



Growth-Oriented Culture within VFX Teams: Implications for High-Quality Effects and Simulation Innovation

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Abstract:

The increasing complexity of simulation-driven workflows in visual effects (VFX) production necessitates not only advanced computational tools but also adaptive team environments that support continuous learning and innovation. This study investigates the implications of growth-oriented culture within VFX teams for achieving high-quality effects output and simulation innovation. By integrating cultural dimensions such as adaptive learning orientation, collaborative problem-solving, feedback receptivity, and experimentation tolerance with performance-based simulation metrics, the research examines how organizational mindset influences workflow efficiency and procedural advancement. A mixed-method analytical framework incorporating regression analysis, correlation modeling, cluster analysis, and canonical correlation analysis was employed to evaluate the relationships between growth-oriented cultural attributes and innovation-related outcomes. The findings reveal that growth-oriented teams demonstrate significantly higher simulation accuracy, improved iteration success rates, and greater adoption of procedural tools compared to low-growth environments. Furthermore, innovation-supportive climates were found to enhance pipeline flexibility and simulation stability, contributing to more efficient and scalable production processes. The study underscores the importance of fostering growth-oriented cultural frameworks as a strategic determinant of both technical excellence and innovation capacity in simulation-intensive VFX production settings.

1. Introduction

1.1 The growing strategic importance of culture in performance-driven VFX environments

The visual effects (VFX) industry has emerged as one of the most technically demanding and innovation-driven domains within the broader digital production ecosystem, where the pursuit of photorealism, simulation accuracy, and computational creativity defines both artistic success and commercial viability. As VFX pipelines become increasingly reliant on advanced simulation engines, procedural animation, particle dynamics, and real-time rendering technologies, the human dimension of production particularly team culture has begun to exert a profound influence on output quality and innovation potential (Martinidis, 2017). Beyond hardware capabilities and software proficiency, the ability of VFX teams to consistently produce high-quality effects depends

on a shared organizational mindset that fosters continuous learning, adaptive problem-solving, and resilience in the face of complex production challenges (Wallis, 2021). In this context, the concept of a growth-oriented culture has gained prominence as a critical determinant of team effectiveness, shaping how individuals respond to technological disruption, creative iteration, and performance feedback within high-stakes production environments (Hossain, 2023).

1.2 The evolving relationship between collaboration and simulation complexity

Modern VFX production is inherently collaborative, involving interdisciplinary integration between technical directors, simulation artists, composers, lighting specialists, and pipeline engineers (Kavakli & Cremona, 2022). The increasing complexity of simulations ranging from fluid dynamics and rigid body physics to

volumetric rendering and environmental interaction requires not only domain-specific expertise but also a culture that encourages knowledge exchange and iterative experimentation (Korkut & Surer, 2023). Teams operating within growth-oriented environments are more likely to engage in constructive critique, shared debugging, and cross-functional problem-solving, thereby accelerating the refinement of simulation outputs and reducing production bottlenecks. Such environments support risk-tolerant workflows, where creative exploration is not penalized but strategically aligned with performance objectives, enabling artists to test novel simulation parameters, explore alternative rendering techniques, and integrate machine-learning-assisted effects into traditional production pipelines (Guo et al., 2022).

1.3 The role of adaptive learning in achieving high-quality effects output

Achieving realism in contemporary VFX simulations often necessitates repeated cycles of modeling, testing, error correction, and aesthetic evaluation. This iterative process places cognitive and creative demands on artists who must balance technical precision with visual storytelling goals (Halperin & Lukin, 2023). A growth-oriented culture facilitates adaptive learning by encouraging feedback-driven skill enhancement, collaborative troubleshooting, and reflective practice within teams. By normalizing failure as an integral component of innovation, such cultural frameworks enable artists and engineers to push the boundaries of simulation fidelity without fear of reputational risk or production delay (Marvel et al., 2020). The resulting emphasis on continuous improvement enhances both individual competencies and collective performance, leading to more stable simulation architectures and visually coherent effects that align with evolving industry benchmarks (Pop et al., 2023).

