

# TWIAD: Why You Should Index SAND2016-0692C In Write-Optimized Databases

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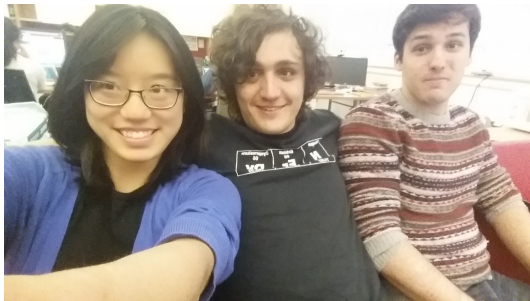
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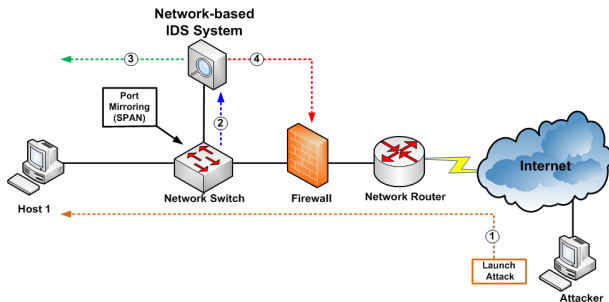
October 28, 2015

# TWIAD: Why You Should Index Your IDS Data In Write-Optimized Databases

**T**WIAD the  
**W**rite-Optimized  
**I**P  
**A**ddress  
**D**atabase

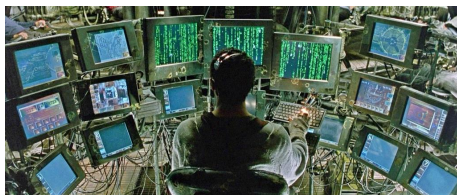


# TWIAD: Why You Should Index Your **IDS** Data In Write-Optimized Databases



IDS as we know it

# TWIAD: Why You Should Index Your **IDS Data** In Write-Optimized Databases



Your friendly neighborhood IDS  
monitor

- State of the art IDS involves logging all connections across the network
- Logging is bad but indexing is too slow
- Why should an IDS do the work of maintaining a database?

# TWIAD: Why You Should **Index Your IDS Data In** Write-Optimized **Databases**

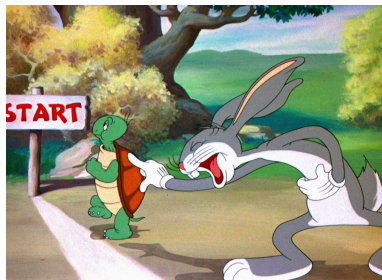
- Indexes facilitate faster queries
- “The right read problem is a write problem”
- Slow queries indicate bad/insufficient indexes



Indexing vs Logging

# TWIAD: Why You Should Index Your **IDS Data In** Write-Optimized **Databases**

- The pipe can be very big
- Connection logs generate quickly
- Lower budget = better
- How can we keep up and store all of this data in multiple indexes?



At first, logging is faster

# TWIAD: Why You Should Index Your IDS Data In Write-Optimized **Databases**

Goal: Ingest and query network data quickly

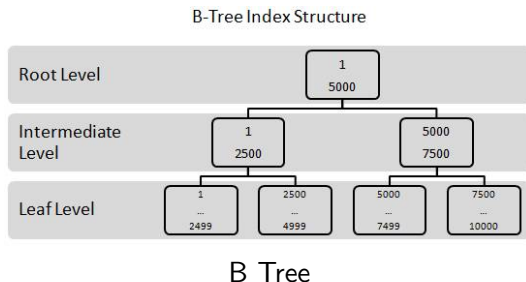
- More indexes expedite query but slow insertion
- Age-old tradeoff
- What if we could ingest more quickly to offset this tradeoff?



