

# Corporate Governance: An Evaluative and Evolutionary Approach

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**Abstract:** Corporate governance is essential for optimizing performance and ensuring the sustainability of contemporary organizations. This article offers an in-depth analysis that combines evaluative and evolutionary approaches to provide a comprehensive understanding of governance practices. The evaluative approach is based on the examination of key indicators such as financial profitability, operational transparency, stakeholder satisfaction, risk management, and environmental responsibility. Complementing this, the evolutionary approach utilizes analytical tools like Markovian modeling to capture the dynamics and evolution of governance practices over time, thereby offering an integrated and temporal perspective on the continuous improvement of organizations.

**Keywords:** Corporate governance, organizational sustainability, evaluative approach, Markovian modeling, dynamic approach.

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## 1. Introduction

Corporate governance, a fundamental pillar of modern organizational management, plays a central role in orchestrating relationships between various stakeholders and the strategic direction of companies. It extends beyond merely managing shareholders' interests, encompassing social, environmental, and ethical dimensions that have become essential in evaluating organizational performance today (Shleifer & Vishny, 1997; Freeman, 1984). Since the 1980s, research in corporate governance has evolved around two main axes: improving the theoretical understanding of governance models and developing standardized measures for objective evaluation of practices (Jensen & Meckling, 1976). However, traditional financial tools, often focused on financial performance indicators, have been criticized for their short-term approach and lack of consideration for non-financial criteria that are equally crucial to a holistic assessment of governance (Eisenhardt, 1989; Williamson, 1985). In light of these limitations, the evaluative approach, which focuses on analyzing performance through indicators such as financial profitability and operational transparency, remains an indispensable method for identifying the strengths and weaknesses of governance structures (Aguilera & Cuervo-Cazurra, 2004). Nevertheless, this approach, while essential, often suffers from a static perspective, offering a limited view of the temporal dynamics and the evolution of governance practices over time (Larcker & Tayan, 2011). It is here that the evolutionary approach, integrating sophisticated analytical tools such as Markov chains, becomes particularly significant. These models capture the transformations of governance practices over a given period, tracking transitions between different states of governance and anticipating future challenges based on the analysis of past trends (Fudenberg & Tirole, 1991). Markovian modeling, in particular, emerges as a powerful tool for understanding the dynamics of governance practices and adjusting policies accordingly, offering a longitudinal perspective that complements traditional static evaluations.

Thus, the central issue addressed by this article lies in the comprehensive and dynamic assessment of corporate governance quality, considering not only current weaknesses but also future challenges and opportunities for continuous improvement. To address this issue, this article proposes an integrative approach combining evaluative and evolutionary methods, thereby offering a more complete understanding of corporate governance. The objective is to provide an assessment of current governance practices through key indicators while integrating a temporal dimension through Markovian modeling, allowing for tracking the evolution of governance practices and anticipating future transformations.

The originality of this article lies in this dual approach, which combines static evaluation with an evolutionary perspective, providing an innovative response to existing literature. Indeed, where traditional analyses often limit themselves to point-in-time evaluations, this research proposes a methodology that not only assesses current performance but also predicts future trajectories of governance practices. This contribution fills a gap in the literature by providing an analytical framework applicable to various organizational contexts, aimed at optimizing the management of governance challenges in an ever-changing environment.

This article is structured to offer a comprehensive and nuanced analysis. It begins with a literature review that explores corporate governance theories as well as evaluative and evolutionary approaches. Then, it presents a theoretical and conceptual framework defining key concepts and the interaction between these two approaches before detailing the methodology used. The study's results, both static and dynamic, are then presented, followed by a discussion of the theoretical and practical implications. Finally, the article concludes by summarizing the main contributions and opening up perspectives for future research on corporate governance.

## 2. Literature review

### 2.2 Evolution of corporate governance

Corporate governance has undergone significant transformation over the past decades, evolving in response to economic, regulatory, and social changes. Initially centered on maximizing shareholder interests, corporate governance has gradually broadened to adopt a more holistic perspective, considering a wider range of stakeholders and addressing social and environmental concerns.

The earliest academic work on corporate governance dates back to the 1930s with the publication of "The Modern Corporation and Private Property" by Berle and Means (1932), which highlighted the separation between ownership and control in large corporations. This separation laid the groundwork for governance debates by emphasizing potential conflicts of interest between shareholders and executives. However, it was from the 1980s onward that corporate governance research truly flourished, particularly with the emergence of agency theory and the development of mechanisms aimed at aligning the interests of executives with those of shareholders (Jensen & Meckling, 1976).

In the following decades, corporate governance became more complex due to the internationalization of markets, evolving stakeholder expectations, and intensified regulations. Today, corporate governance encompasses not only economic dimensions but also social and environmental responsibilities in response to growing pressures for sustainable and ethical practices (Aguilera, Rupp, Williams, & Ganapathi, 2007).

### 2.3 Governance theories

Corporate governance theories have sought to explain the structures and practices that resolve inherent conflicts of interest between ownership and control, as well as manage relationships among various stakeholders.

