

SQL Database Monitoring Tools: Selection and Workflow Design



- Define scope: selecting tooling and designing monitoring workflows



- Core skills for DBAs, platform engineers, and data engineers



- We cover: vocabulary, signals, tool landscape, and selection framework



- Also workflow design, cost control, and future observability trends



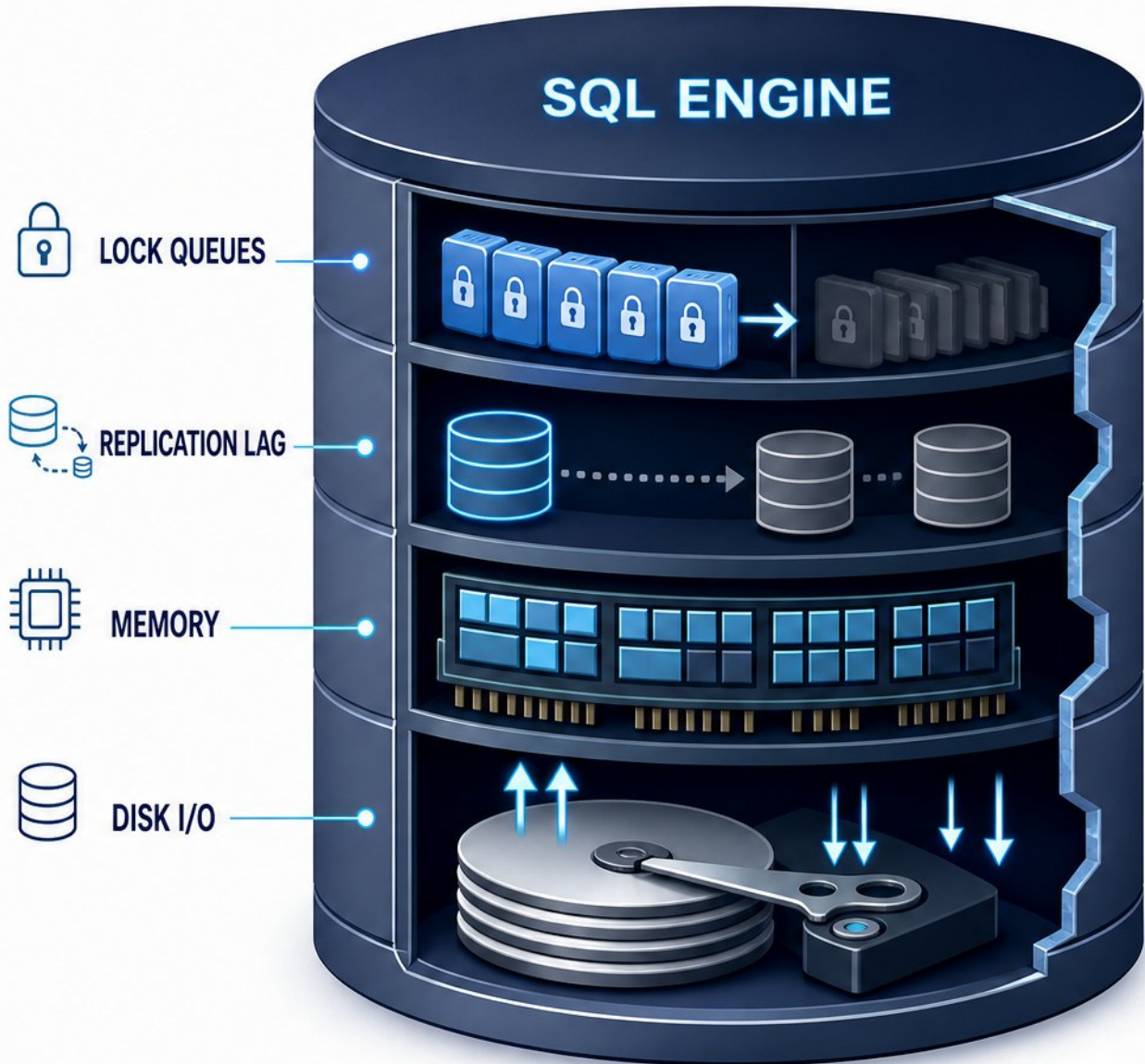
Core Vocabulary: Monitoring, Observability, and Telemetry

