

A MULTI-CRITERIA DECISION-MAKING FRAMEWORK FOR PRODUCTIVITY MEASUREMENT IN CALIBRATION FACILITIES BASED ON OMAX AND AHP

Abstract

Calibration facilities are required to maintain measurement accuracy while meeting operational targets; however, productivity monitoring is often fragmented because multiple performance ratios must be evaluated simultaneously. This study proposes a multi-criteria decision-making framework that integrates the Objective Matrix (OMAX) for productivity measurement with the Analytical Hierarchy Process (AHP) for deriving ratio weights, applied as a case study in the Calibration & Measurement Facility of an automotive manufacturing company. AHP pairwise comparisons (aggregated using the geometric mean) were used to prioritize four productivity ratios, producing weights of 0.30 (R1), 0.40 (R2), 0.23 (R3), and 0.06 (R4), with a strong consistency result ($CR = 0.004 < 0.10$), indicating reliable judgments. The resulting weights were embedded into OMAX scoring to compute monthly productivity indices for January–July 2025. The OMAX results show substantial fluctuations: productivity decreased in January (–46%), improved sharply in February (+109.67%), dropped again in March (–79%) and April (–60%), recovered in May (+16%), peaked in June (+204%), and declined in July (–28%). Overall, the integrated OMAX–AHP framework provides a transparent and consistent way to (i) weight productivity ratios based on expert priorities and (ii) track productivity dynamics over time, supporting managerial diagnosis and targeted improvement in calibration operations.

Keywords:

Objective Matrix (OMAX); Analytical Hierarchy Process (AHP); productivity measurement; multi-criteria decision making; calibration facility; manufacturing operations.

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Introduction:

3 Productivity measurement has become a central managerial requirement as organizations face stronger pressure to
4 deliver faster, more reliable, and more cost-effective services while maintaining compliance and risk control. In
5 highly regulated technical environments, productivity cannot be interpreted only as “output volume”, but must
6 reflect multi-dimensional performance such as timeliness, resource efficiency, service quality, and conformance to
7 standards so that improvement decisions are evidence-based and defensible. Contemporary performance-
8 measurement literature also emphasizes the need for measurement systems that are adaptable to uncertainty and
9 dynamic operational contexts, rather than relying on single indicators or static reporting routines [1].

10 Calibration facilities represent a special case of productivity management. Besides operational targets (throughput,
11 turnaround time, utilization), calibration laboratories must ensure metrological traceability and measurement
12 reliability, which are fundamental to the credibility and comparability of results. Traceability is not merely an
13 administrative requirement; it is a technical chain linking measurements to standards with known uncertainties, and
14 different traceability chains can directly affect measurement uncertainty and risk exposure [2]. In practice,
15 calibration facilities often struggle to balance the competing needs of maintaining technical competence, managing
16 calibration schedules, controlling cost, and meeting customer expectations especially under accreditation
17 frameworks such as ISO/IEC 17025, where competence, impartiality, and consistent operation are explicitly
18 required and audited [3]. Empirical and review-based studies in the ISO/IEC 17025 context show that operational
19 performance improvement is tightly connected to systematic management practices and critical success factors
20 (CSFs), including process control, competence management, documentation discipline, and continual improvement
21 mechanisms [3].

22 Given this multi-constraint environment, productivity measurement in calibration facilities requires tools that can (i)
23 incorporate multiple operational criteria, (ii) convert heterogeneous indicators into a single interpretable index, and
24 (iii) assign criterion importance transparently. The Objective Matrix (OMAX) method is widely used to convert
25 multiple productivity ratios into a structured matrix score and aggregate them into a productivity index, enabling
26 organizations to detect which criteria drive performance deterioration or improvement over time. Recent applied
27 studies show OMAX can support partial-to-total productivity analysis and can be paired with structured
28 improvement actions [4]. However, OMAX requires weighting of criteria; if weighting is ad-hoc, the resulting
29 productivity index may be questioned and less useful as a decision basis.

