



**In-Memory
Computing**
S U M M I T

NORTH
AMERICA
2018

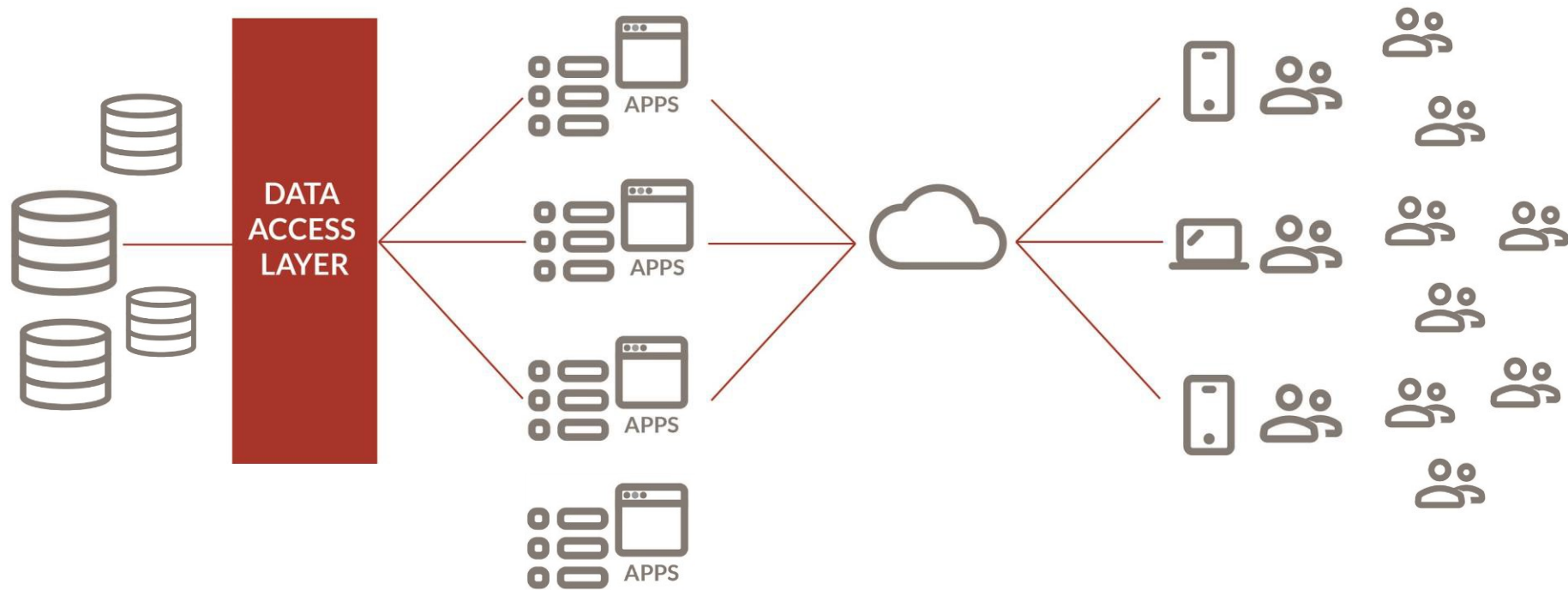
Improve Application Performance with SQL Auto-Caching, No Code changes

Roland Lee
Heimdall Data

Agenda

- 1) Database Proxy Introduction
- 2) Demo




Web scale Challenges



More users and applications implies

- Higher data volumes
- More tables, changes to schema
- **Higher latency responses**

Database Proxy Vendors

Feature	 HEIMDALL DATA	 ScaleArc	 MariaDB	ProxySQL
Automated Failover	✓	✓	✓	✓
SQL Read/Write Splitting	✓	✓	✓	✓
Automated Cache invalidation	✓			
Reduces network latency	✓			

