

A Roadmap for AI Applications in the Fastener and Fastening Tools Industry

數位智造的啟航-
AI在緊固工具與扣件業的應用藍圖

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I. Introduction: From Experience-Driven to Data-Driven

The fastener and fastening tools industry has long built its competitive advantage on mature manufacturing processes, a comprehensive supply chain, and customization capabilities. However, in the face of product premiumization, tight market lead times, rising quality demands, and supply chain restructuring, traditional operational models—which rely on manual experience, scheduling, and inspection—are now reaching their limits. The value of AI lies not in replacing human labor, but in transforming the expertise of departments such as sales, R&D, production, quality assurance, and warehousing into data that can be analyzed, predicted, and continuously optimized. From intelligent quoting, drawing analysis, generative design, production scheduling, forging force monitoring, and AOI visual inspection, to smart logistics and AI-powered customer service, AI is gradually integrating the entire value chain. Fastening tools and fasteners both fall under the category of fastening products, but there are some differences in the focus of AI implementation. Fastener manufacturing centers on optimizing processes such as cold forging, heat treatment, tooling, and quality inspection; fastening tools, on the other hand, focus on product design, assembly, functional testing, and after-sales service. However, both sectors can leverage AI to integrate ERP, MES, AOI, Vision AI, and corporate knowledge bases, thereby establishing an AI-driven smart manufacturing system that spans order processing, design, production, quality assurance, logistics, and customer service. The value chains for AI applications in the fastener tools and fastener industries are summarized in **Table 1**.

Table 1. Overview of AI Applications in the Fastening Tools and Fasteners Industry

Manufacturing Value Chain	Current Condition	AI Module	AI Technology	Expected Benefits
Order Processing and Business Operations	Quotations rely on manual processing, and order information is scattered	AI Quotations, Customer Service, Email Analysis, Order Forecasting & Comparison	LLM, OCR, RAG, Similar Order Search	Improving Order Processing Efficiency and Response Times to Customers
R&D	Mold design relies on experience, and managing drawings can be challenging	AI-Generated Design, Mold Design, Drawing Search, AI Knowledge Base, AI BOM	Multimodal Models, Vector Retrieval, CAD/CAE Integration	Shorten design and development timelines
Production Management	Scheduling and material planning rely heavily on production planning.	AI Scheduling, Data Analysis, AI ERP Decision-Making, AI Smart Middle Platform, AI Production Utilization Analysis	APS, Optimization Algorithms, Digital Twin	Improving Utilization Rates and On-Time Delivery
Molding Process	Mold adjustment and equipment monitoring rely on experienced master craftsmen	AI Mold Adjustment Guidance, AI Forging Force Monitoring, AI Mold Maintenance and Calibration, Sensory Force Analysis	Sensors, Edge AI, Machine Learning	Stabilize process quality and reduce unplanned downtime

Manufacturing Value Chain	Current Condition	AI Module	AI Technology	Expected Benefits
Heat Treatment and Process Control	Lack of real-time analysis for process parameter adjustments	AI SPC, Process Parameters, Quality Prediction, Anomaly Alerts	Time Series Analysis, Forecasting Models, Statistical Process Control	Improving the Consistency of Heat Treatment Quality
Quality Management	Relying on manual sampling and visual inspections	AOI, Visual Inspection, Metallographic Inspection, Defect Detection, AI Defect Classification, Quality History	Vision AI, Deep Learning, Image Classification	Improving Testing Efficiency and Traceability Capabilities
Warehousing and Logistics	Limitations on inventory, picking, and shipping efficiency	AI Inventory Management, AI Smart Shelving, AI Logistics Optimization, Wire Management, AI Picking	AMR Scheduling, Path Optimization, AMR	Reducing Inventory Costs and Logistics Lead Times
Integrated Cross-Domain	System data is independent of one another	AI Agent, Cross-System Integration	Agent, API, RAG, Knowledge Graph	Establish a Cross-Departmental Smart Decision-Making Platform
Note: LLM (Large Language Model), OCR (Optical Character Recognition), RAG (Retrieval-Augmented Generation), API (Application Programming Interface), AMR (Autonomous Mobile Robot), AOI (Automated Optical Inspection)				