30 To address this, Analytical Hierarchy Process (AHP) offers a rigorous multi-criteria decision-making mechanism for
31 deriving weights through pairwise comparisons, making the weighting logic explicit and auditable. Recent
32 applications integrating OMAX and AHP demonstrate that the combined approach improves interpretability:
33 OMAX provides the scoring and index structure, while AHP provides defensible weights reflecting expert priorities
34 [4], [5]. A key methodological requirement in AHP is the consistency of expert judgments; recent research confirms
35 the central role of consistency evaluation (e.g., consistency ratio thresholds) to ensure that pairwise comparisons are
36 reliable enough to support valid weighting outcomes [6].

37 Therefore, this study proposes a multi-criteria decision-making framework for productivity measurement in
38 calibration facilities by integrating OMAX and AHP. The framework is designed to: (1) structure productivity
39 indicators relevant to calibration operations, (2) compute a composite productivity index using OMAX, and (3)
40 derive criterion weights using AHP with consistency checking, thereby strengthening the transparency and
41 defensibility of improvement decisions. This approach is expected to be particularly relevant for calibration facilities
42 operating under ISO/IEC 17025-driven demands for reliable, traceable, and consistent service delivery [3], while
43 still pursuing operational efficiency and resource optimization [7].

44 **Related Works and Literature Review:**

45 Recent work suggests that productivity is no longer treated as a single, straightforward construct. Instead,
46 researchers increasingly view it as a multi-dimensional outcome shaped by the realities of modern industrial
47 operations. In many settings, productivity cannot be explained by output and labor alone; it is also influenced by
48 quality performance, reliability, risk controls, and the demands of regulatory compliance[1]. In response,
49 performance measurement has gradually moved toward approaches that combine operational metrics with expert
50 judgment, so that “what matters” in practice is captured alongside what is easiest to count[7].

51

52 **Using OMAX for productivity measurement**

53 The Objective Matrix (OMAX) method remains widely used when organizations need to evaluate productivity
54 across several ratios at once. Its strength lies in the way it structures scoring and consolidates different indicators
55 into a single index, which makes productivity changes easier to monitor over time. Recent empirical studies report
56 that OMAX has been applied in both manufacturing and service environments to trace productivity shifts, reveal
57 bottlenecks, and identify areas where improvement efforts should be prioritized[4], [5]. Kotimah and Aryanny, for
58 instance, show that OMAX can highlight underperformance in specific dimensions even when total output appears
59 unchangeduseful for managerial diagnosis rather than simple reporting.

60 A recurring point in the literature, however, is that OMAX results depend heavily on the weighting scheme used for
61 the productivity ratios. When weights are selected arbitrarily or without a clear rationale, the resulting index can
62 become difficult to justify and less persuasive for managerial decision-making[5]. This concern has motivated
63 researchers to integrate OMAX with more formal multi-criteria decision-making (MCDM) techniques, where the
64 logic behind weighting can be made explicit[8].

65 **Integrating AHP into productivity and performance assessment**

66 Among MCDM methods, the Analytical Hierarchy Process (AHP) is one of the most commonly adopted tools for
67 assigning weights to performance criteria. AHP works by structuring the decision problem into a hierarchy and
68 deriving priority weights from pairwise comparisons. In practice, this allows expert judgments to be used in a
69 disciplined way rather than informally. Recent studies indicate that AHP-based weighting can strengthen the
70 transparency and robustness of productivity evaluation, particularly when multiple indicators must be reconciled
71 into a single assessment[6].

72 Putri and Sahbana report that introducing AHP into productivity analysis makes the weighting process more
73 systematic and improves confidence in the conclusions drawn from the model. Salomon et al. also emphasize the
74 importance of checking the consistency ratio (CR) to ensure that expert comparisons are logically coherent and
75 therefore defensible. Taken together, these findings support pairing AHP with OMAX, especially in contexts where
76 decisions are subject to scrutiny and must be clearly justified[9], [10].

77 **Productivity measurement in calibration facilities and ISO/IEC 17025**

78 Calibration and measurement facilities face a different set of constraints compared with typical production units.
79 Beyond efficiency targets, calibration laboratories must manage uncertainty, maintain traceability, and comply with
80 international standards such as ISO/IEC 17025. Recent evidence highlights that metrological traceability has
81 practical and strategic implications because it shapes measurement risk, operational credibility, and customer
82 trust[2].