IMDG or Database Proxy



Amazon
ElastiCache



Pivotal
GemFire®



redis



hazelcast

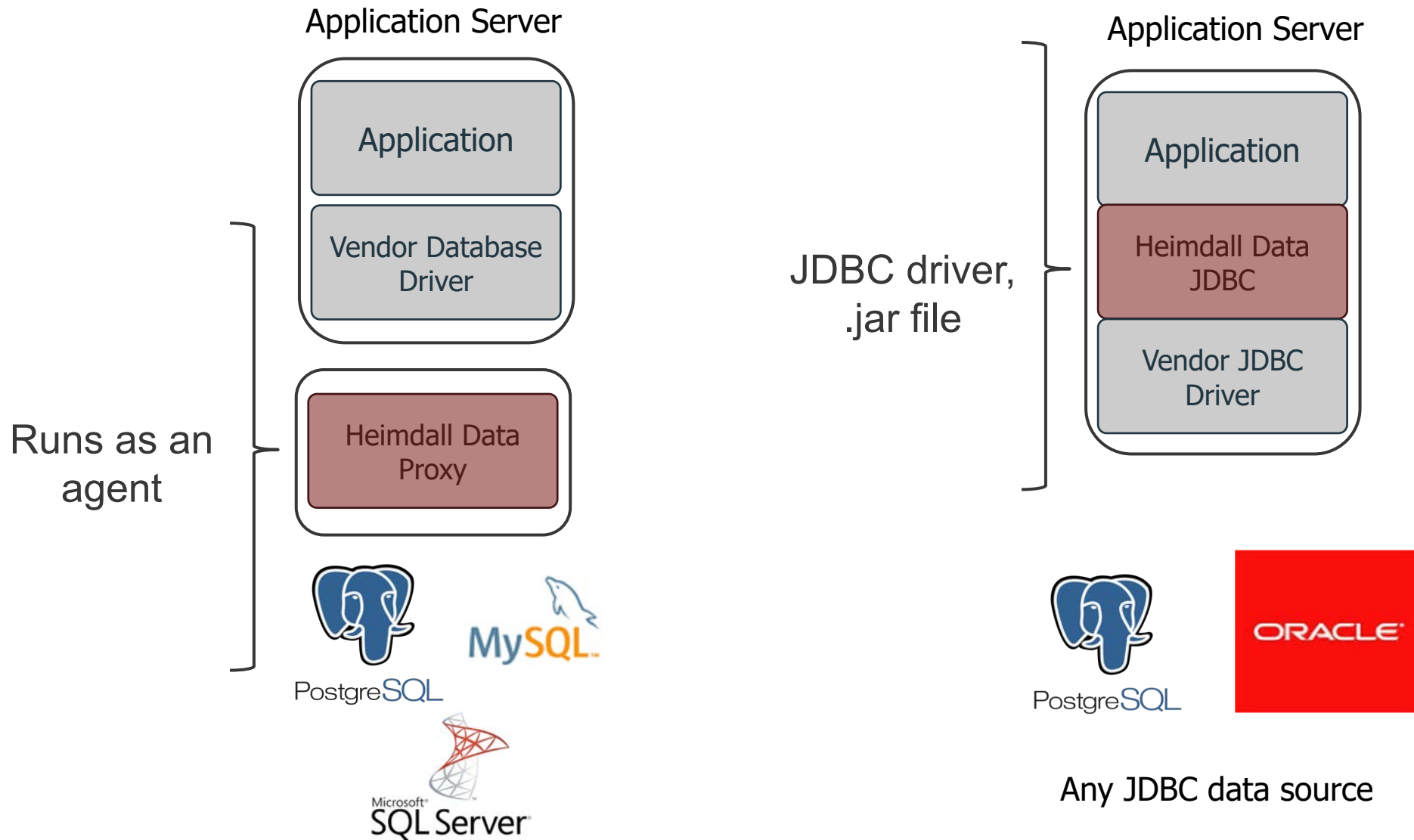


