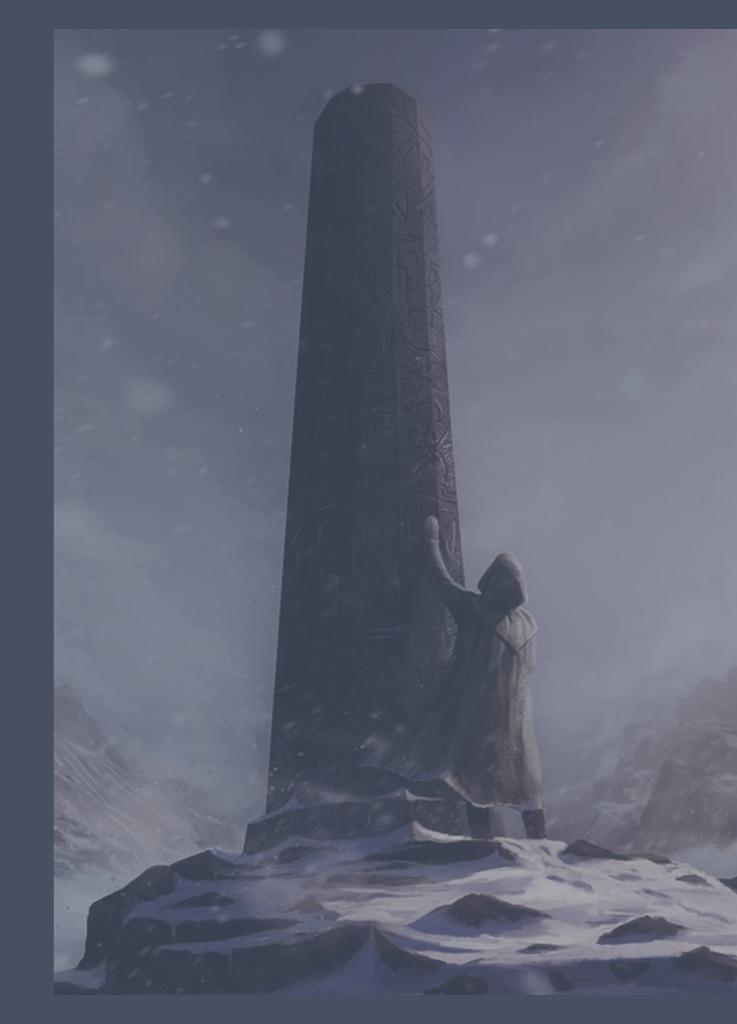
### FROM TO MCROSYSTEMS JONAS BONER **@JBONER** LIGHTBEND

## WE HAVE BEEN SPOILED BY THE ALMIGHTY NONOLITH



## KNOCK. KNOCK. WHO'S THERE?

## WE CAN'T MAKE THE HORSE FASTER

51

GEC. H. HOLZBOG JEFFERSONVILLE, IND.