- **Agency theory**, introduced by Jensen and Meckling (1976), is one of the most influential. It posits that executives, as agents of shareholders, may have interests that diverge from those of the company's owners. This potential conflict of interest necessitates governance mechanisms to align executives' interests with those of shareholders, such as financial incentives, board oversight, and external takeovers.
- **Stakeholder theory** (Freeman, 1984) broadens the governance perspective by asserting that companies must consider the interests of all stakeholders, not just shareholders. This theory emphasizes the importance of balancing the interests of different stakeholders, including employees, customers, suppliers, and the community, to ensure the long-term sustainability of the company.
- **Institutional theory**, on the other hand, focuses on the influence of institutions, such as regulations, social norms, and cultural expectations, on governance practices (DiMaggio & Powell, 1983). This theory argues that companies adopt certain governance practices not only for economic efficiency but also to gain the legitimacy necessary for survival in a given institutional environment.

### 2.4 Evaluative approach

The evaluative approach to corporate governance relies on analyzing various indicators that measure the quality of governance practices. Historically, the focus has been on financial indicators, such as financial profitability, return on investment (ROI), and shareholder value (Fama & Jensen, 1983). These indicators provide a direct measure of economic performance and managerial effectiveness in the interest of shareholders.

Beyond financial indicators, operational transparency has become a crucial aspect of governance evaluation. The frequency and clarity of financial reporting, as well as the disclosure of business practices, are key elements in maintaining stakeholder trust and minimizing information asymmetry (Healy & Palepu, 2001).

More recent approaches integrate non-financial dimensions, such as stakeholder satisfaction, risk management, and environmental responsibility. These indicators reflect a more balanced view of organizational performance, considering not only financial outcomes but also the social and environmental impact of governance practices (Eccles, Ioannou, & Serafeim, 2014).

## 2.5 Evolutionary approach

The evolutionary approach to corporate governance aims to capture the temporal dynamics of governance practices, focusing on how these practices evolve over time in response to internal and external changes. One of the most promising tools in this approach is Markovian modeling, which allows for the modeling of transitions between different states of governance over a given period.

Markov chains are particularly useful for analyzing decision-making processes in uncertain contexts, where current decisions probabilistically influence future outcomes. By applying these models to corporate governance, researchers can track the evolution of governance practices, identify transitions to more or less effective governance states, and anticipate future challenges based on the analysis of past trends (Hamilton & Zeckhauser, 2004).

Integrating the evolutionary approach into governance evaluation enables a more dynamic analysis, accounting for the transformations of practices and strategies in an ever-changing environment. This perspective is essential for understanding not only current performance but also the future trajectories of companies in an increasingly complex context (Tirole, 2001).

## 3. Methodology: Corporate governance evaluation and evolution Process

### 3.1 Definition of evaluation criteria

The evaluation of corporate governance requires a structured methodological approach, enabling a systematic and reproducible analysis of organizational performance. This section presents a rigorous evaluation framework based on two pillars: measurement through quantitative and qualitative indicators and evolutionary analysis to track changes in governance practices over time. This framework integrates evaluative and evolutionary approaches while considering sectoral specifics and organizational contexts.

Corporate governance evaluation is multidimensional and relies on specific criteria that directly impact organizational performance and long-term value creation. These criteria, recognized in academic literature, include:

- **Financial Profitability:** Measures a company's ability to generate profits relative to its resources and investments. Typical indicators include net profit margin, return on investment (ROI), and return on equity (ROE).
- **Operational Transparency:** Essential for maintaining stakeholder trust and ensuring clear communication about internal practices. It includes the frequency of financial reporting and disclosure of business practices.
- **Stakeholder Satisfaction:** Crucial for organizational sustainability, stakeholder satisfaction includes shareholders, employees, and customers, measured by indicators like employee retention rate, satisfaction surveys, and shareholder dividends.

- **Risk Management:** Evaluates the effectiveness of identifying, assessing, and mitigating risks. Indicators include the frequency of internal audits and the implementation of risk management processes.
- **Environmental Responsibility:** Increasingly integrated into governance evaluations due to the growing importance of sustainable practices. Indicators include carbon emission reductions and the use of renewable energy.

### 3.2 Definition of performance indicators

In order to conduct a rigorous evaluation of corporate governance, the selection and definition of performance indicators are crucial. Indicators must be specific, measurable, and quantifiable to allow for objective evaluation and ensure that the assessment process is both systematic and reproducible. Each criterion is assessed using precise indicators, as described below:

#### 3.2.1 Financial profitability

Financial profitability is a fundamental criterion, reflecting an organization's ability to generate profit relative to its resources and investments. The indicators selected for this criterion provide a clear picture of the financial health and efficiency of the enterprise. These indicators are widely used in management and finance studies (Fama & French, 1992), as they provide a direct view of the economic performance of a company and allow reliable sectoral comparisons. They also meet the expectations of financial stakeholders, including shareholders and institutional investors.

