requirements

- Hours of Operation
 - □ Business hours vs. all of the time
 - intranet service vs. web services
 - shift workers vs. all-around the world customers
- Connectivity Requirements
 - □ Online vs. offline applications so response time can be important!
- Tight/Loose coupling of app and DBMS
 - □ Synchronous vs. asynchronous data updates



Availability

- Definition in operation hours
 - □ Av = "up time" / "total time"
 - "up time" = the system is up and operating
 - More practical def.
 - Av = (total time down time) / total time
- Down time
 - □ Scheduled reboot, SW/HW upgrade, ...
 - Unscheduled HW/SW failure, security breaches, network unavailability, power outage, disasters, ...
 - □ Non-functional app requirements response time
- For "true" high-availability, down time is not distinguished



Nines

- Availability as percentage of uptime
 - \square Class of nines: $c = [-\log_{10}(1 Av)]$

Assuming 24/7 operation:

Nine class	Availability	Downtime per year	Downtime per month	Downtime per week
1	90%	36.5 days	72 hours	16.8 hours
2	99%	3.65 days	7.20 hours	1.68 hours
3	99.9%	8.76 hours	43.8 minutes	10.1 minutes
4	99.99%	52.56 minutes	4.32 minutes	1.01 minutes
5	99.999%	5.26 minutes	25.9 seconds	6.05 seconds
6	99.9999%	31.5 seconds	2.59 seconds	0.605 seconds
7	99.99999%	3.15 seconds	0.259 seconds	0.0605 seconds

Source: Wikipedia.org



Scalability

- Providing access to a number of concurrent users
- Handling growing amounts of data without losing performance
- With acceptable latency!



Need for Distributing Data

- Brings data closer to its user
- Allows site independence
- Separates
 - Online transaction processing
 - □ Read-intensive applications
- Can reduce conflicts during user requests
- Process big data



Scalability: Solutions

- Scaling Up vertical scaling → vendor dependence
 - Increasing RAM
 - Multiprocessing
- Scaling Out horizontal scaling
 - □ Replication
 - □ Read-only standby servers
 - Server federations / clusters / data distribution



Horizontal Scaling

- Systems are distributed across multiple machines or nodes
 - □ Commodity machines → cost effective
 - Often surpasses scalability of vertical approach
- Fallacies of distributed computing by Peter Deutsch
 - Network
 - Is reliable, secure, homogeneous
 - Topology does not change
 - Latency and transport cost is zero
 - Bandwidth is infinite
 - □ One administrator

Source: https://blogs.oracle.com/jag/resource/Fallacies.html



Brewer's CAP Theorem

- Consistency
 - After an update, all readers in a distributed system see the same data
 - All nodes are supposed to contain the same data at all times
 - E.g. in multiple instances, all writes must be duplicated before write operation is completed.
- Availability
 - □ Every request receives a response
 - about whether it was successful or failed
- Partition Tolerance
 - System continues to operate despite arbitrary message loss or failure of part of the system.



Brewer's CAP Theorem

Only 2 of 3 guarantees can be given in a "shareddata" system.

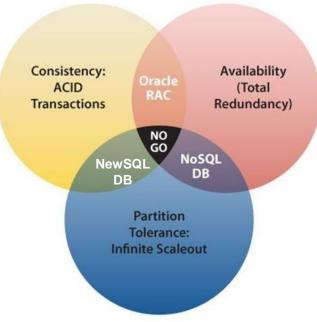
□ Proved by Nancy Lynch in 2002

ACID

- provides Availability and Consistency
- E.g. database on a single machine

BASE

 provides Availability and Partition tolerance



Source: http://bigdatanerd.wordpress.com

- Reality: you can trade a little consistency for some availability
- ☐ E.g. distributed database



NewSQL

- Distributed database system that scales out
- CP system
 - trades availability for consistency when partition happens
- MySQL cluster, Google Spanner, VoltDB, ...
 - In fact, master-master replication with data sharding