83 Studies on ISO/IEC 17025-accredited laboratories further suggest that strong operational performance is closely
84 associated with systematic management practices such as standardized procedures, competence management, and
85 continuous improvement routines[3]. Velásquez et al. also show that calibration planning plays a meaningful role in
86 productivity outcomes, affecting turnaround times and overall resource utilization.

87 Even so, much of the existing work in calibration settings tends to focus on scheduling optimization, compliance
88 assessment, or uncertainty analysis, rather than integrated productivity measurement frameworks. Comparatively
89 few studies combine productivity indexing with structured weighting methods specifically for calibration facilities.
90 This gap indicates the need for an approach that evaluates operational productivity and technical performance
91 together within a single, coherent framework[11], [12].

92 **Research gap and study contribution**

93 Three gaps stand out. First, although OMAX is widely accepted for productivity measurement, many applications
94 still rely on subjective or ad-hoc weighting. Second, while AHP is well established for multi-criteria weighting, its
95 integration with productivity index models in calibration contexts remains limited. Third, empirical productivity
96 studies centered on calibration facilities are still relatively scarce, despite the critical role these facilities play in
97 safeguarding measurement reliability and overall industrial quality[13], [14].

98 To address these gaps, this paper proposes an integrated productivity measurement framework for calibration
99 facilities based on OMAX and AHP. The approach combines OMAX scoring with AHP-derived weights and
100 consistency testing, offering a clearer rationale for weighting, stronger methodological discipline, and more
101 actionable results for managers[15], [16]. This is particularly relevant for ISO/IEC 17025-oriented environments,
102 where productivity decisions must be balanced against reliability requirements and regulatory expectations.

103 **Methods:**

104 **Research Design**

105 This study employs a quantitative case study to measure and monitor productivity performance in a Calibration &
106 Measurement Facility within an automotive manufacturing company. The proposed framework integrates the
107 Objective Matrix (OMAX) method to compute a composite productivity index and the Analytical Hierarchy Process
108 (AHP) to determine the relative importance (weights) of productivity ratios (criteria). This integration is intended to
109 improve objectivity and transparency by combining ratio-based measurement with a structured multi-criteria
110 weighting approach.

111 The methodological stages include:

112 (1) identifying productivity ratios relevant to calibration activities,

113 (2) deriving ratio weights using AHP,

114 (3) converting ratio achievements into standardized OMAX scores and computing productivity indices, and

115 (4) analyzing productivity trends across the observation period.

116 **Case Study and Data Collection**

117 Operational data were obtained as historical records for the calibration unit covering January–July 2025, consisting
118 of calibration output, defect counts, number of workers, and total working hours. The dataset includes monthly
119 calibration output (units), defect (units), number of workers (persons), and working hours (hours).

120 In addition to operational data, expert judgments were collected using structured pairwise-comparison
121 questionnaires to support AHP weighting among the four productivity ratios. The pairwise comparisons were later
122 aggregated using the geometric mean to obtain a single comparison matrix.

123 **Productivity Ratios Definition**

124 The OMAX criteria in this study are defined as four productivity ratios reflecting effectiveness, efficiency, and
125 quality in calibration operations. The ratios are aligned with the calibration context: R1 effectiveness, R2 labor
126 efficiency, R3 time efficiency, and R4 quality/defect.

127 Let, for month t :

128 Q_t : number of instruments successfully calibrated (output)

129 P_t : number of instruments planned/scheduled for calibration

130 L_t : number of workers involved

131 H_t : total working hours used

132 D_t : number of calibration defects/nonconforming results

133 Based on the operational meaning described in the dataset processing section (output vs plan, output per labor,
134 output per hour, and defects as quality indicator), the ratios are formulated as:

135 (1) Effectiveness ratio (R1)

$$R1_t = \frac{Q_t}{P_t}$$

136 (2) Labor efficiency ratio (R2)

$$R2_t = \frac{Q_t}{L_t}$$

137 (3) Time efficiency ratio (R3)

$$R3_t = \frac{Q_t}{H_t}$$

138 (4) Quality/defect ratio (R4) cost criterion

139 Because R4 represents defect/quality (higher defects indicate poorer quality), it is treated as a cost-type criterion:

$$R4_t = \frac{D_t}{Q_t}$$

140 The monthly computed ratio values for January–July 2025 are provided in the ratio processing results table.

141 Analytical Hierarchy Process (AHP) for Weighting

142 AHP Hierarchy

143 The AHP structure consists of:

144 Goal: determine ratio weights for productivity measurement

145 Criteria: $R1, R2, R3, R4$ (productivity ratios)

146 A supporting explanation of the hierarchy (including sub-criteria such as calibration output, workers, hours, and defect) is provided as the basis for constructing the pairwise matrix.

148 Pairwise Comparison and Aggregation

149 Experts compare criteria using Saaty's scale (1–9). The individual matrices are aggregated into a single matrix using the geometric mean, as shown by the geometric mean weighting table.

151 Let the aggregated pairwise comparison matrix be:

$$A = [a_{ij}], a_{ij} > 0, a_{ji} = \frac{1}{a_{ij}}$$

152 Priority Weights

153 Weights are derived by normalizing the matrix and averaging row priorities (or equivalently approximating the principal eigenvector). The final criterion weights used for OMAX weighting are:

$$w = [w_{R1}, w_{R2}, w_{R3}, w_{R4}] = [0.30, 0.40, 0.23, 0.06]$$

155 These weights are reported as the ratio-weight output used in OMAX computation.

156 Consistency Test

157 Judgment consistency is verified using:

$$CI = \frac{\lambda_{\max} - n}{n - 1}, \quad CR = \frac{CI}{RI}$$

158 where n is the number of criteria and RI is the Random Index. The calculated values are $\lambda_{\max}=4.010$, $CI=0.003$, and
159 $CR=0.004$, and the weighting is accepted because $CR<0.1$.

160 Objective Matrix (OMAX) Productivity Measurement

161 OMAX Scoring Scale (0–10) and Level Setting

162 OMAX converts each ratio achievement into a standardized score (0–10), where level 0 represents the worst
163 performance, level 3 is the standard/average, and level 10 is the best/target.

164 The level table for each ratio (R1–R4) is defined and used as the scoring reference in OMAX computation.

165 Let $x_{i,t}$ be the actual value of ratio i in month t . Let W_i , S_i , and T_i denote the worst (level 0), standard (level 3), and
166 target/best (level 10) values for ratio i . Because the OMAX scale is based on these three anchors, the score
167 $s_{i,t} \in [0,10]$ can be computed by linear interpolation:

168 Benefit-type criteria (higher is better): $i \in \{R1, R2, R3\}$

$$s_{i,t} = \begin{cases} 0, & x_{i,t} \leq W_i \\ 3 \cdot \frac{x_{i,t} - W_i}{S_i - W_i}, & W_i < x_{i,t} < S_i \\ 3 + 7 \cdot \frac{x_{i,t} - S_i}{T_i - S_i}, & S_i \leq x_{i,t} < T_i \\ 10, & x_{i,t} \geq T_i \end{cases}$$

169 Cost-type criterion (lower is better): $R4$

$$s_{R4,t} = \begin{cases} 10, & x_{R4,t} \leq T_{R4} \\ 3 + 7 \cdot \frac{S_{R4} - x_{R4,t}}{S_{R4} - T_{R4}}, & T_{R4} < x_{R4,t} \leq S_{R4} \\ 3 \cdot \frac{W_{R4} - x_{R4,t}}{W_{R4} - S_{R4}}, & S_{R4} < x_{R4,t} < W_{R4} \\ 0, & x_{R4,t} \geq W_{R4} \end{cases}$$

170 Composite OMAX Performance Score

171 After scoring, the monthly OMAX performance score is computed as the weighted sum of the ratio scores:

$$Performance_t = \sum_{i=1}^4 w_i s_{i,t}$$

172 where w_i denotes the AHP-derived weight of ratio i and $s_{i,t}$ is the OMAX score (0–10) of ratio i in month t .

