

Management Framework

Manager of Software Engineering Systems

January 29, 2026

Abstract

This document serves as a comprehensive guide covering the Management Technical Knowledge Framework, Leadership Skills, and Standard Operating Procedures.

Contents

| | | |
|----------|--|-----------|
| 1 | Introduction | 1 |
| 1.1 | Summary | 1 |
| 1.2 | Guiding Principles | 1 |
| 2 | Management Technical Knowledge Framework | 2 |
| 2.1 | Technical Area: Artificial Intelligence (AI) and Machine Learning | 2 |
| 2.1.1 | Technical Competency Map: AI and Intelligent Systems | 3 |
| 2.1.2 | Fundamental Knowledge Tree | 3 |
| 2.2 | Technical Area: Data & Analytics | 4 |
| 2.2.1 | Technical Competency Map: Data Engineering & Systems | 5 |
| 2.2.2 | Knowledge Tree | 5 |
| 2.3 | Technical Area: Software Engineering | 6 |
| 2.3.1 | Technical Competency Map: Software Engineering & Systems | 7 |
| 2.3.2 | Knowledge Tree | 7 |
| 2.4 | Integration and Synergy Matrix | 8 |
| 2.4.1 | Software Engineering + Artificial Intelligence: The Operational Intelligence Layer | 8 |
| 2.4.2 | Data Engineering + Artificial Intelligence: The Learning Ecosystem | 8 |
| 2.4.3 | Software Engineering + Data Engineering: The System Backbone | 9 |
| 2.4.4 | Final "Engineering Judgment" Check for Managers | 9 |
| 2.5 | Implementable Management Tools | 9 |
| 2.5.1 | Technical Audit Guide: "The Drill-Down Questions" | 9 |
| 2.5.2 | Technical Debt Radar (TDR) | 10 |
| 2.5.3 | Estimation Validation Protocol (The Reality Check) | 10 |
| 2.5.4 | Post-Mortem Audit Sheets | 10 |
| 2.5.5 | Post-Mortem Close Incident - Executive View | 11 |
| 2.5.6 | Efficiency Matrix: "The Decision Filter" | 11 |
| 2.5.7 | Implementation Suggestions | 12 |
| 2.6 | Validation Milestone: "End-to-End AI Project" | 12 |
| 2.6.1 | How to Structure It | 12 |
| 3 | Leadership Skills Framework | 13 |
| 3.1 | Introduction | 13 |
| 4 | Standard Operating Procedure | 14 |
| 4.1 | Introduction | 14 |

List of Figures

List of Tables

| | | |
|-----|---|----|
| 2.1 | AI Competency Map | 3 |
| 2.2 | Data Engineering Competency Map | 5 |
| 2.3 | Software Engineering Competency Map | 7 |
| 2.4 | Technical Debt Radar | 10 |
| 2.5 | Post-Mortem Executive Matrix | 11 |
| 2.6 | Efficiency Matrix | 12 |

Chapter 1

Introduction

This document details the comprehensive **Management Framework**, designed to provide a structured approach to technical leadership and management.

It is composed of three key pillars:

1. **Management Technical Knowledge Framework (MTKF)**: A guide to the technical competencies required for effective management, moving beyond basic literacy to engineering judgment.
2. **Leadership Skills Framework**: A structure for the soft skills and leadership qualities necessary to inspire and guide teams.
3. **Standard Operating Procedures (SOP)**: A collection of standardized processes to ensure consistency and efficiency in operations.

Together, these frameworks provide the necessary tools for a manager to bridge the gap between business strategy and technical execution.

Chapter 2

Management Technical Knowledge Framework

2.1 Introduction

2.1.1 Summary

This TKF establishes the standard of technical competency required for a manager to act as an effective bridge between business objectives and technological execution. It is rooted in the transition from "digital literacy" to "engineering judgment," enabling the autonomous assessment of risks, costs, and viability without total dependency on specialists.

2.1.2 Guiding Principles

1. **Priority of Judgment over Syntax:** The focus is on understanding how and why systems work, gaining a high-level conceptual grasp of how they are coded.
2. **Architectural Vision:** The ability to view the entire system (End-to-End) and its underlying dependencies.
3. **Trade-off Evaluation:** Every technical decision involves a compromise between cost, time, quality, and scalability.
4. **Technological Agnosticism:** A focus on fundamental principles that transcend specific vendors or programming languages.

2.2 Technical Area: Artificial Intelligence (AI) and Machine Learning

To make informed decisions about AI, a manager must understand how data is "weighted" and how computation is optimized.

