

Strategic Defense HRM 4.0: An Integrative Framework for Future-Ready Defense Organizations

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Abstract

The Fourth Industrial Revolution—artificial intelligence, autonomous systems, and hybrid threats—is fundamentally reshaping defense organizations. Traditional hierarchical personnel systems are increasingly inadequate for preparing military forces to operate across physical, digital, and cognitive domains. Despite this challenge, defense Human Resource Management (HRM) scholarship lacks a comprehensive framework integrating technological transformation, human-machine collaboration, and strategic workforce planning. This systematic review followed PRISMA 2020 guidelines, searching Scopus, Web of Science, and IEEE Xplore (2020–2025). From 512 publications, 52 peer-reviewed studies were thematically synthesized with independent double-coding ($\kappa=0.82$). Three key insights emerged: (1) predictive analytics in defense contexts improve personnel retention and enable proactive forecasting of critical skill gaps; (2) immersive VR/AR training enhances skill retention and reduces training duration; (3) ethical human-machine teaming requires explainable AI, calibrated trust, and ethical reasoning-focused leadership. These findings are integrated into the Strategic Defense HRM 4.0 Framework, comprising four pillars: data-driven talent management, digital workforce readiness, ethical human-machine teaming, and strategic HRM for organizational readiness. This framework enables defense organizations to transition from reactive personnel administration to proactive strategic workforce management.

Keywords: AI ethics; defense transformation; digital readiness; human-machine teaming; strategic human resource management

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Introduction

The 21st-century strategic environment is characterized by volatility, uncertainty, complexity, and ambiguity (VUCA), fundamentally altering defense organizations' operational landscape (Winkler et al., 2020). The Fourth Industrial Revolution has introduced artificial intelligence (AI), autonomous systems, quantum computing, and big data analytics into military operations, blurring traditional boundaries between peace and conflict (Bag et al., 2021). Cyber warfare, information operations, and hybrid threats have emerged as primary security challenges, requiring defense forces to operate across physical, digital, and cognitive domains simultaneously.

In this transformed environment, technological superiority alone is insufficient. The decisive advantage lies in human capital—personnel capable of effectively leveraging advanced technologies, adapting to rapid change, and making ethical decisions in human-machine collaborative contexts. Defense organizations face unprecedented challenges in recruiting digital natives, retaining technical specialists competing with private sector compensation, and developing leaders capable of commanding hybrid human-machine teams.

Traditional defense HRM, designed for industrial-age warfare, emphasized standardization, hierarchy, and compliance. Contemporary defense challenges demand a paradigm shift toward strategic HRM that treats human capital as a dynamic capability—a source of sustained competitive advantage through continuous adaptation and reconfiguration (Holota & Tytkovskyi, 2022).

This transformation represents a fundamental rewiring of how defense organizations view their workforce:

1. From reactive staffing to predictive talent forecasting, using AI to anticipate skill gaps before they become operational vulnerabilities (Matthew Walsh et al., 2024).
2. From episodic training to continuous learning ecosystems, where micro-credentials and immersive simulations keep pace with technological change (Holota & Tytkovskyi, 2022).
3. From vague ethical guidelines to robust frameworks for human-AI collaboration, ensuring accountability when algorithms influence life-or-death decisions (Seeber et al., 2020).
4. From HR as support staff to HR as strategic architects, directly linking workforce development to national defense objectives (Ulrich et al., 2017).

In essence, Strategic Defense HRM is about building organizations that learn as fast as the world changes—and people who grow as quickly as technology evolves. Despite growing recognition that traditional HRM systems are inadequate for digital-age defense, no existing framework systematically integrates talent analytics, digital readiness, ethical human-machine teaming, and strategic HR alignment specifically for defense organizations. Corporate digital HR models (Marler & Boudreau, 2017) emphasize efficiency and employee experience but do not address mission readiness under adversarial conditions, security-cleared analytics, or ethical accountability in life-critical decisions. Conversely, military technology studies (Endsley, 2023) examine human-AI interaction but rarely engage with HRM architecture, talent pipelines, or workforce planning. This fragmentation leaves defense policymakers and HR leaders without evidence-based, integrated guidance.