- **Monitoring:** threshold-based alerts for known failure modes
- **Observability:** ad-hoc querying to investigate novel failures
- **Telemetry pillars:** metrics, logs, and traces
- SLOs, SLIs, and error budgets drive reliability discussions
- Monitoring detects; observability explains root cause

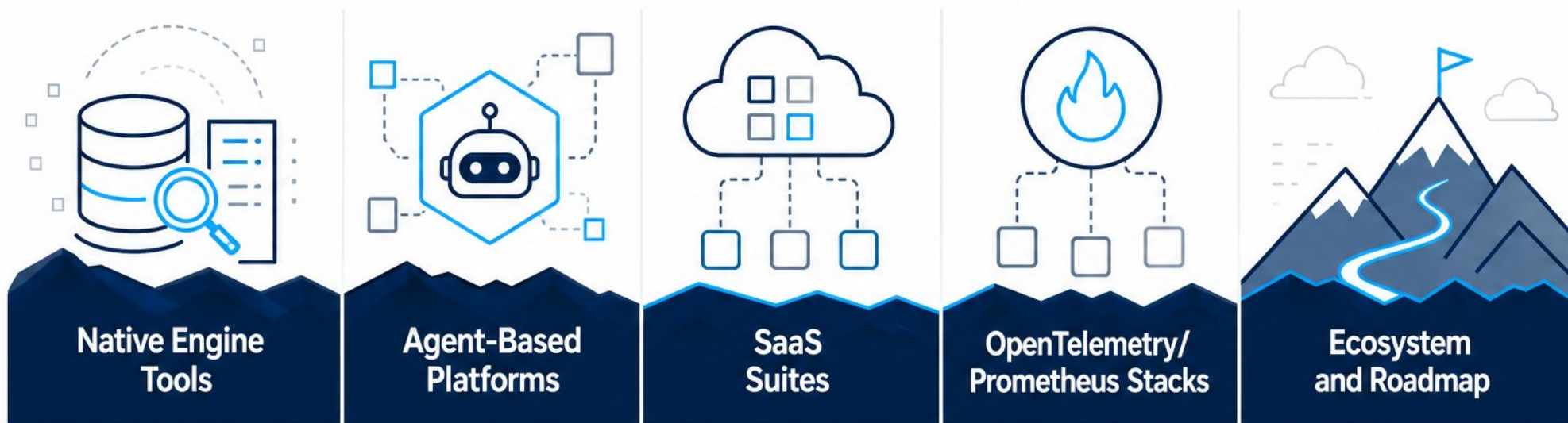


What SQL Databases Require Monitoring

- Track query latency, throughput, locks, disk I/O, memory, connections, replication lag, and index health
- Map signals to user impact: timeouts, split-brain replication, and data corruption
- Engine-specific views: `pg_stat_*`, SQL Server Query Store, MySQL Performance Schema
- Managed SQL limits: RDS, Aurora, Cloud SQL, Azure SQL expose telemetry via cloud APIs



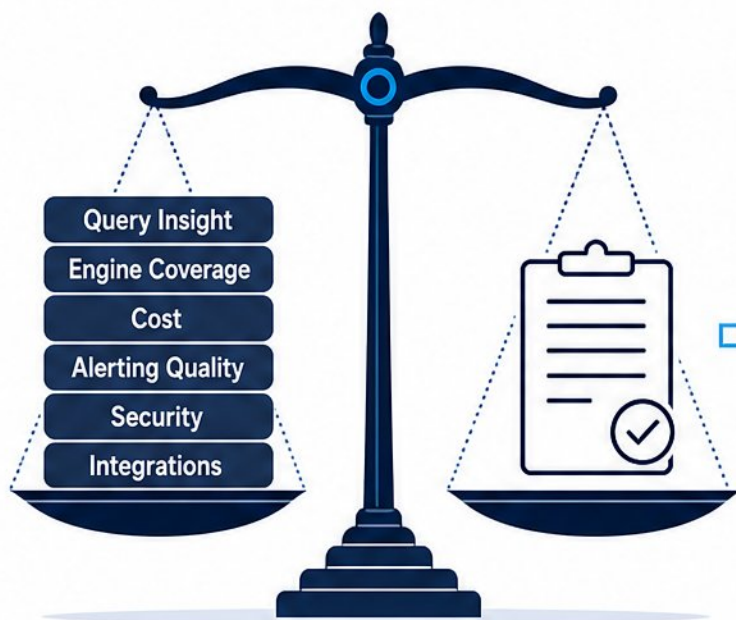
Monitoring Tool Categories and the 2026 Landscape



