

SYMBOLIC ANALYSIS OF WEAK CONCURRENCY SEMANTICS IN MODERN DATABASE PROGRAMS

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INTRODUCTION (MAIN SUBJECT: DATABASE SYSTEMS)



- Data is ubiquitously managed by databases



- Data is ubiquitously managed by databases
- Well-defined abstractions
 - data definition language (DDL)
 - data manipulation language (DML)





- Data is ubiquitously managed by databases
- Well-defined abstractions
 - data definition language (DDL)
 - data manipulation language (DML)
- Single machine





- Data is ubiquitously managed by databases
- Well-defined abstractions
 - data definition language (DDL)
 - data manipulation language (DML)
- Single machine
- Strong concurrency guarantees





- Single machine = poor performance



- Single machine = poor performance
- Amazon DynamoDB
 - Cluster of nodes on the cloud





DC1



- Single machine = poor performance
- Amazon DynamoDB
 - Cluster of nodes on the cloud
- Designed to scale



DC1



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- Clustering architectures
 - Partitioning
 - Replicating



DC1



- Clustering architectures
 - Partitioning
 - Replicating









- Clustering architectures
 - Partitioning
 - Replicating





- Clustering architectures
 - Partitioning
 - Replicating









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 - Partitioning
 - Replicating







WEAK CONCURRENCY SEMANTICS

- Clustering architectures
 - Partitioning
 - Replicating
- Weak concurrency guarantees





WEAK CONCURRENCY SEMANTICS

- Clustering architectures
 - Partitioning
 - Replicating
- Weak concurrency guarantees
 - Programmer is exposed to concurrency



Correctness?

Performance?





WEAK CONCURRENCY SEMANTICS

- Clustering architectures
 - Partitioning
 - Replicating
- Weak concurrency guarantees
 - Programmer is exposed to concurrency
- Concurrency bugs
 - prevalent
 - dangerous
 - alerted database community

ACIDRain: Concurrency-Related Attacks on Database-Backed Web Applications

Todd Warszawski, Peter Bailis Stanford InfoLab

ABSTRACT

In theory, database transactions protect application data from corruption and integrity violations. In practice, database transactions frequently execute under weak isolation that exposes programs to a range of concurrency anomalies, and programmers may fail to correctly employ transactions. While low transaction volumes mask many potential concurrency-related errors under normal operation determined adversaries can exploit them programmatically for fun and profit. In this paper, we formalize a new kind of attack on database-backed applications called an ACIDRain attack, in which an adversary systematically exploits concurrency-related vulnerabil ities via programmatically accessible APIs. These attacks are not theoretical: ACIDRain attacks have already occurred in a handful of applications in the wild, including one attack which bankrupted a popular Bitcoin exchange. To proactively detect the potential for ACIDRain attacks, we extend the theory of weak isolation to analyze latent potential for non-serializable behavior under concurrent web API calls. We introduce a language-agnostic method for detecting potential isolation anomalies in web applications, called Abstract Anomaly Detection (2AD), that uses dynamic traces of database accesses to efficiently reason about the space of possible concurrent interleavings. We apply a prototype 2AD analysis tool to 12 popular self-hosted eCommerce applications written in four languages and deployed on over 2M websites. We identify and verify 22 critical ACIDRain attacks that allow attackers to corrupt store inventory over-spend gift cards, and steal inventory.

1. INTRODUCTION

For decades, database systems have been tasked with maintaining application integrity despite concurrent access to shared state [39]. The serializable transaction concept dictates that, if programmers correctly group their application operations into transactions, application integrity will be preserved [34]. This concept has formed the cornerstone of decades of database research and design and has led to at least one Turing award [2,40].