ORACLE®

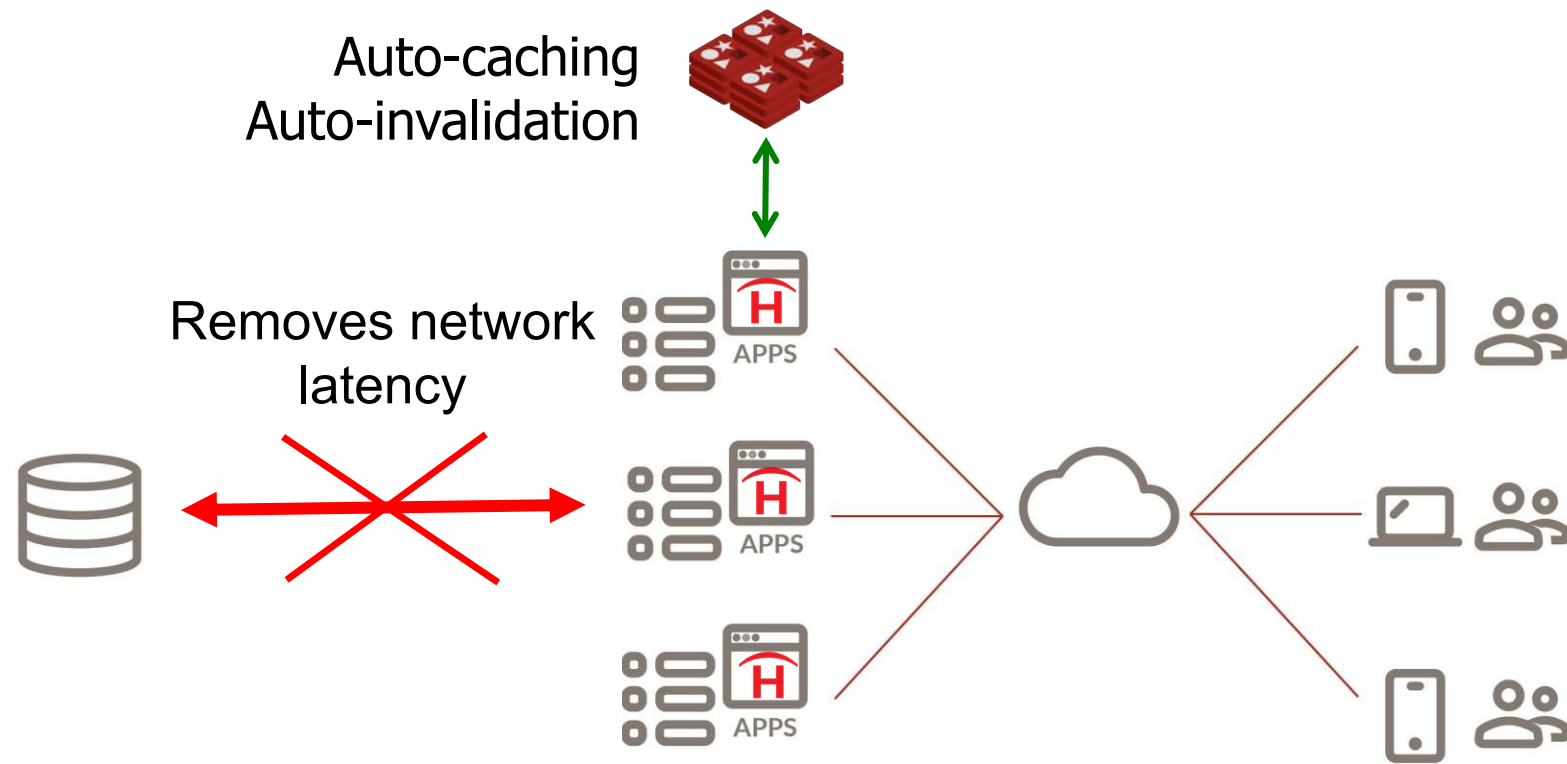
- Best scale & performance
- Greenfield applications
- Requires code changes