- **Return on Investment (ROI):** This indicator is calculated as the ratio of net profit to total investment, reflecting how effectively an organization is using its investments to generate profit. A higher ROI indicates better financial efficiency.
- **Net Profit Margin:** This is the ratio of net profit to total revenue, representing the percentage of revenue that is converted into profit. It provides insight into the operational efficiency and cost management of the organization.
- **Return on Equity (ROE):** ROE measures the profitability generated by the shareholders' equity. It is calculated as net income divided by shareholders' equity and indicates how well the company is using equity to generate profits.

#### 3.2.2 Operational Transparency

Operational transparency is critical for maintaining stakeholder trust and ensuring clear communication regarding internal practices. This criterion is assessed through indicators that measure both the frequency and quality of the organization's reporting practices. Healy & Palepu (2001) demonstrated that transparency improves decision-making and risk management, reducing information asymmetry. This is fundamental in complex markets where decisions must be based on reliable and accessible information.

- **Frequency of financial reports:** This indicator measures the number of financial reports published on an annual or quarterly basis. Regular reporting reflects a commitment to transparency and accountability.

- **Clarity of communications:** This is a qualitative score based on an evaluation of the transparency of financial reports and business practices. It assesses how clearly and effectively the organization communicates its operations to stakeholders.

### 3.2.3 Stakeholder satisfaction

Stakeholder satisfaction is a key indicator of organizational sustainability. It reflects the company's ability to meet the needs and expectations of its shareholders, employees, and customers. Freeman's (1984) stakeholder theory states that companies that effectively meet stakeholders' expectations increase their chances of long-term success. Satisfaction indicators reflect organizational stability and effective governance in managing the conflicting interests of stakeholders.

- **Employee satisfaction rate:** This is the proportion of satisfied employees relative to the total number of employees. It provides insights into the organization's internal environment and its ability to retain talent.
- **Customer retention rate:** This indicator measures customer loyalty by tracking the percentage of customers who continue to engage with the company over a given period. High retention rates suggest strong customer satisfaction and brand loyalty.

### 3.2.3 Risk management

Effective risk management is essential for an organization to identify, evaluate, and mitigate potential risks. The indicators for this criterion focus on the robustness of the company's internal controls and risk management processes. Power (2007) points out that risk management has become a central component of modern corporate governance. Well-defined risk indicators allow for effective monitoring of vulnerability factors in complex environments.

- **Number of successful internal audits:** This indicator measures the number of internal audits conducted without significant findings. A higher number of successful audits indicates strong internal controls and effective risk management.
- **Frequency of risk management audits:** The regularity of risk management audits is a measure of the company's proactive approach to identifying and mitigating risks. Frequent audits suggest a strong commitment to maintaining robust risk management practices.

### 3.2.3 Environmental Responsibility

With the increasing importance of sustainable practices, environmental responsibility has become a crucial criterion in the evaluation of corporate governance. The indicators selected for this criterion reflect the organization's commitment to reducing its environmental impact. Eccles and Serafeim (2013) demonstrated that companies adopting sustainable practices have better long-term financial performance. The integration of environmental criteria into governance responds to growing expectations from investors and regulators for sustainable business practices.

- **Reduction in carbon emissions:** This indicator tracks the decrease in greenhouse gas emissions over time. It measures the effectiveness of the company's efforts to reduce its carbon footprint.
- **Use of Renewable energy:** This indicator measures the adoption of sustainable energy sources and their proportion in the company's total energy consumption. A higher percentage indicates a stronger commitment to environmental sustainability.

Each of these indicators is carefully selected to align with the overall goals of the organization and the specific contexts in which it operates. They are designed to provide actionable insights into the organization's performance, allowing for targeted improvements in governance practices. The selection and definition of these indicators are grounded in established best practices and academic literature, ensuring that the evaluation process is both comprehensive and robust.

Moreover, it is important to note that the indicators should be reviewed and updated periodically to reflect changes in organizational goals, industry standards, and external regulatory requirements. This iterative approach ensures that the evaluation process remains relevant and aligned with the evolving landscape of corporate governance.

### 3.3 Evolutionary approach: Markovian modeling

In this section, we explore the application of Markovian modeling to understand the dynamics of an enterprise's performance over time (Ghahramani, 2022; Koller & Friedman, 2021; Lattimore & Szepesvári, 2020; Rabi & Shin, 2023). By defining discrete levels of governance, from weak to exemplary, we construct a Markov chain where each state represents a specific level of performance. The transition probabilities between these states are estimated based on historical data, reflecting the likelihood of an enterprise moving from one level of governance to another over successive periods. This approach provides valuable insights into the stability and progression of corporate governance and helps in predicting future performance trends. The resulting transition matrix and visualizations offer a comprehensive view of how enterprises navigate through various governance levels, facilitating better strategic decision-making. To do this we will need to define the following elements:

- **State space:** In our modeling, the state space  $\mathcal{S}$  represents the different levels of overall governance performance of a company. This space is defined by a finite set of governance levels, denoted as  $(s_1, s_2, s_3, s_4, s_5)$ , where:

$$\mathcal{S} = \{s_1, s_2, s_3, s_4, s_5\}.$$

Each state ( $s_i$ ) in  $\mathcal{S}$  corresponds to a specific performance level:

- ( $s_1$ ): Weak Governance (low compliance with criteria, low overall score),
- ( $s_2$ ): Medium Governance (partial compliance with criteria),
- ( $s_3$ ): Satisfactory Governance (acceptable compliance with criteria, but improvements possible),
- ( $s_4$ ): High Governance (strong compliance with criteria),



- ( $s_5$ ): Exemplary Governance (full compliance with criteria, best possible performance).

