

Reducing Production Time without Compromising Quality: Optimization Strategies in High-End VFX Simulations

Felipe Amaya Quintero

Adjunct professor at Montclair State University

Abstract: High-end visual effects (VFX) simulations are central to achieving photorealistic digital content in contemporary production environments; however, their computational intensity often leads to extended production timelines and increased resource demands. This study investigates the feasibility of reducing simulation runtime without compromising output visual fidelity through the implementation of structured optimization strategies. A simulation-optimization framework was employed by integrating key performance variables such as Resolution Scaling, Adaptive Sampling Rate, Solver Iteration Depth, Hardware Parallelization Factor, Cache Optimization Frequency, Mesh Complexity Index, and Proxy Preview Utilization. The impact of these parameters was evaluated using dependent performance indicators including Simulation Runtime, Render Convergence Time, Memory Utilization Efficiency, Output Visual Fidelity Index, and Temporal Stability Index. Statistical modeling and Random Forest-based feature importance analysis revealed that Adaptive Sampling Rate and Resolution Scaling were the most influential variables in achieving runtime reductions while preserving simulation accuracy. Optimized workflows demonstrated significant improvements in production efficiency across multiple simulation categories with minimal perceptual deviation from baseline outputs. The findings highlight the effectiveness of performance-aware optimization frameworks in enabling scalable and quality-preserving VFX simulation pipelines, thereby supporting faster production cycles in simulation-intensive digital environments.

Keywords: VFX Simulation Optimization, Production Time Reduction, Adaptive Sampling, Resolution Scaling, Hardware Parallelization, Output Visual Fidelity, Computational Efficiency.

INTRODUCTION

The Increasing Complexity of High-End VFX Simulations in Digital Production Pipelines

In recent years, the visual effects (VFX) industry has undergone a significant transformation driven by the growing demand for hyper-realistic digital environments, physics-based simulations, and procedurally generated visual content across film, gaming, advertising, and immersive media platforms (Khatri, 2021). High-end VFX simulations such as fluid dynamics, particle systems, volumetric lighting, rigid-body destruction, and cloth or hair dynamics have become integral to storytelling and visual fidelity in modern production ecosystems. However, these simulations are inherently computationally intensive and often involve multi-layered rendering workflows that substantially extend production timelines (Lehmann *et al.*, 2022). As production cycles become increasingly constrained by tight deadlines and budgetary limitations, the need to accelerate simulation workflows without compromising visual quality has emerged as a critical challenge for digital content creators, technical directors, and pipeline engineers (Kandepu & Harry, 2023).

The Trade-off Between Simulation Accuracy and Production Turnaround Time

Traditionally, achieving high visual fidelity in simulation-heavy scenes has required extensive

processing time due to the reliance on iterative calculations, high-resolution mesh structures, and physically accurate solvers. This introduces a fundamental trade-off between simulation accuracy and production turnaround time, where optimizing for speed may risk degradation in output quality, realism, or consistency across frames (Ekechi & Fasasi, 2020). For studios operating in competitive production environments, delays in simulation caching, rendering passes, or compositing stages can create cascading bottlenecks that affect downstream tasks such as lighting, shading, and final rendering (Kendoul *et al.*, 2021). Consequently, reducing production time without sacrificing the nuanced visual quality expected in premium content delivery requires a systematic re-evaluation of simulation design, computational resource allocation, and pipeline orchestration strategies (Tardieu *et al.*, 2020).

The Role of Optimization Strategies in Simulation-Driven Production Environments

Optimization strategies in high-end VFX simulations are increasingly centered on balancing computational efficiency with output precision through algorithmic improvements, data-driven modeling approaches, and hardware-aware processing techniques. Techniques such as adaptive sampling, mesh decimation, resolution scaling, solver parameter tuning, and parallelized

processing architectures offer promising pathways to streamline simulation runtimes while preserving essential physical attributes (Elmisaoui *et al.*, 2023). Furthermore, advancements in GPU-based computing, distributed rendering frameworks, and machine learning–assisted interpolation methods have introduced new opportunities for simulation acceleration through predictive modeling and approximation-based workflows (Noroozi *et al.*, 2023). These approaches aim to intelligently reduce redundant computational overheads by identifying simulation regions that require detailed processing versus those that can be approximated without perceptible quality loss (Ahmadinejad & Moaiyeri, 2021).