- **Execution Foundation:** Mastering the concepts of Tensors and Embeddings. It is not enough to know that AI "understands" text; one must understand that it converts text into mathematical vectors within an N-dimensional space. This

realization allows the manager to grasp why data quality (cleaning) is more critical than the model itself.

- **Authority Metric:** Shifting from asking "Does it work?" to asking "What is our error rate on the validation set vs. the test set?" or "How are we managing gradient drift?"

2.2.1 Technical Competency Map: AI and Intelligent Systems

Table 2.1: AI Competency Map

| Level | AI Modeling & Logic | MLOps & Systems Infrastructure | Ethics, Risk, and ROI | Authority Criterion (The "Check") |
|--------------|--|--|---|---|
| Basic | Distinguishes between Supervised, Unsupervised, and Reinforcement Learning. | Understands that models require GPUs/TPUs and the impact of compute on TCO. | Identifies bias risks and the "AI Hierarchy of Needs" (Data before models). | Evaluates if a problem requires AI or if it is solvable via rule-based engineering. |
| Intermediate | Masters concepts of Embeddings, Tokenization, and Context Windows. | Oversees model/data versioning (DVC) and audits inference latency. | Evaluates hallucination risks and establishes technical content Guardrails. | Validates if the AI solution is technically and financially viable compared to statistical methods. |
| Advanced | Technical criteria to choose between RAG (context) and Fine-Tuning (specialization). | Implements agentic architectures and model quantization to optimize systems. | Audits Technical ROI: Balance between inference cost, latency, and business value. | Decides AI architecture by prioritizing data sovereignty and CI/CD integration. |
| Expert | Designs multi-modal Generative AI systems and hyperparameter optimization. | Leads the full MLOps stack. Optimizes "Cold Start" and agent scalability. | Establishes corporate AI Governance frameworks and advanced ethical/legal compliance. | Defines the AI-First strategy, ensuring AI is a resilient component of the ecosystem. |

2.2.2 Fundamental Knowledge Tree

Level 1: Data and Modeling Fundamentals (Basic)

- **SQL for AI:** Feature Engineering through complex queries.
- **Data Types:** Handling structured, unstructured, and semi-structured data for training.
- **Learning Paradigms:** Practical differentiation between Supervised, Unsupervised, and Reinforcement Learning.
- **Data Lifecycle:** Ingestion, cleaning, and preparation as the critical foundation of AI.

Level 2: Architecture and Orchestration (Intermediate)

- **LLM Concepts:** Management of Tokens, Context Windows, and hallucination mitigation.
- **Embeddings and Vectors:** Transforming concepts into mathematical vectors and storing them in Vector Databases.
- **Pipeline Orchestration:** Using Airflow or Dagster to automate data flows to models.
- **Initial MLOps:** Versioning models and data using tools like DVC (Data Version Control).

Level 3: AI Systems Engineering (Advanced)

- **Optimization Strategies:** Technical choice between RAG (Retrieval-Augmented Generation) for dynamic context vs. Fine-Tuning for specialization.
- **Inference and Deployment:** Optimization of Inference Latency, model quantization, and orchestration with agents (e.g., LangChain/Haystack).
- **Governance and Security:** Implementation of "Guardrails," prompt injection mitigation, and privacy compliance (Data Residency).
- **System Monitoring:** Detecting Model Drift and Data Drift in production environments.

2.3 Technical Area: Data & Analytics

A manager with technical judgment does not simply ask for a "dashboard"; they demand a data architecture that does not collapse under load.

- **Execution Foundation:** Understanding Data Modeling (Dimensional vs. Relational). A manager must know when an ACID database is required (for financial transactions) versus a BASE system (for massive scalability). This knowledge prevents the approval of technologies that could corrupt business integrity.
- **Authority Metric:** Identifying bottlenecks in pipeline latency. If data takes 4 hours to process, the manager must be able to determine if the cause is the transformation logic (Compute-bound) or the network/storage (I/O-bound).

