

# Optimizing Query Performance in Cloud Data Warehousing

A Comparative Analysis of Microsoft Azure Synapse, Amazon Redshift, and Oracle Autonomous Data Warehouse

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# Research Context & Problem

## **Context**

- Cloud data warehouses now essential for modern analytics
- SMEs face challenges selecting optimal platforms
- Performance varies significantly across platforms

## **Research Gap**

- Conflicting benchmark results in literature
- No neutral comparison under identical conditions
- Limited guidance for SME adoption decisions

## **Research Objectives**

1. Benchmark three leading platforms using TPC-DS and TPC-H at 10GB scale
2. Evaluate platform-specific optimization strategies
3. Develop evidence-based decision framework for SMEs

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# Research Methodology

## Seven-Phase Experimental Pipeline



### Benchmarks

- TPC-DS: 99 queries, 24 tables
- TPC-H: 22 queries, 8 tables
- Scale Factor: 10GB (SME-representative)

### Statistical Analysis

- Kruskal-Wallis H-test ( $\alpha=0.05$ ) - Tests whether there are statistically significant differences in median query latency across the three platforms and complexity level
- Dunn's post-hoc comparisons - After Kruskal-Wallis detects differences, Dunn's test identifies which specific pairs of platforms differ from each other.
- Cliff's Delta effect sizes - Quantifies how large the performance difference is between platforms, not just whether it's statistically significant.
- $n=20$  iterations per query

# Platform Configurations

**Strategy:** Minimum production-tier matching to reflect real-world SME adoption constraints

Platform	Configuration	Cost/Hour	Justification
Azure Synapse	DW200c	\$1.60	Vendor minimum Gen2 production tier
Amazon Redshift	2 × RA3.large	\$1.08	Minimum RA3 cluster (32GB per node)
Oracle ADW	4 ECPU	\$1.04	Minimum ADW ECPU configuration

# Baseline Performance Results

## Baseline Performance Results TPC-DS Performance

Platform	p50 (s)	p99 (s)	QPH
Oracle ADW	0.014	0.225	148,072
Redshift	0.019	0.229	191,168
Synapse	4.080	27.342	2,414

## Baseline Performance Results TPC-H Performance

Platform	p50 (s)	p99 (s)	QPH
Oracle ADW	0.016	9.221	21,189
Redshift	0.018	3.271	71,338
Synapse	5.389	22.495	2,110