In practice, the picture is less clear-cut. Some databases, including Oracle's flagship offering and SAP HANA, do not offer serializability as an option at all. Other databases allow applications

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1	def withdraw(amt, user_id):	(a)
2	bal = readBalance(user_id)	
3	if (bal >= amt):	
4	writeBalance(bal – amt, user_id)	

def withdraw(amt, user_id): beginTxn() $bal = readBalance(user_id)$ if (bal >= amt) writeBalance(bal – amt, user_id) commit()

Figure 1: (a) A simplified example of code that is vulnerable to an ACIDRain attack allowing overdraft under concurrent access. Two concurrent instances of the withdraw function could both read balance \$100, check that $100 \ge 999$, and each allow \$99 to be withdrawn, resulting \$198 total withdrawals. (b) Example of how transactions could be inserted to address this error. However, even this code is vulnerable to attack at isolation levels at or below Read Committed, unless explicit locking such as SELECT FOR UPDATE is used. While this scenario closely resembles textbook examples of improper transaction use, in this paper, we show that widely-deployed eCommerce applications are similarly vulnerable to such ACIDRain attacks, allowing corruption of application state and theft of assets.

to configure the database isolation level but often default to nor serializable levels [17, 19] that may corrupt application state [45]. Moreover, we are unaware of any systematic study that examines whether programmers correctly utilize transactions.

For many applications, this state of affairs is apparently satisfac tory. That is, some applications do not require serializable transac tions and are resilient to concurrency-related anomalies [18, 26, 48] More prevalently, many applications do not experience concurrency related data corruption because their typical workloads are not highly concurrent [21]. For example, for many businesses, even a few transactions per second may represent enormous sales volume

However, the rise of the web-facing interface (i.e., API) leads to the possibility of increased concurrency-and the deliberate exploitation of concurrency-related errors. Specifically, given a public API, a third party can programmatically trigger database-backed behavior at a much higher rate than normal. This highly concurrent workload can trigger latent programming errors resulting from incorrect transaction usage and/or incorrect use of weak isolation levels. Subsequently, a determined adversary can systematically exploit these errors, both to induce data corruption and induce un

What Are We Doing With Our Lives? Nobody Cares About Our Concurrency Control Research

Andrew Pavlo Carnegie Mellon University pavlo@cs.cmu.edu

ABSTRACT

Most of the academic papers on concurrency control published in the last five years have assumed the following two design decisions: (1) applications execute transactions with serializable isolation and (2) applications execute most (if not all) of their transactions usin stored procedures. I know this because I am guilty of writing these papers too. But results from a recent survey of database administra tors indicates that these assumptions are not realistic. This survey includes both legacy deployments where the cost of changing the application to use either serializable isolation or stored procedures is not feasible, as well as new "greenfield" projects that not encumbered by prior constraints. As such, the research produced by our community is not helping people with their real-world systems and thus is essentially irrelevant

In this talk/denouncement, I will descend from my ivory tower and argue that we need to rethink our agenda for concurrency control research. Recent trends focus on asking the wrong questions and solving the wrong problems. I contend that the real issues that will have the most impact are not easily solved by more "clever algorithms. Instead, in many cases, they can only be solved by hardware improvements and artificial intelligence.

1. ACKNOWLEDGEMENTS

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2. **BIOGRAPHIES**

Andrew Pavlo is an Assistant Professor of Databaseology in the Computer Science Department at Carnegie Mellon University. At CMU, he is a member of the Database Group and the Parallel Data Laboratory. His work is also in collaboration with the Intel Science and Technology Center for Big Data.

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- Ease of strong semantics
- Performance of cloud-native





- Ease of strong semantics
- Performance of cloud-native
- Language-oriented solutions
 - SMT Solvers
 - Model Checking







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(n)



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CLOTHO	Guided Testing Framework	OOPSLA 2019	
ATROPOS	Automated repair of replication anomalies	PLDI 2021	
LACHESIS	Automated repair of partitioning anomalies	under submission	







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- Performance of cloud-native
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CLOTHO (SYMBOLIC PROGRAM ANALYSIS & GUIDED TESTING)







- Atomicity
- Consistency
- Isolation
- Durability







- Atomicity
- Consistency
- Isolation
- Durability





- Atomicity
- Consistency
- Isolation
- Durability





- Atomicity
- Consistency
- Isolation
- Durability





- Atomicity
- Consistency
- Isolation
- Durability





- Atomicity
- Consistency
- Isolation
- Durability
- Serializability





- ACID guarantees
 - Atomicity
 - Consistency
 - Isolation
 - Durability
- Serializability
- Facilitates program design and reasoning