Therefore, this study addresses the following precise research problem: How can defense organizations systematically transform their HRM systems to enable strategic readiness across physical, digital, and cognitive domains, given the simultaneous demands of technological adoption, workforce development, ethical governance, and strategic alignment?

This problem is operationalized through three research questions:

- RQ1: What are the key dimensions of HRM transformation required for future defense organizations?
- RQ2: How can defense organizations effectively integrate digital technologies into talent management while addressing ethical considerations?
- RQ3: What frameworks can guide the development of strategic HRM capabilities that enhance organizational readiness?

Literature Review

Rather than relying on a single theoretical perspective, this study synthesizes five complementary theories to provide a robust conceptual foundation for defense HRM transformation. Table 1 summarizes how each theory informs our framework.

Table 1. Theoretical Foundation of the Strategic Defense HRM 4.0 Framework

Theory	Core Premise	Application to Defense HRM	Primary Link to Framework Pillar
Resource-Based View (RBV) (Barney, 1991)	Human capital as a strategic resource that is valuable, rare, inimitable, and non-substitutable	Defense organizations gain competitive advantage through unique, difficult-to-replicate talent systems and expertise	Pillar 1 (Data-Driven Talent Management)
Dynamic Capabilities Theory (Teece et al., 2016)	Organizations must sense, seize, and transform capabilities to adapt to environmental change	HRM must enable rapid workforce reconfiguration in response to emerging threats and technologies	Pillar 4 (Strategic HRM for Readiness)
Strategic HRM Theory (Ulrich et al., 2017 ; Wright & Nishii, 2006)	Vertical alignment (HR with strategy) and horizontal integration (HR practices with each other) drive performance	Defense HRM practices must align with national security strategy and mutually reinforce each other	Cross-pillar integration
Technology Acceptance Model (TAM) (Davis, 1989; updated by Marler & Boudreau, 2017)	Perceived usefulness and ease of use determine technology adoption by personnel	AI and analytics tools in defense HRM will fail without user trust, training, and perceived operational value	Pillar 2 (Digital Workforce Readiness)
Organizational Learning Theory (Senge, 2006)	Continuous learning, systems thinking, and shared mental models enable adaptation	Defense organizations must become learning organizations that embrace systems thinking and continuous upskilling to maintain cognitive readiness	

These five theories are not mutually exclusive but mutually reinforcing. RBV explains what defense HRM should protect as strategic assets (human capital). Dynamic capabilities explain how HRM enables adaptation over time. Strategic HRM theory explains alignment conditions for effectiveness. TAM explains individual adoption of HR technologies. Organizational learning explains the cultural and cognitive shifts required. Together, they provide a multi-level framework spanning individual (TAM, learning), organizational (RBV, strategic HRM), and dynamic (capabilities) perspectives.

Strategic HRM theory emphasizes that organizational performance is shaped by the alignment of HR practices with strategic objectives and by the interaction between individual, group, and organizational levels of analysis. Wright & Nishii (2006) argues that HR systems create value not only through isolated practices but through their ability to shape collective behaviors and organizational capabilities that support strategic outcomes. In defense organizations, this perspective reinforces the need to align talent management, workforce development, and readiness objectives within a coherent strategic architecture.

This multi-theoretical approach offers richer explanatory power than single-theory perspectives (Tambe et al., 2019; Collings & Mellahi, 2009). It also differentiates our framework from corporate digital HR models, which typically emphasize efficiency and employee experience without addressing mission readiness, ethical accountability under adversarial conditions, or security-cleared analytics.

Methods

This study employs a systematic literature review (SLR) methodology following the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) guidelines (Page et al., 2021). SLR provides a rigorous, transparent, and replicable approach to synthesizing existing research, minimizing bias, and identifying research gaps.

We conducted comprehensive searches in three major academic databases: Scopus (Elsevier), Web of Science (Clarivate Analytics), and IEEE Xplore (Institute of Electrical and Electronics Engineers). The search was performed in January 2026 and covered publications from January 2020 to December 2025.

