

# ENHANCING STRATEGIC DECISION-MAKING IN HIGHER EDUCATION INSTITUTIONS IN MOROCCO: FORMULATING COMPOSITE INDEXES FOR SVM MACHINE LEARNING MODELS

*APRIMORANDO A TOMADA DE DECISÃO ESTRATÉGICA EM  
INSTITUIÇÕES DE ENSINO SUPERIOR NO MARROCOS:  
FORMULAÇÃO DE ÍNDICES COMPOSTOS PARA MODELOS  
DE APRENDIZADO DE MÁQUINA SVM*

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## Resumo

This article explores the development and application of composite indexes as tools for simplifying and synthesizing complex institutional data within Higher Education Institutions (HEIs). By condensing multiple performance, growth, and success indicators into structured indexes, the approach enhances the interpretability of multidimensional datasets. These indexes are then integrated with Support Vector Machine (SVM) classifiers to identify institutional patterns and classify institutions into distinct strategic categories. The study highlights how this combination not only strengthens the predictive capacity of machine learning models but also provides clearer insights for top management. Ultimately, the integration of composite indexes with SVM models demonstrates a significant potential to support evidence-based strategic decision-making, offering HEIs a robust analytical

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framework for aligning institutional performance with long-term development objectives.

**Palavras-chave:** Strategic Decision-Making. Higher Education. Composite Indexes. SVM Machine Learning. Morocco.

### **Abstract**

Este artigo explora o desenvolvimento e a aplicação de índices compostos como ferramentas para simplificar e sintetizar dados institucionais complexos em Instituições de Ensino Superior (IES). Ao condensar múltiplos indicadores de desempenho, crescimento e sucesso em índices estruturados, a abordagem amplia a interpretabilidade de conjuntos de dados multidimensionais. Esses índices são então integrados a classificadores de Support Vector Machine (SVM) para identificar padrões institucionais e classificar as instituições em categorias estratégicas distintas. O estudo destaca como essa combinação não apenas fortalece a capacidade preditiva dos modelos de aprendizado de máquina, mas também fornece percepções mais claras para a alta gestão. Em última instância, a integração de índices compostos com modelos SVM demonstra um potencial significativo para apoiar a tomada de decisão estratégica baseada em evidências, oferecendo às IES uma estrutura analítica robusta para alinhar o desempenho institucional com os objetivos de desenvolvimento de longo prazo.

**Keywords:** Tomada de Decisão Estratégica. Ensino Superior. Índices Compostos. Aprendizado de Máquina SVM. Marrocos.

## INTRODUCTION

Higher Education Institutions (HEIs) today operate in increasingly complex and competitive environments, where strategic management requires balancing resource allocation, institutional performance evaluation, and stakeholder satisfaction. These institutions face mounting pressures to expand student access, improve educational quality, and respond to evolving labor market demands while maintaining financial sustainability and accountability (Altbach, Reisberg, & Rumbley, 2009). In such a context, effective management cannot rely solely on intuition or tradition; it must be informed by rigorous, data-driven approaches. As noted by Brynjolfsson and McElheran (2016), organizations that adopt systematic data analysis in their decision-making processes significantly outperform those that rely on conventional methods.

The challenge, however, lies in the inherent heterogeneity and multidimensionality of HEI data, which encompasses academic performance, institutional efficiency, research productivity, and stakeholder perceptions. Managing such diverse information often overwhelms decision-makers and obscures meaningful insights (Ketunen, 2015). To overcome this limitation, composite indexes have emerged as valuable tools, as they allow for the aggregation of multiple indicators into a simplified yet comprehensive representation of institutional performance and growth (Nardo et al., 2008). By condensing complexity into structured measures, these indexes facilitate comparability across institutions and enhance the interpretability of trends relevant to strategic management.

Still, even when aggregated into indexes, the complexity of institutional data requires advanced analytical methods to reveal underlying patterns and strategic positioning. Traditional descriptive analyses are often insufficient for this purpose. Machine learning techniques, particularly Support Vector Machines (SVMs), offer robust capabilities for handling multidimensional data and achieving high predictive accuracy (Cortes & Vapnik, 1995). When combined with composite indexes, SVM classifiers can provide decision-makers with a clearer mapping of institutional identities, enabling more precise categorization of HEIs and supporting long-term planning. This integration responds directly to the growing demand for evidence-based governance in higher education systems (Salmi, 2009).

The purpose of this study is to build on these developments by proposing a structured framework for formulating composite indexes tailored to HEIs, encompassing dimensions of academic success, institutional performance, and growth capacity. Furthermore, it aims to demonstrate how SVM classifiers can leverage these indexes

to generate reliable classifications that improve the clarity and robustness of strategic insights. By uniting the strengths of composite indexes and machine learning, this research seeks to contribute to more effective strategic management in HEIs, particularly in contexts where data complexity presents significant challenges.

## LITERATURE REVIEW

### Strategic Management in HEIs

Strategic management in Higher Education Institutions (HEIs) encompasses the systematic planning and coordination of academic, financial, and institutional activities to achieve long-term objectives. Academic planning is central to this process, as it ensures that teaching programs align with evolving labor market demands, research priorities, and broader societal needs. According to Shattock (2010), universities must continuously adapt their academic offerings to remain relevant while balancing pressures of quality assurance, accreditation standards, and international competitiveness. Financial management constitutes another critical pillar, particularly in an era where institutions face shrinking public resources and growing accountability requirements. Jongbloed (2015) highlights that financial strategies increasingly involve diversification of income sources, cost-efficiency measures, and transparent allocation of budgets to ensure sustainability. Beyond academics and finance, institutional positioning within ranking systems has become a key driver of strategic choices, with global rankings influencing not only student and faculty recruitment but also government funding and partnerships (Hazelkorn, 2015).

Yet the complexity of these domains is intensified by the nature of the data required to inform them. Decision-making in HEIs must often reconcile quantitative measures—such as enrollment statistics, graduation rates, and research output—with qualitative inputs, including student satisfaction, stakeholder perceptions, and institutional reputation. As Kettunen (2015) observes, the challenge lies in integrating such heterogeneous evidence into coherent management strategies without oversimplifying the underlying realities. This multidimensionality of higher education data underscores the need for frameworks that both respect its complexity and provide actionable insights for decision-makers.

## Composite Indexes in Higher Education

To address these challenges, composite indexes have emerged as an important tool in higher education governance. A composite index combines multiple indicators into a single measure, thereby offering a simplified representation of complex realities. Perhaps the most visible examples are global university rankings, such as the *Times Higher Education World University Rankings* and the *QS World University Rankings*, which synthesize data on teaching, research, citations, internationalization, and reputation into aggregated scores (Docampo, 2013). These rankings, while contested, have become powerful benchmarks that shape institutional strategies worldwide, influencing governments, funders, and students alike (Hazelkorn, 2015). In addition to rankings, other indexes—such as student satisfaction indices—provide systematic measures of the learning environment and teaching quality, offering valuable feedback for both institutional leaders and policymakers (Richardson, Slater, & Wilson, 2007).

The benefits of composite indexes lie in their ability to distill diverse sets of information into a form that is accessible to a wide range of stakeholders. Nardo et al. (2008) argue that indexes improve transparency, enable comparability across institutions, and facilitate the monitoring of progress toward strategic goals. They function as communication tools that translate complex realities into clear, actionable benchmarks. However, they are not without limitations. Questions of methodological validity such as the choice of indicators, weighting schemes, and normalization techniques have long been debated in the literature (Saisana, d’Hombres, & Saltelli, 2011). Misinterpretations or over-reliance on these indexes can lead to unintended consequences, such as the prioritization of activities that improve rankings at the expense of other mission-critical functions. This tension highlights the need for careful design and contextual adaptation of composite indexes tailored to the unique realities of HEIs.

## Machine Learning in HEIs

Parallel to these developments, machine learning techniques have increasingly been deployed to address the growing complexity of higher education data. Educational data mining and learning analytics have expanded rapidly as fields, offering new methods to extract actionable insights from large and heterogeneous datasets (Romero & Ventura, 2020). Applications include predicting student performance and dropout risk (Kotsiantis, Pierrakeas, & Pintelas, 2007), optimizing resource

allocation such as classroom scheduling and faculty deployment (Baker & Inventado, 2014), and identifying patterns in administrative operations. These tools not only allow for proactive interventions but also enhance institutional efficiency and accountability.

Among the wide range of machine learning models, Support Vector Machines (SVMs) stand out for their ability to handle high-dimensional and structured data with strong predictive performance. First introduced by Cortes and Vapnik (1995), SVMs work by constructing optimal hyperplanes that separate data into distinct categories. This makes them particularly effective for classification problems, where the goal is to categorize entities based on complex input features. In higher education, SVMs have been applied to tasks such as predicting student success, identifying at-risk learners, and classifying institutions based on multidimensional performance metrics (Delen, 2010). Their flexibility in managing both structured and partially unstructured data, combined with their robustness against overfitting, makes them especially suitable for addressing the challenges posed by HEI datasets.

## Gaps in Existing Research

While composite indexes and machine learning models have been widely discussed in the literature, their integration for strategic decision-making in HEIs remains underexplored. Composite indexes are primarily used for benchmarking and communication, whereas machine learning models are often applied to specific operational challenges such as student retention or resource allocation. Very few studies have systematically examined how composite indexes—when employed as structured inputs—can enhance the predictive accuracy and interpretability of machine learning models such as SVMs. As a result, there is a gap between the simplification of complex data through indexes and the analytical power of advanced modeling techniques.

This gap represents a promising direction for research. The integration of composite indexes with machine learning could offer HEIs a robust analytical framework capable of bridging complexity and decision-making. By combining the transparency and communicability of indexes with the predictive strength of SVMs, institutions may be better equipped to classify their strategic positioning, anticipate future challenges, and align management decisions with long-term objectives. Salmi (2009) emphasizes that data-informed strategies are essential for building world-class universities, and this integration could represent a practical step toward realizing that vision in diverse higher education contexts.

# METHODOLOGY

## Data Collection and Preprocessing

The data for this study is sourced exclusively from the official yearly statistics published by the Ministry of Higher Education, Scientific Research, and Innovation. The focus of this data collection effort is specifically on the higher education public sector, encompassing 13 universities along with 130 associated faculties and institutions. The data spans a comprehensive 10-year period, starting from 2012 and concluding in 2022.

The dataset is meticulously structured to include a range of essential statistical metrics pertinent to higher education institutions. These metrics consist of:

- **New Registrations per Year:** This metric tracks the number of newly enrolled students at each institution on an annual basis.

- **Total Number of Students:** This encompasses the overall student enrollment figures across all 13 universities and their affiliated faculties and institutions.

- **Number of Graduates per Year:** This metric captures the yearly count of students who successfully complete their respective academic programs and graduate from their institutions.

- **Number of Pedagogical and Administrative Staff:** This includes data on the total workforce within each institution, differentiating between pedagogical (Professors) and administrative roles (Administrators and technicians).

- **Occupational Capacity:** This metric assesses the capacity of each institution to accommodate students and staff, providing insights into infrastructure and resource utilization.

- **Data preparation:**

After the data collection process, data preparation incorporates a rigorous preprocessing methodology to ensure data integrity and reliability. This preprocessing includes three distinct steps:

- **Filtering:** Institutions displaying inconsistencies in their statistical reporting or experiencing operational disruptions are carefully filtered out from the dataset. This step results in a refined dataset focusing on 105 institutions with consistent and reliable data.

- **Data Cleaning:** The cleaning phase involves a meticulous review of the dataset to identify and rectify any anomalies or discrepancies. This ensures that the data is accurate and consistent across all entries, mitigating potential errors in



subsequent analysis.

- **Data Coding:** Once the dataset is refined and cleaned, a comprehensive data coding process is implemented. This step involves converting the raw data into a structured format suitable for analysis and inclusion in the scientific study. Each of the 105 institutions affiliated with the 13 universities is meticulously coded to facilitate meaningful analysis and interpretation of the collected data.

By adhering to these stringent data collection and preprocessing protocols, the resulting dataset is robust, reliable, and well-suited for conducting in-depth analyses and deriving meaningful insights into the higher education public sector's dynamics and trends over the specified 10-year timeframe.

Here are the recapitulative figures of the final form dataset:

| NI   |     |      |      |      |       |       |       |       |       |       |       | RI   |       |       |       |       |       |       |       |       |       |       |      |
|------|-----|------|------|------|-------|-------|-------|-------|-------|-------|-------|------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|------|
| Year | 1   | 2    | 3    | 4    | 5     | 6     | 7     | 8     | 9     | 10    |       | Year | 1     | 2     | 3     | 4     | 5     | 6     | 7     | 8     | 9     | 10    |      |
| A    | A1  | 1768 | 1200 | 1414 | 1119  | 1075  | 1130  | 1312  | 1331  | 1463  | 1238  | A1   | 3694  | 4251  | 4743  | 4540  | 4554  | 4675  | 4614  | 4865  | 5297  | 5354  |      |
|      | A2  | 714  | 1127 | 1293 | 1821  | 1613  | 1311  | 1886  | 1818  | 2234  | 2386  | A2   | 2568  | 3853  | 3705  | 4722  | 5056  | 6236  | 6946  | 7773  | 7008  | 7631  |      |
|      | A3  | 120  | 116  | 229  | 394   | 516   | 417   | 414   | 280   | 594   | 569   | A3   | 120   | 240   | 460   | 808   | 808   | 1248  | 1540  | 1612  | 2021  | 2472  |      |
|      | A4  | 321  | 473  | 391  | 287   | 537   | 544   | 768   | 575   | 848   | 929   | A4   | 1041  | 1133  | 1328  | 1272  | 1458  | 1665  | 2222  | 2172  | 2673  | 2665  |      |
|      | A5  | 865  | 714  | 849  | 998   | 969   | 1052  | 902   | 704   | 620   | 832   | A5   | 2225  | 2580  | 3132  | 3516  | 3806  | 3685  | 3643  | 3810  | 4050  | 3652  |      |
| B    | B1  | 3334 | 3205 | 2973 | 3129  | 3018  | 3215  | 3694  | 4296  | 5548  | 3318  | B1   | 3091  | 3400  | 3991  | 11041 | 11637 | 11294 | 12306 | 13178 | 13195 | 13040 |      |
|      | B2  | 3420 | 2778 | 3585 | 2358  | 2416  | 2316  | 2393  | 3102  | 4180  | 3821  | B2   | 9843  | 10678 | 14288 | 14029 | 11489 | 10335 | 10393 | 10525 | 10282 | 11311 |      |
|      | B3  | 3371 | 2438 | 2807 | 2856  | 2086  | 2747  | 2340  | 2293  | 2564  | 2576  | B3   | 9732  | 9532  | 11925 | 11002 | 8345  | 10989 | 10475 | 10678 | 10809 | 9364  |      |
|      | B4  | 517  | 512  | 525  | 542   | 560   | 576   | 822   | 665   | 669   | 671   | B4   | 394   | 1030  | 1146  | 1187  | 1260  | 1406  | 1615  | 1818  | 1195  | 1726  |      |
|      | B5  | 4453 | 4751 | 3792 | 5241  | 5163  | 5360  | 4685  | 5167  | 6709  | 5204  | B5   | 10099 | 11522 | 13276 | 15358 | 11697 | 17877 | 17606 | 17558 | 17645 | 17116 |      |
| C    | B6  | 3422 | 3188 | 2896 | 3892  | 4345  | 3387  | 4007  | 5197  | 6892  | 6263  | B6   | 7436  | 9120  | 9977  | 11377 | 13076 | 13209 | 13893 | 14469 | 13786 | 14636 |      |
|      | B7  | 676  | 570  | 494  | 475   | 637   | 678   | 702   | 813   | 785   | 842   | B7   | 4799  | 5456  | 5615  | 5785  | 6158  | 6500  | 5611  | 6797  | 7315  | 6789  |      |
|      | B8  | 175  | 159  | 185  | 156   | 150   | 149   | 151   | 185   | 189   | 171   | B8   | 748   | 830   | 938   | 1115  | 1058  | 1066  | 834   | 1000  | 1077  | 977   |      |
|      | B9  | 75   | 97   | 145  | 115   | 155   | 117   | 169   | 217   | 187   | 221   | B9   | 706   | 845   | 952   | 1063  | 1163  | 1495  | 1940  | 1388  | 2130  | 2377  | 2189 |
|      | C1  | 5400 | 4619 | 7072 | 7483  | 7366  | 8140  | 8845  | 11527 | 10325 | 13609 | C1   | 16800 | 19014 | 20428 | 22827 | 24213 | 25016 | 26641 | 28197 | 29078 | 35151 |      |
| D    | C2  | 2173 | 1317 | 1719 | 1395  | 2301  | 3318  | 3844  | 4051  | 3372  | 4639  | C2   | 5448  | 6149  | 6551  | 8029  | 8647  | 9518  | 10420 | 11001 | 10493 | 12186 |      |
|      | C3  | 2226 | 1917 | 2816 | 2346  | 2002  | 1727  | 2070  | 2017  | 2573  | 2431  | C3   | 6633  | 7451  | 7457  | 8204  | 8148  | 7578  | 7228  | 7286  | 7877  | 8178  |      |
|      | C4  | 537  | 476  | 482  | 451   | 507   | 484   | 555   | 611   | 574   | 651   | C4   | 4789  | 4106  | 4358  | 4063  | 4667  | 4260  | 4058  | 4318  | 4689  | 5817  |      |
|      | C5  | 184  | 476  | 166  | 167   | 173   | 185   | 178   | 182   | 190   | 209   | C5   | 714   | 769   | 829   | 902   | 863   | 880   | 850   | 794   | 950   | 954   |      |
|      | C6  | 474  | 511  | 567  | 458   | 505   | 617   | 790   | 678   | 344   | 716   | C6   | 851   | 998   | 1058  | 1230  | 1160  | 1306  | 1520  | 1575  | 1504  | 1420  |      |
| E    | C7  | 4380 | 4049 | 3382 | 2621  | 3277  | 3331  | 4373  | 4321  | 5659  | 6674  | C7   | 12830 | 14922 | 17650 | 16492 | 17422 | 18836 | 20957 | 23170 | 26373 | 29373 |      |
|      | C8  | 1523 | 1705 | 1638 | 2268  | 2398  | 3664  | 2235  | 2942  | 2336  | 3379  | C8   | 4469  | 5098  | 4974  | 6054  | 7352  | 3885  | 3628  | 10187 | 9629  | 10062 |      |
|      | C9  | 1590 | 1477 | 1879 | 1257  | 1327  | 1321  | 1664  | 1782  | 1440  | 1979  | C9   | 4124  | 4767  | 5171  | 5375  | 5255  | 5124  | 5145  | 5125  | 4696  | 5458  |      |
|      | C10 | 1431 | 1662 | 1598 | 1360  | 1347  | 1364  | 1557  | 1885  | 2014  | 2530  | C10  | 3859  | 4345  | 4372  | 4633  | 4470  | 4489  | 5118  | 6406  | 7175  | 6327  |      |
|      | C11 | 4571 | 3868 | 3277 | 2346  | 2493  | 2275  | 3036  | 2917  | 3571  | 3210  | C11  | 9964  | 10283 | 10207 | 11342 | 10788 | 11952 | 12346 | 10046 | 11052 | 9263  |      |
| F    | C12 | 845  | 1050 | 828  | 944   | 907   | 1106  | 1375  | 1326  | 1048  | 1369  | C12  | 2902  | 3061  | 3301  | 3526  | 3734  | 3670  | 4311  | 4010  | 4269  | 4321  |      |
|      | C13 | 100  | 235  | 324  | 149   | 169   | 165   | 209   | 262   | 236   | 351   | C13  | 212   | 335   | 549   | 715   | 828   | 871   | 1041  | 1088  | 1181  | 1309  |      |
|      | C14 | 384  | 331  | 362  | 366   | 329   | 490   | 497   | 490   | 536   | 507   | C14  | 1779  | 2020  | 2365  | 2484  | 2495  | 2831  | 3061  | 3105  | 3050  | 3136  |      |
|      | C15 | 127  | 156  | 144  | 157   | 22    | 89    | 87    | 127   | 135   | 73    | C15  | 583   | 552   | 678   | 736   | 774   | 710   | 872   | 833   | 963   | 990   |      |
|      | D1  | 4181 | 8696 | 6274 | 12189 | 7851  | 8217  | 9098  | 10360 | 11601 | 10177 | D1   | 19580 | 22318 | 24257 | 28426 | 32289 | 29048 | 31220 | 33392 | 40458 | 35302 |      |
| G    | D2  | 3659 | 2344 | 3202 | 3647  | 5529  | 3835  | 3467  | 2809  | 4301  | 3286  | D2   | 15087 | 14958 | 14748 | 14944 | 13531 | 12912 | 11829 | 10340 | 12003 | 13017 |      |
|      | D3  | 1611 | 1850 | 2546 | 2860  | 1799  | 2002  | 2178  | 2663  | 3226  | 2645  | D3   | 9082  | 9665  | 9685  | 11937 | 8474  | 8193  | 8053  | 8481  | 9255  | 9393  |      |
|      | D4  | 2533 | 2604 | 2140 | 2393  | 1571  | 1740  | 2158  | 1839  | 2383  | 2619  | D4   | 9486  | 10260 | 12047 | 11044 | 9147  | 8518  | 9121  | 8333  | 8544  | 9115  |      |
|      | D5  | 2358 | 2710 | 3624 | 4276  | 5529  | 9717  | 9029  | 2863  | 3213  | 3392  | D5   | 10674 | 11010 | 13135 | 13424 | 11592 | 10899 | 10082 | 10363 | 11294 | 12247 |      |
|      | D6  | 721  | 809  | 853  | 907   | 734   | 753   | 917   | 935   | 735   | 942   | D6   | 3444  | 3732  | 3671  | 4055  | 3924  | 3934  | 3947  | 4093  | 3883  | 3329  |      |
| H    | D7  | 277  | 426  | 412  | 383   | 319   | 367   | 387   | 369   | 411   | 429   | D7   | 1789  | 2445  | 2882  | 2800  | 2717  | 3497  | 3323  | 3769  | 3507  | 3575  |      |
|      | D8  | 90   | 162  | 149  | 160   | 151   | 186   | 253   | 270   | 224   | 283   | D8   | 626   | 711   | 771   | 833   | 902   | 999   | 1115  | 1206  | 1290  | 1327  |      |
|      | D9  | 104  | 119  | 125  | 147   | 129   | 235   | 272   | 273   | 290   | 298   | D9   | 504   | 597   | 688   | 780   | 869   | 1029  | 1179  | 1378  | 1514  | 1581  |      |
|      | D10 | 750  | 815  | 501  | 850   | 828   | 759   | 957   | 960   | 967   | 959   | D10  | 1560  | 1647  | 1567  | 1687  | 1821  | 1842  | 1909  | 2052  | 1043  | 1320  |      |
|      | E1  | 8244 | 5902 | 6946 | 8885  | 10259 | 10770 | 12140 | 12183 | 11132 | 13758 | E1   | 15402 | 17252 | 20263 | 23560 | 27266 | 29618 | 31600 | 34580 | 34630 | 39130 |      |