- Good scalability & performance
- Existing applications, small dev
- **No code changes**

Heimdall Data Software Options



Heimdall Data Distributed Proxy

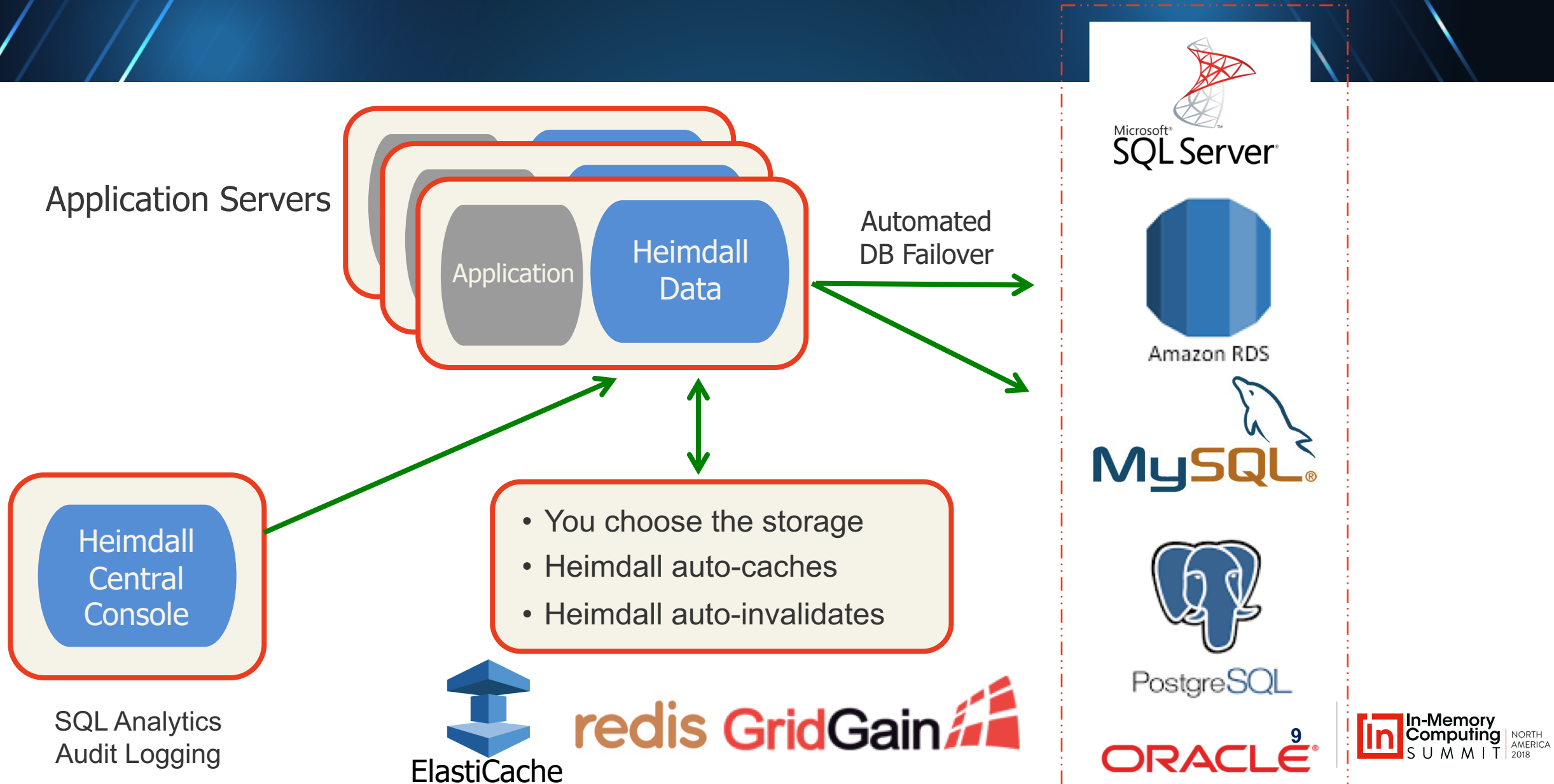


1. Heimdall is installed on each application instance
2. Direct Heimdall to the grid-cache of your choice (e.g. Redis, Hazelcast)
3. Heimdall will auto-cache and auto-invalidate SQL results to the look-aside cache

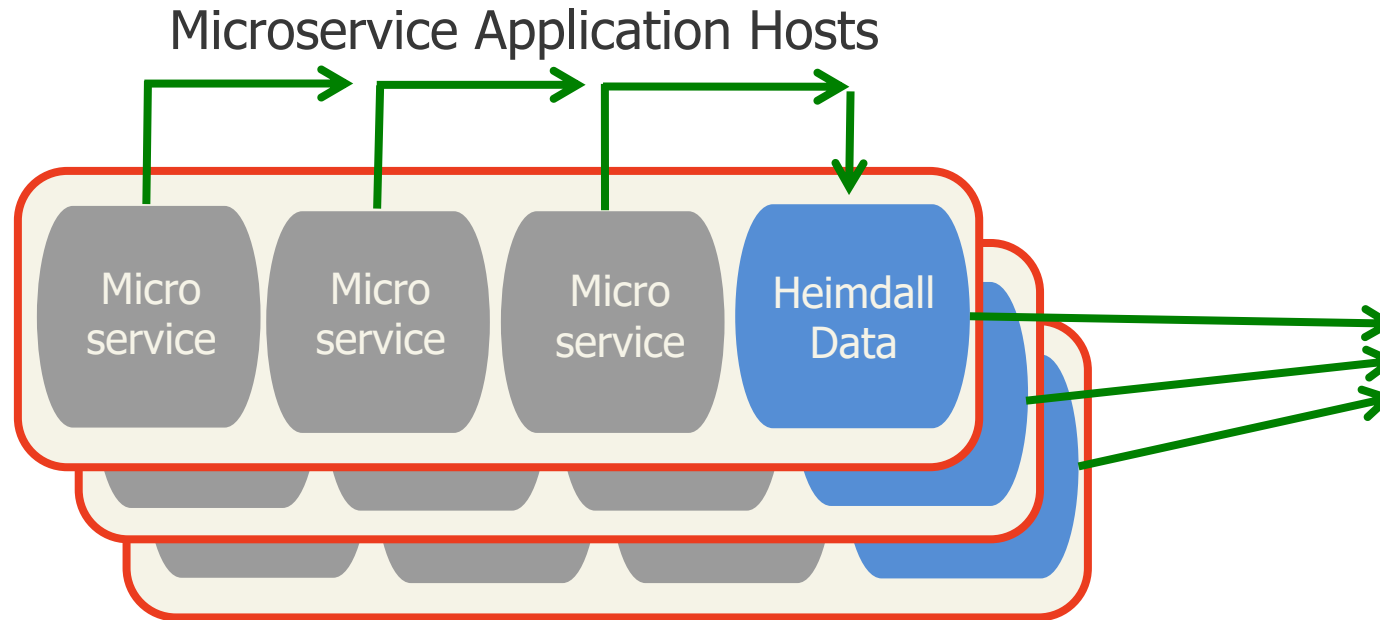
Heimdall Data Uses Cases

Use case	Customer Benefit
SQL Results Caching	<ul style="list-style-type: none">• Auto-caching / Auto-invalidate• NO code changes
Automated failover	<ul style="list-style-type: none">• Faster failover for MySQL & SQL Server AlwaysOn• PGPool-II Replacement
Batch Processing	<ul style="list-style-type: none">• Improves write performance• Batches singleton DML operations• Removes Commit overhead
Auditing for Privacy Compliance	<ul style="list-style-type: none">• Logs data access: Who, what, when• GDPR, SOX, PCI, HIPPA

