How Event-Sourcing & Serverless have super-powered E-commerce systems

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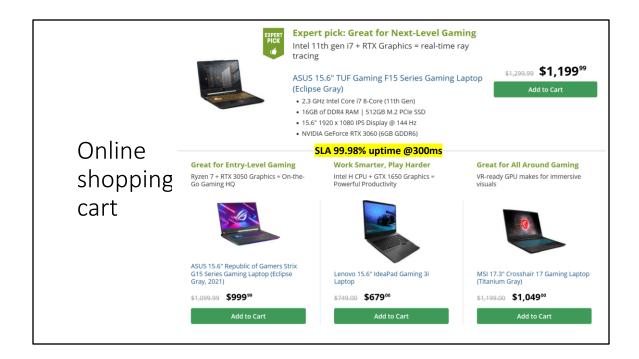
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Agenda

- E-commerce: It's complicated
- Modern principles event sourcing, serverless, change data capture, etc
- A basic implementation in Azure
- Event sourcing discussions and questions (Other topics too)





This is a typical online shopping cart

Has been around for over 2 decades and pretty much looks similar on most ecommerce websites

Customers browse the website for electronics, filter by laptops and then land on this page which

Shows some recommendations, probably powered by some nefarious tracking activity of the customer all over the internet

Item availability is served via an API

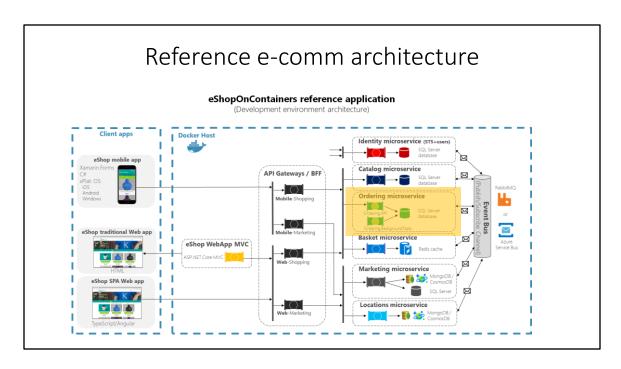
When she checks her fav ecomm website it can't be down or take too long to load Because competitors website is just a tab away SLA of 99.98% 2 hrs downtime per year, 300ms latency

I will dig into one small piece of functionality Customer wants to buy a laptop

The general process here is customer adds the laptop to cart

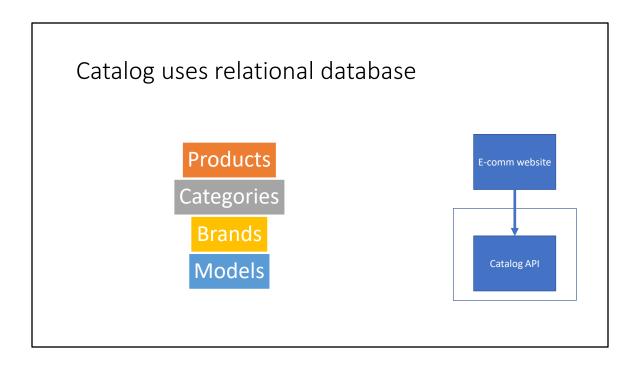
When ready customer checks out online, sets delivery address and pays for the laptop

Most e-commerce system really don't reserve a product until you go into the checkout page
We're freezing at the moment you're about to checkout
IS THE LAPTOP STILL AVAILABLE?
Is it available either online or in a nearby store?



We're used to looking at architectures like eshopcontainers which are pretty awesome for small to medium size businesses You have identity, catalog, ordering, basket and other microservices. But when it comes to larger e-commerce systems each component gets more complicated

And if you're not careful even these can get real gnarly For instance let's drill into the ordering microservice

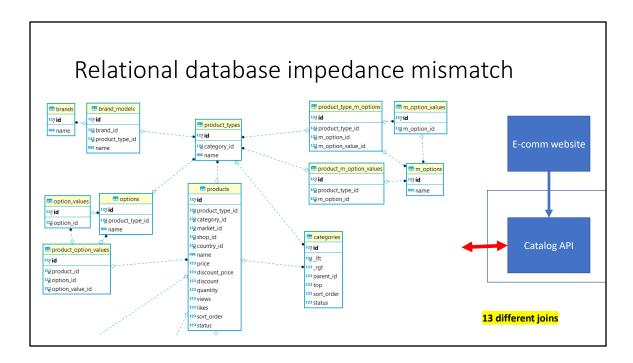


It has to check with all this data:

Products Categories Brands Models

Just to name a few

So if you're using a relational DB, what does this look like?