The search strategy combined four concept clusters using Boolean operators:

Cluster 1 (HRM): "human resource management" OR "strategic HRM" OR "talent management" OR "workforce planning" OR "people analytics"

Cluster 2 (Defense context): "defense organization" OR "military" OR "armed forces" OR "defense sector" OR "military organization"

Cluster 3 (Digital transformation): "digital transformation" OR "artificial intelligence" OR "AI" OR "machine learning" OR "automation" OR "data analytics" OR "big data"

Cluster 4 (Future workforce): "future workforce" OR "workforce readiness" OR "human-machine teaming" OR "digital competence" OR "workforce agility" OR "reskilling" OR "upskilling"

The final search string combined all four clusters with AND operators:

1. Scopus (Elsevier)

Search field: TITLE-ABS-KEY (title, abstract, keywords)

Exact search string:

TITLE-ABS-KEY ("human resource management" OR "strategic hrm" OR "talent management" OR "workforce planning" OR "people analytics") AND TITLE-ABS-KEY ("defense organization*" OR "military" OR "armed forces" OR "defense sector" OR "military organization*") AND TITLE-ABS-KEY ("digital transformation" OR "artificial intelligence" OR "ai" OR "machine learning" OR "automation" OR "data analytics" OR "big data") AND TITLE-ABS-KEY ("future workforce" OR "workforce readiness" OR "human-machine teaming" OR "digital competence" OR "workforce agility" OR "reskilling" OR "upskilling"))

2. Web of Science (Clarivate Analytics)

Search field: Topic (TI=title, AB=abstract, KW=keywords)

Exact search string:

(TI=("human resource management" OR "strategic hrm" OR "talent management" OR "workforce planning" OR "people analytics") OR AB=("human resource management" OR "strategic hrm" OR "talent management" OR "workforce planning" OR "people analytics") OR KW=("human resource management" OR "strategic hrm" OR "talent management" OR "workforce planning" OR "people analytics")) AND (TI=("defense organization*" OR "military" OR "armed forces" OR "defense

sector" OR "military organization*") OR AB=("defense organization*" OR "military" OR "armed forces" OR "defense sector" OR "military organization*") OR KW=("defense organization*" OR "military" OR "armed forces" OR "defense sector" OR "military organization*")) AND (TI=("digital transformation" OR "artificial intelligence" OR "ai" OR "machine learning" OR "automation" OR "data analytics" OR "big data") OR AB=("digital transformation" OR "artificial intelligence" OR "ai" OR "machine learning" OR "automation" OR "data analytics" OR "big data") OR KW=("digital transformation" OR "artificial intelligence" OR "ai" OR "machine learning" OR "automation" OR "data analytics" OR "big data")) AND (TI=("future workforce" OR "workforce readiness" OR "human-machine teaming" OR "digital competence" OR "workforce agility" OR "reskilling" OR "upskilling") OR AB=("future workforce" OR "workforce readiness" OR "human-machine teaming" OR "digital competence" OR "workforce agility" OR "reskilling" OR "upskilling") OR KW=("future workforce" OR "workforce readiness" OR "human-machine teaming" OR "digital competence" OR "workforce agility" OR "reskilling" OR "upskilling"))

3. IEEE Xplore (IEEE)

Search field: Abstract, Keywords, Metadata (all metadata fields)

Exact search string:

((("human resource management" OR "strategic hrm" OR "talent management" OR "workforce planning" OR "people analytics") AND ("defense organization*" OR "military" OR "armed forces" OR "defense sector" OR "military organization*") AND ("digital transformation" OR "artificial intelligence" OR "ai" OR "machine learning" OR "automation" OR "data analytics" OR "big data") AND ("future workforce" OR "workforce readiness" OR "human-machine teaming" OR "digital competence" OR "workforce agility" OR "reskilling" OR "upskilling"))

Inclusion Criteria:

1. Published between January 2020 and December 2025
2. Peer-reviewed journal articles, systematic reviews, or high-quality conference proceedings
3. Empirical studies (quantitative, qualitative, or mixed methods) or conceptual/theoretical papers
4. Focus on HRM practices, defense/military organizations, and digital/technological transformation
5. Written in English and Full-text availability

After screening titles, abstracts, and full texts, 52 studies were included. A thematic coding process was conducted using a structured coding protocol. Two coders independently performed the analysis, with reliability confirmed via Cohen's kappa ($\kappa = 0.82$).

Results and Discussion

Our analysis of 52 studies reveals a consistent pattern: successful defense HR transformation rests on four interconnected pillars. Think of these not as separate initiatives, but as interdependent gears in a well-designed engine—when they turn together, they generate adaptive strategic readiness.