- ▶ Native engine tools and agent-based platforms offer granular visibility
- ▶ SaaS suites and OpenTelemetry/Prometheus stacks trade depth for breadth
- ▶ Agent overhead, cloud support, and cost shape tool selection
- ▶ Leaders: Datadog, Dynatrace, Grafana Cloud, SolarWinds, Redgate Monitor
- ▶ Open-source depth: Percona PMM, pganalyze; plus native views like `pg_stat_*`

Evaluation Criteria and a Repeatable Selection Framework

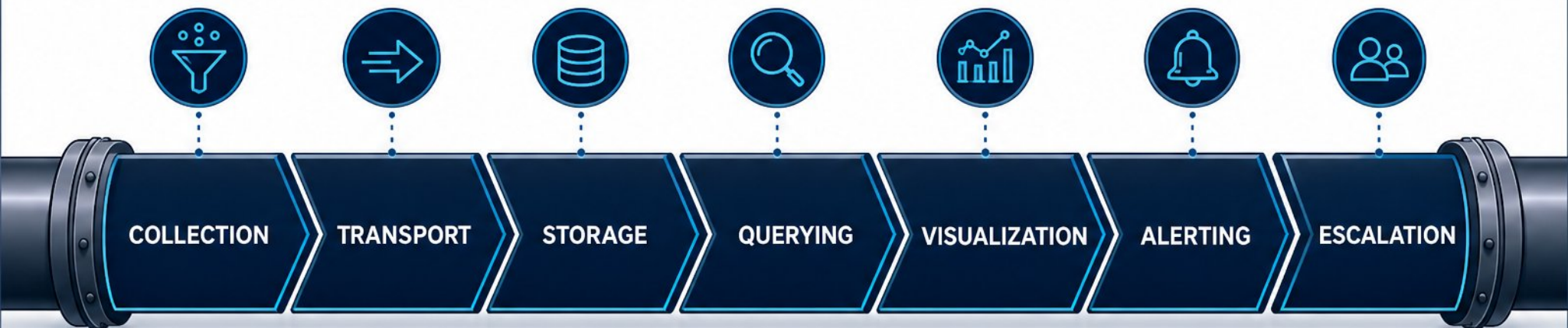
- Weight criteria: query insight, engine coverage, cost, alerting quality, security, integrations
- Check baseline needs of mixed SQL estates: PostgreSQL, MySQL, SQL Server, managed services
- Score tools objectively; avoid bias toward brand or familiarity



EVALUATION CRITERIA (WEIGHTED)					
					

TOOL SHORTLIST (RANKED)	
	
	
	
	
	

Architecting the Monitoring Workflow



- ▶ Define workflow stages: collection, transport, storage, querying, visualization, alerting, escalation
- ▶ Map ownership across DBA, platform, and data engineering roles
- ▶ Set ingestion cadence, cardinality controls, and sampling per metric
- ▶ Balance fidelity and cost with tiered retention policies
- ▶ Preserve high-fidelity traces for high-value templates; sample the rest

Designing Alerts for SLOs, Not Noise



- Thresholds derived from SLOs and error budgets, not raw metric spikes



- Severity levels, aggregation windows, deduplication, and hysteresis reduce alert fatigue



