

# Planning and Optimization of Adaptive Milling Trajectories in Robotic Manipulators

*Ermatova Zarina*<sup>1</sup>

**Abstract:** In the era of advanced automation and smart manufacturing, robotic manipulators are increasingly employed for complex tasks such as adaptive milling. This paper presents an in-depth investigation into the planning and optimization of milling trajectories in robotic manipulators with an emphasis on adaptivity to surface topology and process dynamics. By integrating real-time sensor feedback and trajectory adjustment algorithms, our approach improves machining accuracy, tool life, and overall production efficiency. Simulation results and comparative performance metrics confirm the superiority of the proposed method over conventional fixed-path planning.

**Keywords:** Adaptive Milling, Robotic Manipulators, Trajectory Planning, Path Optimization, Surface Topology, Sensor Feedback, Smart Manufacturing.

## Introduction

Robotic manipulators have transformed modern manufacturing with their versatility, flexibility, and ability to perform precision tasks. In particular, their role in milling operations—traditionally dominated by CNC machines—has gained prominence due to the demand for greater flexibility and cost-effective automation.

Conventional milling techniques often rely on pre-programmed, rigid trajectories that do not account for variations in workpiece geometry or material inconsistency. This inflexibility can result in suboptimal surface finish, reduced productivity. Therefore, there is a growing need for adaptive trajectory planning methods that can dynamically respond to changes in surface topology, tool condition, and environmental disturbances.

This paper focuses on planning and optimizing adaptive milling trajectories in robotic manipulators by integrating real-time feedback and intelligent control strategies. The aim is to enhance machining performance while maintaining precision and safety in dynamic industrial environments.

In adaptive milling, trajectory flexibility plays a crucial role in dealing with uncertainties such as inconsistent surface geometry, thermal deformation, and tool deflection. These factors, if not properly managed, can lead to poor surface finish, dimensional inaccuracies, and increased wear on the tool and manipulator joints.

Robotic manipulators, compared to traditional CNC machines, offer enhanced degrees of freedom and greater spatial flexibility, which are essential for complex surface machining. However, this flexibility also introduces challenges in maintaining machining stability and precision, especially when operating in dynamic or semi-structured environments.

Recent advancements in real-time sensing technologies, such as force-torque sensors and laser profilometers, have enabled closed-loop control of machining operations. This paves the way for adaptive trajectory planning, where the robot can continuously update its toolpath based on sensor data and predefined optimization criteria.

The primary motivation behind this study is to develop a trajectory planning system that can intelligently adapt to surface variations while minimizing tool wear and energy consumption. This is

---

<sup>1</sup> Senior lecturer of Fergana state technical university  
ermatovazarinabonu@gmail.com



achieved by integrating real-time sensor feedback with predictive control algorithms and optimization strategies. The approach is designed to be scalable and suitable for a range of industrial applications, from aerospace component finishing to customized prosthetic manufacturing.

**Methodology.** Our proposed methodology involves three key components:

**System Architecture.** The system consists of a 6-DOF robotic manipulator, a high-speed milling spindle, and an integrated sensor suite including force-torque sensors and laser profilometers. These provide real-time data on cutting conditions and surface profiles.

#### Trajectory Planning Algorithm

A hybrid trajectory planning algorithm was developed using the following steps:

- Initial Path Generation: Based on CAD models using conventional CAM software.
- Surface Scanning: Laser profilometry captures the actual surface deviations.
- Adaptive Correction: The trajectory is modified in real-time using a feedback loop that accounts for tool deflection, surface deviation, and desired machining parameters.
- Optimization: A cost function is minimized based on energy consumption, surface quality, and machining time. This is solved using a genetic algorithm.

#### Control Strategy

A model-predictive control (MPC) approach is used to regulate the robot's joint velocities and end-effector force in response to sensory inputs, ensuring adaptive correction during milling.

#### Literature Review

Several previous works have addressed robotic milling and trajectory planning. For instance, Zhang et al. (2019) developed a method for real-time compensation of robotic tool path errors using force sensors. Similarly, Kim et al. (2020) implemented adaptive control for surface finishing with robots.

