



Cost Modeling and Economic Analysis of Machining Processes

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Abstract — The economics of machining processes has remained a central concern in manufacturing engineering for decades, particularly as global competitive pressures demand tighter cost controls without sacrificing component quality or output rates. This paper presents a structured analysis of cost modeling frameworks applicable to common machining operations including turning, milling, drilling, grinding, and computer numerical control (CNC) machining. The study examines the principal cost constituents — material, labor, tooling, machine overheads, energy consumption, inspection, and scrap — and proposes analytical methods for estimating total production cost per component. Mathematical formulations derived from established machining economics literature, including Taylor's tool life equation, and are employed to evaluate the trade-off between cutting speed, tool life, and associated costs. Economic performance indicators such as break-even point, return on investment (ROI), and cost-benefit ratios are discussed in the context of both small-batch and large-volume production scenarios. Two case studies, one drawn from automotive component manufacturing and another from aerospace machining, illustrate how cost modeling can guide practical optimization decisions. The findings demonstrate that systematic cost estimation, when integrated with process parameter selection, can reduce total machining costs by 15–30% in typical production environments. This paper aims to serve as a practical reference for engineering

students, production engineers, and manufacturing planners engaged in process cost analysis.

Key Words: Machining economics, cost modeling, Taylor's tool life equation, CNC machining, manufacturing cost analysis, tooling cost, break-even analysis, production optimization.

1. INTRODUCTION

Manufacturing is, at its core, an economic activity. The ability to shape raw materials into functional components with dimensional accuracy and acceptable surface quality has always been judged not only by the technical outcome but by what it costs to achieve that outcome. Machining — which encompasses material removal operations like turning, milling, drilling, boring, and grinding — remains the most widely employed family of manufacturing processes in engineering industries worldwide. Despite the growing adoption of near-net-shape forming processes and additive manufacturing techniques, machining continues to be indispensable for finishing, tight-tolerance features, and hard materials where other methods fall short.

The economic significance of machining is substantial. In the automotive sector alone, it has been estimated that machining operations account for 15–20% of the total production cost of a typical engine block or transmission



Housing. In aerospace manufacturing, where materials like titanium alloys, Inconel, and carbon fiber composites demand specialized tooling and slower cutting conditions, machining costs can constitute an even larger share of part cost. Given these figures, it is hardly surprising that manufacturing engineers and researchers have invested considerable effort in developing systematic methods for estimating, predicting, and reducing these costs.

Cost modeling in machining refers to the process of quantitatively representing the various cost components associated with a machining operation or a set of operations. Unlike a simple cost accounting exercise, a proper cost model links process parameters — cutting speed, feed rate, depth of cut, tool geometry — to the economic outcomes of the process. This connection is important because a change in cutting speed, for instance, affects not just machining time but also tool wear rate, surface finish, energy consumption, and ultimately the frequency at which tools must be replaced. Each of these effects has a monetary consequence that feeds back into the total cost per part.

Historically, the economic analysis of machining was dominated by a relatively straightforward optimization problem: find the cutting speed that minimizes cost per component or maximizes production rate. The famous Taylor's tool life equation, published in 1907, provided the theoretical backbone for this analysis and continues to be taught in manufacturing engineering courses as a fundamental tool. Over subsequent decades, the problem has grown considerably more complex. Multi-axis CNC machining centers, automated tool changers, high-pressure coolant systems, in-process inspection, and integrated manufacturing cells have introduced many additional cost contributors that a simple single-variable optimization cannot fully capture.

Modern manufacturing environments also operate under tighter constraints. Just-in-time production schedules require reliable cycle time estimates. Environmental regulations impose additional costs through coolant disposal, energy efficiency requirements, and limits on certain cutting fluids. Demand variability forces production planners to frequently switch between product families, making setup costs more significant than they were in the era of long, stable production runs. All of these factors have intensified interest in comprehensive, multi-factor cost models that go beyond the classical approach.

The objective of this paper is to present a detailed review and analysis of cost modeling methods applicable to machining processes, with emphasis on practical application. The paper covers the principal cost components involved in machining, the mathematical relationships that govern them, and the economic analysis techniques used to evaluate and compare different machining strategies. Two industrial case studies are presented to ground the theoretical discussion in realistic scenarios. The paper also discusses emerging trends — including smart manufacturing, digital twins, and AI-assisted optimization — that are beginning to reshape how machining economics is approached in modern production facilities.

The scope of this work is primarily focused on conventional machining (turning, milling, drilling, grinding) and CNC machining, which together represent the dominant modes of material removal in engineering production. Specialized non-conventional processes such as electrical discharge machining, laser cutting, and water jet machining are briefly mentioned in the context of process comparison but are not analyzed in depth. The intended audience includes

undergraduate and postgraduate students in mechanical and manufacturing engineering, practicing engineers in production and process planning roles, and researchers working on manufacturing productivity and cost analysis.

Section 2 provides an overview of the principal machining processes and their cost characteristics. Section 3 develops the cost modeling framework in detail, including formulas for each cost component. Section 4 presents the economic analysis methodology, covering break-even analysis, ROI calculations, and cost-benefit comparisons. Section 5 focuses specifically on tool life economics and the classical optimization problem. Section 6 addresses CNC machining economics and the impact of automation. Sections 7 and 8 present the case studies and future trends respectively, followed by the conclusions.



2. OVERVIEW OF MACHINING PROCESSES

Before developing cost models, it is worth briefly characterizing the principal machining processes in terms of their operating principles, typical applications, and the economic factors that make each process distinct. The following subsections provide this overview.

2.1 Turning

Turning is performed on a lathe and involves the rotation of the work piece against a stationary single-point cutting tool. The cutting tool traverses axially along the work piece, removing material in a helical chip. Turning is used to produce cylindrical external and internal surfaces, tapered profiles, and threads. It is among the most common machining operations in engineering manufacture — virtually every shaft, spindle, axle, and pin in an automobile or machine tool passes through a turning operation.

From a cost perspective, turning is generally efficient because of its continuous cutting action and high material removal rates. However, tool life is a dominant cost driver because the tool remains in constant contact with the work piece for the entire pass. At higher cutting speeds, thermal loads increase sharply, and tool wear accelerates. The cost-optimal cutting speed for turning must therefore balance productivity against tooling expenditure.

