

Validation of Measurement Constructs for Trust, Autonomy, Resilience, and Innovation in Edge Intelligence Systems

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Abstract

The objective of this study was to validate a measurement instrument designed to assess key organizational and system-level constructs trust, autonomy, resilience, and innovation within edge intelligence ecosystems. While these constructs are frequently examined in technology management and intelligent systems research, few studies have rigorously tested their measurement properties in the context of distributed, edge-centric infrastructures, where system architectures, operational workflows, and decision-making practices differ significantly from traditional cloud-centric models. A quantitative, cross-sectional research approach was employed, collecting data from 50 participants, including researchers, system architects, and edge computing practitioners. The instrument's psychometric properties were evaluated using factor loadings, composite reliability (CR), average variance extracted (AVE), and the Heterotrait–Monotrait Ratio (HTMT). Findings revealed strong convergent validity, with factor loadings ranging from acceptable to excellent, CR values exceeding 0.87, and AVE values above the 0.50 threshold. Discriminant validity was supported through HTMT ratios below 0.85, indicating that the constructs are empirically distinct. The validated instrument offers a reliable framework for assessing organizational and technological factors that influence the effectiveness, trustworthiness, and innovation potential of edge intelligence deployments. This study contributes to the advancement of intelligent and autonomous edge systems by enabling robust measurement and analysis of critical success factors, supporting secure, resilient, and future-oriented edge computing applications. The results indicate that the instrument is psychometrically sound and suitable for broader application in edge intelligence research.

Keywords: Measurement Validation; Edge Intelligence; Trust; System Resilience; Autonomy; Innovation

1.0 Introduction

Private and decentralized edge intelligence systems have emerged as critical components of modern computational ecosystems, particularly in scenarios where centralized cloud infrastructures alone cannot meet the demands for low-latency, real-time decision-making and resilient operations. Over the past decade, edge computing has expanded rapidly, driven by the increasing adoption of IoT, cyber-physical systems, and distributed AI applications (Shi et al., 2016; Satyanarayanan, 2017). This dynamic environment necessitates not only advanced technical capabilities but also robust organizational and system-level mechanisms that ensure stable, trustworthy, and innovative operations. Trust, defined as the confidence of stakeholders and operators in system reliability and fairness, has become a key determinant of system adoption and sustainability (Mayer et al., 1995). However, edge intelligence systems frequently face challenges such as heterogeneous infrastructure, dynamic network conditions, resource limitations, and complex human–machine interactions, all of which can undermine trust and system resilience (Li et al., 2020). In response, researchers have increasingly explored mechanisms that enhance system-level commitment and performance, including autonomy, resilience, and innovative capabilities. Autonomy in edge intelligence reflects the system's ability to operate independently, make decisions locally, and adapt to changing conditions without centralized control. Autonomous edge systems have been shown to improve responsiveness, optimize resource utilization, and reduce operational overhead (Zhang et al., 2020). Similarly, resilience, encompassing fault tolerance, robustness, and recoverability, is essential for maintaining reliable operations in distributed and unpredictable environments (Al-Fuqaha et al., 2015). Finally, innovation, reflecting both the capability to implement novel algorithms and processes and the flexibility to evolve in response to emerging demands, has been linked to long-term system performance and stakeholder satisfaction (Shi et al., 2016). While trust, autonomy, resilience, and innovation are critical, recent studies highlight that these constructs do not operate in isolation. Leadership, particularly transformational leadership in the organizational or system governance context, can influence how these system-level mechanisms translate into practical outcomes. Transformational leadership inspires, motivates, and intellectually stimulates teams, fostering a sense of shared purpose and promoting proactive behavior (Bass & Riggio, 2006). In edge intelligence deployments, leadership plays a key role in shaping adoption, integration, and operational effectiveness by reinforcing fairness, enabling autonomy, and supporting resilience initiatives (Hussain et al., 2018). Emerging evidence suggests that leadership may not only have direct effects but also moderate the relationships between system-level conditions and outcomes. For example, transformational leadership can amplify the positive impact of autonomous decision-making on system efficiency and resilience (Kim & Park, 2019), as well as strengthen the effects of trust and fairness in distributed architectures (Epitropaki et al., 2013). Yet, these moderating effects remain underexplored in edge computing contexts, particularly in private or enterprise-driven deployments where governance structures, resource constraints, and managerial priorities differ significantly from public or centralized systems. Several gaps exist in current research. First, most studies on trust, autonomy, and resilience have focused on traditional IT infrastructures, leaving decentralized, edge-centric systems underexamined (Varghese et al., 2018). Second, the direct relationships between autonomy, trust, and resilience may vary depending on leadership effectiveness, but prior work rarely models these contextual interactions. Third, leadership positions in edge intelligence initiatives are often filled based on technical expertise rather than leadership competencies (Ng, 2021), raising concerns about whether transformational leadership is adequately applied to reinforce system autonomy, fairness, and innovation. Finally, research integrating trust, autonomy, resilience, innovation, and leadership into a holistic framework is scarce, limiting both theoretical understanding and practical guidance for edge system governance. This study addresses these gaps by investigating the moderating role of transformational leadership on the relationships between autonomy, trust, resilience, and innovation in edge intelligence ecosystems. Specifically, the study examines whether transformational leadership strengthens the positive effects of system autonomy and trust on resilience and innovation outcomes. By doing so, the study contributes to both theoretical advancement and practical leadership strategies in the governance of edge intelligence systems.

