

# **“HARNESSING ARTIFICIAL INTELLIGENCE: BALANCING OPPORTUNITY WITH ETHICAL AND REGULATORY RESPONSIBILITY”**

*Research Paper*

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## **“Abstract”**

*Artificial intelligence (AI) is poised to deliver transformative economic and societal benefits, with projections suggesting it could contribute up to \$15.7 trillion to the global economy by 2030. Generative AI alone is expected to create trillions of dollars in annual value across industries. Yet these opportunities are accompanied by profound ethical, regulatory, and governance challenges, including bias, privacy risks, market concentration, and environmental impacts. This paper examines the opportunities and risks of AI, the mechanisms for mitigating those risks through legislation, regulation, and organisational governance, and the importance of embedding ethical principles in AI’s design and deployment. It concludes with an assessment of recent developments and forward-looking predictions for the next one to five years, underscoring the urgent need for a coordinated global response that balances innovation with responsibility.*

*Keywords: Artificial Intelligence, AI Governance, Ethical AI, AI Regulation, Global Standards, Digital Empires, Responsible Innovation.*

# 1 Artificial Intelligence: The Opportunity and the Risks

Artificial intelligence (AI) stands at the forefront of a technological transformation with few historical parallels. Analysts predict that its economic and societal impact could exceed that of previous industrial revolutions, reshaping how organisations operate, innovate, and compete. Estimates suggest that AI could contribute as much as \$15.7 trillion to the global economy by 2030, representing a 14 percent uplift in global GDP over a baseline scenario without AI (PwC, 2017). Generative AI alone could create \$2.6 to \$4.4 trillion annually across diverse sectors (McKinsey & Company, 2023). These advancements are attributable to both improvements in productivity and the introduction of novel products, services, and business models facilitated by artificial intelligence.

The opportunities are broad and transformative. AI applications are already streamlining healthcare diagnostics, accelerating drug discovery, enabling personalised education, and enhancing customer engagement in retail. In finance, AI is improving fraud detection and risk assessment, while in manufacturing it is optimising production processes, reducing waste, and predicting equipment failures before they occur. Across sectors, AI is reshaping the relationship between human expertise and machine capabilities, creating possibilities for both efficiency gains and entirely new forms of value creation.

Recent developments bring into sharp relief the pace and impact of AI. Tech giants, Microsoft, Amazon, Meta, and Alphabet, are planning to invest a staggering US \$340 billion in AI development and data centres in 2025, contributing to a projected US \$7 trillion in global AI investment over five years (Forbes, 2025, Barron's, 2025). Use of AI in at least one business function rose from 55 percent in early 2023 to 78 percent by early 2025. Generative AI usage also climbed, 71 percent now report regular use, up from 65 percent in early 2024 (McKinsey & Company, 2025). In July 2025, Nvidia became the first publicly traded company to reach a market capitalisation of \$4 trillion, underscoring investor confidence in its central role in powering the AI-driven expansion of global data-centre and semiconductor infrastructure (Stanford HAI, 2025; AP News, 2025).

However, training large AI models generates a vast carbon footprint. For example, GPT-3's training emitted 552 metric tonnes of CO<sub>2</sub> (about 123 average cars' lifetime emissions), while data centres are projected to drive electricity use sharply higher. According to the 2025 AI Index Report by Stanford HAI, training compute doubles approximately every five months, dataset sizes every eight months, and power use increases on an annual basis (Stanford HAI, 2025).

The economic potential extends beyond sectoral productivity. Goldman Sachs (2023) forecasts that generative AI could boost annual global GDP growth by 1.5 percentage points over the next decade, while the International Monetary Fund (2025) projects an average 0.5 percent annual lift between 2025 and 2030. Such forecasts, if realised, could represent a structural shift in global economic performance, comparable to the transformative effects of electrification or mass digitisation.

However, these opportunities are inseparable from significant risks and uncertainties. AI systems, particularly those using complex machine learning models, often operate with limited transparency, creating challenges for explainability and accountability. Issues of bias, whether embedded in training data or introduced during model development, threaten to reinforce existing inequalities. The environmental footprint of AI, stemming from the substantial energy and water demands of training large-scale model, raises further questions about sustainability.

Labour market disruption is another pressing concern. While AI can augment human work, there is also potential for large-scale displacement in sectors where routine cognitive or procedural tasks are prevalent. The distribution of AI's economic benefits is unlikely to be uniform, potentially widening the gap between technology leaders and laggards, and between countries with high innovation capacity and those without.

Geopolitical dynamics further complicate the landscape. AI capabilities are increasingly seen as strategic assets, contributing to competition between the United States, China, and the European Union, each with distinct regulatory philosophies and industrial strategies. This divergence raises the risk of fragmented global standards, potentially undermining interoperability and slowing collaborative research. Moreover, the market dominance of a small number of technology companies concentrates power, data, and infrastructure in ways that may limit competition and innovation, while heightening concerns over privacy and democratic accountability.