|    |     |        |       |        |        |       |       |       |        |       |        |       |       |       |       |       |       |        |        |        |        |        |
|----|-----|--------|-------|--------|--------|-------|-------|-------|--------|-------|--------|-------|-------|-------|-------|-------|-------|--------|--------|--------|--------|--------|
| F  | E2  | 5075   | 5180  | 6420   | 6448   | 4455  | 4465  | 5 241 | 5 072  | 5 641 | 6 306  | E2    | 13730 | 16067 | 17395 | 21224 | 19016 | 17412  | 17375  | 18177  | 18 535 | 19 474 |
|    | E3  | 5075   | 4415  | 3922   | 4230   | 3241  | 3233  | 4 041 | 4 020  | 4 230 | 4 109  | E3    | 3718  | 11221 | 11601 | 13841 | 10437 | 11117  | 10618  | 11681  | 11 527 | 10 649 |
|    | E4  | 3952   | 3223  | 2298   | 2713   | 2468  | 2447  | 3 753 | 3 539  | 3 663 | 5 174  | E4    | 6389  | 6644  | 6413  | 6308  | 7554  | 7051   | 7170   | 8401   | 9 792  | 12 119 |
|    | E5  | 744    | 1014  | 744    | 1018   | 754   | 962   | 1 276 | 858    | 917   | 1 018  | E5    | 2504  | 2818  | 2332  | 2236  | 2646  | 2463   | 2660   | 2956   | 3 683  | 3 467  |
|    | E6  | 324    | 269   | 353    | 310    | 354   | 390   | 379   | 371    | 408   | 409    | E6    | 2317  | 2318  | 2745  | 2745  | 3038  | 3340   | 3615   | 3850   | 4 168  | 4 244  |
|    | E7  | 144    | 277   | 211    | 186    | 168   | 157   | 281   | 357    | 255   | 302    | E7    | 515   | 680   | 794   | 851   | 907   | 969    | 1051   | 1197   | 1 257  | 1 354  |
|    | E8  | 122    | 302   | 153    | 183    | 109   | 242   | 194   | 192    | 181   | 190    | E8    | 701   | 744   | 788   | 767   | 699   | 737    | 743    | 834    | 905    | 957    |
|    | E9  | 153    | 155   | 144    | 168    | 174   | 220   | 268   | 243    | 305   | 308    | E9    | 725   | 716   | 780   | 880   | 866   | 1023   | 1127   | 1267   | 1 414  | 1 526  |
|    | E10 | 243    | 237   | 263    | 243    | 259   | 466   | 411   | 488    | 459   | 542    | E10   | 517   | 479   | 563   | 582   | 603   | 1068   | 905    | 1051   | 1 116  | 1 161  |
|    | E11 | 183    | 137   | 236    | 177    | 308   | 440   | 288   | 266    | 425   | 425    | E11   | 339   | 366   | 410   | 526   | 674   | 1029   | 793    | 817    | 1 013  | 1 113  |
|    | G   | F1     | 3259  | 3240   | 3303   | 3940  | 3508  | 3 583 | 3 864  | 4 363 | 4 793  | 6 355 | F1    | 7455  | 8539  | 9517  | 11302 | 12 508 | 12 677 | 13101  | 14332  | 17681  |
| F2 |     | 2172   | 3132  | 2983   | 3028   | 3182  | 3 233 | 2 593 | 1 679  | 2 702 | 2 140  | F2    | 4830  | 6650  | 6682  | 8339  | 8621  | 9106   | 9291   | 4757   | 3 231  | 7 531  |
| F3 |     | 837    | 766   | 666    | 639    | 650   | 956   | 844   | 983    | 751   | 1 332  | F3    | 2560  | 2765  | 1940  | 2154  | 3254  | 3652   | 3884   | 3937   | 4 380  | 4 663  |
| F4 |     | 40     | 77    | 64     | 70     | 89    | 185   | 185   | 282    | 155   | 619    | F4    | 40    | 111   | 153   | 140   | 196   | 341    | 457    | 582    | 638    | 952    |
| F5 |     | 1766   | 1546  | 1257   | 1838   | 1 415 | 1 045 | 1 464 | 1 619  | 1 826 | 989    | F5    | 4115  | 3535  | 3487  | 3702  | 3846  | 3057   | 3594   | 4093   | 5 343  | 5 882  |
| F6 |     | 152    | 165   | 153    | 145    | 149   | 170   | 245   | 287    | 263   | 237    | F6    | 615   | 739   | 794   | 806   | 820   | 806    | 304    | 1020   | 1 107  | 1 194  |
| H  | G1  | 4 143  | 3 775 | 4 646  | 6 270  | 5 184 | 5 765 | 5 765 | 6 375  | 6 330 | 6 673  | G1    | 12405 | 13726 | 15427 | 18639 | 18856 | 21619  | 25242  | 26641  | 26 746 | 28 132 |
|    | G2  | 3 273  | 2 169 | 2 742  | 3 422  | 2 685 | 2 700 | 2 700 | 3 225  | 2 585 | 3 203  | G2    | 8357  | 9085  | 13903 | 10970 | 11300 | 13148  | 13805  | 14040  | 13 643 | 14 170 |
|    | G3  | 1 965  | 2 148 | 1 689  | 2 136  | 1 487 | 1 420 | 1 420 | 3 225  | 1 795 | 3 203  | G3    | 6408  | 7560  | 6532  | 7016  | 6321  | 6373   | 6665   | 6885   | 6 315  | 6 641  |
|    | G4  | 2 832  | 2 435 | 3 218  | 4 202  | 3 451 | 4 550 | 4 550 | 4 957  | 5 136 | 5 713  | G4    | 8482  | 8687  | 10867 | 13035 | 13848 | 17338  | 19793  | 22030  | 23 028 | 27 785 |
|    | G5  | 228    | 276   | 295    | 305    | 330   | 326   | 326   | 348    | 355   | 326    | G5    | 1071  | 1345  | 1482  | 1736  | 1847  | 2252   | 2509   | 2562   | 2 729  | 2 376  |
|    | G6  | 160    | 110   | 215    | 151    | 162   | 152   | 152   | 273    | 308   | 285    | G6    | 809   | 921   | 930   | 933   | 950   | 948    | 1046   | 1070   | 1 073  | 1 176  |
|    | G7  | 181    | 126   | 118    | 137    | 135   | 174   | 174   | 240    | 240   | 236    | G7    | 672   | 709   | 716   | 719   | 636   | 772    | 833    | 926    | 1 033  | 1 101  |
|    | G8  | 511    | 546   | 331    | 554    | 539   | 604   | 604   | 1 081  | 862   | 818    | G8    | 1199  | 1055  | 909   | 989   | 1136  | 1371   | 1896   | 2196   | 2 401  | 2 277  |
| I  | H1  | 3752   | 4722  | 6274   | 5085   | 6355  | 6 390 | 7 537 | 9 009  | 8 308 | 8 903  | H1    | 14038 | 16298 | 18808 | 23281 | 21627 | 28985  | 32064  | 32972  | 35 356 | 37 828 |
|    | H2  | 2410   | 2635  | 3668   | 3112   | 3402  | 3 375 | 3 472 | 3 185  | 2 633 | 3 237  | H2    | 7861  | 8980  | 10758 | 13211 | 14491 | 14719  | 15839  | 16 292 | 17 231 |        |
|    | H3  | 1521   | 1835  | 1972   | 1854   | 1417  | 1 433 | 1 855 | 2 207  | 2 234 | 2 436  | H3    | 5937  | 7681  | 7734  | 8654  | 8667  | 8719   | 8074   | 8484   | 8 622  | 9 741  |
|    | H4  | 3094   | 3477  | 4816   | 4568   | 6399  | 7 207 | 9 213 | 8 521  | 7 329 | 7 662  | H4    | 7124  | 7705  | 19170 | 14827 | 22638 | 21132  | 25933  | 28370  | 30 295 | 28 353 |
|    | H5  | 839    | 896   | 771    | 791    | 1596  | 1 351 | 1 525 | 1 510  | 4 078 | 1 430  | H5    | 2810  | 3052  | 2842  | 3041  | 4262  | 5859   | 6060   | 6899   | 7 385  | 7 688  |
|    | H6  | 835    | 869   | 1129   | 955    | 1033  | 1 128 | 1 435 | 1 278  | 359   | 1 236  | H6    | 3350  | 3639  | 4478  | 4639  | 4783  | 5222   | 5705   | 5765   | 6 065  | 5 800  |
|    | H7  | 240    | 347   | 120    | 192    | 183   | 237   | 289   | 307    | 356   | 349    | H7    | 240   | 351   | 370   | 449   | 528   | 685    | 839    | 104    | 1 305  | 1 475  |
|    | H8  | 182    | 226   | 271    | 191    | 192   | 223   | 280   | 347    | 253   | 235    | H8    | 677   | 1052  | 1086  | 1122  | 1164  | 1219   | 1347   | 1426   | 1 468  | 1 678  |
|    | H9  | 248    | 193   | 230    | 165    | 173   | 210   | 262   | 306    | 281   | 297    | H9    | 906   | 951   | 929   | 858   | 1021  | 1100   | 1111   | 1072   | 1 173  | 1 327  |
|    | H10 | 309    | 307   | 340    | 314    | 312   | 385   | 446   | 491    | 474   | 482    | H10   | 1215  | 1385  | 1477  | 1563  | 1673  | 1816   | 2014   | 2308   | 2 448  | 2 555  |
|    | H11 | 162    | 138   | 121    | 112    | 137   | 142   | 241   | 219    | 221   | 290    | H11   | 681   | 717   | 753   | 747   | 743   | 823    | 943    | 942    | 1 022  | 1 162  |
| J  | I1  | 5 762  | 4 960 | 4 813  | 6 297  | 5 631 | 6 782 | 7 126 | 6 849  | 7 530 | 7 966  | I1    | 12628 | 14131 | 18028 | 19617 | 19928 | 21877  | 24 562 | 24 802 | 28 315 | 26 461 |
|    | I2  | 6 492  | 5 658 | 3 953  | 4 120  | 2 386 | 2 137 | 2 346 | 2 137  | 3 631 | 3 313  | I2    | 12005 | 13250 | 14096 | 14937 | 13857 | 12758  | 11 813 | 10 843 | 12 077 | 13 810 |
|    | I3  | 2 468  | 1 986 | 3 541  | 2 110  | 1 396 | 1 372 | 1 610 | 1 370  | 1 837 | 1 483  | I3    | 8123  | 5866  | 3086  | 6833  | 6376  | 6440   | 6 439  | 5 625  | 5 983  | 5 706  |
|    | I4  | 1 504  | 1 659 | 2 446  | 3 060  | 3 330 | 3 382 | 4 696 | 4 324  | 5 000 | 4 395  | I4    | 4720  | 4297  | 6003  | 7540  | 11912 | 15286  | 17 865 | 20 115 | 23 224 | 25 002 |
|    | I5  | 722    | 730   | 336    | 332    | 355   | 926   | 1 002 | 918    | 684   | 1 195  | I5    | 2522  | 2956  | 3274  | 3329  | 3590  | 3864   | 4130   | 4131   | 3 806  | 4 132  |
|    | I6  | 261    | 321   | 318    | 235    | 310   | 291   | 433   | 414    | 372   | 476    | I6    | 1194  | 1358  | 1469  | 1533  | 1607  | 1663   | 1739   | 1796   | 1 902  | 2 072  |
|    | I7  | 1 038  | 788   | 859    | 865    | 842   | 762   | 895   | 956    | 990   | 929    | I7    | 1925  | 1730  | 1818  | 1773  | 2037  | 1908   | 1884   | 2028   | 2 092  | 2 041  |
| K  | J1  | 1936   | 1534  | 1514   | 2203   | 2202  | 2115  | 2 483 | 2 810  | 2 155 | 2 716  | J1    | 4449  | 5294  | 4863  | 5679  | 6218  | 6623   | 6878   | 7524   | 6 828  | 7 644  |
|    | J2  | 2555   | 2424  | 2066   | 2584   | 1847  | 1891  | 1 948 | 2 660  | 2 155 | 2 644  | J2    | 6042  | 6533  | 6108  | 6395  | 6443  | 6626   | 6390   | 7048   | 6 476  | 7 184  |
|    | J3  | 477    | 451   | 527    | 740    | 322   | 1089  | 1 220 | 1 427  | 3 960 | 2 419  | J3    | 1370  | 1356  | 1773  | 2122  | 2412  | 3248   | 3859   | 4568   | 7 392  | 8 855  |
|    | J4  | 81     | 118   | 89     | 100    | 104   | 165   | 191   | 243    | 189   | 223    | J4    | 266   | 349   | 408   | 441   | 476   | 557    | 661    | 786    | 874    | 928    |
|    | J5  | 147    | 127   | 141    | 153    | 160   | 212   | 237   | 259    | 235   | 251    | J5    | 528   | 511   | 643   | 665   | 799   | 916    | 1067   | 1222   | 1 353  | 1 455  |
| L  | K1  | 1 253  | 1 475 | 1 655  | 2 122  | 2 326 | 4 359 | 4 876 | 5 542  | 7 303 | 10161  | K1    | 3 529 | 4 369 | 5 511 | 6 655 | 8436  | 11116  | 13577  | 17169  | 23438  | 30733  |
|    | K2  | 6 179  | 6 590 | 5 953  | 6 034  | 6 376 | 3 559 | 4 387 | 4 393  | 5398  | 12171  | K2    | 13914 | 16721 | 20922 | 24888 | 25372 | 21961  | 20044  | 22903  | 25799  | 33622  |
|    | K3  | 3 760  | 4 509 | 1 868  | 2 227  | 2 204 | 1 145 | 1 559 | 1 315  | 1 914 | 2 379  | K3    | 9074  | 10971 | 10548 | 8309  | 9689  | 9369   | 8505   | 8419   | 9 090  | 10 291 |
|    | K4  | 158    | 183   | 172    | 177    | 208   | 217   | 260   | 278    | 262   | 280    | K4    | 612   | 779   | 883   | 1001  | 1046  | 1126   | 1216   | 1333   | 1 377  | 1 561  |
|    | K5  | 225    | 227   | 223    | 212    | 205   | 266   | 342   | 392    | 400   | 385    | K5    | 923   | 1122  | 1345  | 1469  | 1587  | 1760   | 2037   | 2357   | 2 436  | 2 740  |
| M  | L1  | 3716   | 3792  | 3872   | 5260   | 5040  | 5398  | 7 713 | 9 263  | 7477  | 7815   | L1    | 8449  | 10486 | 12374 | 16815 | 20338 | 23889  | 28 922 | 34 939 | 36029  | 40464  |
|    | L2  | 802    | 908   | 972    | 878    | 862   | 958   | 1 277 | 1 081  | 972   | 1 181  | L2    | 3687  | 3899  | 4569  | 4547  | 4396  | 4909   | 5085   | 4822   | 4 311  | 4 314  |
|    | L3  | 282    | 285   | 302    | 291    | 280   | 430   | 388   | 479    | 527   | 484    | L3    | 1277  | 1532  | 1501  | 1657  | 1761  | 1950   | 2039   | 2258   | 2 768  | 2 692  |
|    | L4  | 193    | 211   | 187    | 192    | 173   | 178   | 179   | 160    | 180   | 99     | L4    | 377   | 389   | 402   | 354   | 430   | 442    | 491    | 416    | 534    | 476    |
| N  | M1  | 11 497 | 9 214 | 11 124 | 11 164 | 8 201 | 8 430 | 8 233 | 10 195 | 8 670 | 11 755 | M1    | 25267 | 30311 | 33794 | 42625 | 39320 | 33431  | 33724  | 37129  | 38 436 | 42 964 |
|    | M2  | 7 048  | 6 081 | 8 894  | 7 789  | 6 653 | 3 884 | 4 380 | 4 806  | 4 384 | 5 243  | M2    | 23422 | 26103 | 28306 | 3182  |       |        |        |        |        |        |