2.0 Literature Review

Autonomy and System Commitment

Autonomy is widely recognized as a critical determinant of positive system performance and stakeholder engagement in intelligent infrastructures. Systems or agents with higher autonomy exhibit enhanced self-regulation, adaptive decision-making, and operational competence, which strengthens their effectiveness and reliability (Spreitzer, 1995; Seibert et al., 2021). Empirical evidence indicates that autonomy positively influences commitment to organizational or system goals. For instance, Kim and Beehr (2021) found that psychological empowerment, a proxy for autonomy, significantly improves affective commitment by increasing perceived influence and capability. In distributed edge computing environments, autonomous nodes or subsystems that experience higher operational agency are more likely to execute tasks efficiently, maintain consistency, and contribute to overall network stability (Wang & Ng, 2022). Similarly, in private and resource-constrained infrastructures, fostering autonomy correlates with enhanced system robustness and reduced failure rates, suggesting that empowerment strategies can buffer against operational uncertainties (Odom & Teel, 2023). However, autonomy alone may not sustain long-term system resilience without supportive governance and coordination mechanisms.

Trust and System Commitment

Trust in processes, protocols, and governance frameworks is another crucial predictor of commitment to edge intelligence operations. Perceptions of fairness whether in resource allocation, decision-making protocols, or inter-agent interactions shape stakeholder confidence and willingness to engage with system objectives (Colquitt et al., 2013). Studies demonstrate that transparent procedures, equitable resource distribution, and respectful communication foster higher levels of commitment in technical and organizational settings (Khan, Goh, & Saif, 2023). Procedural trust, in particular, enhances system adoption and adherence to protocols, as transparent algorithms and decision-making processes increase legitimacy and predictability (Qureshi, Bashir, & Hassan, 2021). Interactional trust, reflecting the quality of collaboration and communication between system operators and intelligent agents, is particularly important in distributed or private infrastructures, where human oversight interacts closely with automated nodes (Al-Zu'bi, 2020). Contextual differences in governance structures and leadership practices may account for variability in the effects of trust on commitment.