Logging falls flat when you get to queries

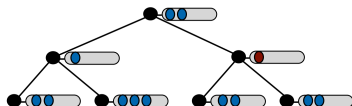
# TWIAD: Why You Should Index Your IDS Data In Write-Optimized Databases

- Start with a B-Tree
- Data stored at the leaves
- Insertion requires tree traversal
- Same cost as query

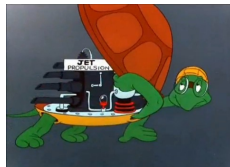




# TWIAD: Why You Should Index Your IDS Data In Write-Optimized Databases



$B^{\epsilon}$  Tree



- Add buffers at every node
- Insert into root buffer
- When buffers fill, flush down one level
- Amortized insertion cost

# TWIAD: Why You Should Index Your IDS Data In Write-Optimized Databases

	insert	point query
<b>Optimal tradeoff</b> (function of $\epsilon=0\dots 1$ )	$O\left(\frac{\log_{1+B^\epsilon} N}{B^{1-\epsilon}}\right)$	$O(\log_{1+B^\epsilon} N)$
<b>B-tree</b> ( $\epsilon=1$ )	$O(\log_B N)$	$O(\log_B N)$
$\epsilon=1/2$	$O\left(\frac{\log_B N}{\sqrt{B}}\right)$	$O(\log_B N)$
$\epsilon=0$	$O\left(\frac{\log N}{B}\right)$	$O(\log N)$

10x-100x faster inserts

Performance Bounds



Clear Winner:  
Indexing With  
Write-Optimization

# TWIAD: Why You Should **Index Your IDS Data In Write-Optimized Databases**



## Common IDS Queries:

- “Have I seen x IP before?”
- “Have I seen x IP between times y and z?”
- “Have I seen x subnet before?”

# TWIAD: Why You Should **Index Your IDS Data In Write-Optimized Databases**

## TWIAD Key Construction:

- Origin IP - 4 bytes
- Timestamp - 8 bytes
- Origin Port - 3 bytes
- Destination IP - 4 bytes
- Destination Port - 3 bytes
- All inserts are stored twice
- Origin and Destination are reversed in the second insert
- Origin and Destination are now interchangeable

22 byte key

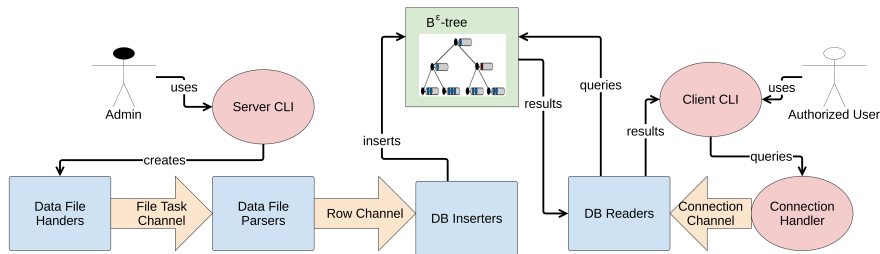
# TWIAD: Why You Should **Index Your IDS Data In Write-Optimized Databases**

## TWIAD Value Fields:

- Protocol - 4 bytes
- Duration - 8 bytes
- Origin Bytes - 8 bytes
- Response Bytes - 8 bytes
- Connection State - 5 bytes
- Origin Packets Bytes (sans header) - 4 bytes
- Response Packets Bytes (sans header) - 4 bytes
- Version - 6 bytes
- IsReversed Bit - 1 byte
- We include most BRO Connection Log fields
- Bytes are payload bytes vs Packets Bytes are all IP level bytes
- 

48 byte value

# TWIAD: Why You Should **Index Your IDS Data In Write-Optimized Databases**



TWIAD Architecture Diagram

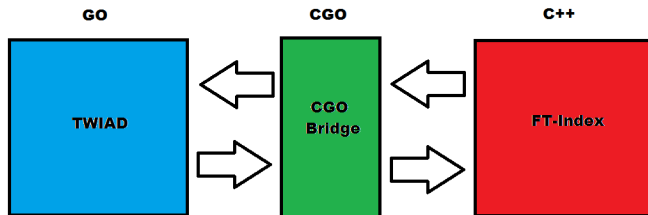
# TWIAD: Why You Should Index Your IDS Data In Write-Optimized Databases

Write Optimized databases should provide a faster infrastructure for storing IDS data

This lends itself to faster intrusion response with less hardware and lower costs

# Preliminary TWIAD Results

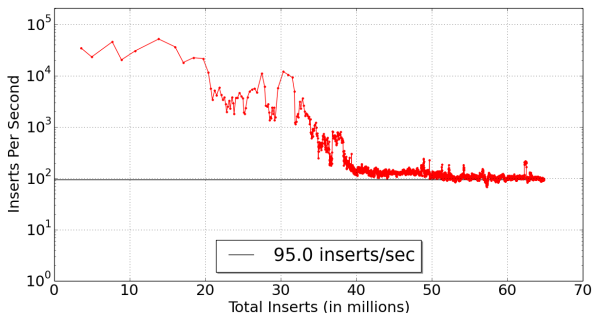
- When designing TWIAD, we decided to build on top of an open source Write-Optimized index: FT-Index
- We wrote TWIAD in GO, though FT-Index is written in C++
- CGO acts as an intermediary to link and execute C/C++ code with GO code
- GO has many convenient features for parallel computation: GoRoutines, Channels, Scheduling, etc.
- However, the CGO bridge to C++ may have hurt our performance