### WE NEED CARS FOR WHERE WE ARE GOING

### BUT DON'T JUST DRINK THE

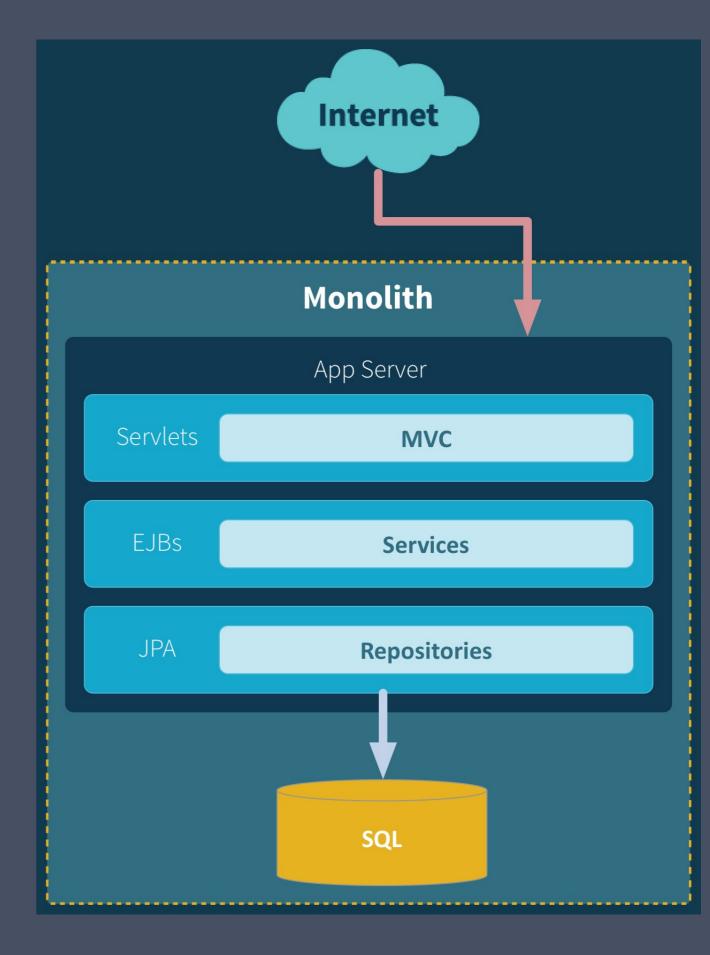


#### THINK FOR YOURSELF

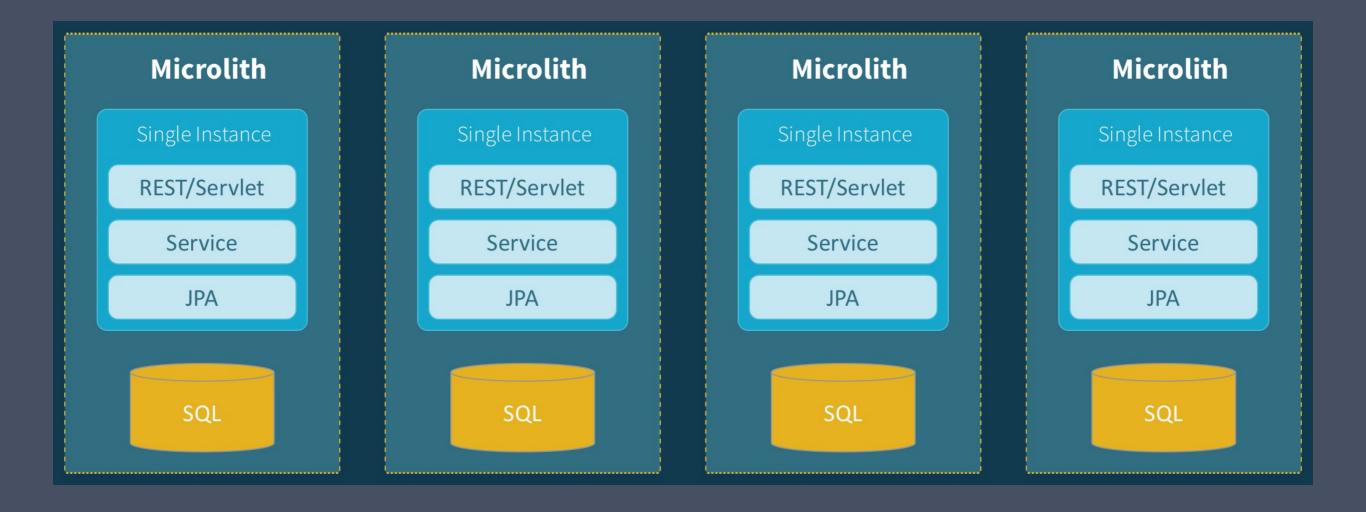
# NO ONE WANTS **NCROSERVICES** IT'S A NECCESSARY EVIL

### ARCHITECTURAL CONTEXT OF MICROSERVICES: DISTRIBUTED SYSTEMS

#### LET'S SAY THAT WE WANT TO SLICE THIS MONOLITH UP



#### TOO MANY END UP WITH AN ARCHITECTURE LIKE THIS



MCROLITH: SINGLE INSTANCE MICROSERVICE -NOT RESILIENT -NOT ELASTIC



ONE ACTOR IS NO ACTOR. ACTORS COME IN SYSTEMS. - CARL HEWITT



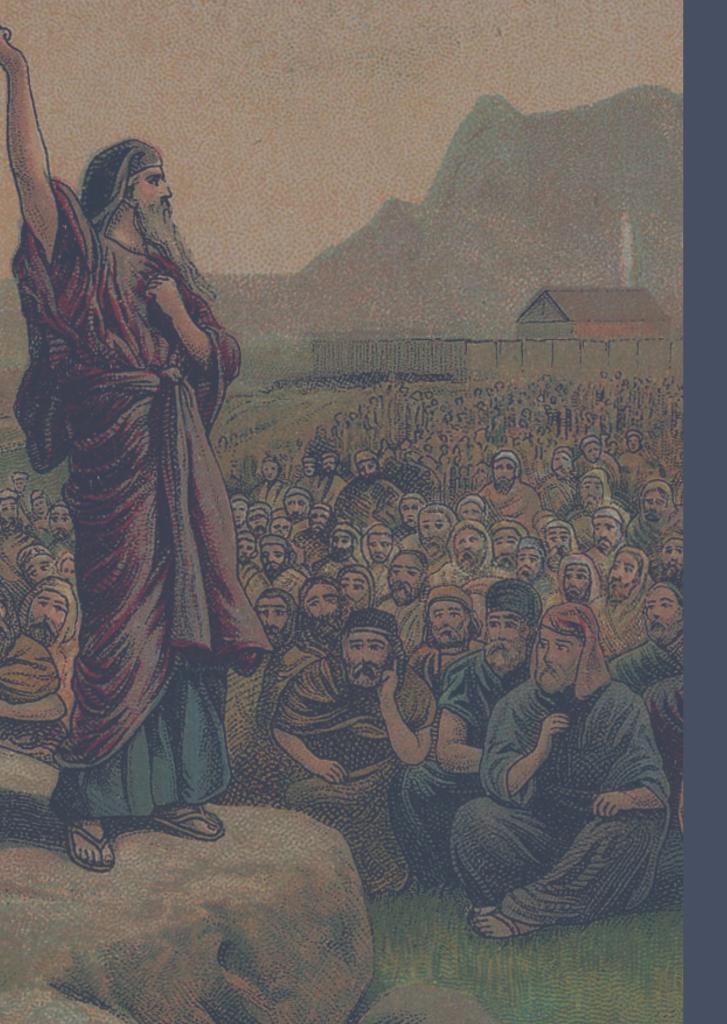
# MICROSERVICES COME IN SYSTEMS

AS SOON AS WE **EXIT THE SERVICE** WE ENTER A WILD OCEAN OF NON-DETERMINISM THE WORLD OF **DISTRIBUTED SYSTEMS** 



#### SYSTEMS NEED TO EXPLOIT REALITY





#### EMBRACE REALITY AND ITS CONSTRAINTS SHALL SET YOU FREE



## INFORMATION HAS LATENCY

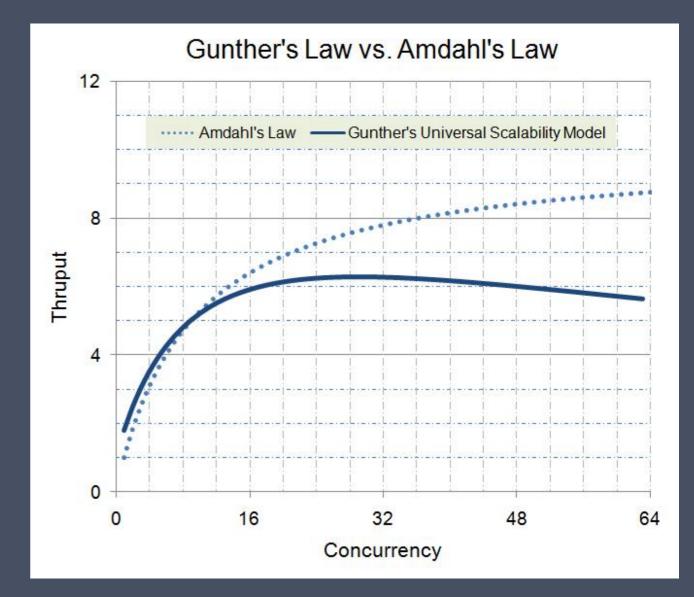
THE CONTENTS OF A MESSAGE ARE **ALWAYS FROM THE PAST!** THEY ARE NEVER **NOW**. - PAT HELLAND



#### WE ARE ALWAYS LOOKING INTO THE PAST



## THE COST OF MAINTAINING THE OF AN ABSOLUTE



## AS LATENCY GETS HIGHER. THE ILLUSION CRACKS EVEN MORE

IN A **DISTRIBUTED** SYSTEM, YOU CAN KNOW WHERE THE WORK IS DONE OR YOU CAN KNOW WHEN THE WORK IS DONE BUT YOU CAN'T KNOW BOTH - PAT HELLAND



