



---

POLICY PAPER • 2026

# HUMAN SOVEREIGNTY IN THE AGE OF AI

*A Strategic Framework for Cognitive Resilience and Human Flourishing*

---

**By Stephen C. Jordan**

Institute for Sustainable Development • Washington, D.C. • [isdus.org](http://isdus.org)

### A NOTE TO INDUSTRY READERS

*This paper argues that artificial intelligence should be deployed in ways that increase human capability, strengthen institutional resilience, and preserve human authority over consequential decisions. While there is an obvious public interest and national security dimension to these objectives, the AI industry has a strong commercial interest in this outcome: public trust is a prerequisite for broad adoption, and trust is harder to earn than capability is to demonstrate. In order to facilitate the integration of AI into modern American life, the industry should benchmark what other industries and business titans have done to build trust with the American people in the past. Andrew Carnegie invested his wealth in the libraries, universities, and civic institutions that equipped a new generation for the industrial economy. The pharmaceuticals and airline industries have also developed important industry-wide initiatives that have contributed to their long-term compact with their customers and key stakeholders. The AI industry faces the same historical moment. The frameworks proposed here for lifelong learning, workforce transition, and community investment are not regulatory burdens; they are the modern version of Carnegie's program — and the condition for sustained public legitimacy and acceptance of the potential disruptive risks that AI may cause as it races to innovate transformative solutions for the American people.*

## EXECUTIVE SUMMARY

Artificial intelligence is already improving scientific work, operational efficiency, and decision support across the economy. It also creates governance challenges of historic scale — and obligations of historic scale. The central question is not whether AI should be adopted, but how to adopt it in ways that expand human capability rather than displace human judgment, distribute its benefits broadly rather than concentrate them narrowly, and equip society for a transition that will be more disruptive, more rapid, and more prolonged than any previous technological transformation in American history.

This paper uses the term intellectual protectionism to describe a policy and institutional posture that preserves human cognitive capacity and sovereignty in domains where AI failure could be catastrophic, irreversible, or constitutionally significant. But it goes further. The paper argues that governance frameworks alone are insufficient: the AI transition requires a social compact — a coordinated set of affirmative investments by industry, government, and civil society — that equips people to thrive in the economy AI is creating. Without that compact, the governance frameworks will be undermined by a political backlash that is already organizing.

The framework is organized around four arguments. First, the hollowing-out analogy: America's offshoring of manufacturing over the past two generations destroyed not just jobs but the ecosystems of tacit knowledge, feedback loops, and engineering depth that made industries governable and improvable. AI adoption risks the same pattern at the level of human cognition. Second, the automation complacency evidence: skill atrophy from over-reliance on automated systems is empirically documented, predictable, and potentially catastrophic in high-stakes domains. Third, the trust problem: voluntary industry self-governance is structurally insufficient, and trust must be made verifiable through institutional architecture rather than asserted through corporate commitments.

Fourth, the social compact imperative: a society transitioning to AI-mediated work at a time when medical advances are simultaneously extending productive lifespans to 80, 90, and beyond requires a fundamental reinvention of how learning, workforce transitions, and community resilience are supported — not a patch on existing programs, but a new architecture of the same ambition as the GI Bill or the Land Grant college system.

The distinction between fact-based operations and normative decision-making runs through all four arguments. AI excels at operational, fact-based labor — pattern recognition, data synthesis, classification, iteration at scale. Human beings are irreplaceable for normative labor — including the exercise of judgment, values, accountability, and the weighing of competing goods. The goal of this framework is to maximize the first while protecting the second, and to ensure that the transition from one to the other is navigated with the human reserve, the institutional depth, and the social investment that a change of this magnitude requires.

### KEY FINDINGS

Human expertise remains strategically valuable even as AI improves, because high-stakes systems need people who can understand, audit, and override automated outputs. The governance challenge is not a single problem but four — skill atrophy, AI system failure, strategic dependence and community insecurity — each requiring different remedies but sharing a common solution principle. Process intelligibility matters more than output legibility in critical domains. Voluntary industry self-governance is structurally insufficient; trust must be built through verifiable institutional architecture. The coming wave of AI-driven displacement will coincide with dramatic increases in human longevity, creating a multi-decade transition requiring lifelong learning infrastructure that does not yet exist. The risk of neo-Luddite backlash is real and is best addressed by proactive social investment, not by dismissing legitimate grievances.