II. The AI Revolution in Business and R&D

a. AI-Powered Automated Order Processing and Smart Quotation System

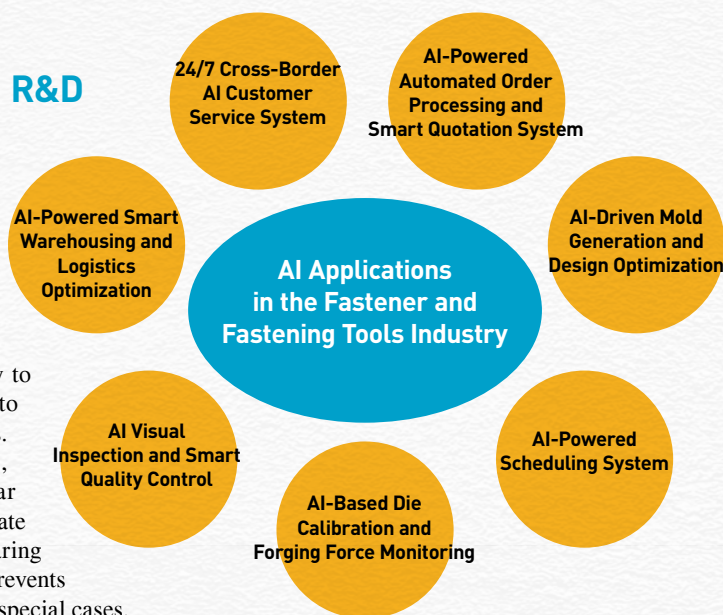
1. From Analyzing Order Requirements in Emails to Smart Quoting: The most time-consuming part of quoting fasteners is interpreting information such as specifications, materials, quantities, delivery times, and certifications from customer emails. AI can combine OCR, multimodal models, and field extraction technology to automatically convert drawings, attachments, and text into structured data, and integrate with ERP or quoting systems. By combining this with historical quotes, material costs, manufacturing labor hours, and inventory data, a “similar order search + cost model” is established to quickly estimate reasonable prices and assess gross margin risks. By comparing similar cases to provide suggested quotes, this approach prevents cost underestimation and allows staff to focus on analyzing special cases.

2. Reducing quotation time and increasing order intake: The benefits of AI implementation can be quantified. In practice, AI is used to analyze emails and compare them with historical orders, shifting the first round of quotations from manual review to AI-assisted processing, which significantly reduces lead time; Customer response time was reduced from 12 hours to 4 hours, the quote success rate increased from 63% to 70%, and the customer complaint resolution rate rose from 50% to 80%, effectively improving response speed, quote accuracy, order intake, and customer trust.

b. AI-Driven Mold Generation and Design Optimization

1. Accelerating Mold Development and Process Planning with Generative Design: Generative AI can analyze fastener drawings and integrate materials, dimensional tolerances, forming conditions, and equipment capabilities to help generate recommendations for mold design, forming processes, and machining methods. In practice, by establishing a mold database and using multimodal models to learn drawing features and process correlations, the system automatically generates machining steps, design guidelines, and preliminary BOMs, while estimating process requirements such as materials, molds, heat treatment, and surface treatment. This significantly reduces manual setup time and the risk of omissions; Generative AI can simultaneously consider design requirements, resources, and manufacturing constraints, making it suitable for the development of high-mix, low-volume products. After implementation, design and materials management processes are centralized, shortening project development timelines by approximately 20%.

2. Establishing a Mold Knowledge Base and Drawing Management System: AI can integrate drawings, modification records, mold service life, failure causes, and craftsmen's experience to create a searchable knowledge base. Through vector databases and RAG technology, designers can quickly search for historical cases by keyword, drawing, or defect, accelerating the transfer of technical knowledge and reducing the risk of rework caused by using outdated drawings or incorrect manufacturing methods.



III. Core AI Applications in Production and Quality Control

a. AI-Powered Scheduling System

1. Integrating Process Information with Dynamic Scheduling Management: Fastener scheduling involves integrated management across processes and resources, covering stages from coil to material feeding, heading, thread rolling, heat treatment, surface treatment, and packaging. Any delay at any station will affect delivery times. AI scheduling first integrates information on orders, materials, machines, molds, and external suppliers to establish a unified data platform, then links with ERP, CRM, and production data to optimize scheduling. In the face of rush orders, material shortages, equipment malfunctions, or delays from external suppliers, the system can instantly reschedule work orders and start times based on real-time shop floor conditions. By combining this with Digital Twin technology to simulate production status, it rapidly evaluates alternative solutions, transforming scheduling into a dynamic management process that is continuously optimized.