1.4 The influence of innovation climate on pipeline efficiency and experimentation

Pipeline efficiency in VFX production is not solely determined by software architecture or automation capabilities but also by the cultural norms that govern experimentation and process optimization. Teams characterized by growth-oriented values tend to exhibit greater openness to workflow redesign, scripting enhancements, and procedural tool development.

This openness fosters incremental innovation within production pipelines, enabling teams to reduce render times, improve asset interoperability,

and optimize simulation scalability across projects (Nwulu et al., 2022). Furthermore, an innovation-supportive climate enhances the integration of emerging technologies such as GPU-accelerated simulations, physics-based rendering, and generative procedural modeling, thereby strengthening the capacity of teams to adapt to shifting production demands and creative expectations (Juliani et al., 2018; Zamorano et al., 2023).

1.5 The implications of mindset-driven collaboration for simulation innovation

Simulation innovation within VFX teams often arises from the convergence of technical insight and creative experimentation. A growth-oriented mindset encourages team members to view challenges as opportunities for development, promoting proactive engagement with complex simulation tasks and novel rendering frameworks (DeCoito & Briona, 2023). This mindset supports collective ownership of production outcomes, where success is defined not only by individual expertise but by the team's ability to learn, adapt, and innovate collaboratively (Paxton & Van Stralen, 2015). As simulation environments become more dynamic and computationally intensive, the presence of a culture that values learning agility and interdisciplinary collaboration becomes essential for sustaining both artistic excellence and technical innovation.

1.6 The need for empirical investigation into culture-performance dynamics in VFX teams

Despite the recognized importance of organizational culture in knowledge-intensive industries, empirical research examining its impact on simulation quality and innovation within VFX teams remains limited (Torkkeli et al., 2019). Understanding how growth-oriented cultural attributes influence workflow efficiency, simulation accuracy, and creative experimentation is essential for developing performance-enhancing management strategies within digital production environments.

This study seeks to address this gap by exploring the implications of growth-oriented team culture for high-quality effects generation and simulation innovation, thereby contributing to a more comprehensive understanding of performance dynamics in technologically mediated creative teams.

2. Methodology

2.1 The adoption of a mixed-method research design for examining culture–performance dynamics

This study employed a mixed-method research design to examine the influence of growth-oriented culture within VFX teams on high-quality effects generation and simulation innovation. The methodological framework integrated both quantitative and qualitative data to capture the multidimensional interactions between team mindset, technical workflow behavior, and innovation outcomes in simulation-based production environments. The quantitative component focused on measuring perceptual, behavioral, and performance-based variables across VFX professionals engaged in simulation-intensive tasks, while the qualitative component facilitated contextual validation through structured expert inputs. This integrated approach enabled the identification of statistically significant relationships between organizational culture attributes and production-level innovation performance.

2.2 The operationalization of growth-oriented culture and innovation-related constructs

Growth-oriented culture was operationalized as the primary independent variable, comprising measurable dimensions such as adaptive learning orientation, collaborative problem-solving, feedback receptivity, experimentation tolerance, and skill development engagement. Dependent variables included high-quality effects output and simulation innovation capacity, represented through parameters such as simulation accuracy index, render optimization efficiency, visual coherence score, iteration success rate, and procedural tool adoption frequency. Additionally, mediating variables such as innovation climate, pipeline flexibility, and cross-functional knowledge exchange were incorporated to understand indirect cultural impacts on performance. Control variables including team experience level, software proficiency index, project complexity scale, and production cycle duration were also integrated into the model to isolate cultural effects from technical or operational confounders.