SERIALIZABILITY

- Serializability is costly



SERIALIZABILITY

- Serializability is costly
- Developers forced to use weaker semantics
 - Transactions are interleaved
 - Asynchronous replication of data











PURDUE UNIVERSITY

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- Serializability is costly
- Developers forced to use weaker semantics
 - Transactions are interleaved
 - Asynchronous replication of data
- Less intuitive behaviors











PURDUE UNIVERSITY
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- Less intuitive behaviors

Serializability Anomalies











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- Online gaming platform



38

EXAMPLE (TABLES)

- Online gaming platform
 - 2 tables

PLAYE	R			
p_id	p_t_id	p_rc	ole	p_stat
TEAM				
t_id	t_name	p_cnt		





EXAMPLE (TABLES)

- Online gaming platform
 - 2 tables

PLAYERp_idp_t_idp_rolep_statTEAMt_idt_namep_cnt







- Online gaming platform
 - 2 tables
 - 3 transactions

PLAYE	R				
p_id	p_t_id		p_role	9	p_stat
TEAM				_	
t_id	t_name	p_	_cnt		





- Online gaming platform
 - 2 tables
 - 3 transactions

addPlayer(p, t)

old_cnt := select p_cnt from TEAM w
Insert (p,t,Ø,Ø) into PLAYER
update TEAM set p cnt=old cnt+1 whe



where
$$t_id=t$$

ere $t_id=t$





- Online gaming platform
 - 2 tables
 - 3 transactions

addPlayer(p, t)

old_cnt := select p_cnt from TEAM w Insert (p,t,Ø,Ø) into PLAYER

update TEAM set p_cnt=old_cnt+1 whe

PlayerByTeam (t)

ps := select * from PLAYER where p_
foreach p in ps do
 print(p.*)







- Online gaming platform
 - 2 tables
 - 3 transactions

addPlayer(p, t)

old_cnt := select p_cnt from TEAM w Insert (p,t,Ø,Ø) into PLAYER

update TEAM set p_cnt=old_cnt+1 whe

PlayerByTeam (t)

ps := select * from PLAYER where p_
foreach p in ps do
 print(p.*)

setPlayer (p, r, s)

update PLAYER set p_role=r and p_stat=p where p_id=p







TEAM		
t_id	t_name	p_cnt
T1	"A"	2

addPlayer(p=P0, t=T1)

Select p_cnt // 2

Update p_cnt = 3

. . .

addPlayer(p=P0, t=T1)

Select p_cnt // 2

Update p_cnt = 3

. . .



TEAM		
t_id	t_name	p_cnt
T1	"A"	2





TEAM		
t_id	t_name	p_cnt
T1	"A"	2





_name	p_cnt
"A"	3



TEAM		
t_id	t_name	p_cnt
T1	"A"	2











TEAM

t_id	t_name	p_cnt	
T1	"A"	3 🗲	

"The column 'p_cnt' reflects the number of players in each team"

- Lost Update







TEAM

t_id	t_name	p_cnt	
T1	"A"	3 🗲	



Lost Update

"The column 'p_cnt' reflects the number of players in each team"





- Serializability anomalies are subtle



- Serializability anomalies are subtle
 - initial state





- Serializability anomalies are subtle
 - initial state
 - transaction arguments





- Serializability anomalies are subtle
 - initial state
 - transaction arguments
 - interleaved order





- Serializability anomalies are subtle
 - initial state
 - transaction arguments
 - interleaved order





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BLACKBOX TESTING





BLACKBOX TESTING

 Run and monitor apps in (semi-) production environment





BLACKBOX TESTING

- Run and monitor apps in (semi-) production environment
 - + real bugs (no false positive)
 - costly
 - too specific
 - no guarantee of coverage
 - manual effort







- Cloud-based testing framework
 - OLTPBench
 - Jepsen







- Cloud-based testing framework
 - OLTPBench
 - Jepsen
- TPC-C Benchmark
 - 5 txn and 9 tables









- Cloud-based testing framework
 - OLTPBench
 - Jepsen
- TPC-C Benchmark
 - 5 txn and 9 tables
 - 21 application-level invariants

