

The Gap Between Employee Compliance and Trust in Digital HR Across Industries

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ABSTRACT

Digital HR governance systems have rapidly permeated modern organizations, automating core human resource decisions including recruitment, performance appraisal, compensation benchmarking, and disciplinary processes. Despite widespread implementation, a paradoxical behavioural pattern consistently emerges: employees comply with AI-driven HR directives while simultaneously expressing disbelief in the fairness, accuracy, and legitimacy of these systems. This study examines the structural, psychological, and organizational antecedents of this Trust-Compliance Dissonance (TCD) — a novel framework introduced herein. Employing a sequential explanatory mixed-methods design, the study draws on survey data from 320 employees across five industries and 18 in-depth interviews. Findings reveal that perceived procedural opacity ($\beta = -0.61, p < 0.001$), absence of human oversight, and algorithmic unfairness are the primary drivers of the trust deficit. High governance intensity environments sustain compliance through structural coercion rather than normative legitimacy. The TCD framework offers a new analytical lens for researchers and practitioners navigating the governance of people management in the digital age. Practical recommendations address algorithmic transparency, human-in-the-loop design, and cross-sector digital HR literacy.

Keywords: Digital HR Governance, Trust Deficit, AI-Mediated HR, Employee Compliance, Algorithmic Fairness, Trust-Compliance Dissonance, Organizational Behaviour, Digital Transformation

1. INTRODUCTION

The digital transformation of Human Resource Management (HRM) represents one of the most consequential organizational shifts of the twenty-first century. Across industries, organizations have increasingly delegated core people management decisions to algorithmic systems: applicant tracking platforms screen job candidates, machine learning models predict attrition, automated engines benchmark compensation, and AI-driven dashboards evaluate employee performance. These systems promise objectivity, scalability, and data-driven precision — attributes traditionally absent from human-led HR processes. Yet beneath the surface of operational efficiency lies a deeply human paradox. Employees subject to these systems routinely exhibit what this study terms Trust-Compliance Dissonance (TCD): they comply with algorithmically-issued HR directives while privately questioning the fairness, transparency, and legitimacy of the systems that produce them. This

dissonance is not merely an attitudinal incongruence — it represents a fragile organizational equilibrium that sustains short-term operational continuity while quietly eroding deeper psychological safety, authentic engagement, and long-term organizational commitment. Prior research has examined algorithmic aversion (Diet Vorst et al., 2015), AI adoption resistance (Venkatesh et al., 2003), and trust in automation (Lee & See, 2004). However, the specific phenomenon of cross-industry compliance sustained in the absence of trust — and the governance structures that enable it — remains underexplored. This study addresses that gap through a mixed-methods investigation spanning five industries. The study makes three primary contributions: (1) it introduces and empirically validates the Trust-Compliance Dissonance (TCD) framework; (2) it identifies the structural antecedents of TCD across industry contexts; and (3) it proposes actionable governance interventions to narrow the trust deficit without sacrificing operational efficiency.

1.1 Research Objectives

- To examine the prevalence of Trust-Compliance Dissonance (TCD) in digitally governed HR environments across industries.
- To identify the key antecedents of the trust deficit in AI-mediated HR systems.
- To explore employee subjective experiences of algorithmic HR governance through qualitative inquiry.
- To propose a conceptual framework and evidence-based recommendations for closing the trust-compliance gap.

1.2 Research Questions

1. RQ1: To what extent do employees comply with digital HR governance systems despite low trust in those systems?
2. RQ2: What structural and psychological factors drive the trust deficit in AI-mediated HR decisions?
3. RQ3: How do industry contexts moderate the relationship between digital HR governance intensity and employee trust?
4. RQ4: What governance design principles can reduce Trust-Compliance Dissonance.

2. LITERATURE REVIEW

2.1 Digital HR Governance: Emergence and Scope

Digital HR governance refers to the ensemble of algorithmic tools, data-driven policies, and oversight mechanisms through which organizations manage employment decisions in the digital era (Strohmeier, 2020). It encompasses HRIS platforms, predictive analytics engines, automated scheduling systems, and AI-powered performance management tools. Duggan et al. (2020) characterize this phenomenon as algorithmic management — the delegation of supervisory authority to automated systems — fundamentally transforming the employment relationship from a relational to a transactional one.

