VoltDB Things you learn as you massively scale...

VOLTDB

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17-Jul-18

VOLTDB

Scaling at the Architectural Level...

How many servers will you need to start?

- HA implies more than one machine
 - With only 2 nodes you need 100% spare capacity
 - With 3 50% spare, with 4 33% spare...
 - So: Don't assume a cluster of two 'monster' servers is optimal.
- Something will be a driving factor. Do not guess this measure it!
 - HA
 - RAM
 - CPU
 - Network
- You may not be able to dictate the size of servers...
 - Example: AWS may require a certain size node for an adequate network
 - Reality check: "Someone Else's Cloud" will have its own selection of available size.



How many servers will you need eventually?

How many spare copies do you need?

- As the number of machines goes up the chance of a failure goes up...
- You have 1 spare copy of data but what if both copies are lost because you lost 2 out of 20 servers?
- Eventually you'll need two spares. When is dependent on your level of paranoia...
- Hybrid approach is to have 'wallflower' nodes that will rapidly join cluster
 - Reduces time spent with only 1 copy from hours to minutes
- Do you reject peak traffic or size for it?
 - You do have a plan for peaks, don't you?



Will you need multiple sites?

- Historically Active-Active was 'science fiction'
- Now it's a common requirement
- Motivation
 - Survivability
 - Latency
 - Ego
- Doesn't help you scale
 - Everybody has to find out about every transaction everywhere
 - Going from Active-Active to Active-Active-Active implies extra work even if new site does nothing



How do you partition the data?

You mean we have to partition?

- For low latency environments with writes partitioning is unavoidable.
- Pick the least awful partition key...
- VoltDB's Materialized views can help...
- Eventual Consistency isn't
 - Side effects of inconsistent reads will propagate way beyond the database before data is made consistent.
- Do you reject peaks or size for them?



Broader Implications...

- System is too complicated to do testing on a laptop:
 - RAM
 - Network
 - CPU
 - ...all non trivial
- Development and Testing costs will spike
- Problems with behavior changing between Dev and Test
- Problems with emulating connected systems in Test



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Scaling Write Intensive Workloads...

Scaling "Writes" isn't like scaling "Reads"...

- Traditionally we scale by adding more of *whatever* is most needed.
- So commodity hardware is great at scaling reads, as reads need CPU, RAM etc
- Some writes scale well e.g. if they are inherently unique and disconnected from anything else.
- But if writes need to be ACID we can't simply have two separate updates to two copies in two places.
- The bottleneck is not a physical resource.
- In this case "Whatever is most needed" is the data itself.
 - Implies you can't solve this problem with hardware



If we tried DB write strategies in a supermarket...

Row Level Locking: Nobody can touch the Orange Juice shelf or any other shelf I'm taking things from until I've finished shopping and checked out!