Invariant	
CR1	
CR2	
CR3	
CR4	
CR5A	
CR5B	
CR6	
CR7A	
CR7B	
CR8	
CR9	
CR10	
CR11	
CR12	
NCR1	
NCR2	
NCR3	
NCR4	
NCR5	
NCR6	
NCR7	



- Cloud-based testing framework
 - OLTPBench
 - Jepsen
- TPC-C Benchmark
 - 5 txn and 9 tables
 - 21 application-level invariants
 - only 14 invariants broken

Invariant	Broken?
CR1	Y
CR2	Y
CR3	Y
CR4	Y
CR5A	Ν
CR5B	Ν
CR6	Y
CR7A	Ν
CR7B	Ν
CR8	Y
CR9	Y
CR10	Y
CR11	Y
CR12	Y
NCR1	Y
NCR2	Y
NCR3	Ν
NCR4	Ν
NCR5	Y
NCR6	Y
NCR7	Ν



ANOTHER SOLUTION?

2. Construct and run anomalous executions to the devs

- KEY IDEA -1. Statically analyze programs and find abstract anomalies (white-box testing)



- Non-serializable executions in a program, iff
 - Cyclic dependencies between txns

ogram, iff ns



- Non-serializable executions in a program, iff
 - Cyclic dependencies between txns
- Executions abstracted by directed dependency graphs



- Non-serializable executions in a program, iff
 - Cyclic dependencies between txns
- Executions abstracted by directed dependency graphs
- Three types of dependency edges



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- Three types of dependency edges
 - write dependency (WW)







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- Executions abstracted by directed dependency graphs
- Three types of dependency edges
 - write dependency (WW)
 - read dependency (WR)





- Non-serializable executions in a program, iff
 - Cyclic dependencies between txns
- Executions abstracted by directed dependency graphs
- Three types of dependency edges
 - write dependency (WW)
 - read dependency (WR)
 - read anti-dependency (RW)







- Non-serializable executions in a program, iff
 - Cyclic dependencies between txns
- Executions abstracted by directed dependency graphs
- Three types of dependency edges
 - write dependency (WW)
 - read dependency (WR)
 - read anti-dependency (RW)
- Example: lost update











- Key idea: reduction to SMT




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- Use efficient SMT-solvers, e.g. Z3



73

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- Finding bounded anomalies against a database abstraction reduced to finding satisfying assignments to a formula arphi



- Key idea: reduction to SMT
- Use efficient SMT-solvers, e.g. Z3
- Axiomatic relations encoded within a decidable fragment of FOL
- Finding bounded anomalies against a database abstraction reduced to finding satisfying assignments to a formula arphi
- Components of the encoding

$\varphi \equiv \varphi_{\text{context}} \land \varphi_{\text{db}} \land \varphi_{\text{dep}} \land \varphi_{\rightarrow \text{dep}} \land \varphi_{\text{anomaly}}$



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- Components of the encoding

$$\varphi \equiv \varphi_{\rm CONTEXT} \land \varphi_{\rm DB} \land$$

conditions satisfied by any execution of any program

$\varphi_{\text{DEP}} \wedge \varphi_{\rightarrow \text{DEP}} \wedge \varphi_{\text{ANOMALY}}$



- Key idea: reduction to SMT
- Use efficient SMT-solvers, e.g. Z3
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conditions satisfied by any execution of any program

database-specific consistency constraints

 $\varphi \equiv \varphi_{\text{CONTEXT}} \land \varphi_{\text{DB}} \land \varphi_{\text{DEP}} \land \varphi_{\rightarrow \text{DEP}} \land \varphi_{\text{ANOMALY}}$



- Key idea: reduction to SMT
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conditions satisfied by any execution of any program

Necessary conditions to establish a dependency relation

database-specific consistency constraints Sufficient conditions to establish a dependency relation



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conditions satisfied by any execution of any program

Necessary conditions to establish a dependency relation

database-specific consistency constraints Sufficient conditions to establish a dependency relation

> Enforces the existence of a dependency cycle





PURDUE UNIVERSITY.