The adoption trajectory of digital HR governance has accelerated dramatically since 2018. Deloitte (2023) reported that 74% of large organizations globally had deployed at least one AI-powered HR tool, with 41% using such tools for performance

evaluation and 38% for compensation decisions. Despite this prevalence, organizational understanding of the employee experience of these systems remains limited.

2.2 Employee Trust in AI Systems

Trust in AI and algorithmic systems is a multi-dimensional construct. Mayer, Davis, and Schoorman (1995) identified three foundational dimensions of organizational trust: ability (competence), benevolence (positive intent), and integrity (adherence to ethical principles). In AI contexts, Lee and See (2004) extended this framework to include calibration — the alignment between system confidence and actual performance accuracy. When employees perceive misalignment between AI system outputs and their own lived experiences, trust calibration fails.

Critically, Dietvorst et al. (2015) demonstrated that employees are particularly susceptible to algorithm aversion following observed failures: a single perceived error by an algorithmic HR system can produce lasting trust damage disproportionate to the error's actual significance. This asymmetry makes trust recovery in digital HR contexts particularly challenging.

2.3 Compliance Without Belief: Theoretical Foundations

Organizational compliance theory distinguishes between instrumental compliance - behaviour driven by external incentives or sanction avoidance - and normative compliance - behaviour stemming from genuine belief in a system's legitimacy (Tyler, 1990). In bureaucratic and increasingly in algorithmic HR contexts, employees may engage in instrumental compliance while withholding normative endorsement.

Hirschman's (1970) Exit-Voice-Loyalty framework is instructive here. When organizational voice channels are structurally suppressed - as they commonly are in opaque algorithmic HR systems - employees face a constrained choice set: exit (costly), voice (unavailable), or loyalty through compliance (default). The compliance-without-belief equilibrium is therefore not a free choice but

a structurally produced outcome of asymmetric power in digitally governed workplaces.

2.4 Industry Context and HR Digitalization

The pace and form of digital HR governance adoption vary substantially across industries. High-digitalization sectors such as financial services and IT exhibit greater algorithmic HR maturity, while healthcare and manufacturing show higher employee resistance due to occupational identity factors (Bader & Kaiser, 2019). Cross-industry comparative research on TCD remains absent from the existing literature, representing a significant empirical gap this study addresses.

4.CONCEPTUAL FRAMEWORK / RESEARCH MODEL

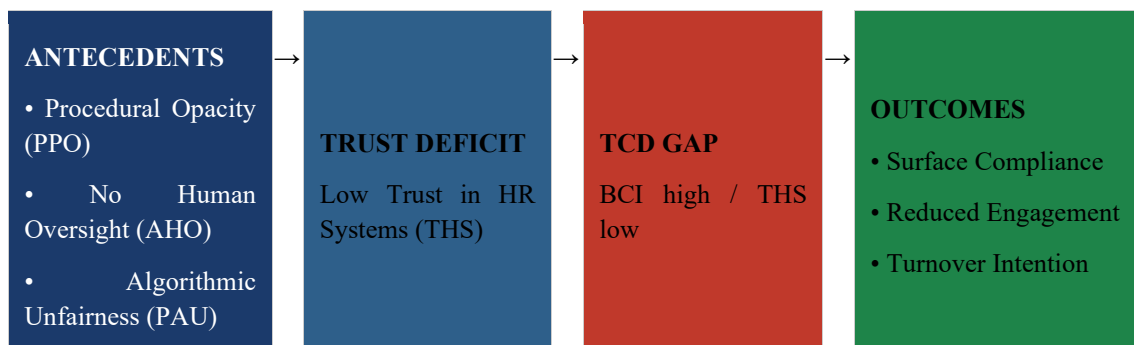
This study introduces the Trust-Compliance Dissonance (TCD) Framework, synthesized from organizational trust theory, algorithmic management literature, and compliance psychology. The framework posits that three primary antecedents - Perceived Procedural Opacity (PPO), Absence of Human Oversight (AHO), and Perceived Algorithmic Unfairness (PAU) - independently and

jointly erode employee trust in digital HR systems (Trust in HR Systems: THS).