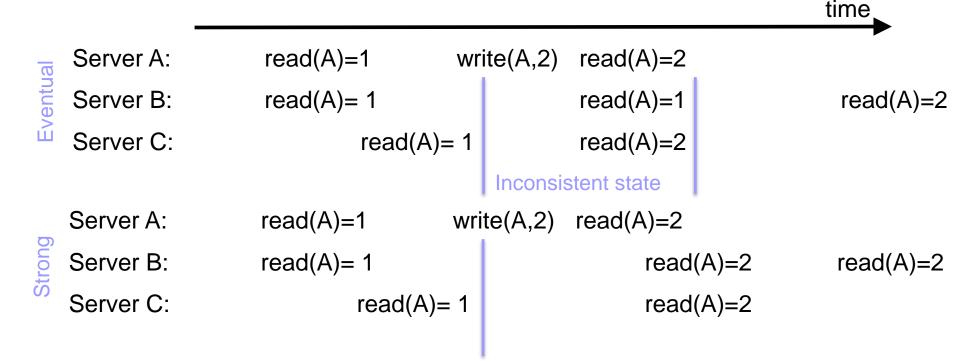
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BASE Properties

- Basically Available
 - Partial failures can occur, but without total system failure
- Soft state
 - □ System is in flux / non-deterministic
 - Changes occur all the time
- <u>Eventual consistency</u> (replica convergence)
 - □ is a liveness guarantee
 - reads eventually return the same value
 - □ is not safety guarantee
 - can return any value before it converges



- Strong (ACID) vs. Eventual (BASE) consistency
- Example:

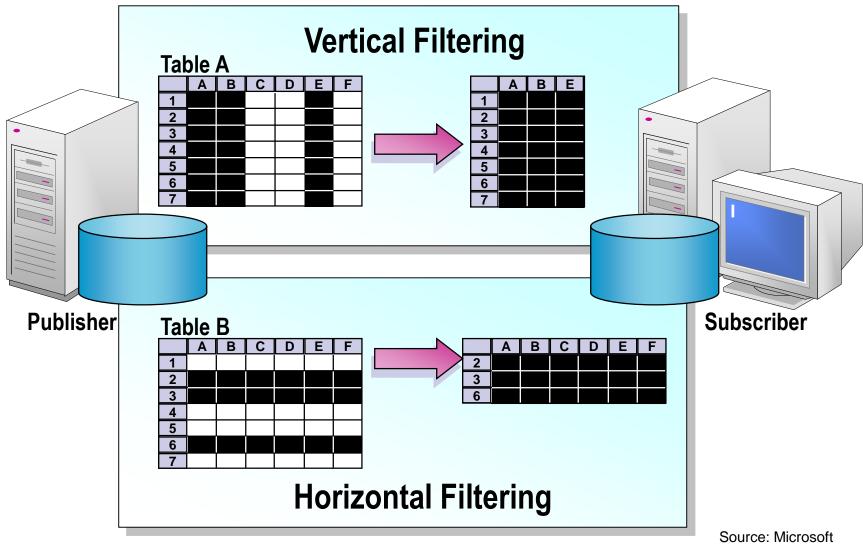




Replication / Distribution Model

- Model of distributing data
 - □ Replication
 - The same data stored in more nodes.
 - □ Filtering data (sharding)
 - The data is partitioned and stored separately
 - Helps avoid replication conflicts when multiple sites are allowed to update data.