- Connect alerts to runbooks, on-call rotations, and post-incident reviews



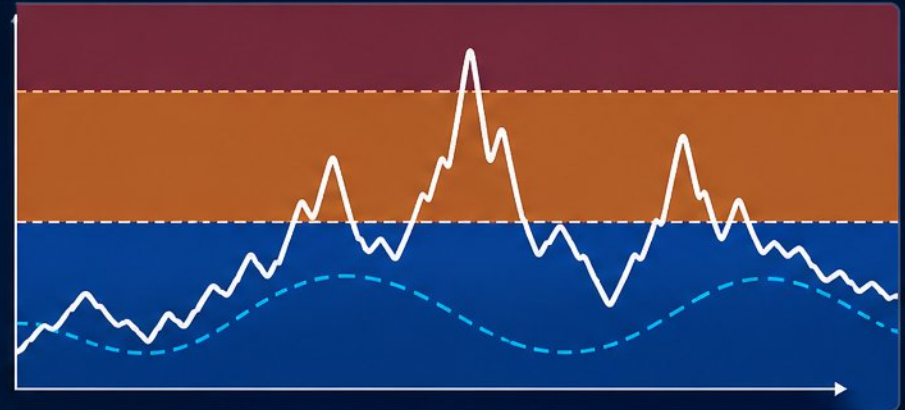
- Dynamic baselines beat static thresholds for cyclical workloads



- Closed-loop learning: incidents refine thresholds and dashboards

Alert Thresholds Driven by SLOs and Error Budgets

- Error budget exhausted
- High risk (approaching budget limit)
- Healthy (within budget)
- Dynamic baseline



Severity levels



Aggregation windows



Deduplication



Hysteresis



Security, Compliance, and Access Controls for Telemetry



Encrypt telemetry in transit and at rest



Limit exposure of query text, credentials, and customer data



Meet GDPR, HIPAA, SOC 2, and audit requirements



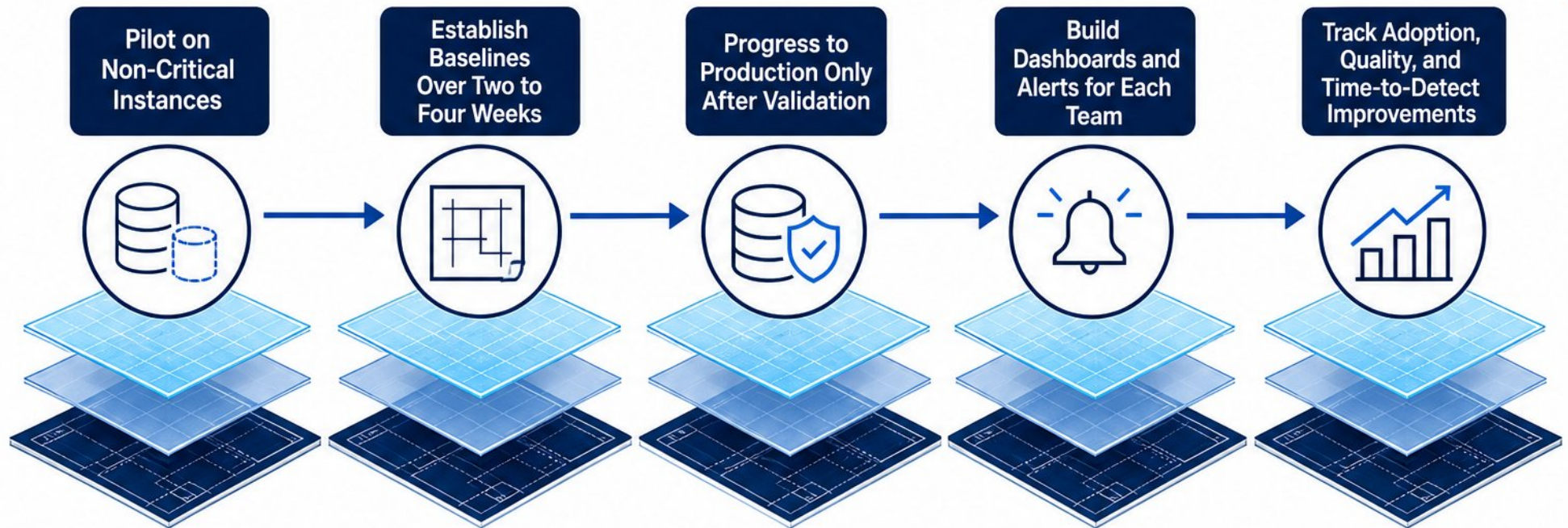
Apply least-privilege access to dashboards and raw logs