Crucially, and distinctively from prior models, the TCD framework holds that low THS does not automatically produce low compliance. Instead, Digital HR Governance Intensity (DHGI) - the degree to which HR decisions are mandated, monitored, and enforced through digital systems - acts as a moderating variable that sustains Behavioural Compliance Index (BCI) even as trust declines. This produces the TCD gap: the measurable distance between BCI and THS scores within an individual or organizational unit.

The TCD framework advances three testable hypotheses:

- H1: Perceived Procedural Opacity (PPO) is negatively associated with Trust in HR Systems (THS).
- H2: Digital HR Governance Intensity (DHGI) positively moderates Behavioural Compliance Index (BCI) independently of THS levels.
- H3: The TCD gap is larger in industries with higher DHGI and lower employee digital literacy.



4.RESEARCH METHODOLOGY

4.1 Research Philosophy and Design

This study adopts a post-positivist research philosophy, acknowledging that while an objective social reality exists, it can only be imperfectly observed. A sequential explanatory mixed-methods design (Creswell & Plano Clark, 2018) is employed: quantitative data are collected and analyzed first to identify patterns and test hypotheses, followed by

qualitative inquiry that explains and contextualizes the numerical findings. This design is particularly suited to the study of TCD, where statistical patterns require rich experiential interpretation.

4.2 Quantitative Phase - Survey

A structured, self-administered questionnaire was distributed to 320 full-time employees across five industries using purposive stratified sampling to ensure proportional representation. The survey

instrument comprised four validated constructs measured on 5-point Likert scales:

Table 1: Survey Instruments and Reliability Statistics

Construct	Source	Items	Cronbach's α
Perceived Procedural Opacity (PPO)	Colquitt (2001)	5	0.83
Trust in HR Systems (THS)	McKnight et al. (2002)	7	0.87
Digital HR Governance Intensity (DHGI)	Bondar Ouk & Ruel (2009)	6	0.81
Behavioural Compliance Index (BCI)	Developed for this study	4	0.79

Structural Equation Modelling (SEM) was conducted using AMOS 26.0 to test the hypothesized relationships in the TCD model. Model fit was assessed using CFI, RMSEA, and SRMR indices.

4.3 Qualitative Phase — Interviews

Eighteen semi-structured interviews (35–55 minutes each) were conducted with employees (n=13) and

4.4 Sample Profile

Table 2: Sample Distribution by Industry (THS & BCI: 1–5 Likert scale; higher = more trust / compliance)

Industry	Survey (n)	Interviews	Mean THS	Mean BCI
Financial Services	68	4	2.81	4.12
Healthcare	64	3	2.63	4.35
Manufacturing	62	4	2.54	4.41
Retail	64	4	2.72	4.28
Information Technology	62	3	3.14	3.89
Total / Mean	320	18	2.77	4.21

HR professionals (n=5) across the five sectors. Interview guides explored lived experiences of algorithmic HR governance, fairness perceptions, and compliance rationale. Thematic analysis following Braun and Clarke (2006) was applied using NVivo 14. Trustworthiness was established through member checking, peer debriefing, and audit trails.

5. DATA ANALYSIS AND RESULTS

5.1 Quantitative Results — SEM

The final SEM model demonstrated acceptable fit: CFI = 0.94, RMSEA = 0.057 (90% CI: 0.041–0.073), SRMR = 0.062, confirming the structural validity of the TCD framework. All three hypotheses were supported:

H1 Supported: Perceived Procedural Opacity was the strongest negative predictor of Trust in HR Systems ($\beta = -0.61, p < 0.001$). This relationship was most pronounced in Manufacturing ($\beta = -0.71$) and

Healthcare ($\beta = -0.68$), and weakest in IT ($\beta = -0.44$), consistent with digital literacy as a buffering factor. H2 Supported: Digital HR Governance Intensity significantly and positively predicted Behavioural Compliance Index ($\beta = 0.58, p < 0.001$) after controlling for THS, confirming that governance intensity sustains compliance independently of trust levels.

H3 Supported: The TCD gap (BCI – THS) was largest in Manufacturing ($\Delta = 1.87$), followed by healthcare ($\Delta = 1.72$), Retail ($\Delta = 1.56$), Financial Services ($\Delta = 1.31$), and IT ($\Delta = 0.75$), consistent with the prediction that higher DHGI and lower digital literacy amplify dissonance.