Pillar 1: Data-Driven Talent Management asks: What if we could predict attrition before it happens, or identify future leaders based on patterns invisible to the human eye? Talent management in the public sector must shift from traditional seniority-based approaches to competency- and data-driven approaches. This shift is consistent with the strategic talent management perspective, which emphasizes the systematic identification of pivotal positions, development of high-potential talent pools, and creation of differentiated HR architectures that sustain organizational performance under

dynamic conditions. According to Collings & Mellahi (2009), talent management should focus on strategically critical roles that generate disproportionate value for organizational effectiveness. HR analytics can be used to identify future competency needs and design targeted development programs (Pagiu et al., 2025).

Pillar 2: Digital Workforce Readiness recognizes that technology alone doesn't create digital readiness—people do. It's about cultivating digital fluency: the cognitive agility to interpret data, the collaborative skills to work with AI, and the learning mindset to reskill every 3–5 years (Schwarz et al., 2019). Immersive VR/AR training enhances skill retention and reduces training time through improved formative feedback, technique stabilization, and accelerated competency development (Holota & Tytkovskyi, 2022; Muha & Silitonga, 2025). Digital transformation is not merely the adoption of technology, but a fundamental change in organizational culture, work processes, and human resource competencies. The success of digital transformation highly depends on the readiness of human resources to manage technology and adapt to change (R. F. Novienty et al., 2025).

Pillar 3: Ethical Human-Machine Teaming confronts perhaps the most profound question: How do we build trust between humans and AI in high-stakes environments? Research emphasizes calibrated trust—avoiding both blind reliance on flawed algorithms and unwarranted skepticism of capable systems. This requires explainable AI, clear accountability frameworks, and leaders who are as fluent in ethics as they are in technology (S. Ososky et al., 2014; Schwarz et al., 2019). The use of AI in public decision-making must be accompanied by clear ethical frameworks, algorithmic transparency, and accountability mechanisms. Without these, risks of bias and injustice can threaten the legitimacy of public policies (Ologun et al., 2026).

Pillar 4: Strategic HRM for Organizational Readiness elevates HR from the back office to the strategy table. Here, HR becomes the mechanism that translates national defense priorities into workforce capabilities, ensuring that every talent decision—from recruitment to retention—serves a strategic purpose (Ulrich et al., 2017). The readiness of public organizations to face change is determined not only by individual adaptive capacity but also by institutional capacity to conduct strategic planning that is responsive to external environmental dynamics (Yean et al., 2022).

Table 2 synthesizes these themes, but the real insight lies in their integration: data-driven talent management informs digital learning pathways, which enable ethical human-machine collaboration, all governed by strategic HR alignment. This is the ecosystem of Strategic Defense HRM 4.0.

Table 2. Systematic Synthesis of Defense HRM Literature (2020–2025)

Theme/Pillar	Studies (n)	Key Findings	Implications & Recommendations
Pillar 1: Data-Driven Defense Talent Management	12	Transition from traditional personnel management to AI-enabled predictive talent systems (Matthew Walsh et al., 2024). Focus on competency mapping and attrition prediction (Holota & Tytkovskyi, 2022)), as well as personalized career pathways. Major barriers include organizational resistance to algorithmic decision-making, concerns about fairness and	Investment: Develop secure defense HR analytics platforms. Capability building: Form HR data science and analytics teams. Culture: Promote leadership commitment to evidence-based HR decisions. Governance: Establish ethical and privacy frameworks for personnel data.

