

## Article

# Analysis of Changes in Innovative Management of Global Insurers in the Pre- and Post-COVID-19 Eras

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**Abstract:** This study aims to determine the changes in the innovative management of global insurers in the pre- and post-COVID-19 eras. The keywords of the CEO messages included in the sustainability reports and annual reports of 91 global insurers were used to perform network analysis based on text mining. Specifically, (1) words related to innovation as emphasized in the CEO message of global insurers were extracted, (2) keywords related to innovation in the past three years, of 2018–2020, were compared, and (3) the characteristics of innovative management were analyzed based on the connotations in these words and their structural interpretation. The results showed that the keywords emphasized by the CEOs varied each year; “market” in 2018, “risk” and “value” in 2019, and “people” in 2020. Furthermore, mapping the open innovation success factors based on the keywords extracted according to eigenvector centrality showed that the key factors exhibiting the highest centrality represented the process in 2018, the provision of resources in 2019, and governance in 2020, indicating changes in innovative management. The significance of this study lies in suggesting a directionality of change in post-COVID-19 innovative management for stakeholders, clients, and employees of global insurers.

**Keywords:** innovation management; CEO message; network analysis; text mining; global insurance industry



**Citation:** Kim, H.J.; Cho, K.T.

Analysis of Changes in Innovative Management of Global Insurers in the Pre- and Post-COVID-19 Eras. *Sustainability* **2022**, *14*, 9976.

<https://doi.org/10.3390/su14169976>

Academic Editors: Alessandro Pagano, Federica Murmura and Elisabetta Savelli

Received: 14 July 2022

Accepted: 7 August 2022

Published: 12 August 2022

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## 1. Introduction

In the New Normal era, encompassing the Fourth Industrial Revolution and the COVID-19 pandemic, the insurance industry, like any other, seeks to transform its management strategies and business directions [1]. The New Normal signifies a phenomenon that has been perceived as abnormal in the past, gradually increasing in popularity and undergoing generalization towards replacement with a new standard [2]. This causes general social reformations and transitions. In economics, it facilitates the conversion to a digital economy whereby non-face-to-face business models and online services would be placed at the core [3].

Various studies have been conducted regarding innovations in the insurance industry, such as the study on the business innovation of digital insurance in the crisis era conducted by analyzing the product, process, organization, and business models [4], and the study on the propagation of marketing innovation in the service industry by reviewing prior cases in the insurance industry (1700–1914) [5]. Additionally, notable findings were reported by a study analyzing the digital trends in the post-COVID-19 insurance industry [6] and a study on InsurTech and digital insurance platforms based on a case of successful digitalization at Ping An Insurance in China [7]. These studies offered insights regarding the necessary management strategies and leadership in the digital era to Asian insurers in active preparation to adopt InsurTech and digital insurance platforms.

It is essential to establish the vision of the insurance company's CEO and reinforce their integrated capability to ensure sustainability [8,9] to adapt to changing business environments and respond to new market conditions. A CEO message containing the CEO's thoughts affects the organizational culture and has a great impact on the social science field [10]. Therefore, the CEO provides a message regarding the status of the company and future strategic initiatives to the company members through a sustainability report (SR). The SR thereby provides a roadmap for sustainable management while being the driving force for strategic communication by the company [10–13].

Studies on CEO messages have been conducted in diverse fields, from finance and accountancy to construction, environment, and car engineering [14–23]. Methods based on text mining are being actively developed and applied. The text mining method was used in Bang [24] to analyze CEO letters at a company in Germany, and in Uhm [25] to analyze the trends of environmental, social, and governance (ESG) management strategies. Kim and Kim [26] surveyed the CEOs of 42 insurance companies in South Korea to determine their converged opinions regarding environmental changes. The results showed that the CEOs constructed long-term strategies to respond to the rapid ongoing changes in socio-economic conditions after the COVID-19 pandemic, with increased interest in fulfilling the insurance industry's social responsibility. The CEOs claimed that insurers should maintain a suitable balance between long-term growth and response to current issues.

Thus, this study identifies the changes in the innovative management of global insurers by analyzing the CEO's messages, an aspect that has been neglected in previous studies. Notably, the effects of CEO messages on the innovative management of insurance companies were analyzed concerning the COVID-19 pandemic in 2019. Therefore, the CEO messages of global insurers in the period of 2018–2020 were comparatively analyzed to identify changes in the keywords. The results were used to derive a directionality of innovative management for how global insurers should respond to and prepare for the New Normal post-COVID-19 era. The findings in this study are expected to be the benchmarked data regarding the CEO's vision and innovative innovation for insurers.

The flow of this study was as follows: First, the CEO messages of global insurers were collected, the keywords were extracted based on the criteria of eigenvector centrality, and the significance of each keyword was determined. Second, the CEO messages of global insurers in the past three years (2018–2020) were collected to comparatively analyze the keywords of CEO messages being emphasized for actual management conditions. Considering that the COVID-19 outbreak occurred in 2019, CEO messages were collected in the pre- and post-COVID-19 eras. Third, the extracted keywords were mapped with the nine reported success factors of open innovation [27]. The network was visualized to analyze the significance and structure of the changes in innovative management.

This paper is organized as follows. Section 2 analyzes the characteristics of the fields investigated in previous studies, and the unique significance of this study is presented. Section 3 describes the data collection, treatment, and analysis from text mining to network analysis and factor mapping. In Section 4, the list of keywords classified according to weight in the centrality analysis and the results of comparing the keywords in the past three years are presented through visualization based on data analysis. Section 5 describes keywords, trends compared across previous studies, and the network structure. Section 6 discusses the limitations of this study and future directions.

## 2. Literature Review

### 2.1. Trend of Research on CEO Messages

CEO messages have a critical impact on the overall socioeconomics of companies as well as on company employees. Therefore, the study of CEO messages has been conducted in the area of company finance. These studies have predicted future finance through the qualitative analysis of descriptive data [14,15] and identified the characteristics of insolvent and going-concern companies by describing security [16,17]. Studies have also suggested ways to prevent financial statement fraud through CEO messages [18,19] and analyze the

CEO message with the quantitative data of company performance using the text mining method for Korean and Japanese chemical businesses [20].

Studies on CEO messages have been conducted in finance and other industrial fields. For example, in tourism, post-COVID-19 CEO messages were analyzed for hospitality businesses (hotels, tours, travel, etc.) [28]. Im et al. [28] analyzed the CEO messages of 57 hospitality companies during the COVID-19 pandemic to detect management strategies. In the environmental field, the ESG-related social responsibility of the CEO in the message and sustainability outcomes have been investigated [20]. Yook [20] applied text mining to analyze the association between the scale of environmental accounting and the CEO message found in the SR or the Corporate Social Responsibility (CSR) report across Korean and Japanese companies. The study showed that the companies with large expenditures related to the environment displayed an interest in CSR activities, that is, problems related to society and the environment. However, companies with small expenditures assessed sustainability-related activities concerning economic factors, such as growth, development, values, and efforts. Other notable studies include the one analyzing CEO communication in order to investigate sustainable organizational strategies [29] and the association between the CEO's message on social responsibility and sustainability outcomes [22].

Text mining has been applied in several studies on CEO messages. In Bang [24], the text mining method was used to analyze the theme of the CEO's message to the shareholders of a German company in each industry. In Uhm [25], the text mining method was used to analyze the changes in the trends of ESG management strategies based on CEO messages. Choi and Cho [21] collected CEO messages from advanced construction companies and used text mining to extract keywords emphasized by the CEO. This study reported on safety management characteristics through a visualization based on network analysis. In addition, Choi and Cho [23] extracted environment-related words emphasized in the CEO messages of advanced automobile companies and compared the tendency of keywords related to the environment, as emphasized by companies with higher and lower performance. Thus, the study showed the differences in the characteristics of environmental management across companies based on the connotations and structural interpretation of the analyzed keywords.

Furthermore, Park [30] performed an automatic text analysis of the New Year's message of the government of North Korea, one of the most important and authoritative documents published by the government, to analyze the changes in North Korea's domestic and foreign policies. Yun et al. [31] extracted and analyzed keywords in unstructured data, such as the CEO speech and greeting, to measure a company's CSR directionality and scale. Yook [22] analyzed the linguistic and syntactic characteristics of CEO messages and reported that readability decreased and positive expressions were used more frequently with a stronger emphasis on future outcomes when the company showed lower sustainability outcomes.