However, most studies rely on offline optimization or simplified kinematic models, lacking the real-time adaptability required for complex surface milling. This paper builds upon those foundations by combining real-time feedback, multi-objective optimization, and predictive control, offering a more robust solution for industrial applications.

#### Optimization Objective Function

The trajectory optimization is formulated as a multi-objective cost function, which considers three main criteria: surface accuracy, tool wear, and energy efficiency. The function is defined as:

$$J = \omega_1 \cdot E_{\text{surf}} + \omega_2 \cdot E_{\text{tool}} + \omega_3 \cdot E_{\text{energy}}$$

Where:

- $J$  — Total cost to minimize
- $E_{\text{surf}}$  — Surface error (difference between desired and actual profile)
- $E_{\text{tool}}$  — Tool wear estimate over the trajectory
- $E_{\text{energy}}$  — Estimated energy consumption
- $\omega_1, \omega_2, \omega_3$  — Weight coefficients representing the priority of each criterion (e.g.,  $\omega_1 + \omega_2 + \omega_3 = 1$ )

This optimization function is solved using a genetic algorithm, which iteratively updates the trajectory parameters to minimize  $J$ .



## Results

Simulation Setup. A simulation environment was developed in MATLAB/Simulink and ROS, using a UR5 robotic arm model. Workpieces with varying surface contours were processed under different scenarios:

- Static pre-planned path
- Adaptive feedback-driven path (proposed method)

## Performance Metrics

Metric	Conventional Path	Adaptive Path (Proposed)
Surface roughness ( $\mu\text{m}$ )	8.5	2.3
Tool wear ( $\text{mm}^3/\text{hour}$ )	0.45	0.18
Milling time (seconds)	180	155
Energy consumption (kWh)	0.72	0.53

## Discussion

The results clearly show that the adaptive approach reduces surface roughness by 73%, decreases tool wear by 60%, and saves 26% of milling time. These improvements validate the effectiveness of our real-time trajectory optimization model. The results of this study clearly demonstrate the advantages of using adaptive trajectory planning in robotic milling operations. Compared to traditional fixed-path methods, the proposed adaptive system achieved significantly improved surface finish, reduced tool wear, and lower energy consumption. These improvements are not only quantitatively measurable but also qualitatively evident in smoother operations and more consistent machining outcomes.

One of the most critical aspects highlighted by this research is the role of real-time feedback in enhancing machining precision. The integration of force-torque sensors and laser profilometers allows the robot to detect minor surface variations and dynamically adjust its toolpath. This closed-loop control reduces the cumulative effects of positioning errors and tool deflection, especially when milling complex or curved surfaces.

Furthermore, the use of a multi-objective optimization algorithm ensures a balanced trade-off between different performance metrics. While some traditional methods prioritize only geometric accuracy or speed, this approach simultaneously considers surface quality, tool life, and energy efficiency. The weight-based cost function provides flexibility, allowing system operators to adjust priorities according to specific production needs.

Compared with existing studies in the literature, our framework outperforms many conventional and semi-adaptive methods. For instance, while Zhang et al. (2019) focused on error compensation using force sensors, their model lacked a predictive control component. Our integration of Model Predictive Control (MPC) enables anticipation of system responses, providing smoother and more efficient trajectory updates. Similarly, unlike approaches that rely solely on offline trajectory correction, our method supports online, real-time adaptation, which is crucial in high-precision manufacturing environments.

From an engineering standpoint, the application of robotic manipulators in milling tasks presents both opportunities and challenges. Their kinematic flexibility enables operations in confined or irregular spaces, but also requires advanced control systems to manage instability and oscillation during high-speed machining. The proposed system addresses this through adaptive velocity control, joint trajectory smoothing, and vibration mitigation strategies.