Mask and redact query text in shared observability stores



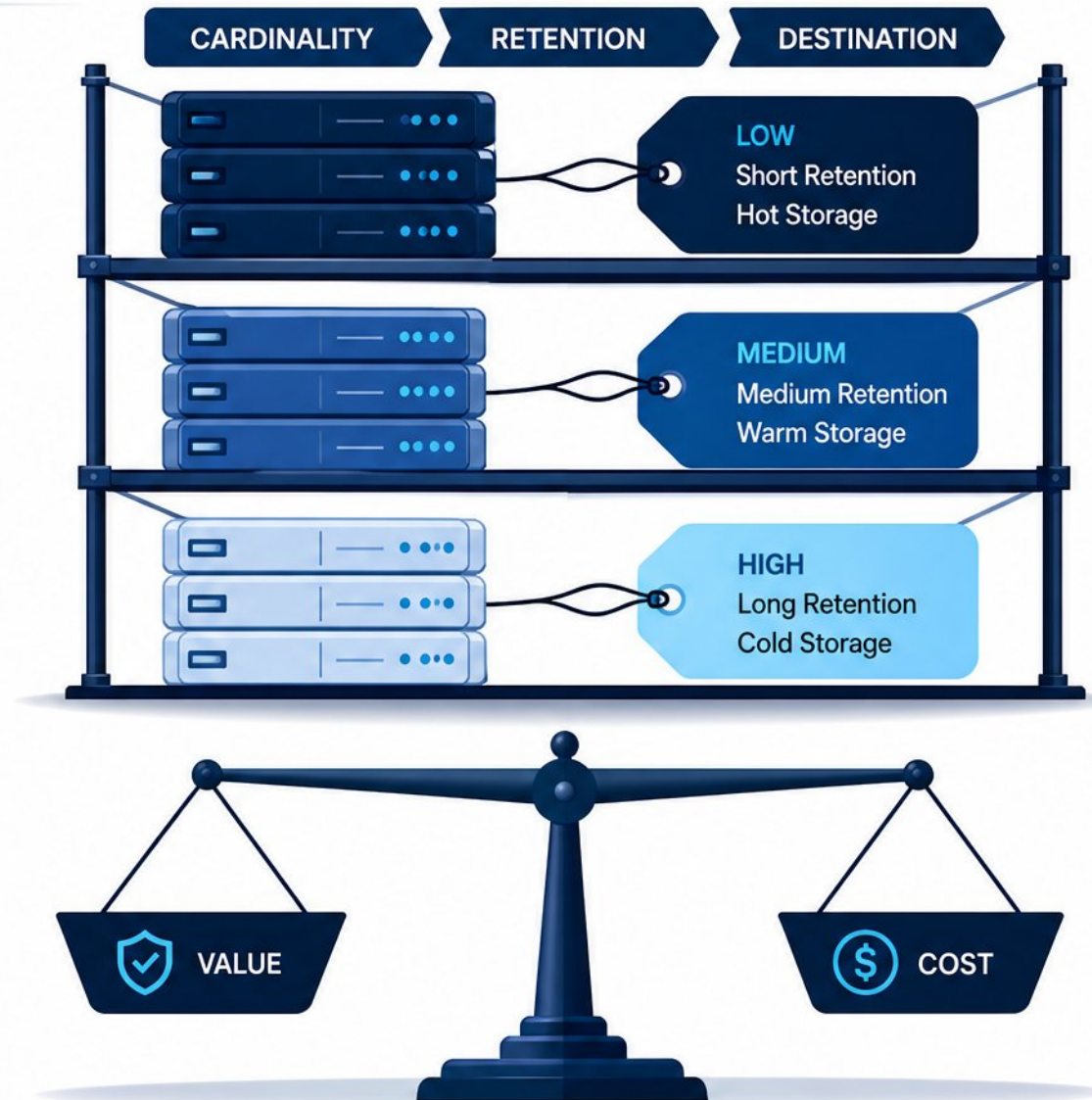
Implementation and Rollout Strategy



- ▶ - Start with pilot on non-critical instances
- ▶ - Establish baselines over two to four weeks
- ▶ - Progress to production only after validation
- ▶ - Build dashboards and alerts for each team
- ▶ - Track adoption, quality, and time-to-detect improvements