Numerous studies on CEO messages have been conducted across diverse industrial fields, while only a few have analyzed the CEO messages in the insurance industry using the text mining method. In addition, few previous studies have analyzed only the CEO messages of the year the study was conducted. However, this study analyzed and defined the changes in innovative management in the insurance industry based on the CEO messages of the past three years.

## 2.2. Trend of Research on Innovative Management

Studies on innovative management have extended to diverse fields, from the definition of open innovation and its methods and innovativeness to company performance concerning open innovation. Open innovation methods involve actively utilizing ideas and technologies from external sources and transferring internal technical resources to enhance the company's performance [32,33]. Open innovation presents a new paradigm in using these methods to maximize values and innovativeness, including new internal and external technologies and intellectual properties. Hence, open innovation activities have

become a critical strategy and means of differentiation for companies to ensure continuous competitive superiority [34].

The continuous emphasis on innovative activities to reinforce the competitiveness of a company [35] has been shown to positively impact innovation and network capacities [36]. Kim and Kim [37] verified the effects of the success factors of innovation and innovation capacity on performance in manufacturing companies. This study applied CEO attitude, support, cooperation, and culture as the success factors of innovation, and R&D, production, and learning capacities as variables of innovation capacity. The results of the study showed that despite the similarity of success factors of innovation at different companies, varying outcomes were obtained depending on the innovation capacity of the company. Jang et al. [38] highlighted the importance of reinforced awareness of innovation so that various innovative activities in manufacturing companies could lead to meaningful company management performance. The study also claimed that there should be continuous and actual adherence to various educational training, and suggested policies to elicit staff participation within the organization to reinforce the awareness of innovation.

Furthermore, the characteristics of managers were reported to have an impact on the cause–effect relationship between innovative and management outcomes. According to Chang and Jang [39], diverse abilities and systematic management activities are essential for a manager to achieve an innovative outcome. However, the manager’s characteristics are critical in the company’s opportunities for growth and development. Hong et al. [40] surveyed innovative small to medium and venture companies to empirically analyze the effects of CEO management style, innovative directivity, and innovative activities on management outcomes. Thus, the results verified the association between CEO management style and management outcomes based on the cognitive process of the innovative directivity of the CEO and innovative activities by the staff.

There are many studies on the relationship between corporate management strategy, organizational culture, and management innovation. Lee [41] studied the relationship between corporate management strategy, organizational culture, and innovation. According to the results, strategy had a significant effect on innovation, whereas organizational culture did not. He suggested that innovation had a significant effect on management performance. Faruk Kalay and Gary Lynn [42] studied the impact of a strategic innovation management method on corporate innovation performance. As a result of the analysis, it was found that innovation strategy, organizational structure, and innovation culture significantly increased corporate innovative performance. Köse et al. [43] studied a case study of an extended interactive innovation management model based on a Turkish insurance company. Based on the results, the company concluded that a series of educational activities such as customer management, communication, and marketing should be carried out to improve employees’ capabilities. It was suggested that the innovation management model proposed to the company be continuously applied in the company’s digital innovation office in consideration of future policies and strategic goals. Numerous studies on innovative management have been conducted, on topics ranging from open innovation to innovativeness. In addition, the effects of innovation capacity on company performance were also investigated. Thus, these studies contribute to defining the current flow of innovative management. However, there is a general lack of studies analyzing the association between innovative management and CEO messages that could represent decision-making by the CEO of an insurance company through text mining.

Thus, this study applies text mining and performs a network analysis to analyze CEO messages and determine the changes in innovation management in insurance.

### *2.3. Uniqueness of the Present Study*

Most studies on the characteristics of CEO messages have been conducted in finance, accounting, and the environmental field. These studies focused mainly on keywords rather than the correlations among them and identifying the sentence properties and patterns. In the insurance industry, studies applying text mining include long-term care insurance for

senior citizens through keyword network analysis [44]; however, only a few studies have investigated CEO messages from the perspective of innovative management. Most studies on innovative management have been conducted using surveys or statistical analysis in the manufacturing industry, with few studies performing network analysis using keywords extracted using text mining from CEO messages.

Compared to previous studies, this study has the following characteristics and significance: (1) the analysis of CEO messages applied the perspective of innovation management to the insurance industry; (2) instead of using the conventional methods of performing a survey or interview, the data of CEO messages reflecting the most recent opinions were collected, and the changes and characteristics of innovative management were comparatively analyzed; (3) the data collection and analysis involved the most recent data from the SR and annual report (AR) so that the latest trend could be identified based on the CEO messages of global insurers; and (4) the findings may serve as practical data regarding the directionality of management strategies and visions according to the changes in the innovative management of global insurers.

### 3. Materials and Methods

#### 3.1. Data Collection

Company selection was based on the Brandirectory list of the top 100 insurance companies of 2020 that reported the annual company rankings [45]. Among the 100 insurance companies (2020), the top 1–100 ranks were selected, and their CEO messages were obtained from their SRs and ARs. Brandirectory is an institution that investigates and reports the company's sales and reputation in each industry, quoted in the reports of prominent consulting companies, such as Deloitte. In addition, Marsden et al. [46] identified the top 100 global financial brands using the Banking 500 league table of Brandirectory. The CEO message in each company's SR and AR is an official message that emphasizes its performance, visions, and policies in the respective year [47]. As the latest year in this study was 2021, SRs or ARs presenting each company's performance for 2020 was selected, and the top 100 companies were selected accordingly. However, publishing these reports is not a legal requirement, so CEO messages were only collected from companies that had published the reports. Thus, the CEO messages were obtained from the SRs and ARs that were published in 2021 and reflect each company's performance in 2020.

We visited the website of each company and downloaded the necessary reports to collect the SRs and ARs of the top 100 companies. Therefore, CEO messages were obtained from 91 reports: SR ( $n = 44$ ) and AR ( $n = 47$ ). Most reports were in English and the native language of the respective country. However, the reports of three companies were written in Hindi, German, and Spanish. Four reports were without a CEO message and two were in secure PDF formats to prevent conversion to TXT formats; nine reports were excluded. The obtained reports were in PDF or HTML format, so they were converted to TXT format to extract the CEO message. Considering the COVID-19 outbreak in 2019, CEO messages for 2018–2020, including the three years in the pre- and post-COVID-19 eras, were collected and analyzed. The flow of the study is shown in Figure 1.

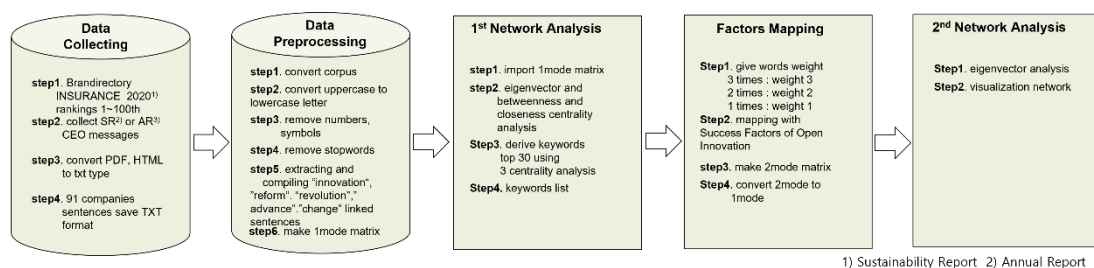


Figure 1. Research procedure and methods.

### 3.2. Data Preprocessing

The data collected in the TXT format are unstructured data composed of text. The data should be pre-processed in order to perform an analysis based on text mining [48]. For the data pre-processing in this study, the R statistical computing software (version 4.0.2) and the NLP-based package were used. Next, complete sentences were divided into words to generate a corpus. For “Innovation” and “innovation”, lowercase letters were used in both instances, as they were identical. Spaces, senseless articles, and prepositions (“a”, “an”, “of”, “the”, and “for”), numbers, and signs without any effects were removed. The names of the top 100 companies, country names and meaningless words (“AIA”, “China”, “year”, etc.) were removed. Analogous terms such as “COVID”, “COVID-19”, and “coronavirus” were extracted and turned into a single word, “COVID”. From complete sentences, those containing the word “innovation” and those containing words analogous to “innovation” were extracted to widen the range of collected words. The words “change” and “advance” were considered to be analogous to “innovation” in Lee [49]. Lee and Yi [50] mentioned that “change”, “reform”, and “revolution” shared an analogous concept with “innovation”. Thus, the sentences containing the word “innovation” or the similar words “change”, “reform”, “revolution”, and “advance” were extracted, and a single TXT file was generated. The keywords and frequency data were extracted to generate the comma-separated values (CSV) in a data frame format.