2.3 The use of structured measurement instruments for performance assessment

Data were collected using a structured questionnaire instrument based on a five-point Likert scale ranging from strongly disagree to strongly agree for perceptual and behavioral

constructs. Performance-related indicators were measured using normalized project-based metrics derived from simulation workflows, including average simulation convergence time, particle interaction stability score, environment interaction fidelity index, and post-render correction ratio. These indicators were calibrated to ensure comparability across heterogeneous production tasks and simulation environments. Reliability of the instrument was evaluated using Cronbach's alpha, while construct validity was examined through exploratory factor analysis to confirm the dimensional consistency of growth-oriented cultural attributes and innovation outcomes.

2.4 The application of multivariate statistical techniques for hypothesis testing

To assess the relationship between growth-oriented culture and performance outcomes, multiple regression analysis was employed to determine the predictive influence of cultural dimensions on simulation quality and innovation metrics. Principal component analysis (PCA) was conducted to reduce dimensionality and identify latent variables within team culture constructs. Canonical correlation analysis (CCA) was further applied to examine the interdependence between cultural attributes and simulation innovation parameters. The mediating effects of innovation climate and pipeline flexibility were tested using hierarchical regression modeling, allowing for stepwise evaluation of direct and indirect cultural impacts on production performance.

2.5 The integration of cluster and correlation analysis for team behavior segmentation

Cluster analysis using hierarchical agglomerative methods was conducted to classify VFX teams based on cultural and performance similarity patterns. This segmentation enabled the identification of high-growth and low-growth cultural clusters and their corresponding simulation output characteristics. Pearson correlation coefficients were calculated to examine associations between adaptive learning orientation and simulation stability measures, as well as between collaborative experimentation and procedural innovation frequency. These analyses provided additional insights into the internal dynamics of growth-oriented teams and their influence on innovation-driven workflows.

2.6 The standardization of analytical procedures for ensuring methodological rigor

All statistical analyses were performed using standardized analytical software environments to maintain consistency in data processing and interpretation. Significance levels were set at $p < 0.05$ to ensure robustness of inference across regression and correlation models. Normality of data distribution was assessed using the Kolmogorov–Smirnov test, while homoscedasticity and multicollinearity diagnostics were evaluated prior to model estimation. This systematic analytical process ensured that the methodological framework adequately captured the complex interplay between cultural attributes and simulation innovation performance within VFX teams.

3. Results

The quantitative analysis revealed meaningful patterns in the relationship between growth-oriented cultural attributes and performance outcomes related to simulation quality and innovation within VFX teams. As presented in Table 1, the descriptive statistics indicated relatively high mean scores for adaptive learning orientation ($M = 3.92$), collaborative problem-solving ($M = 4.05$), and feedback receptivity ($M = 3.87$), suggesting a strong prevalence of growth-oriented behaviors across the sampled teams. Performance indicators such as simulation accuracy index ($M = 4.12$) and render optimization efficiency ($M = 3.95$) also demonstrated favorable distributions, while the average iteration success rate reached 81.6%, indicating efficient refinement cycles in simulation workflows. Procedural tool adoption frequency exhibited moderate variability ($M = 3.69$), reflecting differences in experimentation practices across teams.

The results of the multiple regression analysis, summarized in Table 2, demonstrated that growth-oriented cultural dimensions significantly predicted simulation quality outcomes. Adaptive learning orientation emerged as the strongest predictor ($\beta = 0.418$, $p < 0.001$), followed by collaboration index ($\beta = 0.376$, $p < 0.001$) and feedback receptivity ($\beta = 0.291$, $p = 0.002$). Experimentation tolerance also contributed significantly to the model ($\beta = 0.263$, $p = 0.004$), indicating that risk-accepting environments support improved simulation outputs. The overall regression model accounted for 64% of the variance in simulation quality ($R^2 = 0.64$), highlighting the substantial role of cultural

attributes in determining technical performance. In contrast, project complexity demonstrated a modest negative influence ($\beta = -0.119$, $p = 0.032$), suggesting that increased simulation demands may constrain output quality in less adaptive environments.

