

Article

A Human-Needs-Based Dynamics to Simulate Technology Policy and Its Effects on Both Business Success and Human Happiness

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Abstract: This paper focuses on how human needs are reflected in the market and how several technological and political policies affect the market share of government-supported industries, as well as the satisfaction of human desires and consequent happiness. In this paper, we seek to understand the dynamics of consumer decision-making processes in relation to technology products in the market. In this study, we present a new marketing model based on human needs, wants, and demands, and focus on both holistic and social perspectives. We have shown that human-based policy dynamics and sustainable human happiness can be realized by stimulating national policies for consumer happiness in the human-needs-based sector, e.g., the healthcare industry.

Keywords: human needs; human happiness; market dynamics; consumer decision-making; sustainable market policy

1. Introduction

Recently, improving the quality of life and welfare of citizens has been considered in the development of national and enterprise policies. Happiness is a sign that human needs are met. Average happiness has increased in most countries over the last decade. Still, happiness has not increased everywhere and most countries are far from achieving an optimal level of happiness. There seems to be a need for rethinking how we should shape our society and our lives in a happiness-increasing direction. Many researchers believe that some very powerful technologies will be developed during the coming decades, including biotechnology, nanotechnology, artificial intelligence, and the like. However, little attention has been given to how these can be used to improve human happiness and welfare as well as enhancing our emotional capacities.

Will the nation as a whole achieve greater happiness as a result of increased national policies of science and technology? The government exists for the purpose of serving the people. One of the primary goals of technology policy should be to assist in this pursuit. Therefore this paper focuses on how human needs are reflected in the market and how several technological and political policies affect the market share of government-supported industries, as well as the satisfaction of human desires and consequent happiness.

In this study, we explored factors that affect the relationship between national R&D policy and human happiness. We assumed that social marketing practices based on improving human happiness affect long-term customer satisfaction (since it is known that perceived happiness is the main driver of customer satisfaction) [1].

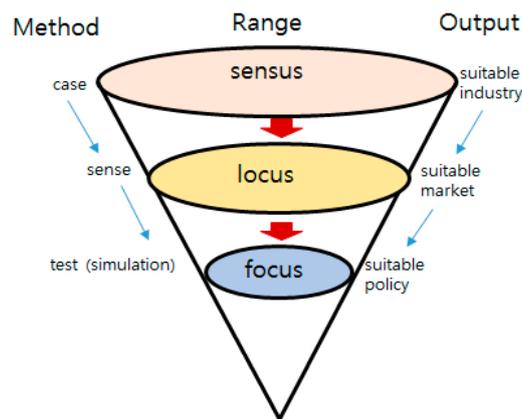


Figure 2. Method, range and output of research.

A first dimension is the *sensus* of human needs—the degree of sensitivity to changes in human needs while we are interacting with the industry. A *locus* of human wants, meanwhile (in Latin, the word means “place”), is a site or location. As is the case with *sensus*, the *locus* can be a specific place or part of the *sensus*. The third dimension, *focus* applies policy simulations to industries and markets that have been discovered in the upper dimension. This will ultimately ensure sustainable growth potential.

3. Research Model

The research model in this paper proceeds as follows (Figure 3).

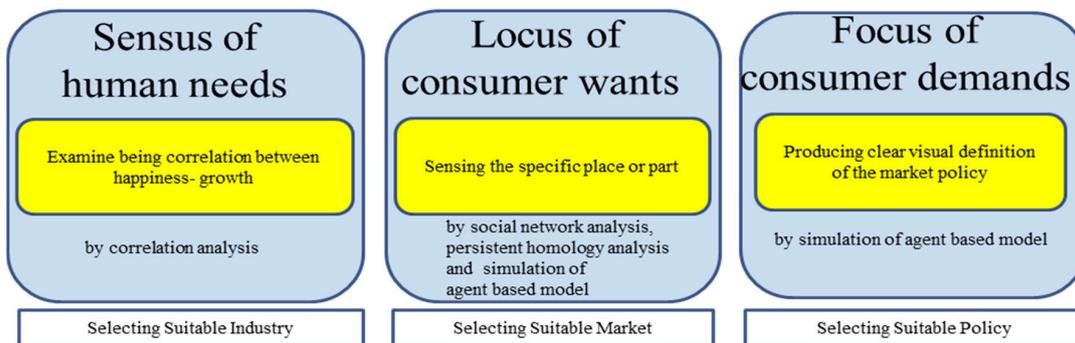


Figure 3. Research model.