173 Productivity Index (PI) Based on Reference Score

174 To facilitate interpretation, the productivity index (PI) is expressed as the percentage deviation of the monthly
175 OMAX performance score from the standard reference value (baseline = 300):

$$PI_t(\%) = \frac{Performance_t - 300}{300} \times 100\%$$

176 Accordingly, negative PI values indicate performance below the reference level, rather than negative productivity.
 177 The absolute OMAX performance score $Performance_t$ remains non-negative because it is derived from weighted
 178 scores mapped onto a 0–10 scale.

179 The reference value of 300 represents the standard OMAX performance level, corresponding to normal operating
 180 conditions. This baseline allows productivity performance to be expressed as a relative deviation from the standard
 181 level, facilitating intuitive interpretation of overperformance and underperformance.

182 **Month-to-Month Performance Variation**

183 Month-to-month dynamics are reported as the percentage change in the OMAX performance score:

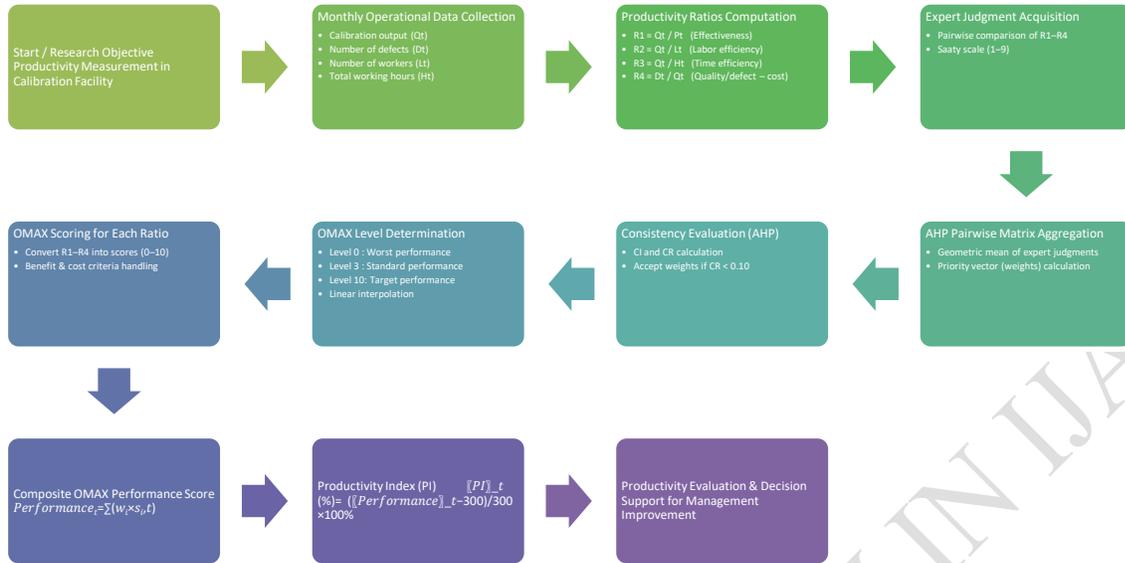
$$\Delta Performance_t = \frac{Performance_t - Performance_{t-1}}{Performance_{t-1}} \times 100\%$$

184 **Integrated OMAX–AHP Framework**

185 The integrated framework uses AHP to provide validated criterion weights and OMAX to provide standardized
 186 scoring and a composite index. The calibration context emphasizes that OMAX indicators include output, defects,
 187 labor, and working time, and each is weighted based on its influence on total productivity.

188 **Flow Diagram**

189 The workflow starts by collecting monthly operational data from the calibration facility, including calibration
 190 output, defect counts, the number of workers, and total working hours. These data are used to compute the
 191 productivity ratios (R1–R4) representing effectiveness, labor efficiency, time efficiency, and quality/defect
 192 performance. Expert judgments are then gathered to build the AHP pairwise comparison matrices, which are
 193 aggregated using the geometric mean. Criterion weights are subsequently derived and validated through a
 194 consistency test ($CR < 0.10$). Next, OMAX performance levels are specified and each ratio achievement is
 195 converted into a standardized score. The monthly OMAX performance score is then obtained through weighted
 196 scoring, and it is finally expressed as a productivity index (PI) relative to the reference value (baseline = 300) to
 197 support trend analysis and managerial decision-making.