- **Transition matrix K:** The transition matrix is the central tool of this modeling. It quantifies the transition probabilities between different governance states, estimated from historical data. For example, the probability of a company moving from a "medium compliance" state to a "high compliance" state can be determined based on performance history and changes in governance practices. In this article the transition matrix will be noted  $K = K(i, j)$  is a square matrix is defined later and where  $i$  and  $j$  represent the indices for each state.
- **The Markov chain:** The Markov chain  $\mathbf{X} = \{X_t\}_{t \geq 0}$  is a discrete-time stochastic process that models the evolution of the company's governance performance over time.

The Markov chain  $X$  is defined on the state space  $\mathcal{S}$  and satisfies the Markov property, i.e., the probability of transition to a future state ( $s_j$ ) depends only on the current state ( $s_i$ ) and is independent of previous states:

$$P(X_{t+1} = s_j \mid X_t = s_i, X_{t-1} = s_{t-1}, \dots, X_0 = s_0) = P(X_{t+1} = s_j \mid X_t = s_i) = K(i, j),$$

where  $K(i, j) = \widehat{p}_{ij}$  represents the transition probability from state ( $s_i$ ) to state ( $s_j$ ).

- **Estimation of transition matrices based on observed frequencies**

The transition matrices ( $K(i, j)$ ) model the probabilities of moving between different levels of performance. The estimated transition probability ( $p_{ij}$ ) from state ( $i$ ) (current governance level) to state ( $j$ ) (future governance level) is calculated from observed historical data as follows:

$$K(i, j) = \widehat{p}_{ij} = \frac{n_{ij}}{n_i} = \frac{\sum_{t=0}^{T-1} n_{ij}(t)}{\sum_{t=0}^{T-1} n_i(t)},$$

where  $n_{ij}$  is the number of observed transitions from state ( $i$ ) to state ( $j$ ) during the observation period  $T$ .  $n_i$  is the total number of times the company is in state ( $i$ ) during the observation period  $T$ , and  $T$  represents the total observation period.

The transition matrix  $K$  captures the probabilities that a company will move from one governance level to another between successive periods, providing a quantitative framework to evaluate governance practice dynamics and predict future performance trends.

### 3.4 Markov Chain Simulation for Performance Evaluation

To further enhance the robustness of the performance evaluation, we incorporate a Markov chain model to simulate the stationary distribution of governance practices. The stationary distribution reflects the long-term probabilities of the system being in various states of governance, providing insights into the expected future performance of the company.

The stationary distribution  $\boldsymbol{\pi}$  of a Markov chain is a probability vector that satisfies the following equation:

$$\boldsymbol{\pi}K = \boldsymbol{\pi},$$



where  $K$  is the transition matrix of the Markov chain  $X$ . This distribution provides the long-term probabilities of a company being in each state of governance (e.g., low, medium, high compliance) under the assumption that the system has reached equilibrium.

The stationary distribution is particularly useful in predicting the steady-state behavior of a company's governance practices. By simulating the stationary distribution, we can determine the likelihood that a company will maintain or improve its governance performance over time.

For example, if a company currently exhibits medium compliance with governance criteria, the stationary distribution can indicate the probability of this compliance level improving to high or dropping to low over time. This information is crucial for stakeholders to make informed decisions regarding long-term strategies and investments.

The insights gained from the stationary distribution can be used to refine governance strategies. Companies can identify areas where interventions are necessary to shift the probability distribution towards more desirable states, such as high compliance or exemplary governance. By focusing on the most impactful indicators, companies can drive long-term improvements in governance and overall performance.

The simulation process involves the following steps:

**Step1. Initialization:** Define the initial distribution of the company's governance state based on current performance indicators.

**Step2. Transition matrix estimation:** Estimate the transition probabilities between different states of governance using historical data.

**Step3. Simulation:** Run multiple iterations of the Markov chain model to simulate the evolution of governance practices over time.

**Step4. Stationary Distribution Calculation:** Calculate the stationary distribution to assess the long-term probabilities of being in each state.

#### 4. Results and Discussion

The governance performance indicators of a major Moroccan company operating in the logistics and technology distribution sector were analyzed over the period from 2013 to 2023. The indicators utilized for this analysis included Return on Investment (ROI), Net Profit Margin, Return on Equity (ROE), report frequency, employee and customer satisfaction, successful audits, risk management, carbon emissions reduction, and renewable energy usage. The raw data collected is presented in Table 1.