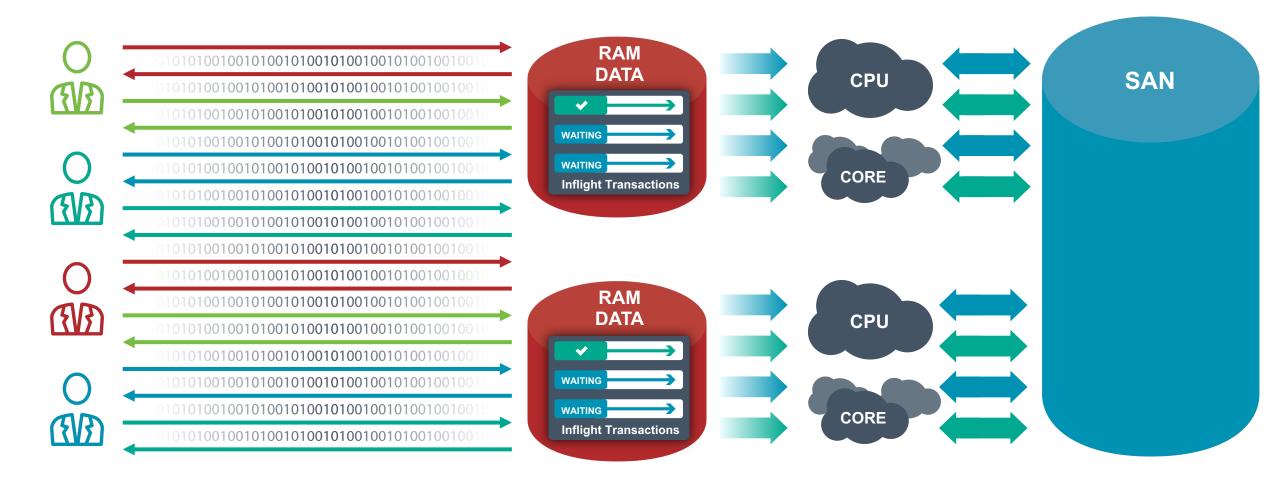
Eventual Consistency: I take Orange Juice, then pay for it, but it vanishes from my shopping cart and moves to someone else's as I put my bags in my car. The staff deny this happened.

Optimistic Updates: I buy my Orange Juice but are pulled over by security as I attempt to drive away. They refund my money and take the Juice off me, then tell me to try again.



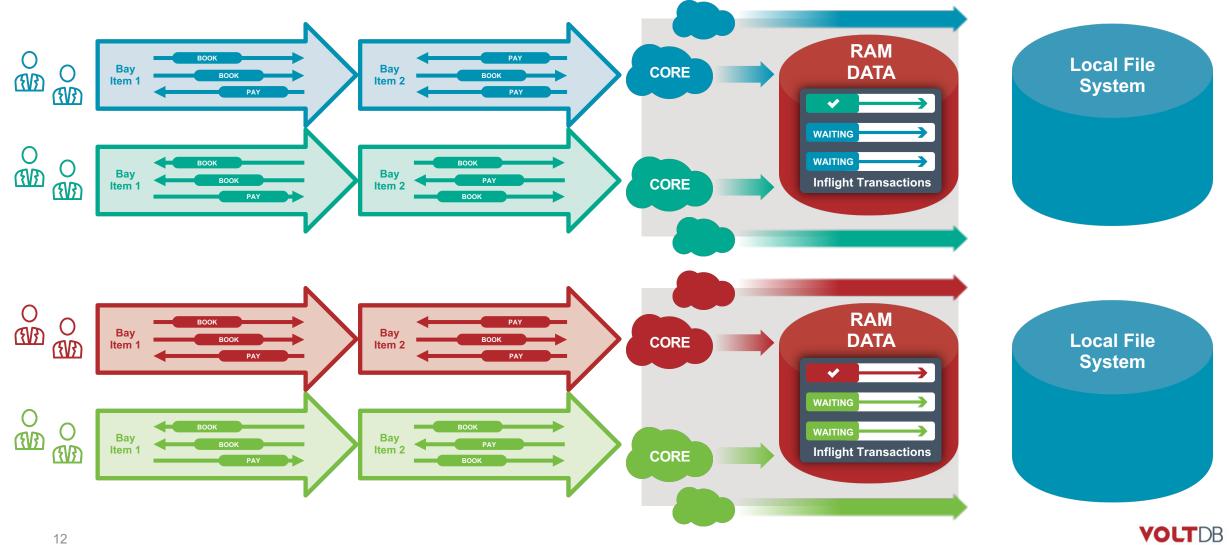


RDBMS - What Actually Happens – Part 2





How VoltDB works



VOLTDB

Scaling in the real world...

Or "6 things I wish I knew before I started"

1. Ludic Fallacy



"Ludic Fallacy" – Mistaking a game for reality...

Our model can never perfectly match reality.

Which means that no matter how 'well trained' it is, there will be a scenario which the model oversimplifies or otherwise fails to cope with.