198 **Figure 1. Workflow of the proposed OMAX-AHP productivity measurement framework.**
199

200 **Data Processing Tools**

201 All computations (ratio calculations, AHP weighting and consistency checking, OMAX scoring, and productivity
202 index computation) were implemented using spreadsheet-based tools to ensure traceability and reproducibility.

203 **Results and Discussion:**

204 **Results**

205 **AHP Weighting Results**

206 The Analytic Hierarchy Process (AHP) was used to translate expert judgments into a clear set of weights for the four
207 productivity ratios (R1-R4). The weighting results assign 0.30 to R1 (effectiveness), 0.40 to R2 (labor efficiency),
208 0.23 to R3 (time efficiency), and 0.06 to R4 (quality/defect). This weighting pattern suggests that, in the context of
209 the calibration facility, labor efficiency (R2) is viewed as the primary driver of productivity, with effectiveness (R1)
210 and time efficiency (R3) playing secondary roles, while the defect ratio (R4) contributes relatively less to the overall
211 assessment.

212 The consistency test supports the soundness of these judgments. The analysis produced $\lambda_{max} = 4.010$, $CI = 0.003$,
213 and $CR = 0.004$, which is far below the commonly accepted threshold of 0.10. Therefore, the pairwise comparisons
214 are considered consistent, and the resulting weights can be confidently applied in the subsequent OMAX
215 calculations.

216 **OMAX Productivity Index Results**

217 Using the weights obtained from AHP, we then applied the Objective Matrix (OMAX) method to compute the
218 calibration facility's monthly productivity index for January-July 2025. The resulting performance scores and
219 productivity indices are reported in Table 1.

220 Table 1. Monthly OMAX performance score, productivity index PI (%) based on baseline 300, and month-to-month
221 performance change (Jan-Jul 2025)

Month (2025)	OMAX Performance Score	Productivity Index (PI, %)	Performance change vs. previous month (%)
January	162	-46.00	-
February	629	109.67	288.27
March	63	-79.00	-89.98
April	120	-60.00	90.48
May	348	16.00	190.00
June	912	204.00	162.07
July	215	-28.00	-76.43

222

223 The productivity results fluctuate markedly across the observation period. The strongest performance was recorded
 224 in June, when the productivity index reached 204%, showing a pronounced improvement relative to the reference
 225 level. By contrast, productivity fell to its lowest point in March, with an index of -79%, suggesting that the facility
 226 experienced its most severe operational shortfall during that month.

227 **Productivity Trend Analysis**

228 The month-to-month results show pronounced swings, which suggests that productivity in the calibration facility is
 229 highly responsive to day-to-day operational conditions. The most notable improvement occurred between January
 230 and February, when productivity rose by 288.27%, pointing to a rapid rebound from a weak starting point possibly
 231 driven by better workload handling, staffing alignment, or smoother process execution. The trend then reversed
 232 sharply from February to March, with productivity dropping by 89.98%, representing the steepest decline observed
 233 during the study period.

234 After March, the pattern becomes one of intermittent recovery followed by renewed decline. Productivity improved
 235 from April to May, surged strongly in June, and then fell again in July. Taken together, these movements indicate
 236 that productivity was not sustained at a stable level over time. Instead, it appears to be shaped by shifting
 237 performance across key drivers' effectiveness, labor utilization, time efficiency, and quality/defect outcomes.

238 Overall, these findings underline the practical value of the integrated OMAX-AHP approach: it does not only
 239 summarize performance into a single index, but also makes it possible to track productivity dynamics clearly,
 240 identifying both high-performing periods and critical downturns that merit managerial attention.

241 **Discussion**

242 The findings suggest that the proposed OMAX-AHP integrated framework can capture productivity changes in a
 243 calibration facility within an automotive manufacturing setting in a meaningful way. By pairing a ratio-based
 244 productivity measure (OMAX) with a transparent, multi-criteria weighting procedure (AHP), the approach offers a
 245 clearer and more informative picture than relying on a single productivity indicator.