**Table1 :** Raw governance performance Data (2013-2023)

Year	ROI (%)	Net Profit Margin (%)	ROE (%)	Report Frequency	Clarity Score	Employee Satisfaction (%)	Customer Retention (%)	Successful Audits	Risk Management Audits	Carbon Emissions Reduction (%)	Renewable Energy Usage (%)
2013	8.2	14.5	10.1	4	8.1	78.4	70.3	8	4	3.5	7.1
2014	7.8	15.0	9.8	3	7.8	82.1	68.9	9	5	4.2	8.5

2015	8.1	15.2	10.3	3	8.4	80.7	72.5	7	4	4.0	6.9
2016	8.4	15.7	9.9	4	7.9	75.3	75.2	9	5	4.5	7.3
2017	8.0	14.8	10.5	3	7.6	77.6	78.4	8	4	4.3	8.1
2018	7.9	15.5	10.2	3	8.0	80.1	76.0	8	3	4.1	7.7
2019	8.3	15.1	9.7	4	8.2	83.2	79.5	9	4	4.8	8.9
2020	4.3	8.1	5.2	2	6.1	59.3	51.6	6	3	2.5	4.2
2021	3.9	7.5	5.0	1	5.6	53.7	47.9	5	2	2.0	3.5
2022	7.1	12.3	7.8	3	6.9	65.4	62.7	7	3	3.5	6.1
2023	7.5	12.0	8.1	2	7.4	71.2	66.8	8	4	4.0	7.2

To enable a more rigorous assessment of governance performance, these raw data were converted into scores on a scale from 1 to 5, based on predefined benchmarks for each indicator (Table 2). This transformation facilitates a clearer comparative analysis across the years and aids in identifying trends.

**Table 2: Governance Performance Scoring Criteria**

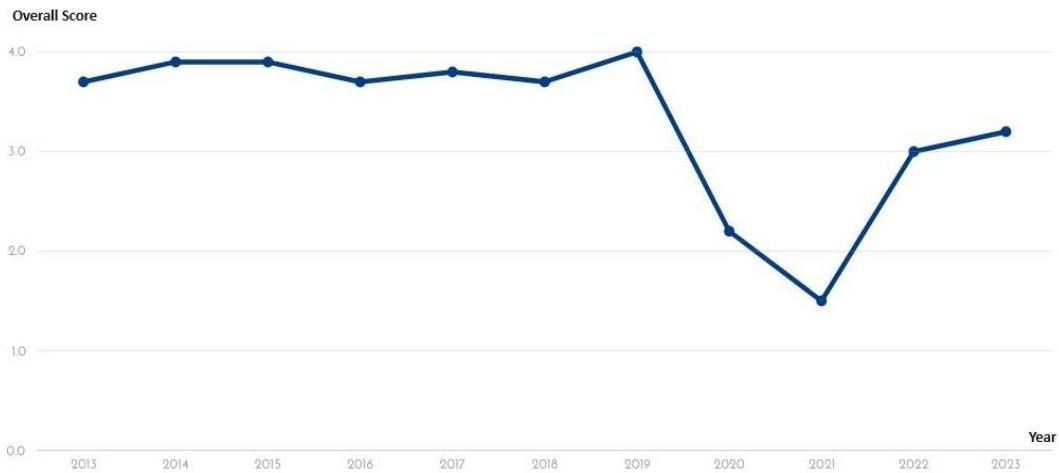
Indicator	Score 1	Score 2	Score 3	Score 4	Score 5
<b>Return on Investment (ROI %)</b>	ROI ≤ 5%	5% < ROI ≤ 6.5%	6.5% < ROI ≤ 8%	8% < ROI ≤ 9.5%	ROI > 9.5%
<b>Net Profit Margin (%)</b>	Net Profit Margin ≤ 8%	8% < Net Profit Margin ≤ 10%	10% < Net Profit Margin ≤ 12%	12% < Net Profit Margin ≤ 14%	Net Profit Margin > 14%
<b>Return on Equity (ROE %)</b>	ROE ≤ 5%	5% < ROE ≤ 7.5%	7.5% < ROE ≤ 10%	10% < ROE ≤ 12.5%	ROE > 12.5%
<b>Report Frequency</b>	1 report/year	2 reports/year	3 reports/year	4 reports/year	≥ 5 reports/year
<b>Clarity Score</b>	Clarity Score ≤ 5	5 < Clarity Score ≤ 6.5	6.5 < Clarity Score ≤ 8	8 < Clarity Score ≤ 9.5	Clarity Score > 9.5
<b>Employee Satisfaction (%)</b>	Satisfaction ≤ 55%	55% < Satisfaction ≤ 65%	65% < Satisfaction ≤ 75%	75% < Satisfaction ≤ 85%	Satisfaction > 85%
<b>Customer Retention (%)</b>	Retention ≤ 55%	55% < Retention ≤ 65%	65% < Retention ≤ 75%	75% < Retention ≤ 85%	Retention > 85%
<b>Successful Audits</b>	0 to 2 successful audits	3 to 4 successful audits	5 to 6 successful audits	7 to 8 successful audits	≥ 9 successful audits
<b>Risk Management Audits</b>	0 to 1 risk management audits	2 risk management audits	3 risk management audits	4 risk management audits	≥ 5 risk management audits
<b>Carbon Emissions Reduction (%)</b>	Reduction ≤ 2%	2% < Reduction ≤ 3.5%	3.5% < Reduction ≤ 5%	5% < Reduction ≤ 6.5%	Reduction > 6.5%
<b>Renewable Energy Usage (%)</b>	Usage ≤ 5%	5% < Usage ≤ 6.5%	6.5% < Usage ≤ 8%	8% < Usage ≤ 9.5%	Usage > 9.5%

