#### **lemonvolt**



McKinsey's View and Enterprise Reality

# -EMONVOLT

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#### **Executive Summary**

The enterprise AI landscape in 2025 is defined by a profound asymmetry: adoption is ubiquitous, but maturity is scarce. According to McKinsey & Company's State of AI 2025 report, nearly 8 in 10 companies have integrated AI into at least one business function, and 7 in 10 have experimented with generative AI. Yet fewer than 1% of executives characterize their deployments as "mature," and more than 80% report no measurable enterprise-wide EBIT impact.

This divergence—between widespread activity and shallow outcomes—is the Al gap. It signals that technology adoption alone is insufficient to unlock structural value.

#### **Three Levers Define the Divide**

#### **Leadership** Ownership

Companies
where AI strategy
is owned at the
CEO or board
level scale faster
and report
stronger EBIT
outcomes. Only
28% of surveyed
organizations
have such
governance in
place.

#### Workflow Redesign

Embedding Al
into existing
processes yields
incremental
gains;
redesigning
those workflows
around Al
unlocks nonlinear value. Just
21% of firms
have taken this
step.

#### Scaling Discipline

of organizations
track AI KPIs,
and only onethird follow
structured
adoption
roadmaps. This
is the single
biggest barrier
between
experimentation
and measurable
enterprise
outcomes.

#### **Source Authority Overview**

Understanding the scale and credibility of the underlying research is critical to interpreting the Al gap accurately.

The State of Al 2025 report by McKinsey & Company is not an isolated pulse survey — it is one of the largest

The report is based on a global survey of 1,491 respondents spanning 101 countries, weighted by GDP contribution to ensure proportional representation of

structured analyses of enterprise AI adoption to date.



# KEY FINDING ANALYSIS

#### Finding 1

Among all organizational attributes analyzed, CEO or board-level ownership of Al governance shows the strongest correlation with EBIT impact. Yet only 28% of companies currently place Al oversight at this level. The majority delegate responsibility to IT or digital functions, which correlates with significantly weaker financial performance

#### **Why It Matters**

- Board-level ownership allows AI to be integrated into capital allocation, risk appetite, and corporate strategy, rather than operating as a sidecar initiative.
- Scaling AI requires coordination across product, data, security, compliance, and go-to-market functions. This level of orchestration is rarely achievable without senior leadership backing.
- Executive ownership communicates seriousness to regulators, partners, investors, and top-tier talent.

#### **Second Order Effects**

- Organizations with low leadership ownership risk fragmented tooling, conflicting priorities, and underfunded governance — a recipe for value dilution.
- Conversely, early executive sponsorship accelerates ecosystem formation — partnerships, capital allocation, and compliance alignment — which compounds over time.
- This is why larger enterprises are already widening their lead: their scale allows for board-level AI strategy as an enterprise architecture, not a technology project.

#### Finding 2

Fewer than 20% of organizations track KPIs for Al initiatives, and only one-third maintain structured adoption roadmaps. The majority of Al projects lack formal scaling frameworks. This "missing middle" between pilots and platforms is where value is leaking most rapidly.

#### **Why It Matters**

- KPI tracking translates localized experimentation into enterprise learning loops.
- Roadmaps enable reuse of infrastructure and governance across functions, lowering marginal costs per deployment.
- Structured adoption ensures alignment between technical capability and operational readiness.

#### **Second Order Effects**

- Firms without scaling discipline will face "pilot fatigue" rising costs, stagnant ROI, and talent attrition.
- Scaling maturity compounds: once an organization operationalizes one AI workflow successfully, subsequent deployments become exponentially cheaper and faster.
- This explains why market leaders are pulling away even when technology access is democratized.

#### Finding 3

Nearly half of organizations surveyed report negative consequences from AI use — including inaccuracy, cybersecurity incidents, and IP breaches. Only 27% review all AI outputs, while an equal proportion review almost none

#### Why It Matters

- As Al moves from experimentation to production, the probability and cost of failure increase.
- Jurisdictions including the EU and U.S. are tightening AI governance requirements, particularly for explainability and privacy.
- Firms that can demonstrate oversight will hold a commercial advantage in competitive procurement.

#### **Second Order Effects**

- Lack of robust risk governance doesn't just create operational exposure – it limits addressable market.
- Firms with mature TRiSM frameworks can command trust premiums in regulated industries, effectively converting governance into a commercial asset.
- Regulatory tailwinds will turn weak oversight from a "technical debt" into a market exclusion mechanism.