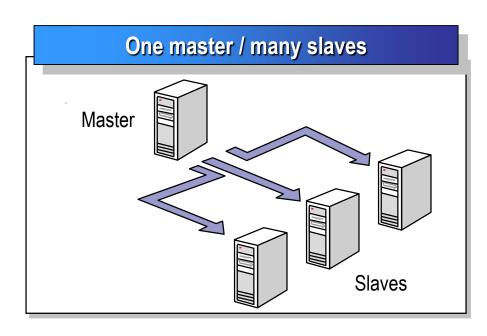
Filtering Data





Distribution Model: Replication

- Master-slave model
 - □ Load-balancing of read-intensive queries
- Master node
 - □ manages data
 - ☐ distributes changes to slaves
- Slave node
 - □ stores data
 - □ queries data
 - no modifications to data

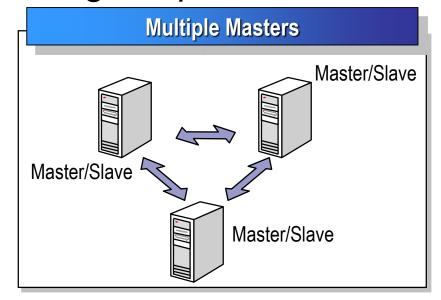


Distribution Model: Replication

- Master-master model
 - □ Typically with filtering data
 - Master for a subset of data
 - Slave for the rest

Consistency needs resolving of update

conflicts



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Master-master Model



Orders (Master A)				
Primary Key				
Area	ld	Order_no	Qty	
1	1000	~	15	
1	3100	~	22	
2	1000	~	32	
2	2380	~	8	
3	1000	~	7	
3	1070	~	19	



Master/Slave

Master/Slave

Orders (Master B)					
Primary Ke	ЭУ				
Area	ld	Order_no	Qty		
1	1000	~	15		
1	3100	~	22		
2	1000	~	32		
3	2380	~	8		
3	1000	~	7		
3	1070	~	19		

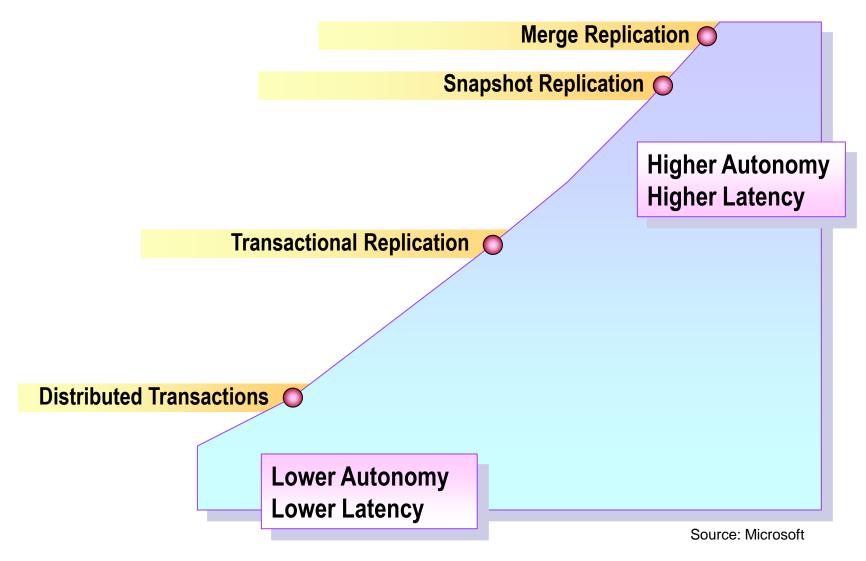


Master/Slave

Orders (Master C)					
Primary Key					
Area	ld	Order_no	Qty		
1	1000	~	15		
1	3100	~	22		
2	1000	~	32		
2	2380	~	8		
3	1000	~	7		
3	1070	~	19		

Source: Microsoft







Replication Types

- Distributed Transactions
 - For "real" master-master model, ensures consistency
 - □ Low latency, high consistency
- Transactional Replication
 - □ Replication of incremental changes
 - Minimal latency (typically online)
 - Conflicts solves using shared locks



Replication Types

- Snapshot Replication
 - Periodic bulk transfer of new snapshots of data
 - Intermediate updates to data might be unnoticed by "subscribers"
 - □ Data changes substantial but infrequent
 - □ Slaves are read-only
 - ☐ High latency is acceptable