Based on these criteria, the performance scores over the 2013-2023 period are presented in Table 3, which reflects the annual governance performance and enables a comprehensive analysis of trends over the decade.

**Table 3:** Governance Performance Scores (2013-2023)

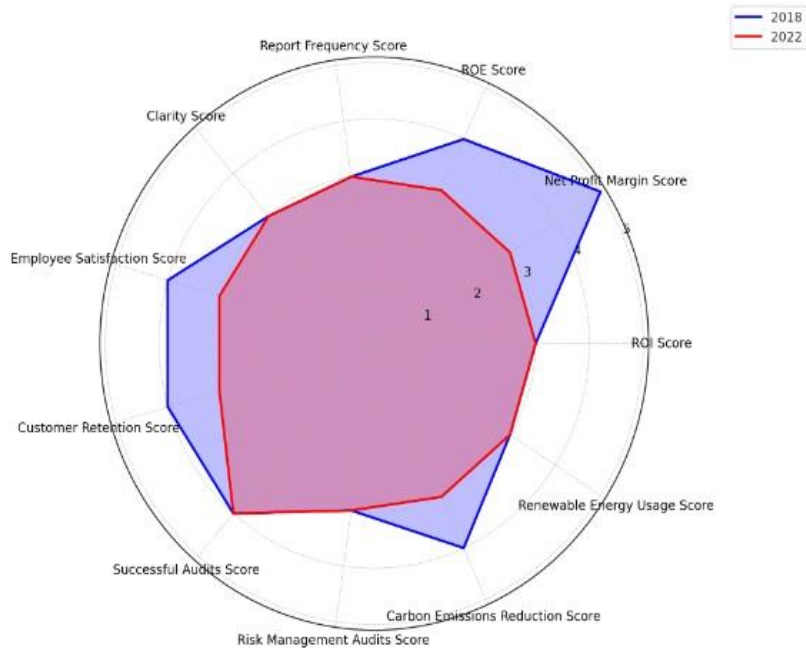
Year	ROI (%)	Net Profit Margin (%)	ROE (%)	Report Frequency	Clarity Score	Employee Satisfaction (%)	Customer Retention (%)	Successful Audits	Risk Management Audits	Carbon Emissions Reduction (%)	Renewable Energy Usage (%)
2013	4	5	4	4	3	4	3	4	4	3	3
2014	3	5	3	3	3	4	3	5	5	4	4
2015	4	5	4	3	4	4	4	4	4	4	3
2016	4	5	3	4	3	3	4	5	5	4	3
2017	4	5	4	3	3	4	4	4	4	4	4
2018	3	5	4	3	3	4	4	4	4	4	4
2019	4	5	3	4	4	4	4	5	4	5	4
2020	2	2	2	2	2	2	2	3	3	2	2
2021	1	2	2	1	2	1	1	2	2	1	1
2022	3	4	3	3	2	3	3	4	3	3	3
2023	3	4	3	2	3	3	3	4	4	4	3

The graph (Figure 1), which illustrates the evolution of global scores from 2013 to 2023, clearly depicts a significant dip in overall scores during 2020 and 2021, largely due to the impact of COVID-19, followed by a recovery in 2022. The overall score, which consolidates performance across various indicators, reveals a general upward trend from 2013 to 2019, culminating in a peak in 2019. This makes 2019 a benchmark year for optimal performance before the pandemic-induced decline.