Data is stored is in a disassembled form aka shredded

For each product item in the list

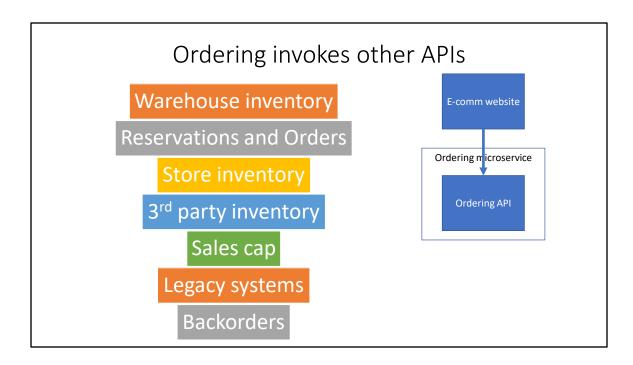
data is read by reassembling objects

This is the object-relational "impedance mismatch."

The workaround is object-relational mapping frameworks, which are inefficient at best, problematic at worst.

In this example at the very least you have to invoke data from a minimum 13 different tables every time you product details

So you have a request traveling from the website to the services to 13 different tables and then back

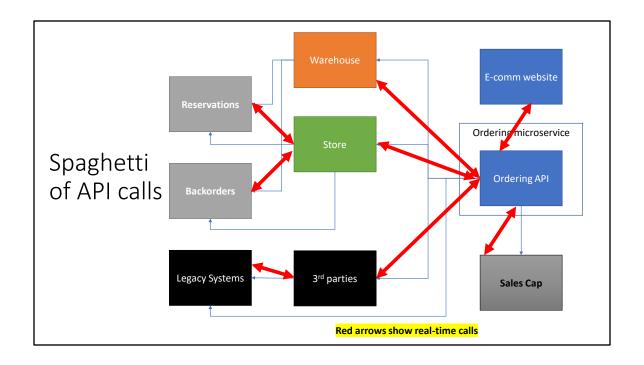


In this example the real life version of the ordering microservice is actually an entire org

You hear about supply chain these days all the time: When the e-comm website invokes the global ordering API It has to check with all these microservices:

Warehouse inventory
Existing reservations and orders
Store inventory – floor and backroom
3rd party inventory – partners and suppliers
Sales cap – like seasonal limits for XBoxes
Legacy systems which may also contain duplicates that need to be checked
Backorders that were not fulfilled

So what does this look like?



This is somewhat it looks like

Each is represented by a service (each box is a system)

Each may be maintained by one or more teams,

Each has it's own service or more likely multiple services and databases

What happens when customer wants to place the order for the the laptop Ordering API call

Which calls warehouse inventory

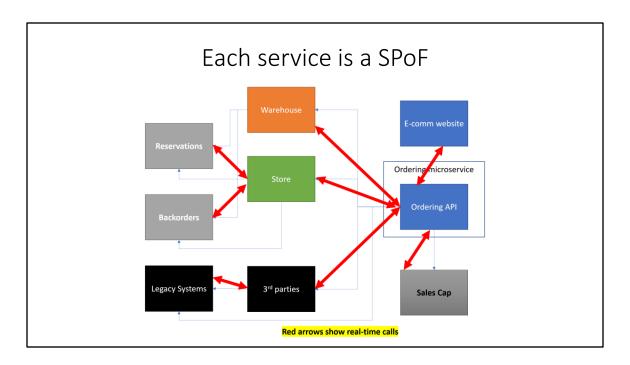
Not finding anything it calls the store inventory where it finds a match