2. Improving Equipment Utilization and Production Decision-Making Efficiency: AI-powered intelligent scheduling can effectively improve equipment utilization, material availability, and on-time delivery rates. In practice, after implementation, equipment utilization increased from 85% to 95%, material availability in scheduling rose from 80% to 88%, capacity utilization improved from 65% to 85%, and the time required to generate reports was reduced from 2 hours to 5 minutes. AI scheduling does not replace production planning; rather, through data integration, anomaly alerts, and simulation analysis, it helps production planners transition from reactive troubleshooting to proactive decision-making.

b. AI-Based Die Calibration and Forging Force Monitoring

1. Optimizing forming processes by combining sensors with forging force analysis: Processes such as heading, thread rolling, and tapping have long relied on master craftsmen to make judgments based on sound, vibration, temperature rise, and forming conditions. The key to AI implementation lies in using external sensors to collect process signals—such as pressure, displacement, vibration, temperature, and current—and correlating them with product dimensional data and defect records, thereby enabling a phased transition to digitalization without the need for a complete equipment overhaul. The force waveform during forging reflects process stability. When material anomalies, die wear, positioning deviations, or insufficient lubrication occur, AI can analyze the signals in real time to identify abnormalities. Combined with optical measurement, geometric calculations, and equipment compensation mechanisms, it provides die adjustment recommendations, transforming the die adjustment process from an experience-driven approach to data-driven management.

2. Shifting from post-production inspection to in-process prevention: Traditional quality control often identifies quality issues only after a product is completed, which can easily lead to waste in downstream processing and materials. By integrating AI with forging force monitoring, optical measurement, and model analysis, real-time alerts for process deviations can be issued during the heading or thread rolling stages, elevating quality management from “post-production inspection” to “in-process prevention.” At the same time, by incorporating predictive maintenance mechanisms, equipment failure trends can be identified in advance, reducing tooling costs, downtime, and the risk of batch scrapping.

c. AI Visual Inspection and Smart Quality Control

1. AI Identifies Product Defects and Improves Inspection Efficiency: The trend in fastener quality inspection is toward AOI combined with AI. AI compensates for the limitations of rule-based inspection in recognizing complex textures and multiple defects. It performs image classification and anomaly detection for cracks, burrs, missing teeth, forming abnormalities, surface scratches, and plating defects, and can also be applied to metallographic structure identification. In practice, after implementing AI-based metallographic analysis, inspection time has been reduced from 1–2 hours to 15–30 minutes, labor requirements have decreased by 70%–80%, and human interpretation errors have dropped from 15%–20% to 2%–3%. Combining forging force monitoring with AI visual inspection can further reduce the missed-defect rate from 10% to below 5%. The key performance indicators before and after the implementation of AI in fastener and fastening tool manufacturing processes are summarized in *Table 2*.

2. Establishing Real-Time Quality History and Traceability Capabilities: By combining AI inspection with process monitoring, it is possible to integrate process parameters, mold status, inspection results, and anomaly records to establish a comprehensive quality history and traceability mechanism. This capability meets the quality traceability requirements of high-reliability industries such as automotive, aerospace, and semiconductors, and helps incorporate AI analysis results into quality management and continuous improvement processes.

Table 2. Pre- and Post-Implementation Performance Indicators for AI Adoption in Fastener Tooling and Fastener Manufacturing Processes

Indicator	Pre-Implementation	Post-Implementation	Performance
Quote Response Time	~12 hrs	~4 hrs	Faster quoting process, improved timeliness in securing orders
Quote Success Rate	~63%	~70%	Understanding specifications leads to more accurate quotations
Customer Complaint Resolution Rate	~50%	~80%	Real-time customer service responses and status updates