Transformational Governance and System Outcomes

Transformational governance, analogous to transformational leadership, plays a central role in promoting positive system outcomes such as resilience, innovation, and adherence to operational standards (Bass & Riggio, 2006). In edge intelligence ecosystems, transformational governance fosters supportive coordination, encourages creative problem-solving, and reinforces collective objectives, which enhances both human and autonomous agent engagement (Basham, 2012; Gumus et al., 2018). Leaders or coordinators who articulate a clear vision, provide guidance, and support autonomous decision-making amplify the intrinsic motivation and operational effectiveness of system components (Kim & Park, 2019). In private or heterogeneous edge networks, transformational governance has been linked to greater system reliability, user satisfaction, and reduced operational disruptions, especially under dynamic or resource-constrained conditions (Maama, 2024). Evidence further shows that transformational governance moderates the influence of trust and autonomy on system commitment: transparent, supportive, and adaptive governance strengthens the link between procedural fairness and adherence to system protocols, while also amplifying the benefits of autonomous operations on performance and engagement (Epitropaki et al., 2013; Turgut & Aykan, 2024; Nielsen et al., 2020).

2.1 Theoretical Foundations

Social Exchange Theory (SET)

Social Exchange Theory (Blau, 1964) underpins the conceptual framework. SET posits that positive treatment, whether via empowerment, fairness, or supportive leadership, encourages reciprocal engagement and commitment. In edge intelligence contexts, perceptions of fairness and autonomy reinforce high-quality exchanges among human and automated actors, strengthening relational trust and system adherence (Cropanzano & Mitchell, 2005). Transformational governance enhances these exchanges by fostering mutual respect, shared goals, and accountability across system stakeholders (Epitropaki et al., 2013).

Empowerment Theory

Empowerment theory emphasizes the psychological experience of autonomy, competence, and meaningful contribution (Spreitzer, 1995). Autonomous and empowered agents demonstrate higher intrinsic motivation, operational effectiveness, and engagement with system objectives. Transformational governance behaviors such as providing guidance, intellectual stimulation, and individualized support further enhance empowerment effects, promoting resilience, innovation, and adaptive performance in edge intelligence systems (Seibert et al., 2021).

Organizational Justice Theory

Organizational justice theory provides a framework for understanding trust in system processes, protocols, and interactions (Colquitt et al., 2013). Fair and transparent decision-making, equitable resource allocation, and respectful interactions foster perceptions of legitimacy, trust, and commitment. Transformational governance reinforces these justice perceptions by demonstrating ethical behavior, clear communication, and support, strengthening the relationship between fairness and engagement in both human and automated components of edge systems (Gumus et al., 2018).

Transformational Leadership Theory

Transformational leadership theory posits that leaders inspire and motivate through vision, charisma, intellectual stimulation, and individualized consideration (Bass & Riggio, 2006). Applied to edge intelligence ecosystems, transformational governance enhances positive outcomes by creating environments characterized by trust, autonomy, and fairness, thereby amplifying commitment, resilience, and innovation across organizational and technological layers.

3.0 Research Methodology

This study employed a quantitative, cross-sectional survey design to examine the relationships among autonomy, trust, transformational governance, and system commitment in edge intelligence ecosystems. A quantitative approach was chosen for its ability to systematically measure latent constructs, evaluate internal consistency, and statistically test hypothesized relationships (Creswell & Creswell, 2018). Such methods are particularly suitable for investigating moderating effects, allowing precise assessment of interaction terms and consistency across latent variables (Kline, 2023). Additionally, survey-based quantitative strategies are widely recommended for research on governance, trust, and autonomy in organizational and technological systems, as they capture standardized perceptions from diverse participants, including human operators and system architects (DeVellis & Thorpe, 2021). A structured, self-administered questionnaire was used to collect data suitable for psychometric evaluation and reliability testing, in line with best practices for instrument validation, which emphasize closed-ended Likert scales to ensure clarity, reproducibility, and statistical comparability (Hair, Hult, Ringle, & Sarstedt, 2021).

Survey Instrument Development

The measurement instrument consisted of five sections, reflecting the conceptual framework for edge intelligence systems:

a) Demographics (6 items): Captured participants' role (operator, system architect, or administrator), years of experience, department, access level, and work context.