246 **Interpretation of AHP Weighting Results**

247 The AHP results place labor efficiency (R2) as the strongest contributor to the overall productivity assessment,
 248 followed by effectiveness (R1) and time efficiency (R3), while the defect ratio (R4) receives the smallest weight.
 249 This pattern is in line with recent work in productivity and operations management, which consistently highlights
 250 workforce utilization as a major determinant of performance in technical service settings especially when outcomes
 251 depend on specialized skills, tacit knowledge, and expert judgment.

252 In a calibration facility, labor efficiency has practical consequences: it shapes throughput, influences turnaround
253 time, and ultimately affects service reliability. The relatively large weight assigned to R2 therefore suggests that
254 shifts in staffing availability, skill mix, or workload allocation are more likely to drive productivity swings than
255 short-term changes in defect rates. This interpretation also fits the broader ISO/IEC 17025 context, where
256 competence management and effective deployment of qualified personnel are frequently emphasized as key
257 conditions for sustaining both operational performance and compliance.

258 It is also important to interpret the low weight of R4 carefully. A smaller weight does not mean quality is ignored;
259 rather, it may indicate that defect levels are comparatively less variable than the other drivers over the period
260 observed. In accredited calibration environments, defect-related outcomes are often constrained by standardized
261 procedures, verification steps, and quality assurance controls. As a result, defects may contribute less to short-term
262 productivity variation than factors related to labor and process execution, even though quality remains fundamental
263 to the integrity of calibration services.

264 **Analysis of OMAX Productivity Fluctuations**

265 The OMAX analysis shows pronounced month-to-month swings, suggesting that productivity in the calibration
266 facility is highly dependent on operational conditions. The improvement observed in February points to a period
267 when resources were used more effectively and the process ran with better coordination. In contrast, the sharp drop
268 in March highlights how quickly productivity can deteriorate when the operation is disrupted whether due to
269 workload imbalance, staffing limitations, or less effective scheduling.

270 The strong performance recorded in June illustrates what the facility can achieve when the main productivity drivers
271 especially labor efficiency and effectiveness are managed well. However, the declines that follow high-performing
272 periods indicate that these gains are not yet consistently maintained. This pattern suggests that productivity
273 improvements may still depend on temporary conditions rather than being supported by stable control practices or
274 standardized routines that sustain performance over time.

275 Overall, the observed fluctuation aligns with recent discussions in the technical service and operations literature,
276 which note that productivity improvements often occur in bursts unless they are reinforced by continuous
277 monitoring, structured feedback, and disciplined follow-up actions. In this regard, the integrated OMAX–AHP
278 framework is useful not only for tracking when performance rises or falls, but also for clarifying which criteria are
279 most responsible for the changes making it easier to focus managerial attention on the true drivers of variation.

280 **Contribution of the Integrated OMAX–AHP Framework**

281 Compared with using OMAX on its own, combining it with AHP strengthens the methodological foundation of the
282 analysis. The integration replaces subjective or ad-hoc weighting with a structured procedure, supported by a
283 consistency check, so the weighting decisions are easier to justify. This matters in calibration facilities, where
284 productivity evaluations can carry practical consequences for example, informing resource allocation, shaping
285 improvement priorities, and supporting management reviews in environments aligned with ISO/IEC 17025.

286 The integrated approach also makes the results easier to act on. OMAX produces a single productivity index, but
287 AHP clarifies the logic behind that index by showing which ratios matter most and how strongly they influence the
288 overall score. With this added transparency, managers can concentrate improvement efforts on the highest-impact
289 levers such as workforce deployment and process effectiveness rather than spreading attention evenly across all
290 indicators. This type of “measurement plus prioritization” aligns with broader multi-criteria decision-making
291 perspectives that emphasize turning performance data into actionable guidance.

292 Finally, applying the OMAX–AHP framework in a calibration facility contributes to the productivity measurement
293 literature by extending it beyond the settings where it is most commonly tested (e.g., production lines or output-

294 oriented manufacturing processes). Calibration operations are technically constrained and quality-critical by design,
295 and demonstrating that the framework remains usable in this context helps broaden the empirical evidence for
296 OMAX–AHP integration.