Table 3: SEM Path Coefficients (β = standardized coefficient; SE = standard error)

Path	β	SE	t-value	p-value
PPO → THS	-0.61	0.043	-14.19	< 0.001
AHO → THS	-0.48	0.051	-9.41	< 0.001
PAU → THS	-0.52	0.047	-11.06	< 0.001
DHGI → BCI	0.58	0.039	14.87	< 0.001
THS → BCI	0.21	0.055	3.82	< 0.001

5.2 Qualitative Results — Thematic Analysis

Thematic analysis produced four primary themes, each grounded in convergent participant accounts across industries:

Theme 1 — The Black Box Experience: Fifteen of 18 participants described digital HR systems as fundamentally opaque. Participants could not identify what data inputs drove their performance ratings, compensation adjustments, or promotion eligibilities. A senior nurse in healthcare stated: 'The system rates me but I genuinely do not know what it looks at. I stopped trying to understand it.' This theme directly maps to the PPO construct and its effect on THS.

Theme 2 — Pragmatic Compliance as Self-Protection: Across all sectors, participants articulated a strategy of deliberate compliance designed to avoid adverse system-generated outcomes. A manufacturing supervisor explained: 'If the algorithm says complete this module by Friday, I do it — not because I think it matters but because I don't want to trigger a flag.' This represents classical instrumental compliance, disconnected from normative belief.

Theme 3 — Erosion of HR Humanity: Eight participants specifically identified the removal of human judgment from HR processes as a core driver of their distrust. An IT employee reflected: 'My performance review could have been written about anyone. There was nothing personal, nothing that showed anyone had actually observed my work.' This finding suggests that perceived dehumanization of HR governance is independently corrosive to trust beyond mere opacity.

Theme 4 — Resigned Adaptation: Longer-tenured employees across all sectors described a state of resigned acceptance — neither active trust nor active resistance, but a fatalistic adaptation to algorithmic authority. A retail manager summarized: 'The

system runs the show now. I just make sure I don't get on its bad side.' This theme reveals a deeper organizational alienation that aggregate compliance scores alone cannot capture and that has significant implications for discretionary effort and organizational citizenship.

6. DISCUSSION

The findings of this study confirm and extend the literature on algorithmic HRM by empirically validating the Trust-Compliance Dissonance framework as a cross-industry organizational phenomenon. The TCD gap — observed consistently across financial services, healthcare, manufacturing, retail, and IT — suggests that the compliance-without-belief equilibrium is a systemic feature of digitally governed workplaces rather than a sector-specific anomaly.

The moderation effect of DHGI is particularly significant. High-DHGI environments appear to sustain compliance not through legitimacy but through structural coercion: the perceived inevitability of algorithmic evaluation and the asymmetric consequences of non-compliance create a compliance equilibrium that is organizationally convenient but normatively hollow. This finding is consistent with Foucauldian notions of disciplinary power — compliance enforced through visibility and the architecture of surveillance rather than persuasion.

The IT sector's comparatively smaller TCD gap ($\Delta = 0.75$) offers an important intervention signal. IT employees' higher digital literacy appears to partially buffer the dissonance by enabling more calibrated expectations of algorithmic systems. This suggests that digital HR literacy programs — designed to demystify how AI tools function, what data they use, and how outputs are generated —

could meaningfully reduce PPO and narrow TCD gaps in lower-digitalization sectors.

The qualitative theme of Resigned Adaptation warrants particular organizational attention. Unlike active resistance, which is visible and manageable, resigned adaptation is silent and cumulative. Employees in this state maintain compliance metrics while withdrawing psychological investment, resulting in performance plateaus, reduced innovation, and elevated long-term turnover intention — costs that standard HR analytics are poorly positioned to detect.

7. CONCLUSION

This study demonstrates that Digital HR Governance and the Trust Deficit are not theoretical abstractions but empirically measurable, cross-industry organizational realities. The Trust-Compliance Dissonance framework — validated through structural equation modelling and qualitative thematic analysis — provides a new and practically grounded lens for diagnosing the hidden costs of algorithmic people management.