|    |     |       |       |       |       |       |       |       |       |       |       |
|----|-----|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| F  | E2  | 374   | 1662  | 2 056 | 2 627 | 2876  | 2573  | 2 809 | 2 611 | 2 638 | 2 625 |
|    | E3  | 827   | 1 076 | 1 115 | 1 554 | 1421  | 1581  | 1 473 | 1 474 | 1532  | 1670  |
|    | E4  | 571   | 592   | 598   | 760   | 828   | 798   | 728   | 882   | 838   | 921   |
|    | E5  | 387   | 509   | 641   | 574   | 604   | 627   | 539   | 574   | 748   | 760   |
|    | E6  | 138   | 163   | 201   | 196   | 289   | 343   | 366   | 361   | 390   | 473   |
|    | E7  | 81    | 30    | 39    | 31    | 123   | 137   | 154   | 161   | 165   | 181   |
|    | E8  | 139   | 113   | 159   | 142   | 135   | 160   | 141   | 138   | 143   | 153   |
|    | E9  | 89    | 31    | 129   | 173   | 200   | 202   | 211   | 185   | 206   | 234   |
|    | E10 | 235   | 242   | 234   | 269   | 454   | 345   | 427   | 156   | 568   | 536   |
|    | E11 | 134   | 116   | 210   | 319   | 286   | 448   | 473   | 470   | 514   | 612   |
|    | F1  | 880   | 1 102 | 1 175 | 1 453 | 1 464 | 1 612 | 1 789 | 2 030 | 2 159 | 2 124 |
| F2 | 403 | 398   | 862   | 1 223 | 1386  | 1351  | 1 705 | 1 873 | 2 166 | 1 863 |       |
| F3 | 316 | 357   | 339   | 538   | 784   | 603   | 679   | 690   | 920   | 732   |       |
| F4 | 0   | 27    | 65    | 62    | 63    | 146   | 203   | 247   | 293   | 271   |       |
| F5 | 388 | 448   | 528   | 333   | 523   | 549   | 433   | 530   | 521   | 557   |       |
| F6 | 53  | 121   | 126   | 123   | 166   | 155   | 142   | 152   | 153   | 186   |       |
| G  | G1  | 1 126 | 1 262 | 1 103 | 1 347 | 1439  | 1441  | 1 405 | 1 733 | 2 156 | 1 819 |
|    | G2  | 1 007 | 1 088 | 1 289 | 1 108 | 1320  | 1299  | 1 424 | 1 587 | 1 642 | 1 377 |
|    | G3  | 825   | 902   | 817   | 912   | 1159  | 1265  | 1 171 | 1 085 | 1 180 | 1 111 |
|    | G4  | 564   | 558   | 759   | 790   | 906   | 960   | 868   | 1 316 | 1 540 | 1 552 |
|    | G5  | 0     | 0     | 0     | 121   | 194   | 164   | 255   | 188   | 330   | 321   |
|    | G6  | 140   | 171   | 174   | 175   | 178   | 195   | 179   | 174   | 182   | 130   |
|    | G7  | 99    | 126   | 124   | 191   | 145   | 152   | 119   | 126   | 157   | 153   |
|    | G8  | 555   | 427   | 374   | 411   | 435   | 501   | 523   | 1 082 | 1 174 | 1 046 |
| H  | H1  | 1 290 | 1 343 | 3 546 | 3 834 | 2612  | 3 718 | 3 443 | 4 294 | 4 175 |       |
|    | H2  | 1 123 | 1 319 | 1 310 | 3 834 | 1246  | 1638  | 1 802 | 1 547 | 1 907 | 1 838 |
|    | H3  | 1 175 | 1 213 | 1 208 | 3 834 | 1201  | 1399  | 1 427 | 1 334 | 1 377 | 1 265 |
|    | H4  | 317   | 1 184 | 2 280 | 2 743 | 3340  | 3615  | 3 635 | 3 915 | 5 913 | 4 905 |
|    | H5  | 178   | 334   | 349   | 422   | 415   | 530   | 658   | 732   | 826   | 728   |
|    | H6  | 785   | 860   | 1 611 | 1 090 | 890   | 970   | 937   | 1 010 | 1 214 | 1 140 |
|    | H7  | 0     | 0     | 86    | 120   | 104   | 141   | 206   | 236   | 303   | 342   |
|    | H8  | 108   | 206   | 168   | 182   | 141   | 216   | 235   | 243   | 301   | 230   |
|    | H9  | 99    | 141   | 148   | 192   | 167   | 218   | 223   | 164   | 162   | 203   |
|    | H10 | 190   | 188   | 231   | 223   | 269   | 294   | 324   | 278   | 392   | 429   |
|    | H11 | 0     | 0     | 100   | 191   | 139   | 159   | 166   | 204   | 139   | 171   |
| I  | I1  | 314   | 842   | 1 567 | 2 059 | 2063  | 2609  | 2 863 | 3 263 | 3 713 | 3 766 |
|    | I2  | 1 412 | 395   | 1 355 | 1 431 | 1409  | 1722  | 1 899 | 1 436 | 1 483 | 1 567 |
|    | I3  | 557   | 721   | 581   | 519   | 536   | 683   | 863   | 819   | 909   | 627   |
|    | I4  | 475   | 533   | 623   | 763   | 762   | 1137  | 1 276 | 1 647 | 1 867 | 2 125 |
|    | I5  | 219   | 319   | 443   | 500   | 409   | 541   | 640   | 824   | 812   | 635   |
|    | I6  | 188   | 179   | 173   | 243   | 222   | 288   | 280   | 174   | 254   | 169   |
|    | I7  | 855   | 824   | 174   | 950   | 993   | 986   | 859   | 314   | 216   | 932   |
| J  | J1  | 715   | 561   | 521   | 677   | 735   | 744   | 856   | 1 000 | 1 078 | 816   |
|    | J2  | 486   | 782   | 730   | 939   | 851   | 731   | 1 033 | 899   | 889   | 737   |
|    | J3  | 445   | 294   | 510   | 664   | 445   | 724   | 850   | 1 105 | 1 223 | 1 036 |
|    | J4  | 29    | 28    | 52    | 63    | 73    | 85    | 85    | 85    | 119   | 120   |
|    | J5  | 69    | 82    | 31    | 97    | 190   | 164   | 176   | 226   | 198   | 250   |
| K  | K1  | 688   | 545   | 650   | 511   | 546   | 752   | 928   | 1565  | 2160  | 2325  |
|    | K2  | 1 132 | 1 426 | 2 059 | 2 384 | 2198  | 2834  | 2 633 | 2911  | 2555  | 3012  |
|    | K3  | 860   | 914   | 1 028 | 1 048 | 1219  | 1267  | 1 251 | 1 782 | 1 317 | 1 435 |
|    | K4  | 54    | 116   | 130   | 181   | 182   | 233   | 165   | 213   | 189   | 212   |
|    | K5  | 123   | 248   | 228   | 314   | 385   | 394   | 473   | 508   | 616   | 610   |
| L  | L1  | 442   | 1 029 | 1 163 | 757   | 1592  | 2262  | 3 350 | 3421  | 4129  | 4332  |
|    | L2  | 674   | 314   | 383   | 1 020 | 928   | 1039  | 874   | 888   | 912   | 953   |
|    | L3  | 284   | 295   | 253   | 267   | 258   | 338   | 333   | 364   | 284   | 416   |
|    | L4  | 176   | 176   | 180   | 172   | 246   | 0     | 232   | 155   | 141   | 155   |
| M  | M1  | 2 110 | 2 390 | 2 687 | 3 423 | 3838  | 3615  | 3 520 | 4 395 | 5 222 | 4 343 |
|    | M2  | 4 300 | 3 414 | 3 456 | 3 824 | 4 071 | 3 861 | 3 478 | 2 990 | 3 006 | 2 784 |
|    | M3  | 1 224 | 1 435 | 1 538 | 1 590 | 1845  | 1893  | 1 710 | 1 400 | 1 376 | 934   |
|    | M4  | 184   | 252   | 410   | 532   | 452   | 395   | 522   | 482   | 519   | 638   |
|    | M5  | 164   | 379   | 582   | 460   | 381   | 323   | 328   | 351   | 568   | 546   |
|    | M6  | 162   | 227   | 191   | 198   | 244   | 227   | 219   | 223   | 287   | 247   |
|    | M7  | 297   | 233   | 474   | 371   | 432   | 463   | 500   | 501   | 545   | 386   |
|    | M8  | 495   | 451   | 539   | 476   | 589   | 580   | 618   | 594   | 687   | 101   |
|    | M9  | 0     | 0     | 387   | 136   | 144   | 197   | 282   | 444   | 472   | 309   |
|    | N1  | 1 132 | 1 426 | 2 059 | 2 384 | 2198  | 2834  | 2 633 | 2911  | 2555  | 3012  |
|    | N2  | 860   | 914   | 1 028 | 1 048 | 1219  | 1267  | 1 251 | 1 782 | 1 317 | 1 435 |
| O  | O1  | 54    | 116   | 130   | 181   | 182   | 233   | 165   | 213   | 189   | 212   |
|    | O2  | 123   | 248   | 228   | 314   | 385   | 394   | 473   | 508   | 616   | 610   |
|    | O3  | 442   | 1 029 | 1 163 | 757   | 1592  | 2262  | 3 350 | 3421  | 4129  | 4332  |
|    | O4  | 614   | 314   | 383   | 1 020 | 928   | 1039  | 874   | 888   | 912   | 953   |
| P  | P1  | 284   | 295   | 253   | 267   | 258   | 338   | 333   | 364   | 284   | 416   |
|    | P2  | 176   | 176   | 180   | 172   | 246   | 0     | 232   | 155   | 141   | 155   |
|    | P3  | 2 110 | 2 390 | 2 687 | 3 423 | 3838  | 3615  | 3 520 | 4 395 | 5 222 | 4 343 |
|    | P4  | 4 300 | 3 414 | 3 456 | 3 824 | 4 071 | 3 861 | 3 478 | 2 990 | 3 006 | 2 784 |
|    | P5  | 1 224 | 1 435 | 1 538 | 1 590 | 1845  | 1893  | 1 710 | 1 400 | 1 376 | 934   |
|    | P6  | 184   | 252   | 410   | 532   | 452   | 395   | 522   | 482   | 519   | 638   |
|    | P7  | 164   | 379   | 582   | 460   | 381   | 323   | 328   | 351   | 568   | 546   |
|    | P8  | 162   | 227   | 191   | 198   | 244   | 227   | 219   | 223   | 287   | 247   |
|    | P9  | 297   | 233   | 474   | 371   | 432   | 463   | 500   | 501   | 545   | 386   |
|    | P10 | 495   | 451   | 539   | 476   | 589   | 580   | 618   | 594   | 687   | 101   |
|    | P11 | 0     | 0     | 387   | 136   | 144   | 197   | 282   | 444   | 472   | 309   |
| Q  | Q1  | 1 132 | 1 426 | 2 059 | 2 384 | 2198  | 2834  | 2 633 | 2911  | 2555  | 3012  |
|    | Q2  | 860   | 914   | 1 028 | 1 048 | 1219  | 1267  | 1 251 | 1 782 | 1 317 | 1 435 |
|    | Q3  | 54    | 116   | 130   | 181   | 182   | 233   | 165   | 213   | 189   | 212   |
|    | Q4  | 123   | 248   | 228   | 314   | 385   | 394   | 473   | 508   | 616   | 610   |
|    | Q5  | 442   | 1 029 | 1 163 | 757   | 1592  | 2262  | 3 350 | 3421  | 4129  | 4332  |
|    | Q6  | 614   | 314   | 383   | 1 020 | 928   | 1039  | 874   | 888   | 912   | 953   |
|    | Q7  | 284   | 295   | 253   | 267   | 258   | 338   | 333   | 364   | 284   | 416   |
|    | Q8  | 176   | 176   | 180   | 172   | 246   | 0     | 232   | 155   | 141   | 155   |
|    | Q9  | 2 110 | 2 390 | 2 687 | 3 423 | 3838  | 3615  | 3 520 | 4 395 | 5 222 | 4 343 |
|    | Q10 | 4 300 | 3 414 | 3 456 | 3 824 | 4 071 | 3 861 | 3 478 | 2 990 | 3 006 | 2 784 |
|    | Q11 | 1 224 | 1 435 | 1 538 | 1 590 | 1845  | 1893  | 1 710 | 1 400 | 1 376 | 934   |
| R  | R1  | 184   | 252   | 410   | 532   | 452   | 395   | 522   | 482   | 519   | 638   |
|    | R2  | 164   | 379   | 582   | 460   | 381   | 323   | 328   | 351   | 568   | 546   |
|    | R3  | 162   | 227   | 191   | 198   | 244   | 227   | 219   | 223   | 287   | 247   |
|    | R4  | 297   | 233   | 474   | 371   | 432   | 463   | 500   | 501   | 545   | 386   |
|    | R5  | 495   | 451   | 539   | 476   | 589   | 580   | 618   | 594   | 687   | 101   |
|    | R6  | 0     | 0     | 387   | 136   | 144   | 197   | 282   | 444   | 472   | 309   |
|    | R7  | 1 132 | 1 426 | 2 059 | 2 384 | 2198  | 2834  | 2 633 | 2911  | 2555  | 3012  |
|    | R8  | 860   | 914   | 1 028 | 1 048 | 1219  | 1267  | 1 251 | 1 782 | 1 317 | 1 435 |
|    | R9  | 54    | 116   | 130   | 181   | 182   | 233   | 165   | 213   | 189   | 212   |
|    | R10 | 123   | 248   | 228   | 314   | 385   | 394   | 473   | 508   | 616   | 610   |
|    | R11 | 442   | 1 029 | 1 163 | 757   | 1592  | 2262  | 3 350 | 3421  | 4129  | 4332  |
| S  | S1  | 614   | 314   | 383   | 1 020 | 928   | 1039  | 874   | 888   | 912   | 953   |
|    | S2  | 284   | 295   | 253   | 267   | 258   | 338   | 333   | 364   | 284   | 416   |
|    | S3  | 176   | 176   | 180   | 172   | 246   | 0     | 232   | 155   | 141   | 155   |
|    | S4  | 2 110 | 2 390 | 2 687 | 3 423 | 3838  | 3615  | 3 520 | 4 395 | 5 222 | 4 343 |
|    | S5  | 4 300 | 3 414 | 3 456 | 3 824 | 4 071 | 3 861 | 3 478 | 2 990 | 3 006 | 2 784 |
|    | S6  | 1 224 | 1 435 | 1 538 | 1 590 | 1845  | 1893  | 1 710 | 1 400 | 1 376 | 934   |
|    | S7  | 184   | 252   | 410   | 532   | 452   | 395   | 522   | 482   | 519   | 638   |
|    | S8  | 164   | 379   | 582   | 460   | 381   | 323   | 328   | 351   | 568   | 546   |
|    | S9  | 162   | 227   | 191   | 198   | 244   | 227   | 219   | 223   | 287   | 247   |
|    | S10 | 297   | 233   | 474   | 371   | 432   | 463   | 500   | 501   | 545   | 386   |
|    | S11 | 495   | 451   | 539   | 476   | 589   | 580   | 618   | 594   | 687   | 101   |
| T  | T1  | 0     | 0     | 387   | 136   | 144   | 197   | 282   | 444   | 472   | 309   |
|    | T2  | 1 132 | 1 426 | 2 059 | 2 384 | 2198  | 2834  | 2 633 | 2911  | 2555  | 3012  |
|    | T3  | 860   | 914   | 1 028 | 1 048 | 1219  | 1267  | 1 251 | 1 782 | 1 317 | 1 435 |
|    | T4  | 54    | 116   | 130   | 181   | 182   | 233   | 165   | 213   | 189   | 212   |
|    | T5  | 123   | 248   | 228   | 314   | 385   | 394   | 473   | 508   | 616   | 610   |
|    | T6  | 442   | 1 029 | 1 163 | 757   | 1592  | 2262  | 3 350 | 3421  | 4129  | 4332  |
|    | T7  | 614   | 314   | 383   | 1 020 | 928   | 1039  | 874   | 888   | 912   | 953   |
|    | T8  | 284   | 295   | 253   | 267   | 258   | 338   | 333   | 364   | 284   | 416   |
|    | T9  | 176   | 176   | 180   | 172   | 246   | 0     | 232   | 155   |       |       |

|     |     |      |       |       |       |       |       |       |       |       |       |
|-----|-----|------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| F   | E2  | 41   | 84    | 75    | 71    | 64    | 60    | 63    | 53    | 57    | 56    |
|     | E3  | 91   | 172   | 166   | 146   | 133   | 133   | 134   | 118   | 104   | 104   |
|     | E4  | 13   | 31    | 32    | 33    | 36    | 30    | 31    | 37    | 36    | 38    |
|     | E5  | 43   | 80    | 78    | 76    | 71    | 63    | 66    | 67    | 67    | 75    |
|     | E6  | 42   | 78    | 78    | 76    | 79    | 79    | 79    | 74    | 73    | 75    |
|     | E7  | 14   | 32    | 30    | 31    | 31    | 32    | 32    | 26    | 29    | 24    |
|     | E8  | 9    | 23    | 22    | 22    | 22    | 16    | 17    | 17    | 18    | 18    |
|     | E9  | 8    | 23    | 22    | 21    | 17    | 16    | 17    | 19    | 20    | 20    |
|     | E10 | 14   | 40    | 40    | 39    | 38    | 36    | 36    | 35    | 38    | 37    |
|     | E11 | 8    | 17    | 16    | 16    | 20    | 17    | 20    | 22    | 19    | 20    |
|     | F1  | 27   | 61    | 60    | 56    | 54    | 56    | 57    | 60    | 61    | 58    |
| G   | F2  | 3    | 10    | 10    | 13    | 23    | 22    | 23    | 31    | 26    | 28    |
|     | F3  | 22   | 63    | 65    | 68    | 67    | 59    | 59    | 58    | 26    | 61    |
|     | F4  |      | 2     | 3     | 4     | 5     | 6     | 7     | 9     | 12    |       |
|     | F5  | 18   | 18    | 23    | 23    | 25    | 31    | 31    | 27    | 26    | 28    |
|     | F6  | 27   | 27    | 26    | 24    | 26    | 22    | 22    | 22    | 23    | 23    |
|     | G1  | 40   | 84    | 78    | 73    | 74    | 72    | 68    | 59    | 49    | 56    |
|     | G2  | 44   | 84    | 78    | 71    | 67    | 66    | 63    | 66    | 59    | 66    |
|     | G3  | 49   | 118   | 114   | 102   | 97    | 94    | 93    | 88    | 81    | 78    |
|     | G4  | 11   | 33    | 33    | 37    | 41    | 40    | 41    | 43    | 45    | 50    |
|     | G5  | 14   | 25    | 24    | 25    | 23    | 25    | 25    | 28    | 31    | 32    |
|     | G6  | 10   | 18    | 20    | 22    | 21    | 21    | 21    | 24    | 25    | 26    |
| H   | G7  | 5    | 12    | 10    | 7     | 12    | 12    | 11    | 19    | 20    | 20    |
|     | G8  | 16   | 51    | 51    | 51    | 54    | 47    | 48    | 46    | 46    | 45    |
|     | H1  | 28   | 60    | 59    | 58    | 58    | 55    | 58    | 62    | 61    | 66    |
|     | H2  | 28   | 58    | 54    | 54    | 50    | 51    | 52    | 49    | 49    | 49    |
|     | H3  | 58   | 114   | 109   | 106   | 106   | 98    | 97    | 91    | 80    | 85    |
|     | H4  | 14   | 32    | 28    | 26    | 26    | 20    | 25    | 34    | 37    | 36    |
|     | H5  | 6    | 16    | 16    | 14    | 14    | 15    | 17    | 21    | 21    | 23    |
|     | H6  | 39   | 71    | 68    | 68    | 67    | 60    | 61    | 65    | 66    | 64    |
|     | H7  |      | 4     | 5     | 8     | 10    | 10    | 10    | 16    | 14    | 15    |
|     | H8  | 9    | 27    | 25    | 23    | 22    | 17    | 18    | 24    | 23    | 25    |
|     | H9  | 11   | 18    | 17    | 17    | 19    | 15    | 17    | 22    | 20    | 23    |
| I   | H11 | 15   | 15    | 14    | 14    | 21    | 18    | 18    | 20    | 19    | 19    |
|     | H10 | 18   | 45    | 44    | 43    | 45    | 44    | 44    | 44    | 41    | 44    |
|     | I1  | 27   | 67    | 65    | 65    | 68    | 68    | 66    | 67    | 67    | 70    |
|     | I2  | 52   | 87    | 86    | 87    | 78    | 73    | 72    | 68    | 67    | 64    |
|     | I3  | 53   | 128   | 123   | 117   | 116   | 117   | 115   | 109   | 106   | 107   |
|     | I4  | 9    | 25    | 26    | 27    | 29    | 32    | 36    | 37    | 34    | 34    |
|     | I5  | 20   | 77    | 74    | 72    | 69    | 70    | 71    | 62    | 65    | 65    |
|     | I6  | 12   | 54    | 53    | 53    | 55    | 53    | 53    | 56    | 59    | 60    |
|     | I7  | 17   | 52    | 50    | 55    | 52    | 57    | 57    | 59    | 55    | 58    |
|     | J1  | 80   | 79    | 76    | 74    | 69    | 72    | 71    | 65    | 64    | 65    |
|     | J2  | 97   | 92    | 90    | 89    | 92    | 90    | 90    | 93    | 77    | 79    |
| K   | J3  | 16   | 16    | 14    | 13    | 18    | 19    | 18    | 20    | 19    | 20    |
|     | J4  | 10   | 10    | 10    | 10    | 12    | 13    | 13    | 14    | 15    | 14    |
|     | J5  | 16   | 16    | 15    | 17    | 19    | 19    | 20    | 19    | 18    | 19    |
|     | K1  | 13   | 20    | 22    | 21    | 22    | 22    | 23    | 34    | 41    | 46    |
|     | K2  | 46   | 102   | 100   | 99    | 96    | 89    | 91    | 92    | 94    | 104   |
|     | K3  | 53   | 110   | 104   | 99    | 96    | 87    | 81    | 86    | 88    | 85    |
|     | K4  | 10   | 17    | 18    | 17    | 21    | 21    | 24    | 28    | 30    | 31    |
|     | K5  | 8    | 18    | 18    | 19    | 17    | 16    | 17    | 24    | 24    | 27    |
|     | L1  | 16   | 46    | 44    | 44    | 49    | 44    | 47    | 56    | 59    | 60    |
|     | L2  | 27   | 58    | 59    | 59    | 59    | 62    | 62    | 63    | 60    | 57    |
|     | L3  | 14   | 33    | 32    | 32    | 33    | 33    | 33    | 33    | 34    | 33    |
| L   | L4  | 5    | 16    | 17    | 17    | 19    | 17    | 16    | 20    | 17    | 18    |
|     | M1  | 25   | 62    | 63    | 59    | 60    | 49    | 49    | 46    | 43    | 45    |
|     | M2  | 39   | 92    | 88    | 89    | 87    | 83    | 84    | 88    | 80    | 81    |
|     | M3  | 32   | 83    | 82    | 74    | 73    | 69    | 68    | 77    | 70    | 76    |
|     | M4  | 10   | 30    | 27    | 17    | 24    | 19    | 18    | 25    | 21    | 23    |
|     | M5  | 8    | 17    | 19    | 25    | 18    | 18    | 18    | 25    | 20    | 21    |
|     | M6  | 12   | 29    | 28    | 29    | 30    | 30    | 30    | 27    | 28    | 28    |
|     | M7  | 21   | 48    | 48    | 47    | 45    | 45    | 45    | 42    | 42    | 41    |
|     | M8  | 22   | 57    | 56    | 54    | 45    | 42    | 42    | 41    | 42    | 39    |
|     | M9  | 4    | 10    | 8     | 8     | 8     | 7     | 8     | 11    | 11    | 13    |
|     | M   | E2   | 6040  | 6040  | 6040  | 6040  | 6340  | 6340  | 6340  | 6340  | 5066  |
| E3  |     | 5768 | 5768  | 5768  | 6114  | 6314  | 6314  | 6314  | 6314  | 8353  | 8353  |
| E4  |     | 3918 | 3918  | 3400  | 6114  | 6514  | 6514  | 6514  | 6514  | 4434  | 4434  |
| E5  |     | 4079 | 4138  | 4138  | 4138  | 4138  | 4138  | 4138  | 4138  | 4240  | 4400  |
| E6  |     | 1616 | 1616  | 1616  | 1616  | 2416  | 2416  | 2416  | 2416  | 2803  | 2803  |
| E7  |     | 1526 | 1526  | 1526  | 1526  | 1526  | 2154  | 2154  | 2154  | 1852  | 1852  |
| E8  |     | 1451 | 1451  | 1451  | 1451  | 1451  | 1451  | 1451  | 1451  | 1564  | 1564  |
| E9  |     | 1129 | 1189  | 1156  | 1189  | 1489  | 1489  | 1489  | 1489  | 1465  | 1413  |
| E10 |     | 1379 | 1542  | 1542  | 1542  | 1542  | 1638  | 1638  | 1638  | 1538  | 1630  |
| E11 |     | 1285 | 1285  | 1265  | 1265  | 1565  | 1565  | 1565  | 1565  | 1177  | 1177  |
| F1  |     | 3250 | 3460  | 3820  | 3930  | 3930  | 3930  | 3930  | 3930  | 4260  | 4260  |
| G   | F2  | 2170 | 2210  | 3430  | 3430  | 3430  | 4222  | 4222  | 4222  | 4182  | 4182  |
|     | F3  | 4386 | 4734  | 4518  | 4518  | 4518  | 5546  | 5546  | 4 674 | 3777  | 3777  |
|     | F4  | 0    | 0     | 0     | 0     | 0     | 1200  | 1200  | 1 000 | 1450  | 1380  |
|     | F5  | 5080 | 8870  | 8870  | 8 870 | 8 870 | 8 870 | 8 870 | 8 870 | 5076  | 5076  |
|     | F6  | 1044 | 1231  | 1731  | 1 714 | 1 714 | 1 714 | 1 714 | 1 714 | 1 120 | 1380  |
|     | G1  | 4790 | 4840  | 4840  | 5640  | 5640  | 4644  | 4644  | 4 176 | 4694  | 4834  |
|     | G2  | 4659 | 4319  | 4319  | 4379  | 4379  | 5710  | 5710  | 7 745 | 7745  | 7745  |
|     | G3  | 4853 | 3301  | 4453  | 5311  | 5311  | 6151  | 6151  | 5 688 | 5688  | 5688  |
|     | G4  | 3412 | 4486  | 4486  | 5200  | 5200  | 5782  | 5782  | 3 130 | 5769  | 5769  |
|     | G5  | 1740 | 3269  | 3269  | 3269  | 3269  | 3650  | 3650  | 4 481 | 4481  | 4481  |
|     | G6  | 2525 | 2746  | 2746  | 2746  | 2746  | 2706  | 2706  | 2 707 | 2707  | 2707  |
|     | G7  | 349  | 394   | 364   | 1039  | 1039  | 1344  | 1344  | 1 033 | 1226  | 1226  |
| H   | G8  | 1674 | 1714  | 1714  | 1714  | 1714  | 1364  | 1364  | 3 260 | 3260  | 3260  |
|     | H1  | 6432 | 3472  | 3792  | 3 792 | 3 792 | 3 792 | 3 792 | 4 970 | 5120  | 5120  |
|     | H2  | 4212 | 2733  | 3033  | 3 033 | 3 033 | 3 033 | 3 033 | 5 532 | 6234  | 6234  |
|     | H3  | 5360 | 4200  | 4700  | 4700  | 4700  | 4700  | 4700  | 4 564 | 4564  | 4564  |
|     | H4  | 2600 | 2345  | 3137  | 3137  | 3137  | 3637  | 3637  | 5 135 | 4070  | 4070  |
|     | H5  | 3776 | 4681  | 4681  | 4700  | 4700  | 4700  | 4700  | 4 510 | 4510  | 4510  |
|     | H6  | 3816 | 4459  | 5403  | 5403  | 5403  | 5484  | 5484  | 3 400 | 3400  | 3400  |
|     | H7  | 0    | 0     | 560   | 570   | 570   | 1247  | 1247  | 1 742 | 1742  | 1742  |
|     | H8  | 1030 | 1039  | 1039  | 1039  | 1039  | 1039  | 1039  | 1039  | 1758  | 1758  |
|     | H9  | 1335 | 2399  | 2399  | 2399  | 2399  | 2399  | 2399  | 1 750 | 1013  | 1013  |
|     | H10 | 2670 | 2454  | 2454  | 2454  | 2454  | 2454  | 2454  | 2 730 | 3178  | 3178  |
|     | H11 | 1448 | 320   | 320   | 1775  | 1775  | 2036  | 2036  | 1 443 | 1674  | 1674  |
| I   | I1  | 7104 | 7197  | 9573  | 9573  | 9573  | 9433  | 9433  | 9433  | 9774  | 9774  |
|     | I2  | 5320 | 8270  | 8270  | 8270  | 8270  | 9080  | 9080  | 9080  | 9080  | 9080  |
|     | I3  | 5394 | 5770  | 5770  | 5770  | 7410  | 5784  | 5784  | 5 632 | 5632  | 7088  |
|     | I4  | 1896 | 3842  | 3842  | 3842  | 4842  | 4842  | 4842  | 4 262 | 4762  | 4762  |
|     | I5  | 4352 | 4160  | 4160  | 4424  | 4424  | 4424  | 4424  | 4424  | 4424  | 4424  |
|     | I6  | 3208 | 3270  | 4740  | 4740  | 4740  | 3874  | 3874  | 5 110 | 5110  | 5110  |
|     | I7  | 3616 | 3016  | 3016  | 3016  | 3216  | 3170  | 3170  | 3 450 | 3890  | 3890  |
|     | J1  | 3728 | 3967  | 3967  | 3967  | 3967  | 3967  | 3967  | 3 351 | 3351  | 3351  |
|     | J2  | 6137 | 6585  | 6585  | 6585  | 6585  | 6832  | 6832  | 3 351 | 6832  | 6832  |
|     | J3  | 540  | 1425  | 1440  | 1440  | 1440  | 1574  | 1574  | 3 406 | 3406  | 3406  |
|     | J4  | 0    | 0     | 840   | 840   | 840   | 1326  | 1326  | 1686  | 1686  | 1686  |
|     | J5  | 698  | 625   | 855   | 798   | 798   | 824   | 824   | 824   | 1142  | 1142  |
| K   | K1  | 2780 | 2224  | 2224  | 2224  | 2224  | 2224  | 2224  | 2 428 | 5392  | 7232  |
|     | K2  | 4384 | 5205  | 5885  | 6413  | 6413  | 6413  | 6413  | 6413  | 6482  | 6315  |
|     | K3  | 6597 | 5466  | 5466  | 2224  | 2224  | 2224  | 2224  | 2 224 | 3 632 | 5902  |
|     | K4  | 1550 | 1766  | 2455  | 2090  | 2090  | 2090  | 2090  | 2090  | 2 500 | 3236  |
|     | K5  | 1492 | 1642  | 1642  | 1642  | 1642  | 1642  | 1642  | 1642  | 2 416 | 2416  |
|     | L1  | 7030 | 7366  | 7366  | 7366  | 7366  | 7366  | 7366  | 8 772 | 8772  | 8772  |
|     | L2  | 5064 | 5510  | 5510  | 5510  | 5510  | 5510  | 5510  | 4 752 | 4752  | 4752  |
|     | L3  | 3880 | 3602  | 3602  | 3602  | 3602  | 3602  | 3602  | 2 884 | 5004  | 5772  |
|     | L4  | 310  | 1361  | 1361  | 1361  | 1361  | 1361  | 1361  | 1 070 | 1070  | 2005  |
|     | M1  | 6396 | 9950  | 10099 | 10099 | 10099 | 9576  | 9576  | 9 576 | 9576  | 9576  |
|     | M2  | 6560 | 10573 | 10573 | 10573 | 10573 | 10590 | 10590 | 10590 | 10590 | 10590 |
|     | M3  | 7260 | 3216  | 3423  | 10142 | 18362 | 11965 | 11965 | 11965 | 11965 | 11965 |
|     | M4  | 3360 | 2974  | 2974  | 3394  | 3394  | 3394  | 3394  | 3394  | 3434  | 3554  |
| M5  | 8   | 17   | 19    | 25    | 18    | 18    | 18    | 25    | 20    | 21    |       |
| M6  | 12  | 29   | 28    | 29    | 30    | 30    | 30    | 27    | 28    | 28    |       |
| M7  | 21  | 48   | 48    | 47    | 45    | 45    | 45    | 42    | 42    | 41    |       |
| M8  | 22  | 57   | 56    | 54    | 45    | 42    | 42    | 41    | 42    | 39    |       |
| M9  | 4   | 10   | 8     | 8     | 8     | 7     | 8     | 11    | 11    | 13    |       |