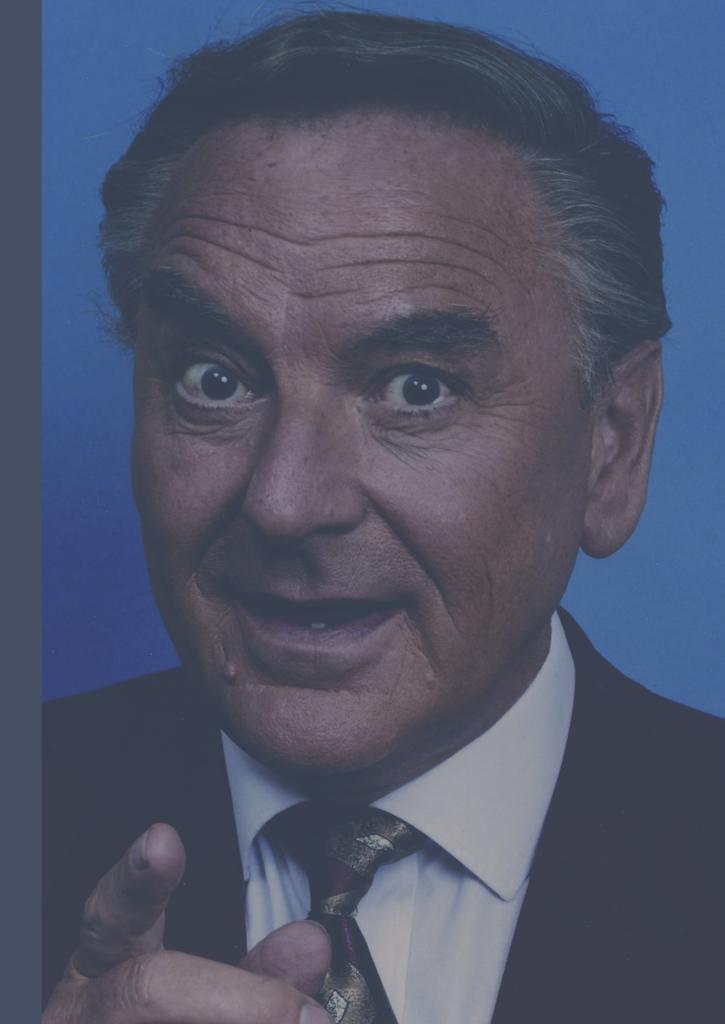
# STRIVE TO MINIMIZE COUPLING & COMMUNICATION

WORDS ARE VERY UNECESSARY. THEY CAN ONLY DO HARM. ENJOY THE SILENCE BY MARTIN

- ENJOY THE SILENCE BY MARTIN Gore (Depeche Mode)



#### SILENCE IS NOT ONLY GOLDEN, IT IS SELDOM MISQUOTED. - BOB MONKHOUSE



## WE HAVE TO RELY ON EVENTUAL CONSISTENCY BUT RELAX IT'S HOW THE WORLD WORKS

## NO ONE WANTS EVENTUAL CONSISTENCY. IT'S A NECESSARY EVIL IT'S NOT COOL. IT'S USEFUL.

## TWO HELPFUL TOOLS **1. REACTIVE DESIGN 2. EVENTS-FIRST DDD**

## REACTIVE PROGRAMMING VS REACTIVE SYSTEMS

### REACTIVE PROGRAMMING CAN HELP US MAKING THE INDIVIDUAL SERVICE INSTANCE HIGHLY PERFORMANT AND EFFICIENT

## REACTIVE SYSTEMS CAN HELP US BUILDING DISTRIBUTED SYSTEMS THAT ARE ELASTIC AND RESILIENT





**Viktor Klang** @viktorklang

"To get consistent & fast response in a real time system, you've to work asynchronously [...] never wait on something happening." @hintjens



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#### ASYNCHRONOUS

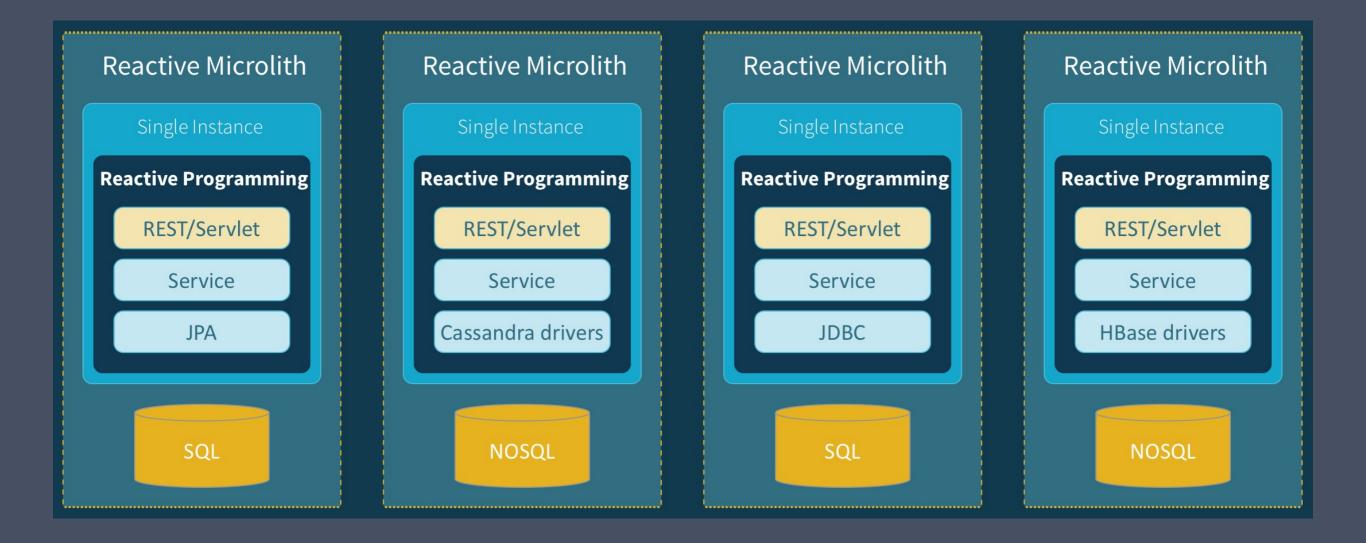
### ASYNC IO-IS ABOUT NOT BLOCKING THREADS ASYNC COMM-IS ABOUT NOT BLOCKING REQUESTS