#### **Strategic Inflection**

In the next 24-36 months:

**Experimenters will plateau**—trapped in pilots, efficiency loops, and mounting risk.

**Orchestrators will compound**—translating early infrastructure and governance investments into sustained performance gains.

The strategic agenda for leadership is therefore shifting from "Should we adopt AI?" to "How do we operationalize it at scale: securely, profitably, and defensibly?

#### Imperatives for Leadership

Treat AI as a board-level agenda, not just IT initiative

Re-architect value chains rather than bolting AI onto legacy steps

Direct capital toward governance, data infrastructure, and workforce readiness

Make TRiSM (Trust, Risk, and Security Management) visible, measurable, and contractual

Build internal capability and redesign workforce strategy around augmentation, not substitution

# BUSINESS IMPACT ASSESSMENT

## Short-Term Implications (6–12 Months)

#### **Adoption without Integration**

Al adoption will continue to accelerate across marketing and sales, customer operations, and software development — domains where copilots, workflow automation, and content generation can be applied most quickly. McKinsey reports that while 78% of firms now use Al in at least one function, over 80% see no enterprise-level profit impact. This dynamic is reinforced by Stanford University's Al Index echoes this gap: model performance is rising and costs are falling, yet organizational absorption still trails technical capability by 24–36 months.

#### **Systemic Risk**

Many business units are experiencing efficiency lifts: marketing teams cutting campaign cycle times, software teams accelerating development with Al-assisted code, service teams using chatbots for first-contact deflection. Yet at the enterprise level, the lack of scaling discipline and workflow redesign fragments value. Local optimization is not translating into structural transformation. Meanwhile, nearly half of surveyed organizations report negative consequences from Al use — including inaccuracies, security lapses, and IP risks

# Mid-Term Implications (1-2 Years)

#### **Regulatory and Procurement Pressure Rises**

By 2027, Gartner forecasts that half of enterprise procurement processes will include TRISM (Trust, Risk & Security Management) requirements. This means companies without risk maturity will face barriers to market access — particularly in finance, healthcare, defense, and other regulated domains.

This aligns with McKinsey's finding that risk oversight is currently patchy and often delegated too low in the organizational hierarchy.

#### **Workforce Redistribution Accelerates**

McKinsey's data shows growing demand for data scientists, ML engineers, and compliance professionals, coupled with shrinking demand for transactional service roles. Stanford University projects a global shortfall of 2 million Al-proficient professionals by 2027, creating a winner-takes-most dynamic in talent markets.

Companies that delay workforce strategy will lose twice:

- Once in execution capacity.
- Again in talent costs as the labor market tightens.

#### **Market Share Shifts Begin**

Sectors like pharmaceuticals, aerospace, semiconductors, and financial services will begin regulated to capture disproportionate market share through operational speed, regulatory advantages, and innovation cycles. Service-heavy workflow industries without redesign will see margin compression as they compete on cost rather than differentiated capability.

## Long Term Implications (3-5 Years)

#### **Industry Boundaries Begin to Blur**

When decision velocity increases and cost structures fall, industry boundaries start to shift:

- Pharma firms can operate like tech companies with rapid product iteration and shorter go-to-market cycles.
- Semiconductor design cycles collapse, allowing new entrants to compete at lower scale thresholds.
- Aerospace firms leverage simulation for compliance, shortening traditionally fixed certification timelines.

#### Al Readiness Becomes a Prerequisite for Survival

By this stage, Al maturity will be a baseline requirement for participation in many industries.

- Firms without board-level AI governance will struggle to comply with regulatory standards.
- Organizations lacking telemetry, explainability, and TRiSM frameworks will be excluded from key procurement ecosystems.
- Companies that failed to reskill early will face irreversible talent gaps.

#### **Cost Structures Are Rewritten**

- Surrogate models in R&D will compress simulation cycles from days to seconds.
- Agentic AI systems will automate high-frequency decisionmaking in logistics and manufacturing.
- Service-heavy industries face disintermediation or commoditization if they fail to evolve beyond basic automation.

# IMPLEMENTATION REQUIREMENTS

#### **Technology Requirements**

#### **Data Infrastructure**

The first priority is a unified, high-quality data backbone. Platforms such as Snowflake, Databricks Lakehouse, Google BigQuery, or AWS Redshift allow data consolidation across silos, ensuring that Al systems have consistent access to accurate information.

Real-time data ingestion tools like **Kafka** or **Fivetran** help transform batch processes into live intelligence flows, enabling faster and context-aware decision making.

To maintain trust and compliance, enterprises should integrate governance suites such as **Collibra, Alation, or Azure Purview** for lineage tracking, metadata control, and regulatory reporting.