2.3.1 Technical Competency Map: Data Engineering & Systems

Table 2.2: Data Engineering Competency Map

| Level | Modeling Logic & Structure | Pipeline Engineering & Flow | Systems Security & Governance | Authority Criterion (The "Check") |
|--------------|---|---|---|---|
| Basic | Understands normalization (1NF to 3NF) and the difference between SQL and NoSQL. | Identifies ETL/ELT phases and Batch processing concepts. | Familiar with encryption, masking, and Privacy-by-Design principles. | Detects if a data structure compromises software integrity or scalability. |
| Intermediate | Masters dimensional modeling and columnar storage (Parquet/Avro). | Audits DAGs in Airflow/Dagster. Manages retries and engineering dependencies. | Defines RBAC policies and audits access logs and platform security. | Identifies if a failure is Compute-bound (logic) or I/O-bound (infra/network). |
| Advanced | Chooses between Warehouse and Lakehouse based on the CAP Theorem and consistency needs. | Implements CDC (Change Data Capture) and Stream processing (Kafka/Flink). | Implements Data Lineage and observability for error traceability. | Refutes architectures that do not support horizontal scaling or real-time AI training. |
| Expert | Leads Data Mesh design. Masters distributed systems and eventual consistency. | Orchestrates multi-cloud architectures and optimizes tiered storage costs. | Establishes data sovereignty and global compliance (GDPR/HIPAA) at the systems level. | Defines the Data-as-a-Product strategy, ensuring data is ready for AI and software consumption. |

2.3.2 Knowledge Tree

Level 1: Structure and Query Fundamentals (Basic)

- **Advanced SQL:** Mastery of Window Functions and CTEs for complex data analysis.
- **Data Modeling:** Differentiation between Star Schema and Snowflake Schema.
- **Normalization & Types:** Understanding 1NF through 3NF and the difference between Relational (Postgres) and NoSQL (Mongo/Cassandra) databases.
- **Lifecycle:** Identification of ETL phases and Batch Processing concepts.

Level 2: Orchestration and Real-Time Flows (Intermediate)

- **Systems Orchestration:** Auditing flows in Airflow or Dagster, including retry handling and dependency management.
- **Replication Engineering:** Implementation of Change Data Capture (CDC) for real-time synchronization.
- **Efficient Storage:** Understanding columnar storage (Parquet/Avro) for massive query optimization.

- **Systems Diagnosis:** Ability to identify if a delay is Compute-bound (transformation) or I/O-bound (network/disk).

Level 3: Distributed Architecture and Governance (Advanced)

- **High Availability Systems:** Choosing between Data Warehouse and Data Lakehouse based on eventual consistency and the CAP Theorem.
- **Data Security:** Implementation of encryption at rest and in transit, alongside granular access control (RBAC).
- **Lineage and Quality:** Using Data Lineage to track the origin of failures and ensure data observability.
- **Platform Governance:** Designing Data Mesh architectures and ensuring global regulatory compliance (GDPR/HIPAA) at the infrastructure level.

2.4 Technical Area: Software Engineering

To oversee engineering, a manager must understand architectural constraints and the actual lifecycle of the code.

- **Execution Foundation:** Understanding the CAP Theorem (Consistency, Availability, Partition Tolerance). A manager must know that you cannot have all three simultaneously in a distributed system. This underpins decisions on whether a system should be "always available" or "always accurate."
- **Authority Metric:** Evaluation of Cyclomatic Complexity, System Health, and Code Coverage. If the team delivers quickly but test coverage is at 20%, the manager must be able to understand and evaluate issues of latency, resource consumption, and Disaster Recovery (DR) capability. Furthermore, the manager must recognize that they are accumulating a technical debt "time bomb."

2.4.1 Technical Competency Map: Software Engineering & Systems

Table 2.3: Software Engineering Competency Map

| Level | Architecture & Systems Design | Lifecycle & DevOps (CI/CD) | Proactive Quality & Observability | Authority Criterion (The "Check") |
|--------------|---|---|--|--|
| Basic | Distinguishes Monoliths from Microservices and the Client-Server model. | Familiar with Git flows and containers (Docker). Understands minimum infrastructure. | Reads logs and differentiates between application errors (400s) and server/system errors (500s). | Identifies if a team follows engineering standards or is generating technical debt. |
| Intermediate | Understands Asynchronous (Queues/Events) vs. Synchronous (REST/gRPC) communication. | Manages CI/CD pipelines and Infrastructure as Code (Terraform/IaC). | Audits test coverage and cyclomatic complexity to prevent systemic failures. | Challenges designs that lack exception handling, retries, or redundancy. |
| Advanced | Designs for High Availability. Applies resilience patterns (Circuit Breaker, Saga). | Implements progressive deployments (Blue-Green/Canary) and masters orchestration (K8s). | Establishes SLOs/SLIs. Uses advanced telemetry (Traces/Metrics) to prevent outages. | Decides when to pause the commercial roadmap to prioritize system stability and health. |
| Expert | Leads transformations to Event-Driven Architectures and Serverless. | Optimizes Cloud TCO through efficient architecture, not just by billing costs. | Establishes a culture of blameless Post-mortems and continuous resilience improvement. | Acts as the final authority during crises, determining the critical recovery path for the ecosystem. |

2.4.2 Knowledge Tree

Level 1: Engineering Fundamentals (Basic)

- **Design Patterns:** Singleton, Factory, Observer, and their impact on maintainability.
- **Asynchronous Programming:** Handling threads, promises, and concurrency for system efficiency.
- **Networking Fundamentals:** Protocols (HTTP/S, TCP/IP), DNS, and basic load balancing.
- **Advanced Versioning:** Branching strategies (Trunk-based development) and Pull Request policies.