# GO ASYNCHRONOUS BION-BLOCKING - MORE EFFICIENT USE OF RESOURCES - MINIMIZES CONTENTION ON Shared Resources

#### ALWAYS APPLY BACK-PRESSURE A FAST SYSTEM SHOULD NOT OVERLOAD A SLOW SYSTEM



#### LET'S APPLY REACTIVE PROGRAMMING TO OUR MICROLITHS



#### WE NEED TO EXTEND OUR MODELS OF

#### **1. ASYNCHRONOUS MESSAGING** (N-M)**2. STREAMING (1-1) 3. SYNCHRONOUS REQUEST/ REPLY (1-1)**



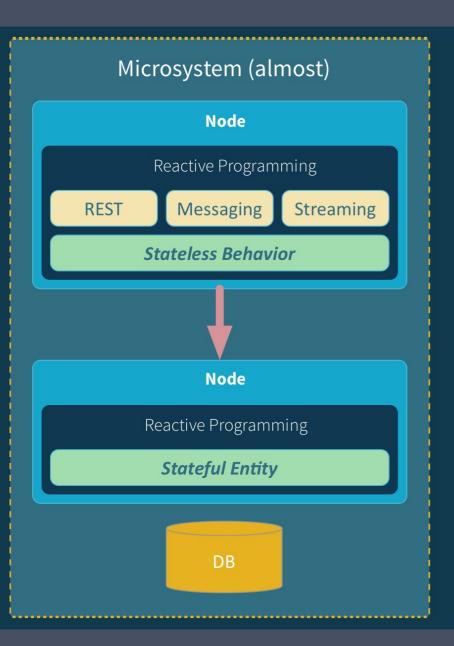
## WE'RE GETTING THERE. BUT WE STILL HAVE A SINGLE INSTANCE MICROSERVICE -NOT SCALABLE -NOT RESILENT