### 3.3. First Network Analysis

The method of analyzing sentences and their contents is referred to as language network analysis, in contrast to the conventional text analysis, where only the word frequency is analyzed [51]. The analysis, thus, allows for the extraction of concepts that reflect various intrinsic characteristics of a message expressed in texts, the identification of semantic relationships among words, and the quantification of relationships across specific concepts [51].

A network is a structure of connected nets that indicates the interactions among relational attributes [52]. The node referring to the agent indicates the words, and the link indicates the relationship between two relational words [53]. In conventional text analysis, the basic relationships and structures can be identified [54], and the form of analysis is identical to social network analysis (SNA) [53]. However, directionality is excluded in the analysis of messages because the words are interconnected [55].

Density and centrality are the most frequently used analyses to identify the influence of network structures and components [56]. In density analysis, the connection between the nodes in the network is analyzed, and a high-density network exhibits a rapid distribution of data as data exchange is activated [57]. In contrast, centrality analysis is the most widely used quantitative analysis to identify the keywords of the network [52]. Gephi version 0.9.2 was used in this study, as it is widely used across studies based on the great visualization function and the open-source centrality analysis, whereby the node importance is defined, and the key nodes are determined and ranked [58].

In centrality analysis, the types of centralities are as follows: degree, betweenness, closeness, and eigenvector [59]. For degree centrality, the centrality of a specific node at the center of the network was measured [52]. A node showing many degree lines has a high degree of centrality because it indicates the exchange of popularity, interest, attention, needs, and demands with other nodes [52]. The calculation of degree centrality uses the maximum number of connectable lines of the degree. The degree centrality analysis defines data directionality, whereas degree centrality is classified into in-degree and out-degree [59]. The interpretation of in-degree and out-degree centrality indicates authority or popularity for the former and activity for the latter [60].

For betweenness centrality, the mediation of one node with other nodes in the network was measured [61]. While a node with centralized connections exhibits degree centrality, a node taking the position of a mediator among different nodes through only a few

lines of degree without popularity, authority, interest, or demands exhibits betweenness centrality [52].

For closeness centrality, the measurement is based on all degrees, from direct to indirect, identified by degree centrality [62]. The difference between closeness and betweenness centrality is that the latter shows the number of nodes mediated by a specific node in the network [52]. Hence, it is difficult to identify the influence of a given node solely based on closeness and betweenness centrality.

For eigenvector centrality, the node's centrality is used to measure its influence [59]. Moreover, in eigenvector centrality, not simply nodes in connection but also their weight and centrality in the vicinity are considered [59]. Thus, considering these, the analyses in this study applied betweenness, closeness, and eigenvector centralities.

### 3.4. Mapping of the Nine Success Factors of Open Innovation

Words were extracted to include the top 30 words with the highest levels of betweenness, closeness, and eigenvector centralities, and the centrality values were listed. Based on the three centrality results, the importance of the node was measured based on the weight given to the word as follows [21]: For instance, taking B: betweenness centrality, C: closeness centrality, and E: eigenvector centrality, the weight for  $B \cap C \cap E$  is 3, the weight for  $B \cap C$  or  $B \cap E$  or  $C \cap E$  is 2, and the weight for B or C or E is 1 [21].

After the list of words with weighted values was generated, the indices were mapped and classified to analyze the characteristics of the collected words. Among the success factors of open innovation, three that reflect the latest research trends were selected [27].

The mapping analysis in this study used the following nine success factors of open innovation suggested by Durst and Stähle [27]: (1) relational issues, (2) involvement of people in open innovation, (3) governance, (4) facilitators, (5) resource provisions, (6) strategy, (7) open innovation process, (8) leadership, and (9) culture. Considering the terms analogous to "innovation", sentences containing the words "innovation", "change", "reform", "revolution", and "advance" were extracted. They were then mapped by identifying the sense of these words, sentences, and contexts.

The nine success factors of open innovation suggested by Durst and Stähle [27] are defined as follows: (1) relational issues are the overall communication skills and trust; (2) people involved in the open innovation process are the participants with devotion and motivation; (3) governance is the mechanism and structure; (4) facilitators are the promoters; (5) resource provision is the distribution of time, workforce, and resources; (6) strategy is the tactics applied; (7) open innovation process is the process unique to open innovation in contrast to closed innovation; (8) leadership is the capacity of an organization related to leadership and management, and (9) culture is the unique organizational culture of open innovation in contrast to closed innovation.

The final factor mapping was confirmed through interviews with ten experts, including workers with a career length of ten years in open innovation at a large company, accelerators fostering start-ups for at least ten years, and investors who locate and review potential start-ups. The interview period was from 12 February to 4 March 2022. Factor mapping based on interviews with ten experts was undertaken to ensure the objectivity of the data. Each word was mapped onto three success factors for open innovation. For multi-mapping factors with four or more mapped factors, the priority was set based on eigenvector centrality to select the top three factors. The mapping data with set priority after multiple mappings were converted to 2-mode alignment to exhibit word relationships.

### 3.5. Second Network Analysis

This study aims to identify which of the nine success factors of open innovation exhibited the greatest influence in the past three years (2018–2020) and determine the variation in the characteristics of innovation management through network analyses. In the secondary network analysis, 2-mode alignment was performed on the keywords mapped with the nine success factors of open innovation, followed by conversion to a 1-mode

alignment. Hence, the 2-mode alignment with two entities in different rows and columns was converted to a 1-mode alignment to identify correlations with the success factors of open innovation. In addition, the entities in identical rows and columns were classified as a single entity, and the MMULT and TRANSPOSE functions in Microsoft Excel were applied [63].

Eigenvector centrality analysis was performed to determine the key nodes [64]. The top ten keywords from the data for three years (2018–2020) were comparatively analyzed based on eigenvector centrality. Networking and links to the key node were interpreted for the nine success factors of open innovation. The results were visualized using Gephi version 0.9.2 software [54]. Through eigenvector centrality analysis, the success factors of open innovation and the characteristics of their relationships are visualized.

Table 1 presents the results of this study. The words from the data for the three years from 2018 to 2020 were compared based on keyword analysis in Category 1. The words extracted from Category 1 were mapped to the nine success factors for open innovation in Category 2. A list of keywords with weight values was summarized.

**Table 1.** Research topics of the study.

Category	Subcategory	Topic	Content Analysis
1	1-a	Keyword Analysis	Comparison of the top 91 insurance companies in 2018
	1-b	Keyword Analysis	Comparison of the top 91 insurance companies in 2019
	1-c	Keyword Analysis	Comparison of the top 91 insurance companies in 2020
2	2-a	Nine Factors of Open Innovation	Comparison of the top 91 insurance companies in 2018
	2-b	Nine Factors of Open Innovation	Comparison of the top 91 insurance companies in 2019
	2-c	Nine Factors of Open Innovation	Comparison of the top 91 insurance companies in 2020

## 4. Results

### 4.1. Word-Based Network Analysis

The top 30 words with the highest betweenness, closeness, and eigenvector centralities, as shown in Tables 2–4, were extracted [21]. In all three years (2018–2020), the words “business”, “development”, “technology”, “company”, “growth”, “development”, and “product” were all linked with the top ten words. These are the top ten words in all centrality analyses; they are the most critical. Nevertheless, this study focused on keywords that varied according to year.