Another important discussion point is the scalability and industrial applicability of the approach. The modular nature of the framework allows for integration with various robotic arms, sensors, and tool types, making it adaptable to a wide range of applications—from aerospace and automotive parts to biomedical implants and artistic engraving.



However, several limitations must be acknowledged. First, the system currently relies on pre-trained optimization parameters, which may not generalize well to all material types or tool geometries. Second, the response time of the adaptation loop is constrained by sensor latency and computation speed. Future implementations could benefit from edge computing and GPU acceleration to further reduce reaction time.

Finally, while the simulations provide strong validation of the proposed method, physical experimentation on actual robotic hardware will be necessary to fully assess durability, long-term reliability, and environmental robustness. These steps are planned as part of future work.

## Conclusion

This paper presented a comprehensive framework for adaptive trajectory planning and optimization in robotic milling applications. By leveraging sensor integration, hybrid planning algorithms, and predictive control, the system significantly improves machining quality and efficiency.

Future work includes experimental validation on physical setups and integration of AI-based learning mechanisms to predict optimal milling parameters based on historical data. Moreover, the integration of sensor feedback with real-time control strategies enables the robotic system to respond dynamically to machining disturbances, which significantly improves reliability and adaptability. This ensures consistent product quality, especially in scenarios involving variable workpiece geometries or challenging material properties.

The proposed framework also supports scalability, making it applicable not only to small-scale manufacturing cells but also to large-scale automated production lines. As manufacturing continues to evolve towards Industry 4.0, systems that incorporate adaptive intelligence, such as the one proposed in this study, will become increasingly vital.

In future research, the incorporation of machine learning algorithms could further enhance the system's ability to predict optimal parameters based on historical performance data. Additionally, extending this approach to multi-robot coordination in collaborative milling tasks may open new possibilities in flexible and distributed manufacturing environments.

## References

1. Zhang, H., Li, X., & Wang, J. (2019). Real-time compensation for robotic milling errors using force sensors. *International Journal of Advanced Manufacturing Technology*, 102(1-4), 123–135.
2. Kim, D., Park, Y., & Lee, C. (2020). Adaptive robotic surface finishing using online correction and force feedback. *IEEE Transactions on Industrial Electronics*, 67(6), 4890–4900.
3. Tang, L., & Zhao, R. (2018). Milling path planning for freeform surfaces using hybrid optimization. *Robotics and Computer-Integrated Manufacturing*, 54, 50–59.
4. Liu, Y., & Chen, J. (2021). A review on robot-based machining systems: Architecture, control, and applications. *Robotics*, 10(1), 8.
5. ISO 9283:1998. (1998). *Manipulating industrial robots – Performance criteria and related test methods*.
6. Rajput, R.K. (2021). *Industrial automation and control*. S. Chand Publishing.
7. Ahror K., Shohjahon U. Analysis of a program that provides additional services for visitors to resorts //Miasto Przyszłości. – 2024. – T. 48. – C. 264-269.
8. Yusubjanovich S. N., Muminjonovich K. A. Trikotaj to 'qimalarining shakl saqlash xususiyatlarini raqamli baholash usullari //Al-Farg'oniy avlodlari. – 2024. – T. 1. – №. 1. – C. 57-61.
9. Эрматова З. К. Актуальность преподавания языка программирования с++ в высших учебных заведениях //Al-Farg'oniy avlodlari. – 2023. – T. 1. – №. 4. – C. 237-241.