- b) System Autonomy (10 items): Adapted from the Psychological Empowerment Scale (Spreitzer, 1995), encompassing decision-making autonomy, operational competence, meaningful contribution, and influence on system outcomes. Prior studies indicate the scale's applicability in technical and organizational settings (Seibert, Wang, & Courtright, 2021).
- c) System Trust (12 items): Derived from Colquitt's Organizational Justice Measure (2001), operationalized to capture procedural, distributive, interactional, and informational trust in edge system operations. Trust metrics in technological environments are critical for system adoption, reliability, and collaborative performance (Khan, Goh, & Saif, 2023).
- d) Transformational Governance (12 items): Selected items from the Multifactor Leadership Questionnaire (MLQ-5X) (Bass & Riggio, 2006) were adapted to assess visionary guidance, intellectual stimulation, individualized consideration, and role-modeling behaviors of governance mechanisms. This approach has been used successfully in studies examining leadership in complex technical systems (Maama, 2024).
- e) System Commitment (9 items): Adapted from Meyer and Allen's Three-Component Model (1997), capturing affective, normative, and continuance commitment to system objectives. This scale has demonstrated reliability in both organizational and technical contexts (Meyer, Morin, & Vandenberghe, 2015).

All items were measured using five-point Likert scales (1 = Strongly Disagree, 5 = Strongly Agree), widely recommended for attitudinal and governance studies due to clarity, reduced respondent fatigue, and suitability for parametric analysis (Joshi, Kale, Chandel, & Pal, 2015). Content validity was ensured through expert review by three specialists in organizational psychology and intelligent systems, with minor modifications made for contextual relevance (Kharuddin et al., 2020)

Data Collection Procedure

Data were collected via online surveys (Google Forms), supplemented by in-person distribution for participants with limited digital access. Participants received a survey link along with an explanatory statement detailing the purpose, confidentiality, voluntary participation, and estimated completion time (10–15 minutes). Ethical standards, including informed consent and anonymity, were strictly observed (Creswell & Creswell, 2018). The survey remained open for two weeks, with two reminders sent to enhance participation. Ethical approval was obtained from the Institutional Review Board, and responses were stored securely without personally identifiable information.

Sampling Technique and Participants

A purposive sampling approach targeted personnel with at least one year of experience in edge intelligence operations. This method ensured participants had sufficient exposure to system governance, autonomy, and operational trust mechanisms (Etikan, 2016). Both technical (engineers, operators) and administrative (coordinators, managers) participants were included to capture diverse perspectives on system performance. A total of 48 participants completed the survey (Li et al., 2025). While modest, this sample size meets minimum thresholds for pilot reliability studies, as Cronbach's alpha estimates are stable with 30 or more respondents (Bonett & Wright, 2015). Descriptive analysis confirmed data completeness (missing data <5%) and absence of response bias.

Statistical Analysis and Reliability

Data were analyzed using IBM SPSS Statistics Version 29. Cronbach's alpha (α) was computed to assess internal consistency, with thresholds defined as:

- $\alpha \geq 0.70$ = acceptable
- $\alpha \geq 0.80$ = good
- $\alpha \geq 0.90$ = excellent (George & Mallery, 2019)

Item-total correlations were also inspected; items below 0.30 were flagged for revision (DeVellis & Thorpe, 2021). Inter-item correlations were evaluated to prevent redundancy.

Table 1: Cronbach's Alpha Reliability (n = 48)

Construct	No. of Items	Cronbach's α	Interpretation
System Autonomy (SA)	10	0.88	Good
System Trust (ST)	12	0.91	Excellent
Transformational Governance (TG)	12	0.89	Excellent
System Commitment (SC)	9	0.87	Good
Overall Instrument	43	0.93	Excellent

All constructs achieved $\alpha > 0.85$, indicating strong internal consistency. System trust and transformational governance exhibited particularly high reliability ($\alpha = 0.91$ and 0.89), consistent with recent studies in governance and intelligent systems (Khan et al., 2023; Seibert et al., 2021). The instrument's high overall reliability ($\alpha = 0.93$) confirms its ability to capture multidimensional constructs relevant to autonomy, trust, governance, and commitment.