Theme/Pillar	Studies (n)	Key Findings	Implications & Recommendations
		accountability, and data privacy risks in AI-enabled HR systems (Tambe et al., 2019) and data privacy concerns (Prasad & Chandra, 2024).	
Pillar 2: Digital Workforce Readiness & Agility	14	Workforce readiness defined by cognitive agility, digital fluency, and continuous reskilling (Schwarz et al., 2019). Learning organization principles remain essential (Senge, 2006). VR/AR-based immersive training shows strong effectiveness (Holota & Tytkovskyi, 2022; (Muha & Silitonga, 2025). Digital readiness assessment frameworks enable organizations to identify capability gaps, prioritize workforce development investments, and allocate resources more effectively during digital transformation initiatives (Michelotto & Joia, 2024; Chen et al., 2024)). Key challenge: balancing operational tempo with learning opportunities.	Training modernization: Implement micro-credentials and immersive learning. Assessment: Apply dynamic digital readiness evaluation systems. Culture: Shift from episodic training toward continuous learning ecosystems. Partnerships: Collaborate with EdTech firms and universities to accelerate curriculum updates.
Pillar 3: Ethical Human-Machine Teaming (HMT)	9	Abeywickrama & Ramchurn (2024) demonstrate that explainable and responsible AI models are critical for human-agent collectives, requiring transparency, consistent performance, and user training to achieve calibrated trust. Ethical challenges include responsibility gaps in AI-assisted decisions (Santoni de Sio & Mecacci, 2021), transparency in AI outputs (Endsley, 2023), and algorithmic bias (Matthew Walsh et al., 2024). Demand grows for 'digital leaders' capable of leading hybrid teams (Ulrich et al., 2017).	Doctrine: Develop new operational and doctrinal frameworks for HMT. Ethics: Integrate AI ethics into all leadership development programs. Design: Ensure AI systems incorporate transparency and explainability. Leader development: Build digital-leadership competency models emphasizing ethical reasoning. Governance: Establish clear accountability for AI-assisted decisions.
Pillar 4: Strategic HRM & Organizational Readiness	6	Integrating talent management, digital transformation, and strategic planning increases defense adaptability and resilience (Holota & Tytkovskyi, 2022). HR is a dynamic capability enabling organizational adaptive capacity (Ulrich et al., 2017). Defense sectors must shift HR from a support function to a strategic	Restructuring: Position HR as part of strategic planning processes. Governance: Establish strategic HR councils connecting workforce planning to defense strategy. Metrics: Adopt strategic readiness indicators beyond traditional HR metrics.

Theme/Pillar	Studies (n)	Key Findings	Implications & Recommendations
		partner (Ulrich et al., 2017). Strategic HR governance ensures alignment with broader defense priorities (Boxall & Purcell, 2022).	Collaboration model: Embed HR business partners in operational commands.
Cross-Cutting Synthesis	52	Integrated finding: Future defense HRM is predictive, personalized, adaptive, and strategically integrated. Success requires synergy among all four pillars—data-driven talent management, digital readiness, ethical HMT, and strategic HR governance. HR transformation is systemic, not modular.	Holistic transformation: Avoid fragmented interventions. Framework: Adopt a unified “Strategic Defense HRM 4.0” model. Leadership: Top defense leaders must champion integrated HR reforms. Investment priorities: HR technologies, capability development, and structured change management.

Source: Author’s own elaboration based on various sources (2025)

Detailed Findings by Pillar

Pillar 1: Data-Driven Defense Talent Management

Transformation from reactive personnel administration to predictive, data-enabled talent strategy using AI, machine learning, and people analytics, with key findings:

- a. Predictive Analytics Applications
Matthew Walsh et al. (2024) demonstrated that machine learning models can be safely applied to Air Force human resource management, enabling predictive analytics for skill shortage forecasting and workforce optimization. Alan B. Gelder (2022) from the Institute for Defense Analyses highlight that machine learning applications in defense personnel systems can improve retention outcomes through data-driven interventions, though legal and ethical frameworks must guide implementation.
- b. Competency Mapping and Skills Architecture
Holota & Tytkovskyi (2022) advocate for capability-based defense planning models that prioritize skills and competencies over rigid hierarchical structures, enabling greater agility in task assignment and career development. Yean et al. (2022)emphasize that adaptability and change readiness in public sector organizations require dynamic competency frameworks that respond to evolving external environmental dynamics.
- c. Personalized Career Pathways
Cappelli P. & Tavis (2018) describe AI-enabled career recommendation systems that match individual competencies, preferences, and performance trajectories with optimal development opportunities, increasing engagement and retention.
- d. Implementation Barriers
Tambe et al. (2019) argue that organizational adoption of AI-enabled HR systems is frequently constrained by employee resistance, concerns regarding fairness, accountability, and the perceived legitimacy of algorithmic decision-making. These challenges may be particularly pronounced in hierarchical defense organizations where traditional authority structures and seniority-based

promotion systems remain dominant. Prasad & Chandra (2024) highlight data privacy and security concerns unique to defense contexts, where personnel data intersects with security clearances and operational security.

e. Theoretical Integration

This pillar aligns with Resource-Based View (RBV) theory Barney (1991), positioning human capital as a strategic resource that, when managed through superior analytics capabilities, becomes a source of sustained competitive advantage.