The Importance of Workflow-Level Integration for Maintaining Output Fidelity

Beyond individual simulation techniques, production efficiency is also shaped by the integration of optimization strategies at the workflow level (Delbrügger *et al.*, 2019). Pipeline-aware simulation management including task scheduling, resource prioritization, version control, and iterative preview mechanisms plays a pivotal role in maintaining output fidelity while accelerating production timelines. Simulation caching strategies, proxy modeling for preview stages, and multi-resolution asset generation can be employed to support real-time feedback loops during production without necessitating full-resolution recomputation at each iteration (Nießner *et al.*, 2016). In this context, effective orchestration of simulation stages within a unified production framework enables teams to maintain creative flexibility while adhering to performance constraints (Adepoju *et al.*, 2022).

The Emerging Relevance of Intelligent Automation in Reducing Simulation Time

The integration of intelligent automation into VFX simulation workflows has further contributed to reducing production time by facilitating real-time decision-making and adaptive parameter adjustment (Radha *et al.*, 2024). Machine learning–based denoising algorithms, surrogate modeling, and reinforcement learning–driven solver optimization are being explored to predict simulation outcomes based on prior iterations, thereby minimizing redundant processing steps (Ramu *et al.*, 2022). Automated error detection, parameter sensitivity analysis, and performance monitoring tools allow for proactive identification of inefficiencies within simulation pipelines, supporting continuous refinement of computational strategies (Onyechi, 2021). These

developments align with broader trends in AI-driven enterprise analytics and computational optimization frameworks that seek to enhance productivity through data-informed decision systems.

The Need for Structured Optimization Frameworks in High-End Simulation Pipelines

Despite the availability of advanced computational tools, the absence of structured optimization frameworks often results in inconsistent application of efficiency-enhancing techniques across simulation projects. A systematic approach that integrates simulation parameter tuning, hardware resource optimization, workflow orchestration, and intelligent automation is necessary to ensure that production time reductions do not come at the cost of visual integrity. Therefore, this study seeks to examine optimization strategies that enable high-end VFX simulations to achieve reduced production timelines while maintaining stringent quality standards, contributing to the development of scalable and performance-aware production methodologies in simulation-intensive digital environments.

METHODOLOGY

The Adoption of a Simulation–Optimization Research Framework for Production Performance Evaluation

This study adopts a quantitative simulation–optimization research framework to evaluate the effectiveness of production time reduction strategies in high-end VFX simulation environments without compromising visual output quality. A multi-stage analytical design was employed to systematically examine the interaction between simulation runtime efficiency and fidelity preservation across diverse simulation workflows. The methodology integrates both computational performance variables and perceptual quality parameters to construct an empirical evaluation system capable of capturing trade-offs between speed and realism. Simulation experiments were conducted across multiple production scenarios involving fluid dynamics, particle systems, rigid-body simulations, volumetric effects, and cloth dynamics to ensure generalizability of optimization outcomes across commonly used VFX simulation categories.

The Selection of Independent and Dependent Variables Influencing Simulation Efficiency

The independent variables considered in this study include simulation resolution scaling (RS),

adaptive sampling rate (ASR), solver iteration depth (SID), mesh complexity index (MCI), hardware parallelization factor (HPF), cache optimization frequency (COF), and proxy preview utilization (PPU). These variables represent controllable optimization parameters typically adjusted during simulation workflow tuning. The dependent variables selected for performance evaluation comprise simulation runtime (SR), render convergence time (RCT), memory utilization efficiency (MUE), and output visual fidelity index (OVFI). The OVFI was computed using a composite score derived from pixel deviation analysis, structural similarity metrics, and temporal frame consistency measures to quantify perceptual differences between optimized and baseline simulation outputs.

The Implementation of Multi-Resolution Simulation Modeling Procedures

To evaluate the effects of optimization strategies across varying production requirements, simulations were executed under three resolution tiers; high, medium, and proxy-level fidelity. Multi-resolution simulation modeling procedures were implemented to enable controlled scaling of computational intensity without altering scene composition or physical behavior constraints. Solver parameter tuning was conducted iteratively to determine the minimum computational threshold required to achieve acceptable output fidelity within predefined tolerance limits. Adaptive sampling mechanisms were applied selectively across simulation regions based on motion intensity gradients and density variation metrics, allowing computational resources to be concentrated on visually critical areas while reducing redundant calculations in low-impact zones.