### POLICY RECOMMENDATIONS

The paper proposes twelve actions across four actors. Industry should fund a Human Cognitive Advancement Initiative (\$2B annual charitable investment modeled on Carnegie's library program), develop AI psychological safety tools, and establish a nonprofit engagement arm for community-level implementation. Government should establish a Lifelong Learning Transformation Fund (\$7B, approximately 10% of the Department of Education budget, reallocated to reinvent education for 18–80-year-old learners), a Workforce Transition Program modeled on the GI Bill. A Community Disruption Fund (\$5–10B for small towns and urban areas facing acute displacement) should also be considered by either sector. Cross-sector partnerships should build AI governance and auditing standards through the NIST framework and establish a stakeholder and community engagement platform. All of this should be governed by an accountability architecture with public reporting, pre-specified adjustment triggers, and independent oversight designed to build structural trust and confidence in the industry.

This is a pro-AI position, pro-industry position, pro-constitutional position and pro-human position at the same time. It is also the only position that takes seriously both what AI could accomplish and what is required — in governance, in social investment, and in human development — to ensure that it does.

## I. THE HOLLOWING: HOW AMERICA LOST ITS INDUSTRIAL MIND

From the 1970s onward, American firms increasingly moved production abroad in search of lower costs. The assumption was that the United States could keep design, engineering, and innovation while other countries handled the physical work. That assumption proved disastrously incomplete.

The overlooked factor was tacit knowledge — the embodied, experience-based judgment that cannot be fully codified in manuals, patents, or algorithms. As philosopher Michael Polanyi observed, we know more than we can tell. When factories closed or moved abroad, the feedback loop between engineers and production floors was severed. Process and product innovation, once intertwined, drifted apart. Over time, American firms lost not only manufacturing capability but also the engineering depth needed to push the technological frontier. There is an extensive body of literature distinguishing two types of technological innovation: transformative breakthroughs that create entirely new industries, and the incremental, adjacent improvements that deepen and extend a technology's capabilities. The United States has consistently excelled at the first. But the second kind — small innovation — is equally important and often more durable, creating competitive moats that compound over time. When the underlying engineering base moves elsewhere, the capacity for small innovation follows.

The semiconductor industry provides the starkest illustration. By separating chip design from fabrication and transferring the latter to Taiwan and South Korea, U.S. companies assumed they could retain the “brains” while outsourcing the “hands.” Two decades later, manufacturing excellence proved inseparable from engineering excellence. When supply-chain disruptions hit during the COVID-19 pandemic, the United States found itself strategically dependent on foreign production of the most critical component of modern technology. The CHIPS and Science Act — committing \$52.7 billion to rebuild domestic semiconductor capacity — stands as an expensive admission that preserving ecosystems is far cheaper than reconstructing them.

***Once capability disappears, restoring it is slow, expensive, and sometimes impossible. The same lesson applies to cognition.***

Industrial ecosystems behave like biological ones: remove a keystone species, and the surrounding system degrades. The deeper lesson is that knowledge is not a static stock; it is a living practice sustained by use. Lose the practice, lose the knowledge. Lose the knowledge, lose the ability to govern and improve the systems built upon it. America now stands at risk of repeating this mistake — not in physical production, but in the cognitive domains that underpin national power. Strategic capability depends on keeping the full ecosystem of know-how alive, not just the top layer of visible output. That is the core meaning of intellectual protectionism.

## II. AUTOMATION COMPLACENCY: WHEN SKILLS ATROPHY

High-reliability industries already understand the danger of over-automation. The foundational academic literature on human-automation interaction, established by Parasuraman and Riley in their landmark 1997 paper in *Human Factors*, identifies four failure modes: use, misuse, disuse, and abuse. The most consequential for the age of AI may be misuse causing humans to disuse their own capabilities — specifically, the automation complacency that develops when systems are reliable enough that operators lose the habit of independent verification. Parasuraman and Riley found that complacency occurs even in expert practitioners and cannot be overcome through simple practice: the design of the human-machine system itself must preserve active human engagement.

On June 1, 2009, Air France Flight 447 departed Rio de Janeiro for Paris carrying 228 passengers and crew. It never arrived. The Airbus A330 crashed into the Atlantic Ocean, killing everyone aboard. The aircraft was mechanically sound. The crew was experienced. The disaster stemmed from skill atrophy induced by over-reliance on automation. When pitot tubes iced over and the autopilot disconnected, the pilots — long accustomed to highly reliable fly-by-wire systems — struggled to diagnose inconsistent airspeed readings and recover from a stall. One pilot held back-pressure on the sidestick for most of the four-minute descent, an input the others failed to override. Manual flying skills had degraded through disuse; on long-haul routes, pilots might spend fewer than ten minutes per flight in manual control. The aviation industry responded by strengthening manual-flight training requirements and emphasizing hand-flying proficiency — the direct model for the Pilot-Hours Principle this paper recommends.

A parallel example appeared in naval operations. As GPS navigation became universal, the U.S. Navy discovered that officers were losing the ability to perform celestial navigation and manual charting. Naval academies reintroduced celestial navigation training in 2015 to ensure that sailors could operate if satellite systems were compromised or destroyed — a recognition that skill preservation is a national security requirement, not a nostalgic preference.