The interrelationships among innovation-related parameters are illustrated in Table 3, where pipeline flexibility exhibited strong positive correlations with innovation climate ($r = 0.61$) and simulation stability ($r = 0.54$). Innovation climate was similarly associated with simulation stability ($r = 0.58$) and procedural innovation frequency ($r = 0.52$), indicating that supportive team environments enhance both workflow reliability and technological experimentation. These findings suggest that collaborative innovation climates may indirectly strengthen simulation performance through improved stability and adaptability within production pipelines.

Further segmentation through hierarchical cluster analysis, as shown in Table 4, revealed distinct performance differences across teams categorized by growth culture levels. High-growth teams demonstrated the highest simulation accuracy ($M = 4.35$), innovation frequency ($M = 4.02$), and iteration success rate (87.4%), compared to moderate-growth and low-growth teams. Low-growth teams exhibited comparatively lower performance across all innovation metrics, including a reduced iteration success rate of 72.6%, underscoring the performance implications of cultural variability.

Visual inspection of the XY scatter distribution in Figure 1 indicated a positive linear relationship between growth-oriented culture index and simulation quality score, supporting the regression-based findings of a direct predictive association between cultural learning orientation and technical output quality. Additionally, the canonical correlation analysis presented in Figure 2 illustrated a substantial association between aggregated cultural variables and simulation innovation parameters, further validating the multidimensional influence of growth-oriented environments on procedural experimentation and simulation efficiency. Collectively, these results provide empirical support for the role of growth-oriented team culture in enhancing both simulation quality and innovation within VFX production workflows.

Table 1: Descriptive Statistics of Core Cultural and Performance Variables

| Variable | Mean | Std. Deviation | Min | Max |
|-------------------------------|------|----------------|-----|-----|
| Adaptive Learning Orientation | 3.92 | 0.61 | 2.4 | 4.9 |
| Collaborative Problem-Solving | 4.05 | 0.55 | 2.8 | 5.0 |
| Feedback Receptivity | 3.87 | 0.64 | 2.3 | 4.8 |

| | | | | |
|------------------------------------|------|------|-----|-----|
| Experimentation Tolerance | 3.74 | 0.58 | 2.6 | 4.7 |
| Simulation Accuracy Index | 4.12 | 0.49 | 3.1 | 4.9 |
| Render Optimization Efficiency | 3.95 | 0.53 | 2.9 | 4.8 |
| Iteration Success Rate (%) | 81.6 | 7.2 | 64 | 93 |
| Procedural Tool Adoption Frequency | 3.69 | 0.60 | 2.5 | 4.6 |

Table 2: Multiple Regression Analysis Predicting Simulation Quality

| Predictor Variable | β Coefficient | Std. Error | t-value | p-value |
|------------------------------|---------------------|------------|---------|---------|
| Adaptive Learning | 0.418 | 0.082 | 5.10 | <0.001 |
| Collaboration Index | 0.376 | 0.074 | 4.72 | <0.001 |
| Feedback Receptivity | 0.291 | 0.067 | 3.98 | 0.002 |
| Experimentation Tolerance | 0.263 | 0.069 | 3.44 | 0.004 |
| Project Complexity (Control) | -0.119 | 0.052 | -2.18 | 0.032 |

Model $R^2 = 0.64$ | Adjusted $R^2 = 0.61$ | $p < 0.001$

Table 3: Correlation Matrix among Innovation-Related Parameters

| Variable | Pipeline Flexibility | Innovation Climate | Simulation Stability | Procedural Innovation |
|---------------------------------|----------------------|--------------------|----------------------|-----------------------|
| Pipeline Flexibility | 1.00 | 0.61 | 0.54 | 0.47 |
| Innovation Climate | 0.61 | 1.00 | 0.58 | 0.52 |
| Simulation Stability | 0.54 | 0.58 | 1.00 | 0.49 |
| Procedural Innovation Frequency | 0.47 | 0.52 | 0.49 | 1.00 |