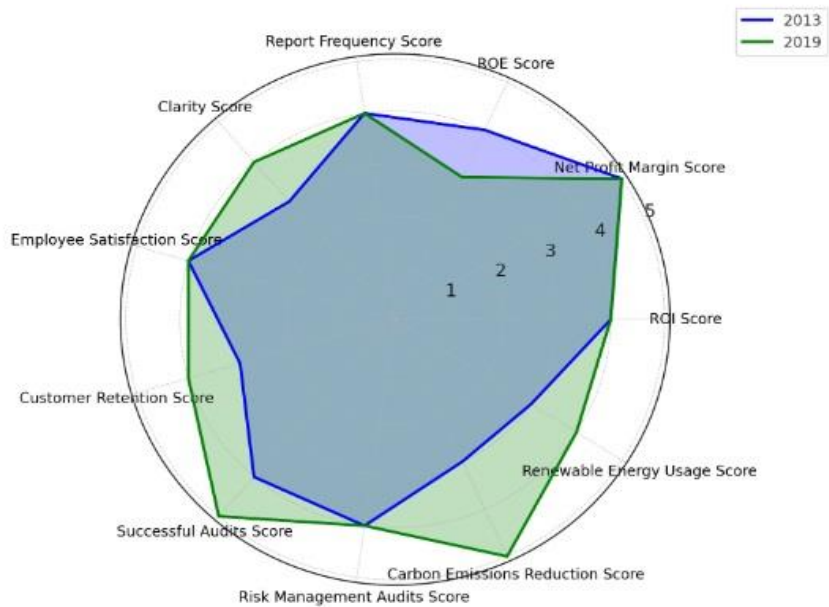


**Figure 1:** Global scores over time.

The radar charts (Figure 2 and Figure 3), comparing the company's governance scores between 2018 and 2022, and 2013 and 2019, respectively, highlight the variations in performance across key indicators. These visualizations enable the identification of areas where the company has made progress or faced challenges over these periods.



**Figure 2:** Comparison of governance scores (2018 vs 2022).



**Figure 3:** Comparison of governance scores (2013 vs 2019).

The analysis of Global Scores and the application of Markovian modeling provide a deep and evolving understanding of corporate governance performance over time. By defining discrete governance levels,

ranging from "Weak" to "Exemplary," this evolutionary approach allows us to model the probable transitions between these states through a Markov chain. Each state in the governance space represents a specific performance level, and the transition matrix (Table 4), which we estimated from historical data, reflects the probabilities of movement between these states.

**Table 4 :** Transition matrix.

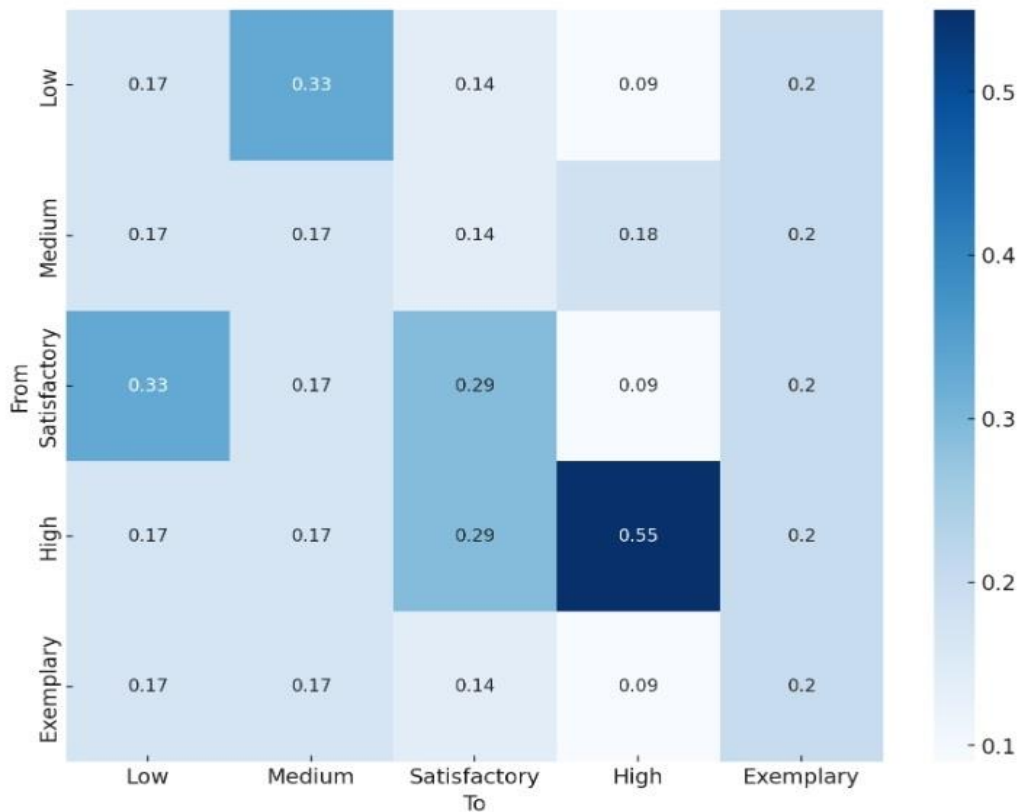
<b>From \ To</b>	<b>Low</b>	<b>Medium</b>	<b>Satisfactory</b>	<b>High</b>	<b>Exemplary</b>
<b>Low</b>	0.17	0.33	0.14	0.09	0.20
<b>Medium</b>	0.17	0.17	0.14	0.18	0.20
<b>Satisfactory</b>	0.33	0.17	0.29	0.09	0.20
<b>High</b>	0.17	0.17	0.29	0.55	0.20
<b>Exemplary</b>	0.17	0.17	0.14	0.09	0.20

This analysis is further reinforced by the simulation of the Stationary Distribution that we calculated with the *Visual Markov Chain Simulator* (Bounnite and Nasroallah (2020)), which reveals the long-term probabilities for the company to remain in each of the governance states.