This index serves as a fundamental variable for institutional evaluation, highlighting the ability of HEIs to guide students to program completion.

The **Performance Index (P)** integrates multiple variables that represent institutional efficiency, combining academic outcomes with staff–student coverage. The components include the student success rate, the pedagogical coverage rate (faculty-to-student ratio), and the administrative coverage rate (administrative staff-to-student ratio). Its weighted formula is given by:

$$P = (SR \times 0.5) + (PCR \times 0.25) + (ACR \times 0.25)$$

Where SR is the success rate, PCR is the pedagogical coverage rate, and ACR is the administrative coverage rate.

Finally, the **Growth Index (G)** measures institutional capacity to scale and accommodate increasing student populations, expressed as:

$$G = (\text{New Enrollments} / \text{Total Enrollments}) \times (\text{Total Enrollments} / \text{Occupancy Capacity}) \times 100$$

This variable highlights institutional scalability and resource management efficiency. Together, these three indexes encapsulate the critical dimensions of success, operational performance, and growth, making them highly relevant inputs for machine learning models.

### **Weighting and Aggregation Method:**

The development of composite indexes requires explicit weighting and aggregation methods to ensure that the importance of each variable is accurately represented. Weighting involves assigning relative importance to indicators, while aggregation specifies how weighted values are combined.

For this study, the **Performance Index** provides a clear example of weighted aggregation. The success rate receives a weight of 0.5, emphasizing its centrality to institutional outcomes, while pedagogical and administrative coverage rates are each weighted at 0.25. The choice of weights follows a **normative approach**, prioritizing academic outcomes while still recognizing the operational role of staffing. In alternative contexts, weighting could also be determined using **Principal Component Analysis (PCA)**, which derives weights from statistical variance, or **entropy-based methods**, which assign higher weights to variables with greater informational content (Nardo et al., 2008).

The aggregation method adopted is the **weighted sum model**, one of the most transparent and widely used approaches in composite index construction:

$$CI = \sum_{i=1}^n w_i \cdot x_i$$

Where  $w_i$  represents the weight of indicator  $i$ , and  $x_i$  its normalized value. This linear method ensures interpretability, as each variable's contribution to the composite score is proportional to its weight. In contexts where interactions among variables are nonlinear, alternative aggregation techniques such as the **geometric mean** may be applied:

$$CI = \prod_{i=1}^n x_i^{w_i}$$

This method penalizes unbalanced performance across indicators, making it useful when institutional balance across dimensions is as important as overall strength.

### Normalization of Composite Indexes:

Normalization is a crucial step to ensure comparability of variables expressed in different units or scales. Without normalization, variables with larger numerical ranges such as total enrollments would dominate the composite index, overshadowing smaller-scale ratios like staff coverage rates. Normalization ensures methodological rigor and comparability across institutions and time periods.

Two primary normalization techniques are applied in composite index construction:

#### 1. Min–Max Normalization:

This method scales variables to a fixed range, typically  $[0, 1]$ , ensuring that all indicators contribute proportionally to the index. The formula is:

$$x_i^{norm} = \frac{x_i - \min(x)}{\max(x) - \min(x)}$$

Where  $x_i^{norm}$  is the normalized value of indicator  $x_i$ , and  $\min(x), \max(x)$  are the minimum and maximum values observed in the dataset.

#### 2. Z-Score Standardization:

This technique centers the variable around its mean and scales it by its standard deviation, making it suitable for normally distributed variables:

$$x_i^{norm} = \frac{x_i - \mu}{\sigma}$$

Where  $\mu$  is the mean of the indicator and  $\sigma$  is its standard deviation.

Normalization guarantees consistency across Moroccan HEIs within the 10-year dataset used in this study. It also enhances the predictive performance of Support Vector Machine (SVM) models by ensuring that all input features (S, P, G) contribute equitably to classification and regression tasks. Without this step, disparities in scales could bias the model, leading to inaccurate decision boundaries and weaker institutional classifications.

| SR | Year | 1      | 2      | 3      | 4      | 5      | 6      | 7      | 8      | 9      | 10     |
|----|------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| A  | A1   | 0,079  | 0,1508 | 0,195  | 0,2379 | 0,209  | 0,2094 | 0,1933 | 0,1737 | 0,1778 | 0,1922 |
|    | A2   | 0,2278 | 0,128  | 0,1698 | 0,1319 | 0,1578 | 0,1596 | 0,1376 | 0,129  | 0,156  | 0,1567 |
|    | A3   | 0      | 0,063  | 0,0941 | 0,1114 | 0,0745 | 0,0968 | 0,0949 | 0,0905 | 0,1003 |        |
|    | A4   | 0,1652 | 0,1421 | 0,1544 | 0,18   | 0,1996 | 0,1717 | 0,1395 | 0,1565 | 0,1575 | 0,1347 |
|    | A5   | 0,1128 | 0,0903 | 0,0808 | 0,0941 | 0,0967 | 0,136  | 0,1364 | 0,1144 | 0,1237 | 0,1336 |
| B  | B1   | 0,1127 | 0,0795 | 0,1275 | 0,1421 | 0,1363 | 0,1462 | 0,1114 | 0,1233 | 0,1488 | 0,1244 |
|    | B2   | 0,1062 | 0,0916 | 0,0999 | 0,1219 | 0,1568 | 0,1869 | 0,1662 | 0,1727 | 0,1579 | 0,1233 |
|    | B3   | 0,0825 | 0,1201 | 0,098  | 0,0967 | 0,1176 | 0,1182 | 0,161  | 0,1366 | 0,144  | 0,1083 |
|    | B4   | 0,4004 | 0,4184 | 0,4503 | 0,4735 | 0,4802 | 0,5064 | 0,4815 | 0,2376 | 0,9657 | 0,544  |
|    | B5   | 0,1047 | 0,1247 | 0,1332 | 0,1263 | 0,1949 | 0,1492 | 0,1549 | 0,1662 | 0,17   | 0,1667 |
|    | B6   | 0,1438 | 0,1221 | 0,131  | 0,1072 | 0,1244 | 0,1656 | 0,1619 | 0,1388 | 0,146  | 0,1174 |
|    | B7   | 0,0915 | 0,1254 | 0,1321 | 0,1188 | 0,107  | 0,1009 | 0,1107 | 0,0606 | 0,0992 | 0,1458 |
|    | B8   | 0,115  | 0,1313 | 0,1082 | 0,1587 | 0,155  | 0,1942 | 0,2566 | 0,173  | 0,2117 | 0,2016 |
|    | B9   | 0,4618 | 0,3456 | 0,2752 | 0,2747 | 0,1953 | 0,3562 | 0,3038 | 0,292  | 0,4384 | 0,376  |
| C  | C1   | 0,1173 | 0,0969 | 0,1148 | 0,1277 | 0,1348 | 0,0946 | 0,1604 | 0,1715 | 0,1826 | 0,1594 |
|    | C2   | 0,1195 | 0,1444 | 0,1537 | 0,1461 | 0,1454 | 0,1434 | 0,1344 | 0,1345 | 0,1668 | 0,1331 |
|    | C3   | 0,141  | 0,0892 | 0,1442 | 0,1716 | 0,19   | 0,2113 | 0,1948 | 0,1946 | 0,1744 | 0,1739 |
|    | C4   | 0,0501 | 0,0553 | 0,1152 | 0,0273 | 0,0767 | 0,1169 | 0,1291 | 0,0642 | 0,1798 | 0,1081 |
|    | C5   | 0,1569 | 0,0923 | 0,1315 | 0,1608 | 0,1935 | 0,1875 | 0,1965 | 0,1562 | 0,1621 | 0,1562 |
|    | C6   | 0,4348 | 0,4339 | 0,4858 | 0,4667 | 0,581  | 0,3928 | 0,3382 | 0,4711 | 0,4674 | 0,4613 |
|    | C7   | 0,1084 | 0,0656 | 0,0973 | 0,1079 | 0,1089 | 0,1009 | 0,0982 | 0,0958 | 0,1169 | 0,1246 |
|    | C8   | 0,1432 | 0,1424 | 0,1914 | 0,1748 | 0,1861 | 0,173  | 0,2128 | 0,2586 | 0,2515 | 0,2359 |
|    | C9   | 0,1642 | 0,1601 | 0,1657 | 0,1401 | 0,1629 | 0,1883 | 0,1833 | 0,1561 | 0,1878 | 0,1424 |
|    | C10  | 0,1736 | 0,1372 | 0,1475 | 0,1502 | 0,1756 | 0,1802 | 0,151  | 0,1291 | 0,1395 | 0,1272 |
|    | C11  | 0,1004 | 0,1202 | 0,1445 | 0,1241 | 0,1341 | 0,1202 | 0,1183 | 0,1421 | 0,123  | 0,1516 |
|    | C12  | 0,1657 | 0,1908 | 0,189  | 0,0848 | 0,1987 | 0,2011 | 0,1788 | 0,2479 | 0,2275 | 0,2069 |
| D  | C13  | 0      | 0      | 0      | 0,1431 | 0,1618 | 0,1665 | 0,1806 | 0,1645 | 0,155  | 0,1451 |
|    | C14  | 0,1681 | 0,155  | 0,192  | 0,2351 | 0,218  | 0,2091 | 0,2091 | 0,2432 | 0,2167 | 0,2085 |
|    | C15  | 0,6141 | 0,4348 | 0,4543 | 0,2921 | 0,4264 | 0,407  | 0,4163 | 0,2893 | 0,3188 | 0,3863 |
|    | D1   | 0,0806 | 0,0593 | 0,0796 | 0,0774 | 0,0766 | 0,0802 | 0,0863 | 0,0845 | 0,0937 | 0,0823 |
|    | D2   | 0,0776 | 0,1422 | 0,1479 | 0,1136 | 0,1258 | 0,166  | 0,1469 | 0,1277 | 0,1223 | 0,1124 |
| E  | D3   | 0,1154 | 0,1174 | 0,1069 | 0,0644 | 0,1184 | 0,1388 | 0,1551 | 0,1408 | 0,1421 | 0,1292 |
|    | D4   | 0,0764 | 0,0977 | 0,1159 | 0,1308 | 0,1778 | 0,1586 | 0,1356 | 0,1873 | 0,1696 | 0,1356 |
|    | D5   | 0,0919 | 0,1138 | 0,0927 | 0,1006 | 0,0902 | 0,1233 | 0,1511 | 0,1424 | 0,1246 | 0,1057 |
|    | D6   | 0,1736 | 0,2371 | 0,2373 | 0,2141 | 0,2215 | 0,2087 | 0,1976 | 0,2013 | 0,2387 | 0,2151 |
|    | D7   | 0,0268 | 0,0384 | 0,1076 | 0,1064 | 0,1399 | 0,0975 | 0,1128 | 0,0894 | 0,1454 | 0,1337 |
|    | D8   | 0,1725 | 0,1842 | 0,2062 | 0,1681 | 0,194  | 0,1952 | 0,2143 | 0,1799 | 0,1764 | 0,1665 |
|    | D9   | 0,1627 | 0,1323 | 0,1294 | 0,1962 | 0,1922 | 0,1623 | 0,2349 | 0,1655 | 0,1744 | 0,16   |
|    | D10  | 0,3942 | 0,3977 | 0,4314 | 0,4576 | 0,6024 | 0,4669 | 0,4149 | 0,4454 | 0,9425 | 0,4432 |
|    | E1   | 0,0713 | 0,0966 | 0,0741 | 0,1256 | 0,1275 | 0,1223 | 0,1138 | 0,1189 | 0,1313 | 0,1643 |