10. Хайитов А. М. МЕХАТРОННАЯ СИСТЕМА УПРАВЛЕНИЯ ДВИЖЕНИЕМ МАНИПУЛЯТОРА ПРИ ОКРАСКЕ ИЗДЕЛИЙ //Universum: технические науки. – 2025. – Т. 2. – №. 5 (134). – С. 21-23.
11. O'G, Hayitov Azizjon Mo'Minjon. "ROBOT MANIPULYATORLARIDA NOANIQ YUZALARNI AVTOMATIK PARDOZLASHNI TAKOMILLASHTIRISH UCHUN ADAPTIV TRAEKTORIYA SHAKLLANTIRISH ALGORITMLARI." *Al-Farg'oniy avlodlari* 1.2 (2025): 175-181.
12. Hayitov A., Kayumov A. CURRENT AND FUTURE CHALLENGES OF SOFT-WARE ENGINEERING FOR ANDROID APPLICATIONS //Conference on Digital Innovation: "Modern Problems and Solutions".–2023.
13. Hayitov A., Kayumov A. ANDROID APPLICATION TESTING AND CODE RE-FACTORING //Conference on Digital Innovation: "Modern Problems and Solutions".–2023.
14. Kayumov A. TECHNOLOGIES OF TECHNICAL MACHINE EXPERTISE //Journal of technical research and development. – 2023. – Т. 1. – №. 3. – С. 96-99.
15. Kayumov A. СОЗДАНИЕ НА ОСНОВЕ ЭКСПЕРТНОЙ СИСТЕМЫ ПРОГРАММЫ ОЦЕНКИ ЭФФЕКТИВНОСТИ ТЕКСТИЛЬНЫХ МАШИН //Потомки Аль-Фаргани. – 2023. – Т. 1. – №. 2. – С. 49-52.
16. Kayumov, A., & Maxamadjonov, N. MOBIL QURILMALAR ISHLAB CHIQISH FANINI O 'QITISHDA SUNIY INTELLEKTNING RO 'LI. In *Conference on Digital Innovation: "Modern Problems and Solutions"*.–2023.
17. Qaxramonovna E. Z. et al. TO'QIMACHILIK SANOATIDA ZAMONAVIY TEXNOLOGIYALAR VA ULARNING DASTURIY TA'MINOTI //FAN, TA'LIM, MADANIYAT VA INNOVATSIYA JURNALI| JOURNAL OF SCIENCE, EDUCATION, CULTURE AND INNOVATION. – 2024. – Т. 3. – №. 5. – С. 38-46.
18. Qaxramonovna E. Z. et al. LINTER QURILMASINING TUZILISHI VA ISHLASH TEXNOLOGIYASI //FAN, TA'LIM, MADANIYAT VA INNOVATSIYA JURNALI| JOURNAL OF SCIENCE, EDUCATION, CULTURE AND INNOVATION. – 2024. – Т. 3. – С. 5.
19. Эрматова З. К. АКТУАЛЬНОСТЬ ПРЕПОДАВАНИЯ ЯЗЫКА ПРОГРАММИРОВАНИЯ C++ В ВЫСШИХ УЧЕБНЫХ ЗАВЕДЕНИЯХ //Al-Farg'oniy avlodlari. – 2023. – Т. 1. – №. 4. – С. 237-241.
20. Ermatova Z. ZAMONAVIY DASTURIY MAHSULOTLAR YARATISH VA SIFATINI YAXSHILASHDA DASTURLASH TILLARINI O 'QITISHNING O 'RNI //Research and implementation. – 2023.
21. Каюмов А. М. Определение параметров испытаний в зависимости от типа трикотажного полотна //Universum: технические науки. – 2025. – Т. 3. – №. 3 (132). – С. 30-32.
22. Шарибаев Н. Ю., Каюмов А. М. Численная оценка формосохраняющих свойств тканей посредством анализа изображений //Universum: технические науки. – 2024. – Т. 4. – №. 3 (120). – С. 33-36.
23. Зулунув Р. М., Каюмов А. М. Идентификация и сортировка текстиля для автоматизированной обработки с помощью ближней инфракрасной спектроскопии //Universum: технические науки. – 2024. – Т. 1. – №. 3 (120). – С. 38-41.
24. Kayumov A., Sobirov M., Musayev K. Methods of fabric defect detection using expert systems-a systematic literature review //E3S web of conferences. – EDP Sciences, 2024. – Т. 538. – С. 04015.
25. Zulunov R. et al. Detecting mobile objects with ai using edge detection and background subtraction techniques //E3S Web of Conferences. – EDP Sciences, 2024. – Т. 508. – С. 03004.

