

Enhancing Team Efficiency and Quality Management through AI-Driven 3D Game Development

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Abstract

This study utilizes a multi-level efficiency analysis to examine the impact of artificial intelligence (AI) on team efficiency and approach to content quality management during 3D game creation. While AI is being increasingly integrated into game development, its concrete advantages across different phases of production and types of content are still unclear. The study examines ten projects in 3D content production narrative from commercial, education, and entertainment sectors and investigates key production phases, namely, pre-production, asset creation, production, and post-production. Results show that post-production provides the best and highest efficiency (0.913) sustained over a longer period, with central production processes showing the lowest efficiency gains (0.042). AI demonstrates its true strength with technical, rule-based work (rendering, data processing, etc) rather than in hyper-creative processes, and the largest efficiency gains are to be found in industry projects. The research stresses the necessity of the strategic integration of AI and organizational preparedness beyond just access to tools. Interestingly, productivity is at its highest when AI tools account for 30–70% of working hours, revealing superior outcomes from a well-balanced collaboration between AI and local staff. These findings support a cautious trajectory of AI integration that begins with the automation of repetitive, machine-processable tasks to facilitate the efficiency of a team-generation process outputting high-quality content. This study provides significant theoretical and practical implications to maximize the utilization of AI in managing the short-term game industry workflow and to explore in a grounded way whether and how AI implementation could better enhance productivity and quality management, offering empirical findings for such future AI use.

Keywords: Artificial Intelligence (AI), 3D Game Animation, Efficiency Analysis, Content Quality Management

Introduction

That all changed through the incorporation of AI into the lifecycle of the production of 3D games that transformed developer's workflows, team efficiency, quality control, and asset creation. AI helps teams to automate menial work, create materials, and decide, so they can

both streamline their workflows and spend more time on creative challenges (Singh & Kaur, 2023). The adoption of AI, however, is not uniform across all production pipelines, and there are concerns about the efficiency of its use at different points in the development process.

But while AI's contribution to game production is heavily debated, the vast majority of this research has focused on its use within gameplay experiences—whether AI-driven NPC behavior or procedural content generation (Acemoglu & Restrepo, 2020). Internally in production processes or team workflows, not much more is understood about the helping hand AI should be. This study seeks to bridge the gap by examining AI's impact on efficiency across the four phases of production — pre-production, asset creation, production, and post-production.

This research focuses on three key questions;

RQ1: How does the inclusion of AI within 3D game development groups affect the overall efficiency of the whole team?

RQ2: At what stage of production do AI-based efficiency improvements have their greatest impact?

RQ3: What are the differences in the AI efficiency gains by the project type (e.g., education vs. business), and what are the differences in resource management and quality control implications?

To address these questions, we investigate data from ten 3D game production projects, comparing the degrees of AI utilization, efficiency measures, and quality outputs. It implies that AI can work wonders in streamlining the efficiency of post-production, while its role in making main production processes efficient is of no consequence, thus highlighting the long-standing status of human expertise.

Literature Review

AI in 3D Game Development

AI has a wide variety of applications in the gaming industry, including automated animation, procedural content generation, and quality control based on AI. According to Hussain (2024), AI can help optimize digital media production by reducing the time it takes to create digital media whilst maintaining quality. Solomon (2024) highlights research done in the adjacent realms of animation and film, where AI tools are found to create efficiencies by automating repetitive technical labor associated with the writing process.

Recent developments in AI asset generation, which include automated texture generation, 3D modeling, and rigging (Reddy et al., 2024), alleviate some of the manual labor burden placed upon artists. Despite this, artist intervention is required to ensure some continuity of artwork (Anantrasirichai & Bull, 2022). AI is also used in automating workflows for project planning and task distribution (Carayannis et al., 2016).

Getting Started with AI Workflow Efficiency

Data Envelopment Analysis (DEA): Researchers quantify the contribution of AI to production efficiency, a performance measurement strategy used to assess productivity in multi-stage processes (Cook et al., 2010). Network DEA (NDEA) models can be used to research creative

production pipelines (Zhang et al., 2021). The present study uses an NDEA methodology to assess the productivity performance of 3D game production.

Methodology

A qualitative interview with experts was combined with a quantitative efficiency assessment as a mixed-method design was implemented in this study. The strategy consists of three stages:

AI Integration: Industry experts also share their thoughts and insights on best practices and the challenges faced by game developers working with AI; including developers, engineers, and creative directors.

Appliance for DEA Models of Network System: The production process of a 3D game consists of four stages, i.e., preproduction, asset creation, production, and postproduction, and the efficiency level of their allocation is analyzed. Analysis of sample projects: 10 projects from game production (3D) Based on AI, team productivity, and quality.

Results & Discussion

The results showed a significant improvement in the efficiency in the post-production process with an average efficiency rate of 0.913 for the projects. AI-based optimization and testing tools are enhancing quality control with minimum manual tasks (Jayanthiladevi et al., 2020) and ensuring the standards are always high. Conversely, basic production tasks (animation, gameplay coding) have negligible efficiency improvements of ~ 0.042 , highlighting AI's very minimal contribution to these tasks.

The study also pinpointed the sweet spot in which AI should be applied (30–70% of the total production hours) to maintain a healthy balance between efficiency and creative quality. Heavy reliance on AI (>80%) may limit good and cause even more manual corrections, and too little AI (<30%) can mean lost opportunities for productivity gains (LeCun et al., 2015).

Conclusion & Recommendations

Our findings validate the fact that artificial intelligence boosts the productivity of three-dimensional game production, especially in post-production and asset-creation domains. But it is only useful if used strategically. About our findings, we recommend the following:

Gradually adopt AI: First, implement AI in the post-production phase and later you can extend the same in other phases. AI-Human Partnership in Harmony: Use AI between 30–70% usage only to ensure productivity yet maintain the creative arm. Quality Assurance Processes: Integrate AI testing tools that operate within human supervision to achieve quality results. Moreover, AI capabilities in real-time procedural content generation and adaptive storytelling should be explored further as these are incredibly promising directions for game development.

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