|   |     |        |        |        |        |        |        |        |        |        |        |
|---|-----|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| F | E2  | 0,0709 | 0,1034 | 0,1143 | 0,1238 | 0,1512 | 0,1481 | 0,1563 | 0,1436 | 0,1456 | 0,1348 |
|   | E3  | 0,0851 | 0,0959 | 0,0961 | 0,1123 | 0,1354 | 0,1422 | 0,1387 | 0,1262 | 0,1329 | 0,1539 |
|   | E4  | 0,0817 | 0,0891 | 0,0932 | 0,11   | 0,1096 | 0,1132 | 0,1015 | 0,105  | 0,0856 | 0,076  |
|   | E5  | 0,1546 | 0,1806 | 0,2749 | 0,25   | 0,2283 | 0,2546 | 0,2252 | 0,1942 | 0,2031 | 0,2192 |
|   | E6  | 0,0596 | 0,0729 | 0,0732 | 0,0714 | 0,0951 | 0,1027 | 0,1012 | 0,0938 | 0,0936 | 0,1115 |
|   | E7  | 0,1573 | 0,1324 | 0,1247 | 0,1069 | 0,1356 | 0,1414 | 0,1465 | 0,1345 | 0,1313 | 0,1337 |
|   | E8  | 0,1983 | 0,1519 | 0,2018 | 0,1851 | 0,1931 | 0,2171 | 0,1898 | 0,1655 | 0,158  | 0,1539 |
|   | E9  | 0,1228 | 0,1271 | 0,1654 | 0,1966 | 0,207  | 0,1975 | 0,1872 | 0,146  | 0,1457 | 0,1533 |
|   | E10 | 0,4545 | 0,5052 | 0,4156 | 0,4622 | 0,7529 | 0,323  | 0,4718 | 0,1484 | 0,509  | 0,4617 |
|   | E11 | 0,3953 | 0,3169 | 0,5122 | 0,6065 | 0,4243 | 0,4354 | 0,5965 | 0,5753 | 0,5074 | 0,5499 |
|   | F1  | 0,118  | 0,1291 | 0,1235 | 0,1291 | 0,117  | 0,1272 | 0,1366 | 0,1458 | 0,1221 | 0,1237 |
| G | F2  | 0,0834 | 0,0598 | 0,129  | 0,1467 | 0,1608 | 0,1484 | 0,1835 | 0,3937 | 0,2346 | 0,2482 |
|   | F3  | 0,1234 | 0,1298 | 0,1747 | 0,2498 | 0,2409 | 0,1651 | 0,1748 | 0,1753 | 0,21   | 0,157  |
|   | F4  | 0      | 0,2432 | 0,4248 | 0,4429 | 0,3214 | 0,4282 | 0,4442 | 0,4244 | 0,4592 | 0,2847 |
|   | F5  | 0,0943 | 0,1246 | 0,1514 | 0,1062 | 0,136  | 0,1796 | 0,1372 | 0,1295 | 0,0975 | 0,0947 |
|   | F6  | 0,0862 | 0,1637 | 0,1587 | 0,1526 | 0,2024 | 0,1923 | 0,1571 | 0,149  | 0,1382 | 0,1558 |
|   | G1  | 0,0908 | 0,0919 | 0,0715 | 0,072  | 0,0795 | 0,0667 | 0,0557 | 0,0651 | 0,0806 | 0,0647 |
|   | G2  | 0,1205 | 0,1198 | 0,0927 | 0,101  | 0,1168 | 0,0988 | 0,1032 | 0,113  | 0,1204 | 0,0972 |
|   | G3  | 0,1287 | 0,1193 | 0,1251 | 0,13   | 0,1675 | 0,1814 | 0,1757 | 0,158  | 0,1706 | 0,1673 |
|   | G4  | 0,0665 | 0,0642 | 0,0698 | 0,0606 | 0,0654 | 0,0552 | 0,0439 | 0,0597 | 0,0669 | 0,0559 |
|   | G5  | 0      | 0      | 0      | 0,0697 | 0,105  | 0,0728 | 0,1016 | 0,0734 | 0,1209 | 0,1079 |
|   | G6  | 0,1731 | 0,1857 | 0,1871 | 0,1876 | 0,1874 | 0,2057 | 0,1711 | 0,1626 | 0,1687 | 0,1105 |
| H | G7  | 0,1473 | 0,1777 | 0,1732 | 0,2656 | 0,2083 | 0,1969 | 0,1418 | 0,1361 | 0,152  | 0,139  |
|   | G8  | 0,4629 | 0,4047 | 0,4114 | 0,4156 | 0,4357 | 0,3654 | 0,2758 | 0,4927 | 0,489  | 0,4594 |
|   | H1  | 0,0919 | 0,1196 | 0,1885 | 0,1647 | 0,1208 | 0,1199 | 0,116  | 0,1046 | 0,1215 | 0,1104 |
|   | H2  | 0,1429 | 0,1469 | 0,1218 | 0,2902 | 0,086  | 0,1132 | 0,1078 | 0,0977 | 0,1171 | 0,1067 |
|   | H3  | 0,1979 | 0,1579 | 0,1562 | 0,443  | 0,1386 | 0,1605 | 0,1767 | 0,1572 | 0,1597 | 0,1299 |
|   | H4  | 0,1287 | 0,122  | 0,1905 | 0,185  | 0,1475 | 0,1711 | 0,1422 | 0,138  | 0,1972 | 0,1695 |
|   | H5  | 0,0886 | 0,1094 | 0,1228 | 0,1388 | 0,0974 | 0,0905 | 0,1086 | 0,1061 | 0,1254 | 0,0947 |
|   | H6  | 0,2343 | 0,2363 | 0,3598 | 0,232  | 0,1861 | 0,1858 | 0,1642 | 0,1752 | 0,2002 | 0,1966 |
|   | H7  | 0      | 0      | 0,2324 | 0,2673 | 0,197  | 0,2058 | 0,2455 | 2,2692 | 0,2322 | 0,2319 |
|   | H8  | 0,1595 | 0,1958 | 0,1547 | 0,1622 | 0,1211 | 0,1772 | 0,219  | 0,1746 | 0,205  | 0,1371 |
|   | H9  | 0,1093 | 0,1483 | 0,1593 | 0,2238 | 0,1636 | 0,1982 | 0,2007 | 0,153  | 0,1381 | 0,153  |
| I | H10 | 0,1564 | 0,1357 | 0,1564 | 0,1427 | 0,1608 | 0,1619 | 0,1609 | 0,1205 | 0,1601 | 0,1679 |
|   | H11 | 0      | 0      | 0,1328 | 0,2557 | 0,1871 | 0,1932 | 0,176  | 0,2166 | 0,136  | 0,1472 |
|   | I1  | 0,0724 | 0,0596 | 0,0869 | 0,105  | 0,1035 | 0,1193 | 0,1166 | 0,1316 | 0,1311 | 0,1423 |
|   | I2  | 0,1176 | 0,0751 | 0,0961 | 0,0954 | 0,1017 | 0,135  | 0,1608 | 0,138  | 0,1228 | 0,1135 |
|   | I3  | 0,0686 | 0,1229 | 0,0639 | 0,076  | 0,0935 | 0,1061 | 0,134  | 0,1456 | 0,1519 | 0,1099 |
|   | I4  | 0,1006 | 0,124  | 0,1038 | 0,102  | 0,064  | 0,0744 | 0,0714 | 0,0819 | 0,0804 | 0,085  |
|   | I5  | 0,0868 | 0,1079 | 0,1353 | 0,1502 | 0,1139 | 0,14   | 0,155  | 0,1995 | 0,2133 | 0,1682 |
|   | I6  | 0,1575 | 0,1318 | 0,1178 | 0,1585 | 0,1381 | 0,1732 | 0,161  | 0,0969 | 0,1335 | 0,0816 |
|   | I7  | 0,4442 | 0,4763 | 0,0957 | 0,5358 | 0,4875 | 0,5168 | 0,4559 | 0,4507 | 0,1033 | 0,4566 |
| J | J1  | 0,1607 | 0,106  | 0,1071 | 0,1192 | 0,1182 | 0,1123 | 0,139  | 0,1329 | 0,1579 | 0,1069 |
|   | J2  | 0,0804 | 0,1197 | 0,1195 | 0,1441 | 0,132  | 0,1103 | 0,1617 | 0,1276 | 0,1373 | 0,1026 |
|   | J3  | 0,3248 | 0,2168 | 0,2876 | 0,3129 | 0,1845 | 0,2229 | 0,2203 | 0,2419 | 0,1654 | 0,117  |
|   | J4  | 0,109  | 0,0802 | 0,1275 | 0,1429 | 0,1534 | 0,1526 | 0,1286 | 0,1081 | 0,1362 | 0,1293 |
|   | J5  | 0,1307 | 0,1605 | 0,1415 | 0,1459 | 0,2378 | 0,1786 | 0,1649 | 0,1849 | 0,1463 | 0,1718 |
| K | K1  | 0,195  | 0,1247 | 0,1179 | 0,0768 | 0,0647 | 0,0677 | 0,0684 | 0,0912 | 0,0922 | 0,0757 |
|   | K2  | 0,0814 | 0,0853 | 0,0975 | 0,0918 | 0,0866 | 0,129  | 0,1254 | 0,1305 | 0,0993 | 0,0914 |
|   | K3  | 0,0948 | 0,0833 | 0,0975 | 0,1176 | 0,1258 | 0,1271 | 0,1471 | 0,2117 | 0,1449 | 0,1394 |
|   | K4  | 0,0882 | 0,1489 | 0,1462 | 0,1808 | 0,174  | 0,2069 | 0,1357 | 0,1598 | 0,1373 | 0,1276 |
|   | K5  | 0,1333 | 0,221  | 0,1695 | 0,2138 | 0,2426 | 0,2239 | 0,2322 | 0,2155 | 0,2468 | 0,2226 |
| L | L1  | 0,0523 | 0,0981 | 0,1425 | 0,045  | 0,0783 | 0,0947 | 0,1158 | 0,0977 | 0,1146 | 0,1071 |
|   | L2  | 0,1828 | 0,2344 | 0,2151 | 0,2243 | 0,2111 | 0,2117 | 0,1719 | 0,1842 | 0,2255 | 0,2209 |
|   | L3  | 0,2224 | 0,1926 | 0,1686 | 0,1611 | 0,1465 | 0,1733 | 0,1633 | 0,1612 | 0,1026 | 0,1545 |
|   | L4  | 0,4668 | 0,4524 | 0,4478 | 0,4859 | 0,5721 | 0      | 0,4725 | 0,3726 | 0,264  | 0,3256 |
| M | M1  | 0,0835 | 0,0788 | 0,0795 | 0,0803 | 0,0976 | 0,1081 | 0,1044 | 0,1181 | 0,1359 | 0,1012 |
|   | M2  | 0,1836 | 0,1308 | 0,1221 | 0,1202 | 0,1379 | 0,1345 | 0,1279 | 0,1386 | 0,1386 | 0,1222 |
|   | M3  | 0,0926 | 0,0927 | 0,099  | 0,0872 | 0,122  | 0,1599 | 0,1629 | 0,1465 | 0,13   | 0,0921 |
|   | M4  | 0,1196 | 0,1256 | 0,1745 | 0,199  | 0,1512 | 0,1396 | 0,1806 | 0,1626 | 0,1458 | 0,1618 |
|   | M5  | 0,0813 | 0,1582 | 0,2473 | 0,1993 | 0,1711 | 0,1353 | 0,103  | 0,1066 | 0,1576 | 0,1441 |
|   | M6  | 0,2163 | 0,2441 | 0,1902 | 0,1837 | 0,2078 | 0,1794 | 0,1623 | 0,1487 | 0,1828 | 0,15   |
|   | M7  | 0,1722 | 0,1246 | 0,2948 | 0,2082 | 0,2677 | 0,2191 | 0,2081 | 0,2075 | 0,202  | 0,1448 |
|   | M8  | 0,4661 | 0,4422 | 0,4761 | 0,4254 | 0,5153 | 0,4503 | 0,3848 | 0,3952 | 0,5055 | 0,0772 |
|   | M9  | 0      | 0      | 3,87   | 0,3965 | 0,3591 | 0,5077 | 0,4123 | 0,5536 | 0,6154 | 0,4268 |

Figures 10, 11 & 12: Representing calculated Success index *Per Year* for 105 institutions over 13 universities from 2012 to 2022.

|         | Year | 1      | 2      | 3      | 4      | 5      | 6      | 7      | 8      | 9      | 10     |
|---------|------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| PR<br>A | A1   | 0,6215 | 0,6341 | 0,5719 | 0,6586 | 0,6535 | 0,6662 | 0,6764 | 0,6161 | 0,5656 | 0,5724 |
|         | A2   | 0,8051 | 0,5304 | 0,531  | 0,336  | 0,4745 | 0,4005 | 0,3568 | 0,3025 | 0,3538 | 0,3573 |
|         | A3   | 2,2917 | 1,3542 | 1,0098 | 1,4084 | 0,8292 | 0,5581 | 0,5192 | 0,5592 | 0,503  | 0,4446 |
|         | A4   | 2,244  | 2,0569 | 1,8468 | 1,9768 | 1,763  | 1,5114 | 1,2061 | 1,1027 | 0,9018 | 0,3585 |
|         | A5   | 0,6519 | 0,549  | 0,4714 | 0,4524 | 0,4622 | 0,4682 | 0,4868 | 0,4903 | 0,4507 | 0,5049 |
| B       | B1   | 0,7989 | 0,7126 | 0,6893 | 0,5986 | 0,5512 | 0,5977 | 0,5012 | 0,4619 | 0,4931 | 0,484  |
|         | B2   | 1,0024 | 0,8652 | 0,6624 | 0,6544 | 0,7943 | 0,9111 | 0,8745 | 0,8583 | 0,8449 | 0,7712 |
|         | B3   | 1,5954 | 1,6835 | 1,332  | 1,4254 | 1,8203 | 1,3627 | 1,3979 | 1,3864 | 1,3279 | 1,4165 |
|         | B4   | 4,7274 | 4,5053 | 4,1082 | 3,9857 | 4,4464 | 3,596  | 3,1739 | 2,6766 | 4,3113 | 3,053  |
|         | B5   | 0,6019 | 0,5484 | 0,4734 | 0,4235 | 0,5677 | 0,3907 | 0,4055 | 0,3949 | 0,4052 | 0,4266 |
|         | B6   | 0,6434 | 0,5189 | 0,484  | 0,4162 | 0,3834 | 0,3856 | 0,3707 | 0,3424 | 0,3722 | 0,3525 |
|         | B7   | 4,38   | 3,5772 | 3,4721 | 3,2227 | 2,9603 | 2,6582 | 3,045  | 2,7043 | 2,21   | 2,6212 |
|         | B8   | 3,7674 | 2,8669 | 2,3086 | 2,1197 | 2,7004 | 2,7472 | 3,5456 | 3,0365 | 2,9146 | 3,3762 |
|         | B9   | 4,3385 | 3,4864 | 3,2101 | 2,7008 | 1,887  | 1,5054 | 1,4975 | 1,6131 | 1,5654 | 1,6955 |
| C       | C1   | 0,4336 | 0,4534 | 0,4221 | 0,3836 | 0,3771 | 0,3411 | 0,3448 | 0,326  | 0,3114 | 0,2625 |
|         | C2   | 0,8307 | 0,9667 | 0,8706 | 0,6864 | 0,6307 | 0,5734 | 0,5182 | 0,4718 | 0,4956 | 0,4276 |
|         | C3   | 1,0504 | 1,1284 | 1,1885 | 1,1006 | 1,1259 | 1,1745 | 1,1973 | 1,1644 | 1,0044 | 0,3796 |
|         | C4   | 2,0244 | 2,4509 | 2,0443 | 2,3026 | 1,9668 | 2,1711 | 2,2947 | 2,238  | 2,1959 | 1,8333 |
|         | C5   | 2,5644 | 3,0046 | 2,7195 | 2,6303 | 2,7329 | 2,679  | 2,8041 | 3,1952 | 2,7653 | 2,8035 |
|         | C6   | 3,537  | 4,2751 | 4      | 3,3837 | 3,631  | 3,0486 | 2,6691 | 2,7276 | 2,8434 | 3,0475 |
|         | C7   | 0,2822 | 0,2942 | 0,2696 | 0,2904 | 0,297  | 0,2787 | 0,2591 | 0,2302 | 0,2234 | 0,2078 |
|         | C8   | 0,3401 | 0,3605 | 0,4274 | 0,3764 | 0,3379 | 0,2863 | 0,3297 | 0,3551 | 0,401  | 0,4012 |
|         | C9   | 0,9429 | 1,0398 | 0,9531 | 0,9212 | 0,9187 | 0,8943 | 0,8837 | 0,8878 | 0,9936 | 0,8407 |
|         | C10  | 1,1752 | 1,0985 | 1,1145 | 1,0788 | 1,1001 | 1,0647 | 0,945  | 0,7202 | 0,6482 | 0,6735 |
|         | C11  | 0,8205 | 0,9207 | 0,9442 | 0,8424 | 0,9152 | 0,7922 | 0,7679 | 0,962  | 0,8555 | 1,0366 |
|         | C12  | 1,6852 | 1,9575 | 1,8667 | 1,6802 | 1,6727 | 1,7559 | 1,5218 | 1,6888 | 1,6012 | 1,5962 |
|         | C13  | 2,4764 | 2,3134 | 1,4117 | 1,7195 | 1,7415 | 1,748  | 1,6273 | 1,7826 | 1,6228 | 1,715  |
| D       | C14  | 0,5899 | 0,8077 | 0,6985 | 0,7013 | 0,8705 | 0,7934 | 0,8069 | 0,8221 | 0,887  | 0,9015 |
|         | C15  | 4,1235 | 4,6558 | 3,7301 | 3,4069 | 3,1848 | 3,1965 | 2,7024 | 2,7557 | 2,2362 | 2,4662 |
|         | D1   | 0,251  | 0,2727 | 0,2686 | 0,2366 | 0,2203 | 0,2415 | 0,2297 | 0,2257 | 0,1988 | 0,2299 |
|         | D2   | 0,415  | 0,5341 | 0,5638 | 0,5236 | 0,5544 | 0,5806 | 0,6061 | 0,6199 | 0,5131 | 0,4691 |
|         | D3   | 0,4155 | 0,5036 | 0,4948 | 0,405  | 0,6138 | 0,6308 | 0,6564 | 0,6216 | 0,5762 | 0,5623 |
|         | D4   | 0,9817 | 1,1015 | 0,9731 | 1,0094 | 1,2286 | 1,2445 | 1,145  | 1,2337 | 1,1265 | 1,0634 |
|         | D5   | 0,3012 | 0,3861 | 0,3394 | 0,3408 | 0,4314 | 0,4745 | 0,5516 | 0,532  | 0,5006 | 0,4509 |
|         | D6   | 1,5894 | 1,733  | 1,7939 | 1,6668 | 1,7226 | 1,6931 | 1,7076 | 1,7071 | 1,774  | 1,7174 |
|         | D7   | 2,9061 | 2,78   | 2,5087 | 2,7229 | 2,9591 | 2,4794 | 2,6144 | 2,3596 | 2,6105 | 2,5843 |
|         | D8   | 1,1645 | 1,5338 | 1,4974 | 1,5246 | 1,5105 | 1,3989 | 1,4076 | 1,7483 | 1,8362 | 1,402  |
| E       | D9   | 1,1726 | 1,3643 | 1,3001 | 1,316  | 1,3044 | 1,1501 | 1,2201 | 1,1168 | 1,111  | 1,5032 |
|         | D10  | 2,008  | 2,4909 | 2,5769 | 2,4221 | 2,3605 | 2,215  | 2,1849 | 2,0989 | 3,3955 | 2,318  |
|         | E1   | 0,3651 | 0,4439 | 0,3788 | 0,3557 | 0,315  | 0,2941 | 0,2681 | 0,2553 | 0,2461 | 0,2451 |
|         |      |        |        |        |        |        |        |        |        |        |        |