**Table 2.** List of centrality analysis results for deducing words in 2018.

Ranking	Word	Betweenness Centrality	Word	Closeness Centrality	Word	Eigenvector Centrality
1	business	4804.488063	business	0.588068	business	1
2	customer	3045.830799	development	0.560976	development	0.91335
3	development	3006.858327	customer	0.553476	customer	0.901219
4	technology	1993.420767	technology	0.541885	market	0.872153
5	market	1963.880573	market	0.539062	technology	0.856544
6	company	1843.926569	company	0.528061	growth	0.751172
7	management	1140.855644	growth	0.518797	product	0.703894
8	product	908.97019	product	0.512376	company	0.674082
9	growth	845.876406	management	0.509852	management	0.633944
10	risk	756.291668	service	0.498795	service	0.633375
11	value	746.141142	value	0.497596	value	0.608037
12	strategy	609.53948	industry	0.492857	industry	0.602606
13	service	577.47129	transformation	0.488208	transformation	0.597111
14	term	452.405025	environment	0.480278	capability	0.510997
15	data	431.08478	strategy	0.475862	strategy	0.496044
16	sale	428.188955	opportunity	0.469388	environment	0.486918



Table 2. Cont.

Ranking	Word	Betweenness Centrality	Word	Closeness Centrality	Word	Eigenvector Centrality
17	spirit	404.437356	capability	0.461024	opportunity	0.486013
18	transformation	333.93174	focus	0.45898	regulation	0.387869
19	world	297.28158	rate	0.457965	data	0.3806
20	operation	255.973072	structure	0.456954	sale	0.37222
21	environment	254.700688	sale	0.452954	partner	0.367701
22	process	248.690713	risk	0.45098	focus	0.365345
23	regulation	239.846801	system	0.45	rate	0.354932
24	client	237.350276	data	0.445161	risk	0.344397
25	channel	229.903887	interest	0.444206	structure	0.340511
26	effort	228.080061	impact	0.442308	client	0.337535
27	initiative	224.905099	regulation	0.441365	system	0.321511
28	culture	224.872637	channel	0.437632	channel	0.318865
29	partner	212.654668	model	0.433962	need	0.308033
30	sustainability	209	client	0.43215	effort	0.280447

Table 3. List of centrality analysis results for deducing words in 2019.

Ranking	Word	Betweenness Centrality	Word	Closeness Centrality	Word	Eigenvector Centrality
1	business	8371.367625	business	0.639752	business	1
2	customer	2659.814331	customer	0.544974	development	0.626061
3	technology	2099.376483	risk	0.540682	risk	0.625209
4	company	1926.713614	service	0.517588	customer	0.622442
5	risk	1681.044729	company	0.515	technology	0.609292
6	product	1396.166852	development	0.513716	growth	0.535411
7	development	1365.241474	growth	0.513716	service	0.513819
8	growth	1104.056215	product	0.512438	value	0.496417
9	industry	968.452601	industry	0.509901	product	0.478197
10	service	792.181964	technology	0.507389	company	0.473282
11	value	632.786426	transformation	0.506143	transformation	0.462239
12	experience	585.136041	value	0.503667	industry	0.461913
13	capability	540.519468	capability	0.491647	model	0.434704
14	governance	352.950319	model	0.47907	capability	0.418697
15	environment	316.448428	environment	0.469248	core	0.376881
16	issue	288.663233	society	0.46712	platform	0.345275
17	solution	283.951597	experience	0.466063	structure	0.315581
18	transformation	275.971188	structure	0.462921	strategy	0.300243
19	effort	261.444299	effort	0.453744	effort	0.272083
20	society	256.24405	platform	0.450766	environment	0.266141
21	system	226.090031	area	0.444924	experience	0.251166
22	rate	209.200767	core	0.443966	operation	0.244525
23	pace	205.277778	time	0.443966	data	0.240208
24	investment	195.02384	governance	0.44206	society	0.23769
25	country	175.610864	strategy	0.44206	advancement	0.236087
26	board	171.421597	leader	0.439232	empowerment	0.220581
27	world	160.416323	world	0.436441	high quality	0.216134
28	data	146.900778	policy	0.436441	project	0.210711
29	platform	129.365021	future	0.433684	future	0.197233
30	market	127.642152	impact	0.433684	point	0.195956

**Table 4.** List of centrality analysis results for deducing words in 2020.

Ranking	Word	Betweenness Centrality	Word	Closeness Centrality	Word	Eigenvector Centrality
1	customer	6971.114138	customer	0.52488	business	1
2	business	6605.773401	management	0.519873	management	0.998436
3	management	5801.812297	business	0.514151	customer	0.904793
4	company	4467.55627	company	0.514151	service	0.895345
5	technology	4315.287062	people	0.49696	development	0.895133
6	community	4134.016543	service	0.493213	company	0.808335
7	service	4031.550403	technology	0.489521	technology	0.792498
8	people	3862.612437	development	0.483728	people	0.770994
9	issue	3297.511231	risk	0.471861	risk	0.744962
10	development	2308.395574	system	0.471182	value	0.739666
11	opportunity	1950.489972	value	0.468481	opportunity	0.694328
12	market	1790.931952	opportunity	0.466476	capability	0.660783
13	system	1765.997475	product	0.46383	system	0.632842
14	sustainability	1742.206902	industry	0.462518	product	0.615687
15	employee	1696.12015	growth	0.461864	strategy	0.594059
16	product	1630.748383	strategy	0.459916	operation	0.590444
17	growth	1525.617263	issue	0.45927	industry	0.55721
18	risk	1485.531893	capability	0.456067	growth	0.5426
19	support	1404.88417	operation	0.454167	enhance	0.475768
20	impact	1312.170166	transformation	0.452282	transformation	0.453461
21	industry	1283.459258	time	0.449794	society	0.450704
22	society	1268.450378	society	0.439516	high quality	0.448089
23	strategy	1226.178128	model	0.435419	challenge	0.436487
24	value	1203.475365	enhance	0.433687	model	0.429346
25	culture	884.391275	challenge	0.433113	project	0.40401
26	area	854.328831	market	0.43254	time	0.403747
27	capability	823.765979	project	0.429698	issue	0.389959
28	effort	816.924729	effort	0.428571	team	0.364789
29	health	776.608341	high quality	0.424125	market	0.360719
30	transformation	704.124137	investment	0.420308	effort	0.359729

Table 2 shows that the word “risk” had a low eigenvector centrality but a high betweenness centrality in 2018. The words “partner”, “client”, and “channel” appeared within the top 30. The word “market” was among the top keywords indicating market growth and expandability in 2018. The growth of the expanded market could be seen through digitalization. Based on this, the potential growth in the market or extended scope could be considered. Notably, as part of digitalization, the process related to insurance payment could be introduced and improved using the banking software and data analysis upon the insurance payment review and insurance accident investigation [3]. This implies that enhanced digital services and channels within established systems could offer new experiences to clients.

Table 3 shows that the word “risk” had high betweenness, closeness, and eigenvector centralities in 2019, along with the newly emerged word, “value”. The words “platform” and “future” appeared within the top 30. The word “risk” indicates the emphasis on market regulations and the social, cultural, and economic risks. The appearance of the word “platform” predicted the activated insurance sales through various platforms, including online and mobile channels, while new and potential clients were expected to emerge [3]. Furthermore, the conversion of the work environment towards higher efficiency during the COVID-19 pandemic implied the start of the new system application regarding the employment process, work environment, and training policies to suit the telecommuting environment.

Table 4 shows that the word “people” emerged among the top keywords in 2020, compared to 2018 and 2019. The words “growth”, “challenge”, and “team” appeared within the top 30. In 2020, the word “people” emphasized the need for professionalism

regarding the motivation of the team, fluent communication among team members, and their collaboration capacity underlying innovation. With the increased demand for telecommuting due to the COVID-19 pandemic, the key human resources should be efficiently developed, maintained, and utilized. The resources should be distributed based on needs. The words that emphasized the change in the future work environment were linked.

#### 4.2. Weight Loading on the Words

According to the centrality results, the classification and analysis of the words with weight values are as follows: words ranked within the top 30 for the highest levels of betweenness, closeness, and eigenvector centralities were given a weight value of 3. Words ranked within the top 30 for the highest levels of the two centrality types were assigned a weight value of 2. Words ranked within the top 30 for the highest level of at least one of the three centrality types were assigned a weight value of 1 [21].