But we immediately have to check the reservations and backorders to make sure the item is still available

And then we have to check 3rd party API which has to make a call to legacy systems in this case to check for duplicates

After all said and done we still have to check to make sure the item falls within sales cap

We make multiple calls including a ton of synchronous API calls until we can finally compute the answer

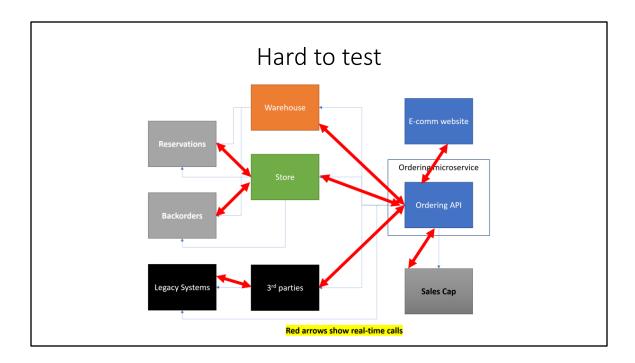


Some of the problems with this architecture

System has 99.98% uptime and 300ms response means each of these systems has to be about better than that

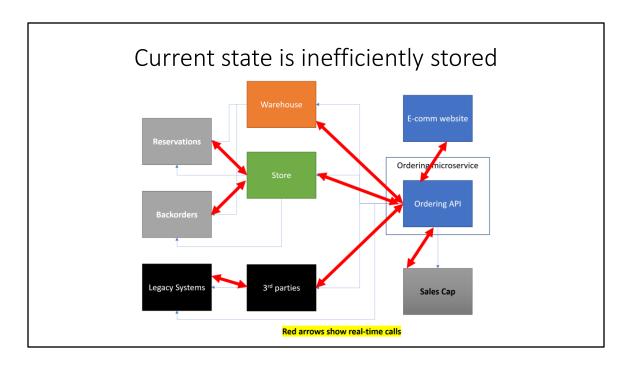
Which means 5 9s of uptime and around 20-50ms latency

Additionally for every call, each of these services needs to work correctly Even if one service goes down you risk giving the wrong answer to the customer Aka one service takes down the entire system



Hard to test the entire dependency tree

Sure you can do unit tests which covers a small unconnected surface the integration tests are a nightmare, hard to execute, complicated and expensive



Each of these systems was designed internally in a traditional manner

As changes happen the state of each system, usually in relational databases has to be mutated in place

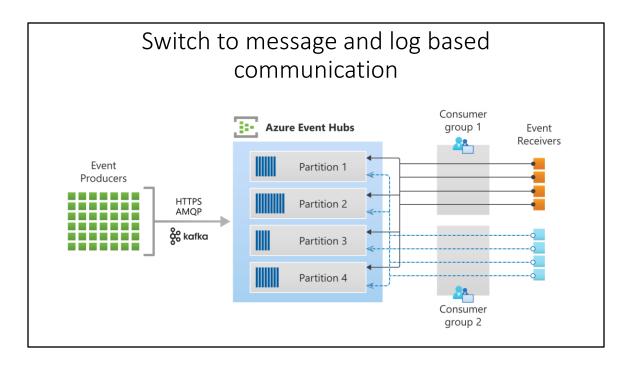
We're assuming that there are no outages and everything stays up all the time

What happens when a component goes down or a bug is introduced

This is code, distributed unreliable, expensive code

To put it bluntly, servers are not reliable and we have to account for downtime

Modern principles – event sourcing, serverless, change data capture, caching, etc

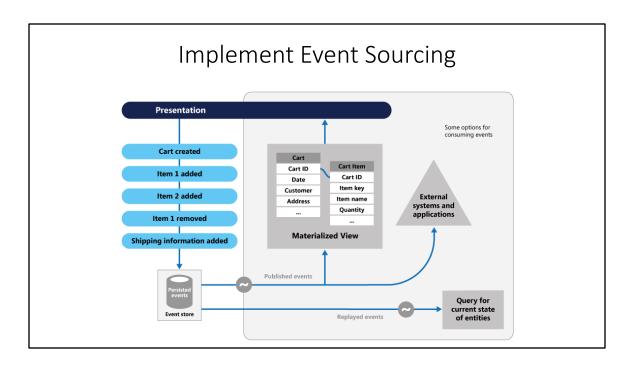


Systems comm with each other over http or messages using queue Log based pub sub systems like Event Hub and Kafka messages — order and retained for later consmption using checkpoiints