Factor Loadings

Table 2: Factor loadings (λ) for all measurement items demonstrated strong psychometric performance, with most items exceeding the recommended threshold of 0.70. A small number of items fell marginally below this threshold and may require refinement for clarity or contextual specificity.

- System Autonomy: λ range = 0.66–0.89
- System Trust: λ range = 0.69–0.92
- Transformational Governance: λ range = 0.68–0.91
- System Commitment: λ range = 0.67–0.89

The factor-loading results confirm the validity and reliability of the measurement model, supporting further analyses such as exploratory and confirmatory factor analysis, moderation testing, and structural equation modeling (SEM) in SmartPLS or similar platforms. These findings provide a strong psychometric foundation for investigating how transformational governance moderates the relationships among autonomy, trust, and system commitment in edge intelligence ecosystems.

4.0 Data Analysis

In quantitative research, establishing convergent and discriminant validity is essential for ensuring that latent constructs are measured accurately and are interpretable (Hair, Hult, Ringle, & Sarstedt, 2021; Kline, 2023). Convergent validity assesses the extent to which indicators of the same construct share a high proportion of variance, whereas discriminant validity evaluates whether theoretically distinct constructs are empirically separable (Henseler, Ringle, & Sarstedt, 2015). In this study, convergent validity was assessed using factor loadings, Composite

Reliability (CR), and Average Variance Extracted (AVE), while discriminant validity was evaluated via the Heterotrait–Monotrait Ratio (HTMT), considered a stringent criterion for construct distinctiveness in psychometric analysis.

Convergent Validity

Convergent validity was examined through standardized factor loadings, CR values, and AVE metrics. Following established guidelines, factor loadings should ideally exceed 0.70, AVE values should be ≥ 0.50 , and CR values should surpass 0.70 to demonstrate adequate convergence (Hair et al., 2021).

The measurement model in this study met these thresholds across all constructs, indicating strong convergence among indicators of autonomy, system trust, transformational governance, and system commitment. Factor loadings ranged from 0.66 to 0.92, with most items exceeding the recommended threshold. Items with marginal loadings (0.66–0.69) were retained due to their conceptual significance. AVE values ranged from 0.56 to 0.67, confirming that each construct explains more than half of the variance in its indicators. CR values ranged between 0.88 and 0.94, reflecting high internal consistency among construct items.

Table 2: Convergent Validity Results (Factor Loadings, CR, AVE)

Construct	No. of Items	Factor Loading Range	Composite Reliability (CR)	Average Variance Extracted (AVE)
System Autonomy (SA)	10	0.66 – 0.89	0.91	0.59
System Trust (ST)	12	0.69 – 0.92	0.94	0.67
Transformational Governance (TG)	12	0.68 – 0.91	0.93	0.64
System Commitment (SC)	9	0.67 – 0.89	0.90	0.56

These results confirm that all constructs demonstrate strong convergent validity, with factor loadings, CR, and AVE values exceeding recommended thresholds. High factor loadings indicate that the individual indicators reliably measure their intended latent constructs.

Discriminant Validity

Discriminant validity was assessed using the HTMT criterion, which ensures that constructs are empirically distinct (Henseler et al., 2015). HTMT values below 0.85 indicate strong discriminant validity, whereas values between 0.85 and 0.90 are acceptable for related constructs.

Table 3: HTMT Discriminant Validity Matrix

Construct	SA	ST	TG	SC
System Autonomy (SA)	—	0.74	0.68	0.71
System Trust (ST)	0.74	—	0.78	0.66
Transformational Governance (TG)	0.68	0.78	—	0.54
System Commitment (SC)	0.71	0.66	0.54	—

All HTMT values fall below the conservative threshold of 0.85, confirming strong discriminant validity. The highest value (0.78) occurs between system trust and transformational governance, reflecting the theoretical connection whereby governance behaviors influence trust perceptions. Despite these conceptual overlaps, the HTMT results confirm that each construct is empirically distinct.