The challenge, therefore, is not simply to accelerate AI adoption, but to ensure that its development and deployment are safe, ethical, and aligned with societal values. As the pace of innovation accelerates, the window for establishing effective safeguards narrows. Failure to address these risks in a timely and coordinated manner could erode public trust, stifle innovation, and ultimately diminish the long-term benefits AI promises to deliver.

## **2 Mitigation of Risks: Governance, Regulation, and Ethics**

As artificial intelligence technologies advance in scale, complexity, and societal impact, the question of how to mitigate their associated risks has become increasingly urgent. Regulatory frameworks, industry standards, and internal governance mechanisms must evolve in parallel with technological innovation if AI is to be deployed responsibly and sustainably.

The governance challenge is multidimensional. It encompasses technical issues such as safety, robustness, and transparency; social concerns including bias, discrimination, and privacy; and economic factors such as market concentration and labour displacement. Addressing these issues requires not only formal regulation but also a broader culture of ethical responsibility that spans both public and private sectors.

### **2.1 Legislative and regulatory frameworks**

Governments and international bodies are beginning to define structured approaches to AI oversight. The European Union's AI Act represents the most comprehensive legislative effort to date, employing a risk-based approach that imposes stricter requirements on high-risk systems while encouraging innovation in low-risk applications. The United States, by contrast, favours a sector-specific, decentralised regulatory model, while China combines industrial policy with state-centred oversight and mandatory security reviews. The United Kingdom has taken an innovation-first stance, encouraging industry-led standards within existing legal frameworks.

These differing approaches illustrate the challenge of establishing globally consistent standards. Bradford's *Digital Empires* (2023) characterises this as a struggle between three dominant regulatory philosophies: the 'Brussels effect' of the EU, which exports stringent legal norms through market power; the 'Washington approach', which prioritises innovation and market freedom; and the 'Beijing model', which embeds AI development within a framework of state

control. Each model reflects its jurisdiction's political values, institutional structures, and strategic priorities, yet their divergence risks fragmenting the global AI ecosystem.

Fragmentation not only complicates cross-border compliance but can also hinder collaborative research and the scaling of beneficial AI solutions. International coordination, perhaps analogous to the Paris Agreement on climate change, may be necessary to harmonise baseline safety, transparency, and accountability standards while preserving space for innovation.

## 2.2 Ethical governance and organisational responsibility

Legislation, however, cannot by itself guarantee responsible AI. Governance must also be embedded at the organisational level, ensuring that ethical considerations are integrated into every stage of the AI lifecycle, from data collection and model training to deployment and monitoring. Frameworks such as the NIST AI Risk Management Framework, ISO 42001, and Singapore's AI Verify offer practical mechanisms for embedding ethics and risk controls into development processes.

Yet research shows that many companies' AI ethics statements are more performative than operational. Dotan et al. (2024) highlight the risk of "ethics washing," where public declarations are not matched by concrete implementation. To avoid this, organisations should be required, or at least strongly incentivised, to disclose how ethical principles are operationalized, through measurable indicators, independent audits, and clear lines of accountability.

Ethical governance extends beyond compliance into the domain of organisational culture. Traits such as integrity, transparency, and fairness are difficult to codify yet essential for long-term trust. Leadership plays a pivotal role in modelling these values and ensuring that decision-making processes reflect them. Embedding ethics into performance metrics, incentives, and professional development can help translate abstract principles into day-to-day practices.

## 2.3 Addressing market concentration and power asymmetries

Another critical aspect of AI governance is the concentration of market power in a small number of global technology companies. Firms such as Alphabet, Amazon, Microsoft, and Meta control not only the most advanced AI models but also the vast computing infrastructure and proprietary datasets that underpin them. While their resources allow rapid innovation, their dominance can limit competition, restrict access for smaller firms, and concentrate decision-making over technologies with far-reaching societal consequences.

Policy responses might include antitrust enforcement, data portability mandates, and open innovation incentives. By ensuring a more equitable distribution of AI capabilities, such measures could broaden participation in AI development and help prevent the entrenchment of monopolistic control.

## 2.4 Balancing regulation with innovation

A recurring concern is that excessive regulation might slow innovation, while too little could allow harmful applications to proliferate unchecked. The goal, therefore, is to strike a balance between protective oversight and innovation enablement. Regulatory sandboxes, public-private partnerships, and sector-specific advisory councils offer ways to test new AI applications under controlled conditions, refining both the technology and the rules that govern it before large-scale

deployment.

Ultimately, effective risk mitigation in AI requires three interlocking components:

1. Clear, enforceable regulations aligned with democratic values and human rights.
2. Robust internal governance supported by measurable, transparent practices.
3. A culture of ethical responsibility that permeates organisational decision-making.