If you have an outage or bug, you can reset checkpoint to point in time before the issue

Replace http API calls with event hub or kafka

What problem does this solve? Drastically improves mean time to recovery



Events are facts about something that happened in the world

Doesn't change by fact already happened

Events are 1st class citizens

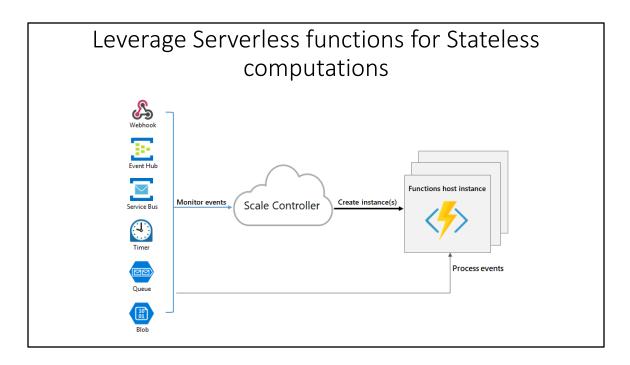
Dat source is ordered set of events

Current state is ordered set of events – aggregating stream

Bank account is result of every depostit and withdraawael

Time machine you can see the state for any point in time and walk step by step for every state

What problem does this solve? Easy to replicate state at any point in time. Good for debggin and analytics



All business logic is stateless funitons with zero external dependencies

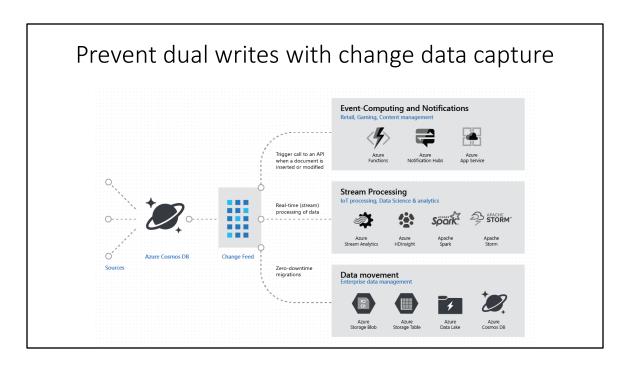
Collect all state you need up front and pass into code as parameters
The function is Predictabile and atomic in nature
No random outcomes or partial results
Real world failures will keep code from running but won't cause uncertain state

Domain logic Easy to unit test wide set of scenarios Stateless code is easy to test Integration tests only needed for testing connection between services

You must also consider a service like Function for the following reasons:

- 1. Atomic in operation
- 2. Scaleable
- 3. Integration code is abstracted away

You're getting some state values, you run your calculation and provide results



Once business logic is complete then you have to write

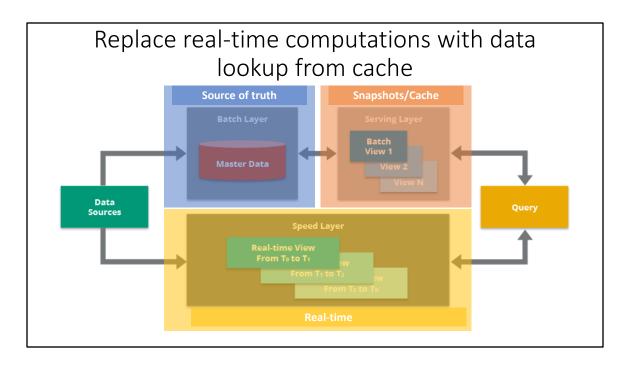
Write-once (source of truth)

Don't write to a database and then notify downstream consumers in the same process about changes – don't

Dual write exposes us to failure and inconsistent state

Change data capture is the safe way to do this – only after commit event is published You will never lose an event