Heimdall Data Architecture



Heimdall Data for Microservices



Heimdall Data Solution

DB Failover

Connection Pooling

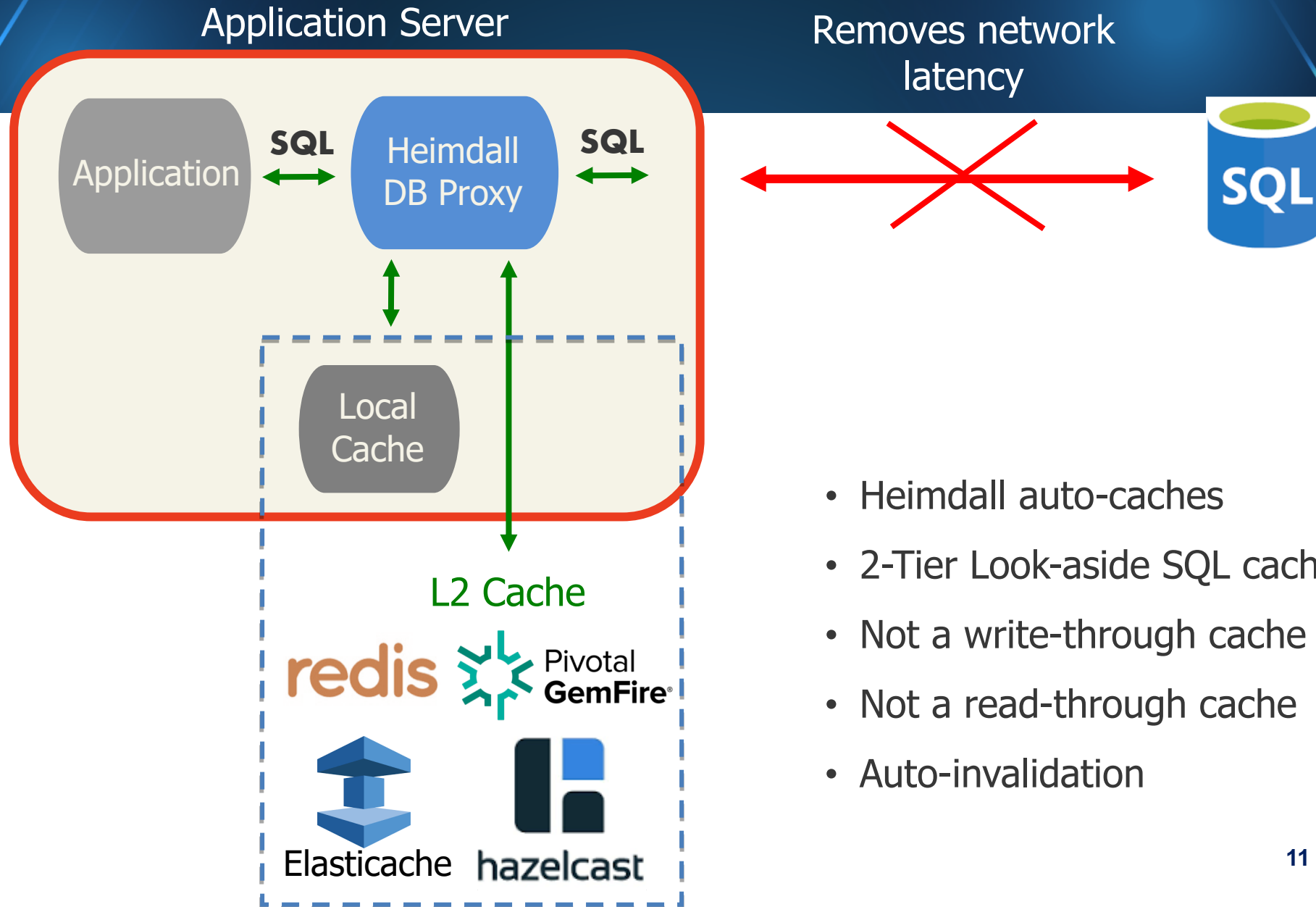
Caching

SQL Analytics

Audit logging

ORACLE[®] 10

Application Side – SQL Caching



What to Cache?

Uses real-time analysis and statistics on:

- Query frequency and variability
- Relative performance of Cache vs. Database

Provides:

- Auto-cache only if there is a performance benefit
- Cache recommendations and benefits