Level 2: Architecture and Resilience (Intermediate)

- **Stability Patterns:** Circuit Breakers, Retries, and Bulkheads to avoid cascading failures.

- **Infrastructure as Code (IaC)**: Concepts of Terraform or CloudFormation to manage "systems" via code.
- **Caching Strategies**: Implementation of Redis/Memcached and invalidation policies.
- **System-to-System Communication**: Service Mesh (Istio), message queues (RabbitMQ/Kafka), and APIs (REST vs. gRPC).

Level 3: Platform Engineering and Operations (Advanced)

- **Distributed Architectures**: Microservices vs. Modular Monolith and state management.
- **Total Observability**: Implementing the three pillars (Logs, Metrics, Traces) and managing SLIs/SLOs.
- **Systems Security**: DevSecOps, secret management, and vulnerability scanning within the pipeline.
- **CAP Theorem**: Strategic decision-making between Consistency, Availability, and Partition Tolerance.

2.5 Integration and Synergy Matrix

A manager must perceive how these areas converge:

2.5.1 Software Engineering + Artificial Intelligence: The Operational Intelligence Layer

It is not just about "calling an API"; it is about integrating AI resiliently into the system's lifecycle.

- **CI/CD Integration**: How do we automate model deployment and validation (Model A/B Testing) within our existing software pipelines?
- **Resilience Patterns**: If the AI model fails or exceeds the response time (timeout), does the software system have a Circuit Breaker or a fallback response to avoid degrading the user experience?
- **AI Observability**: Are we monitoring AI inference latency with the same rigor as we monitor software microservices?

2.5.2 Data Engineering + Artificial Intelligence: The Learning Ecosystem

The model is a direct reflection of the data infrastructure that feeds it.

- **Training Support**: Does our data architecture (Data Lakehouse/Stream Processing) allow for real-time re-training or context retrieval (RAG) without saturating transactional systems?

- **Lineage and Integrity:** Can we trace exactly which data points produced a specific AI prediction to audit for bias or logic errors?
- **Ingestion Data Quality:** Are there validation filters in the data pipelines to prevent "garbage data" from reaching the model and corrupting the inference?

2.5.3 Software Engineering + Data Engineering: The System Backbone

The infrastructure that allows information to flow toward the user and other services.

- **Proportional Scalability:** Do our data API and persistence layer scale elastically alongside software application traffic to avoid bottlenecks?
- **Synchronization and Consistency:** Given data duplication in microservices, how does software engineering ensure there are no critical inconsistencies between the software and the data warehouse?
- **Integral Security:** Are the same security policies and secret management (IaC) applied to both database access and software deployment?

2.5.4 Final "Engineering Judgment" Check for Managers

If we update our data schema tomorrow, will it break our AI's inference, and does our Software CI/CD pipeline have the observability to catch it before the user does?

2.6 Implementable Management Tools

2.6.1 Technical Audit Guide: "The Drill-Down Questions"

Use these questions when a solution seems "too simple" or a project is stalled to expose structural weaknesses.

In Architecture Reviews (Software & Cloud)

- **On Coupling:** "If we shut down service X for maintenance, exactly which parts of system Y will stop working, and how will they handle the lack of response?"
- **On Scalability:** "What is our current breaking point in terms of concurrent users, and which resource (CPU, Memory, I/O) will be exhausted first?"
- **On Security:** "How are we managing secrets (API keys, passwords) in the code, and what happens if a developer loses their access?"

In Data & AI Projects

- **On Quality:** "What percentage of our input data is null or inconsistent, and how does that affect the standard deviation of our results?"
- **On AI/GenAI:** "How are we mitigating prompt injection risks, and what is our Human-in-the-loop mechanism for validating model outputs?"

- **On Storage:** "Why did we choose this specific database engine for this access pattern? Is it optimized for reads or writes?"

2.6.2 Technical Debt Radar (TDR)

A visual tool to quantify the risk that the technical team might be "hiding" under the rug of speed.