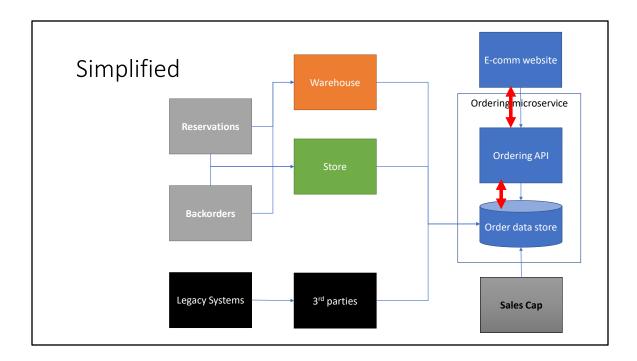
Cosmos db change feed



When you know possible set of inputs in advance Instead of real-time computation, pre-computed cache of results Real-time is expensive

This is an example of lambda architecture

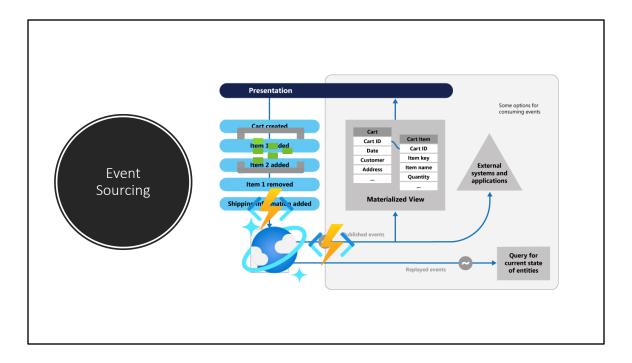
Save snapshots, builds updated stream and updates cache pUblish snapshots as they occur Downstream can consume real time or snapshots Embrace eventual consistency



Real time calls in red

Using duality of code and data,
Stream events and messages over Event Hub
Instead of dependent service pulling the info,
Source system pushes data changes
From service arch to event driven arch
All components are async
Messages flowing from left to right
Trading real time comp to pre-computed data for supply chain
Hot path to calculate
2 service calls with 3 9s SLA with 150 ms

Event sourcing implementation in Azure with Serverless

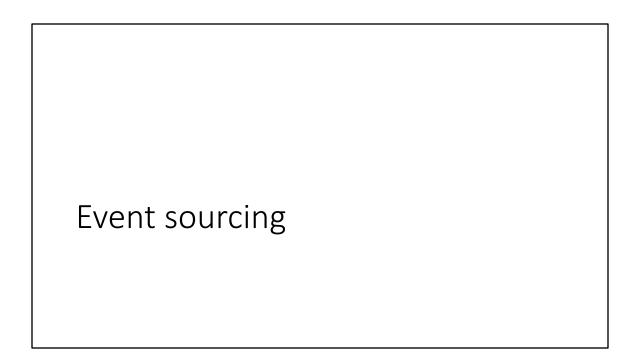


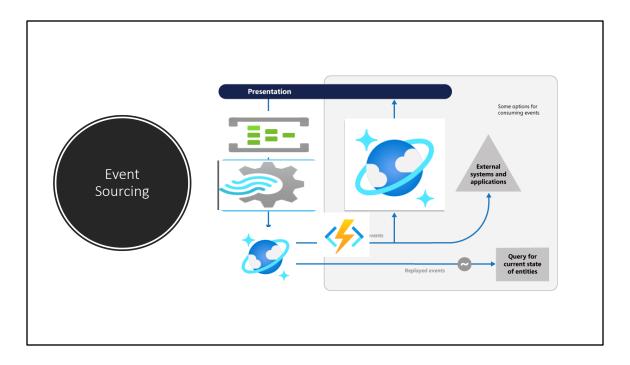
Instead of storing just the current state of the data in a domain, use an append-only store to record the full series of actions taken on that data. The store acts as the system of record and can be used to materialize the domain objects.

This can simplify tasks in complex domains,

by avoiding the need to synchronize the data model and the business domain, while improving performance, scalability, and responsiveness.

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