These elements, working together, can help ensure that AI systems are not only technically sound but also socially beneficial, protecting both individual rights and the broader public interest.

## 2.5 Case studies of governance failures in data-driven technologies

AI models are trained on vast amounts of data, which can include structured, unstructured and synthetic data. The models are subject to key concerns such as privacy risks and data breaches, algorithmic bias and unfairness, security threats (deep fakes etc), shadow AI and opacity, and regular and regulatory gaps.

A comprehensive understanding and awareness of these factors are essential for the thorough development and implementation of mitigation strategies. Failure to do so may result in significant consequences, including punitive fines, damage to brand reputation, and negative fluctuations in stock prices.

Despite their sectoral differences, the following three cases share key governance failures: lack of transparency, inadequate validation, insufficient independent oversight, and weak post-deployment monitoring. In each instance, the consequences were amplified by organisational incentives that prioritised speed, market advantage, or political influence over ethical responsibility and public safety.

Technological innovation often advances at a faster pace than governance frameworks, creating opportunities for breakthroughs but also exposing significant risks when validation, transparency, and accountability are insufficient.

### 2.5.1 Facebook

The Facebook–Cambridge Analytica scandal, which highlighted the risks of lax data governance, centred on the unauthorised harvesting of personal data from millions of Facebook users, which was then exploited for political profiling and targeted advertising. Public disclosure of the breach in 2018 triggered intense scrutiny of Facebook’s privacy practices, leading to multiple investigations, fines (including a \$5 billion penalty from the US Federal Trade Commission), and global debates on data protection, consent, and the ethical use of personal information in AI-driven analytics (Cadwalladr et al, 2018).

### 2.5.2 Theranos

Theranos’s collapse offers a cautionary example of the dangers posed when technological claims outpace scientific validation. Founded in 2003, the company claimed to have developed a device capable of running hundreds of blood tests from just a few drops of blood. In practice, the technology was inaccurate, and the majority of tests were conducted using commercially available machines rather than the company’s proprietary device (Carreyrou, 2018).

The concealment of these failings from investors, regulators, and patients resulted in significant financial and reputational damage, alongside the risk of harm to individuals receiving incorrect medical results. The parallels with high-risk AI applications are clear: both involve complex, safety-critical systems whose performance is difficult for outsiders to verify. Provisions in the EU AI Act requiring pre-market conformity assessments, post-market monitoring, and independent auditing are designed to prevent the kind of opaque, unvalidated deployment seen in Theranos (European Commission, 2024).

### 2.5.3 Boeing 737 MAX

The Boeing 737 MAX crisis demonstrates how inadequate safety validation and poor risk communication can result in catastrophic consequences. Two crashes (Indonesia in 2018 and Ethiopia in 2019) claimed 346 lives and were traced to the Maneuvering Characteristics Augmentation System (MCAS), a software system intended to stabilise the aircraft. MCAS relied on input from a single angle-of-attack sensor, without sufficient redundancy, and could repeatedly push the aircraft's nose down in response to erroneous data (Joint Authorities Technical Review, 2019).

Investigations revealed that Boeing had minimised regulatory scrutiny by presenting MCAS as a minor modification, avoiding extensive pilot training requirements. Internal communications indicated awareness of the risks but a prioritisation of cost and market deadlines over safety (US House of Representatives, 2020). In the context of AI governance, the 737 MAX case underlines the critical importance of system transparency, human oversight, and robust testing under realistic operational conditions. The emphasis on high-risk system risk management and post-deployment monitoring in the EU AI Act parallels the aviation sector's eventual regulatory reforms, reinforcing that software-driven systems, whether in aircraft or AI, require rigorous, ongoing safety assurance.

## 3 Important Recent Developments and Predictions for the Next 1–5 Years

The pace of progress in artificial intelligence is accelerating at a rate that is both remarkable and challenging for policymakers, businesses, and society to absorb. Over the past two years, advances in generative AI, multimodal models, and domain-specific AI systems have demonstrated capabilities that would have seemed aspirational only a short time ago. At the same time, these breakthroughs have intensified debates about governance, ethics, and long-term societal impact.

### 3.1 Key developments in the AI landscape

#### 3.1.1 Generative AI at scale

Since the public release of advanced large language models (LLMs) in late 2022, generative AI has expanded from text-based outputs to multimodal capabilities, integrating language, vision, audio, and increasingly complex reasoning functions (McKinsey & Company, 2023). This has enabled new applications in drug discovery, design automation, software development, and customer engagement, while also raising concerns about deepfakes, misinformation, and intellectual property rights.

### 3.1.2 Advances in multimodal AI

Recent AI systems such as GPT-5o, Google's Gemini 2.0, and Anthropic's Claude 4.0 demonstrate the ability to process and generate across multiple data types simultaneously. This allows seamless interaction between text, images, audio, and video, making AI more versatile but also expanding the potential vectors for misuse, particularly in disinformation and manipulation.