Heimdall Query Analytics

Heimdall Data Console

demoa.heimdalldata.com:8087/#/analytics

Very cacheable. 700 μ s per query

Rank	One-Click Optimize	Query	Server Time (%)	Duplicate Query & Response (%)	Cache Index	Cache Time (s)	Cache Hit (%)	Count	Query Response Time (μ s)	Result Retrieval Time (μ s)	Average Result Size
1	Cache	SELECT user_id, meta_key, meta_value FROM wp_usermeta WHERE user_id IN (?) ORDER BY umeta_id ASC	13	100	100	86400	0.0	2.1k	697.2	730.7	1k
2	Cache	SELECT * FROM wp_users WHERE ID = ?	13	100	100	86400	0.0	2.1k	693.7	733.2	310.0
3	Cache	SELECT * FROM wp_posts WHERE ID = ? LIMIT ?	8.9	100	100	86400	0.0	1.4k	756.1	836.8	2.9k
4	Cache	SELECT option_name, option_value FROM wp_options WHERE autoload = ?	7.1	100	100	86400	0.0	805.0	1026.5	1123.9	21.9k
5	Cache	SELECT option_value FROM wp_options WHERE option_name = ? LIMIT ?	6.8	100	100	86400	0.0	1.1k	691.7	707.6	6.9
6	Cache	SELECT wp_posts.* FROM wp_posts WHERE ?=? AND ((YEAR(wp_posts.post_date) = ? AND MONTH(wp_posts.post_date) = ? AND DAYOFMONTH(wp_posts.post_date) = ?) AND wp_posts.post_name = ? AND wp_posts.post_type = ? ORDER BY wp_posts.post_date DESC	5.4	100	100	86400	0.0	773.0	811.6	886.1	3k
7	Cache	SELECT t.*, tt.*, tr.object_id FROM wp_terms AS t INNER JOIN wp_term_taxonomy AS tt ON t.term_id = tt.term_id INNER JOIN wp_term_relationships AS tr ON tr.term_taxonomy_id = tt.term_taxonomy_id WHERE tt.taxonomy IN (?, ?, ?) AND tr.object_id IN (?) ORDER BY t.name ASC	5.1	100	100	86400	0.0	804.0	733.1	784.2	909.4
8	Cache	SELECT post_id, meta_key, meta_value FROM wp_postmeta WHERE post_id IN (?) ORDER BY meta_id ASC	4.7	100	100	86400	0.0	804.0	680.7	700.2	1.0
9	Cache	SELECT wp_posts.* FROM wp_posts WHERE ID IN (?, ?, ?, ?, ?)	2.4	100	100	86400	0.0	340.0	813.1	915.2	13k
10	Cache	SELECT wp_comments.* FROM wp_comments WHERE comment_ID IN (?, ?, ?, ?, ?)	2.3	100	100	86400	0.0	351.0	771.2	840.0	5.9k
11	Cache	SELECT wp_posts.ID FROM wp_posts WHERE ?=? AND wp_posts.post_type = ? AND ((wp_posts.post_status = ?)) ORDER BY wp_posts.post_date DESC LIMIT ?, ?	2.3	100	100	86400	0.0	339.0	791.5	811.0	120.0
		SELECT SQL_CALC_FOUND_ROWS wp_comments.comment_ID FROM wp_comments WHERE (

Heimdal Cache Rules Analytics

Before: Query response time 700 μ s per query. Significant repetitive database load.

Rank	One-Click Optimize	Query	Server Time (%)	Duplicate Query & Response (%)	Cache Index	Cache Time (s)	Cache Hit (%)	Count	Query Response Time (μ s)	Result Retrieval Time (μ s)	Average Result Size
1	Cache	SELECT user_id, meta_key, meta_value FROM wp_usermeta WHERE user_id IN (?) ORDER BY umeta_id ASC	13	100	100	86400	0.0	2.1k	697.2	730.7	1k
2	Cache	SELECT * FROM wp_users WHERE ID = ?	13	100	100	86400	0.0	2.1k	693.7	733.2	310.0
3	Cache	SELECT * FROM wp_posts WHERE ID = ? LIMIT ?	8.9	100	100	86400	0.0	1.4k	756.1	836.8	2.9k

After: Cached response in 70 μ s per query. Reduced database load.

20	Caching	SELECT user_id, meta_key, meta_value FROM wp_usermeta WHERE user_id IN (?) ORDER BY umeta_id ASC	1.8	100	100	86400	99.9	55k	68.4	68.4	1.9
21	Caching	SELECT * FROM wp_users WHERE ID = ?	1.6	100	100	86400	99.9	54.9k	59.5	59.5	1.3
22	Caching	SELECT * FROM wp_posts WHERE ID = ? LIMIT ?	1.4	100	100	86400	97.8	35.4k	77.8	79.3	63.5