Table 2.4: Technical Debt Radar

| Risk Category | Status (R/A/G) | Business Impact | Required Action |
|-----------------|----------------|---|---|
| Obsolescence | Red | Security risk due to deprecated library versions. | Immediate migration plan. |
| Lack of Testing | Yellow | Increase in bugs with every new deployment. | Freeze features; raise coverage to 80%. |
| Documentation | Red | Total dependency on specific individuals (Key person risk). | Mandatory architecture docs this week. |
| Infrastructure | Green | Optimized costs and fluid scalability. | Maintain preventive monitoring. |

2.6.3 Estimation Validation Protocol (The Reality Check)

Technical teams are often either too optimistic or overly cautious. Use the 3-Layer Estimation Method to audit a schedule:

1. **Integration Layer:** "How much of this time is pure development vs. integration with external systems we don't control?"
2. **Testing/QA Layer:** "Are we including bug-fixing time after the first QA round, or just the 'Happy Path' time?"
3. **Deployment Layer:** "How automated is the deployment? If it's manual, add a 20% buffer for human error."

2.6.4 Post-Mortem Audit Sheets

AI Incident Audit

- **Failure Identification:** Was it a hallucination, data bias, or a compute infrastructure error?
- **Input Data Analysis:** What is the difference (vector distance) between current production data and the original training set?
- **Output Validation:** Did the content "Guardrails" work? Why did they allow the erroneous output?
- **Inference Efficiency:** Was the incident caused by excessive Inference Latency or an agent orchestration error?

Data & Analytics Incident Audit

- **Pipeline Breaking Point:** At which node of the Directed Acyclic Graph (DAG) did the load fail (Ingest, Transform, or Load)?
- **Bottleneck Nature:** Did the system fail due to lack of compute (Compute-bound) or transfer network limitations (I/O-bound)?
- **Integrity & Consistency:** Were ACID properties violated during a transaction, or was there an eventual consistency failure?
- **Access Audit:** Was there a failure in RBAC policies that allowed data corruption?

Software Development Incident Audit

- **Error Isolation:** Did the failure originate in a specific microservice or was it a cascading failure due to tight coupling?
- **System Resilience:** Did the Circuit Breaker pattern trigger, or did the system exhaust resources on failed connections?
- **Observability:** Why was the error not visible on dashboards before the client reported it? (Missing logs or traces).
- **Deployment Governance:** Did the incident occur after a manual deployment or a failure in the automated CI/CD pipeline?

2.6.5 Post-Mortem Close Incident - Executive View

Complete this matrix after every incident to ensure objective decision-making.

Table 2.5: Post-Mortem Executive Matrix

| Evaluation Criterion | Technical Team Response | Manager Validation (Authority Check) |
|----------------------|-----------------------------------|---|
| Root Cause | Explanation of technical failure. | Was the systemic "Why" identified, or just the symptom? |
| Reversibility | How long did the Rollback take? | Could the change be undone in under 5 minutes? |
| Detection | How did we find out? | Was observability proactive or reactive? |
| Cost of Failure | Impact estimation. | Was the cost of prevention lower than the current impact? |

2.6.6 Efficiency Matrix: "The Decision Filter"

Use this filter before approving any major technical change.

Table 2.6: Efficiency Matrix

| Filter | Control Question | If the answer is NO... |
|---------------|--|---|
| Reversibility | Can we undo this in under 5 mins if it fails? | Demand a documented Rollback plan. |
| Observability | Will we see the failure before the client calls? | The ticket is not "Done". |
| Cost-Benefit | Is compute/storage cost proportional to value? | Request data architecture optimization. |

2.6.7 Implementation Suggestions

- **The Daily Sync:** Use one random question from the Audit Guide each week to maintain high technical standards.
- **The Steering Committee:** Present the Technical Debt Radar to justify infrastructure investment over new features.
- **Quarterly Planning:** Use the Reality Check Protocol to commit to dates that can actually be met, protecting your reputation.

2.7 Validation Milestone: "End-to-End AI Project"

This is the ultimate milestone to validate the knowledge acquired through this framework.

2.7.1 How to Structure It

- **The Problem:** The manager must identify a real-world problem (either internal or client-facing).
- **The Solution:** They must technically justify the selection of tools across Software Engineering, Data, and AI.
- **Evaluation Criteria:**
 - **Trade-off Analysis:** The manager's ability to explain the compromises made (e.g., why they chose RAG over Fine-tuning, or why they selected a specific database engine).
 - **Holistic Design:** The ability to design and understand a solution that effectively integrates all three required domains.

Chapter 3

Leadership Skills Framework

3.1 Introduction

This framework outlines the essential leadership skills required for effective management.

Chapter 4

Standard Operating Procedure

4.1 Introduction

This section details the standard operating procedures for the organization.

Bibliography