1. Ludic Fallacy – An Example

Mapping apps are reportedly directing people fleeing the Southern California wildfires to areas that are on fire



- Out-of-control wildfires are raging in and near Los Angeles.
- Mapping apps, which are frequently designed to help users route around traffic, are in some cases reportedly directing drivers into fire-affected zones.

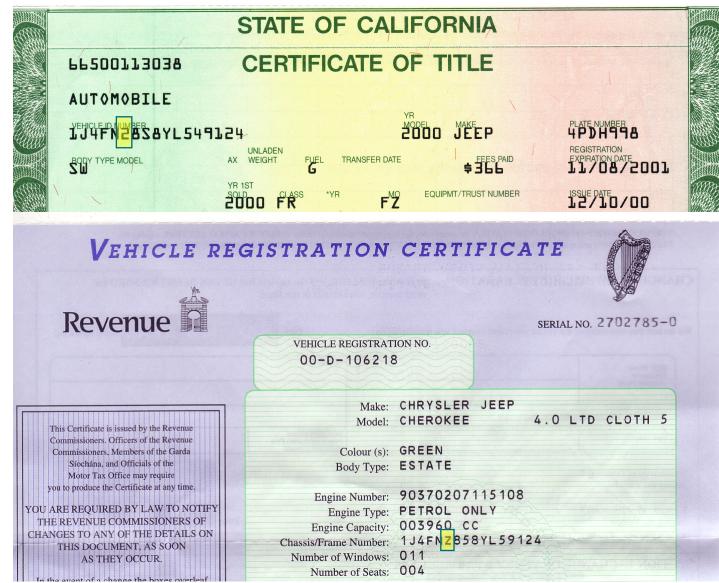


Emergency crews in Southern California block a roadway as flames spread from a wind-driven brush fire. REUTERS/Gene Blevins

 The Los Angeles Police Department is warning those in the area to cease using such apps.



2. Your Data Is Always Slightly Wrong



Real world data streams are always imperfect.

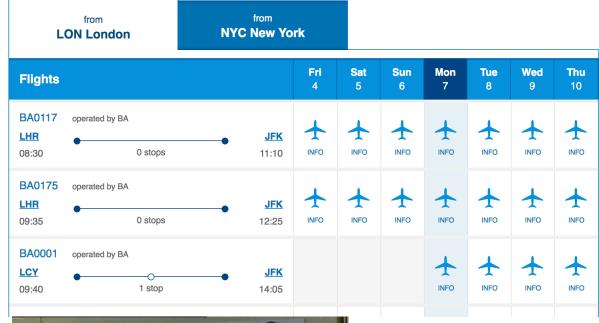
Example: The chassis / VIN number of an automobile can <u>never</u> change, <u>ever</u>!

Information about the 'ghost' vehicle went was sent to the police, insurance industry, stats agency....



3. Merging multiple data streams is hard

Goal: Predict flight delays.





"The Late Arrival Of The Incoming Aircraft"

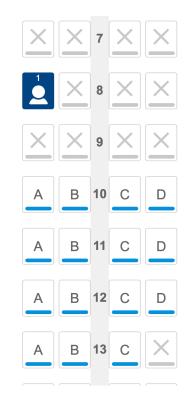


Raw TAF

KJFK 070809Z 0708/0812 36004KT P6SM SCT025 BKN040 FM071400 04009KT P6SM SCT035 BKN050 FM071800 15010G15KT P6SM SCT035 BKN050 FM080100 09009KT P6SM SCT030 BKN100 FM080900 05005KT P6SM SCT020 SCT100

Raw METAR

KJFK 070951Z 35006KT 10SM FEW060 BKN250 13/11 A3000 RMK AO2 SLP159 T01280106 KJFK 070851Z 35005KT 10SM FEW060 BKN250 12/11 A2998 RMK AO2 SLP152 T01220106 53013 KJFK 070751Z 36004KT 10SM FEW055 BKN250 13/11 A2996 RMK AO2 SLP146 T01330106 KJFK 070651Z 36008KT 10SM SCT024 BKN055 14/11 A2996 RMK AO2 SLP144 T01440111