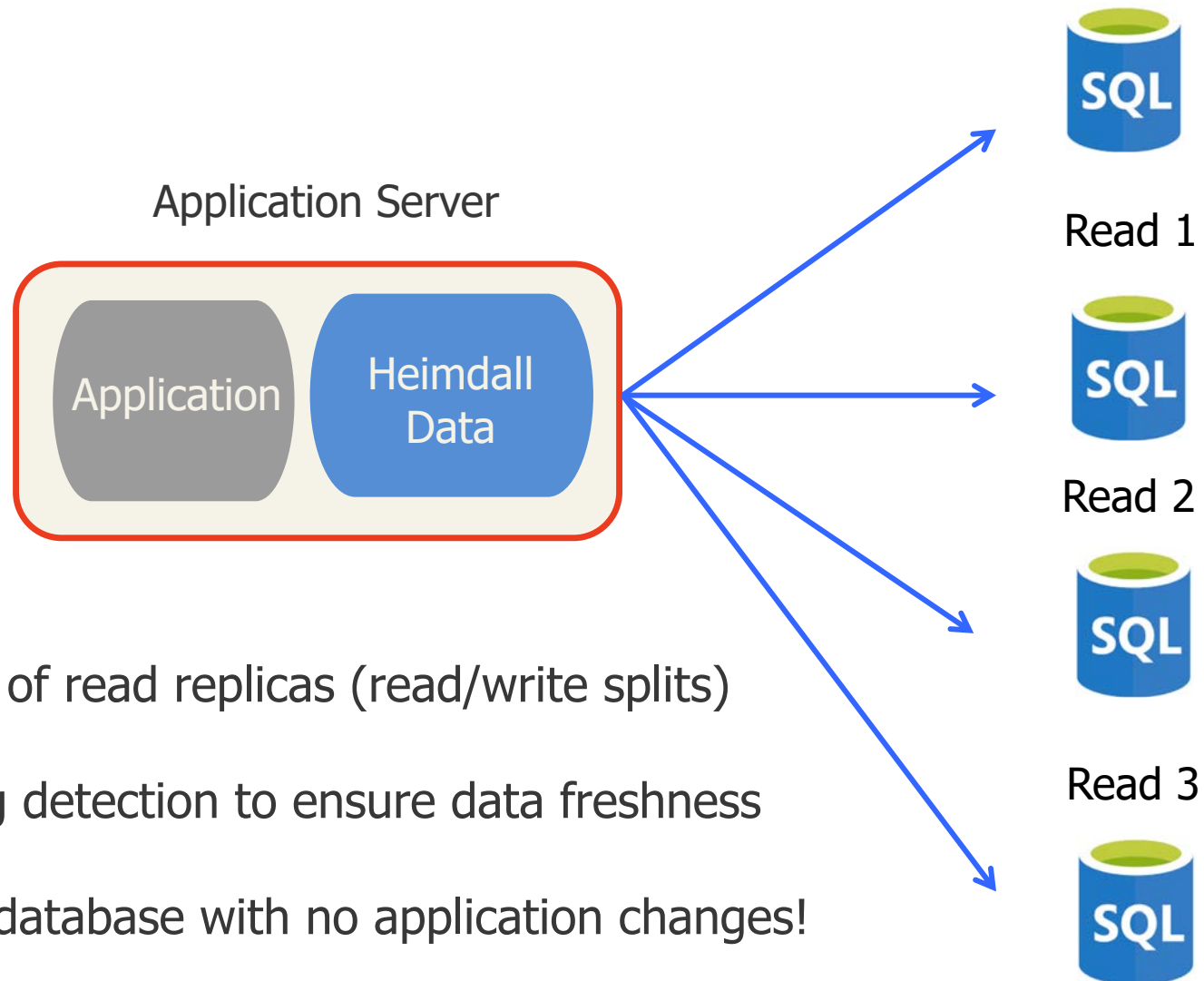
**In-Memory
Computing**
S U M M I T

NORTH
AMERICA
2018

Other use cases

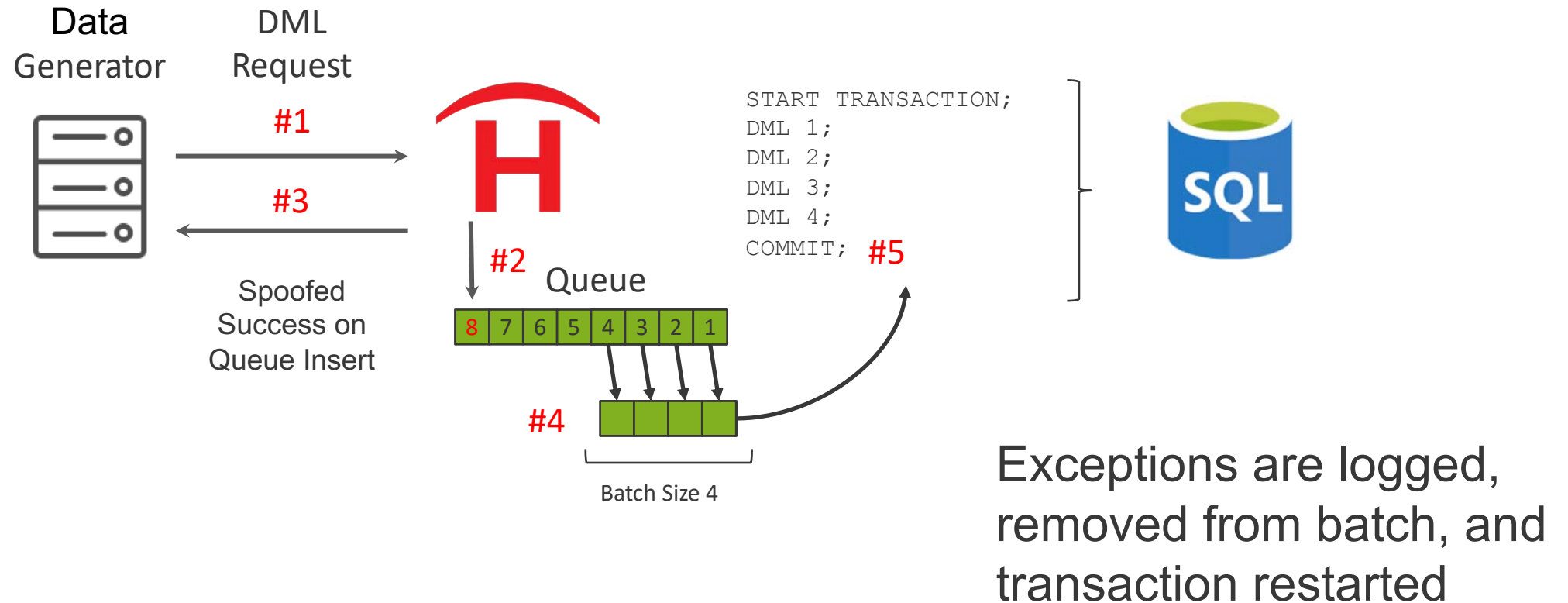
Other use cases

Use Case #1: Read / Write Splitting



1. Automate use of read replicas (read/write splits)
1. Replication lag detection to ensure data freshness