***The more reliable automated systems become, the less frequently humans practice the skills needed to intervene when those systems fail.***

As AI systems assume more routine reasoning, drafting, classification, and synthesis work, human operators face the same dynamic across security, law, medicine, intelligence, defense, engineering, and finance. The ability to recognize an AI error can matter as much as the ability to produce a fast answer. This is why process intelligibility — understanding how a system reaches its conclusions, not merely whether its conclusions are correct — is foundational to meaningful human oversight. And it is why the social compact for lifelong learning proposed in Section VII is not a peripheral add-on to this framework but central to it: a society cannot maintain the human reserve that AI governance requires unless it continuously invests in the cognitive capacities that automation is simultaneously eroding.

### III. THE CASE FOR COGNITIVE RESILIENCE

This paper does not argue for rejecting AI but rather maximizing its value while minimizing its risks. Anthropic CEO Dario Amodei, in his 2024 essay “Machines of Loving Grace,” offers one of the most serious cases for AI’s transformative potential: a world in which AI could compress decades of biomedical progress into years, expand access to expertise that is currently scarce, and address problems that have resisted human effort for generations. The Stanford HAI AI Index Report 2025 documents the pace of this change: AI benchmark performance advanced by 49 to 67 percentage points in a single year across leading tests, global private investment in AI hit a record \$252 billion, and AI-enabled medical devices approved by the FDA have grown from 6 in 2015 to 223 by 2023. That potential is worth pursuing. The question this paper asks is whether the institutional and human foundations required to govern, correct, and steer such a technology are being built alongside it.

The governance challenge has three parts. The first is human cognitive atrophy: if people no longer practice core reasoning tasks, they may no longer be able to supervise the systems that perform them nor have the emotional defenses to cope with them. The same Stanford report documents a 56 percent rise in reported AI-related incidents in a single year — a record 233 cases, including deepfake harm events and AI systems implicated in a teenager’s death.

The second is AI system failure: models can hallucinate, behave unpredictably under edge conditions, or be manipulated through adversarial inputs. In the modern art of war, targeting supply chains, critical infrastructures, and other civil society assets — like AI — that enable warfighting capabilities, are just as much targets as the warfighting capabilities themselves. Even for civilian purposes, AI system failure could be catastrophic for health care, utility and financial services that require virtually 100% reliability.

The third is strategic dependence: when essential AI services concentrate in a small number of providers, societies become vulnerable to outages, capture, or coercive leverage. The recent Anthropic-Department of War dispute illustrates this dynamic. The Department of War could not operate with conditional AI technical support that could be withdrawn in critical moments. It works in reverse as well, because of the massive buying power of the federal government and the pressure it could put on service providers to invade privacy and use AI for surveillance and other purposes that might curtail civil liberties.

These problems share a common solution principle: preserve human sovereign capacity. In constitutional terms, that means keeping consequential authority legible, reviewable, and ultimately accountable to human decision-makers. In practical terms, it means designing systems that can be audited, backed up, and overridden. The facts-values distinction is central to understanding where this matters most. AI excels at descriptive labor — pattern recognition, data synthesis, classification, iteration at scale. Human beings remain essential in prescriptive labor — judgment, values, accountability, and decisions for which someone must answer. The boundary between these categories is not always fixed: at sufficient scale, descriptive outputs can shape prescriptive decisions, which is precisely why the boundary must be actively managed. AI should inform decisions; it should not constitute them. This is where the accounting industry’s GAAP and FASB support structure might be useful. A standards body like FASB could oversee

General AI Application Principles to establish protocols for AI uses and protections as the technology evolves.

## IV. STRATEGIC DEPTH IN HUMAN INTELLIGENCE

National resilience depends on strategic depth: the ability to absorb shocks and continue operating under degraded conditions. The United States maintains this depth in military forces, energy systems, and industrial supply chains through redundancy, reserves, and domestic capability. The NIST AI Risk Management Framework, released in 2023 and widely adopted as the U.S. government's reference standard for AI governance, explicitly identifies de-skilling and over-reliance as human-AI interaction risks that organizations must actively manage across the AI lifecycle. Intellectual protectionism is the strategic complement to that framework: where NIST focuses on the AI system itself, this paper focuses on the human reserve that must exist alongside it.

A government that relies on AI for legal interpretation, intelligence analysis, financial risk assessment, or critical policy formulation must still be able to explain, review, and override those systems. Otherwise, public authority is being exercised through tools that are not fully transparent to the public or accountable to law. Democratic self-government requires that consequential decisions remain reviewable by human institutions that can assign responsibility. AI can assist those institutions, but it should not become the hidden source of binding authority where liberty, property, due process, or public safety are at stake. This may require information security offices to not only develop process record-keeping and AI audit capabilities, but also to maintain critical skills to replace or work around AI failures.