As shown in Table 5, a total of 19 words, including “business” and “management”, were in the Weight 3 clusters in 2018, while 10 words, including “industry” and “partner”, were in the Weight 2 clusters and 11 words including “process” and “mentality” were in the Weight 1 cluster. In 2019, a total of 17 words including “technology” and “advancement” were in the Weight 3 clusters, 10 words including “structure” and “strategy” were in the Weight 2 clusters, and 19 words including “investment” and “market” were in the Weight 1 cluster. In 2020, 22 words including “people” and “opportunity” were in the Weight 3 clusters, 7 words including “challenge” and “project” were in the Weight 2 clusters, and 10 words including “community” and “sustainability” were in the Weight 1 cluster.

**Table 5.** Weighted words classified according to the results of centrality analysis in 2018~2020.

	2018	2019	2020
Cluster	Word	Word	Word
Weight 3	business, development, customer, market, technology, growth, product, company, management, service, value, transformation, strategy, environment, data, sale, risk, client, channel	business, development, risk, customer, technology, growth, service, value, product, company, transformation, industry capability, platform, effort, environment, society	business, management, customer, service, development, company, technology, people, risk, value, opportunity, system, product, strategy, industry, growth, issue, transformation, society, market, effort, capability
Weight 2	capability, opportunity, regulation, partner, focus, rate, structure, system, effort, industry	model, core, structure, strategy, experience, future, governance, world, data, area	operation, enhance, challenge, high quality, model, project, time
Weight 1	need, spirit, world, operation, process, initiative, culture, sustainability, interest, impact, model	operation, advancement, empowerment, high quality project, point, issue, solution, system, rate, pace, investment, country, board, market, time, leader, policy, impact	team, community, sustainability, employee support, impact, culture, area, health, investment

#### 4.3. Mapping of the Words with the Success Factors of Open Innovation

The weighted words and success factors of open innovation were mapped according to the weighted cluster. The results are presented in Appendix A. Table A1 shows the correlations of the words in the Weight 3 cluster for 2018 and the nine success factors of open innovation. Among the nine success factors, governance, culture, and strategy were multi-mapping. Table A2 shows the correlations of the words in the Weight 2 cluster with facilitators, leadership, and strategy in multi-mapping. Table A3 shows the correlations of the words in the Weight 1 cluster, with people displaying a lack of mapping.

The results of mapping the weighted words for 2019 and the success factors of open innovation are shown in Tables A4–A6. Table A4 shows the correlations of the words in the Weight 3 cluster with all factors in mapping and relational issues, culture, and strategy in multi-mapping. Table A5 shows the correlations of the words in the Weight 2 cluster with culture and strategy in multi-mapping. Table A6 shows the correlations of the words in the Weight 1 cluster with relational issues, people, and processes in multi-mapping.

The results of mapping the weighted words for 2020 and the success factors of open innovation are shown in Tables A7–A9. Table A7 shows the correlations of the words in the Weight 3 cluster with all factors in mapping and governance, process, and strategy in multi-mapping. Table A8 shows the correlations of the words in the Weight 2 cluster, with facilitators and leadership being unmapped. Table A9 shows the correlations of the words in the Weight 1 cluster, with governance and process in multi-mapping and leadership being unmapped.

#### 4.4. Factor-Based Network Analysis and Visualization

After mapping the weighted words with the success factors of open innovation, eigenvector centrality analysis was performed to identify the correlations among the success factors, and the network was visualized [65]. The network for 2018 based on the eigenvector centrality analysis is presented in Figure 2. The key node is process, with strong links to the green nodes of governance and strategy. The network for 2019 is presented in Figure 3. The key node is provision of resources, with strong links to the purple nodes of relational issues, people, and culture. The network for 2020 is shown in Figure 4. The key node is governance, with strong links to the purple nodes of people, culture, and process. The peripheral node of all three years (2018–2020) was leadership, which displayed the weakest links.



Figure 2. Network visualization of open innovation characteristics in 2018.



Figure 3. Network visualization of open innovation characteristics in 2019.



Figure 4. Network visualization of open innovation characteristics in 2020.

5. Discussion

5.1. Keywords Extraction and Variation

The keywords exhibiting significant variations across the three years were extracted from the keyword analysis. As shown in Table 6, the new keywords linked within the top 10 words of the highest eigenvector centrality were “market” in 2018, “risk” and “value” in 2019, and “people” in 2020. Based on these keywords that newly emerged in the CEO message within the top 10 words of each year, the changes in the innovative management of global insurers could be examined. An example sentence containing the newly emerged keywords in the CEO message is as follows.

“There is a broad prospect for product and service innovation in line with the market trend . . . ”.

“to the operational guideline of “prioritizing business value, strengthening sales force, achieving stable growth, upgrading technology, . . . as our strategic core, making new strides in shouldering corporate social responsibility, quality development, technology empowerment, reform and transformation, and risk prevention and control . . . ”.

“We have a commitment to provide new value our customers and society expects from us . . . ”.

“We kept investing in digital innovations and foundational technologies, in people and in our businesses . . . ”

Table 6. Keywords list in 2018~2020.

2018		2019		2020	
Ranking	Word	Ranking	Word	Ranking	Word
1	business	1	business	1	business
2	development	2	development	2	management
3	customer	3	<b>risk</b>	3	customer
4	<b>market</b>	4	customer	4	service
5	technology	5	technology	5	development
6	growth	6	growth	6	company
7	product	7	service	7	technology
8	company	8	<b>value</b>	8	<b>people</b>
9	management	9	product	9	risk
10	service	10	company	10	value

As the example depicts, emphasis was placed on the importance of advancing into the market in 2018, before the COVID-19 outbreak. This indicates that a foundation was established to advance towards digitalization by applying digital services reflecting market needs. In addition, insurers responded to the flow of the market by applying digital technologies such as AI-based underwriting automation, and developing platforms that allow mobile insurance subscriptions. Therefore, the insurers that had entered the digital transition period were moving from the simple provision of insurance services towards

the use of digital technologies to provide risk prevention services, leading a paradigm shift [66].

With the outbreak of COVID-19 in 2019, the emphasis was on the importance and value of risk management. Durst and Stähle [27] claimed that resource provision included keywords such as risk, potential client search, environment, and value; these keywords were extracted in this study. This indicates that insurers created novel values according to client needs and delivered new values to clients against competitors to achieve client value innovation. The insurers establish a virtuous cycle that provides the driving force towards client values and, through innovation, strive to contribute to enhancing enterprise values. The insurers also increase their dependency on new technologies and data sources to reconsider efficiency, reinforce cyber security, and improve the overall capacity of the organization. Changes such as process simplification and client-tailored service provision through automation ultimately reflect the insurers' will to improve client experience [67].

After the COVID-19 outbreak, the focus was placed on the importance of people. The COVID-19 pandemic highlighted communication with employees and their participation, motivation, and creation of a new organizational culture. This indicates that the responsibility for decision-making based on project execution goals and issues has been emphasized [31]. In addition, the competition for employing able individuals deepened in the circumstances after the COVID-19 outbreak, posing challenges to acquiring human resources. The competition for recruiting individuals with strong IT development and data analysis capabilities was notably high, consistent with the digital transition. Deloitte [67] showed that insurers were faced with the challenge of discovering and fostering talented individuals, especially human resources, in technological fields, along with maintaining and advancing the digitalized operation system.

### *5.2. Characteristics of Innovative Management*

In this study, based on the success factor mapping of open innovation for each keyword with analyzed centrality, inferences were drawn regarding the trends, characteristics, and changes in emphasis in open innovation management. The finding in this study is that the digitalization target of the insurers moved beyond the application of the conventional processes and business models towards the overall digital transition over three years, consistent with Kim and Kim [26]. In addition, Kim and Kim [26] reported that CEOs gradually recognized the importance of ESG, which is also consistent with this study. However, this study was distinguished from Kim and Kim's [26] on the CEOs of Korean insurers, where the CEOs recognized the importance of social responsibility (S), emphasizing the three values of ESG.

Mapping the keywords and the success factors of open innovation showed that the emphasis was on the process in 2018. As Figure 2 shows, the process strongly links to governance and strategy. Governance indicates the mechanism and structure of the open innovation process, while strategy indicates the execution of the innovation process. This lent a degree of support to previous studies reporting that insurers strive to develop insurance packages by applying different techniques and understanding the technology life cycle [27,31]. It can thus be presumed that insurers perceived changes in market mechanisms and the start of digital services.