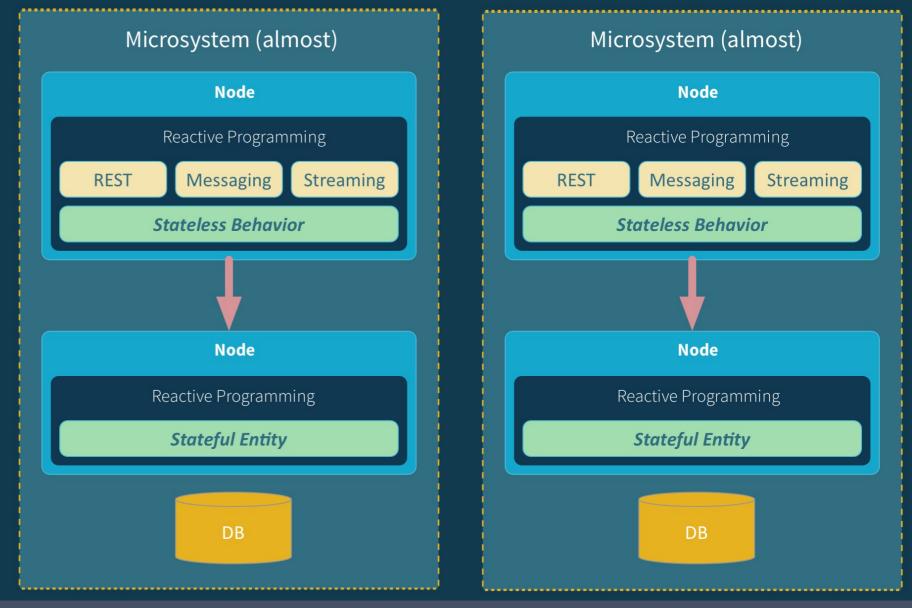
## MICROSERVICES COME AS SYSTEMS

## EACH MICROSERVICE NEEDS BE DESIGNED AS A DISTRIBUTED SYSTEM A MCROSYSTEM

## WE NEED TO MOVE FROM MCROLTHS TO MCROSYSTEMS

### SEPARATE THE STATELESS BEHAVIOR FROM THE STATEFUL ENTITY TO SCALE THEM INDIVIDUALLY



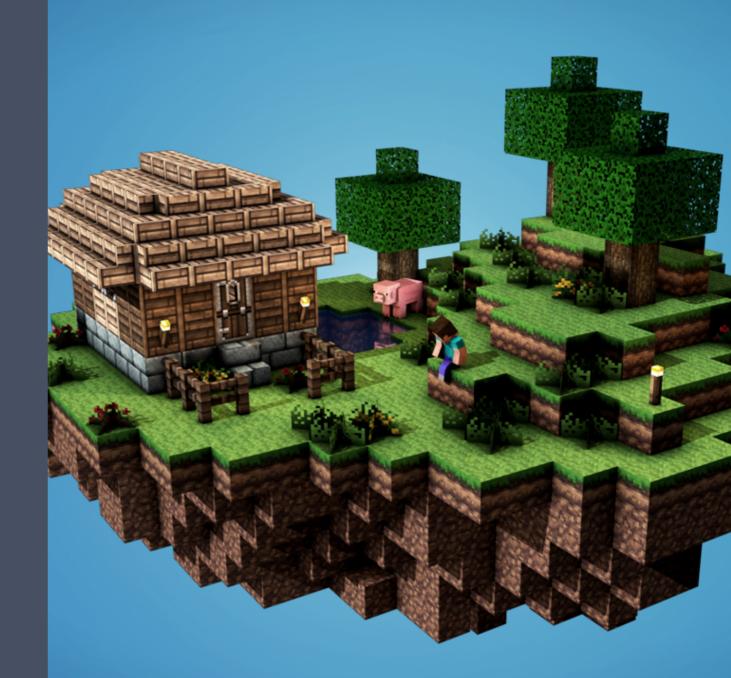


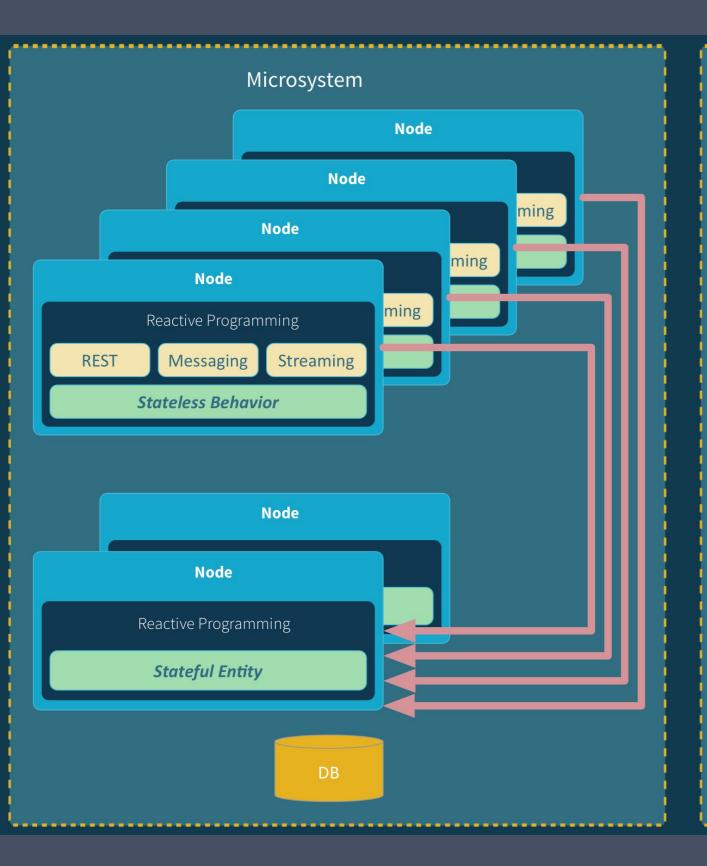
# SCALING (STATELESS) BEHAVIOR

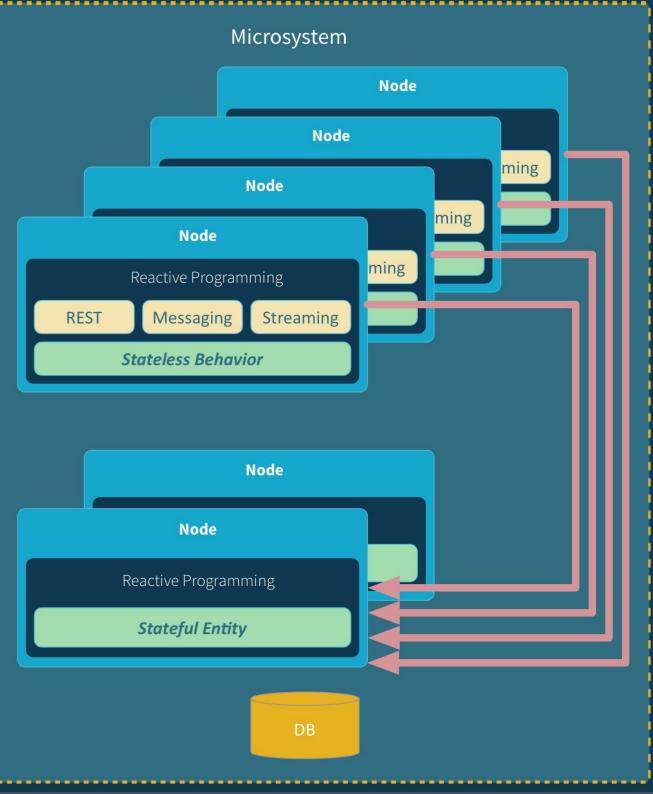
## SCALING (STATEFUL) ENTITIES

#### THERE IS NO SUCH THING AS A "Stateless" Architecture It's just someone else's problem

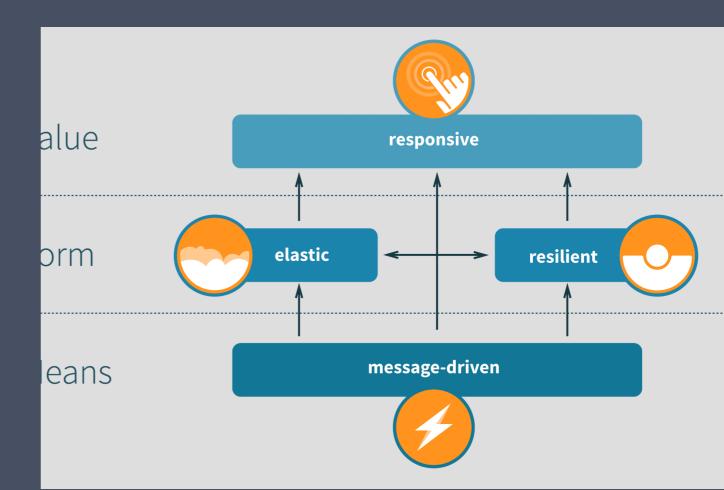
THE ENTITY CAN BECOME AN **ESCAPE ROUTE** FROM REALITY. **A SAFE ISLAND OF DETERMINISM AND STRONG CONSISTENCY** 





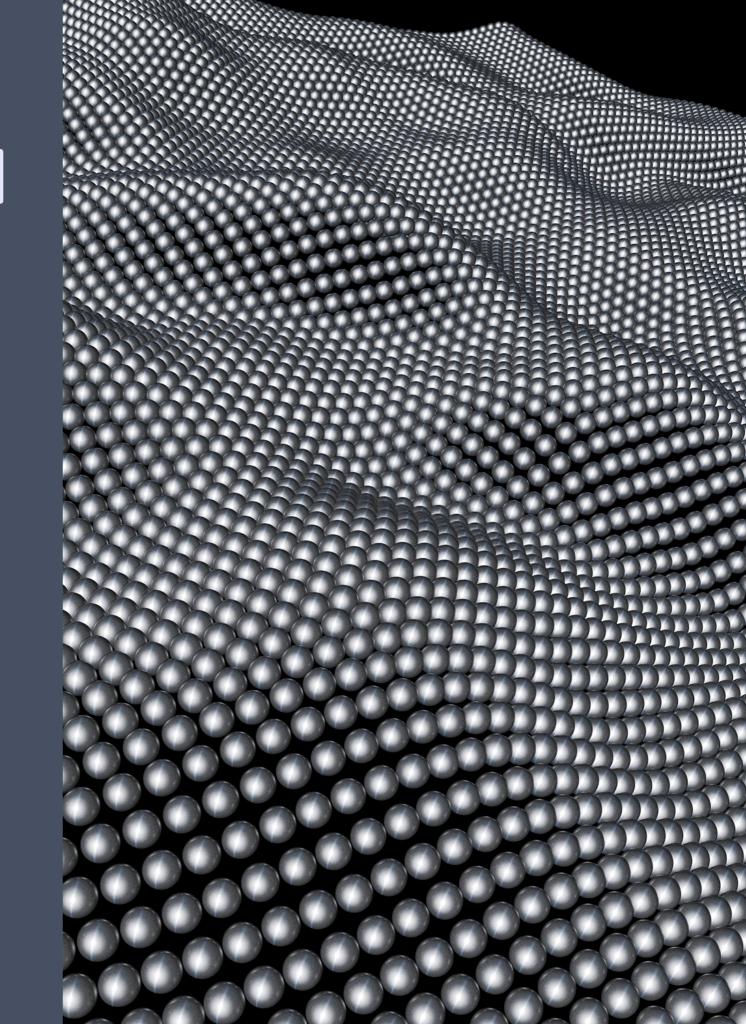


**REACTIVE SYSTEMS HELPS MAKING THE** MICROSERVICE (MICROSYSTEM) **RESILIENT AND ELASTICALLY SCALABLE** LEARN MORE AT REACTIVEMANIFESTO.ORG