297 **Managerial Implications**

298 From a practical standpoint, the results imply that productivity improvement in calibration facilities is likely to
299 benefit most from stronger labor-efficiency management for example, balancing workloads, aligning tasks with staff
300 competencies, and improving scheduling. Using the proposed framework to track productivity over time also helps
301 managers spot early signs of performance deterioration, so corrective actions can be taken before issues escalate.

302 Beyond day-to-day operations, the framework can function as a structured decision-support aid for management
303 reviews, internal audits, and continuous improvement initiatives. Because the weighting process is transparent and
304 supported by a consistency check, the resulting productivity assessment is easier to explain and defend both to
305 internal stakeholders and, when needed, to external auditors.

306 **Limitations and Future Research**

307 While this study offers useful insights, several limitations should be noted. First, the findings are drawn from a
308 single facility and cover only a seven-month period, so the results may not fully represent productivity behavior in
309 other calibration settings or over longer time horizons. Second, the weighting stage depends on expert judgments.
310 Although the consistency test indicates that these judgments are coherent, the priorities may still reflect local
311 conditions, practices, or managerial preferences specific to the case context.

312 Future work could strengthen and extend the framework in several ways. Applying it over a longer period and across
313 multiple calibration facilities would allow more robust comparisons and improve generalizability. Methodologically,
314 the framework could also be expanded using approaches such as fuzzy AHP or other hybrid MCDM techniques to
315 better account for uncertainty in expert assessments. Finally, connecting productivity indices with additional
316 outcomes such as measurement uncertainty indicators and customer satisfaction measures would provide a more
317 comprehensive view of calibration performance and its operational impact.

318 **Conclusion:**

319 This study developed and implemented an integrated OMAX–AHP framework to quantify and interpret productivity
320 performance in a calibration facility within an automotive manufacturing setting. The main motivation was to
321 strengthen conventional productivity measurement by addressing one of its most common weaknesses how criterion
322 weights are assigned. By combining OMAX with AHP, the weighting stage becomes more transparent and
323 defensible, rather than relying on informal or purely subjective judgment.

324 The AHP results highlight labor efficiency as the most influential contributor to productivity, followed by
325 effectiveness and time efficiency, while the defect ratio plays a comparatively smaller role in the overall index. The
326 consistency test confirms that the underlying expert comparisons are coherent, which supports the credibility of the
327 resulting weight structure. When incorporated into OMAX, these weights produce productivity indices that are
328 straightforward to interpret and useful for tracking month-to-month performance movements.

329 Across the January–July 2025 period, the OMAX analysis shows clear fluctuations, including marked improvements
330 and sharp downturns. This pattern indicates that productivity in calibration operations is sensitive to changing
331 operational conditions particularly how resources are utilized and how work is organized. A practical advantage of
332 the proposed approach is that it helps managers go beyond simply noting that productivity has changed. It also
333 clarifies which criteria are most responsible for the change, making it easier to prioritize corrective actions and focus
334 improvement efforts on the most influential drivers.

335 From an implementation standpoint, the framework functions as a structured decision-support tool for calibration
336 and measurement facilities working under technical and regulatory requirements such as ISO/IEC 17025. Because it
337 links operational records to an explicit weighting logic, the output can support routine performance reviews, internal
338 audit discussions, and continuous improvement programs especially in environments where performance judgments
339 must be clearly explained and justified.

340 Several limitations should be acknowledged. The study is based on a single case and a relatively short observation
341 window, which may constrain broader generalization. In addition, although AHP includes a consistency check, the
342 weights still reflect expert priorities that may be shaped by local practices and contextual conditions. Future research
343 could apply the framework across multiple facilities and longer time horizons, or extend it using advanced MCDM
344 variants (e.g., fuzzy or hybrid approaches) to better accommodate uncertainty and operational variability. Even with
345 these limitations, the present findings indicate that the OMAX–AHP integration provides a practical, credible, and
346 effective approach for productivity measurement and monitoring in calibration operations.

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