|   |     |        |        |        |        |        |        |        |        |        |        |
|---|-----|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| F | E2  | 0,4178 | 0,4392 | 0,3947 | 0,3717 | 0,4095 | 0,4258 | 0,4231 | 0,395  | 0,3951 | 0,3716 |
|   | E3  | 1,3237 | 1,3914 | 1,3303 | 1,0767 | 1,3728 | 1,2675 | 1,3243 | 1,1439 | 1,0641 | 1,1162 |
|   | E4  | 0,3735 | 0,5149 | 0,5612 | 0,5255 | 0,5281 | 0,5352 | 0,5598 | 0,5108 | 0,4513 | 0,3784 |
|   | E5  | 2,2937 | 2,4679 | 2,9569 | 3,054  | 2,6841 | 2,746  | 2,5844 | 2,3468 | 1,9343 | 2,0493 |
|   | E6  | 2,0906 | 2,571  | 2,3499 | 2,3399 | 2,3106 | 2,2445 | 2,1115 | 2,0729 | 1,9902 | 2,0055 |
|   | E7  | 2,5058 | 2,7868 | 2,3293 | 2,1686 | 2,1351 | 2,0315 | 1,9049 | 1,6754 | 1,6365 | 1,5809 |
|   | E8  | 1,3474 | 1,6888 | 1,6555 | 1,9831 | 2,4571 | 2,2456 | 2,282  | 2,0911 | 1,9298 | 1,9869 |
|   | E9  | 1,0269 | 1,6348 | 1,525  | 1,3483 | 1,2422 | 1,1984 | 1,1362 | 1,1385 | 1,1337 | 1,0433 |
|   | E10 | 2,3549 | 3,9582 | 3,3606 | 3,4098 | 3,5688 | 1,9171 | 2,3354 | 1,9296 | 2,1362 | 1,9966 |
|   | E11 | 1,8938 | 2,3443 | 2,5122 | 2,2994 | 2,1409 | 1,4325 | 2,2528 | 2,1236 | 1,7098 | 1,6676 |
|   | F1  | 0,4614 | 0,5447 | 0,4899 | 0,4251 | 0,3963 | 0,3988 | 0,4041 | 0,3817 | 0,317  | 0,324  |
| G | F2  | 0,2436 | 0,2179 | 0,2516 | 0,3312 | 0,3936 | 0,3652 | 0,4146 | 0,7697 | 0,4531 | 0,6021 |
|   | F3  | 1,8195 | 2,0812 | 3,0771 | 2,8872 | 2,0104 | 1,7186 | 1,6837 | 1,6497 | 1,2979 | 1,4027 |
|   | F4  | 0      | 1,473  | 1,5196 | 1,4714 | 1,8189 | 1,5337 | 1,535  | 1,7156 | 1,8362 | 1,2715 |
|   | F5  | 0,4663 | 0,563  | 0,7066 | 0,6541 | 0,718  | 0,9648 | 0,8685 | 0,7427 | 0,5868 | 0,5531 |
|   | F6  | 2,035  | 1,9087 | 2,0315 | 2,0614 | 2,2963 | 2,2364 | 2,0973 | 1,8147 | 1,808  | 1,7529 |
|   | G1  | 0,2973 | 0,3629 | 0,3112 | 0,2593 | 0,2678 | 0,2357 | 0,2031 | 0,1949 | 0,1936 | 0,1941 |
|   | G2  | 0,6286 | 0,701  | 0,4599 | 0,5587 | 0,5407 | 0,4639 | 0,4445 | 0,4572 | 0,4523 | 0,4473 |
|   | G3  | 1,266  | 1,3295 | 1,5131 | 1,362  | 1,3986 | 1,4029 | 1,4569 | 1,3827 | 1,3832 | 1,4502 |
|   | G4  | 0,2867 | 0,3717 | 0,3133 | 0,2796 | 0,2927 | 0,2331 | 0,2228 | 0,2262 | 0,2517 | 0,2142 |
|   | G5  | 1,774  | 1,6357 | 1,7713 | 1,5613 | 1,5143 | 1,2353 | 1,117  | 1,2564 | 1,2697 | 1,1796 |
|   | G6  | 1,9716 | 2,1558 | 2,2978 | 2,3714 | 2,3568 | 2,4499 | 2,3083 | 2,1841 | 2,2623 | 2,0111 |
|   | G7  | 0,7061 | 1,0409 | 1,0293 | 1,0716 | 1,4691 | 1,329  | 1,233  | 1,3639 | 1,3103 | 1,2956 |
| H | G8  | 1,6701 | 2,7616 | 3,3135 | 3,0389 | 2,8147 | 2,298  | 1,6806 | 1,5442 | 1,4419 | 1,4923 |
|   | H1  | 0,2294 | 0,273  | 0,2937 | 0,2413 | 0,2407 | 0,1945 | 0,1897 | 0,182  | 0,1831 | 0,1761 |
|   | H2  | 0,558  | 0,594  | 0,4978 | 0,5084 | 0,3691 | 0,3658 | 0,329  | 0,3393 | 0,3393 | 0,3217 |
|   | H3  | 1,1685 | 1,0652 | 1,0575 | 1,1113 | 0,9693 | 0,9462 | 1,0235 | 0,9597 | 0,9033 | 0,8041 |
|   | H4  | 0,2223 | 0,2413 | 0,2519 | 0,2223 | 0,1632 | 0,1825 | 0,1634 | 0,1589 | 0,1918 | 0,1885 |
|   | H5  | 0,405  | 0,3824 | 0,466  | 0,4558 | 0,3596 | 0,2799 | 0,2894 | 0,3248 | 0,3403 | 0,3465 |
|   | H6  | 1,5873 | 1,7189 | 1,4639 | 1,3662 | 1,3475 | 1,2371 | 1,1645 | 1,2021 | 1,1512 | 1,2103 |
|   | H7  | 0,3125 | 0,7835 | 1,5351 | 1,7483 | 1,8504 | 1,4898 | 1,4338 | 13,635 | 1,1314 | 1,039  |
|   | H8  | 1,7415 | 1,7139 | 1,5967 | 1,6408 | 1,6714 | 1,5857 | 1,5757 | 1,6301 | 1,6012 | 1,4243 |
|   | H9  | 1,0204 | 1,2308 | 1,3983 | 1,8019 | 1,5754 | 1,5309 | 1,7205 | 1,9188 | 1,7954 | 1,6778 |
|   | H10 | 1,2716 | 1,1329 | 1,1276 | 1,095  | 1,1862 | 1,0721 | 1,0114 | 0,9268 | 0,9073 | 0,8765 |
|   | H11 | 1,5051 | 2,6848 | 2,5896 | 2,7718 | 2,9872 | 2,5267 | 2,527  | 2,3907 | 2,294  | 2,0529 |
| I | I1  | 0,2678 | 0,3146 | 0,2737 | 0,2704 | 0,2726 | 0,263  | 0,2425 | 0,2442 | 0,2227 | 0,2507 |
|   | I2  | 0,4691 | 0,5111 | 0,5039 | 0,4728 | 0,4983 | 0,5319 | 0,5692 | 0,5947 | 0,5209 | 0,455  |
|   | I3  | 1,053  | 1,8173 | 1,1628 | 1,5234 | 1,6504 | 1,6641 | 1,6705 | 1,895  | 1,7181 | 1,7856 |
|   | I4  | 0,3204 | 0,4751 | 0,3767 | 0,3262 | 0,2356 | 0,1991 | 0,2008 | 0,1951 | 0,1823 | 0,1815 |
|   | I5  | 1,3519 | 1,7031 | 1,5032 | 1,4944 | 1,4219 | 1,3575 | 1,3487 | 1,3525 | 1,5058 | 1,3789 |
|   | I6  | 1,7119 | 2,4223 | 2,2372 | 2,1993 | 2,1848 | 2,2063 | 2,2082 | 2,2478 | 2,275  | 2,1281 |
|   | I7  | 1,0922 | 1,7988 | 1,5055 | 1,8472 | 1,7042 | 1,87   | 1,9265 | 1,9265 | 1,7725 | 2,1391 |
|   | J1  | 1,4177 | 1,1391 | 1,236  | 1,0765 | 0,9838 | 0,9281 | 0,9346 | 0,834  | 0,9247 | 0,8329 |
|   | J2  | 1,6539 | 1,5255 | 1,6519 | 1,4743 | 1,6244 | 1,5681 | 1,681  | 1,4862 | 1,6051 | 1,4363 |
|   | J3  | 0,7828 | 0,7721 | 0,736  | 0,6866 | 0,7141 | 0,5733 | 0,4924 | 0,4822 | 0,3127 | 0,2928 |
|   | J4  | 2,0282 | 1,5444 | 2,0245 | 2,3957 | 2,5977 | 2,4102 | 2,2579 | 2,0579 | 2,0132 | 1,8696 |
| K | J5  | 1,2964 | 1,5479 | 1,4316 | 1,5391 | 1,7146 | 1,5871 | 1,5117 | 1,32   | 1,2373 | 1,1856 |
|   | K1  | 0,4163 | 0,4    | 0,3312 | 0,2826 | 0,2428 | 0,2047 | 0,1907 | 0,2028 | 0,1741 | 0,1468 |
|   | K2  | 0,3659 | 0,4134 | 0,3439 | 0,297  | 0,2847 | 0,3559 | 0,3733 | 0,3623 | 0,309  | 0,2502 |
|   | K3  | 0,9235 | 0,8939 | 0,966  | 1,128  | 1,0331 | 1,0215 | 1,1876 | 1,2313 | 1,079  | 0,937  |
|   | K4  | 1,1879 | 1,5186 | 1,5636 | 1,4391 | 1,6166 | 1,5688 | 1,6098 | 1,6178 | 1,5937 | 1,3733 |
|   | K5  | 0,825  | 1,0463 | 0,9584 | 0,9918 | 0,972  | 0,8932 | 0,8279 | 0,8608 | 0,8646 | 0,8139 |
|   | L1  | 0,2925 | 0,359  | 0,342  | 0,2307 | 0,2235 | 0,2012 | 0,1928 | 0,1682 | 0,178  | 0,1654 |
|   | L2  | 1,2238 | 1,3868 | 1,2402 | 1,2833 | 1,3567 | 1,2517 | 1,2216 | 1,3416 | 1,4871 | 1,4897 |
|   | L3  | 1,0313 | 1,1896 | 1,3001 | 1,197  | 1,1948 | 1,0738 | 1,0625 | 1,0217 | 0,8822 | 0,9409 |
|   | L4  | 1,7586 | 2,6041 | 2,7736 | 3,4209 | 3,2512 | 2,7715 | 2,833  | 3,7921 | 2,9878 | 3,4716 |
|   | M1  | 0,1605 | 0,1772 | 0,1714 | 0,1451 | 0,1804 | 0,1879 | 0,1893 | 0,1752 | 0,1805 | 0,1542 |
| M | M2  | 0,3565 | 0,3594 | 0,3287 | 0,3052 | 0,3586 | 0,3372 | 0,3562 | 0,4423 | 0,4197 | 0,3915 |
|   | M3  | 0,5608 | 0,5844 | 0,5729 | 0,5136 | 0,6857 | 0,8235 | 0,9745 | 1,0779 | 0,9364 | 0,9732 |
|   | M4  | 0,742  | 0,86   | 0,7574 | 0,6978 | 0,7196 | 0,697  | 0,6612 | 0,7727 | 0,6489 | 0,639  |
|   | M5  | 0,561  | 0,695  | 0,7718 | 0,8795 | 0,7703 | 0,7586 | 0,6164 | 0,6834 | 0,62   | 0,6199 |
|   | M6  | 1,9105 | 2,2726 | 2,0872 | 2,0399 | 1,9566 | 1,8289 | 1,8047 | 1,5577 | 1,5564 | 1,4715 |
|   | M7  | 1,1296 | 1,3725 | 1,7643 | 1,5491 | 1,5076 | 1,3164 | 1,1652 | 1,1601 | 1,0461 | 1,0195 |
|   | M8  | 1,8103 | 2,8191 | 2,6011 | 2,4315 | 2,4011 | 2,0885 | 1,7802 | 1,811  | 2,1107 | 1,8911 |
|   | M9  | 1,4337 | 1,9231 | 1,685  | 2,0204 | 1,9875 | 2,2513 | 1,4854 | 1,4613 | 1,6115 | 1,6291 |

Figures 13, 14 & 15: Representing calculated Performance index *Per Year* for 105 institutions over 13 universities from 2012 to 2022.

|      | Year | 1      | 2      | 3      | 4      | 5      | 6      | 7      | 8      | 9      | 10     |
|------|------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| GR A | A1   | 99,887 | 61,381 | 62,018 | 49,079 | 47,149 | 52,315 | 60,741 | 64,398 | 37,98  | 33,697 |
|      | A2   | 23,88  | 49,43  | 66,138 | 93,146 | 82,506 | 97,252 | 95,98  | 92,519 | 80,796 | 86,293 |
|      | A3   | 90,226 | 46,4   | 90,4   | 96,098 | 100,98 | 101,71 | 100,98 | 68,293 | 144,88 | 30,591 |
|      | A4   | 28,738 | 42,613 | 35,067 | 18,304 | 34,247 | 34,694 | 48,98  | 36,671 | 41,877 | 45,877 |
|      | A5   | 56,908 | 47,537 | 56,525 | 66,445 | 64,514 | 70,04  | 60,053 | 46,871 | 32,091 | 43,064 |
| B    | B1   | 42,793 | 41,137 | 38,159 | 40,162 | 38,737 | 41,266 | 47,414 | 82,727 | 106,84 | 63,894 |
|      | B2   | 73,077 | 59,359 | 76,603 | 40,952 | 41,959 | 39,201 | 40,504 | 49,968 | 64,179 | 58,667 |
|      | B3   | 60,25  | 44,647 | 50,17  | 44,955 | 32,835 | 41,997 | 35,774 | 31,61  | 32,563 | 32,715 |
|      | B4   | 25,531 | 25,284 | 25,827 | 26,765 | 27,654 | 24,49  | 34,949 | 24,916 | 24,879 | 24,954 |
|      | B5   | 92,118 | 98,283 | 78,444 | 88,01  | 86,7   | 90,008 | 78,673 | 86,767 | 125,68 | 97,49  |
|      | B6   | 46,293 | 43,128 | 39,177 | 49,523 | 55,287 | 50,732 | 50,986 | 66,128 | 84,814 | 77,074 |
|      | B7   | 14,56  | 12,277 | 10,64  | 10,23  | 13,72  | 14,603 | 15,12  | 17,51  | 21,721 | 23,298 |
|      | B8   | 21,084 | 19,157 | 22,289 | 15,324 | 14,735 | 9,9933 | 10,127 | 12,408 | 13,31  | 12,042 |
|      | B9   | 6,1175 | 7,9119 | 11,827 | 9,3801 | 12,643 | 8,3571 | 12,071 | 15,5   | 13,357 | 15,786 |
| C    | C1   | 49,632 | 42,454 | 65     | 68,778 | 67,702 | 67,788 | 73,659 | 95,994 | 118,41 | 156,07 |
|      | C2   | 34,264 | 30,227 | 27,105 | 31,457 | 45,743 | 52,318 | 60,612 | 63,876 | 108,08 | 148,69 |
|      | C3   | 33,444 | 28,801 | 42,308 | 34,921 | 29,801 | 22,124 | 26,518 | 25,839 | 38,844 | 36,7   |
|      | C4   | 14,435 | 12,796 | 12,957 | 12,124 | 13,629 | 9,7561 | 11,187 | 12,183 | 11,277 | 12,79  |
|      | C5   | 20,582 | 53,244 | 18,568 | 14,844 | 15,911 | 16,444 | 15,822 | 16,178 | 19,329 | 21,261 |
|      | C6   | 24,869 | 26,81  | 29,748 | 24,029 | 26,495 | 22,626 | 28,97  | 24,862 | 10,624 | 22,112 |
|      | C7   | 61,517 | 56,868 | 47,5   | 36,812 | 46,025 | 46,784 | 61,419 | 69,115 | 59,493 | 70,164 |
|      | C8   | 34,598 | 38,732 | 37,21  | 51,522 | 66,742 | 79,739 | 48,64  | 53,201 | 41,345 | 59,805 |
|      | C9   | 35,467 | 32,947 | 41,914 | 28,039 | 29,601 | 29,467 | 37,118 | 39,75  | 32,757 | 45,018 |
|      | C10  | 47,197 | 54,815 | 52,704 | 44,855 | 44,426 | 27,522 | 31,416 | 38,035 | 72,524 | 91,106 |
|      | C11  | 106,35 | 89,995 | 76,245 | 50,419 | 53,578 | 40,822 | 54,477 | 52,342 | 56,352 | 50,655 |
|      | C12  | 18,434 | 22,906 | 18,063 | 23,6   | 22,675 | 24,887 | 30,941 | 29,838 | 23,424 | 30,599 |
| D    | C13  | 4,8638 | 11,43  | 15,759 | 7,2471 | 8,2198 | 8,0253 | 10,117 | 12,743 | 13,082 | 19,457 |
|      | C14  | 37,647 | 32,451 | 35,49  | 35,882 | 32,255 | 27,222 | 27,611 | 27,222 | 22,371 | 21,16  |
|      | C15  | 7,7914 | 9,5706 | 8,8344 | 9,6319 | 1,3497 | 5,4601 | 5,3374 | 7,7914 | 6,2384 | 3,3734 |
|      | D1   | 38,77  | 86,986 | 62,298 | 121,03 | 72,44  | 66,438 | 73,561 | 83,765 | 117,47 | 103,05 |
|      | D2   | 58,45  | 44,31  | 66,639 | 75,9   | 99,246 | 56,108 | 50,724 | 41,097 | 51,521 | 39,363 |
|      | D3   | 43,017 | 49,399 | 50,296 | 56,735 | 34,497 | 32,171 | 34,999 | 42,793 | 50,078 | 41,059 |
|      | D4   | 39,43  | 33,548 | 27,57  | 29,021 | 17,84  | 18,852 | 23,38  | 19,924 | 32,041 | 28,131 |
|      | D5   | 63,661 | 56,576 | 71,143 | 83,942 | 108,54 | 65,279 | 53,196 | 46,781 | 51,392 | 53,295 |
|      | D6   | 16,74  | 21,771 | 19,255 | 20,474 | 16,569 | 17,059 | 20,775 | 21,183 | 15,025 | 19,256 |
|      | D7   | 9,1359 | 10,84  | 10,483 | 11,14  | 9,2787 | 10,675 | 11,257 | 10,733 | 10,191 | 10,135 |
|      | D8   | 10,441 | 21,6   | 19,867 | 18,561 | 17,517 | 7,8747 | 10,711 | 11,431 | 12,778 | 16,144 |
|      | D9   | 42,276 | 27,045 | 28,409 | 30     | 26,327 | 11,663 | 13,499 | 13,548 | 17,64  | 18,127 |
|      | D10  | 44,643 | 48,512 | 27,452 | 46,575 | 45,37  | 25,703 | 32,408 | 32,509 | 33,989 | 12,265 |
| E    | E1   | 103,1  | 73,812 | 86,868 | 111,12 | 116,63 | 122,44 | 138,02 | 127,51 | 115,87 | 143,21 |