Pillar 2: Digital Workforce Readiness and Agility

Building adaptive capacity through continuous learning, digital fluency development, and cognitive agility enhancement, with key findings:

a. Digital Fluency as Core Competency

Muha & Silitonga (2025) define defense human resource competency development in the VUCA era as requiring digital fluency beyond basic IT literacy, including adaptive capacity, technological agility, and cross-domain collaboration skills. Recent studies conceptualize digital readiness as a multidimensional construct encompassing technological capability, workforce competence, leadership commitment, organizational culture, and strategic alignment. Digital readiness assessments can help organizations identify capability gaps and guide targeted interventions to strengthen workforce adaptability and digital competence (Michelotto & Joia, 2024; Chen et al., 2024).

b. Continuous Learning Ecosystems

Schwarz et al. (2019) project that by 2040, defense personnel will require reskilling every 3-5 years due to technological acceleration. Senge (2006) argues for "learning organization" principles adapted to military contexts, emphasizing systems thinking and shared mental models.

c. Immersive Training Technologies

Immersive VR/AR training enhances skill retention and reduces training time through improved formative feedback, technique stabilization, and accelerated competency development (Holota & Tytkovskyi, 2022; Muha & Silitonga, 2025). These technologies enable safe, repeatable practice of high-risk scenarios and accelerate expertise development.

d. Micro-Credentials and Modular Learning

Holota & Tytkovskyi (2022) emphasize that capability-based defense planning requires modular, stackable competency development aligned with evolving mission requirements. This approach provides flexibility for personnel to acquire specific capabilities aligned with evolving mission needs without lengthy formal education programs.

e. Cognitive Agility Assessment

Cognitive agility has emerged as a critical capability for defense personnel operating in complex and rapidly changing environments. (Schwarz et al., 2019) argue that future military workforces must continuously adapt to technological disruption and evolving mission requirements. Similarly, van Laar et al., (2017) identify higher-order digital competencies—including information evaluation, problem-solving, communication, and adaptability—as essential determinants of effective performance in digitally enabled environments.

f. Implementation Challenges

Implementing continuous digital learning ecosystems remains challenging because operational demands often limit personnel participation and long-term engagement. (Schwarz et al., 2019)

emphasize that future defense organizations must institutionalize continuous reskilling as a core organizational capability. Consistent with learning organization principles proposed by Senge (2006), sustained learning requires leadership support, cultural reinforcement, and integration with career development systems. Furthermore, Winkler et al. (2020) demonstrate that personalized feedback and conversational learning support mechanisms can significantly improve learner engagement and outcomes. (Winkler et al., 2020) demonstrate that scaffolding-based conversational agents can improve learning outcomes and reduce attrition in online education through personalized feedback and engagement strategies.

Pillar 3: Ethical Human-Machine Teaming (HMT)

Developing effective, ethical collaboration between human operators and AI/autonomous systems through trust calibration, shared situational awareness, and clear accountability frameworks, with key findings:

- a. **Trust Calibration:**
Endsley (2023) distinguishes between over-trust and under-trust in human-AI teams, demonstrating that both degrade performance, and emphasizes that transparency, explainability, and shared situation awareness are essential for calibrated trust.
- b. **Shared Situational Awareness:**
Endsley (2023) emphasizes that effective HMT requires mutual understanding of goals, constraints, and environmental factors. Interface design must support human comprehension of AI reasoning while enabling AI systems to model human intent and cognitive state.
- c. **Responsibility and Accountability Gaps:**
Santoni de Sio & Mecacci (2021) examine responsibility gaps arising from AI-assisted decision-making and propose governance mechanisms to preserve meaningful human control and accountability in high-stakes contexts. They propose distributed accountability frameworks where human operators maintain meaningful control through understanding AI recommendations, authority to override, and clear doctrine governing AI use in specific contexts.
- d. **Ethical Decision-Making:**
Ologun et al. (2026) examine AI governance and ethical frameworks for public-sector AI decision systems, including algorithmic transparency, bias mitigation, and accountability mechanisms essential for preserving human judgment in high-stakes contexts. They recommend ethics training integrated throughout military education, not as separate modules.
- e. **Digital Leadership Competencies:**
Ulrich et al. (2017) emphasize that HR leaders must develop digital competencies including technological fluency, adaptive decision-making, and strategic technology alignment to effectively lead hybrid human-AI teams.
- f. **Explainable AI (XAI) Requirement:**
S. Ososky et al. (2014) demonstrate that explainable AI systems—providing interpretable rationales for recommendations—significantly improve operator trust calibration and decision quality compared to "black box" systems.