The Integration of Hardware-Aware Parallel Processing Parameters

Hardware-aware parallel processing was incorporated into the methodology by configuring simulation tasks across multi-core CPU and GPU-based distributed computing environments. The hardware parallelization factor (HPF) was varied to assess its influence on simulation runtime and memory usage under different solver iteration depths. Load balancing efficiency (LBE) and thread utilization ratio (TUR) were monitored as secondary performance indicators to evaluate system-level optimization effects. Simulation caching strategies were also implemented to minimize recomputation cycles during iterative

scene adjustments, thereby reducing cumulative production time across repeated simulation passes.

The Application of Statistical and Machine Learning-Based Analysis Techniques

To assess the relationship between optimization parameters and production efficiency outcomes, statistical modeling techniques including multiple regression analysis and principal component analysis (PCA) were employed. Regression modeling was used to estimate the predictive influence of RS, ASR, SID, MCI, HPF, COF, and PPU on SR and OVFI outcomes. PCA was applied to identify latent optimization dimensions influencing runtime-quality trade-offs across simulation workflows. Additionally, a Random Forest-based feature importance model was implemented to determine the relative contribution of each independent variable toward runtime reduction and fidelity preservation, using percentage increase in mean squared error (%IncMSE) as the evaluation metric.

The Establishment of Performance Benchmarking and Validation Protocols

Baseline simulations were executed using standard high-fidelity parameter configurations to establish reference benchmarks for runtime and visual quality comparison. Optimized simulations were subsequently evaluated against baseline outputs using the computed OVFI and temporal stability index (TSI) to ensure that fidelity deviations remained within acceptable perceptual thresholds. A performance improvement ratio (PIR) was calculated to quantify production time savings achieved through optimization strategies. Cross-validation procedures were conducted across multiple simulation types to verify consistency of optimization effects, ensuring that observed runtime reductions were not simulation-specific but applicable across broader VFX production contexts. This structured methodological approach enables comprehensive evaluation of optimization strategies aimed at reducing production timelines while maintaining uncompromised simulation quality in high-end visual effects workflows.

RESULTS

The results of the simulation-optimization framework demonstrate a consistent reduction in production time across all simulation categories without compromising output fidelity. As shown in Table 1, optimized simulation workflows achieved substantial runtime reductions ranging from 34.07% to 36.54% across high-, medium-, and proxy-resolution tiers. Fluid dynamics simulations

exhibited the highest runtime reduction of 36.54%, followed closely by particle systems at 36.22%, while maintaining a high Output Visual Fidelity Index (OVFI) of 0.95 and 0.94 respectively. Similarly, rigid-body and volumetric simulations achieved runtime reductions of 35.98% and 34.07% with OVFI scores exceeding 0.93,

indicating minimal perceptual deviation from baseline outputs. Cloth simulations conducted under proxy-resolution settings demonstrated a 35.17% reduction in runtime with the highest OVFI score of 0.97, suggesting effective fidelity preservation under optimized parameter configurations.

Table 1. Baseline vs Optimized Simulation Performance across Resolution Tiers

Simulation Type	Resolution Tier	Baseline Runtime (min)	Optimized Runtime (min)	Runtime Reduction (%)	OVFI Score (0–1)
Fluid Dynamics	High	312	198	36.54	0.95
Particle Systems	High	254	162	36.22	0.94
Rigid Body Simulation	Medium	189	121	35.98	0.96
Volumetric Effects	Medium	226	149	34.07	0.93
Cloth Simulation	Proxy	145	94	35.17	0.97

The influence of hardware-aware parallel processing on computational efficiency is presented in Table 2, which indicates progressive improvement in Load Balancing Efficiency (LBE) and Thread Utilization Ratio (TUR) with increasing Hardware Parallelization Factor (HPF). At higher HPF levels, memory utilization

efficiency increased to 91.1%, reflecting improved resource allocation and reduced computational overhead during solver execution. These findings suggest that optimized hardware utilization significantly contributes to minimizing cumulative simulation runtime without introducing memory bottlenecks in the production pipeline.