In 2019, the emphasis was on resource provision. As shown in Figure 3, the provision of resources is strongly linked to relational issues, people, and culture, in structural terms. The COVID-19 era appears to have highlighted the response to external risks and decision-making based on risks, social connections, economics, and partnerships, emphasizing the time and availability of resources for open innovation. In addition, this supports a previous study [27] that emphasized the balance between innovation and general tasks, communication skills, and organizational culture, including trust. Thus, insurers provide an environment to boost productivity, including flexible work options consistent with telecommuting due to the COVID-19 pandemic.

In 2020, the emphasis was on governance. As Figure 4 shows, governance exhibits strong structural links to people, culture, and process. Therefore, education and culture, such as networking and knowledge sharing, were highlighted in 2020. This indicated a phase of restructuring the workforce and organization and monitoring market fluctuations as a strategy to respond to the COVID-19 pandemic. The insurers recruited MZ-generation staff to replace the aging workforce with the long-term goals of human resource management. In addition, this coincided with a previous study [68] emphasizing more vigorous recruitment of individuals with advanced data analysis and automation technology skills to promote more rapid and effective digital innovation.

Furthermore, Governance is one of the ESG; “Environmental”, “Social”, and “Governance” structures, a factor related to the sustainable growth of enterprise [31]. This implied that insurers valued society and the environment and strived to provide positive impressions to clients. This indicates that, as in previous studies [69–71], the culture in which clients show preference toward products of companies that contribute to society has spread. In contrast, companies could enhance their values through positive reviews by stakeholders and investors. With the continuation of such a virtuous cycle of clients and companies, potential sales increases and sustainable growth are anticipated.

## 6. Conclusions

This study investigated official CEO messages through text-mining analysis of the SR or AR, rather than through a survey or interview of CEOs. Network analysis was performed on text mining and the mapping of innovation factors using the CEO messages of global insurers. Furthermore, the characteristics of innovative management were identified. The significance of this study lies in the extraction of keywords through the analysis of the CEO messages of global insurers, and in the detection of the changes in open innovation success factors in the past three years.

The implications of the results of this study are as follows. The CEO messages contain words related to the directionality of innovative management, and reflect the current circumstances faced by the company. Thus, this indicates the strong interest and will of each CEO in regards to innovation. The CEO’s message is significant in suggesting the directionality of innovative management to stakeholders, clients, and employees, as it contains diverse words directly connected to innovation. From a practical point of view, it can be used to provide data that can be used by stakeholders, customers, and executives and employees to understand various market contexts such as the digital environment and the impact of climate change after COVID-19. In addition, it can be used to provide monitoring data to lead the insurance industry for insurers that are aware of the importance of corporate social responsibility. The significance of this study is as follows. First, from a theoretical perspective, the actual and official CEO messages were examined through keyword and network analyses with visualized results, rather than through statistical quantification, to suggest a novel research method regarding the directionality of innovative management strategies. Second, from a practical perspective, insights to lead healthy financial culture were provided through financial innovation, the enhanced competitive edge in financial markets, and the continuous appeal of future growth engines. Finally, from a political perspective, insights were provided to resolve the issues of sales channel diversification and information asymmetry, consistent with the digital transition. The analysis of the CEO messages of insurers and the entire finance industry, including banks and stock markets, may allow the directionality of innovative management to be determined regarding the study of financial policies.

In this study, CEO messages (including letters and statements) in the SRs or ARs were collected for the top 100 insurance companies in 2020, as announced by Brandirectory in their annual report. Furthermore, CEO messages were collected from companies ranked within the top 100 in 2020 for three consecutive years over 2018–2020. Thus, a limitation of this study is that the company rankings in 2018 and 2019 were not reflected. There could have been companies that had entered or fallen out of the top 100 in these two years. In

addition, as the study investigated the last three years, while the COVID-19 pandemic persisted, future studies may extend the period for further investigation.

More studies should be conducted on CEO messages regarding innovation. First, CEO messages regarding innovation may be analyzed for the upper and lower groups of open innovation characteristics. Through this, the differences in perspectives on innovation between the upper and lower groups will be discovered. Based on the results, the directionality of innovation from market leaders can be discussed. Second, CEO messages of companies that have continuously maintained their top ranks may be analyzed. The insurance company rankings are updated each year by Brandirectory, with the change in ranking highlighted based on the rank of the previous year. Companies that have maintained the top ranks for three or more years can be presumed to be executing sustainability management. Analyzing the CEO messages of such companies will be significant, as the companies may be leaders in sustainability management and innovation. Finally, CEO messages may be analyzed by country and continent. Such studies will allow insurers to benchmark the directionality of strategy across different countries. Notably, a significant insight may be gained for developing the market advancement strategy of global insurers.

**Author Contributions:** Conceptualization, H.J.K.; methodology, H.J.K.; software, H.J.K.; validation, K.T.C.; formal analysis, H.J.K.; writing—original draft preparation, H.J.K.; writing—review and editing, H.J.K. and K.T.C.; supervision, K.T.C. All authors have read and agreed to the published version of the manuscript.

**Funding:** This research received no external funding.

**Institutional Review Board Statement:** Not applicable.

**Informed Consent Statement:** Not applicable.

**Data Availability Statement:** Not applicable.

**Conflicts of Interest:** The authors declare no conflict of interest.

## Appendix A

**Table A1.** Mapping list of safety performance factors and Weight 3 words in 2018.

Word	A	B	C	D	E	F	G	H	I
business		X				X	X		
development				X		X	X		
customer				X	X	X			
market			X	X					X
technology		X			X	X			
growth							X	X	X
product			X		X				X
company	X	X						X	
management			X		X				X
service			X		X				
value	X	X							
transformation	X				X		X		
strategy							X	X	X
environment					X	X	X		
data				X	X		X		
sale		X		X					X
risk	X	X			X				
client	X	X							X
channel		X					X		X

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.



**Table A2.** Mapping list of safety performance factors and Weight 2 words in 2018.

Word	A	B	C	D	E	F	G	H	I
industry			X		X				X
capability			X				X	X	
opportunity							X	X	X
regulation		X						X	X
partner	X			X			X		
focus					X		X	X	
rate	X	X							X
structure					X	X			X
system	X	X				X			
effort			X				X	X	

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.

**Table A3.** Mapping list of safety performance factors and Weight 1 words in 2018.

Word	A	B	C	D	E	F	G	H	I
need			X			X			X
spirit					X			X	X
world	X	X							X
operation			X		X	X			
process	X				X			X	
initiative			X					X	X
culture	X	X			X				
sustainability		X	X						X
interest		X						X	X
impact						X	X		X
model	X	X			X				

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.

**Table A4.** Mapping list of safety performance factors and Weight 3 words in 2019.

Word	A	B	C	D	E	F	G	H	I
business	X		X						X
development					X	X			X
risk	X	X	X		X				
customer		X		X	X				
technology	X				X		X		
growth					X		X		X
service			X	X					X
value	X			X				X	
product			X					X	X
company	X		X		X				
transformation			X		X				X
industry			X	X				X	
capability		X	X						X
platform				X			X		X
effort	X		X				X		
environment			X	X	X				
society	X			X	X				

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.

**Table A5.** Mapping list of safety performance factors and Weight 2 words in 2019.

Word	A	B	C	D	E	F	G	H	I
model				X			X		X
core						X	X		X
structure					X	X			X
strategy					X			X	X
experience			X		X				X
future			X		X				X
world	X		X		X				
area	X				X				X
data					X	X	X		
governance		X			X				X

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.

**Table A6.** Mapping list of safety performance factors and Weight 1 words in 2019.

Word	A	B	C	D	E	F	G	H	I
operation		X		X		X			
advancement			X				X		X
empowerment			X			X			X
high quality			X	X		X			
project				X		X			X
point				X		X			X
issue		X		X					X
solution		X		X					X
system			X		X	X			
rate		X	X			X			
pace			X		X			X	
investment	X		X						X
country		X		X					X
board		X				X		X	
market	X			X		X			
time			X	X	X				
leader				X	X			X	
policy		X	X					X	
impact		X		X	X				

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.