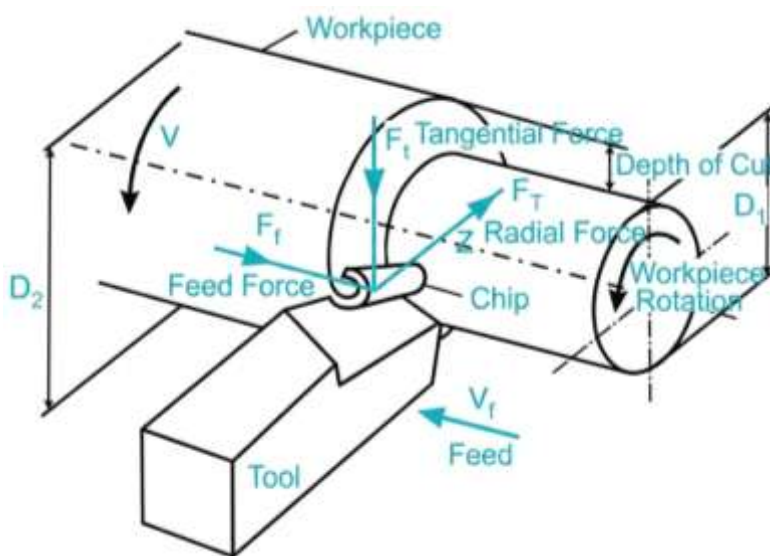


Fig. 1: Schematic of a Single-Point Turning Operation on a Centre Lathe

2.2 Milling

Milling employs a rotating multi-tooth cutter to remove material from a work piece that is fed against it. The interrupted cutting action — each tooth engages the work piece briefly and then exits — gives milling a significantly different thermal and mechanical profile compared to turning. Face milling, peripheral (slab) milling, end milling, and slot milling are the most common variants. Applications span from flat surface generation on large structural components to the intricate pockets and contours of mould and die cavities.

The economics of milling are somewhat more complex than turning because multiple cutting edges are involved. While multi-tooth cutters can achieve high material removal rates, the interrupted cut also means each tooth experiences periodic impact loading, which can cause chipping in brittle tool materials. The cost per cutting edge, insert indexing frequency, and cutter body investment are all relevant cost items that a milling cost model must include.

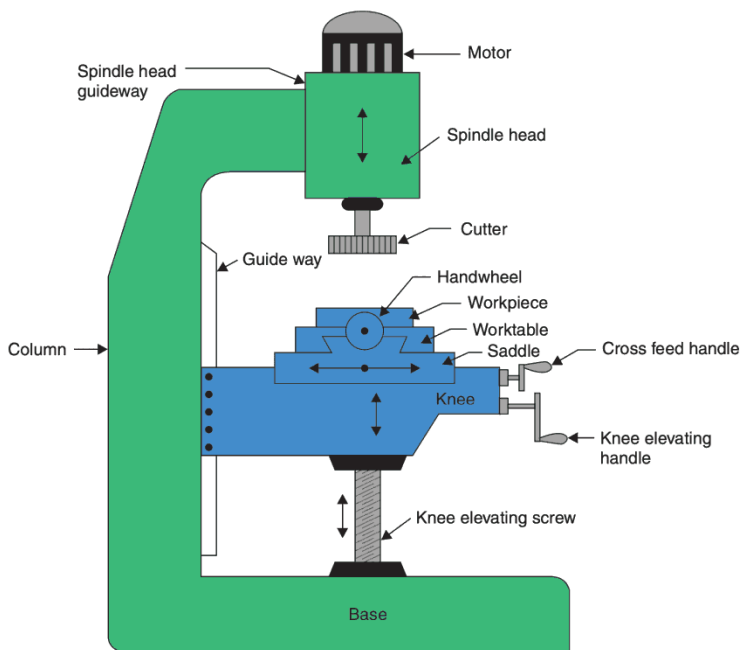


Fig. 2: Face Milling and End Milling Operations — Tool and work piece Engagement

2.3 Drilling

Drilling is one of the most frequently performed machining operations, used to create round holes in work pieces by rotating a multi-fluted drill against the stationary or slowly rotating part. While seemingly straightforward, drilling presents its own cost challenges: drill breakage in difficult materials, chip evacuation in deep holes, tool regrinding economics, and the dimensional accuracy limitations of standard twist drills all need to be considered. In high-volume production — for example, engine block oil passages or printed circuit board vias — drilling cost per hole can become a critical optimization variable.

2.4 Grinding

Grinding uses an abrasive wheel to remove small amounts of material from a work piece surface, producing fine surface finishes and tight dimensional tolerances that are not achievable with single-point or multi-point cutting tools in a single pass. Cylindrical grinding, surface grinding, and internal grinding are the main categories. Grinding wheel cost (wheel wear, dressing frequency, and wheel replacement) and the relatively low material removal rate are the primary economic concerns. Grinding is typically a finishing operation and is often evaluated as a cost-added step after rough and semi-finish turning or milling.

2.5 CNC Machining

Computer numerical control (CNC) machining encompasses any of the above processes performed on machines whose axes are controlled by pre-programmed numerical commands. CNC turning centers, vertical and horizontal machining centers, and five-axis machining centers allow complex geometries to be produced with minimal operator intervention. The economic case for CNC machining rests on its ability to reduce setup time, eliminate operator skill variability, integrate tool management, and maintain consistent quality over long production runs. Capital investment in CNC equipment is substantially higher than conventional machines, and this must be factored into the hourly machine rate used in cost calculations.

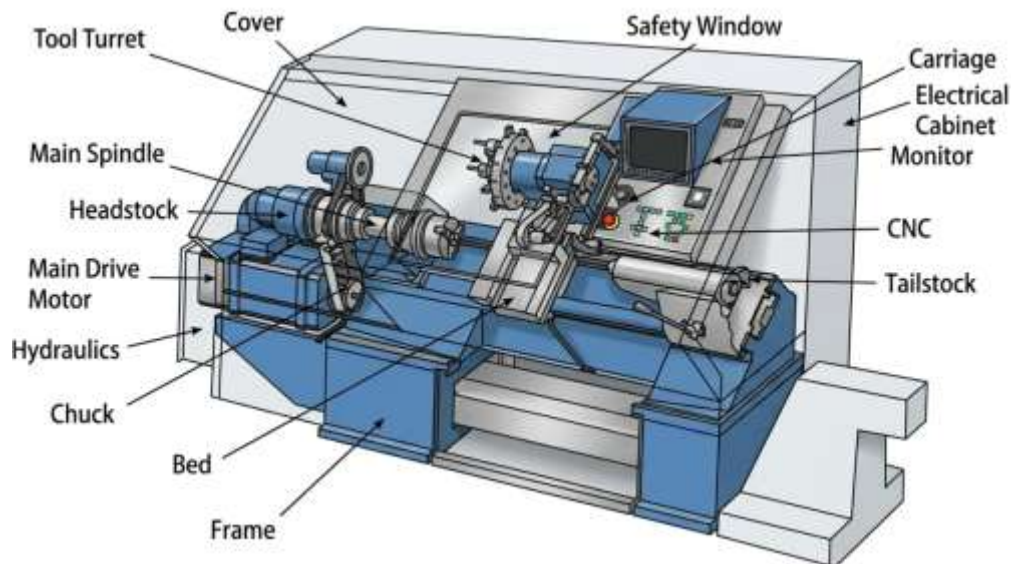


Fig. 3: Typical 3-Axis CNC Vertical Machining Center — Major Components

Table 1: Comparison of Principal Machining Processes

Process	Material Removal Rate	Surface Finish (Ra)	Primary Cost Driver	Typical Applications
Turning	High	0.8–6.3 μm	Tool wear / cycle time	Shafts, pins, spindles
Milling	Medium–High	0.8–3.2 μm	Cutter cost / setup time	Flat surfaces, pockets, contours
Drilling	Medium	1.6–6.3 μm	Drill breakage / depth	Holes in structural parts
Grinding	Low	0.025–0.8 μm	Wheel dressing / low MRR	Precision finishing
CNC Machining	High	0.4–3.2 μm	Machine hourly rate / programming	Complex geometries, automation