Section 4.1 examines the *sensus* of human needs, which is the industry setting that best fits the people of the country. Here, Korea is taken as an example. Section 4.2 examines the *locus* of human wants, sensing the specific place or part in the *sensus* by social network analysis, persistent homology analysis, and simulation of an agent-based model [4]. Section 4.3 examines the *focus* of demands, meaning the state or quality of having or producing a clear visual definition of market policy by simulation of an agent-based model.

4. Methods and Data

4.1. *Sensus of Human Needs*

We analyzed the relationship between national happiness and bio-innovation indices across 56 countries, which were matched to each other using the investigation reported by ITIF and the “World Happiness Report” (Table 1) [5].

Table 1. National happiness index and bio-innovation index.

Country	Happiness *	R&D Composite Score	Government Health R&D	Government R&D as a Share of GDP	Extent of Price Controls	Country	Happiness	R&D Composite Score	Government Health R&D	Government R&D as a Share of GDP	Extent of Price Controls **
Denmark	1	1	2	3	3	Spain	29	6	4	24	3
Switzerland	2	13	13	10	1	Slovakia	30	27	14	37	2
Iceland	3	17	35	2	2	Malaysia	31	41	40	41	2
Norway	4	2	6	8	3	Italy	32	24	21	26	2
Finland	5	28	41	7	2	Japan	33	34	38	21	2
Canada	6	32	33	20	2	Russia	34	39	51	14	2
Netherlands	7	23	28	18	2	Poland	35	14	10	31	2
New Zealand	8	25	22	27	2	South Korea	36	21	39	6	2
Australia	9	31	18	38	3	Lithuania	37	43	45	42	2
Sweden	10	3	9	4	2	Slovenia	38	15	16	16	2
Israel	11	52	55	29	1	Peru	39	46	37	53	2
Austria	12	12	25	1	2	Latvia	40	51	47	55	2
United States	13	7	12	5	1	Romania	41	47	44	46	2
Costa Rica	14	53	52	44	2	Estonia	42	16	24	11	2
Germany	15	22	34	13	2	Hong Kong	43	50	50	43	3
Brazil	16	37	43	22	2	Turkey	44	56	56	48	2
Belgium	17	48	53	28	2	Indonesia	45	44	30	51	1
Ireland	18	38	36	35	3	Philippines	46	54	48	56	2
Mexico	19	35	27	40	1	China	47	26	19	32	3
Singapore	20	9	11	9	1	Hungary	48	18	15	25	2
United Kingdom	21	4	1	30	3	Portugal	49	8	8	19	2
Chile	22	33	17	50	2	Vietnam	50	40	31	52	2
Argentina	23	36	32	34	1	Greece	51	29	26	33	2
Czech Republic	24	19	23	17	2	South Africa	52	30	20	39	3
Colombia	25	11	3	49	2	India	53	42	49	23	3
France	26	20	29	12	3	Kenya	54	10	5	45	2
Thailand	27	49	42	54	3	Ukraine	55	45	46	36	2
Taiwan	28	5	7	15	1	Bulgaria	56	55	54	47	2

* This is a ranking of the sampled countries, not the actual order; ** Extent of Price Controls: 1, low; 2, moderate; 3, high; Sources: [6,7].

We examine the aggregate relationship by structural modeling using the PLS-SEM algorithm [8]. The hypothesis of this analysis is as follows.

Hypothesis 1. *Healthcare capabilities positively influence the long-term performance of technology investments, which is regarded as a critical factor of human happiness, and work by deploying resources with the capacity to create value (Figure 4).*



Figure 4. The direct and mediated relationships that enhance human happiness.

The results of confirmatory factor analysis (CFA) and construct reliability (CR) are shown in Tables 2 and 3. They were estimated to be wider than a 99% confidence level [9].

Table 2. Average variance extracted (AVE).

	Original Sample	Sample Mean	Standard Deviation	T Statistics	p Value
Healthcare Investment	0.517	0.544	0.045	9.597	0.000
Technology Investment	0.834	0.831	0.050	16.657	0.000

Table 3. Composite reliability.