Indicator	Pre-Implementation	Post-Implementation	Performance
Equipment Utilization Rate	~85%	~95%	More stable scheduling and resource utilization
Scheduling Material Availability Rate	~80%	~88%	More accurate material preparation and start times
Capacity Utilization Rate	~65%	~85%	AI scheduling and utilization analysis to address bottlenecks
Product Yield Rate	~95%	~99%	Moving process monitoring upstream to reduce batch defects
Missed-detection rate	~10%	<5%	AOI and AI recognition reduce omissions
Metallographic Inspection Time	1-2 hrs	15-30 mins	Significantly faster assessment of heat treatment quality
Errors in Manual Interpretation	15-20%	2-3%	Metallographic expert knowledge stabilized in a model
Days of Inventory Turnover	~75 days	~52 days	Order forecasting and inventory level optimization
Total Cost from Order Receipt to Delivery	Reference value 100	~85-90	Total production costs have decreased by approximately 10-15%

IV. AI Applications in Logistics and Customer Service

a. AI-Powered Smart Warehousing and Logistics Optimization

1. Dynamic Planning of Shipping Routes and Inventory Management: In the fastener logistics industry, the most common challenges are frequent rush orders, a wide variety of items, and scattered information—not a lack of warehouse space. When handling rush orders, consolidated shipments, or early shipments, manual scheduling can easily lead to conflicts in picking, container loading, and shipping sequences. AI can integrate order priorities, storage locations, packaging requirements, ERP inventory data, and shipping schedules to plan optimal shipping routes. In practice, after implementing AI-driven inventory management and order forecasting, inventory accuracy improved to 95%, excess inventory levels decreased by approximately 30%, and inventory turnover days were reduced from 75 to 52, effectively improving material flow and supply chain management.

2. Reduce the risk of shipping errors and enhance supply chain resilience: When fastener products are mixed up, missing, or mislabeled, it can easily lead to rework, shipping costs, and customer complaints. AI can integrate information on orders, warehouse locations, picking, packaging, container loading, and shipping to verify quality inspection status, packaging requirements, and container consolidation needs in advance, thereby reducing the risk of shipping errors and establishing end-to-end logistics traceability.

b. 24/7 Cross-Border AI Customer Service System

1. AI-powered customer service addresses cross-time zone needs: Since the European and American markets are the primary customers for fasteners, time zone differences often cause delays in responses; AI customer service can prioritize routine tasks such as delivery date inquiries, order status updates, product specifications, and common customer complaints. Once the system is integrated with ERP, MES, and shipping systems, order, production, and shipping statuses can be checked in real time. Combined with RAG and the corporate knowledge base, this enables centralized handling of customer complaints, quality assessment criteria, and document templates, thereby enhancing cross-time-zone service.

2. Improving Customer Complaint Resolution Efficiency and Customer Satisfaction: Practical case studies show that after implementing 24/7 AI customer service, the customer complaint resolution rate increased from 50% to 80%. AI customer service integrates order status and quality records, streamlines processing workflows, and enhances information transparency.

V. Summary: The Integrated Benefits of AI Adoption

a. Moving from isolated improvements to end-to-end optimization: Start by implementing AI in high-impact areas such as intelligent quoting, drawing analysis, scheduling, and quality inspection, then gradually integrate it across forming, heat treatment, quality assurance, logistics, and customer service. Once data across these systems is interconnected, AI can evolve from a standalone tool into a cross-departmental decision-making platform, continuously building up the company's knowledge and data assets.

b. Reducing Total Production Costs from Order Receipt to Delivery: Practical case studies show that AI can simultaneously improve quoting efficiency, shorten development timelines, optimize scheduling, increase yield rates, reduce inspection oversights, improve inventory management, and accelerate the resolution of customer complaints, resulting in a reduction of approximately 10% to 15% in total production costs from order receipt to delivery. The adoption of intelligent solutions has increased output value, reduced operating costs, and driven corporate investment. ■

Next Article Focus: As the roadmap for AI applications becomes more comprehensive, the challenges of implementation also become more apparent. These include the digitization of legacy equipment, the structuring of experiential knowledge, ERP/MES integration, and issues related to organizational culture and talent development. Most of these challenges stem from hardware, data, processes, and organizational factors, rather than AI technology itself. The next article will further explore the practical challenges and feasible solutions for implementing AI in fastening tools and fasteners.