3. COST MODELING IN MACHINING

Cost modeling in machining is the systematic representation of all expenditures associated with producing a component through material removal operations. A well-constructed cost model serves multiple purposes: it enables comparison between alternative process routes, supports make-or-buy decisions, provides input data for quotation and pricing, and identifies cost reduction opportunities. The general structure of a machining cost model can be represented as a sum of individual cost elements, each driven by specific process variables.



Machining Cost Flowchart

Breakdown of Total Cost per Component

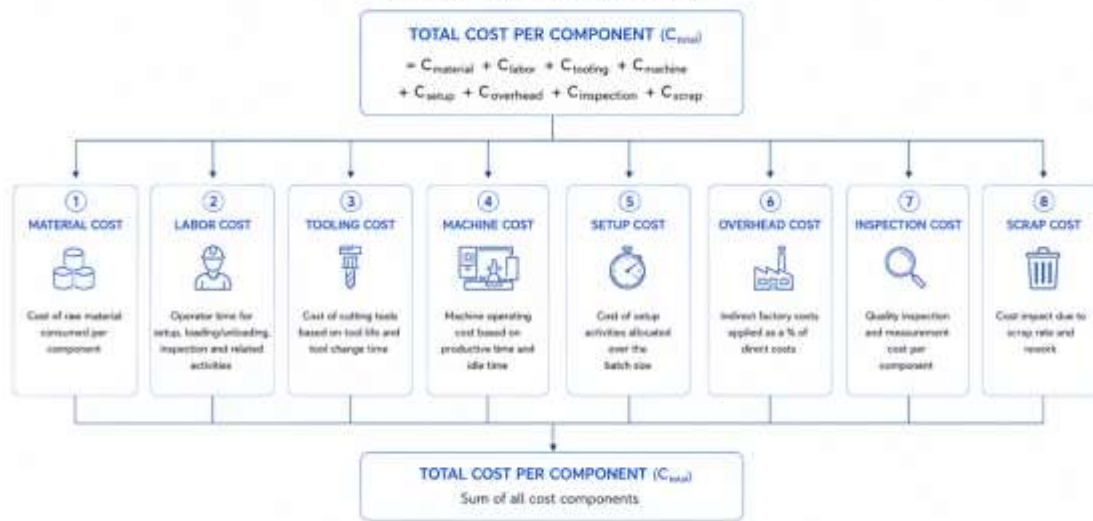


Fig. 4: Flowchart Showing the Breakdown of Machining Cost Components

The total cost per component (C_{total}) is commonly expressed as:

$$C_{total} = C_{material} + C_{labor} + C_{tooling} + C_{machine} + C_{setup} + C_{overhead} + C_{inspection} + C_{scrap}$$

3.1 Material Cost

Material cost is typically the most straightforward element to estimate. It is computed as the cost of raw material consumed per component, accounting for the buy-to-fly ratio — the ratio of starting billet weight to final component weight. In aerospace machining, this ratio can exceed 10:1 for difficult-to-cut alloys, meaning more than 90% of the expensive raw material is removed as chips. The material cost per component is:

$$C_{material} = M_{billet} \times C_{unit_material} + C_{cutting_fluid}$$

where M_{billet} is the mass of the starting material (kg) and $C_{unit_material}$ is the material cost per kilogram. Cutting fluid costs, while sometimes treated as overhead, can be allocated per component based on consumption rates.

3.2 Labor Cost

Labor cost depends on whether the operation is machine-controlled (where the operator monitors but does not perform cutting) or manually controlled. In CNC machining, labor cost per component is primarily a function of the operator's hourly rate and the total operator time per component, which includes active participation in setup, loading/unloading, and inspection:

$$C_{labor} = (t_{setup}/n_{batch} + t_{load} + t_{inspect}) \times R_{labor}$$

where t_{setup} is setup time per batch, n_{batch} is batch size, t_{load} is loading/unloading time per component, $t_{inspect}$ is operator inspection time, and R_{labor} is the labor rate per unit time.

3.3 Machining Time and Tooling Cost

Machining time is the foundation on which machine operating cost, tooling cost, and energy cost are all computed. For a simple turning operation, the basic machining time formula is:

$$t_m = (\pi \times D \times L) / (1000 \times V_c \times f)$$

where D is workpiece diameter (mm), L is length of cut (mm), V_c is cutting speed (m/min), and f is feed rate (mm/rev). For milling:



$$t_m = L_{path} / (f_z \times z \times N)$$

where L_{path} is tool path length, f_z is feed per tooth, z is number of teeth, and N is spindle speed (rpm).

Tooling cost per component is a function of tool life and the cost to replace or re-grind the tool. If a tool costs C_{tool} and lasts for T_{life} minutes of cutting, and each component requires t_m minutes of machining:

$$C_{tooling} = (t_m / T_{life}) \times (C_{tool} + t_{change} \times R_{labor})$$

The term $t_{change} \times R_{labor}$ accounts for the operator time spent changing the tool when it reaches end of life.

3.4 Machine Operating Cost

The machine operating cost per component is determined by the machine hourly rate (which itself incorporates capital depreciation, maintenance, and energy) multiplied by the productive time the machine is engaged per component:

$$C_{machine} = (t_m + t_{idle}) \times R_{machine}$$

The machine hourly rate $R_{machine}$ is typically computed by finance departments and reflects the purchase price of the machine, its expected working life, maintenance costs, and floor space charges. For a CNC machining center costing ₹80 lakhs with a 10-year working life and 4000 operating hours per year, the depreciation component alone works out to approximately ₹200 per hour, to which maintenance and energy charges must be added.

3.5 Setup Cost

Setup cost is often underestimated in small-batch and job-shop environments. It includes fixture design and fabrication, machine configuration, first-piece inspection, and CNC program verification. Per-component setup cost decreases with batch size:

$$C_{setup_per_part} = (t_{setup} \times R_{setup}) / n_{batch}$$

This relationship makes setup cost reduction strategies — quick-change tooling, modular fixtures, offline programming — economically significant in low-to-medium volume production. Setup cost per part can easily exceed machining cost per part in a batch of 10–20 components.