Cost Control and Cardinality Management

- Control costs with cardinality limits, sampling, and retention tiers
- Keep telemetry that detects, explains, or validates incidents
- Attribute cost per database, telemetry type, and owner
- Total cost formula: license + implementation + triage over 3 years
- Right-size instances and decommission idle resources to reduce spend



Operationalizing Monitoring: Dashboards, Runbooks, and Ownership



- Use SRE overview, DBA forensics, engineering self-service dashboards



- Link alerts to diagnostic queries, steps, escalation in runbooks



- Define ownership for collection, storage, alerting, dashboards



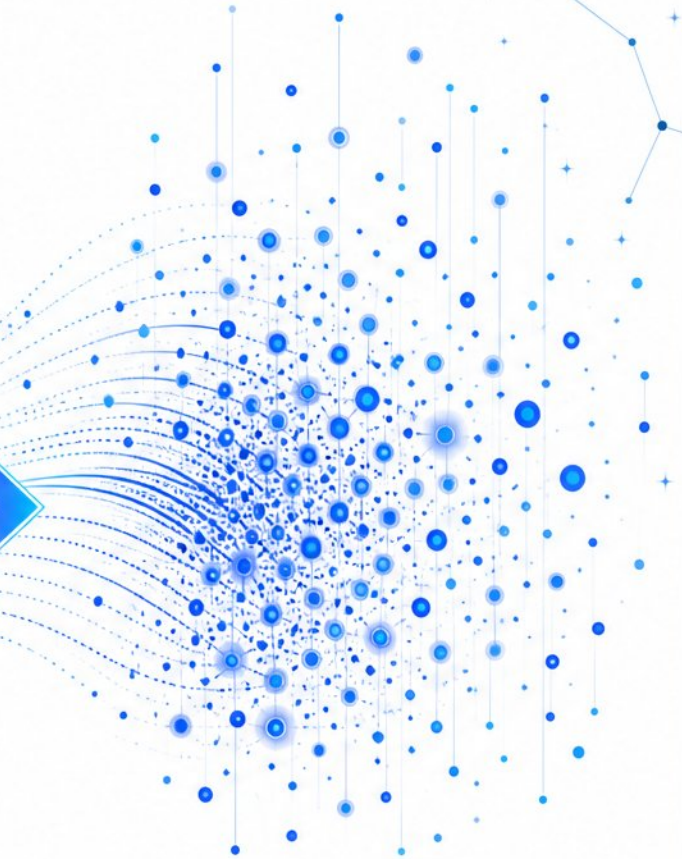
- Review incidents to refine thresholds, dashboards, workflows

Future Trends and the Path to Observability 2.0

- AI-assisted anomaly detection and predictive cost models
- OpenTelemetry as the de facto collection standard
- Observability 2.0: raw, high-cardinality events, query-time slicing
- Shift from pre-aggregated metrics to arbitrary dimension grouping
- Plan for Performance Insights EOL and Database Insights migration
- Consolidate agents and collectors across the SQL estate



PRE-AGGREGATED
METRIC CUBES



RAW, HIGH-CARDINALITY
EVENT STREAMS

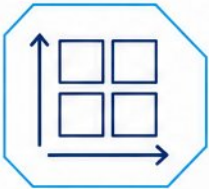
Workshop: Scoring Your Current Monitoring Stack



- Apply the weighted evaluation framework from module 4 to your own SQL estate



- Run a gap analysis: missing signals, alert noise, ownership, compliance, cost overruns



- Prioritize next steps with an impact/effort matrix



- Output: one agreed action per team for the next sprint



CRITERIA	CURRENT STATE	GAPS	ACTION NOTES
			
			
			
			
			

Key Takeaways and Recommended Next Steps



- Monitoring detects known failures; observability diagnoses novel ones; both are complementary



- Tool selection driven by weighted criteria, not vendor familiarity



- Workflow design, alerting, and cost control are ongoing practices



- Next: baseline telemetry, run a tool trial, establish SLO-based alerting on one critical DB

RECOMMENDED NEXT STEPS



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