	Original Sample	Sample Mean	Standard Deviation	T Statistics	p Value
Healthcare Investment	0.642	0.635	0.096	6.680	0.000
Technology Investment	0.910	0.908	0.030	30.382	0.000

In Table 4, all the direct relationships postulated in this analysis were found to be positive and significant (at a 0.001 level of significance), except the hypothesis that proposed a positive influence of healthcare investment on human happiness, which was found to be insignificant.

Table 4. Direct path coefficients.

	Original Sample (β)	Standard Deviation	T Statistics	p Value
Healthcare Investment -> Human Happiness	-0.290	0.184	1.582	0.114
Technology Investment -> Human Happiness	0.655	0.149	4.382	0.000
Healthcare Investment -> Technology Investment	0.739	0.040	18.556	0.000

Figure 5 shows that “Healthcare Investment” indirectly affects “Human Happiness” through “Technology Investment”. Therefore, this analysis established healthcare investment as one of the critical requirements of human happiness.

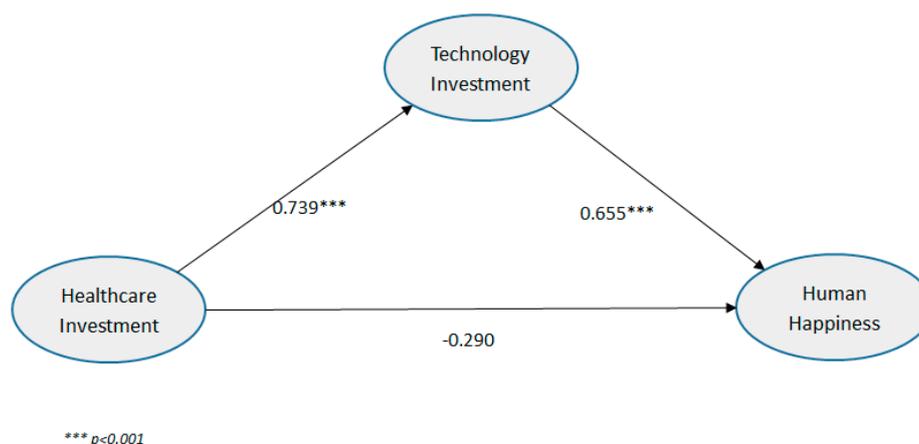


Figure 5. Path network among the various factors.

4.2. Locus of Consumer Wants

4.2.1. Bio-Issue Clustering

We analyzed co-occurrence patterns between significant words to identify potential market needs within the healthcare industry. Our text data consist of the titles of healthcare sectors found in the BRIC (Biology Research Information Center) database (<http://www.ibric.org>) from 2009 to 2015 (Figure 6) [10].

We first performed a persistent homology analysis to determine the number of clusters. The reason for this analysis is that there is no objective criterion for how many clusters should be set in the cluster analysis. The notion of “persistent homology” is the algorithms for topological invariants across various scales of a topological manifold in TDA (Topology Data Analysis) [11]. A “barcode” plot is a qualitative visualization of a homology group. The x -axis in Figure 7 represents the scale parameter “ e ” and the y -axis is an arbitrary ordering of homology generators. In Figure 7, the straight line that survives to the end is meaningful even though the y -axis increases clockwise, while short lines are just topological noise [12]. The number of meaningful long lines is 12. This is because we have to choose the place where the gap between lines is the largest (Figure 7).

Here, we can see 12 of the main topics in these word clusters. The vertical distance represents the degree to which semantic relationships were included. The co-occurrence words-based, social network analysis is one of the simplest topic models and, as such, is very widely used. The model is considered psychological, which means such models represent a sort of social perspective (Figure 8).

4.2.2. Hysteresis Effect Model for Bio-Issue Analysis

We use the cluster above to construct a hysteresis effect model to see how the relationship between elements changes with changes in external circumstances. Here, hysteresis means that the state of the object depends on the state of the past. In other words, depending on the strength of the system in the past, the recovery paths are different. In the figure, “a” represents the path of the smallest bond and “b” represents the path for the large bond (Figure 9).