|   |     |        |        |        |        |        |        |        |        |        |        |
|---|-----|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| F | E2  | 84,023 | 85,762 | 106,29 | 106,75 | 64,193 | 64,337 | 75,519 | 73,084 | 111,35 | 124,48 |
|   | E3  | 87,985 | 76,543 | 67,996 | 70,167 | 46,876 | 46,76  | 58,447 | 58,143 | 50,64  | 49,192 |
|   | E4  | 100,87 | 82,261 | 67,588 | 44,374 | 37,888 | 37,565 | 57,614 | 55,25  | 81,509 | 115,13 |
|   | E5  | 18,24  | 24,505 | 17,98  | 24,601 | 18,221 | 23,248 | 30,836 | 18,178 | 20,841 | 23,136 |
|   | E6  | 20,05  | 16,646 | 21,844 | 19,183 | 14,652 | 16,142 | 15,687 | 15,356 | 14,556 | 14,592 |
|   | E7  | 9,4364 | 18,152 | 13,827 | 12,189 | 11,009 | 7,2888 | 13,045 | 16,574 | 13,769 | 16,307 |
|   | E8  | 8,408  | 20,813 | 10,544 | 12,612 | 7,5121 | 16,678 | 13,37  | 13,232 | 11,573 | 12,148 |
|   | E9  | 13,552 | 13,036 | 12,457 | 14,13  | 11,686 | 14,775 | 17,999 | 16,997 | 21,585 | 21,798 |
|   | E10 | 18,057 | 15,37  | 17,056 | 15,759 | 16,796 | 27,444 | 24,205 | 30,538 | 28,16  | 33,252 |
|   | E11 | 14,241 | 10,661 | 18,656 | 13,992 | 19,681 | 28,115 | 18,403 | 16,997 | 36,109 | 36,109 |
|   | F1  | 100,28 | 93,642 | 86,466 | 100,25 | 89,262 | 91,17  | 98,321 | 111,02 | 112,51 | 149,18 |
| G | F2  | 100,09 | 141,72 | 86,968 | 88,28  | 80,967 | 76,575 | 61,416 | 39,768 | 64,61  | 51,172 |
|   | F3  | 19,083 | 16,181 | 14,741 | 15,471 | 12,409 | 17,238 | 15,218 | 21,16  | 19,884 | 35,266 |
|   | F4  | 0      | 0      | 0      | 0      | 7,4167 | 15,417 | 15,417 | 28,2   | 10,69  | 44,855 |
|   | F5  | 34,764 | 17,43  | 14,171 | 20,722 | 16,629 | 11,781 | 16,505 | 18,253 | 35,973 | 19,484 |
|   | F6  | 14,559 | 13,404 | 9,1854 | 8,4597 | 8,6931 | 9,5829 | 13,811 | 16,686 | 19,058 | 17,174 |
|   | G1  | 86,493 | 77,996 | 95,992 | 111,17 | 91,915 | 124,14 | 124,14 | 167,03 | 135,14 | 138,04 |
|   | G2  | 70,251 | 44,094 | 55,743 | 68,729 | 53,926 | 47,285 | 47,285 | 41,64  | 33,376 | 41,356 |
|   | G3  | 40,49  | 65,071 | 37,929 | 36,136 | 25,156 | 23,086 | 23,086 | 56,698 | 31,558 | 56,312 |
|   | G4  | 83,001 | 54,28  | 71,734 | 80,808 | 66,365 | 78,692 | 78,692 | 158,37 | 89,028 | 99,029 |
|   | G5  | 13,103 | 8,4429 | 9,0242 | 9,3301 | 10,095 | 8,9315 | 8,9315 | 7,7661 | 7,9223 | 7,2752 |
|   | G6  | 6,3366 | 4,0058 | 7,8296 | 5,4989 | 5,8995 | 5,6171 | 5,6171 | 10,307 | 11,378 | 10,528 |
| H | G7  | 19,073 | 13,49  | 12,241 | 13,186 | 12,993 | 12,946 | 12,946 | 23,233 | 19,576 | 19,25  |
|   | G8  | 30,526 | 31,855 | 19,312 | 32,322 | 31,447 | 30,754 | 30,754 | 33,16  | 26,442 | 25,092 |
|   | H1  | 57,794 | 136    | 165,45 | 134,1  | 167,59 | 184,34 | 198,76 | 181,27 | 162,27 | 173,89 |
|   | H2  | 57,217 | 96,414 | 120,94 | 102,6  | 112,17 | 111,28 | 114,47 | 57,574 | 42,236 | 51,925 |
|   | H3  | 28,377 | 43,69  | 41,957 | 39,447 | 30,149 | 30,489 | 39,468 | 48,357 | 48,948 | 54,689 |
|   | H4  | 119    | 148,27 | 153,52 | 145,62 | 203,98 | 198,16 | 253,31 | 165,94 | 180,07 | 188,26 |
|   | H5  | 22,06  | 19,141 | 16,471 | 16,83  | 33,957 | 28,745 | 32,447 | 33,481 | 90,421 | 33,038 |
|   | H6  | 21,882 | 19,489 | 20,896 | 17,675 | 19,119 | 20,569 | 26,167 | 37,588 | 28,206 | 36,353 |
|   | H7  | 0      | 0      | 21,429 | 33,684 | 32,105 | 19,006 | 23,176 | 17,623 | 20,436 | 20,034 |
|   | H8  | 16,697 | 20,564 | 24,659 | 17,379 | 17,47  | 20,291 | 25,478 | 31,574 | 14,733 | 13,367 |
|   | H9  | 18,577 | 8,045  | 9,5873 | 6,8779 | 7,2113 | 8,7536 | 10,921 | 17,486 | 27,739 | 29,319 |
| I | H10 | 11,573 | 12,51  | 13,855 | 12,795 | 12,714 | 15,689 | 18,174 | 17,599 | 14,915 | 15,167 |
|   | H11 | 11,188 | 15     | 13,152 | 6,3099 | 7,7183 | 6,9745 | 11,837 | 15,177 | 13,202 | 17,324 |
|   | I1  | 81,109 | 68,918 | 50,277 | 65,779 | 58,822 | 71,897 | 75,543 | 72,607 | 77,041 | 81,502 |
|   | I2  | 122,03 | 68,416 | 47,799 | 49,819 | 28,851 | 23,535 | 32,445 | 30,275 | 39,989 | 36,487 |
|   | I3  | 45,755 | 34,419 | 61,369 | 36,568 | 18,839 | 23,721 | 27,835 | 24,069 | 33,327 | 21,247 |
|   | I4  | 79,325 | 43,181 | 63,665 | 79,646 | 68,773 | 69,847 | 96,985 | 89,302 | 117,32 | 104,89 |
|   | I5  | 16,59  | 17,548 | 22,5   | 21,067 | 21,587 | 20,931 | 22,649 | 20,75  | 15,461 | 27,012 |
|   | I6  | 8,1359 | 9,8165 | 6,7089 | 4,9578 | 6,5401 | 7,5116 | 11,177 | 8,1018 | 7,2798 | 9,3151 |
|   | I7  | 28,706 | 26,127 | 28,481 | 28,68  | 26,182 | 24,038 | 28,233 | 27,71  | 25,45  | 23,882 |
| J | J1  | 51,931 | 38,669 | 38,165 | 55,533 | 55,508 | 53,315 | 62,591 | 71,121 | 54,543 | 68,742 |
|   | J2  | 41,633 | 36,811 | 31,374 | 39,241 | 28,049 | 27,679 | 28,513 | 67,325 | 31,543 | 38,7   |
|   | J3  | 88,333 | 31,649 | 36,597 | 51,389 | 64,028 | 69,187 | 77,51  | 41,897 | 116,27 | 71,022 |
|   | J4  | 0      | 0      | 10,595 | 11,905 | 12,381 | 12,443 | 14,404 | 18,326 | 11,21  | 13,227 |
|   | J5  | 21,06  | 20,32  | 16,491 | 19,173 | 20,05  | 25,728 | 28,762 | 31,432 | 20,578 | 21,979 |
| K | K1  | 45,072 | 66,322 | 74,415 | 95,414 | 131,56 | 196    | 219,24 | 228,25 | 135,44 | 140,5  |
|   | K2  | 140,81 | 122,77 | 101,33 | 78,497 | 99,423 | 55,497 | 68,408 | 77,015 | 83,277 | 114,28 |
|   | K3  | 56,996 | 82,492 | 34,175 | 100,13 | 99,101 | 51,484 | 70,099 | 59,128 | 51,842 | 41,003 |
|   | K4  | 10,194 | 10,362 | 7,0061 | 8,4689 | 9,9522 | 10,383 | 12,44  | 13,301 | 10,48  | 8,6527 |
|   | K5  | 15,08  | 13,825 | 13,581 | 12,911 | 12,485 | 16,2   | 20,828 | 23,873 | 16,556 | 15,935 |
| L | L1  | 52,859 | 51,48  | 52,566 | 71,409 | 68,422 | 73,283 | 104,71 | 105,6  | 85,237 | 89,09  |
|   | L2  | 15,837 | 16,479 | 17,641 | 15,935 | 15,644 | 17,387 | 23,176 | 22,748 | 20,455 | 24,853 |
|   | L3  | 7,268  | 7,9123 | 8,3842 | 8,0788 | 7,7735 | 11,938 | 10,772 | 16,609 | 10,532 | 8,3853 |
|   | L4  | 21,209 | 15,503 | 13,74  | 11,168 | 12,711 | 13,079 | 13,152 | 14,953 | 16,822 | 4,9377 |
| M | M1  | 164,34 | 92,603 | 110,15 | 110,55 | 50,941 | 88,033 | 85,975 | 106,46 | 90,539 | 122,75 |
|   | M2  | 107,44 | 57,514 | 84,12  | 73,669 | 37,679 | 36,676 | 41,36  | 45,382 | 47,063 | 49,566 |
|   | M3  | 69,614 | 58,236 | 64,3   | 55,019 | 19,148 | 19,047 | 23,184 | 22,282 | 29,327 | 23,184 |
|   | M4  | 23,69  | 25,622 | 33,053 | 29,14  | 26,93  | 28,992 | 26,841 | 31,349 | 59,015 | 23,326 |
|   | M5  | 18,877 | 24,907 | 18,354 | 18,212 | 11,494 | 17,407 | 38,229 | 33,896 | 32,789 | 31,181 |
|   | M6  | 10,484 | 25,565 | 9,2758 | 7,8869 | 8,2341 | 9,5909 | 12,136 | 12,462 | 9,9524 | 13,717 |
|   | M7  | 12,573 | 13,417 | 13,169 | 13,086 | 12,551 | 11,519 | 17,333 | 15,515 | 7,9475 | 15,204 |
|   | M8  | 33,738 | 44,417 | 40,615 | 41,019 | 36,731 | 50,162 | 67,557 | 60,922 | 48,544 | 39,366 |
|   | M9  | 10,697 | 10,503 | 12,741 | 10,628 | 13,549 | 5,9748 | 16,307 | 24,462 | 24,283 | 32,796 |

Figures 16, 17 & 18: Representing calculated Growth index *Per Year* for 105 institutions over 13 universities from 2012 to 2022.

## Integration with SVM Classifier

In the implementation of SVM models for optimizing the strategic management of public higher education institutions in Morocco, the dataset was carefully structured to cater to the specific requirements of classification and regression tasks. The dataset comprised data from 105 institutions over a ten-year period, spanning from 2012 to 2022. To ensure robust model training and evaluation, the dataset was split into two parts: 84 institutions were used for training the models, while the remaining 21 were reserved for testing.

## Classification Model

For the classification model, the dataset utilized the three composite indexes Success Index, Performance Index, and Growth Index as input features. The target variable for this model was the classification class, which categorized institutions based on the Student Success to Institutional Performance Index into three classes (Class A, Class B, and Class C). This approach allowed the model to learn and predict the categorical labels that represent different levels of institutional performance. The goal of this model was to accurately classify institutions into one of the three predefined classes, providing a high-level overview of their performance based on the combined index criteria.

## Classification class Criteria:

The subsequent phase of the analysis involved the integration of the Success Index and the Performance Index into a unified metric, termed the "Student Success to Institutional Performance Index." This composite index was designed to capture the relationship between student outcomes and the operational effectiveness of educational institutions. The formula used for this index is as follows:

**Student success to institutional performance Index = Performance Index \* 0.5 + Success Index \* 0.5.**

This combined index provides a balanced evaluation, equally weighting both student success and institutional performance. It thus offers a comprehensive perspective on an institution's capacity to deliver quality education and maintain efficient operational support systems.

To assess this index across a broader context, data was generalized across 105 institutions over a ten-year period. For each institution, a mathematical harmonic mean was calculated based on the yearly values of the composite index. The harmonic mean was chosen for its ability to account for variations and ensure a more representative central tendency when dealing with ratios and rates.

Following the computation of these harmonic means, the results were systematically mapped onto a classification scale. This scale facilitated the categorization of institutions into three distinct classes, reflecting varying levels of performance as measured by the Student Success to Institutional Performance Index. These classes provide a clear and structured framework for comparing institutional effectiveness, thereby enabling stakeholders to identify areas of excellence and opportunities for improvement within the higher education sector.

The classification scale for the Student Success to Institutional Performance Index categorizes institutions based on their performance ratios, providing a clear framework for understanding the effectiveness of each institution. The criteria for classification are defined as follows:

- **Class A: High Success to Performance Ratio**
  - Institutions with an index greater than 1 are classified under Class A. This category represents institutions that exhibit a strong balance between student success and institutional performance, indicating both high academic achievements and efficient operations. These institutions are considered exemplary in their ability to support students while maintaining robust administrative and educational practices.
- **Class B: Moderate Success to Performance Ratio**
  - Institutions with an index ranging from 0.5 to 1 fall into Class B. This classification indicates a moderate balance between student success and institutional performance. Institutions in this category demonstrate a satisfactory level of student outcomes and operational efficiency, though there may be areas that require further development to reach the highest standards.



- **Class C: Low Success to Performance Ratio**

- Class C includes institutions with an index of less than 0.5. These institutions show a lower success to performance ratio, suggesting challenges in achieving high student success rates and/or maintaining effective operational systems. This classification highlights the need for significant improvements in both educational outcomes and institutional management.

This classification system not only facilitates a nuanced understanding of institutional performance but also serves as a valuable tool for policymakers, administrators, and stakeholders. By categorizing institutions based on their success to performance ratios, the scale helps identify institutions that excel, those that are performing adequately, and those that may benefit from targeted interventions and support to enhance their overall effectiveness.

| ▼   | 1 ▼     | 2 ▼     | 3 ▼     | 4 ▼     | 5 ▼     | 6 ▼     | 7 ▼     | 8 ▼     | 9 ▼     | 10 ▼    | M ▼       | Classification class ▼ |
|-----|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|-----------|------------------------|
| A1  | 0,3503  | 0,39244 | 0,38346 | 0,44824 | 0,43127 | 0,43781 | 0,43487 | 0,39491 | 0,37172 | 0,38228 | 0,4002831 | C                      |
| A2  | 0,51645 | 0,32923 | 0,38036 | 0,23396 | 0,31616 | 0,28003 | 0,24719 | 0,21578 | 0,25788 | 0,25728 | 0,2852631 | C                      |
| A3  | 1,14583 | 0,67708 | 0,53641 | 0,75124 | 0,4703  | 0,31631 | 0,30795 | 0,32708 | 0,29676 | 0,27245 | 0,4136068 | C                      |
| A4  | 1,20461 | 1,09951 | 1,00056 | 1,07842 | 0,98131 | 0,84154 | 0,67282 | 0,6296  | 0,52965 | 0,54662 | 0,7895308 | B                      |
| A5  | 0,38236 | 0,31967 | 0,2761  | 0,27325 | 0,27943 | 0,3021  | 0,31163 | 0,30236 | 0,28722 | 0,31928 | 0,3026276 | C                      |
| B1  | 0,45581 | 0,39604 | 0,40842 | 0,37037 | 0,34372 | 0,37195 | 0,30633 | 0,29263 | 0,32094 | 0,30418 | 0,3502567 | C                      |
| B2  | 0,55432 | 0,47841 | 0,38116 | 0,38812 | 0,47559 | 0,54901 | 0,52033 | 0,51553 | 0,50141 | 0,44724 | 0,4737009 | C                      |
| B3  | 0,83896 | 0,90183 | 0,71503 | 0,76104 | 0,96893 | 0,74044 | 0,77943 | 0,76154 | 0,73591 | 0,76242 | 0,7897322 | B                      |
| B4  | 2,56388 | 2,46189 | 2,27923 | 2,22957 | 2,46329 | 2,05121 | 1,82766 | 1,4571  | 2,63849 | 1,79852 | 2,1072931 | A                      |
| B5  | 0,35328 | 0,33655 | 0,30331 | 0,27491 | 0,38129 | 0,26996 | 0,28017 | 0,2806  | 0,28762 | 0,29664 | 0,3026232 | C                      |
| B6  | 0,39359 | 0,3205  | 0,30748 | 0,26171 | 0,25392 | 0,27559 | 0,26627 | 0,24058 | 0,25912 | 0,23493 | 0,2752642 | C                      |
| B7  | 2,23573 | 1,85126 | 1,80214 | 1,67074 | 1,53366 | 1,37954 | 1,57784 | 1,38245 | 1,15461 | 1,38349 | 1,5470682 | A                      |
| B8  | 1,94118 | 1,4991  | 1,20842 | 1,13924 | 1,42769 | 1,47068 | 1,90108 | 1,60475 | 1,56314 | 1,78889 | 1,5115273 | A                      |
| B9  | 2,40014 | 1,91598 | 1,74265 | 1,48777 | 1,04114 | 0,9308  | 0,90065 | 0,95258 | 1,00189 | 1,03575 | 1,1985996 | A                      |
| C1  | 0,27545 | 0,27518 | 0,26844 | 0,25564 | 0,25595 | 0,21787 | 0,25261 | 0,24879 | 0,24698 | 0,21092 | 0,2489639 | C                      |
| C10 | 0,6744  | 0,61784 | 0,631   | 0,6145  | 0,63786 | 0,62247 | 0,54802 | 0,42464 | 0,39383 | 0,40035 | 0,5346353 | B                      |
| C11 | 0,46041 | 0,52047 | 0,54436 | 0,48325 | 0,52468 | 0,45622 | 0,44306 | 0,55206 | 0,48921 | 0,59408 | 0,5026808 | B                      |
| C12 | 0,92548 | 1,07416 | 1,02787 | 0,88252 | 0,93573 | 0,97847 | 0,85032 | 0,96833 | 0,91432 | 0,90153 | 0,941634  | B                      |
| C13 | 1,23821 | 1,15672 | 0,70583 | 0,93128 | 0,95169 | 0,95723 | 0,90394 | 0,97358 | 0,88887 | 0,9301  | 0,944132  | B                      |
| C14 | 0,37901 | 0,48131 | 0,44524 | 0,4682  | 0,54429 | 0,50124 | 0,508   | 0,53261 | 0,55189 | 0,55501 | 0,4904345 | C                      |
| C15 | 2,36878 | 2,54529 | 2,09218 | 1,84952 | 1,80556 | 1,80176 | 1,55935 | 1,52251 | 1,27752 | 1,42652 | 1,7468988 | A                      |
| C2  | 0,47508 | 0,55554 | 0,51217 | 0,41627 | 0,38802 | 0,3584  | 0,3263  | 0,30315 | 0,33117 | 0,28036 | 0,3762686 | C                      |
| C3  | 0,5957  | 0,60881 | 0,66632 | 0,63603 | 0,65795 | 0,69289 | 0,69604 | 0,67952 | 0,58944 | 0,57673 | 0,6371057 | B                      |
| C4  | 1,03727 | 1,25311 | 1,07972 | 1,16496 | 1,02175 | 1,14401 | 1,21193 | 1,15105 | 1,18783 | 0,97073 | 1,1153997 | A                      |
| C5  | 1,36064 | 1,54844 | 1,42551 | 1,39551 | 1,46321 | 1,43324 | 1,50029 | 1,67569 | 1,46368 | 1,47982 | 1,4700993 | A                      |
| C6  | 1,9859  | 2,35446 | 2,24291 | 1,9252  | 2,10603 | 1,72071 | 1,50362 | 1,59937 | 1,65542 | 1,7544  | 1,8470925 | A                      |
| C7  | 0,1953  | 0,17989 | 0,18344 | 0,19919 | 0,20292 | 0,18984 | 0,17863 | 0,163   | 0,17015 | 0,16622 | 0,1819066 | C                      |
| C8  | 0,24166 | 0,25147 | 0,30941 | 0,2756  | 0,26197 | 0,22964 | 0,27127 | 0,30681 | 0,32625 | 0,31857 | 0,2755078 | C                      |
| C9  | 0,55353 | 0,59991 | 0,55942 | 0,53065 | 0,54082 | 0,54133 | 0,53348 | 0,52195 | 0,59072 | 0,49153 | 0,5446725 | B                      |
| D1  | 0,16582 | 0,166   | 0,17414 | 0,15699 | 0,14844 | 0,16085 | 0,15803 | 0,15512 | 0,14624 | 0,15611 | 0,1583802 | C                      |
| D10 | 1,20112 | 1,44429 | 1,50415 | 1,43983 | 1,48147 | 1,34093 | 1,2999  | 1,27217 | 2,16898 | 1,3806  | 1,4192929 | A                      |
| D2  | 0,24629 | 0,33813 | 0,35586 | 0,31859 | 0,34007 | 0,37333 | 0,37649 | 0,37379 | 0,31771 | 0,29075 | 0,3278259 | C                      |
| D3  | 0,26547 | 0,31053 | 0,30085 | 0,23471 | 0,36609 | 0,38481 | 0,40576 | 0,38121 | 0,35913 | 0,34579 | 0,3260929 | C                      |
| D4  | 0,52907 | 0,59956 | 0,54449 | 0,57011 | 0,70318 | 0,70154 | 0,64031 | 0,71052 | 0,64803 | 0,59951 | 0,6181799 | B                      |
| D5  | 0,19658 | 0,24998 | 0,21604 | 0,22074 | 0,2608  | 0,29893 | 0,35134 | 0,33721 | 0,31258 | 0,27833 | 0,2629521 | C                      |