***Maintaining strategic depth in human intelligence is a constitutional imperative, not merely a technical preference.***

This framing is compatible with innovation. It does not require banning AI from critical domains. It requires building enough human capability, documentation, and fallback capacity that AI remains a tool under law rather than a system above it. The more consequential the decision, the more important human review becomes — and the more important it is that the people doing that reviewing have not had their competence hollowed out by years of delegation to machines they no longer fully understand.

## V. THE TRUST PROBLEM AND ITS SUBSTITUTES

Public concern about AI is real and persistent. In a 2025 Pew Research Center survey, 50 percent of U.S. adults said they were more concerned than excited about AI, while only 10 percent said they were more excited than concerned. Globally, the median share expressing more concern than excitement was 34 percent. The industry is operating in an environment where trust cannot be assumed; it has to be earned.

The trust problem is structural, not merely perceptual. The leading AI developers are simultaneously the entities warning about AI risk and the entities building the systems that create it, while their founders accumulate unprecedented wealth and influence in the process. The sarcasm that greets their public safety commitments is structurally earned. A company that issues a safety framework while retaining unilateral authority to override that framework's own recommendations has not solved the trust problem; it has documented it. OpenAI's Preparedness Framework Version 2, published in April 2025, requests systematic evaluation of only 3 of the 24 risk categories identified in the academic literature, permits deployment of systems with "medium" capability for what OpenAI itself defines as severe harm (more than 1,000 deaths or \$100 billion in damages), and concentrates final authority in the CEO. An independent academic analysis found that the framework "does not guarantee any AI risk mitigation practices." This pattern is familiar: financial services, pharmaceuticals, and environmental protection all went through periods where industry self-regulation was promoted as sufficient, only for significant failures to demonstrate the need for external oversight.

***We cannot make AI leaders trustworthy by asking them to be. We can build institutional architecture that makes trustworthiness verifiable.***

There are three possible responses to the trust deficit. Authentic trust requires demonstrated willingness to absorb real costs for stated principles. When Anthropic refused to remove Claude's ethical constraints from a Pentagon contract — resulting in the Trump administration declaring Anthropic a "supply chain risk" and ordering federal agencies to stop using Claude for reasons discussed earlier — the action was credible precisely because it was costly. That is what genuine commitment looks like, and the consistency in its application helps people understand what Anthropic's moral code is with clarity even when there is ideological conflict. They are removing uncertainty from what their products will or will not do.

Structural trust requires governance architecture that does not depend on the personal integrity of founders: third-party auditing bodies with genuine authority, liability frameworks that expose companies to legal consequences for harms, and regulatory structures modeled on the FDA and FAA, where safety is an enforced standard with consequences for failure. We do not trust pharmaceutical companies because their executives have good values; we trust pharmaceutical products because they have been subjected to independent review by parties with no financial interest in the outcome.

Trust substitutes are what governance provides while authentic and structural trust are being built — which describes part of what is needed at the present moment. They include disclosure requirements, transparency about known failure modes, human-readable audit logs, mandatory red-teaming before deployment, and the AI auditor profession this paper advocates. Trust substitutes do not require trusting the companies building the systems; they create independent verification capacity.

In his 2005 book *Radical Evolution*, Joel Garreau first mapped the three camps in the technology debate: the "Heaven" techno-optimists as embodied by Ray Kurzweil, the "Hell" catastrophists as embodied by Bill Joy, and the Prevail pragmatists as embodied by Jaron Lanier. He argued that technology's trajectory depends on the institutional choices societies make while the technology

is young. Trust substitutes are those choices. The AI industry is at the same juncture that aviation, pharmaceuticals, and finance reached before they earned genuine public trust: the frameworks for building that trust exist, and the political will to require them is what is currently lacking. This is not a liberal or a conservative idea, it is a market-enabling one.

## VI. A POLICY AGENDA FOR COGNITIVE RESILIENCE

The policy response should be principle-based, not rule-based. AI changes quickly, and a static list of prohibited uses will age badly. Durable principles are better: preserve human autonomy, ensure meaningful review, require intelligibility where consequences are serious, and maintain the ability to recover when systems fail. The following agenda is organized across three streams, each targeting a distinct dimension of the governance challenge.

### **Stream One: Addressing Human Cognitive Atrophy**

A principle-based domain classification system should distinguish between contexts by consequence rather than by technology alone. Tier One includes domains where AI failure could be catastrophic, irreversible, time-critical, or constitutionally significant — nuclear command and control, biological threat assessment, systemic financial risk management, and constitutional legal interpretation. These require parallel human-run systems and routine AI-off drills. Tier Two includes consequential but recoverable domains: clinical medicine, intelligence analysis, military planning, and policy research. Tier Three, where AI substitution is broadly benign, requires monitoring but not active intervention. Special attention should be given to interdependencies and back doors through which compromises in lower-tier systems could affect higher-tier assets.