This model is a representation of the major factors. The nodes in this model represent the symptoms of major needs for the market. These factors are pharmaceutical investment, clinical infrastructure, bio-technology development, health food development, R&D investment, expanding commercialization, the development of medical equipment, new drug infrastructure, healthcare development, fostering new industries, technology development, and medical complex development. These issues have a direct causal influence on other factors; this is called the causal network perspective [13].

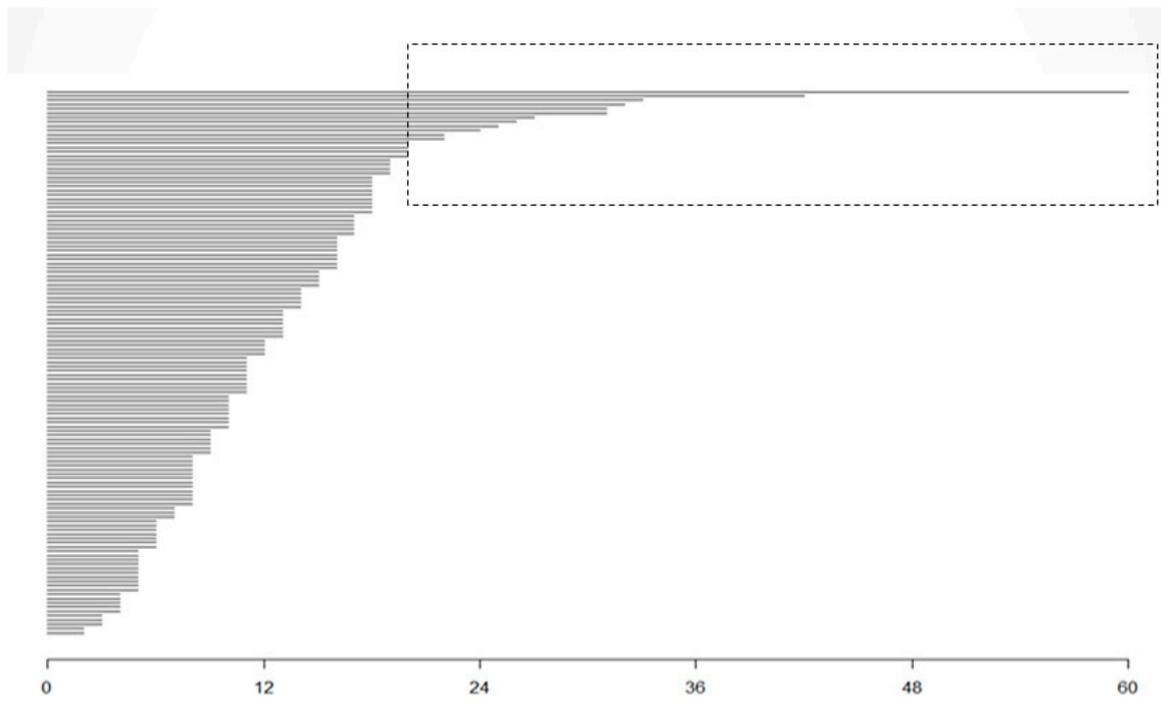


Figure 7. Barcode plot.

The ABM (agent-based model) for market needs management is based on a partial correlation between each factor signal. At each time step, the probability of each needs signal is calculated, which depends on the total activation of neighboring nodes in the previous period as well as the specific parameters. The individual application parameters depend on the slope and intercept of each factor. The activation probability for each factor can be expressed as follows:

$$\text{logit}(Y) = \ln(\text{odds}) = \ln\left(\frac{\pi}{1-\pi}\right) = \alpha X - \alpha\beta.$$

If we solve $\text{logit}(Y)$ for p —Probability($Y = 1 \mid X = x$, a specific value of X), we obtain:

$$\pi = \frac{e^{(\alpha X - \alpha\beta)}}{1 + e^{(\alpha X - \alpha\beta)}} = \frac{1}{e^{-(\alpha X - \alpha\beta)}}.$$

where π is once again the probability of the event, $a (= \alpha)$ and $b (= \alpha\beta)$ are regression coefficients, and X is a set of predictors. Parameters “ a ” (steepness) and “ b ” (intercept) are typically estimated by the maximum likelihood methods. On the other hand, the parameter “ a ” applied to the individual factor represents the slope of the probability function as shown in the above equation, and it can be multiplied by the total needs amount of the entire network to determine the entire needs change. The parameter “ b ” is the needs-based threshold value (resistance value). Market needs with a higher threshold value require a greater intensity of stress to activate a signal with a lower threshold value (Table 5).