**Table A7.** Mapping list of safety performance factors and Weight 3 words in 2020.

Word	A	B	C	D	E	F	G	H	I
business	X	X	X						
management	X				X				X
customer				X		X		X	
service				X	X	X			
development				X		X			X
company	X	X							X
technology		X		X		X			
people		X		X	X				
risk		X	X						X
value			X		X				X
opportunity		X			X				
system		X	X			X			
product		X			X				X
strategy		X					X		X
industry			X		X				X

**Table A7.** *Cont.*

Word	A	B	C	D	E	F	G	H	I
growth					X	X			X
transformation					X	X			
society			X		X				X
capability			X		X		X		
issue			X		X				X
market			X	X					X
effort		X			X	X			

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: process, G; Facilitators, H: Leadership, I: Strategy.

**Table A8.** Mapping list of safety performance factors and Weight 2 words in 2020.

Word	A	B	C	D	E	F	G	H	I
operation	X	X				X			
enhance			X		X				X
high quality			X	X		X			
challenge		X				X			X
model	X				X				X
project		X	X						X
time	X		X		X				X

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: process, G; Facilitators, H: Leadership, I: Strategy.

**Table A9.** Mapping list of safety performance factors and Weight 1 words in 2020.

Word	A	B	C	D	E	F	G	H	I
team		X				X			X
community			X			X			X
sustainability			X		X	X			
employee				X		X	X		
support	X	X			X				
impact			X		X				X
culture		X			X		X		
area	X	X	X						
health	X					X	X		
investment	X	X		X					

A: Provision of resources, B: Governance, C: Relational issues, D: People, E: Culture, F: Process, G; Facilitators, H: Leadership, I: Strategy.

## References

- Kim, G.D.; Kim, Y.J. Status and Challenges of Digital Transformation in Insurance Industry. *Korea Insur. Res. Inst.* **2021**, *514*, 1–8.
- Choi, S.W. A Study of the Development Aspect of Storytelling in the New Normal Era. *Korean Soc. Cult. Converg.* **2021**, *43*, 557–570. [[CrossRef](#)]
- Kim, K.S.; Seong, Y.J.; Han, Y.S. Post-COVID-19, Industrial Strategy in the New Normal Era. *Gyeonggi Res. Inst.* **2020**, *411*, 1–25.
- Nicoletti, B. *Digital Insurance: Business Innovation in the Post-Crisis Era*; Springer: Berlin/Heidelberg, Germany, 2016.
- Pearson, R. Towards an historical model of services innovation: The case of the insurance industry, 1700–1914. *Econ. Hist. Rev.* **1997**, *50*, 235–256. [[CrossRef](#)]
- Kirov, S. Pandemic Digitalization of the Insurance Business. *Izv. J. Union Sci.-Varna Econ. Sci. Ser.* **2021**, *10*, 72–79.
- Lee, J.W.; Oh, S.I. Analysis of Success Cases of InsurTech and Digital Insurance Platform Based on Artificial Intelligence Technologies: Focused on Ping An Insurance Group Ltd. in China. *KIIS* **2020**, *26*, 71–90.
- Hambrick, D.C.; Fukutomi, G.D. The seasons of a CEO's tenure. *Acad. Manag. Rev.* **1991**, *16*, 719–742. [[CrossRef](#)]
- Chah, D.O. A study of recent research on CEO leadership. *Korean Assoc. Pers. Adm.* **2005**, *29*, 205–258.
- Amernic, J.H.; Craig, R.J. Guidelines for CEO-speak editing the language of corporate leadership. *Strategy Leadersh.* **2007**, *35*, 25–31. [[CrossRef](#)]
- Na, H.J.; Lee, K.C.; Choi, S.U.; Kim, S.T. Exploring CEO messages in sustainability management reports: Applying sentiment mining and sustainability balanced scorecard methods. *Sustainability* **2020**, *12*, 590. [[CrossRef](#)]

12. Aerts, W.; Cormier, D.; Magnan, M. Corporate environmental disclosure, financial markets and the media: An international perspective. *Ecol. Econ.* **2008**, *64*, 643–659. [CrossRef]
13. Thomas, J. Discourse in the marketplace: The making of meaning in annual reports. *J. Bus. Commun.* **1973**, *34*, 47–66. [CrossRef]
14. Kloptchenko, A.; Eklund, T.; Karlsson, J.; Back, B.; Vanharanta, H.; Visa, A. Combining data and text mining techniques for analysing financial reports. *Intell. Syst. Account. Financ. Manag. Int. J.* **2004**, *12*, 29–41. [CrossRef]
15. Charteris-Black, J.; Ennis, T. A comparative study of metaphor in Spanish and English financial reporting. *Engl. Specif. Purp.* **2001**, *20*, 249–266. [CrossRef]
16. Shirata, C.Y.; Takeuchi, H.; Ogino, S.; Watanabe, H. Extracting key phrases as predictors of corporate bankruptcy: Empirical analysis of annual reports by text mining. *J. Emerg. Technol. Account.* **2011**, *8*, 31–44. [CrossRef]
17. Choi, J.W.; Han, H.S.; Lee, M.Y.; Ahn, J.M. The Prediction of Corporate Bankruptcy Using Text-mining Methodology. *Product. Rev.* **2015**, *29*, 201–228. [CrossRef]
18. Humphery-Jenner, M.L.; Powell, R.G. Firm size, takeover profitability, and the effectiveness of the market for corporate control: Does the absence of anti-takeover provisions make a difference? *J. Corp. Financ.* **2011**, *17*, 418–437. [CrossRef]
19. Gupta, R.; Gill, N.S. Financial statement fraud detection using text mining. *Editor. Pref.* **2012**, *3*, 189–191.
20. Yook, K.H. The Relationship between Environmental Accounting and CEO Message Using Text Mining Technique—Comparative Study of Korea and Japan. *J. Korean Assoc. Mod. Jpn.* **2017**, *56*, 387–400.
21. Choi, Y.G.; Cho, K.T. Analysis of safety management characteristics using network analysis of CEO messages in the construction industry. *Sustainability* **2020**, *12*, 5771. [CrossRef]
22. Yook, K.H. CEOs Talk about Social Responsibilities and Sustainable Performance: Applications of Text Mining Approach. *Korea Acc.* **2018**, *27*, 253–279. [CrossRef]
23. Choi, Y.K.; Cho, K.T. Analysis of Environmental Management Characteristics Using Network Analysis of CEO Communication in the Automotive Industry. *Sustainability* **2021**, *13*, 11987. [CrossRef]
24. Bang, K.W. A Text Mining for CEO’s Letters—A Study of Specialized Text Types through the Case of German Companies. *Inst. Humanit. Chosun Univ.* **2021**, *62*, 371–401.
25. Uhm, S.W.; Lee, S.S. Analysis of ESG Management Strategy Trends through Text Mining of CEO Messages. *KEPAS* **2021**, *1*, 208–209.
26. Kim, S.J.; Kim, Y.M. 2021 KIRI CEO Survey. *Korea Insur. Res. Inst.* **2021**, *2021*, 17.
27. Durst, S.; Stähle, P. Success factors of open innovation—a literature review. *Int. J. Bus. Res. Manag.* **2013**, *4*, 111–131.
28. Im, J.; Kim, H.; Miao, L. CEO letters: Hospitality corporate narratives during the COVID-19 pandemic. *Int. J. Hosp. Manag.* **2021**, *92*, 102701. [CrossRef]
29. Siano, A.; Tuccillo, C.; Conte, F.A. CEO Communication in Sustainability Organizations: Activities and Tools. Available online: <https://www.researchgate.net/publication/259053875> (accessed on 2 January 2022).
30. Park, C.S. Discovering Meaningful Trends in the Inaugural Addresses of North Korean Leader via Text Mining. *J. Inf. Technol. Appl. Manag.* **2019**, *26*, 43–59.
31. Yun, J.H.; Lee, J.H. A Study on Generalization of Corporate CSR Keywords Using Text Mining. *KMIS Int. Conf.* **2021**, *21*, 31–48.
32. Chesbrough, H.W. *Open Innovation: The New Imperative for Creating and Profiting from Technology*; Harvard Business School Press: Boston, MA, USA, 2003.
33. Kim, S.K.; Jang, B.Y.; Lee, Y.J.; Song, J.G.; An, D.H.; Lee, K.H.; Choi, J.S. Open Innovation: Theory, Practices, and Policy Implications. *STEPI* **2008**, *12*, 1–330.
34. Yang, C.H. An Empirical Study on the Success Factor and Performance of Management Innovation. *Korean Ind. Econ. Assoc.* **1999**, *12*, 63–84.
35. Calantone, R.J.; Cavusgil, S.T.; Zhao, Y. Learning orientation, firm innovation capability, and firm performance. *Ind. Mark. Manag.* **2002**, *31*, 515–524. [CrossRef]
36. Kyung, J.S. A Study on Impact of Innovative and Network Competency of Small-Medium Enterprises on Performance. *Korean Ind. Econ. Assoc.* **2021**, *34*, 363–383.
37. Kim, H.; Kim, N.J. The Effect of Innovation Success Factors and Innovation Capabilities on Performance in Manufacturing Firm. *J. Korean Prod. Oper. Manag. Soc.* **2009**, *20*, 99–125.
38. Jang, K.S.; Kim, Y.B.; Koo, I.S. A Study on the Relationship between Innovation Activity and Business Results -Focused on the Small and Medium Manufacturing Enterprises. *J. Korean Soc. Qual. Manag.* **2010**, *38*, 512–520.
39. Chang, S.I.; Jang, K.S. A Study on the Influence of CEO’s Management Characteristics of SMEs on Innovation Performance and Management Performance. *K Bus. Manag. J.* **2020**, *4*, 43–67.
40. Hong, J.H.; Cho, Y.S.; Park, H.H.; Choi, J.H. The Impact of CEO Business Propensity, Innovation Orientation and Employee Innovation Activity on Business Performances: Focused on the Small, Medium and Venture Enterprises. *J. Commod. Sci. Technol.* **2012**, *30*, 19–32.
41. Lee, C.D. The Causal Relationship among Strategy, Organization Culture, Innovation. *Korean J. Bus. Adm.* **2011**, *24*, 3441–3460.
42. Kalay, F.; Gary, L.Y.N.N. The impact of strategic innovation management practices on firm innovation performance. *Res. J. Bus. Manag.* **2015**, *2*, 412–429. [CrossRef]