3.6 Energy and Overhead Costs

Energy consumption during machining can be estimated from the specific energy (energy per unit volume of material removed) and the material removal rate:

$$P_{cutting} = u \times MRR \quad \text{and} \quad C_{energy} = (P_{total} / MRR) \times V_{removed} \times C_{electricity}$$

where u is specific cutting energy (J/mm^3), MRR is material removal rate (mm^3/s), $V_{removed}$ is total volume removed per component, and $C_{electricity}$ is the cost per kWh. Specific energy values for common materials range from around $1.5 J/mm^3$ for aluminum alloys to $4\text{--}6 J/mm^3$ for steels and up to $10 J/mm^3$ for nickel superalloys.

Overhead costs include factory rent, lighting, air conditioning, administrative support, and quality management system costs. These are usually applied as a percentage surcharge on direct costs:

$$C_{overhead} = (C_{labor} + C_{machine}) \times OH_{rate}$$

where OH_{rate} is the overhead rate, commonly 60–150% of direct labor in manufacturing industries.

3.7 Scrap and Rework Costs

Scrap and rework costs are often treated as probabilistic cost elements based on historical defect rates. If the fraction of defective components is p_{defect} and each scrap part results in a material and machining cost write-off of C_{comp} :

$$C_{scrap} = p_{defect} \times C_{comp}$$



Rework cost is similarly a function of rework probability, rework operation time, and associated machine and labor costs. In precision industries, scrap rates of even 1–2% can significantly affect the economics of a machining operation, particularly when expensive raw materials are involved.

Table 2: Typical Cost Breakdown for a Turned Steel Component (Batch of 100)

Cost Element	Cost per Component (₹)	% of Total Cost
Material cost	85.00	28.3%
Labor cost (direct)	40.00	13.3%
Tooling cost	30.00	10.0%
Machine operating cost	60.00	20.0%
Setup cost (amortized)	25.00	8.3%
Energy cost	12.00	4.0%
Overhead	35.00	11.7%
Inspection & scrap (3%)	13.00	4.4%
Total	300.00	100%

4. ECONOMIC ANALYSIS OF MACHINING PROCESSES

Having established the cost structure, this section discusses the economic analysis methods used to evaluate the viability and comparative performance of machining operations. Economic analysis goes beyond cost estimation — it places machining operations in a business context by evaluating profitability, investment returns, and the conditions under which one process or strategy is economically preferred over another.

4.1 Break-Even Analysis

Break-even analysis identifies the production volume at which total revenues equal total costs, yielding zero profit or loss. In the context of machining process selection, it is often used to determine at what batch size a higher-investment process (such as CNC machining) becomes more economical than a lower-investment but less efficient conventional process.

Let $C_{\text{fixed_CNC}}$ and $C_{\text{fixed_Conv}}$ denote the fixed costs (setup + amortized machine investment per production run) of CNC and conventional machining respectively, and $C_{\text{var_CNC}}$ and $C_{\text{var_Conv}}$ be the variable (per-component) costs. The break-even quantity n^* is found from:

$$C_{\text{fixed_CNC}} + n \times C_{\text{var_CNC}} = C_{\text{fixed_Conv}} + n \times C_{\text{var_Conv}}$$

$$n^* = (C_{\text{fixed_CNC}} - C_{\text{fixed_Conv}}) / (C_{\text{var_Conv}} - C_{\text{var_CNC}})$$

For production volumes below n^* , the lower-investment process is preferred; above n^* , the more automated process justifies its higher fixed costs through lower variable costs. This analysis is routinely performed when evaluating whether to invest in new CNC equipment for a particular product family.

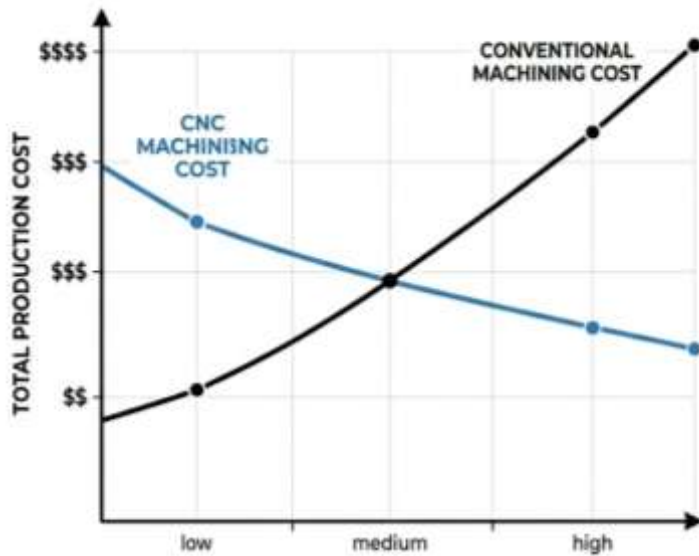


Fig. 5: Break-Even Analysis — CNC Machining vs. Conventional Machining Cost Curves

4.2 Return on Investment (ROI)

For capital investment in machining equipment, ROI measures the annual benefit derived from the investment relative to its cost:

$$ROI (\%) = [(Annual\ Savings\ or\ Profit) / (Investment\ Cost)] \times 100$$

Consider a workshop currently producing 50,000 components per year by conventional turning at ₹300 each. An investment of ₹25 lakhs in a CNC turning center reduces per-component cost to ₹220 by cutting machining time and reducing scrap. Annual saving = $50,000 \times (300 - 220) = ₹40$ lakhs. $ROI = (40/25) \times 100 = 160\%$ per year. Payback period = $25/40 = 7.5$ months. This type of calculation is what drives capital investment decisions in manufacturing firms.

4.3 Productivity Analysis

Production rate — components produced per unit time — is a key economic metric that directly affects revenue generation capacity. For a single machining operation, production rate Q (components/hour) is:

$$Q = 60 / (t_m + t_{load} + t_{change_avg})$$

where $t_{change_avg} = (t_m / T_{life}) \times t_{change}$ is the average tool change time prorated over all components. Maximizing Q — maximum production rate criterion — leads to a different optimal cutting speed than minimizing cost per component, because it ignores cost and focuses purely on throughput. The two criteria — minimum cost and maximum production rate — define a range of economically practical cutting speeds known as the high efficiency machining range.

Table 3: Economic Performance Indicators in Machining

Indicator	Formula / Basis	Decision Criterion
Cost per component	$C_{total} = \sum C_i$	Minimize
Production rate	$Q = 60 / (t_m + t_{aux})$	Maximize
Break-even quantity	$n^* = \Delta C_{fixed} / \Delta C_{var}$	Choose process for given volume
Return on Investment	$(Annual\ saving / Investment) \times 100$	Higher is better; typically $> 30\%$



Payback period	Investment / Annual saving	Shorter is better; typically < 3 years
Machine utilization	Productive time / Available time	Target > 70–80%

4.4 Cost-Benefit Analysis and Decision Making

Cost-benefit analysis extends beyond simple ROI by accounting for both quantifiable and qualitative benefits. In a machining context, benefits might include improved surface quality (reducing post-process finishing), reduced scrap rates (saving material and rework labor), shorter lead times (improving customer responsiveness), and reduced operator health risks (through better coolant management or reduced manual handling). These benefits, even when difficult to quantify precisely, can be assigned approximate monetary values using industrial benchmarks and included in the analysis.