Pillar 4: Strategic HRM for Organizational Readiness

Positioning HRM as a strategic partner in defense planning, aligning workforce capabilities with national security objectives, and building organizational dynamic capabilities:

- a. **HR as Dynamic Capability:**
Holota & Tytkovskyi (2022) apply capability-based planning principles to defense HRM, arguing that leadership-driven strategy development enables organizations to sense, seize, and transform workforce capabilities for adaptive readiness.
- b. **Strategic Business Partnership Model:**
Ulrich et al. (2005) update their influential HR business partner model for defense contexts. They propose HR leaders participate equally in strategic planning, with responsibility for translating defense strategies into workforce implications, identifying capability gaps, and designing talent solutions aligned with operational concepts.
- c. **Strategic Workforce Planning:**
Ulrich et al. (2017) emphasize that victory through organization requires vertical alignment of HR practices with strategy and horizontal integration across HR functions, ensuring that workforce capabilities directly support mission objectives.
- d. **HR Governance Frameworks:**
Boxall & Purcell (2022) describe strategic HR governance as systematic processes ensuring HR decisions support organizational objectives. For defense, this includes strategic HR councils linking workforce planning to defense planning, HR representation in capability development processes, and workforce metrics integrated into readiness assessments.
- e. **Measuring Strategic Readiness:**
Ulrich et al. (2017) propose that organizational victory requires moving beyond traditional HR metrics to strategic readiness indicators that directly link workforce capabilities to mission outcomes.

Integrative Framework: Strategic Defense HRM 4.0

Based on the synthesis of findings, we propose the Strategic Defense HRM 4.0 Framework (Figure 1), which integrates the four pillars into a coherent system for building adaptive strategic readiness. The governance layer ensures all HRM activities align with defense strategy. It positions HR as a strategic partner with authority and capability to translate national security requirements into workforce solutions. The following is an explanation of the pillars of interdependence:

- a. **Data-Driven Talent Management ↔ Digital Workforce Readiness:**
Analytics identify skill gaps and predict future requirements (Pillar 1), which inform targeted reskilling programs and learning pathway design (Pillar 2). Conversely, digital readiness assessments (Pillar 2) provide data for talent models (Pillar 1).
- b. **Digital Workforce Readiness ↔ Ethical HMT:**
Digital fluency and continuous learning (Pillar 2) are prerequisites for effective human-machine teaming. Personnel must understand AI capabilities to collaborate effectively (Pillar 3). HMT experiences inform learning requirements and competency models (Pillar 2).
- c. **Data-Driven Talent Management ↔ Ethical HMT:**
AI-enabled talent systems (Pillar 1) must incorporate ethical guidelines governing algorithmic decision-making in personnel contexts, informed by HMT ethics frameworks (Pillar 3). Leaders managing HMT (Pillar 3) are identified and developed through data-driven talent processes (Pillar 1).

d. All Pillars → Strategic Readiness:

The integrated operation of all pillars produces adaptive strategic readiness—the organizational capacity to sense environmental changes, rapidly reconfigure capabilities, and maintain mission effectiveness under uncertainty.