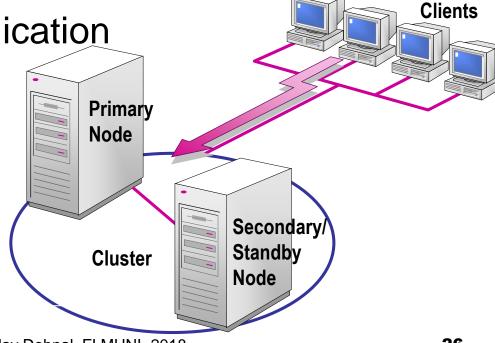
Replication Types

- Merge Replication
 - Autonomous changes to replicated data are later merged
 - Does not guarantee transactional consistency, but converges
 - □ Default and custom conflict resolution rules
 - □ Adv: Nodes can update data offline, sync later
 - □ Disadv: Changes to schema needed.



Maintaining High-Availability

- Standby server
 - □ Shared disk failover (NAS)
 - ☐ File system replication (DRBD)
 - □ Transaction log shipping
 - □ Trigger-based replication
 - □ Statement-Based Replication Middleware



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Log-shipping Standby Server

- Also called warm standby
- Primary node
 - serves all queries
 - □ in permanent archiving mode
 - Continuous sending of WAL records to standby servers
- Standby server
 - serves no queries
 - □ in permanent recovery mode
 - Continuous processing of WAL records arriving from primary node
- Log shipping can be synchronous/asynchronous
- Disadvantage: all tables are replicated typically
- Advantage: no schema changes, no trigger definitions

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Failover

- If primary fails, standby server begins failover.
 - Standby applies all WAL records pending,
 - marks itself as primary,
 - □ starts to serve all queries.
- If standby fails, no action taken.
 - After becoming online, catch-up procedure is started.
- Heartbeat mechanism
 - to continually verify the connectivity between the two and the viability of the primary server

Failover

- Failover by standby succeeded
 - New standby should be configured
 - □ Original primary node becomes available
 - □ → inform it that it is no longer the primary
 - do so-called STONITH (Shoot The Other Node In The Head),
 - otherwise serious data corruption/loss may occur
 - □ Typically old primary becomes new standby



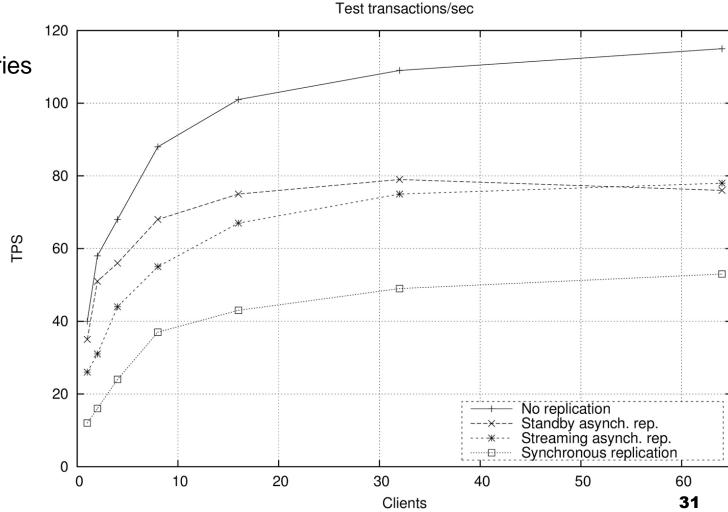
Primary and Standby Servers

- Swap primary and standby regularly
 - □ To verify recovery steps
 - □ To do necessary maintenance on standby server
 - SW/HW upgrades, ...

PostgreSQL: Replication

■ TPC Benchmark B

scale factor 1 1 trans. = 5 queries





Recommended Practices

- Maximize availability at each tier of the application
- Keep standby servers on a different subnet
- Independent power supply to the primary server
- Test whether your availability solution works