For example, if a new CNC machining strategy reduces surface roughness R_a from $3.2\ \mu\text{m}$ to $0.8\ \mu\text{m}$ and eliminates a downstream grinding operation, the saving includes: elimination of grinding machine hourly cost, reduction in grinding wheel cost, and release of shop floor space. Such a change might yield ₹50–80 per component in additional savings that would not appear in a naive machining-only cost comparison.

5. TOOL LIFE AND ECONOMIC OPTIMIZATION

The relationship between cutting speed and tool life is central to the economics of machining. Increasing cutting speed reduces machining time (and thus machine and labor cost per component) but also accelerates tool wear, increasing tooling cost and tool change interruptions. Finding the optimal cutting speed that balances these opposing effects is one of the classical problems of manufacturing engineering.

5.1 Taylor's Tool Life Equation

F.W. Taylor's empirical equation, established through extensive experimental work in the late 19th and early 20th centuries, relates cutting speed V_c to tool life T_{life} :

$$V_c \times T_{\text{life}}^n = C_T$$

where n is the Taylor exponent (typically 0.1–0.5 depending on the tool-workpiece combination) and C_T is a constant equal to the cutting speed giving a tool life of 1 minute. For high-speed steel tools cutting medium carbon

steel, $n \approx 0.125$; for cemented carbide tools, n is typically 0.25–0.33; for ceramic and cubic boron nitride (CBN) tools, n can reach 0.4–0.5. The extended Taylor equation includes feed rate and depth of cut:

$$V_c \times T_{\text{life}}^n \times f^a \times d^b = C_{\text{extended}}$$

where f is feed rate, d is depth of cut, and a , b are additional empirical exponents determined through cutting experiments.

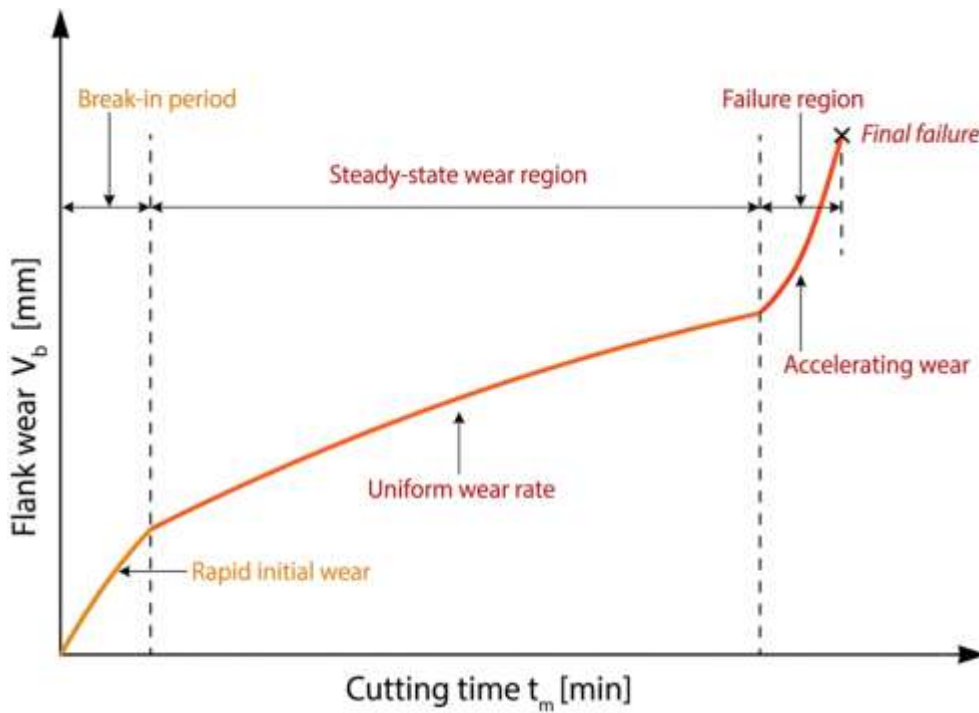


Fig. 6: Flank Wear Progression Curves at Different Cutting Speeds

5.2 Optimum Cutting Speed for Minimum Cost

Substituting Taylor's equation into the per-component cost expression and differentiating with respect to cutting speed, the minimum cost cutting speed V_{opt_cost} can be derived analytically. The optimum tool life T_{opt} for minimum cost is:

$$T_{opt(cost)} = (1/n - 1) \times [t_{change} + C_{tool} / R_{labor}]$$

The corresponding optimal cutting speed is then found from Taylor's equation using $T_{opt(cost)}$. This is the speed at which the rate of decrease in machining cost (from shorter cycle time) exactly equals the rate of increase in tooling cost. Similarly, the optimum tool life for maximum production rate is simply:

$$T_{opt(rate)} = (1/n - 1) \times t_{change}$$

Since C_{tool} / R_{labor} is always positive, $T_{opt(rate)} < T_{opt(cost)}$, meaning the maximum production rate speed is always higher than the minimum cost speed. The region between these two speeds defines the economically and productively acceptable range for cutting speed selection.

Table 4: Tool Life and Cost Data for Carbide Turning of Medium Carbon Steel

Cutting Speed (m/min)	Tool Life T (min)	t _m per Component (s)	C _{tooling} per Part (₹)	C _{machine} per Part (₹)	Total Variable Cost (₹)
80	95	48	8.50	22.00	30.50
100	55	38	13.20	17.50	30.70
120	30	32	21.00	14.70	35.70
150	14	26	38.50	12.00	50.50
180	7	21	65.00	9.70	74.70

The data in Table 4 illustrates the classic trade-off. At 80 m/min, long tool life keeps tooling cost low, but slow machining time means high machine cost. At 120 m/min, the combined cost reaches its minimum. Beyond 150



m/min, rapidly increasing tooling cost dominates and total cost climbs sharply. This tabular analysis replicates the theoretical minimum-cost condition derived analytically from Taylor's equation.

6. CNC MACHINING ECONOMICS

The adoption of CNC machining centers has fundamentally altered the economics of manufacturing. Unlike conventional machines where operator skill directly determines cycle time and quality, CNC machines execute pre-programmed toolpaths with repeatability that a human operator cannot consistently match. This shift has moved cost emphasis from direct labor toward capital depreciation, programming time, and tooling management — a transition that has significant implications for cost modeling.