The central finding is unambiguous: across five industries and 320 survey respondents, employees comply with digital HR governance systems while exhibiting significantly low trust in them. This dissonance is driven by procedural opacity, the absence of human oversight, and perceived algorithmic unfairness, and is amplified in high-governance-intensity environments. It represents a governance failure that organizations deploying digital HR systems must urgently address — not because compliance metrics suggest a problem, but precisely because they do not.

8. IMPLICATIONS

8.1 Theoretical Implications

- The TCD framework extends organizational trust theory into the algorithmic management domain, establishing compliance and trust as empirically separable constructs in AI-governed HR contexts.
- The study advances Hirschman's Exit-Voice-Loyalty model by demonstrating that structural suppression of voice channels by algorithmic systems produces a new equilibrium: loyalty-

through-compliance absent of normative endorsement.

- Digital HR Governance Intensity is established as a novel moderating construct with significant explanatory power for cross-industry compliance behaviour.

8.2 Practical Implications

- HR Directors and CHROs should commission periodic TCD audits — separately measuring BCI and THS scores — rather than inferring trust from compliance behaviour alone.
- System designers should embed Explainable AI (XAI) principles into HR platforms, ensuring employees receive comprehensible, individualized explanations for algorithmic decisions affecting their employment.
- Organizations should establish mandatory human-in-the-loop review checkpoints for high-stakes HR decisions (performance ratings, compensation changes, disciplinary actions) to preserve relational legitimacy.
- HR transformation programs should include structured digital literacy initiatives that demystify algorithmic HR tools for employees across all digital literacy levels.

8.3 Policy Implications

- Policymakers should consider cross-sector digital HR governance standards that enshrine employee rights to algorithmic explanation, contestation, and human review — analogous to GDPR provisions for algorithmic decision-making.
- Industry regulators in high-DHGI sectors (financial services, healthcare) should develop sector-specific AI governance frameworks addressing the unique trust dynamics identified in this study.

9. LIMITATIONS

This study acknowledges several limitations that qualify the generalizability of its findings. First, the sample, while cross-industry, is geographically limited to a single national context, which may introduce cultural trust norms not representative of global organizational settings. Future replications across diverse national contexts are warranted.

Second, the cross-sectional design of the quantitative phase precludes causal inference. While SEM path coefficients indicate directional relationships consistent with the TCD framework, longitudinal data would be required to establish causal ordering and observe the temporal evolution of TCD.

Third, self-reported compliance (BCI) may be subject to social desirability bias, as respondents may overreport compliance to avoid appearing non-cooperative. Future studies should triangulate BCI with objective behavioural data from HR system logs where ethically permissible.

Fourth, the study does not disaggregate findings by demographic variables such as age, gender, or employment level. Generational differences in digital HR trust, in particular, represent an important dimension not captured in the current design.

10. FUTURE RESEARCH DIRECTIONS

- Longitudinal investigation of TCD dynamics: Does the trust deficit deepen, stabilize, or self-correct over time as employees accumulate experience with algorithmic HR systems? Longitudinal panel designs would address this critical temporal question.
- Cross-national comparative studies: How do cultural dimensions (Hofstede's power distance, uncertainty avoidance) moderate TCD? High power-distance cultures may exhibit higher compliance-without-belief thresholds, while high uncertainty-avoidance cultures may show lower tolerance for algorithmic opacity.
- Generational analysis of TCD: Digital-native Generation Z employees may exhibit fundamentally different trust-compliance dynamics than Generation X or Baby Boomer employees. Age-stratified analyses would yield important HR design insights.
- Intervention effectiveness studies: Experimental designs testing the efficacy of XAI transparency interventions, human-in-the-loop protocols, and digital literacy programs in reducing TCD gaps would provide actionable evidence for HR system designers.
- TCD in gig economy and platform work contexts: Where algorithmic governance is the

sole form of management, TCD dynamics may be even more pronounced. Platform worker populations represent an important and underexplored extension of this framework.