### 3.1.3 Open-source AI models

A notable shift has been the release of high-performance open-source AI models by organisations such as Meta (LLaMA 3) and Mistral AI. While open models can democratise access and accelerate innovation, they also make powerful AI capabilities available to actors with fewer constraints, increasing the potential for harmful applications (Holistic AI, 2024).

### 3.1.4 Sector-specific AI integration

Industries such as finance, healthcare, and manufacturing have moved beyond pilot projects to operational deployments. In banking, generative AI is projected to improve efficiency by nearly 46 per cent (Reserve Bank of India, 2025). In healthcare, AI-enhanced diagnostics and predictive analytics are improving patient outcomes while lowering costs. However, integration at scale is also revealing gaps in workforce readiness and regulatory clarity.

### 3.1.5 Geopolitical competition and strategic alliances

AI capabilities are increasingly viewed as strategic assets in the competition between the United States, China, and the European Union, with significant investments in AI research, cloud infrastructure, and semiconductor manufacturing (Bradford, 2023). This dynamic is leading to strategic alliances and competitive positioning that will influence both the pace of innovation and the global regulatory environment.

## 3.2 Predictions for the next 1–5 years

### 3.2.1 Economic impact and productivity gains

McKinsey & Company (2023) projects that AI could deliver \$2.6 to \$4.4 trillion annually in economic value, with adoption in at least one business function already at 78 per cent in early 2025 (McKinsey & Company, 2025). Goldman Sachs (2023) estimates a 1.5 percentage point annual boost to global GDP growth over the next decade, with the most significant near-term gains expected in knowledge-intensive industries such as finance, legal services, and scientific research.

### 3.2.2 Regulatory maturation and divergence

Within five years, the EU AI Act is likely to serve as a global reference point, with other jurisdictions adopting compatible or competing frameworks. The 'Brussels effect' may encourage convergence on certain principles, such as transparency, risk management, and rights protection, but divergence in enforcement models will persist, reflecting underlying political and economic differences (Bradford, 2023).

### 3.2.3 Shift towards hybrid human–AI workflows

In most professional sectors, AI will augment rather than replace human labour, particularly where complex judgement, empathy, and contextual reasoning are required. However, task automation will continue to disrupt administrative and routine analytical roles. The net employment effect will depend on the speed of workforce reskilling and the extent to which organisations create new roles that capitalise on human–AI collaboration (International Monetary Fund, 2025).

### 3.2.4 Heightened focus on AI safety and alignment

Growing awareness of advanced AI risks, particularly from frontier models with emergent capabilities, will drive investment in safety research, red-teaming, and model interpretability. This is likely to be coupled with more formalised global cooperation on AI safety, potentially through the expansion of initiatives such as the AI Safety Summit held in the UK in 2023.

### 3.2.5 Environmental considerations in AI development

Environmental sustainability is becoming a governance priority in AI: rising energy and water demands, especially from model training and inference, call for transparent lifecycle reporting and resource-aware AI development. Incentives for green AI techniques, such as model compression, pruning, and neuromorphic architecture, and transparency in energy/water usage are increasingly vital (Mistral, 2025; Tabbakh, 2024; Rózycki, 2025; Vogginger et al., 2024).

## 4 Conclusion

The increasing investment in artificial intelligence is taking place amidst concerns regarding the long-term return on investment of Generative AI models, the inflationary effects of substantial capital expenditures on energy grids and utility costs, and the potential exacerbation of wealth inequality. An alternative perspective suggests that the foundational Generative AI technology represents a genuinely transformative, general-purpose technology. While these significant investments may be susceptible to speculative excess, they are considered essential for constructing foundational infrastructure that will ultimately deliver substantial economic and societal benefits, albeit over the long term.

Artificial intelligence offers the potential for unprecedented economic, social, and technological advancement. Yet the realisation of these benefits is not guaranteed. Without effective governance, ethical safeguards, and international cooperation, AI could exacerbate inequalities, entrench monopolistic power, and undermine trust in both institutions and technology.

The next one to five years will be decisive. They will determine whether AI's trajectory is defined by its capacity to drive inclusive growth and human flourishing or by the failure to manage its risks responsibly. Achieving the former will require a holistic approach: robust legislation, transparent and accountable governance, organisational cultures grounded in ethical responsibility, and global collaboration that bridges competing regulatory philosophies.

The path we choose will not only shape the evolution of AI but will also influence the broader social contract between technology, business, and society. It is imperative for all stakeholders, including governments, industry leaders, researchers, and civil society, to ensure that the potential

of artificial intelligence is harnessed in a manner that is equitable, sustainable, and aligned with the values we collectively uphold.

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