Selected professions should require periodic demonstration of AI-independent competence — the Pilot-Hours Principle. Medical licensing boards should require diagnostic reasoning without AI support as a condition of certification renewal. Intelligence analyst training should include substantial periods of unassisted analysis. Officer training programs at the service academies should devote significant curriculum to strategic reasoning without AI tools, not as a rejection of those tools but as insurance against their failure. Bar admission requirements should test legal reasoning that cannot be delegated to AI-generated briefs.

Higher education and professional training should emphasize how AI systems work, when they work, and when they do not — building process intelligibility rather than mere tool use. A new profession of AI auditors should be developed for high-stakes domains: domain experts trained to evaluate whether system outputs are trustworthy in context and whether the underlying process is fit for purpose. Industry and government should jointly fund the equivalent of a FASB standards body for General AI Application Protocols. This shared professional governance architecture is analogous to what exists in finance and medicine.

### **Stream Two: Addressing AI System Failure**

High-stakes AI should undergo adversarial testing before deployment — red-teaming, sandboxing, and evaluation under degraded or manipulated conditions. AI systems deployed in

critical domains should provide human-readable documentation of known failure modes, confidence limits, and conditions of valid use. Critical institutions should maintain non-digital or air-gapped fallback procedures for essential functions. Where AI systems connect to sensitive networks, architectures should include quarantine, segmentation, and shutdown capability to prevent cascading failures.

### **Stream Three: Addressing Strategic Dependence**

The legal framework should protect human autonomy, privacy, accountability, and review rights. Consequential AI decisions should remain subject to meaningful human oversight. Federal procurement is one of the most powerful levers available: if the government requires auditability, human override, backup capability, and documentation as conditions of purchase, the market will build more resilient systems. Publicly funded knowledge — scientific research, government datasets, legal records, military doctrine — should be treated as a strategic commons governed with reciprocity, auditability, and security protections.

## **VII. THE SOCIAL COMPACT: SHARED OBLIGATIONS FOR A JUST TRANSITION**

Governance frameworks address what institutions must do to make AI safe and accountable. They do not address the potential threats to livelihoods, communities, and life trajectories that may be disrupted by the technology's deployment. That is a separate and equally important question, and it is the one most likely to determine whether the AI transition succeeds or fails politically.

The neo-Luddite risk is real and should not be dismissed as mere technophobia. The original Luddites were not anti-technology in the abstract; they were workers experiencing the costs of industrial transformation without access to its benefits. The AI equivalent is already forming: Hollywood writers and actors won contract provisions limiting AI in production; Amazon and Google employees staged walkouts over AI contracts with military implications; creative professionals are weaponizing copyright litigation against AI training data; and advocacy organizations are mobilizing resentment over the concentration of AI's economic gains in a small number of firms and geographies. If these grievances are not addressed through proactive investment, they will be addressed through political disruption — moratoria, prohibitive regulation, or the kind of backlash that set back nuclear power for a generation.

There is also a less-discussed structural factor that makes the social compact more urgent than in any previous technological transition: the coincidence of AI disruption with the extension of human longevity. Medical advances in rejuvenation, disease prevention, and health maintenance are extending productive lifespans toward 80, 90, and beyond. A worker disrupted at 45 in the age of AI may need three or four complete professional reinventions over a 60-year remaining career. The conventional workforce transition program — designed to retrain a 52-year-old for a decade before retirement — is structurally insufficient for this reality. What is required is not a transition program but a new architecture of lifelong learning: a system that treats education not

as a phase of life that ends at 22 but as a continuous capacity that must be supported, accessible, and affordable throughout a working life that may span six decades.

***Andrew Carnegie, having built a fortune on the labor of industrial transformation, invested that wealth in the institutions that equipped a new generation for a new economy. The AI industry faces the same historical moment.***

The Carnegie parallel is instructive and morally precise. Carnegie did not build libraries out of guilt; he built them because he understood that industrial wealth was built on a workforce whose capacities needed to be continuously renewed, and that it was in his, and his industry’s interests to invest in shared human capital. The AI industry is in an analogous position. What some might perceive as marketing – AI will save labor costs in industries ranging from law to cybersecurity – can also be perceived as a technology that is displacing workers at a scale and speed that has no recent precedent. The social compact proposed here is not a tax or a penalty; it is the modern version of Carnegie’s program — and can serve as a basis for sustained public legitimacy that commercial success at scale requires.