CLOTHO: DIRECTED TEST GENERATION AND REPLAY

- Static analysis engine for java programs
- Automatic replay of anomalous executions







EXPERIMENTAL RESULTS

- 7 benchmarks of various complexity analyzed
- Serializability anomalies found and replayed in 5 benchmark



ity analyzed replayed in 5 benchmark



EXPERIMENTAL RESULTS

- 7 benchmarks of various complexity analyzed
- Serializability anomalies found and replayed in 5 benchmark



17 anomalies per application (avg)



CASE STUDY: TPC-C

- 22 anomalies mapped to invariant violations
 - All invariants were broken
 - Only 3 anomalies did NOT violate any invariant

Invariant	Blackbox	CLOTHO
CR1	Y	Y
CR2	Y	Y
CR3	Y	Y
CR4	Y	Y
CR5A	N	Y
CR5B	Ν	Y
CR6	Y	Y
CR7A	Ν	Y
CR7B	N	Y
CR8	Y	Y
CR9	Y	Y
CR10	Y	Y
CR11	Y	Y
CR12	Y	Y
NCR1	Y	Y
NCR2	Y	Y
NCR3	Ν	Y
NCR4	N	Y
NCR5	Y	Y
NCR6	Y	Y
NCR7	Ν	Y



ATROPOS

(REPAIRING REPLICATION ANOMALIES)



REPLICATION ANOMALIES

- Serializability anomaly

addPlayer(*p*, *t*)

old_cnt := select p_cnt from TEAM Insert (p,t,Ø,Ø) into PLAYER update TEAM set p_cnt=old_cnt+1 wh

where
$$t_id=t$$

ere $t_id=t$



REPLICATION ANOMALIES

- Serializability anomaly

addPlayer(*p*, *t*)

old_cnt := select p_cnt from TEAM **Insert** $(p, t, \emptyset, \emptyset)$ **into** PLAYER update TEAM set p_cnt=old_cnt+1 where t_id=t



cross-datacenter edge







REPLICATION ANOMALIES

- Serializability anomaly
- Replication anomalies can be eliminated using consistency annotations

```
addPlayer(p, t)
Consistent {
  old cnt := select * from TEAM wh
  Insert (p,t,Ø,Ø) into PLAYER
  update TEAM set p_cnt=old cnt+1 where t id=t
```

TEAM



cross-datacenter edge







- Serializability anomaly
- Replication anomalies can be eliminated using consistency annotations
- Equivalent program without dependency cycles

addPlayer(p, t)

Co	nsisten	it {					
	old_cn	t :=	selec	t *	from	TEAM	whe
	Insert	(p,t	,Ø,Ø)	int	to PLZ	AYER	
}	update	TEAM	set	p_cr	nt=old	d_cnt-	⊦1 w

PLAYEF	8			TEAM
p_id	p_t_id	p_role	p_stat	t_id





- Equivalent program without dependency cycles
 - Schema Refactoring







- Equivalent program without dependency cycles
 - Schema Refactoring
 - Program Refactoring







- Equivalent program without dependency cycles
 - Schema Refactoring
 - Program Refactoring









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- Equivalent program without dependency cycles
 - Schema Refactoring
 - Program Refactoring
- Only keep track of changes to the balance
- No shared item \square No dependency

```
addPlayer(p, t)
```

old_cnt := select p_cnt from TEAM where t id=t **Insert** $(p, t, \emptyset, \emptyset)$ **into** PLAYER Insert (t,Ø,uuid(),+1) into TEAM P CNT LOG

t_id t_name	UUID p	_cnt_change
-------------	--------	-------------







- Equivalent program without dependency cycles
 - Schema Refactoring
 - Program Refactoring
- Only keep track of changes to the balance
- No shared item rightarrow No dependency







REPAIRED PROGRAM'S EXECUTION



t_id	t_name	UUID	p_cnt_change
1	"A"	U3	2





REPAIRED PROGRAM'S EXECUTION



TEAM_P_CNT_LOG

t_id	t_name	UUID	p_cnt_change
1	"A"	U3	2



t_id	t_name	UUID	p_cnt_change
1	"A"	U1	1
1	"A"	U3	2
1	"A"	U4	1



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REPAIRED PROGRAM'S EXECUTION



TEAM			
t_id	t_name	p_cnt	
T1	"A"	4	



TEAM_P_CNT_LOG

t_id	t_name	UUID	p_cnt_change
1	"A"	U3	2



t_id	t_name	UUID	p_cnt_change
1	"A"	U1	1
1	"A"	U3	2
1	"A"	U4	1







- Program P' is a sound refactoring of P iff:



- Program **P**' is a sound refactoring of **P** iff: for any execution E' of P' there is an execution E of P s.t.