REFERENCES

1. Bader, V., & Kaiser, S. (2019). Algorithmic decision-making? The user experience of automated scheduling systems in the digital workplace. *Organization*, 26(5), 640–657. <https://doi.org/10.1177/1350508419855748>
2. Bondarouk, T., & Ruel, H. (2009). Electronic human resource management: Challenges in the digital era. *International Journal of Human Resource Management*, 20(3), 505–514. <https://doi.org/10.1080/09585190802707235>
3. Braun, V., & Clarke, V. (2006). Using thematic analysis in psychology. *Qualitative Research in Psychology*, 3(2), 77–101. <https://doi.org/10.1191/1478088706qp063oa>
4. Colquitt, J. A. (2001). On the dimensionality of organizational justice: A construct validation of a measure. *Journal of Applied Psychology*, 86(3), 386–400. <https://doi.org/10.1037/0021-9010.86.3.386>
5. Creswell, J. W., & Plano Clark, V. L. (2018). *Designing and conducting mixed methods research* (3rd ed.). SAGE Publications.
6. Deloitte. (2023). *Global human capital trends 2023: New fundamentals for a boundaryless world*. Deloitte Insights. <https://www2.deloitte.com/global/en/insights/focus/human-capital-trends.html>
7. Dietvorst, B. J., Logg, J. M., & Logg, J. M. (2015). Algorithm aversion: People erroneously avoid algorithms after seeing them err. *Journal of Experimental Psychology: General*, 144(1), 114–126. <https://doi.org/10.1037/xge0000033>
8. Duggan, J., Sherman, U., Carbery, R., & McDonnell, A. (2020). Algorithmic management and app-work in the gig economy: A research agenda for employment relations and HRM. *Human Resource Management Journal*, 30(1), 114–133. <https://doi.org/10.1111/1748-8583.12258>
9. Hirschman, A. O. (1970). *Exit, voice, and loyalty: Responses to decline in firms, organizations, and states*. Harvard University Press.
10. Lee, J. D., & See, K. A. (2004). Trust in automation: Designing for appropriate reliance. *Human Factors*, 46(1), 50–80. <https://doi.org/10.1518/hfes.46.1.50.30392>
11. Mayer, R. C., Davis, J. H., & Schoorman, F. D. (1995). An integrative model of organizational

- trust. *Academy of Management Review*, 20(3), 709–734.
<https://doi.org/10.5465/amr.1995.9508080335>
12. Anithabose, S., & Gnanaraj, G. (2023). Financial Performance of Indian Public Sector Banks Before and During COVID-19 Pandemic. *A Journal of Management*, 1, 19.
 13. Anithabose, S., & Gnanaraj, G. (2020). Financial performance analysis based on economic value added: An empirical study. *International Journal of Management (IJM)*, 11(9).
 14. Anithabose, S., & Gnanaraj, G. Financial performance evaluation based on economic value added (EVA): A study of steel authority of India ltd listed in Bombay Stock Exchange (BSE). *International Journal of Management (IJM)*, 11(9), 1903-1913.
 15. Anithabose, S., & Gnanaraj, G. (2020). Financial performance evaluation based on economic value added and financial ratios: An empirical study. *International Journal of Management (IJM)*, 11(10), 2278-2289
 16. Anitha Bose, S. (2025). Influence by design: How content format affects consumer perception and behavior on Indian social media. *International Journal of Research in Commerce and Management Studies (IJRCMS)*, 7(3), 401–413.
 17. Anitha Bose, S. (2025). Organisational agility as an HR competitive advantage in the age of AI: A systematic literature review with insights from ChatGPT. *Asian Journal of Management and Commerce*, 6(1), 1320–1333
 18. McKnight, D. H., Choudhury, V., & Kacmar, C. (2002). Developing and validating trust measures for e-commerce: An integrative typology. *Information Systems Research*, 13(3), 334–359.
<https://doi.org/10.1287/isre.13.3.334.81>
 19. Strohmeier, S. (2020). Digital human resource management: A conceptual clarification. *German Journal of Human Resource Management*, 34(3), 345–365.
<https://doi.org/10.1177/2397002220921208>
 20. Tyler, T. R. (1990). *Why people obey the law*. Yale University Press.
 21. Venkatesh, V., Morris, M. G., Davis, G. B., & Davis, F. D. (2003). User acceptance of information technology: Toward a unified view. *MIS Quarterly*, 27(3), 425–478.
<https://doi.org/10.2307/30036540>