6.1 CNC Machine Hourly Rate

The CNC machine hourly rate is a composite cost that encompasses depreciation, maintenance, tooling management systems, coolant, compressed air, and operator monitoring. A representative breakdown is:

CNC Cost Component	Basis	Estimated Rate (₹/hr)
Machine depreciation	₹80 lakh, 10yr, 4000 hr/yr	200
Maintenance & repairs	5% of capital/yr	100
Energy (15 kW avg)	₹8/kWh × 15 kW	120
Tooling management overhead	Estimated	60
Operator (1 per 2 machines)	₹350/hr ÷ 2	175
Overhead allocation	80% of direct costs	364
Total CNC Hourly Rate		~1,019

These figures are indicative and vary significantly by region, machine type, and production volume. The important point is that at approximately ₹1,000/hr, a CNC machining center must be kept productively engaged — idle time is expensive. Machine utilization (ratio of productive cutting time to available time) is therefore a key economic performance metric for CNC facilities, and improving utilization from 60% to 80% on a machine costing ₹1,000/hr generates ₹200/hr of additional productive output.

6.2 CNC vs. Conventional Machining — Economic Comparison

Parameter	Conventional Machining	CNC Machining
Capital investment	₹5–15 lakh	₹25–200 lakh
Machine hourly rate	₹200–400/hr	₹800–2000/hr
Setup time	4–8 hrs	1–2 hrs (with offline programming)
Cycle time consistency	Variable (operator-dependent)	Repeatable ±2–5%
Operator skill required	High	Medium (programming skill)
Economic batch size	10–500	50–100,000+
Surface finish consistency	Moderate	High



Complex geometry	Limited	High (up to 5-axis)
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6.3 Industry 4.0 and Smart Machining Economics

The integration of Industry 4.0 technologies — IoT sensors, machine learning, digital twins, and cloud-based manufacturing execution systems — is beginning to add a new layer to machining economics. Real-time tool condition monitoring, for instance, can extend tool life by 20–30% by preventing premature tool changes and catastrophic tool failures. Predictive maintenance reduces unplanned downtime, which in a high-cost CNC environment can save thousands of rupees per incident. AI-driven cutting parameter optimization has demonstrated cost reductions of 10–20% in controlled studies by dynamically adjusting feed and speed to maintain target tool wear rates across varying workpiece material conditions.

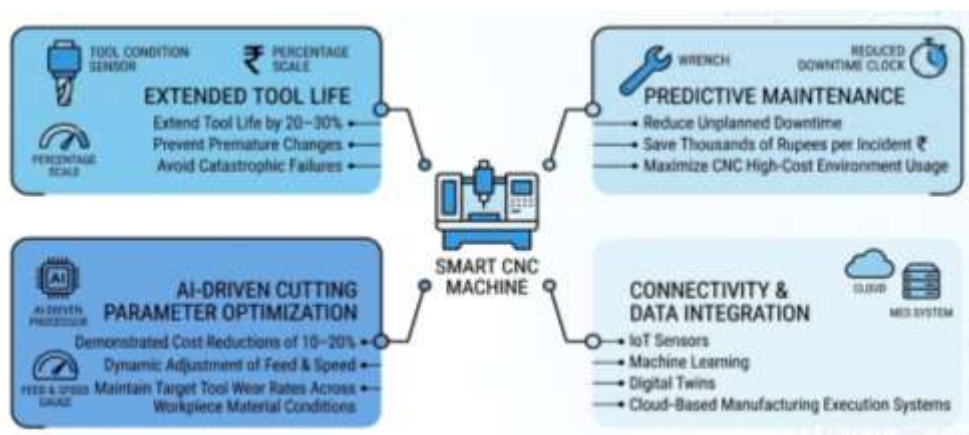


Fig. 7: Industry 4.0 Integration in Machining — Economic Benefit Areas

7. CASE STUDIES

Case Study 1: Automotive Crankshaft Machining Cost Optimization

Problem Statement: An automotive component manufacturer in Pune produces 20,000 crankshafts per year from EN24 alloy steel. The existing conventional turning and grinding process line has an average per-component cost of ₹1,850. The plant management is evaluating whether to invest in a CNC turning center with live tooling (cost: ₹75 lakhs) that would combine turning and milling operations into a single setup, reducing total operation time.

Production Details: Each crankshaft requires 28 minutes of conventional turning across three separate machines, 12 minutes of manual milling for the keyways and oil hole drilling, and 18 minutes of cylindrical grinding. Total conventional cycle time: 58 minutes per component. The CNC turning center with live tooling is expected to reduce combined turning and milling time to 32 minutes, with grinding remaining at 18 minutes (total: 50 minutes on conventional grinder). Setup time reduction from 3 hours to 45 minutes per batch of 50 is also expected.



Cost Calculations:

Cost Element	Current Process (₹/component)	After CNC Investment (₹/component)
Material cost	620	620
Machining (turning + milling)	480	310
Grinding	240	240
Setup cost (amortized, batch 50)	85	25
Tooling cost	120	90
Inspection	45	35
Overhead (75%)	450	330
Total	2,040	1,650

Economic Analysis: Per-component cost reduction = ₹2,040 – ₹1,650 = ₹390. Annual saving at 20,000 units = ₹390 × 20,000 = ₹78 lakhs. Investment = ₹75 lakhs. ROI = (78/75) × 100 = 104% per year. Payback period = 75/78 = 0.96 years (approximately 11.5 months). Additional benefits include reduced floor space usage (three machines replaced by one), lower WIP inventory, and improved delivery reliability. The investment is clearly justified.

Optimization Analysis: Using Taylor's equation data for EN24 machined with PVD-coated carbide inserts ($n = 0.28$, $C_T = 380$ m/min), the minimum cost cutting speed is calculated as:

$$T_{opt(cost)} = (1/0.28 - 1) \times [0.5 + 450/350] = 3.57 \times 1.79 \approx 6.4 \text{ min}$$

$$V_{opt(cost)} = 380 / (6.4^{0.28}) = 380 / 1.67 \approx 228 \text{ m/min}$$

This speed, combined with a feed rate of 0.25 mm/rev and depth of cut of 2.5 mm, gives a material removal rate of approximately 570 mm³/s — consistent with the productivity targets for the production volume.