SOUNDNESS

- Program **P**' is a sound refactoring of **P** iff:
 - for any execution E' of P' there is an execution E of P s.t.
 - E and E' preserve the containment relation (\subseteq) between initial and final DB states





SOUNDNESS

- Program **P**' is a sound refactoring of **P** iff:
 - for any execution E' of P' there is an execution E of P s.t.
 - E and E' preserve the containment relation (\sqsubseteq) between initial and final DB states
 - equivalent values returned





EVALUATION (APPLICABILITY + EFFECTIVENESS)

- 9 benchmarks
 - Number of statically identified anomalous access pairs

ONE benchmark (avg 12%)

THREE benchmarks (avg 12%)

Original + Eventual Consistency Original + Causal Consistency Original + Repeatable Read Refactored + Eventual Consistency



ALL benchmarks (avg 73%)





EVALUATION (APPLICABILITY + EFFECTIVENESS)

- 9 benchmarks
 - Number of statically identified anomalous access pairs

ONE benchmark (avg 12%)

THREE benchmarks (avg 12%)

Original + Eventual Consistency Original + Causal Consistency Original + Repeatable Read Refactored + Eventual Consistency



ALL benchmarks (avg 73%)





EVALUATION (PERFORMANCE)

- 3 benchmarks
- Latency and throughput
 - EC, SC: original program (eventual consistency/serializability)
 - AT_EC, AT_SC: refactored program (eventual consistency/serializability)



16 **US Cluster** Latency (x100 ms) 12 AT-EC AT-SC + EC SC 75 100 125 0 25 50 Number of Clients

SEATS

SmallBank



104



EVALUATION (PERFORMANCE)



SmallBank

- EC, SC: original program + eventual consistency and serializability

- AT_EC, AT_SC: original program + eventual consistency and serializability





LACHESIS

(REPAIRING PARTITIONING ANOMALIES)



106

PARTITIONING BASICS

- Partitioned Database Clusters
 - Same data-center deployment







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PARTITIONING BASICS

- Partitioned Database Clusters
 - Same data-center deployment
 - Subset of records at each node




PARTITIONING BASICS

- Partitioned Database Clusters
 - Same data-center deployment
 - Subset of records at each node
- Partitioning Policy \prod
 - Performance





PARTITIONING BASICS

- Partitioned Database Clusters
 - Same data-center deployment
 - Subset of records at each node
- Partitioning Policy ∏
 - Performance





PARTITIONING BASICS

- Partitioned Database Clusters
 - Same data-center deployment
 - Subset of records at each node
- Partitioning Policy ∏
 - Performance
- Schism: [Curino et.al]

Schism: a Workload-Driven Approach to **Database Replication and Partitioning**

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Sam Madden madden@csail.mit.edu



111

- Correctness?







- Correctness?
- single-partition access = isolated & atomic







113

- Correctness?
- Single-partition access = Isolated & Atomic
- Eliminate concurrency anomalies by program refactoring







- Correctness?
- Single-partition access = Isolated & Atomic
- Eliminate concurrency anomalies by program refactoring
- Partitioning-aware symbolic execution







- Symbolic analysis engine
 - program P0 (transactions & schema)
 - partitioning relation

Static Anomaly P0 Detection



- Symbolic analysis engine
 - program P0 (transactions & schema)
 - partitioning relation
- Annotation & Runtime
 - Isolated { }









- Symbolic analysis engine
 - program P0 (transactions & schema)
 - partitioning relation
- Annotation & Runtime
 - Isolated { }

PlayerByTeam (t)