# Preliminary TWIAD Insertion Results

- TWIAD started with an insertion rate on the order of 10,000 inserts/second
- TWIAD's insertion rate then declined until it steadied out on the order of 100 inserts/second
- These results are from a desktop with 32GB of RAM and a 3.4GHz processor



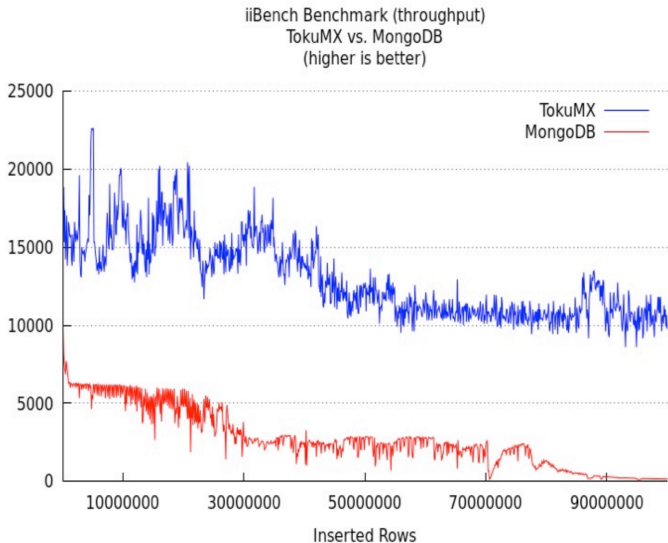
# Preliminary TWIAD Query Results

Table: TWIAD Querying 43M Entry DB For Fixed IP Over Time Range

Number of Entries Returned	Time (ms)
0	1318
1	14
35	192
1665	192
2132	222

- TWIAD boasts sub-second query responses to non-empty queries on a nontrivial DB
- TWIAD has order of one second query responses to empty queries on a nontrivial DB

# TokuMX Results

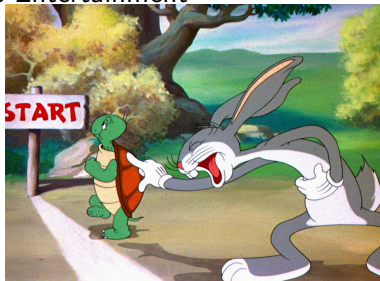


# Lessons Learned & Future Work

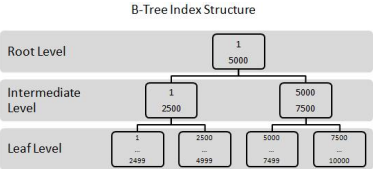
- Executing across several languages impacts performance
- Write Optimized indexes demand familiarity and tuning
- We plan to experiment with TWIAD running on top of alternative write-optimized indexes
- Also plan to experiment with inserting into multiple indexes and investigating more interesting IDS queries

# Credits

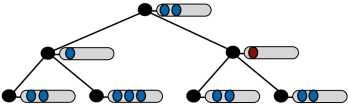
## Warner Brothers Entertainment



# Credits



simple-talk.com



Bender and Kuszmaul

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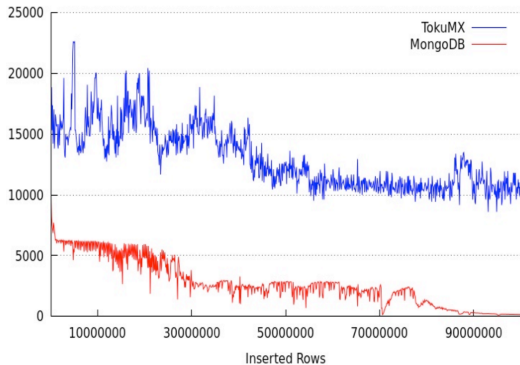
Bender and Kuszmaul



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# Credits

iiBench Benchmark (throughput)  
TokuMX vs. MongoDB  
(higher is better)



Bender and Kuszmaul