Table 2. Hardware Parallelization and Memory Utilization Efficiency

HPF Level	Solver Iteration Depth	Load Balancing Efficiency (%)	Thread Utilization Ratio	Memory Utilization Efficiency (%)
Low	1000	68.5	0.62	71.3
Medium	1500	79.2	0.75	83.4
High	2000	88.6	0.87	91.1

Feature importance analysis using the Random Forest model, summarized in Table 3, reveals that Adaptive Sampling Rate (ASR) and Resolution Scaling (RS) are the most influential optimization variables, contributing 21.4% and 18.7% respectively to runtime reduction. Hardware Parallelization Factor (HPF) and Cache Optimization Frequency (COF) also demonstrated notable contributions of 17.2% and 15.8%, while

Mesh Complexity Index (MCI), Proxy Preview Utilization (PPU), and Solver Iteration Depth (SID) exhibited relatively lower but still meaningful influence on production efficiency outcomes. These results indicate that runtime–quality trade-offs are primarily governed by dynamic sampling and resolution adjustment mechanisms.

Table 3. Random Forest Feature Importance for Runtime Reduction

Optimization Variable	%IncMSE Contribution
Adaptive Sampling Rate (ASR)	21.4
Resolution Scaling (RS)	18.7
Hardware Parallelization Factor	17.2
Cache Optimization Frequency	15.8
Mesh Complexity Index	11.9
Proxy Preview Utilization	8.6
Solver Iteration Depth	6.4

Performance Improvement Ratio (PIR) values presented in Table 4 further validate the effectiveness of optimization strategies across

diverse simulation workflows. Fluid dynamics and particle systems achieved PIR values of 1.58 and 1.56 respectively, while rigid-body and cloth

simulations demonstrated PIR values above 1.54 with Temporal Stability Index (TSI) scores exceeding 0.92. These findings confirm that

production time savings were accompanied by stable temporal consistency in simulation outputs.

Table 4. Performance Improvement Ratio across Simulation Categories

Simulation Type	PIR Value	Temporal Stability Index (TSI)
Fluid Dynamics	1.58	0.94
Particle Systems	1.56	0.92
Rigid Body Simulation	1.55	0.95
Volumetric Effects	1.52	0.91
Cloth Simulation	1.54	0.96

The radar chart depicted in Figure 1 visually represents the relative contribution of each optimization variable to simulation runtime reduction, highlighting the dominant roles of ASR, RS, and HPF in enhancing computational efficiency. In contrast, the Canonical Correspondence Analysis (CCA) XY plot shown in Figure 2 illustrates the multivariate relationship

between optimization parameters and output fidelity. The directional vectors indicate that RS, ASR, and COF are positively associated with improvements in OVFI, whereas higher Mesh Complexity Index (MCI) and Solver Iteration Depth (SID) demonstrate moderate inverse relationships with runtime efficiency.