#### **Model Lifecycle Management**

McKinsey's top performers operationalize model development through standardized MLOps or ModelOps frameworks. Tools such as **MLflow, Kubeflow, Vertex AI, or SageMaker** provide versioning, deployment pipelines, and rollback capabilities that eliminate the ad-hoc nature of experimentation.

Monitoring and telemetry layers—using platforms like Evidently AI or WhyLabs—are essential to detect performance drift and maintain reliability once models are in production.

#### Trust, Risk, and Security Management (TRISM)

Bias-detection libraries such as **IBM AI Fairness 360**, explainability tools like **SHAP** or **LIME**, and identity-governance systems such as Okta or **Azure AD IAM** help build transparent, auditable systems.

By 2027, Gartner projects that half of all enterprise buyers will demand proof of such controls in procurement—making TRiSM maturity a commercial necessity.

#### Organizational Redesign

#### **Establish a Chief Al Officer (CAIO)**

The CAIO or equivalent executive function should be established to unify strategy, risk, and technical governance under one mandate, directly reporting to the CEO or board. This single line of accountability would replace fragmented ownership between IT, data, and business units.

#### **Form Cross-Functional Al Pods**

Instead of central data science teams working in isolation, leading firms create pods composed of a data scientist, ML engineer, process owner, and change lead.

Each pod owns a measurable business outcome—such as reducing claim-processing time or improving supply-chain forecast accuracy—ensuring that every Al initiative ties back to EBIT

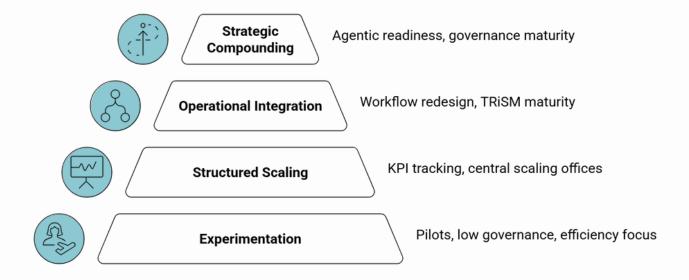
#### Institutionalize a Scaling Office

McKinsey's data shows that fewer than 20 percent of companies track Al KPIs systematically. A dedicated Scaling Office enforces adoption roadmaps, monitors performance metrics, and standardizes governance across departments. This office becomes the nerve center connecting

experimentation with enterprise execution

#### **Maturity Assessment**

Based on McKinsey's maturity indicators and cross-validation with Gartner, Deloitte, and Stanford Al Index findings, industries can be positioned along a four-step maturity ladder:



- At the first stage, Experimentation, Al is used mainly for task automation—isolated pilots that deliver short-term efficiency but little strategic value.
- In the second stage, organizations begin building reusable
   systems and performance metrics, yet impact remains localized.
- By the third stage, **AI becomes embedded within redesigned workflows** and governed by formal TRiSM frameworks; value creation becomes repeatable.
- Finally, at stage four, AI maturity aligns with business design itself—governance, data, and talent operate as one system, producing compounding returns over time.

The distance between each stage is less about technology and more about management discipline—the ability to scale trust, not just tools.

# **STRATEGIC RECOMMENDATIONS**

# Immediate Priorities (Next 90 Days)

#### 1 Elevate Al Governance to the Board

- Assign formal accountability to the CEO or a designated Chief
   Al Officer who reports directly to the board.
- Set up a standing "Al Governance Council" that includes leaders from risk, finance, technology, and HR. This body defines enterprise principles, prioritizes use cases, and approves budgets aligned with business strategy.

#### Define the Enterprise Al Mandate

Publish a concise, organization-wide statement that clarifies why AI exists in your context — whether to improve efficiency, open new markets, or de-risk operations.

#### 3 Identify Measurable KPIs

Catalogue where AI is already in use, outline model-approval procedures, and define how fairness, bias, and explainability are handled. This early visibility prepares the ground for full TRiSM (Trust, Risk, and Security Management) maturity later.

#### Communicate Internally and Train Leadership

Run concise executive sessions explaining Al's operating principles and governance model. When leaders use the same vocabulary, adoption friction drops.

# Strategic Initiatives (6–12 Months)

#### 1 Redesign One Core Workflow End-to-End

- Choose a high-volume, high-impact process for example, underwriting, customer onboarding, or software QA.
- Strip away redundant steps and rebuild the workflow with Al embedded natively.
- Document performance baselines and compare pre- and post-Al metrics.