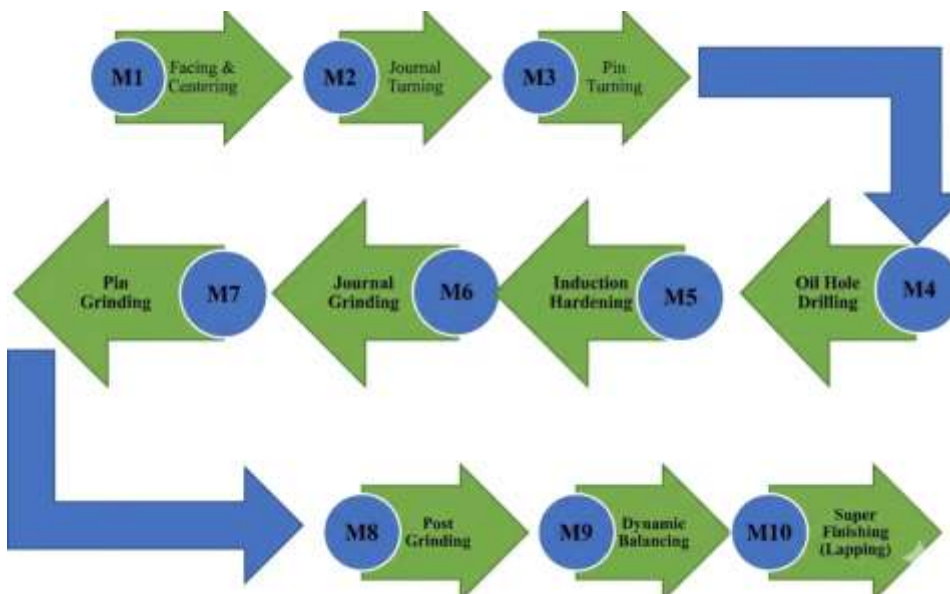


Fig. 8: Optimized Crankshaft Machining Process Flow — Before and After CNC Integration



Case Study 2: Aerospace Titanium Bracket Machining — Cost Analysis

Problem Statement: A tier-2 aerospace supplier machines titanium alloy (Ti-6Al-4V) structural brackets using 5-axis CNC milling. The component has a buy-to-fly ratio of 6.5:1 — starting billet weight is 2.6 kg; finished component weighs 0.4 kg. The customer requires $Ra \leq 0.8 \mu\text{m}$ on critical surfaces and GD&T tolerances of $\pm 0.05 \text{ mm}$ on bolt hole positions. The manufacturer needs to determine the cost per component and identify optimization opportunities.

Production Data:

Parameter	Value
Material (Ti-6Al-4V)	₹3,200/kg; billet mass 2.6 kg
5-axis CNC hourly rate	₹2,400/hr
Total machining time (rough + semi + finish)	4.5 hrs
Setup time per batch of 10	6 hours
Carbide end-mill cost (PCD coated)	₹4,500 per tool
Tool life (Ti at 40 m/min)	45 min (rough), 90 min (finish)
CMM inspection time	35 min per component
Scrap rate (historical)	2.5%

Cost Calculation per Component:

Material cost = $2.6 \times 3,200 = ₹8,320$. Machine cost = $4.5 \times 2,400 = ₹10,800$. Setup cost per component = $(6 \times 2,400) / 10 = ₹1,440$. Tooling cost: rough milling uses 3 tool changes $\times ₹4,500 / 45 \text{ min} \times 150 \text{ min} = ₹4,500$; finishing uses 2 tool changes $\times ₹4,500 / 90 \text{ min} \times 60 \text{ min} = ₹1,500$. Total tooling = ₹6,000. Inspection = $(35/60) \times 2,400 = ₹1,400$. Overhead at 80% of machine + labor = ₹9,792. Scrap provision = $2.5\% \times 37,752 = ₹944$. Total estimated cost per component $\approx ₹38,696$.

This cost highlights why titanium aerospace parts are expensive: material alone accounts for 21% and the 4.5-hour cycle on a ₹2,400/hr machine contributes 28%. Optimization identified three opportunities: (1) increasing rough milling feed by 15% through optimized toolpath (saving 0.3 hrs); (2) switching to a high-pressure coolant system to extend tool life by 35%, reducing tooling cost by ₹1,350 per part; (3) reducing scrap rate to 1.2% through improved in-process gauging. Combined, these changes reduce total cost to approximately ₹35,200 — a saving of 9.1%.



Fig. 09: Cost Distribution for Ti-6Al-4V Aerospace Bracket Machining



8. FUTURE TRENDS AND CHALLENGES

8.1 Sustainable and Green Machining

Environmental sustainability is increasingly influencing machining economics. Traditional flood cooling uses large volumes of cutting fluid that must be filtered, stored, disposed of, and continuously replenished — all at cost. Minimum quantity lubrication (MQL), dry machining, and cryogenic cooling with liquid nitrogen or CO₂ are emerging as economically competitive alternatives for certain material-process combinations. While the upfront investment in MQL or cryogenic systems adds to capital cost, the elimination of coolant purchase, filtration, and disposal costs can yield net savings of ₹15–40 per component in high-volume production. Regulatory pressure on coolant disposal is making clean machining strategies increasingly attractive from a compliance cost perspective.

8.2 Digital Twins and Predictive Cost Modeling

The digital twin concept — a virtual model of a physical machining process that is continuously updated with real sensor data — offers transformative potential for machining economics. A validated digital twin can predict tool wear, surface roughness, and dimensional deviation in real time, allowing adaptive control of cutting parameters to maintain quality without over-engineering the process. From an economic standpoint, digital twins enable predictive maintenance (reducing unplanned stoppages), adaptive toolpath optimization (reducing machining time), and virtual process validation (reducing expensive physical trials). Industry reports suggest digital twin implementation in machining can reduce total production cost by 8–15%.

8.3 AI-Assisted Process Optimization

Machine learning algorithms are being applied to machining parameter optimization with increasing sophistication. Unlike classical optimization methods that rely on analytical models (such as Taylor's equation), ML approaches can capture complex, nonlinear interactions between cutting conditions, workpiece material variability, machine condition, and economic outcomes. Reinforcement learning has shown particular promise for adaptive cutting parameter control in variable-condition environments. The economic benefit is primarily through consistent operation near the minimum-cost point without requiring extensive experimental characterization of each new material or geometry.

8.4 Additive-Subtractive Hybrid Manufacturing

The growing integration of additive manufacturing (AM) with subsequent machining (hybrid manufacturing) is creating new cost modeling challenges and opportunities. AM can near-net-shape a complex part, dramatically reducing the volume of material removed in machining and improving the buy-to-fly ratio. For titanium and nickel superalloy components where material cost dominates, this hybrid approach can significantly reduce total cost despite the additional AM processing expense. Developing cost models that span both the AM and machining phases — accounting for AM deposition rates, heat treatment, support structure removal, and final machining — is an active area of research.

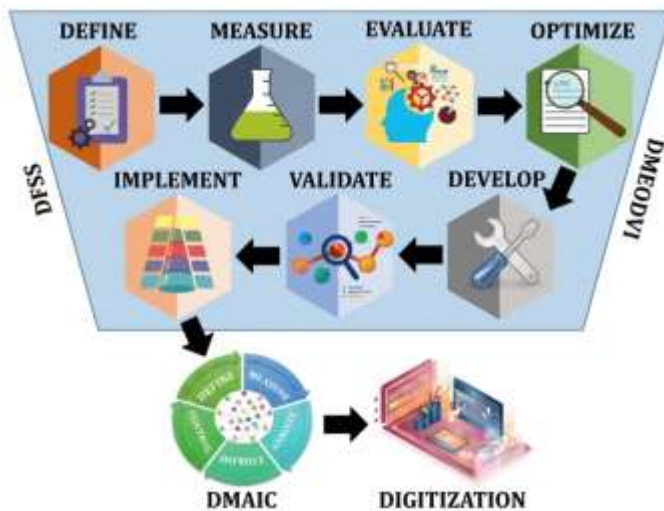


Fig. 10: Technology Roadmap for Advanced Machining Economics

9. CONCLUSION

This paper has presented a structured treatment of cost modeling and economic analysis as applied to machining processes, spanning from the fundamental cost element formulations through classical tool life optimization to contemporary CNC economics and emerging digital manufacturing trends. Several important conclusions can be drawn from this analysis.