### Oracle ADW

Lowest median latency, exceptional consistency

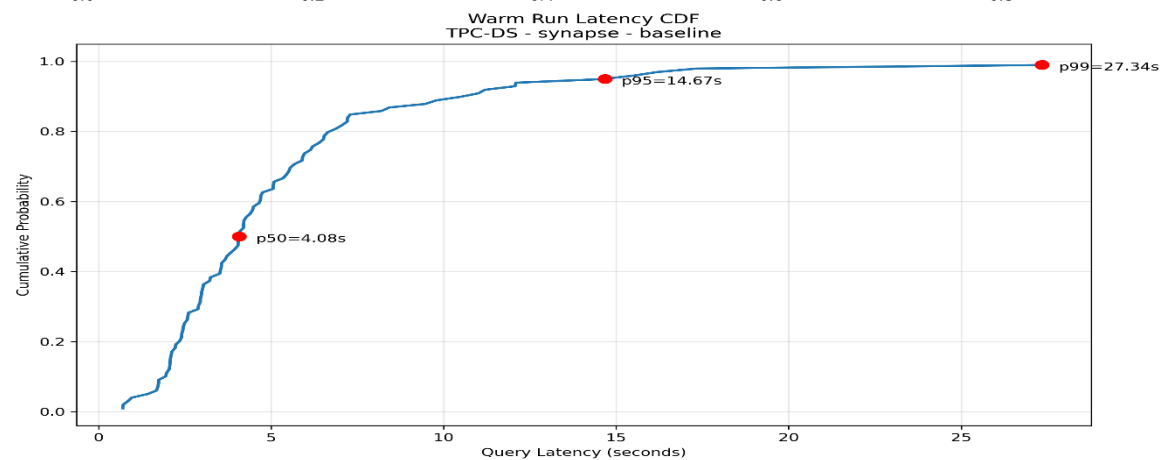
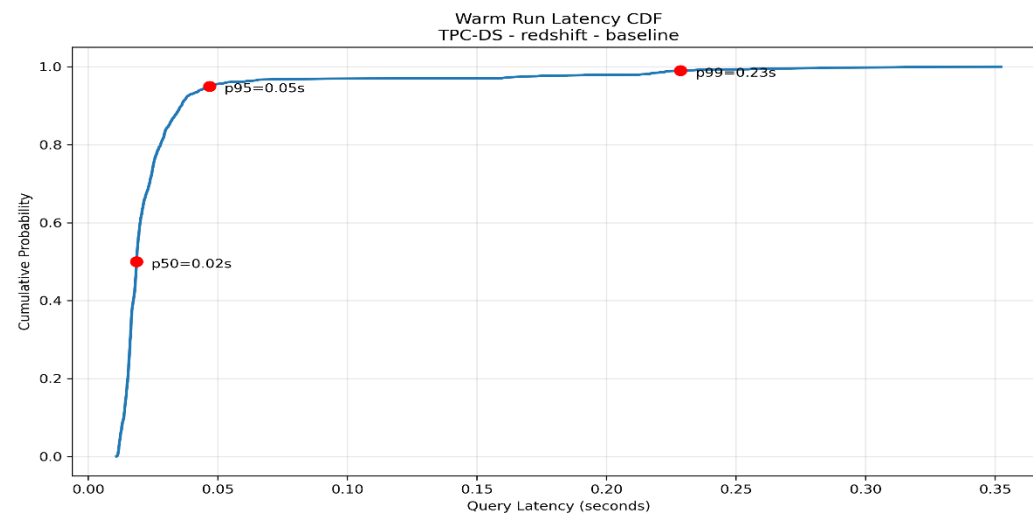
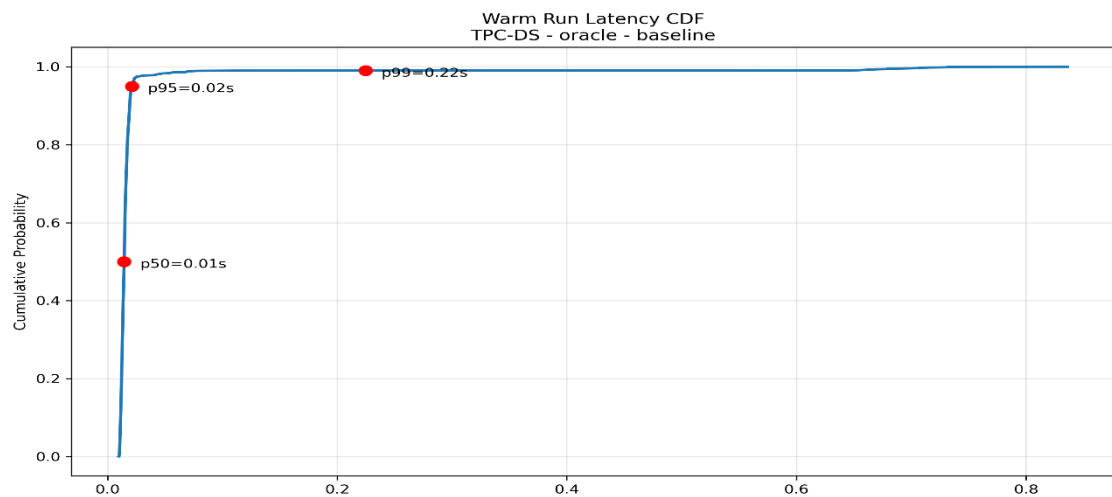
### Redshift

Competitive performance, superior TPC-H throughput

### Synapse

DW200c inadequate for production workloads

# Baseline Performance latency Results



# Optimization Intervention Results

## 1. Compression Tuning

-0.97%

Oracle ADW (TPC-DS)

-3.38%

Redshift (TPC-DS)

-49.23%

Synapse (TPC-H)

## 2. Distribution Key and Partition Optimization

-3.98%

Redshift (TPC-DS)

Top: query 20 (-30.90%)

+123.01%

Synapse (TPC-DS)

79 queries degraded

+1.32%

Oracle Partitioning (TPC-DS)

Mixed results (query 34 -98.12%, query10 +37.00%)

## 3. Materialized Views

-0.61%

Oracle (TPC-DS)

-9.60%

Redshift (TPC-H)

+1.16%

Synapse (TPC-DS)

**Key Finding:** Optimization effectiveness highly variable and platform-dependent. Resource constraints (Synapse DW200c) can negate traditional optimization strategies.