#### 2 Establish a Scaling Office

- Create a cross-functional unit under the Chief Al Officer to manage adoption velocity.
- Responsibilities include maintaining the enterprise Al roadmap, enforcing KPI tracking, standardizing tooling, and ensuring reuse of data assets.

The Scaling Office becomes the nerve center that connects technical teams with business outcomes.

#### **3 Begin Workforce Transition**

- Launch structured reskilling programs for engineers, analysts, and business users to learn Al-assisted workflows.
- Introduce hybrid roles AI product manager, workflow architect, compliance translator — to ensure every project has both technical and operational stewardship.

# Long-Term Institutionalization (1–3 Years)

#### 1 Institutionalize Traceability

- Develop a full "Al bill of materials" a ledger capturing data sources, model versions, decision rights, and ownership for each deployed system.
- Integrate this ledger with ModelOps and compliance dashboards.

#### Invest in Agentic Readiness

 Prepare for semi-autonomous decision systems by upgrading to event-driven architectures, introducing policy-aware APIs, and simulating agentic workflows in controlled environments.

This foresight prevents future retrofits when autonomous decisioning becomes mainstream around 2028.

#### Reshape Workforce Architecture

 Design job frameworks that specify how humans and machines collaborate. Instead of replacing headcount, define augmentation ratios — such as "one analyst to four Al models."

Embed these ratios into HR planning and performance reviews to normalize AI use as a productivity lever.

#### Measurement and Success Signals

#### Within 90 Days:

- Al governance established at board level.
- Organization-wide AI mandate published.
- · Initial KPIs defined and tracked.
- Guardrail framework visible across business units.

#### Within 12 Months:

- One core workflow redesigned end-to-end.
- Scaling Office fully operational.
- TRiSM implemented in at least two major use cases.
- Reskilling program covering ten percent of workforce.
- Early EBIT impact measurable in P&L.

#### Within 3 Years:

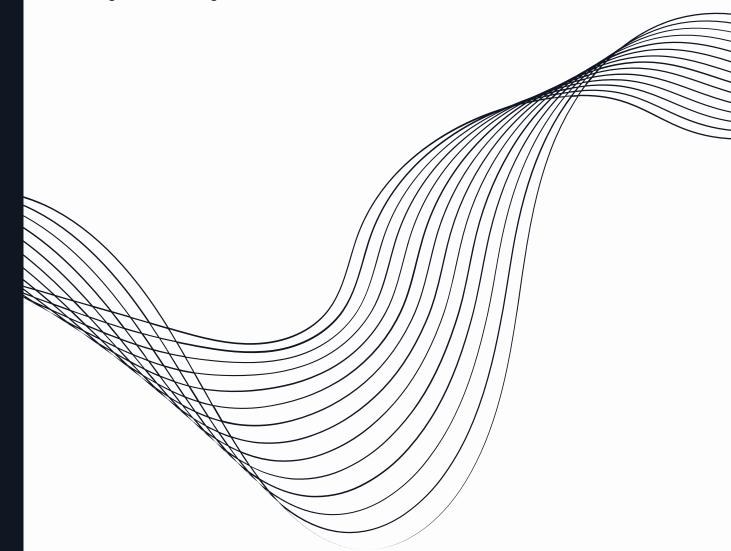
- Traceability and telemetry systems integrated with financial reporting.
- Agentic pilots running safely in production.
- Workforce architecture standardized around human-Al collaboration.
- Al governance codified as part of annual corporate disclosures.

#### **Conclusion**

Every decade brings a new tool that promises to change how companies work. Most do—but only for the few that know how to use them well.

Al is no different. The technology is powerful, but it doesn't fix weak judgment, scattered processes, or poor incentives. What McKinsey's numbers really show is that success still depends on the same old fundamentals: clear accountability, disciplined execution, and patience.

The firms turning AI into real returns aren't the ones chasing every new model. They're the ones wiring it quietly into decisions, measuring results, and learning as they go. In time, that steady discipline—not early adoption—will decide who actually gains from the age of intelligence.



The following report is an asset of



#### **About LemonVolt**

Lemonvolt is reimagining recruitment by replacing outdated hiring systems with autonomous AI agents that run talent acquisition end-to-end. Our Agentic Hiring Platform gives businesses the efficiency and scale of a large recruiting team without the overhead. We empower HR and leadership teams with next-gen tools that cut cost, time-to-hire, and complexity, helping them compete with greater agility in fast-moving tech and IT markets. Anchored in innovation and a Silicon Valley–inspired approach, Lemonvolt is driving the future of hiring with intelligent, transformative technology.

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