The nonprofit sector is the critical missing link in most AI governance discussions. It is the connective tissue between national policy and community reality. Nonprofits, including faith-based organizations, are often the drivers of where trust is actually built or lost at the neighborhood level. A charitable investment that deliberately funds mediating organizations gives the social compact something that no government program can manufacture: locally legitimate institutions that can meet communities where they are, adapt programs to local conditions, and generate the bottom-up feedback that distinguishes what works from what looks good in a policy brief. This is the model that made Carnegie’s library program work: not federal administration but local institutions, nationally funded.

The following table summarizes the social compact framework across four actors. The investments are large but not implausible: the Lifelong Learning Transformation Fund represents 10 percent of the Department of Education’s current budget, reallocated rather than added. The Community Disruption Fund is comparable to a single medium-sized federal infrastructure program. The industry charitable investment, distributed across the major AI developers, represents a fraction of their annual revenue growth. The question is not whether these resources exist; it is whether the political will exists to direct them before the backlash forces a less orderly response.

| ACTOR    | INITIATIVE                             | MECHANISM & INVESTMENT  |
|----------|--|---|
| INDUSTRY | Human Cognitive Advancement Initiative | Consortium-funded, philanthropic model. Minimum \$2B annual charitable investment — the modern Carnegie program — channeled through mediating nonprofits that design, deliver, and improve community-level programs. Targets lifelong learning infrastructure for all ages. |

|                               |   |   |
|-------------------------------|---|---|
|                               | AI Psychological Safety Tools               | A Stay-Safe-Online analogue for AI: tools, curricula, and community resources grounded in developmental psychology, helping individuals and families understand and navigate AI from childhood through older age.   |
|                               | Nonprofit & Community Engagement Arm        | Dedicated funding stream for mediating organizations to build local trust, adapt programs to local conditions, and generate bottom-up feedback that improves what works.  |
| <b>GOVERNMENT OR INDUSTRY</b> | Community Disruption Fund                   | \$5–10B targeted investment in small towns and blighted urban areas facing acute AI-driven economic disruption. Supports local renewal infrastructure: retraining centers, small business transition support, civic technology access, and community anchor institutions.   |
| <b>GOVERNMENT</b>             | Workforce Transition Program                | Federal retraining benefits, portable credentials, and income support for workers displaced by AI automation. Triggered by documented displacement events, not discretionary applications.  |
| <b>CROSS-SECTOR</b>           | Lifelong Learning Transformation Initiative | \$7B annual investment — equivalent to approximately 10% of the existing Department of Education budget, but with a huge ROI in terms of future productivity gains. Targets: reimagined guidance and navigation counseling for all ages; learning R&D and brain plasticity research; expanded high school course diversity; reinvented community college and university capacity for 30–80 year old learners. |
|                               | AI Governance & Auditing Standards          | NIST-anchored public-private standards partnership. Industry funds implementation; government sets principles; independent body certifies compliance. Modeled on aviation, pharmaceutical, and financial oversight architectures.   |
|                               | Stakeholder & Community Engagement Platform | Dedicated cross-sector engagement infrastructure connecting AI developers, regulators, educators, employers, and community organizations. Generates the local legitimacy and feedback loops that national policy cannot manufacture from the top down.  |
| <b>ACCOUNTABILITY</b>         | Metrics, Reporting & Adjustment Triggers    | Annual public reporting on displacement rates, retraining outcomes, community economic indicators, and AI incident data. Pre-specified adjustment triggers: if displacement outpaces transition program capacity by more than 15%, mandatory fund increases activate. Overseen by GAO with independent academic review.   |

## VIII. LEVERAGING EXISTING INSTITUTIONS

The United States does not need to invent an entirely new bureaucracy to begin. The NIST AI Risk Management Framework provides an existing, broadly adopted governance architecture. NIST can extend that work to develop standards for AI evaluation, auditability, documentation, and process-level intelligibility; its 2026 concept note on AI in critical infrastructure is a direct step in this direction. Sector regulators can adapt those standards to their domains: the FDA for medical AI, the SEC for financial systems, the FAA for aviation, DoD for defense applications. Procurement offices can make resilience a purchasing requirement. Courts and legislatures can clarify accountability and human review rights.

DARPA and the national laboratories can conduct adversarial testing and red-team exercises. The GAO and Inspectors General can ensure accountability for federal AI systems and flag cognitive dependency risks before they become irreversible. This distributed governance model avoids the risk of a single centralized AI regulator becoming a bottleneck or a capture target. Each institution brings domain authority and existing relationships with the professionals whose competence must be preserved.

The Lifelong Learning Transformation Initiative is essential for reorganizing the U.S. higher education and workforce development system around a fundamentally different mission. A system designed for 18-to-22-year-olds moving through education once is becoming increasingly obsolete. The future will require a system that serves 18-to-80-year-olds in continuous transition. Community colleges, already the most nimble and accessible institutions in the higher education system, are the natural delivery mechanism for much of this. Employers, unions, and professional associations are natural co-designers.