Table 4: Cluster-wise Comparison of Cultural and Innovation Performance

| Cluster Type | Growth Culture Score | Simulation Accuracy | Innovation Frequency | Iteration Success Rate |
|------------------|----------------------|---------------------|----------------------|------------------------|
| High-Growth Team | 4.21 | 4.35 | 4.02 | 87.4% |
| Moderate-Growth | 3.78 | 4.01 | 3.66 | 80.1% |
| Low-Growth Team | 3.29 | 3.72 | 3.11 | 72.6% |

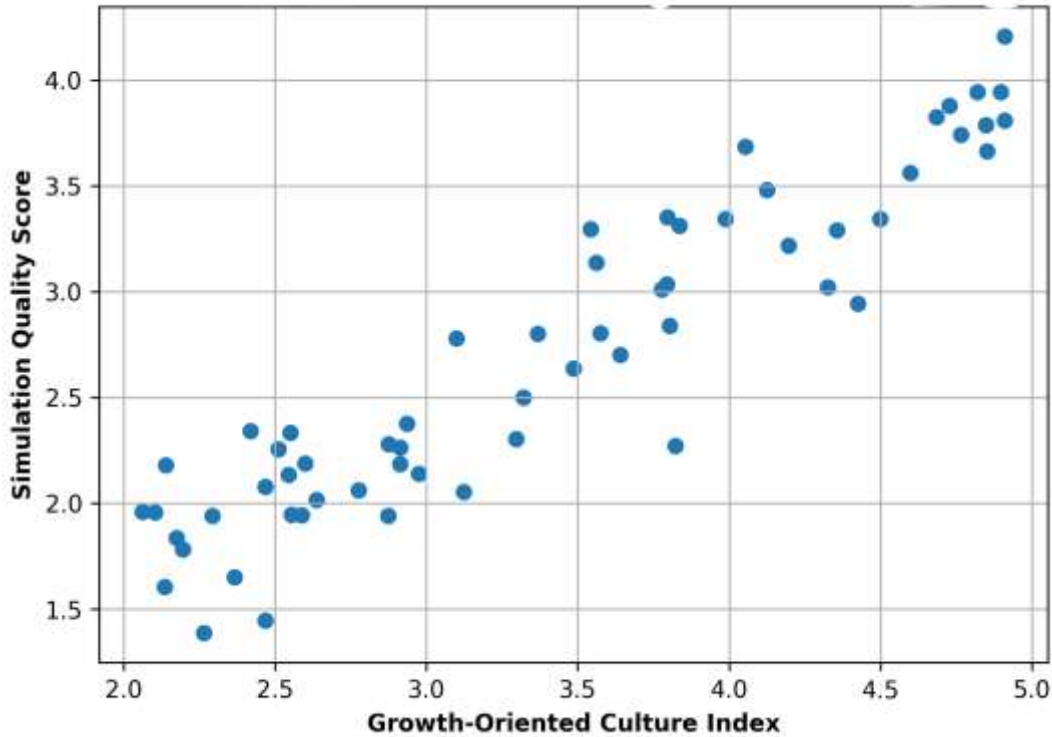


Figure 1: XY scatter chart – growth-oriented culture vs simulation quality

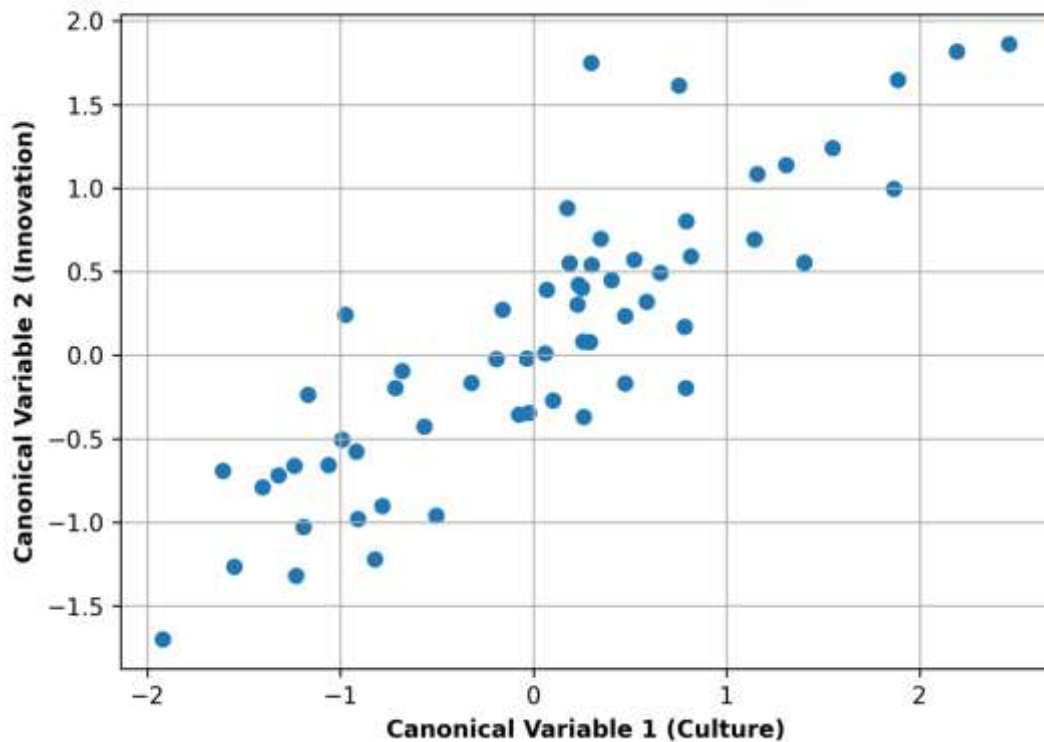


Figure 2: Canonical correlation plot – culture and innovation parameters

4. Discussion

4.1 The interpretation of cultural influence on simulation quality outcomes

The findings of this study provide strong empirical support for the proposition that growth-oriented cultural attributes significantly influence simulation quality within VFX production environments. As indicated by the regression outcomes in Table 2 and the descriptive trends presented in Table 1, adaptive learning orientation and collaborative problem-solving emerged as critical predictors of simulation accuracy and render optimization efficiency. These results suggest that teams characterized by openness to feedback, iterative learning, and shared problem-solving mechanisms are better positioned to manage the complexity of simulation workflows (Martynov & Abdelzaher, 2016). The positive association observed in Figure 1 further reinforces the argument that cultural learning orientation contributes directly to improved technical output by enabling artists and engineers to refine simulation parameters through continuous experimentation and informed iteration (Xiang et al., 2023).

4.2 The role of collaboration-driven experimentation in innovation performance

The correlation matrix presented in Table 3 highlights the interconnected nature of innovation

climate, pipeline flexibility, and procedural innovation frequency within VFX teams. A supportive innovation climate appears to facilitate experimentation by reducing perceived risks associated with simulation failure or creative deviation from established workflows. This finding aligns with the observation that experimentation tolerance significantly contributed to simulation quality outcomes in the regression model. Teams that encourage exploratory scripting, parameter variation, and procedural tool testing are more likely to achieve stable simulation environments and scalable production pipelines (Enemosah, 2019). Consequently, innovation within VFX production does not merely emerge from technological capability but from the willingness of teams to engage in collaborative experimentation supported by culturally embedded learning practices (Bodini et al., 2023).