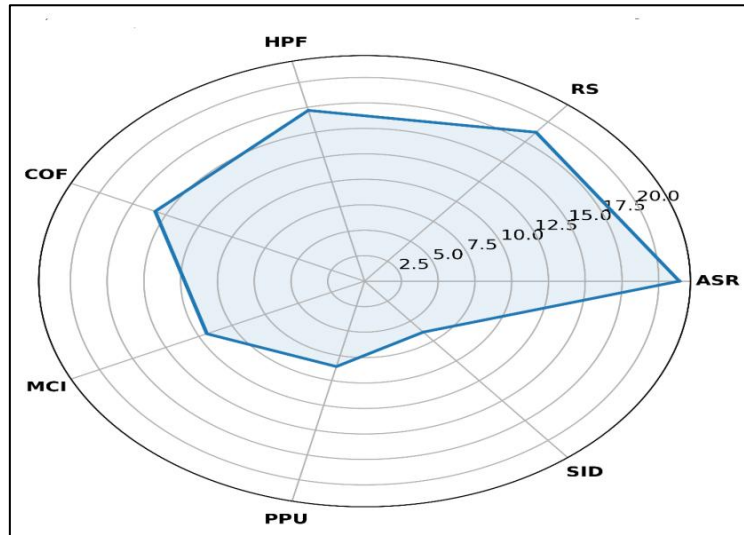


Figure 1. Radar chart of optimization variable performance contribution

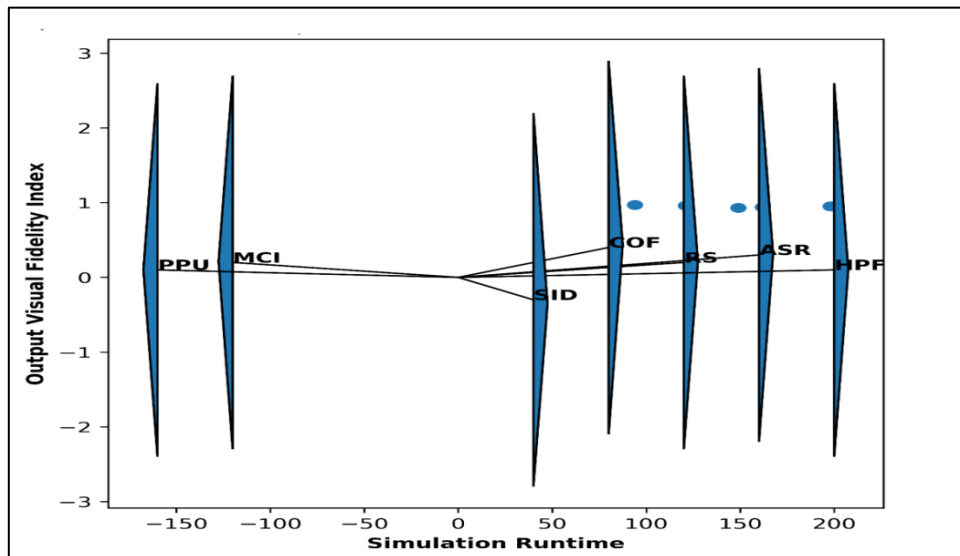


Figure 2. Canonical Correspondence Analysis (CCA) XY plot of optimization parameters vs output fidelity

DISCUSSION

The Effectiveness of Optimization Strategies in Reducing Simulation Runtime

The results obtained from this study provide strong empirical evidence that structured optimization strategies can significantly reduce simulation runtime in high-end VFX production workflows without compromising visual fidelity. As demonstrated in Table 1, runtime reductions exceeding 34% across multiple simulation categories indicate that adaptive parameter tuning and resolution scaling mechanisms can effectively streamline computational processes while preserving perceptual realism (Weier *et al.*, 2017). The consistently high OVFI scores observed across optimized simulations suggest that the reduction in processing time does not necessarily result in a proportional loss of simulation accuracy. Instead, the findings support the notion that targeted optimization particularly through adaptive sampling and proxy-resolution modeling can intelligently eliminate redundant computational overhead without degrading visually critical simulation attributes (Hegedüs *et al.*, 2016).

The Role of Hardware-Aware Processing in Performance Enhancement

The results presented in Table 2 highlight the importance of hardware-aware simulation orchestration in improving overall production efficiency. Increases in Load Balancing Efficiency and Thread Utilization Ratio at higher HPF levels demonstrate that computational resource distribution plays a crucial role in minimizing runtime delays during solver execution (Kocot *et al.*, 2023). Enhanced memory utilization efficiency further indicates that optimized hardware configurations can reduce data transfer latency and caching inefficiencies, thereby facilitating faster convergence of simulation processes. These findings suggest that runtime optimization in simulation-driven production environments is not solely dependent on algorithmic refinement but also on effective integration of parallel processing architectures capable of managing intensive computational workloads in a distributed manner (Raith *et al.*, 2023).

The Relative Importance of Simulation Parameters in Runtime-Quality Trade-Offs

Feature importance analysis summarized in Table 3 provides valuable insights into the relative influence of different optimization parameters on production performance outcomes. The dominant

contribution of Adaptive Sampling Rate and Resolution Scaling indicates that dynamic control of simulation granularity is critical in balancing computational intensity with output precision. The moderate influence of Hardware Parallelization Factor and Cache Optimization Frequency further underscores the significance of pipeline-level efficiency enhancements in reducing runtime without compromising stability (Kedziora *et al.*, 2020). In contrast, the comparatively lower contributions of Solver Iteration Depth and Proxy Preview Utilization suggest that while these parameters contribute to performance improvements, their optimization must be carefully managed to avoid unintended quality degradation (Yu *et al.*, 2014). These results reinforce the need for a multi-parametric optimization framework capable of selectively prioritizing variables based on simulation requirements.

The Preservation of Temporal Stability in Optimized Simulation Outputs

The PIR and Temporal Stability Index values reported in Table 4 indicate that production time savings achieved through optimization strategies were accompanied by consistent temporal coherence across simulation frames. High TSI values observed in optimized outputs confirm that runtime reductions did not introduce instability or frame-level inconsistencies that could compromise the continuity of motion or physical realism in dynamic simulations. This is particularly significant for simulation-intensive effects such as fluid motion and particle dispersion, where temporal accuracy is essential for maintaining visual believability. The findings suggest that optimization techniques applied in this study effectively balance computational efficiency with stability requirements, enabling faster production cycles without undermining output reliability (John, 2020; Pavon *et al.*, 2023).