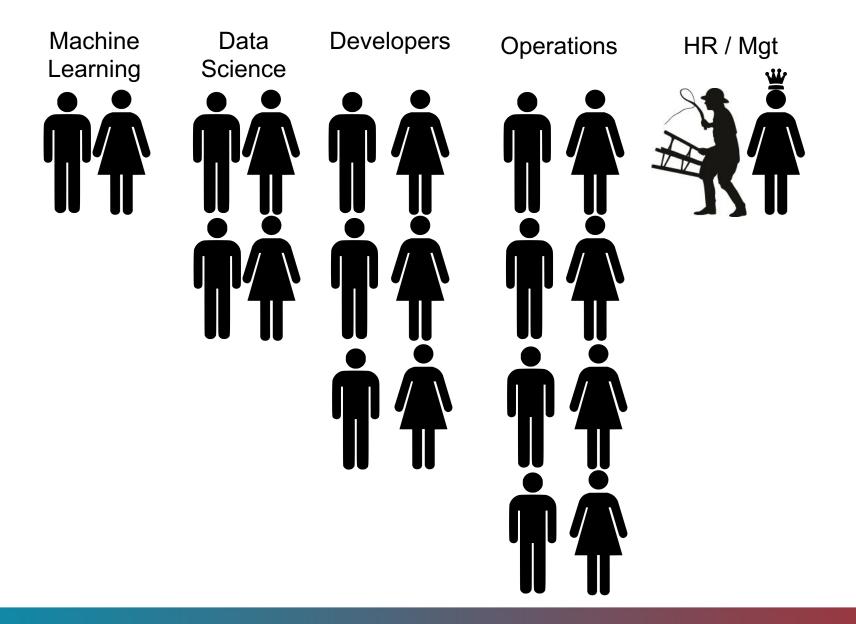


4. As volumes increase, life will get much harder.





5. Loading the data will never finish





6. What happens if time is of the essence?

Traditional Batch / Hadoop Speed: 30 Minutes

Web Server : 3-7 Seconds

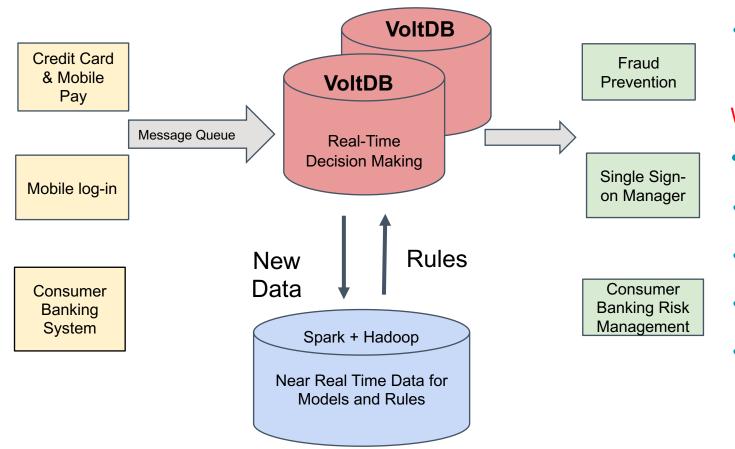
Spark / Kafka : 1-2 Seconds

Traditional OLTP: 5-50 ms

5G Phone Network / VoltDB: 1ms







Application/Use Case

- Fraud Prevention
- Single sign-in of all Huawei phones
- Consumer banking risk management

Why VoltDB?

- > 50% reduction in fraud cases
- >\$15M/year saved from fraud loss
- 10k complex Transactions Per Second
- 99.99% transactions finish < 50ms
- 10x better performance than

traditional fraud detection