First, the total cost of a machined component is the result of multiple interacting cost elements — material, labor, tooling, machine, setup, energy, overhead, and scrap — each driven by specific process variables. A credible cost model must account for all these elements. Simplistic approaches that consider only machining time or labor cost are inadequate for engineering decision-making and can lead to poor process selection or incorrect pricing.

Second, Taylor's tool life equation remains a practically valuable framework for cutting parameter optimization, despite being over a century old. The minimum cost and maximum production rate cutting speeds derived from Taylor's analysis define a rational range for speed selection, and practical experience shows that operation within this range — rather than at arbitrarily chosen speeds — can reduce total machining cost by 15–25% without any capital investment.

Third, the economics of CNC machining are dominated by machine hourly rate and utilization. With typical CNC hourly rates in the range of ₹800–2,000, idle time is extremely expensive. Strategies that improve utilization — offline programming, quick-change tooling, group technology cell layouts — have large economic payoffs. The break-even analysis framework shows clearly that CNC investment is justified above certain batch sizes and becomes increasingly compelling as volume grows.

Fourth, as demonstrated by the two case studies, systematic cost modeling and economic analysis can identify actionable cost reduction opportunities that are not apparent from casual observation. In the automotive crankshaft case, consolidating operations on a CNC turning center yielded a payback period of under 12 months. In the aerospace titanium bracket case, relatively modest improvements in toolpath strategy, coolant delivery, and in-process gauging reduced per-component cost by over 9%.

Fifth, emerging technologies including digital twins, AI-assisted optimization, MQL systems, and hybrid additive-subtractive manufacturing are beginning to reshape the economics of machining at a structural level. Cost models that do not account for these factors will increasingly underestimate the economic case for technology adoption.

For manufacturing students and practitioners, the principal takeaway is that machining is never purely a technical activity — it is always simultaneously an economic one. The engineer who understands both the physics of chip formation and the economics of tool life, setup time, and overhead allocation is far better



equipped to make good decisions than one who knows only half the story. It is hoped that this paper contributes to building that integrated understanding.

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REFERENCES

- [1] Boothroyd, G., and Knight, W.A. (2006). *Fundamentals of Machining and Machine Tools*, 3rd ed. CRC Press, Boca Raton.
- [2] Kalpakjian, S., and Schmid, S.R. (2014). *Manufacturing Engineering and Technology*, 7th ed. Pearson, Upper Saddle River.
- [3] Ghosh, A., and Mallik, A.K. (2010). *Manufacturing Science*, 2nd ed. East-West Press, New Delhi.
- [4] Taylor, F.W. (1907). On the art of cutting metals. *Transactions of ASME*, 28, 31–350.
- [5] Groover, M.P. (2015). *Fundamentals of Modern Manufacturing: Materials, Processes, and Systems*, 5th ed. Wiley, Hoboken.
- [6] Merchant, M.E. (1945). Mechanics of the metal cutting process. *Journal of Applied Physics*, 16(5), 267–275.
- [7] Shaw, M.C. (2005). *Metal Cutting Principles*, 2nd ed. Oxford University Press, New York.
- [8] Armarego, E.J.A., and Brown, R.H. (1969). *The Machining of Metals*. Prentice-Hall, Englewood Cliffs.
- [9] Hitomi, K. (1996). *Manufacturing Systems Engineering*, 2nd ed. Taylor and Francis, London.
- [10] DeGarmo, E.P., Black, J.T., and Kohser, R.A. (2011). *Materials and Processes in Manufacturing*, 11th ed. Wiley, New York.
- [11] Chryssolouris, G. (2006). *Manufacturing Systems: Theory and Practice*, 2nd ed. Springer, New York.
- [12] Ostwald, P.F. (1992). *Engineering Cost Estimating*, 3rd ed. Prentice-Hall, Englewood Cliffs.
- [13] Yeo, S.H., Khoo, L.P., and Neo, S.S. (2000). Tool life prediction and tool replacement strategy for end milling. *International Journal of Advanced Manufacturing Technology*, 16(6), 399–406.
- [14] Cus, F., and Balic, J. (2003). Optimization of cutting process by GA approach. *Robotics and Computer-Integrated Manufacturing*, 19(1–2), 113–121.
- [15] Wang, J., and Mathew, P. (1995). Development of a general tool life equation using a diffusion model. *Journal of Manufacturing Science and Engineering*, 117(2), 243–248.
- [16] Ezugwu, E.O., and Wang, Z.M. (1997). Titanium alloys and their machinability — a review. *Journal of Materials Processing Technology*, 68(3), 262–274.
- [17] Jawahir, I.S., et al. (2016). Cryogenic manufacturing processes. *CIRP Annals*, 65(2), 713–736.
- [18] Iqbal, A., He, N., Li, L., and Dar, N.U. (2007). A fuzzy expert system for optimizing parameters and predicting performance measures in hard milling process. *Expert Systems with Applications*, 32(4), 1020–1027.
- [19] Rajemi, M.F., Mativenga, P.T., and Aramcharoen, A. (2010). Sustainable machining: selection of optimum turning conditions based on minimum energy considerations. *Journal of Cleaner Production*, 18(10–11), 1059–1065.
- [20] Shin, Y.C., and Danai, K. (1992). In-process tool wear estimation through force sensing. *Journal of Engineering for Industry*, 114(2), 153–160.
- [21] Teti, R., Jemielniak, K., O'Donnell, G., and Dornfeld, D. (2010). Advanced monitoring of machining operations. *CIRP Annals*, 59(2), 717–739.



- [22] Herrmann, C., et al. (2004). Life cycle cost based process chain evaluation in the context of sustainability management. Proceedings of CIRP Life Cycle Engineering, Berlin, pp. 31–39.
- [23] Rao, R.V., and Pawar, P.J. (2010). Modelling and optimization of process parameters of wire electrical discharge machining. Proceedings of the Institution of Mechanical Engineers, Part B, 223(11), 1431–1440.
- [24] Kim, D.H., and Lee, C.M. (2014). A study of CNC milling machining parameters for CFRP. International Journal of Advanced Manufacturing Technology, 70(1–4), 591–598.
- [25] Lu, T., and Jawahir, I.S. (2019). A comprehensive evaluation of machining performance: comparing cryogenic machining with dry and MQL-based machining of Ti-6Al-4V alloy. Machining Science and Technology, 23(5), 736–762.