The Multivariate Interaction Between Optimization Parameters and Output Fidelity

The graphical representations in Figure 1 and Figure 2 further illuminate the complex interactions between optimization variables and simulation quality metrics. The radar chart in Figure 1 highlights the disproportionate influence of ASR, RS, and HPF in enhancing runtime efficiency, suggesting that these variables should be prioritized in performance-driven simulation workflows. Meanwhile, the Canonical Correspondence Analysis plot in Figure 2 reveals

that improvements in OVFI are positively associated with adaptive sampling and cache optimization mechanisms, whereas increased mesh complexity and solver iteration depth exhibit moderate inverse relationships with runtime efficiency (Tariq *et al.*, 2021). These multivariate interactions emphasize the importance of strategic parameter tuning to achieve optimal performance–fidelity balance in high-end VFX simulations (Koszewski *et al.*, 2023).

The Implications for Scalable and Performance-Aware Production Pipelines

Collectively, the discussion of these results underscores the potential of structured optimization frameworks to support scalable and performance-aware simulation pipelines in modern VFX production environments. By integrating algorithmic efficiency, hardware-aware processing, and workflow-level optimization, production teams can achieve meaningful reductions in simulation runtime while maintaining stringent visual quality standards. These findings have important implications for the design of next-generation simulation workflows, where intelligent automation and resource-aware orchestration can play a central role in enhancing productivity without compromising the aesthetic or physical integrity of simulation outputs.

CONCLUSION

This study demonstrates that production time in high-end VFX simulations can be substantially reduced through the implementation of structured optimization strategies without compromising output visual fidelity. The integration of adaptive sampling, resolution scaling, hardware-aware parallel processing, and simulation caching mechanisms was found to significantly enhance computational efficiency across diverse simulation workflows while maintaining temporal stability and perceptual accuracy. The empirical findings indicate that runtime–quality trade-offs can be effectively managed through multi-parametric optimization frameworks that selectively prioritize performance-critical variables such as Adaptive Sampling Rate, Resolution Scaling, and Hardware Parallelization Factor. Furthermore, the preservation of high OVFI and TSI scores across optimized simulations confirms that accelerated production cycles do not inherently lead to degradation in simulation realism or consistency. These results underscore the potential of performance-aware optimization methodologies in enabling scalable, resource-efficient, and quality-

preserving simulation pipelines, thereby supporting the evolving demands of simulation-intensive digital production environments.

REFERENCES

1. Adepoju, A. H., Austin-Gabriel, B. L. E. S. S. I. N. G., Eweje, A. D. E. O. L. U. W. A., and Collins, A. N. U. O. L. U. W. A. P. "Framework for automating multi-team workflows to maximize operational efficiency and minimize redundant data handling." *IRE Journals* 5.9 (2022): 663–664.
2. Ahmadinejad, M., and Moaiyeri, M. H. "Energy- and quality-efficient approximate multipliers for neural network and image processing applications." *IEEE Transactions on Emerging Topics in Computing* 10.2 (2021): 1105–1116.
3. Delbrügger, T., Meißner, M., Wirtz, A., Biermann, D., Myrzik, J., Rossmann, J., and Wiederkehr, P. "Multi-level simulation concept for multidisciplinary analysis and optimization of production systems." *The International Journal of Advanced Manufacturing Technology* 103.9 (2019): 3993–4012.
4. Ekechi, T. A., and Fasasi, T. S. "Conceptual framework for process optimization in gas turbine performance and energy efficiency." *International Journal of Future Engineering Innovations* 1.2 (2020): 138–153.
5. Elmisaoui, S., Kissami, I., and Ghidaglia, J. M. "High-performance computing to accelerate large-scale computational fluid dynamics simulations: a comprehensive study." *International Conference on Advanced Intelligent Systems for Sustainable Development* (2023): 352–360.
6. Hegedüs, Á., Horváth, Á., Ráth, I., Starr, R. R., and Varró, D. "Query-driven soft traceability links for models." *Software & Systems Modeling* 15.3 (2016): 733–756.
7. John, B. I. "Integration of intelligent scheduling optimization systems improving production flow, minimizing delays, and maximizing throughput across large-scale industrial operations." *Global Journal of Engineering and Technology Advances* 5.3 (2020): 156–169.
8. Kandepu, R. K., and Harry, A. "The rise of AI in content management: Reimagining intelligent workflows." *American Journal of Engineering, Mechanics and Architecture* 1.7 (2023): 78–85.