A convening or conference focused on human sovereignty and AI — bringing together AI developers, sector regulators, educators, community organizations, and civil society — could help align these actors around shared implementation priorities. Again, there's precedent for it. When the reports about the nation's educational decline came out in the 1980s, a similar Education Summit was convened in 1989 by President George H.W. Bush. The infrastructure for trustworthy AI should be built with the same seriousness that earlier generations applied to aviation safety, pharmaceutical oversight, and cybersecurity. Those industries are not trustworthy because their leaders chose honesty; they are trustworthy because institutions were built that made honesty verifiable and dishonesty costly. What will make them stick is if this is seen not just as a government initiative or an industry initiative, but as a whole of society initiative.

## IX. DESIGNING FOR HUMAN–AI COMPLEMENTARITY

The preceding sections address the risks that AI governance must mitigate and the social obligations it must fulfill. But a durable framework must also articulate what good AI deployment looks like — not as a constraint on technology, but as the standard by which its success should be measured.

Artificial intelligence is best understood as a productivity technology: one that expands the scope, speed, and scale of what human beings can do. The central design challenge is not whether to use AI but how to structure its use so that it consistently elevates human decision-making rather than displacing it. The goal is human-plus-machine, with a clear understanding of what each does best.

### **Division of Labor: From Substitution to Elevation**

AI's comparative advantage lies in pattern recognition across large datasets, rapid synthesis of information, simulation and iteration at scale, and consistency in routine or repetitive tasks. These capabilities are most valuable when they allow human actors to move up the value chain of cognition — away from tasks that can be systematized and toward the tasks that require judgment, context, and accountability. AI is well suited to descriptive labor: tasks grounded in facts, data processing, classification, and pattern identification. Human beings remain essential in prescriptive labor: tasks involving judgment, values, trade-offs among competing goods, and decisions for which someone must be accountable. The objective is not to prevent AI from participating in decision-making processes but to ensure that final authority, especially in consequential domains, remains legible, reviewable, and human.

### **Decision Architecture: Supporting Judgment Under Uncertainty**

Real-world decision-making is shaped by incomplete data, time pressure, competing objectives, and uncertainty about downstream effects. AI systems should be designed to support human judgment by surfacing uncertainty and confidence levels, making underlying assumptions visible, presenting alternative scenarios, and enabling users to weigh trade-offs. Rather than eliminating discretion, the goal should be to structure it — to give decision-makers the information landscape they need to act wisely rather than presenting AI outputs as authoritative conclusions. Decision-makers should be empowered to operate along scales and ranges rather than binary rules, because risk tolerance varies by context, because speed is sometimes properly prioritized over completeness, and because least-worst outcomes are often preferable to theoretically optimal ones that cannot be achieved.

### **Skills Preservation: Capability as a Strategic Asset**

Preserving human capability is not only a defensive measure against AI failure; it is a source of long-term competitive and democratic strength. AI can become one of the most powerful tools for documenting, transmitting, and extending human knowledge across time and communities — preserving endangered languages and cultural traditions, capturing tacit knowledge from declining industries, recording expert practices in medicine, engineering, and craft, maintaining institutional memory across generations. Used intentionally, AI can serve as a repository of

human capability that would otherwise be lost. But skill preservation requires more than archival function: humans must retain the ability to perform and understand the tasks that AI supports. Effective knowledge systems should operate on two levels simultaneously — expert-level documentation that enables specialists to audit, replicate, and refine work, and lay-level documentation accessible to non-specialists and decision-makers. A system that cannot be understood outside a narrow technical class is difficult to oversee; a system that lacks expert grounding is difficult to trust.

### **Education and the Expansion of Human Potential**

The rise of AI does not reduce the need for human education. It increases it. If AI is deployed primarily as a substitute for human effort, educational incentives weaken over time; if deployed as a complement to human capability, the opposite occurs — education becomes more valuable as the returns to judgment, ethics, creativity, and contextual reasoning rise relative to tasks that machines assume. A healthy AI ecosystem must support a wide range of human aspirations. There are communities that seek to preserve traditional ways of life — the Amish model of selective technology adoption is a legitimate exercise of autonomy, not a failure of modernity — and there are societies pursuing the most ambitious frontiers of science and exploration. These are not mutually exclusive. The role of AI should be to expand the feasible set of human outcomes, not to collapse it into a single model of efficiency. A governance framework that accommodates this range — preserving both autonomy and ambition, both privacy and frontier innovation — is more stable, more legitimate, and more aligned with democratic principles. What this means concretely is that education policy in an AI age must hold two things simultaneously: rigorous investment in the technical literacy needed to understand and audit AI systems, and equally rigorous investment in the humanistic capacities — ethical reasoning, historical judgment, civic knowledge, creative and critical thinking — that AI cannot replace and that democratic self-governance cannot function without.