2. Scale out the database with no application changes!

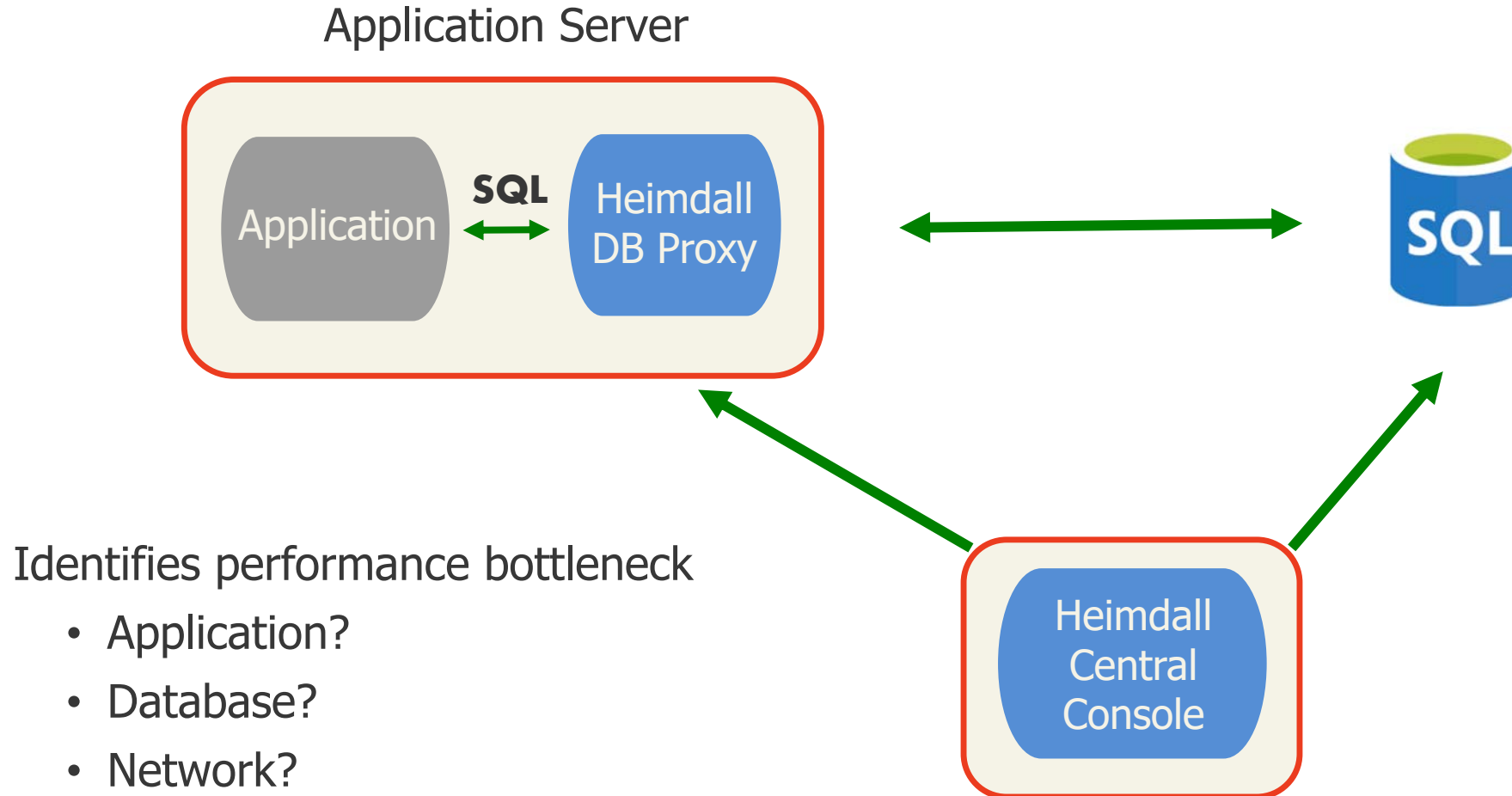
Use Case #2: Async DML Batching



Benefits:

- Lower CPU overhead due to fewer commits
- Improved application response time
- Improved DML scale

Use Case #3: SQL Analytics



Demo



**In-Memory
Computing**
S U M M I T

NORTH
AMERICA
2018

Thank you

Roland Lee
Heimdall Data