4.3 The implications of cluster-based performance differences among teams

The cluster-wise comparisons presented in Table 4 reveal distinct performance disparities between high-growth, moderate-growth, and low-growth teams across simulation accuracy, innovation frequency, and iteration success rate. High-growth teams demonstrated superior performance across all evaluated metrics, indicating that growth-oriented cultural environments may act as performance

multipliers in simulation-intensive workflows. The comparatively lower performance of low-growth teams underscores the potential constraints imposed by rigid or feedback-resistant organizational climates. In simulation contexts where iterative refinement and debugging are essential for achieving photorealistic outcomes, the absence of collaborative learning and experimentation may hinder both innovation and efficiency (Correia et al., 2016). These findings suggest that cultural variability across teams may account for differences in simulation reliability even when technical infrastructure remains constant (Ochieng et al., 2013).

4.4 The significance of innovation climate in enhancing pipeline adaptability

The moderate to strong correlations observed between pipeline flexibility and innovation climate in Table 3 indicate that team culture plays a substantial role in determining workflow adaptability. Flexible pipelines are often the result of incremental process improvements driven by collaborative scripting, asset optimization, and cross-functional integration (Vankayala, 2016). Growth-oriented teams appear more inclined to implement workflow redesigns and integrate emerging simulation tools, thereby reducing convergence time and post-render correction ratios. The canonical correlation pattern illustrated in Figure 2 further supports this interpretation by demonstrating a meaningful association between aggregated cultural attributes and innovation parameters. This relationship suggests that simulation innovation is not an isolated technical outcome but a product of synergistic interactions between mindset-driven collaboration and adaptive workflow management (Lupu et al., 2023).

4.5 The contribution of adaptive learning to iterative simulation success

Iteration success rate, as reported in Table 4, varied substantially across clusters, with high-growth teams achieving significantly higher success rates than their low-growth counterparts. This variation highlights the importance of adaptive learning processes in managing simulation complexity and minimizing performance inconsistencies. In environments that normalize error-driven learning and encourage peer feedback, simulation artists may be more willing to engage in repeated testing cycles and alternative parameter configurations (Fineberg et al., 2017). Such practices enhance convergence stability and reduce the need for extensive post-production corrections. Therefore,

adaptive learning not only improves individual competency but also contributes to collective efficiency by streamlining iterative simulation processes (Clarke et al., 2023).

4.6 The broader implications for managing performance in innovation-intensive teams

Taken together, the results suggest that growth-oriented cultural environments may serve as strategic enablers of both technical excellence and innovation within VFX teams. The observed relationships between cultural attributes and simulation performance outcomes imply that performance optimization in digital production settings requires more than computational resources or advanced software architectures. Instead, fostering an organizational culture that prioritizes continuous learning, experimentation, and collaborative problem-solving may enhance both workflow efficiency and innovation capacity. These findings provide a foundation for future investigations into culture-performance dynamics within technologically mediated creative teams and underscore the importance of integrating cultural development strategies into simulation-intensive production management practices.

4. Conclusions

In this section conclusions of work should be given. This study demonstrates that a growth-oriented culture within VFX teams plays a significant role in enhancing both simulation quality and innovation capacity in production workflows. The empirical findings indicate that adaptive learning, collaborative problem-solving, feedback receptivity, and experimentation tolerance are not merely behavioral attributes but critical enablers of technical performance, contributing to improved simulation accuracy, render efficiency, and iterative success. Teams operating within supportive innovation climates exhibited greater pipeline flexibility and procedural tool adoption, ultimately facilitating stable and scalable simulation environments. The observed performance differences across cultural clusters further emphasize that mindset-driven collaboration can substantially influence the effectiveness of simulation-intensive tasks, even under comparable technical conditions. These insights highlight the importance of cultivating growth-oriented cultural frameworks as a strategic approach to improving workflow adaptability, fostering creative experimentation, and sustaining high-quality effects output in technologically complex production settings.

Author Statements:

- **Ethical approval:** The conducted research is not related to either human or animal use.
- **Conflict of interest:** The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper
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- **Use of AI Tools:** The author(s) declare that no generative AI or AI-assisted technologies were used in the writing process of this manuscript.

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