## **REACTIVE SYSTEMS ARE BASED ON** ASYNCHRONOUS MESSAGE-PASSING

## **ALLOWS DECOUPLING IN** AND



**ALLOWS FOR LOCATION TRANSPARENCY ONE COMMUNICATION ABSTRACTION ACROSS ALL DIMENSIONS OF SCALE**  $\mathsf{CORE} \Longrightarrow \mathsf{SOCKET} \Longrightarrow \mathsf{CPU} \Longrightarrow$  $CONTAINER \implies SERVER \implies$  $\mathsf{RACK} \Longrightarrow \mathsf{DATA} \ \mathsf{CENTER} \Longrightarrow$ SYSTEM

BUT I'LL TAKE MY TIME ANYWHERE. I'M FREE TO SPEAK **MY MIND ANYWHERE. AND I'LL REDEFINE** ANYWHERE. ANYWHERE I ROAM. WHERE I LAY MY HEAD IS HOME.

- WHEREVER I MAY ROAM BY LARS Ulrich. James Hetfield (Metallica)

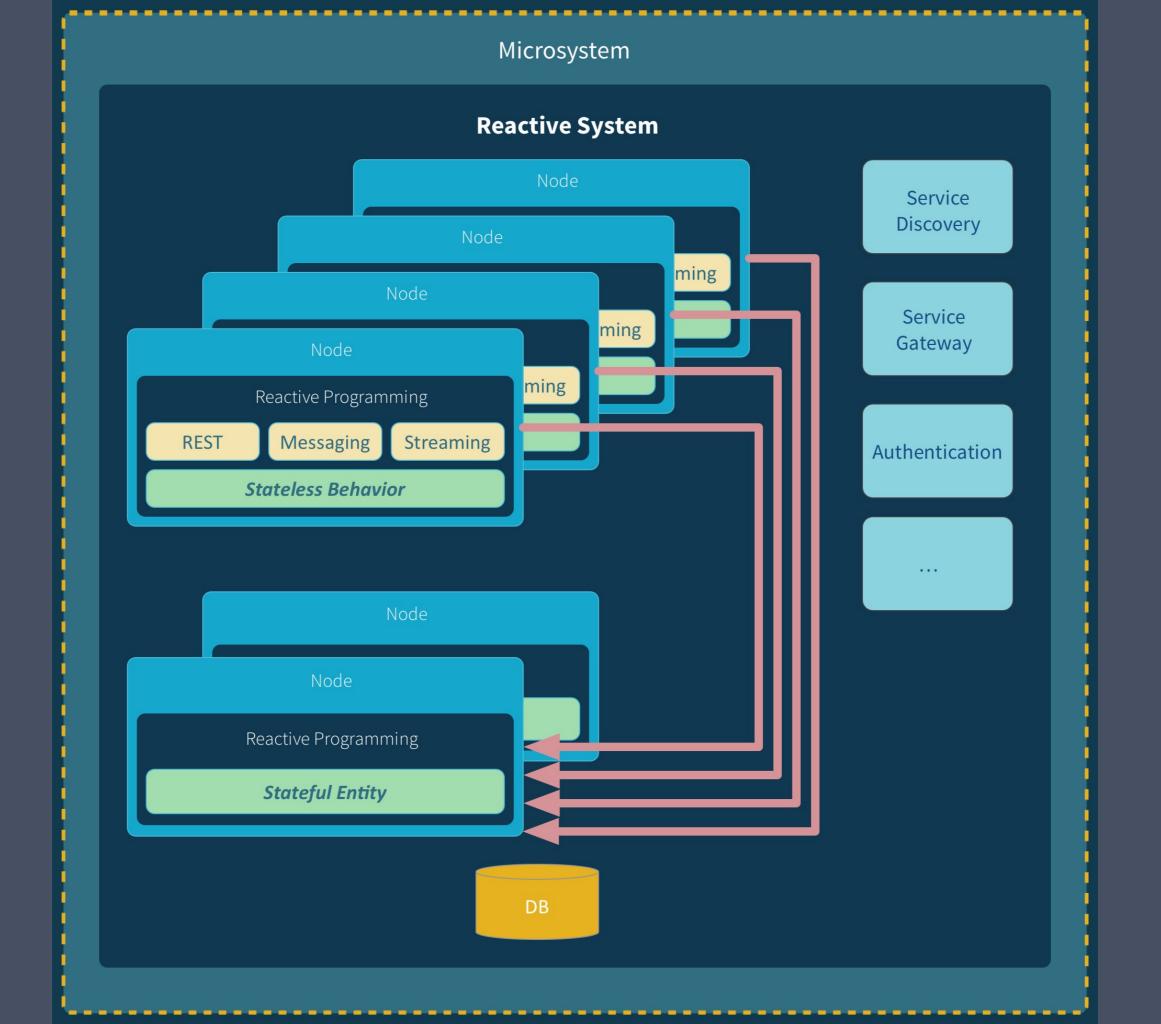


#### THE PATH TOWARDS RESILIENCE

**1. DECENTRALIZED ARCHITECTURE** 2. BULKHEADING **3. REPLICATION 4. FAILURE DETECTION 5. SUPERVISION** ⇒ SELF-HEALING SYSTEMS

#### THE PATH TOWARDS ELASTICITY

DECENTRALIZED ARCHITECTURE
EPIDEMIC GOSSIP PROTOCOLS
SELF-ORGANIZATION
LOCATION TRANSPARENCY
⇒ ELASTIC SYSTEMS



#### THINK IN TERMS OF CONSISTENCY BOUNDARIES

INSIDE DATA: OUR CURRENT PRESENT (STATE) OUTSIDE DATA: BLAST FROM THE PAST (FACTS) BETWEEN SERVICES: HOPE FOR THE FUTURE (COMMANDS) - PAT HELLAND, DATA ON THE INSIDE VS DATA ON THE OUTSIDE