43. Köse, İ.; Güner, Ş.; İlgüzerer, B.; Sisli, M.E. A case study of the extended interactive innovation management model in insurance company. In Proceedings of the 13th European Conference on Innovation and Entrepreneurship (ECIE), Aveiro, Portugal, 20–21 September 2018; Academic Conferences and Publishing International Limited: Kidmore End, UK, 2018; pp. 375–383.
44. Seo, J.G.; Nam, C.M.; Kim, T.H.; Park, S.H. Keyword Network Analysis on Long Term Care Insurance Using Text Mining. *J. Health Inform. Stat.* **2021**, *46*, 257–266. [[CrossRef](#)]
45. Brandirectory. Insurance 100 2020 Top 1–100. Available online: <https://brandirectory.com/rankings/insurance/2020> (accessed on 10 May 2021).
46. Marsden, J.; Briony, T. Brand values: Exploring the associations of symmetry within financial brand marks. *Des. Manag. J.* **2013**, *8*, 62–71. [[CrossRef](#)]
47. Oliver, S. Message from the CEO: A three-minute rule? *Corp. Commun. Int. J.* **2000**, *5*, 158–167. [[CrossRef](#)]
48. Jung, G.H. *A Study of Foresight Method Based on Text Mining and Complexity Network Analysis*; Korea Institute of S&T Evaluation and Planning: Seoul, Korea, 2010.
49. Lee, J.J. The Scope and trends of Innovation Studies. *Korean Manag. Rev.* **1998**, *27*, 1115–1139.
50. Lee, H.S.; Yi, C.G. New Discussion of the Conceptual Definition of Technology Innovation. *KOTIS* **2021**, *24*, 777–798.
51. Lee, S.S. A content analysis of journal articles using the language network analysis methods. *J. Korean Soc. Inf. Manag.* **2014**, *31*, 49–68.
52. Diesner, J.; Carley, K.M.; Tambayong, L. Extracting socio-cultural networks of the Sudan from open-source, large-scale text data. *Comput. Math. Organ. Theory* **2012**, *18*, 328–339. [[CrossRef](#)]
53. Lee, I.W.; Lee, Y.M. Semantic Network Analysis on Core Values and Policy Orientation: By Focusing on Moon Administration Policy Road-map. *Korean Public Adm. Q.* **2019**, *31*, 643–670. [[CrossRef](#)]
54. Hussain, S.; Muhammad, L.J.; Yakubu, A. Mining social media and DBpedia data using Gephi and R. *J. Appl. Comput. Sci. Math.* **2018**, *12*, 14–20. [[CrossRef](#)]
55. Agarwal, U.A.; Datta, S.; Blake-Beard, S.; Bhargava, S. Linking LMX, innovative work behaviour and turnover intentions: The mediating role of work engagement. *Career Dev. Int.* **2012**, *17*, 208–230. [[CrossRef](#)]
56. Bhat, S.S.; Milne, S. Network effects on cooperation in destination website development. *Tour. Manag.* **2008**, *29*, 1131–1140. [[CrossRef](#)]
57. Mrvar, A.; Batagelj, V. Analysis and visualization of large networks with program package Pajek. *Complex Adapt. Syst. Model.* **2016**, *4*, 6. [[CrossRef](#)]
58. Heo, J.E.; Yang, C.H. Applying Network Analysis in Convergent Research Relationships: The Case of High-Tech Convergence Technology Development Program. *J. Korea Technol. Innov. Soc.* **2013**, *16*, 883–912.
59. Hong, J.H.; Yun, H.J. Presidential Candidate’s Speech based on Network Analysis: Mainly on the Visibility of the Words and the Connectivity between the Words. *J. Korea Contents Assoc.* **2014**, *14*, 24–44. [[CrossRef](#)]
60. Shih, H.Y. Network characteristics of drive tourism destinations: An application of network analysis in tourism. *Tour. Manag.* **2006**, *27*, 1029–1039. [[CrossRef](#)]
61. Kim, Y.H.; Kim, Y. *Social Network Analysis*, 4th ed.; ParkYoungsa: Seoul, Korea, 2016; pp. 308–322.
62. Opsahl, T.; Agneessens, F.; Skvoretz, J. Node centrality in weighted networks: Generalizing degree and shortest paths. *Soc. Netw.* **2010**, *32*, 245–251. [[CrossRef](#)]
63. Knoke, D.; Yang, S. *Social Network Analysis*, 3rd ed.; SAGE Publications: New York, NY, USA, 2019; pp. 100–156.
64. Liu, B.; Li, Z.; Chen, X.; Huang, Y.; Liu, X. Recognition and Vulnerability Analysis of Key Nodes in Power Grid Based on Complex Network Centrality. *IEEE Trans. Circuits Syst. II Express Briefs* **2017**, *65*, 346–350. [[CrossRef](#)]
65. Cherven, K. *Mastering Gephi Network Visualization*, 1st ed.; Packt Publishing: Birmingham, UK, 2015; pp. 181–215.
66. Aju Business Daliy. Available online: <https://www.ajunews.com/view/20211012114940983> (accessed on 12 October 2021).
67. Deloitte. Available online: <https://www2.deloitte.com/global/en/insights/industry/financial-services/financial-services-industry-outlooks/insurance-industry-outlook-2021.html> (accessed on 3 December 2020).
68. Rezaee, Z. Business sustainability research: A theoretical and integrated perspective. *J. Account. Lit.* **2016**, *36*, 48–64. [[CrossRef](#)]
69. Tarmuji, I.; Maelah, R.; Tarmuji, N.H. The impact of environmental, social and governance practices (ESG) on economic performance: Evidence from ESG score. *Int. J. Trade Econ. Financ.* **2016**, *7*, 67. [[CrossRef](#)]
70. Yu, E.P.Y.; Guo, C.Q.; Luu, B.V. Environmental, social and governance transparency and firm value. *Bus. Strategy Environ.* **2018**, *27*, 987–1004. [[CrossRef](#)]
71. Lo, K.Y.; Kwan, C.L. The effect of environmental, social, governance and sustainability initiatives on stock value—Examining market response to initiatives undertaken by listed companies. *Corp. Soc. Responsib. Environ. Manag.* **2017**, *24*, 606–619. [[CrossRef](#)]