# Concurrency Scaling Performance

## TPC-DS Throughput Scaling (1 → 20 concurrent sessions)

12.2× Oracle ADW 148K → 1.8M QPH Latency: 0.014s → 0.020s	1.8× Redshift 191K → 354K QPH Latency: 0.019s → 0.201s	1.4× Synapse 2.1K → 3.0K QPH Latency: 5.6s → 63.6s
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### Oracle ADW

- Superlinear throughput scaling
- Minimal latency degradation (1.4×)
- Autonomous resource management
- Excellent p99 stability (0.108s → 0.153s)

### Redshift & Synapse

- Redshift: p99 explosion (0.228s → 1.548s)
- Workload queue limitations evident
- Synapse: 11.3× median latency increase
- Resource constraints prevent scaling

**Implication:** Autonomous optimization (Oracle) provides superior multi-user performance. Manual tuning required for Redshift. Synapse requires higher tier provisioning.

# Statistical Analysis

## Statistical Analysis with Optimizations (TPC-DS):

**Kruskal-Wallis H-test:** Would still reject  $H_0$  ( $p \approx 0.000$ ). The gap between Synapse and others remains enormous.

## Dunn's Post-Hoc Comparisons:

- **Synapse vs. Oracle ADW:** Extremely significant ( $p \approx 0.000$ )
- **Synapse vs. Redshift:** Extremely significant ( $p \approx 0.000$ )
- **Oracle ADW vs. Redshift:** Now more competitive. Oracle has better p50 and p99, but Redshift has higher QPH.

## Cliff's Delta Effect Sizes:

- **Synapse vs. Oracle ADW:**  $\delta \approx +1.0$  (Large)
- **Synapse vs. Redshift:**  $\delta \approx +1.0$  (Large)
- **Oracle ADW vs. Redshift:**  $\delta \approx -0.15$  (Negligible/Small) - Very similar performance profiles

# Key Research Findings

## 1. Performance Disparities

Oracle ADW demonstrated lowest median latency (0.014s TPC-DS) and exceptional consistency. 61× throughput advantage over Synapse's entry-level configuration.

## 2. Optimization Effectiveness Varies

Compression: 3-49% improvements. Distribution keys: -30% to +123% variance. Materialized views: 0.6-68% per-query effects. Results highly dependent on platform maturity and resource provisioning.

## 3. Resource Provisioning Critical

Azure Synapse DW200c proved inadequate for production workloads. Traditional optimization strategies can degrade performance under resource constraints.

## 4. Autonomous vs Manual Tuning

Oracle's autonomous optimization provided consistent performance without manual intervention. Redshift and Synapse require DBA expertise for optimization.

## 5. Workload Sensitivity

Platform performance varies by query complexity. Oracle excels at simple queries (caching), Redshift competitive on moderate joins, all platforms struggle with complex nested subqueries at 10GB scale.

# Platform-Specific Recommendations

## Oracle ADW

Best For:

- Operational simplicity priority
- Multi-user analytics
- Limited DBA expertise

Considerations:

- Higher cost (\$1.04/hr base)
- Provision 8+ ECPUs for TB-scale
- Excellent consistency

## Amazon Redshift

Best For:

- Technical teams
- Cost-conscious deployments
- OLAP-heavy workloads

Considerations:

- Requires DBA tuning
- Distribution key critical
- Best cost efficiency

## Azure Synapse

Best For:

- Azure ecosystem users
- Power BI integration
- Adequate provisioning ( $\geq$ DW500c)

Considerations:

- DW200c inadequate
- Requires tier planning
- Schema design critical

## Decision Framework

- Operational simplicity priority: Oracle ADW (autonomous optimization, compression yields modest improvements; check heavy outliers)
- Cost optimization with technical expertise: Amazon Redshift (Distribution key produced the largest overall improvement in your runs, mean -3.98%)
- Azure ecosystem lock-in: Synapse with  $\geq$ DW500c provisioning (Tuning had mixed effects)
- High concurrency requirements: Oracle ADW (12.2 $\times$  scaling)

# Conclusions & Future Directions

## Limitations

- 10GB scale factor (SME-focused but limits TB/PB generalizability)
- Single region deployment (US-East)
- Temporal validity (October 2025 snapshot)
- Resource imbalance across configurations

## Research Contributions

- ✓ Neutral tri-platform comparison under controlled conditions
- ✓ Explicit cold/warm run separation methodology
- ✓ Marginal attribution of optimization effects
- ✓ Telemetry-driven root cause analysis
- ✓ Evidence-based SME decision framework

## Key Takeaways

**Platform Choice:** No universal winner; depends on workload and expertise

**Provisioning:** Entry-level tiers may prove false economies

**Optimization:** Highly context-dependent; test before deploying

## Future Research Directions

- Larger scale factors (SF100-SF1000) for enterprise validation
- Multi-region performance and cost analysis
- Real-world workload traces beyond synthetic benchmarks
- Cost-per-query economic modeling
- Synapse Spark pool evaluation for complex analytics

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