## PRACTICE EVENTS-FIRST DOMAIN-DRIVEN DESIGN

#### DON'T FOCUS ON THE THINGS-THE NOUNS FOCUS ON WHAT HAPPENS-THE EVENTS

# LET THE **EVENTS DEFINE THE BOUNDED CONTEXT**

### **EVENTS REPRESENT FACTS**

#### TO CONDENSE FACT FROM THE VAPOR OF NUANCE

- NEAL STEPHENSON. SNOW CRASH





#### ...AND WE ARE NOT TALKING ABOUT ALTERNATIVE FACTS



#### UNDERSTAND HOW FACTS ARE CAUSALLY RELATED HOW FACTS FLOW IN THE SYSTEM

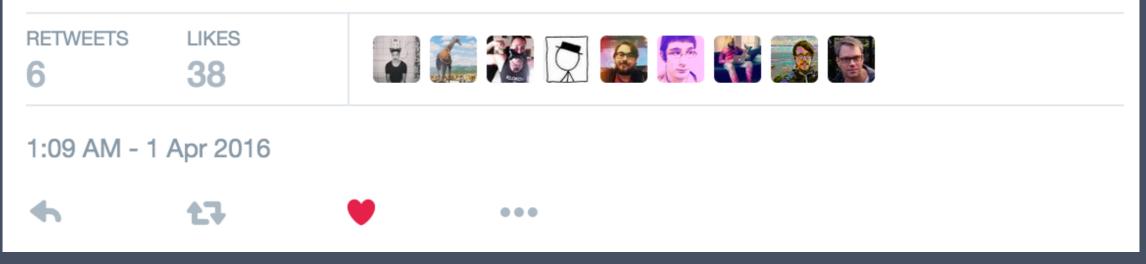


Peter Alvaro @palvaro

🔅 Follo

Following

#### I shall never tire of writing "causality is reachability in spacetime" on the blackboard





#### THE TRUTH IS THE LOG. THE DATABASE IS A CACHE OF A SUBSET OF THE LOG.

- PAT HELLAND

#### Immutability Changes Everything

#### Pat Helland Salesforce.com One Market Street, #300 San Francisco, CA 94105 USA 01(415) 546-5881 phelland@salesforce.com

#### ABSTRACT

There is an inexorable trend towards storing and sending immutable data. We <u>need immutability</u> to coordinate at a distance and we <u>can afford immutability</u>, as storage gets cheaper.

This paper is simply an amuse-bouche on the repeated patterns of computing that leverage immutability. Climbing up and down the compute stack really does yield a sense of déjà vu all over again.

#### 1. INTRODUCTION

It wasn't that long ago that computation was expensive, disk storage was expensive, DRAM was expensive, but coordination with latches was cheap. Now, all these have changed using cheap computation (with many-core), cheap commodity disks, and cheap DRAM and SSD, while coordination with latches gets harder because latch latency loses lots of instruction opportunities. We can now afford to keep immutable copies of lots of data, and one payoff is reduced coordination challenges.

1.1 More Storage, Distribution, & Ambiguity

We have *increasing storage* as the cost per terabyte of disk keeps dropping. This means we can keep lots of data for a long time. We have *increasing distribution* as more and more data and work are spread across a great distance. Data within a datacenter seems "far away". Data within a many-core chip may seem "far away". We have *increasing ambiguity* when trying to coordinate with systems that are far away… more stuff has happened since you've heard the news. Can you take action with incomplete knowledge?

#### 1.2 Turtles All the Way Down [17]

As various technological areas have evolved recently, they have responded to these trends of increasing storage, distribution, and ambiguity by using immutable data in some very fun ways. We will explore how apps use immutability in their ongoing work, how apps generate immutable DataSets for later offline analysis, how SQL can expose and process immutable snapshots, how massively parallel "Big Data" work relies on immutable DataSets. This leads us to looking at the ways in which semantically immutable DataSets may be altered while remaining immutable.

Next, we consider how updatability is layered atop the creation of new immutable files via techniques like LSF (Log Structure File systems), COW (Copy on Write), and LSM (Log Structured Merge trees). We examine how replicated and distributed file systems depend on immutability to eliminate anomalies.

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7th Biennial Conference on Innovative Data Systems Research (CIDR '15) January 4-7, 2015, Asilomar, California, USA.

Next, we discuss how the hardware folks have joined the party by leveraging these tricks in SSD and HDD. See Figure 1. Finally, we look at some trade-offs with using immutable data.

T many, we look at some trade-ons with using minutable data.	
Append-Only Apps	App over Immutable Data: Record Facts then Derive
App Generated DataSets	Generate Immutable Data
Massively Parallel "Big Data"	Read & Write Immutable DataSets
SQL Snapshots & DataSets	Generate Immutable Data
Subjectively Immutable DataSets	Interpret Data as Immutable
LSF, LSM, and COW	Expose Change over Immutable Files by Append
Immutable Files	Replication of Files/Blocks without Update Anomalies
Wear Leveling on SSD	Change via COW to Spread Physical Update Blocks
Shingles on HDD	Change via COW to Allow Large Physical Rewrites
Figure 1. Immutability is a key architectural concept at many layers of the stack.	

#### 2. Accountants Don't Use Erasers

Lots of computing can be characterized as "append-only". This section looks at some of the ways this is commonly accomplished.

#### 2.1 "Append-Only" Computing

May kinds of computing are "Append-Only". Observations are recorded forever (or for a long time). Derived results are calculated on demand (or periodically pre-calculated).

This is similar to a database management system. Transaction logs record all the changes made to the database. High-speed appends are the only way to change the log. From this perspective, the contents of the database hold a caching of the latest record values in the logs. The truth is the log. The database is a cache of a subset of the log. That cached subset happens to be the latest value of each record and index value from the log.

#### 2.2 Accounting: Observed & Derived Facts

<u>Accountants don't use erasers</u> or they go to jail. All entries in a ledger remain in the ledger. Corrections can be made but only by making new entries in the ledger. When a company's quarterly results are published, they include small corrections to the previous quarter. Small fixes are OK! They are append-only, too!

Some entries describe <u>observed facts</u>. For example, receiving a debit or credit against a checking account is an observed fact.

Some entries describe <u>derived facts</u>. Based on the observations, we can calculate something new. For example, amortized capital expenses based upon a rate and a cost. Another example is the current bank account balance with applied debits and credits.