|     |         |         |         |         |         |         |         |         |         |         |           |   |
|-----|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|-----------|---|
| D6  | 0,88153 | 0,98506 | 1,0156  | 0,94044 | 0,97203 | 0,95088 | 0,95262 | 0,95419 | 1,00637 | 0,96621 | 0,9611376 | B |
| D7  | 1,46646 | 1,4032  | 1,30812 | 1,41464 | 1,5495  | 1,28846 | 1,3636  | 1,22453 | 1,37796 | 1,35902 | 1,3706332 | A |
| D8  | 0,66853 | 0,859   | 0,85162 | 0,84634 | 0,85227 | 0,79705 | 0,81099 | 0,96414 | 1,0063  | 0,78429 | 0,834551  | B |
| D9  | 0,66766 | 0,74832 | 0,71475 | 0,75609 | 0,74827 | 0,65622 | 0,72752 | 0,64115 | 0,64267 | 0,83159 | 0,7087769 | B |
| E1  | 0,21822 | 0,27027 | 0,22643 | 0,24063 | 0,22123 | 0,20818 | 0,19099 | 0,1871  | 0,18869 | 0,2047  | 0,2130211 | C |
| E10 | 1,40474 | 2,23173 | 1,8881  | 1,936   | 2,16086 | 1,12008 | 1,40359 | 1,03901 | 1,32258 | 1,22911 | 1,4632282 | A |
| E11 | 1,14454 | 1,3306  | 1,5122  | 1,45295 | 1,28264 | 0,93392 | 1,42465 | 1,34945 | 1,10859 | 1,10872 | 1,2388257 | A |
| E2  | 0,24439 | 0,2713  | 0,25449 | 0,24773 | 0,2804  | 0,28697 | 0,28967 | 0,26934 | 0,27035 | 0,25322 | 0,265901  | C |
| E3  | 0,70439 | 0,74365 | 0,71319 | 0,59447 | 0,7541  | 0,70484 | 0,73152 | 0,63505 | 0,59851 | 0,63508 | 0,6765585 | B |
| E4  | 0,22761 | 0,302   | 0,32723 | 0,31775 | 0,31884 | 0,32421 | 0,33068 | 0,30788 | 0,26843 | 0,22718 | 0,2896216 | C |
| E5  | 1,22414 | 1,32425 | 1,61589 | 1,652   | 1,45616 | 1,5003  | 1,40479 | 1,27047 | 1,06869 | 1,13427 | 1,3394412 | A |
| E6  | 1,0751  | 1,32194 | 1,21157 | 1,20565 | 1,20285 | 1,17358 | 1,10636 | 1,08331 | 1,04187 | 1,05849 | 1,1420231 | A |
| E7  | 1,33155 | 1,45956 | 1,22702 | 1,13778 | 1,13534 | 1,08643 | 1,02569 | 0,90497 | 0,91388 | 0,85727 | 1,078521  | A |
| E8  | 0,77282 | 0,92036 | 0,92862 | 1,08409 | 1,32511 | 1,23134 | 1,23587 | 1,1283  | 1,04392 | 1,07341 | 1,0491487 | A |
| E9  | 0,57483 | 0,88094 | 0,84519 | 0,77244 | 0,72464 | 0,69795 | 0,66171 | 0,64227 | 0,63967 | 0,5983  | 0,6911679 | B |
| F1  | 0,28974 | 0,33687 | 0,30669 | 0,2771  | 0,25668 | 0,263   | 0,27034 | 0,26375 | 0,21954 | 0,22383 | 0,2666611 | C |
| F2  | 0,16351 | 0,13887 | 0,19029 | 0,23891 | 0,27717 | 0,25678 | 0,29908 | 0,58172 | 0,3439  | 0,42514 | 0,2464115 | C |
| F3  | 0,97148 | 1,10552 | 1,6259  | 1,56848 | 1,12569 | 0,94188 | 0,92926 | 0,9125  | 0,754   | 0,77986 | 1,0083569 | A |
| F4  | 0,3503  | 0,39244 | 0,38346 | 0,44824 | 0,43127 | 0,43781 | 0,43487 | 0,39491 | 0,37172 | 0,38228 | 0,4002831 | C |
| F5  | 0,28032 | 0,34381 | 0,42902 | 0,38013 | 0,427   | 0,57221 | 0,50285 | 0,43611 | 0,34218 | 0,32391 | 0,3873204 | C |
| F6  | 1,06057 | 1,0362  | 1,09509 | 1,10701 | 1,24939 | 1,21433 | 1,12721 | 0,98186 | 0,97313 | 0,95436 | 1,0718791 | A |
| G1  | 0,19403 | 0,22742 | 0,19137 | 0,16566 | 0,17364 | 0,15117 | 0,1294  | 0,12396 | 0,1371  | 0,12936 | 0,1571506 | C |
| G2  | 0,37457 | 0,4104  | 0,27633 | 0,32985 | 0,32876 | 0,28135 | 0,27385 | 0,2851  | 0,28634 | 0,27225 | 0,3060196 | C |
| G3  | 0,69737 | 0,7244  | 0,81908 | 0,74601 | 0,78302 | 0,79216 | 0,81632 | 0,77039 | 0,77693 | 0,80873 | 0,7714917 | B |
| G4  | 0,17661 | 0,21797 | 0,19157 | 0,17012 | 0,17905 | 0,14413 | 0,1333  | 0,14296 | 0,15326 | 0,13502 | 0,1611514 | C |
| G5  | 0,88702 | 0,81784 | 0,88563 | 0,81552 | 0,80969 | 0,65409 | 0,60931 | 0,66491 | 0,69531 | 0,64373 | 0,7349615 | B |
| G6  | 1,07231 | 1,17074 | 1,24247 | 1,27947 | 1,27211 | 1,3278  | 1,23972 | 1,17336 | 1,21548 | 1,0608  | 1,1995212 | A |
| G7  | 0,42671 | 0,60931 | 0,60126 | 0,66864 | 0,83872 | 0,76295 | 0,68743 | 0,75    | 0,73112 | 0,7173  | 0,6580521 | B |
| G8  | 1,06651 | 1,58318 | 1,86249 | 1,72725 | 1,62522 | 1,33169 | 0,97824 | 1,01844 | 0,96543 | 0,97585 | 1,2321243 | A |
| H1  | 0,16064 | 0,1963  | 0,24109 | 0,20298 | 0,18075 | 0,15722 | 0,15285 | 0,14328 | 0,15225 | 0,14325 | 0,1684697 | C |
| H10 | 0,71399 | 0,6343  | 0,64201 | 0,61884 | 0,67349 | 0,61702 | 0,58615 | 0,52361 | 0,5337  | 0,52221 | 0,6002161 | B |
| H11 | 0,75257 | 1,3424  | 1,36122 | 1,51372 | 1,58715 | 1,35996 | 1,35154 | 1,30361 | 1,21502 | 1,10004 | 1,2386302 | A |
| H2  | 0,35043 | 0,37046 | 0,30977 | 0,39933 | 0,22752 | 0,23951 | 0,2184  | 0,21846 | 0,22819 | 0,21421 | 0,2627265 | C |
| H3  | 0,68322 | 0,61154 | 0,60687 | 0,77716 | 0,55391 | 0,5533  | 0,60011 | 0,55846 | 0,53152 | 0,46697 | 0,5839082 | B |
| H4  | 0,1755  | 0,18166 | 0,22118 | 0,20367 | 0,15538 | 0,17681 | 0,15278 | 0,14844 | 0,1945  | 0,17899 | 0,1762541 | C |
| H5  | 0,24677 | 0,2459  | 0,29442 | 0,29727 | 0,22847 | 0,18519 | 0,19901 | 0,21547 | 0,23284 | 0,2206  | 0,2317486 | C |
| H6  | 0,91082 | 0,9776  | 0,91185 | 0,79911 | 0,76678 | 0,71141 | 0,66437 | 0,68864 | 0,67568 | 0,70345 | 0,7670134 | B |
| H7  | 0,15625 | 0,39174 | 0,88378 | 1,0078  | 1,02367 | 0,84781 | 0,83969 | 7,95192 | 0,6818  | 0,66542 | 0,5708072 | B |
| H8  | 0,95052 | 0,95485 | 0,87563 | 0,90152 | 0,89626 | 0,88146 | 0,89736 | 0,90235 | 0,9031  | 0,78069 | 0,8919351 | B |
| H9  | 0,56485 | 0,68954 | 0,77879 | 1,01282 | 0,86949 | 0,86455 | 0,96062 | 1,03591 | 0,96675 | 0,91541 | 0,8380882 | B |
| I1  | 0,1701  | 0,18711 | 0,18029 | 0,18768 | 0,18804 | 0,19115 | 0,17954 | 0,18788 | 0,17693 | 0,1965  | 0,1842246 | C |
| I2  | 0,29334 | 0,29311 | 0,3     | 0,28411 | 0,29997 | 0,33344 | 0,365   | 0,36632 | 0,32187 | 0,28423 | 0,3115147 | C |
| I3  | 0,56078 | 0,97012 | 0,61339 | 0,79969 | 0,87194 | 0,88505 | 0,90228 | 1,02031 | 0,93502 | 0,94773 | 0,8207096 | B |
| I4  | 0,21054 | 0,29957 | 0,24025 | 0,21409 | 0,14976 | 0,13674 | 0,13613 | 0,13847 | 0,13134 | 0,13324 | 0,1650221 | C |
| I5  | 0,71937 | 0,90553 | 0,81926 | 0,82232 | 0,7679  | 0,74877 | 0,75182 | 0,77596 | 0,85956 | 0,77354 | 0,7908964 | B |
| I6  | 0,93467 | 1,27706 | 1,1775  | 1,1789  | 1,16148 | 1,18972 | 1,18459 | 1,17233 | 1,20426 | 1,10485 | 1,1514614 | A |
| I7  | 0,76818 | 1,13757 | 0,80061 | 1,19148 | 1,09585 | 1,1934  | 1,19122 | 1,18861 | 0,93786 | 1,29789 | 1,0485259 | A |
| J1  | 0,78922 | 0,62254 | 0,67155 | 0,59786 | 0,55102 | 0,52023 | 0,53678 | 0,48345 | 0,5413  | 0,46987 | 0,5656372 | B |
| J2  | 0,86718 | 0,82259 | 0,88572 | 0,80919 | 0,87816 | 0,83923 | 0,92132 | 0,80686 | 0,87118 | 0,76945 | 0,8448638 | B |
| J3  | 0,55383 | 0,49447 | 0,51184 | 0,49976 | 0,44932 | 0,39809 | 0,35631 | 0,36203 | 0,23908 | 0,20491 | 0,3682986 | C |
| J4  | 1,06861 | 0,81232 | 1,07598 | 1,26327 | 1,37553 | 1,28142 | 1,19327 | 1,08302 | 1,07466 | 0,99946 | 1,1008481 | A |
| J5  | 0,71354 | 0,85421 | 0,78655 | 0,84248 | 0,97622 | 0,8829  | 0,83833 | 0,75245 | 0,6918  | 0,67869 | 0,7918762 | B |
| K1  | 0,30561 | 0,26236 | 0,22455 | 0,17968 | 0,15375 | 0,1362  | 0,12952 | 0,14699 | 0,13312 | 0,11124 | 0,1613569 | C |
| K2  | 0,22362 | 0,24936 | 0,22066 | 0,19439 | 0,18568 | 0,2425  | 0,24935 | 0,24641 | 0,20412 | 0,17077 | 0,2150934 | C |
| K3  | 0,50915 | 0,48861 | 0,53171 | 0,6228  | 0,57945 | 0,57431 | 0,66734 | 0,72146 | 0,61196 | 0,53821 | 0,5766767 | B |
| K4  | 0,63807 | 0,83376 | 0,85489 | 0,80994 | 0,89532 | 0,88788 | 0,87274 | 0,88878 | 0,86547 | 0,75045 | 0,8214812 | B |
| K5  | 0,47914 | 0,63369 | 0,56394 | 0,60279 | 0,60728 | 0,55852 | 0,53007 | 0,53818 | 0,55569 | 0,51825 | 0,5553019 | B |
| L1  | 0,17239 | 0,22857 | 0,24222 | 0,13784 | 0,1509  | 0,14793 | 0,15429 | 0,13295 | 0,14632 | 0,13621 | 0,1584168 | C |
| L2  | 0,70328 | 0,81059 | 0,72768 | 0,75379 | 0,78389 | 0,73167 | 0,69676 | 0,76286 | 0,8563  | 0,8553  | 0,7644123 | B |
| L3  | 0,62686 | 0,69109 | 0,73434 | 0,67909 | 0,67064 | 0,62359 | 0,61292 | 0,59145 | 0,49241 | 0,54773 | 0,6191917 | B |
| L4  | 1,11273 | 1,52828 | 1,6107  | 1,95339 | 1,91163 | 1,38575 | 1,65275 | 2,08233 | 1,62594 | 1,89863 | 1,6240212 | A |
| M1  | 0,122   | 0,12801 | 0,12547 | 0,11272 | 0,13901 | 0,14803 | 0,14685 | 0,14667 | 0,15816 | 0,12771 | 0,1340838 | C |
| M2  | 0,27005 | 0,24511 | 0,22538 | 0,21266 | 0,24822 | 0,23587 | 0,24203 | 0,29044 | 0,27914 | 0,25686 | 0,2485041 | C |
| M3  | 0,3267  | 0,3386  | 0,33595 | 0,30039 | 0,40383 | 0,49174 | 0,56868 | 0,61218 | 0,5332  | 0,53265 | 0,4171066 | C |
| M4  | 0,4308  | 0,49278 | 0,46596 | 0,44839 | 0,43543 | 0,41829 | 0,42093 | 0,46762 | 0,39737 | 0,40043 | 0,4358786 | C |
| M5  | 0,32111 | 0,42662 | 0,50956 | 0,53943 | 0,4707  | 0,44692 | 0,3597  | 0,395   | 0,38884 | 0,382   | 0,4142879 | C |
| M6  | 1,06342 | 1,25833 | 1,1387  | 1,11178 | 1,0822  | 1,00415 | 0,98351 | 0,85317 | 0,86959 | 0,81072 | 0,9995044 | B |
| M7  | 0,65087 | 0,74853 | 1,02954 | 0,87865 | 0,88785 | 0,76775 | 0,68664 | 0,68382 | 0,62407 | 0,58215 | 0,7326141 | B |
| M8  | 1,13818 | 1,63064 | 1,53865 | 1,45845 | 1,45822 | 1,26941 | 1,0825  | 1,10313 | 1,30813 | 0,98415 | 1,2639313 | A |
| M9  | 0,71685 | 0,96154 | 5,7775  | 1,20845 | 1,17332 | 1,37951 | 0,94883 | 1,00748 | 1,11343 | 1,02797 | 1,1197971 | A |

Figures 19, 20 & 21: Representing student success to institutional performance index **Per Year** and the classification classes according to the harmonic mean calculations for 105 institutions over 13 universities from 2012 to 2022.

## Regression Models: M1S, M2P, and M3G

In contrast to the classification model, the regression models were designed to predict the specific values of the composite indexes. The dataset was decomposed into three subsets, each tailored to train a specific regression sub-model:

1. **M1S (Model for Success Index):** This sub-model was trained to predict the Success Index. The inputs included the data elements associated with the composition of the Success Index, such as the number of graduates and total student enrollments. By focusing on these elements, the model aimed to accurately forecast the Success Index based on the institution's performance in facilitating student success.
2. **M2P (Model for Performance Index):** The second sub-model targeted the prediction of the Performance Index. The training dataset for M2P included variables related to the success rate, pedagogical coverage rate, and administrative coverage rate. These inputs were crucial for enabling the model to learn the intricate relationships between institutional operational metrics and the Performance Index, thereby optimizing its predictive capabilities.
3. **M3G (Model for Growth Index):** The third sub-model, M3G, was developed to predict the Growth Index. The input dataset for this model comprised factors influencing institutional growth, such as new student enrollments, total student enrollments, and occupancy capacity. The model leveraged these data points to forecast the Growth Index, offering insights into the institution's ability to manage and sustain growth.

Each of these regression sub-models was specifically designed to maximize the learning process by focusing on the relevant data elements associated with their respective composite index. This decomposition into specialized sub-models allowed for a more targeted and effective prediction, ensuring that each model could capture the nuances and complexities inherent in the different aspects of institutional performance.



The structured approach to dataset partitioning and the differentiation between classification and regression tasks enabled a comprehensive analysis of the institutions. By training on the majority of the data and testing on a separate subset, the models were rigorously validated to ensure their accuracy and generalizability. This methodology not only facilitated precise predictions and classifications but also provided valuable insights for decision-making and strategic management within the public higher education sector in Morocco.

### **SVM models implementation:**

The implementation of the SVM models focuses on developing both classification and regression algorithms to categorize institutions and predict future performance metrics.

1. SVM Classification Model Implementation: The SVM classification model is designed to categorize public higher education institutions into distinct classes based on a composite index of performance metrics. The implementation involves:

- **Model Construction:** The SVM algorithm constructs a decision boundary, known as the hyperplane, which separates the data into different classes. The choice of kernel function such as linear, polynomial, or radial basis function (RBF) is crucial in defining this hyperplane, particularly when the data is not linearly separable.
- **Optimization:** The model's parameters, including the regularization parameter (C) and the kernel-specific parameters, are optimized. The regularization parameter controls the trade-off between achieving a low error on the training data and minimizing the margin width. Kernel parameters, such as gamma in the RBF kernel, determine the influence of individual data points on the decision boundary.

2. SVM Regression Model Implementation: Three SVM regression models are developed to predict the values of three composite indexes, each reflecting different aspects of institutional performance over a ten-year forecast period. The implementation process includes:

- **Model Structure:** The SVM regression, also known as Support Vector Regression (SVR), is configured to predict continuous outcomes. Unlike traditional linear regression, SVR uses a margin of tolerance (epsilon) within which no penalty is given for errors, enhancing its robustness to outliers.
- **Kernel Application:** The choice of kernel function in SVR is similarly critical as in classification, allowing the model to capture non-linear relationships between input features and the target index. The model seeks to fit a regression line (or hyperplane in multi-dimensional space) that best approximates the actual data points, with support vectors defining the margin boundaries.
- **Loss Function:** The SVR model uses an epsilon-insensitive loss function, which ignores errors within a certain range (epsilon). The objective is to minimize a loss function that includes this insensitivity parameter, along with the regularization term, to achieve a balance between the complexity of the model and the accuracy of the predictions.

Both the SVM classification and regression models are implemented using a series of iterative training processes, where the model parameters are adjusted to minimize the classification error and regression prediction error, respectively. These models leverage the strength of SVMs in handling high-dimensional data and capturing complex relationships, making them well-suited for the diverse and multifaceted data of public higher education institutions.

## RESULTS

The empirical testing of Support Vector Machine (SVM) models demonstrates the decisive role of the carefully constructed composite indexes Success Index (S), Performance Index (P), and Growth Index (G) in enhancing both the accuracy and interpretability of institutional evaluations in Moroccan public Higher Education Institutions (HEIs). The integration of these indexes as structured inputs ensured the reliability of classification models and the robustness of regression models,

ultimately reinforcing their capacity to support data-informed strategic decision-making.

## Classification Results

The classification model achieved an **Overall Accuracy (OA) of 95.24%**, indicating that the SVM classifier was able to accurately assign institutions to their respective classes based on the composite indexes. Such a high level of predictive accuracy is rarely achieved in institutional studies that rely on fragmented or unstructured data. The result highlights that the design of the Success, Performance, and Growth Indexes successfully captured the underlying multidimensional features of institutional effectiveness and efficiency.

The OA value underscores the value of **feature engineering through composite indexes**. Instead of relying on raw or heterogeneous variables, the integration of well-structured, normalized, and weighted indexes facilitated the detection of nuanced institutional patterns, thus improving classification precision. For top management, this means that institutional typologies can be identified with high confidence, providing a robust basis for policy interventions, capacity allocation, and targeted support strategies.

## Regression Results

To complement the classification results, regression models were trained separately on each of the three composite indexes, enabling predictive insights into institutional outcomes.

- **Model M1S (Success Index)** achieved a **Mean Squared Error (MSE) of 0.02**, a **Root Mean Squared Error (RMSE) of 0.13**, and a **Mean Absolute Error (MAE) of 0.03**. These results reflect the model's ability to predict student outcomes with high precision, confirming that success is a measurable and stable indicator when integrated into SVM regression frameworks.
- **Model M2P (Performance Index)** demonstrated exceptional accuracy, with **MSE of 0.00**, **RMSE of 0.03**, and **MAE of 0.01**. This near-perfect predictive power reveals that the Performance Index, as a weighted composite of academic and administrative efficiency factors, is the most reliable determinant of institutional behavior. It validates the methodological decision to emphasize weighting and aggregation methods in its construction, particularly the prioritization of student success outcomes (50%) relative to coverage



measures.

- **Model M3G (Growth Index)** reported **MSE of 0.03**, **RMSE of 0.17**, and **MAE of 0.08**. While slightly less precise than the other two models, the results remain within a strong predictive range. The higher error margins may reflect the inherent volatility of institutional growth, which depends on external factors such as demographic dynamics, national education policy, and resource allocation. Nevertheless, the Growth Index still provides valuable insights into capacity management and scalability.

## DISCUSSION

The performance of the SVM models demonstrates the **central importance of composite index design** in supporting advanced machine learning applications within higher education management. The exceptional classification accuracy (95.24%) and the low error margins in regression tasks validate the methodological rigor of combining normalization, weighting, and aggregation strategies in constructing S, P, and G.

From a strategic management perspective, these results translate into actionable benefits for decision-making:

1. **Precision in Typology Identification** – High classification accuracy ensures that institutional classes (e.g., high performance vs. low growth) are defined with reliability, facilitating differentiated governance strategies.
2. **Predictive Insights for Policy** – Regression outputs allow policymakers to forecast institutional trajectories based on current input conditions, making it possible to anticipate risks (e.g., declining success rates) or opportunities (e.g., scalable growth).
3. **Optimization of Resource Allocation** – The Performance Index, with its near-zero prediction error, emerges as the most decisive factor for institutional efficiency, guiding resource prioritization toward pedagogical and administrative staffing as leverage points for improvement.
4. **Support for Data-Informed Governance** – The integration of SVM with composite indexes bridges the gap between quantitative analytics and strategic management, enabling HEIs to transition from descriptive assessments to predictive and prescriptive strategies.

In summary, the careful design and integration of composite indexes significantly enhance the performance of SVM models, demonstrating their potential as

decision-support tools for Moroccan HEIs. By embedding such models into governance frameworks, institutions can move toward **optimized, data-driven, and transparent strategic management**.

## CONCLUSION

The findings of this study underline the transformative role of composite indexes, carefully designed and adapted to the context of Moroccan Higher Education Institutions, in addressing the complexity of decision-making processes. By consolidating heterogeneous data into the Success Index, the Performance Index, and the Growth Index, this research demonstrated that abstract and fragmented variables can be transformed into actionable, comparable, and quantifiable measures. These indexes not only provided clarity in evaluating student outcomes, institutional efficiency, and growth capacity but also served as the backbone for the machine learning framework. Their integration into Support Vector Machine models proved decisive in achieving both exceptionally high classification accuracy, with an overall performance of 95.24%, and strong predictive capabilities in regression analyses, particularly with the Performance Index model, which achieved near-zero error rates.

The results illustrate that the construction of well-defined indexes is not a mere technical exercise, but a strategic necessity for optimizing decision-making in higher education. The methodological rigor applied in weighting, aggregating, and normalizing these indexes ensured their reliability and interpretability, allowing SVM models to uncover patterns that would otherwise remain hidden in raw datasets. This emphasizes the broader implication that the quality of decision-support systems in education is highly dependent on the conceptual soundness of the indicators on which they are built.

From a policy perspective, the adoption of such a framework carries significant potential. Ministries of Higher Education and institutional leaders can utilize these models to classify institutions with confidence, anticipate future trajectories, and allocate resources more efficiently. The Performance Index, for instance, provides a clear and empirically validated signal that staffing and coverage levels represent leverage points for improving institutional effectiveness. Likewise, the Growth Index offers a tool for forecasting scalability challenges, ensuring that expansion policies are aligned with actual capacity. By institutionalizing the use of such composite indexes in governance frameworks, higher education systems can strengthen transparency, accountability, and long-term strategic planning.

Future developments may enhance this framework further, such as by in-

roducing dynamic weighting methods that reflect evolving national priorities or by testing alternative machine learning approaches, including ensemble methods or deep learning, to enrich predictive capabilities. Beyond institutional efficiency and growth, the same analytical logic can be extended to critical areas such as student retention, faculty performance evaluation, or curriculum impact assessment, thus broadening the strategic reach of data-driven decision-making.

In conclusion, this study demonstrates that the real strength of machine learning in higher education management does not solely lie in the sophistication of algorithms, but in the formulation of composite indexes that translate complex realities into structured, comparable, and meaningful indicators. By creating Success, Performance, and Growth Indexes specifically tailored to the Moroccan context and embedding them in SVM models, this research provides both a methodological innovation and a policy-relevant tool that can transform governance practices. The synergy between well-constructed indexes and advanced analytical models thus offers a path toward more informed, efficient, and sustainable strategic management in higher education.



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## LIST OF ABBREVIATIONS

**ACR:** administrative coverage rate  
**HEI's:** Higher Education Institutions  
**MAE:** Mean Absolute Error  
**ML:** Machine Learning  
**MSE:** Mean Squared Error  
**OA:** Overall Accuracy  
**PCA :** Principal Component Analysis  
**PCR:** pedagogical coverage rate  
**RMSE:** Root Mean Squared Error  
**SR:** success rate  
**SVM:** Support Vector Machine

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