### **Complementarity as the Design Standard**

The long-term success of AI will depend less on its raw capability than on how well it is integrated into human systems. A complementarity-based approach provides a clear and testable standard for evaluating AI deployment: does this system increase human capability, or does it obscure it? Does it support judgment, or does it replace it? Does it preserve optionality, or does it create dependency? These are not rhetorical questions; they are measurable design criteria that can inform procurement decisions, professional standards, regulatory requirements, and institutional design. Designing for human-AI complementarity is not an additional objective layered on top of the governance framework. It is the condition under which all the other objectives — economic, scientific, constitutional, and national — can be sustained.

## X. THE STAKES

The point of this framework is to make AI more usable, not less. If the United States wants AI to scale across the economy and the institutions of governance, it has to solve the trust problem that comes with powerful systems operating in high-stakes domains. It also has to solve the legitimacy problem: a technology whose benefits accrue primarily to a small number of wealthy developers and early adopters, while its costs fall on workers, communities, and institutions unprepared for the transition, will not maintain the political conditions necessary for its own continued development.

The manufacturing parallel is instructive. America walked into deindustrialization with open eyes because short-term efficiency gains outweighed the institutional voice warning about long-term fragility. Reversing this has been a bipartisan priority (albeit implemented in different ways), for the past decade. Intellectual protectionism asks that the lesson be applied prospectively — to cognitive capacity, to the human reserve that AI governance requires, and to the communities that will bear the transitional costs of a technology they did not choose and cannot easily avoid.

Those who believe AI's risks are so severe it should be halted are wrong: the technology's potential to address genuine human needs is real, and the right response is not to stop but to build — the auditing professions, the testing requirements, the trust substitutes, the human reserve, the social compact, and the institutional architecture that make powerful AI governable.

***The future of AI should be one in which machines do more of the routine work, while humans keep the authority, judgment, and institutional strength to direct what matters.***

That is a pro-AI position, a pro-industry position, and a pro-constitutional position at the same time. It is also the only position that takes seriously both what AI could accomplish and what is required — in governance, in social investment, and in human development — to ensure that it does.

**Stephen Jordan** is the Executive Director of the Institute for Sustainable Development, an “action tank” committed to building communities that work.

## REFERENCES

1. Amodei, Dario. “Machines of Loving Grace: How AI Could Transform the World for the Better.” [darioamodei.com](https://www.darioamodei.com/essay/machines-of-loving-grace), October 2024. <https://www.darioamodei.com/essay/machines-of-loving-grace>
2. Andreessen, Marc. “The Techno-Optimist Manifesto.” Andreessen Horowitz, October 2023. <https://a16z.com/the-techno-optimist-manifesto/>
3. Bureau d’Enquêtes et d’Analyses pour la sécurité de l’aviation civile (BEA). Final Report on the Accident on 1 June 2009 to the Airbus A330-203 Air France Flight AF 447. July 2012. <https://www.bea.aero/en/investigation-reports/notified-accidents/detail/accident-af447.html>
4. Garreau, Joel. *Radical Evolution: The Promise and Peril of Enhancing Our Minds, Our Bodies — and What It Means to Be Human*. Doubleday, 2005.
5. National Institute of Standards and Technology (NIST). Artificial Intelligence Risk Management Framework (AI RMF 1.0). NIST AI 100-1. U.S. Department of Commerce, January 2023. <https://doi.org/10.6028/NIST.AI.100-1>
6. OpenAI. “Preparedness Framework Version 2.” April 2025. <https://cdn.openai.com/pdf/18a02b5d-6b67-4cec-ab64-68cdfbddebcd/preparedness-framework-v2.pdf>
7. Parasuraman, Raja, and Victor Riley. “Humans and Automation: Use, Misuse, Disuse, Abuse.” *Human Factors* 39, no. 2 (1997): 230–253. <https://doi.org/10.1518/001872097778543886>
8. Pew Research Center. “How Americans View Artificial Intelligence and Its Impact on People and Society.” November 2024. <https://www.pewresearch.org/internet/2024/11/21/how-americans-view-artificial-intelligence-and-its-impact-on-people-and-society/>
9. Stanford Institute for Human-Centered Artificial Intelligence (HAI). Artificial Intelligence Index Report 2025. Stanford University, April 2025. <https://hai.stanford.edu/ai-index/2025-ai-index-report>
10. Thiel, Peter. *Zero to One: Notes on Startups, or How to Build the Future*. Crown Business, 2014.

### Suggested Citation

Jordan, Stephen C. “Human Sovereignty in the Age of AI: A Strategic Framework for Cognitive Resilience and Human Flourishing.” *Institute for Sustainable Development Policy Paper*, 2026. [isdus.org](https://isdus.org)