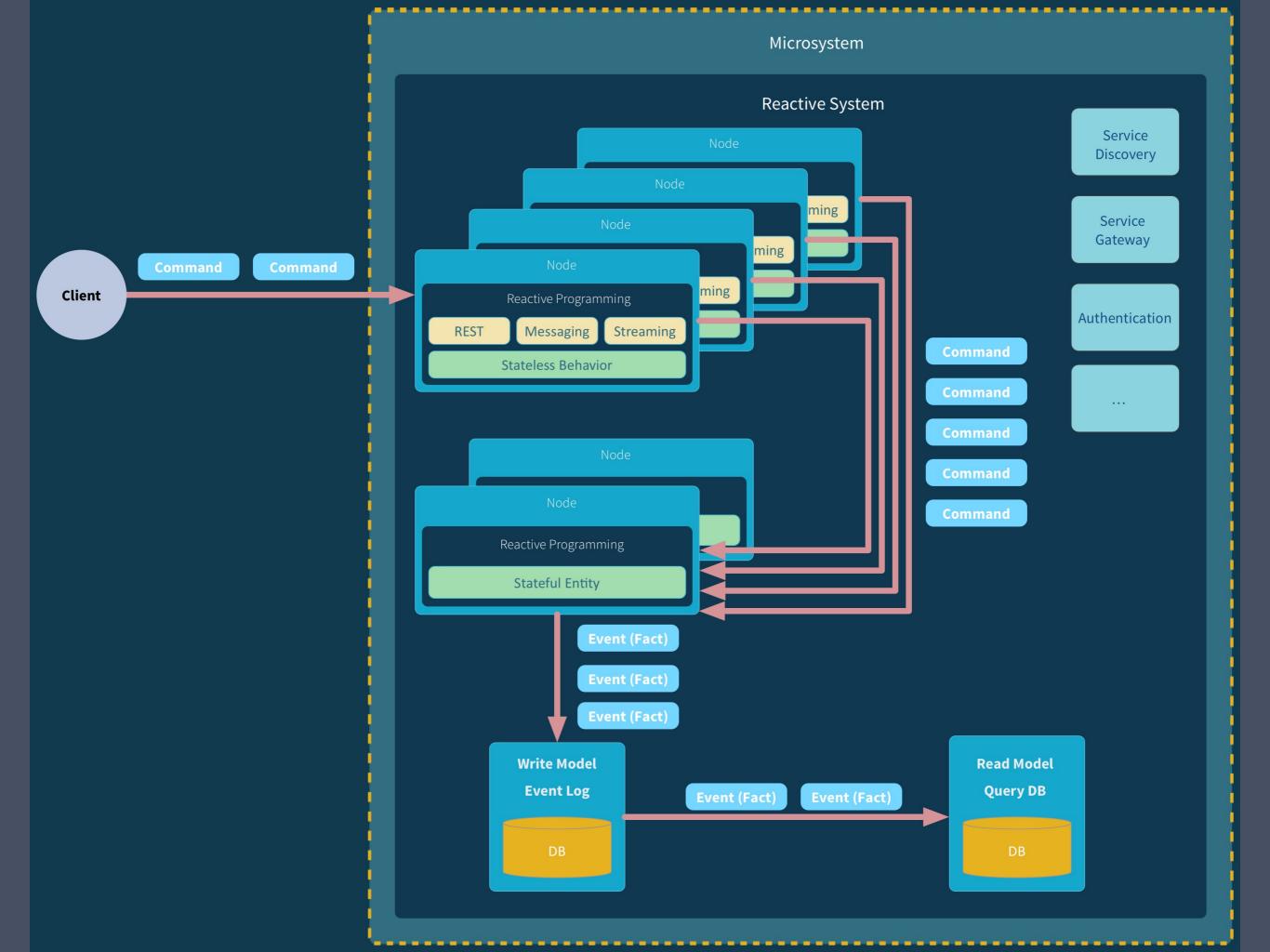




# IS A DATABASE OF THE PAST NOT JUST A DATABASE OF THE PRESENT

## EVENT LOGGING AVOIDS THE INFAMOUS OBJECT-RELATIONAL IMPEDENCE MISMATCH

## **UNTANGLE THE** READ & WRITE MODELS WITH CORS & EVENT SOURCING



# BUT WHAT ABOUT TRANSACTIONS?

**IN GENERAL. APPLICATION** DEVELOPERS SIMPLY DO NOT IMPLEMENT LARGE SCALABLE **APPLICATIONS** ASSUMING DISTRIBUTED TRANSACTIONS. - PAT HELLAND

### Life beyond Distributed Transactions: an Apostate's Opinion Position Paper

Pat Helland

Amazon.Com 705 Fifth Ave South Seattle, WA 98104 USA

PHelland@Amazon.com

The positions expressed in this paper are personal opinions and do not in any way reflect the positions of my employer Amazon.com.

### ABSTRACT

Many decades of work have been invested in the area of distributed transactions including protocols such as 2PC, Paxos, and various approaches to quorum. These protocols provide the application programmer a façade of global serializability. Personally, I have invested a nontrivial portion of my career as a strong advocate for the implementation and use of platforms providing guarantees of global serializability.

My experience over the last decade has led me to liken these platforms to the Maginot Line<sup>1</sup>. In general, application developers simply do not implement large scalable applications assuming distributed transactions. When they attempt to use distributed transactions, the projects founder because the performance costs and fragility make them impractical. Natural selection kicks in...

<sup>1</sup> The Maginot Line was a huge fortress that ran the length of the Franco-German border and was constructed at great expense between World War I and World War II. It successfully kept the German army from <u>directly</u> crossing the border between France and Germany. It was quickly bypassed by the Germans in 1940 who invaded through Belgium.

This article is published under a Creative Commons License Agreement (http://creativecommons.org/licenses/by/2.5/). You may copy, distribute, display, and perform the work, make derivative works and make commercial use of the work, but you must attribute the work to the author and CIDR 2007. 3<sup>rd</sup> Biennial Conference on Innovative DataSystems Research (CIDR) January 7-10, Asilomar, California USA. Instead, applications are built using different techniques which do not provide the same transactional guarantees but still meet the needs of their businesses.

This paper explores and names some of the practical approaches used in the implementations of large-scale mission-critical applications in a world which rejects distributed transactions. We discuss the management of fine-grained pieces of application data which may be repartitioned over time as the application grows. We also discuss the design patterns used in sending messages between these repartitionable pieces of data.

The reason for starting this discussion is to raise awareness of new patterns for two reasons. First, it is my belief that this awareness can ease the challenges of people hand-crafting very large scalable applications. Second, by observing the patterns, hopefully the industry can work towards the creation of platforms that make it easier to build these very large applications.

### 1. INTRODUCTION

Let's examine some goals for this paper, some assumptions that I am making for this discussion, and then some opinions derived from the assumptions. While I am keenly interested in high availability, this paper will ignore that issue and focus on scalability alone. In particular, we focus on the implications that fall out of assuming we cannot have large-scale distributed transactions.

### Goals

This paper has three broad goals:

 <u>Discuss Scalable Applications</u> Many of the requirements for the design of scalable systems are understood implicitly by many application designers who build large systems.



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### IN SUMMARY

**1. DON'T BUILD MICROLITHS** 2. MICROSERVICES COME IN (DISTRIBUTED) SYSTEMS **3. MICROSERVICES COME AS (MICRO)SYSTEMS 4. EMBRACE THE REACTIVE PRINCIPLES 5. EMBRACE EVENT-FIRST DDD & PERSISTENCE 6. PROFIT** 

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