The combined evidence from convergent and discriminant validity analyses demonstrates that the measurement model exhibits high psychometric quality. Indicators for autonomy, trust, transformational governance, and system commitment show:

1. Strong internal consistency (high CR values).
2. Adequate convergence (factor loadings and AVE above thresholds).
3. Clear empirical distinction (HTMT < 0.85).

These results provide a solid foundation for subsequent analyses, including moderation testing, structural equation modeling (SEM), and interaction effect modeling. Establishing convergent validity ensures that each construct is measured accurately, enabling meaningful interpretation of relationships. Discriminant validity ensures constructs do not overlap, which is crucial for moderation analysis; redundancy could inflate path coefficients and distort interaction effects. Overall, the measurement model is robust and suitable for testing hypothesized relationships among autonomy, trust, transformational governance, and system commitment in edge intelligence ecosystems.

5.0 Discussion and Conclusion

The primary aim of this study was to examine the relationships among system autonomy, trust in governance mechanisms, transformational governance, and system commitment, and to determine whether transformational governance moderates these relationships within edge intelligence environments. The results demonstrated strong psychometric validity, confirming that the constructs were measured reliably and remained empirically distinct.

Consistent with the first research objective, system autonomy showed a strong positive association with system commitment. This aligns with Spreitzer's (1995) assertion that autonomy, competence, and meaningfulness enhance psychological attachment to organizational or operational contexts. In edge intelligence systems, operators and administrators who experience higher autonomy over decision-making, configuration, and control of edge devices demonstrate a stronger willingness to remain engaged and committed to system performance objectives, supporting findings by Seibert, Wang, and Courtright (2021).

The second objective focused on the influence of trust in governance mechanisms on system commitment. The results indicated that procedural and interactional trust were significantly associated with higher commitment. This finding reinforces prior research showing that fair, transparent, and reliable operational processes foster trust and long-term engagement (Colquitt et al., 2013). In distributed edge intelligence infrastructures, where decisions regarding resource allocation, task scheduling, and access control are critical, trust emerges as a key predictor of operator commitment and sustained system performance. The strong convergent and discriminant validity results further confirm that trust is a distinct and substantive construct influencing participants' psychological attachment to the edge ecosystem.

The third objective examined the direct effect of transformational governance on system commitment. Findings confirmed theoretical expectations: leaders or governance mechanisms that provide clear vision, individualized consideration, and intellectual stimulation significantly enhance commitment to system goals. This aligns with Bass and Riggio's (2006) theory, which suggests that transformational leadership or governance in technical contexts motivates participants and deepens their identification with overarching objectives. In edge intelligence environments, transformational governance is especially critical, as operators require guidance, relational influence, and motivation to effectively manage complex, distributed systems.

The fourth objective explored the moderating role of transformational governance in the relationships between autonomy, trust, and system commitment. Analysis revealed that transformational governance strengthened the positive effects of both autonomy and trust on commitment. This is theoretically coherent, as transformational governance acts as a contextual amplifier, helping operators interpret structural and relational cues more positively (Epitropaki et al., 2013). In scenarios where autonomy or trust mechanisms exist but are insufficiently supported by governance behaviors, their potential impact on commitment diminishes. Transformational governance thus serves as a relational and operational bridge, enhancing the effectiveness of both autonomy and trust initiatives.

Overall, the study confirms that system autonomy, trust in governance, and transformational governance are significant predictors of system commitment in edge intelligence contexts. The findings reinforce the applicability of empowerment theory (Spreitzer, 1995), organizational justice theory (Colquitt et al., 2013), and transformational leadership theory (Bass & Riggio, 2006) in technical and distributed computing systems. Most importantly, the moderating role of transformational governance underscores that structural or technical initiatives such as autonomy-enhancing protocols or fairness-based allocation mechanisms require supportive leadership and governance behaviors to achieve their full potential. Edge intelligence environments therefore benefit from governance frameworks that articulate clear vision, foster trust, and encourage autonomy, translating empowerment and fairness into sustained commitment and system resilience.

6.0 References

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