- Symbolic analysis engine
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```
playerByTeam (t=T1)
     Select P0
     Select P1
         . . .
```

PlayerByTeam (t)







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PlayerByTeam (t)

ps := select * from PLAYER where p t id=tforeach p in ps do print(p.*)











LACHESIS PIPELINE: CONTINUED

- Symbolic analysis engine
 - program P0 (transactions & schema)
 - partitioning relation
- Annotation & Runtime
 - Isolated { }

PlayerByTeam (t)









- Symbolic analysis engine
 - program P0 (transactions & schema)
 - partitioning relation
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 - Isolated { }
- Partitioning Recommendation
 - Static & efficient
 - Comparable to Schism









- Recommended partitioning policy: $\forall r, r'. r.p_t_id = r'.p_t_id \Rightarrow (r, r') \in \Pi$









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- Recommended partitioning policy: $\forall r, r'. r.p_t_id = r'.p_t_id \Rightarrow (r, r') \in \Pi$
- Partitioning eliminates annotations
- Annotations are costly
- 40% fewer annotations:
 - average of 23% higher throughput & 14% lower latency





129

LIMITATION OF PARTITIONING

- Each table is
 - partitioned for one access pattern
- Conflicting access patterns?
- $\forall r, r'. r.p_t_id = r'.p_t_id \Rightarrow (r, r') \in \Pi$





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LIMITATION OF PARTITIONING

- Refactor the schema:
 - A new table for each anomalous access pattern
 - Rewrite program
 - Repartition





- Running example:
 - additional three tables

g_id p_id

GAME

g_id	g_stat	g_min_player
------	--------	--------------

PLAYER

p_id p_t_id	p_role	p_stat
-------------	--------	--------







- Running example:
 - additional three tables
 - additional one transaction

|--|

GAME

g_id	g_stat	g_min_player
------	--------	--------------

PLAYER



playerByGame (g)

Game := select * from GAME where g_id=g print(game.*)

ids := select p_id from GMP where g_id=g foreach id in ids do

p := select * from PLAYER where p_id=id print(p.*)







- Running example:
 - additional three tables
 - additional one transaction
- Anomaly is not fixed by partitioning

PLAYER

p_id	p_t_id	p_role	p_stat	
P0	1	soldier	on	
P1	1	captain	on	
P2	2	captain	off	

playerByGame (g)

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playerByGame (g)

Game := select * from GAME where g_id=g
print(game.*)

ids := select p_id from GMP where g_id=g
foreach id in ids do

p := select * from PLAYER where p_id=id
print(p.*)

PLAYER_BY_GAMEg_idp_idp_t_idp_rolep_statG1P01soldieronG1P22captainoff







- Running example:
 - additional three tables
 - additional one transaction
- Anomaly is not fixed by partitioning

PLAYER

p_id	p_t_id	p_role	p_stat
P0	1	soldier	on
P1	1	captain	on
P2	2	captain	off

GAME

g_id	g_stat	g_min_player
G1	active	2

playerByGame (g)

Game := select * from GAME where g_id=g
print(game.*)

ids := select p_id from GMP where g_id=g
foreach id in ids do

```
p := select * from PLAYER_BY_GAME
    where p_id=id
```

print(p.*)

PLAYER_BY_GAME

g_id	p_id	p_t_id	p_role	p_stat
G1	P0	1	soldier	on
G1	P2	2	captain	off





IMPLEMENTATION & EMPIRICAL RESULTS

- ZooKeeper runtime
- 11 Benchmarks







IMPLEMENTATION & EMPIRICAL RESULTS

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- 11 Benchmarks
- Number of annotations
 - 40% reduction
 - Additional 32% reduction





IMPLEMENTATION & EMPIRICAL RESULTS

- ZooKeeper runtime
- 11 Benchmarks
- Number of annotations
 - 40% reduction
 - Additional 32% reduction
- Lower latency + Higher throughput



TPC-C benchmark





CONCLUSION



- Modern database systems come with a plethora of complex and subtle concurrency semantics.



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- Modern database systems come with a plethora of complex and subtle concurrency semantics.
- The developers have a hard time reasoning about programs.
- A symbolic program analysis engine to identify and report undesirable behaviors from a given database program
- By analyzing the access patterns of the database program, the program can be rewritten to be optimized for deployment on a particular clustering architecture.
- We report results from empirical experiments to support our claims.





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