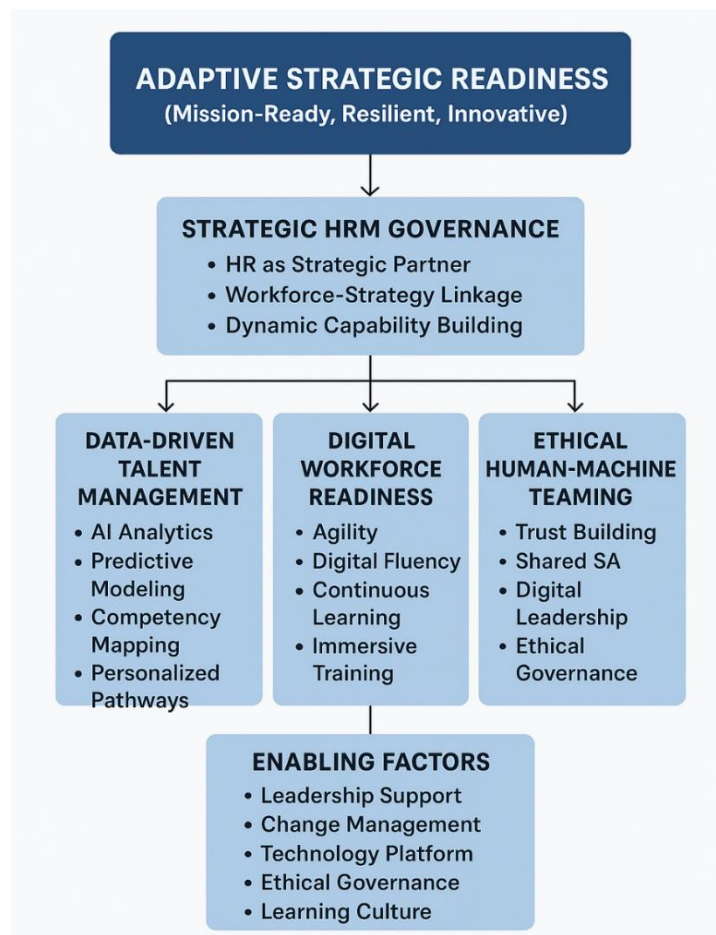


Figure 1. Strategic Defense HRM 4.0 Framework

Theoretical Contributions

This review makes several theoretical contributions to defense HRM literature. Our framework synthesizes:

1. Resource-Based View (RBV): Human capital and HRM systems as strategic resources Barney (1991)
2. Dynamic Capabilities Theory: HR's role in sensing, seizing, and transforming organizational capabilities Teece et al. (2016)
3. Socio-Technical Systems Theory: Interdependence of human and technological elements in defense organizations
4. Ethical Framework Theory: Moral considerations, accountability, and responsibility in AI-assisted decision-making (Floridi et al. (2018); Santoni de Sio & Mecacci)

This multi-theoretical approach provides richer understanding than single-theory perspectives. Our Strategic Defense HRM 4.0 framework differs from existing models:

a. Ulrich's HR Business Partner Model

While adopting Ulrich's strategic partnership concept, we add defense-specific elements: ethical HMT, security-cleared analytics, and integration with operational planning.

b. Corporate Digital HR Models:

Marler & Boudreau (2017) emphasize efficiency and employee experience. Our framework prioritizes mission readiness, ethical accountability, and adaptive capacity under adversarial conditions.

Conclusions

This systematic literature review synthesizes contemporary research to conclude that traditional defense human resource management (HRM) is inadequate for the strategic demands of the Fourth Industrial Revolution. The primary findings, aligned with the research objectives, are as follows: (1) The constructed Strategic Defense HRM 4.0 Framework, integrating four pillars—data-driven talent management, digital workforce readiness, ethical human-machine teaming, and strategic HRM for organizational readiness—provides a comprehensive model for transformation. (2) The framework's multi-theoretical foundation offers robust architecture for understanding this shift from administrative personnel functions to proactive, strategic workforce management. (3) Empirical analysis confirms that successful implementation necessitates concurrent advancements in technology, human capital, ethical governance, and strategic alignment.

The Strategic Defense HRM 4.0 Framework has direct applications for guiding policy modernization, technological investment, and capability development within defense organizations. For practitioners, it serves as a diagnostic tool and implementation roadmap. Future research should address the limitations of this review by investigating non-Western and non-English defense HRM adaptations, validating the framework through longitudinal case studies and quantitative testing, and exploring the specific socio-technical dynamics of human-machine teaming in classified or operational contexts. Further inquiry is also needed into effective change management strategies for overcoming institutional inertia in traditional defense bureaucracies during this transformation.

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