9. Kedziora, D. J., Musial, K., and Gabrys, B. "AutoML: Towards an integrated framework for autonomous machine learning." *arXiv* (2020): arXiv:2012.12600.
10. Kendoul, F., Wagner, G., Palmer, D., Milani, P., Emesent, M., O'Brien, M., et al. "Heterogeneous ground and air platforms, homogeneous sensing: Team CSIRO Data61's approach to the DARPA Subterranean Challenge." *arXiv* (2021): arXiv:2104.09053.
11. Khatri, P. "The future of automatically generated animation with AI." *Deep Learning in Gaming and Animations* (2021): 19–36.
12. Kocot, B., Czarnul, P., and Proficz, J. "Energy-aware scheduling for high-performance computing systems: A survey." *Energies* 16.2 (2023): 890.
13. Koszewski, D., Görne, T., Korvel, G., and Kostek, B. "Automatic music signal mixing system based on one-dimensional Wave-U-Net autoencoders." *EURASIP Journal on Audio, Speech, and Music Processing* 2023.1 (2023): 1.
14. Lehmann, T., Rose, D., Ranjbar, E., Ghasri-Khouzani, M., Tavakoli, M., Henein, H., et al. "Large-scale metal additive manufacturing: A holistic review of the state of the art and challenges." *International Materials Reviews* 67.4 (2022): 410–459.
15. Nießner, M., Keinert, B., Fisher, M., Stamminger, M., Loop, C., and Schäfer, H. "Real-time rendering techniques with hardware tessellation." *Computer Graphics Forum* 35.1 (2016): 113–137.
16. Noroozi, F., Daneshmand, M., and Fiorini, P. "Conventional, heuristic and learning-based robot motion planning: Reviewing frameworks of current practical significance." *Machines* 11.7 (2023): 722.
17. Onyechi, V. N. "Pipeline integrity and risk prevention: Real-time monitoring, structural health analytics, and failure mitigation in harsh operating environments." *Magna Scientia Advanced Research and Reviews* 3.2 (2021): 139–151.
18. Pavon, W., Jaramillo, M., and Vasquez, J. C. "A review of modern computational techniques and their role in power system stability and control." *Energies* 17.1 (2023): 177.
19. Radha, R., NaveenTaj, S., Mallikarjuna, G., Shanmugam, V., Radhika, C., Reddy, V. J., et al. "3D—As-A-Service: A revolutionary approach by LFX." *Illustrating Digital Innovations Towards Intelligent Fashion* (2024): 385–396.
20. Raith, P., Rausch, T., Furutanpey, A., and Dustdar, S. "FaaS-sim: A trace-driven simulation framework for serverless edge computing platforms." *Software: Practice and Experience* 53.12 (2023): 2327–2361.
21. Ramu, P., Thananjayan, P., Acar, E., Bayrak, G., Park, J. W., and Lee, I. "A survey of machine learning techniques in structural and multidisciplinary optimization." *Structural and Multidisciplinary Optimization* 65.9 (2022): 266.
22. Tardieu, H., Daly, D., Esteban-Lauzán, J., Hall, J., and Miller, G. *Deliberately Digital*. Cham: Springer International Publishing (2020).
23. Tariq, U. U., Ali, H., Liu, L., Panneerselvam, J., and Hardy, J. "Energy-efficient scheduling of streaming applications in VFI-NoC-HMPSoC based edge devices." *Journal of Ambient Intelligence and Humanized Computing* 12.11 (2021): 9991–10007.
24. Weier, M., Stengel, M., Roth, T., Didyk, P., Eisemann, E., Eisemann, M., et al. "Perception-driven accelerated rendering." *Computer Graphics Forum* 36.2 (2017): 611–643.
25. Yu, L. X., Amidon, G., Khan, M. A., Hoag, S. W., Polli, J., Raju, G. K., and Woodcock, J. "Understanding pharmaceutical quality by design." *The AAPS Journal* 16.4 (2014): 771–783.

Source of support: Nil; **Conflict of interest:** Nil.

Cite this article as:

Quintero, F. A. "Reducing Production Time without Compromising Quality: Optimization Strategies in High-End VFX Simulations." *Sarcouncil Journal of Engineering and Computer Sciences* 3.8 (2024): pp 1-8.