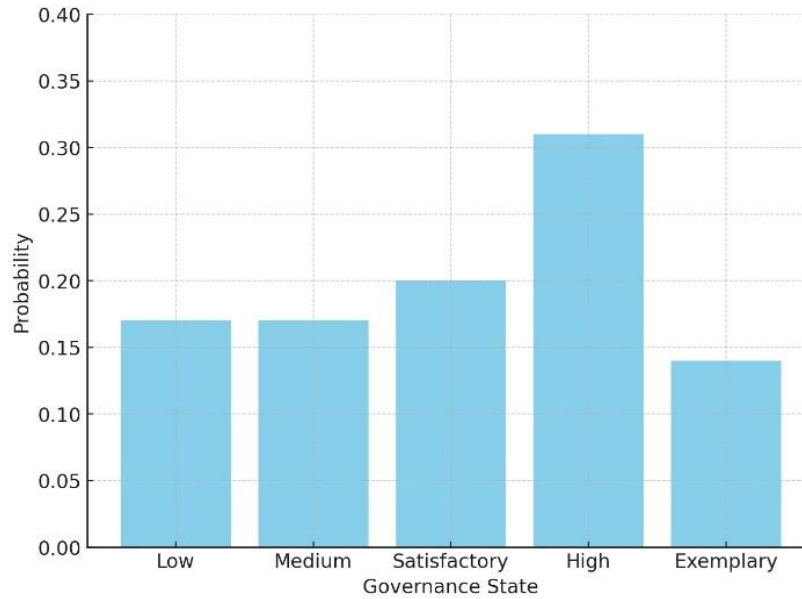
Figure 4: Transition Matrix Visualization, represents the transition probabilities between different governance states, as captured in the transition matrix. This matrix quantifies the likelihood of moving from one governance state to another over successive periods. Each row corresponds to a current state, and each column represents a future state, with the color intensity or size of the points indicating the magnitude of the transition probabilities. The visualization reveals several key dynamics within the enterprise's governance evolution. The relatively high probability of remaining in the "High" governance state (0.55) underscores the stability of enterprises that achieve strong governance practices. Conversely, the transitions from the "Low" governance state are more evenly distributed, suggesting a less predictable progression path, which may indicate challenges in governance improvement for companies starting from a weak position. The "Exemplary" state, while desirable, has a consistently low probability of being reached directly from lower states, emphasizing the difficulty of achieving and maintaining exemplary governance standards. This graph provides a comprehensive overview of the dynamics at play, highlighting both the stability of certain states and the challenges in transitioning to higher governance levels.

In addition, the distribution of governance status is shown in Figure 5, illustrating the long-term probabilities of the enterprise residing in each governance state once equilibrium is achieved. This distribution is derived from the Markov chain model and provides a predictive outlook on the steady-state behavior of governance practices. The stationary distribution graph indicates that the most likely long-term outcome is for the enterprise to maintain a "High" governance state, with a probability of 0.31. This suggests that once strong governance is established, it is likely to be sustained over time. However, the probability of the enterprise remaining in an "Exemplary" governance state is lower

(0.14), reflecting the challenges in not only achieving but also sustaining the highest level of governance. The presence of non-negligible probabilities in the "Low" (0.17) and "Medium" (0.17) states implies that some companies may struggle to improve their governance practices, remaining in lower states over time. This graph provides critical insights into the expected long-term distribution of governance performance, informing strategic decisions aimed at improving or maintaining high governance standards.



**Figure 4:** Transition matrix



**Figure 5:** Stationary distribution of governance states

## 5. Conclusion

This article presents a thorough synthesis of key findings and future directions in corporate governance. By analyzing governance performance over the past decade, we have identified significant trends in both financial and operational metrics, as well as in non-financial areas such as stakeholder satisfaction and environmental responsibility.

The use of Markovian modeling has provided a dynamic perspective on governance practices, revealing historical patterns and potential future scenarios for companies. The transition matrix and stationary distribution highlight the relative stability of firms with high governance standards, while also illustrating the challenges faced by those striving to achieve and maintain excellence. These findings emphasize the value of integrating both static and dynamic evaluations for effective governance management.

In summary, this study offers a robust analytical framework applicable to various organizational contexts, aiming to improve the understanding and management of governance issues in a complex environment. It lays the groundwork for future research, particularly on the impact of strategic interventions on governance transitions. Future work could adapt this model to different sectors and investigate the interplay between governance and long-term corporate performance, considering external factors like regulatory changes and societal shifts. This research thus makes a significant contribution to the literature on corporate governance, enhancing the evaluation methods and analytical tools available to researchers and practitioners alike.



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