Figure 10 shows the hysteresis plot of the ABM. The histogram shows the frequency of activated needs factors. The following is an example of analysis according to two scenarios [14].

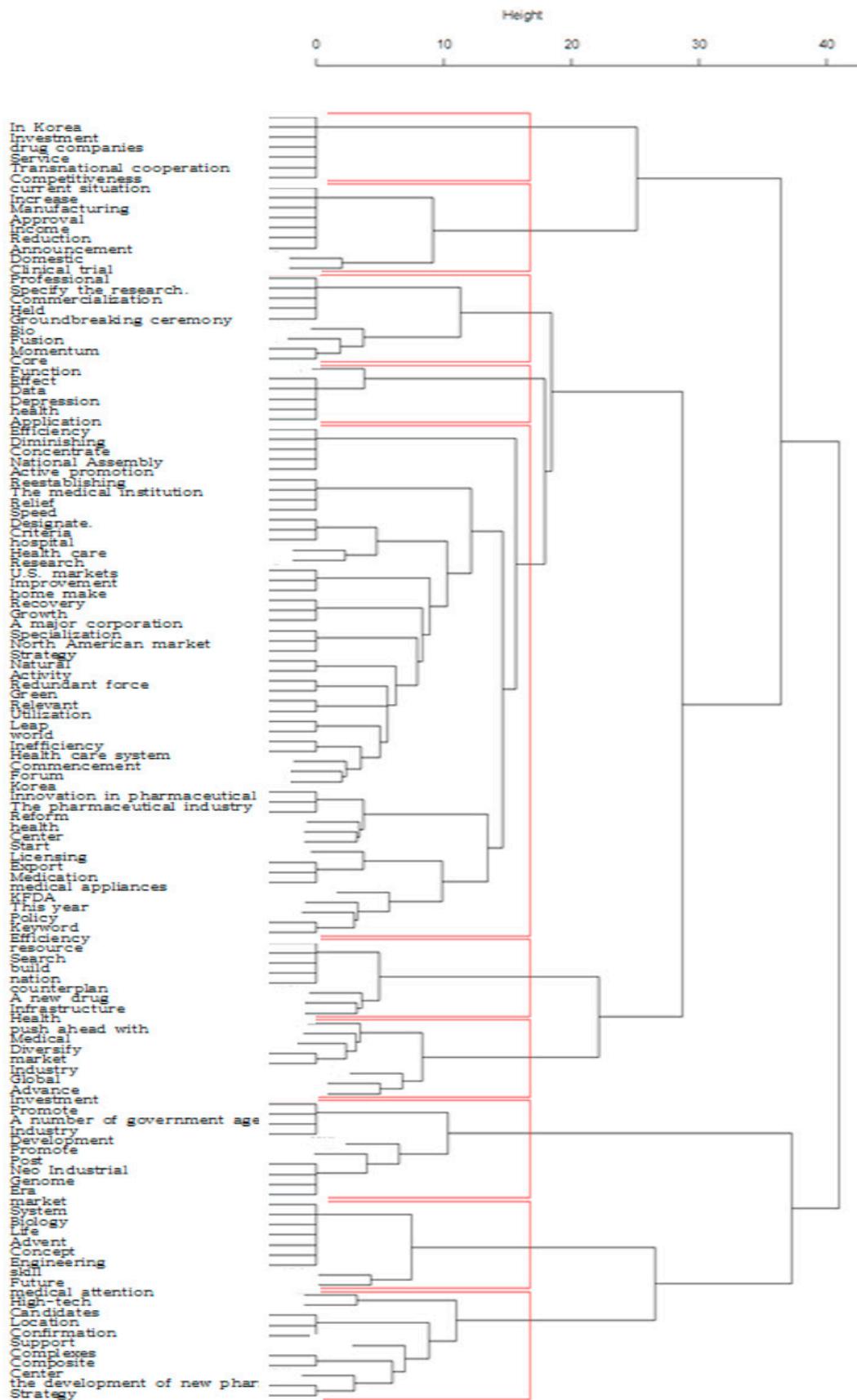


Figure 8. Hierarchical clustering with Euclidean distance.

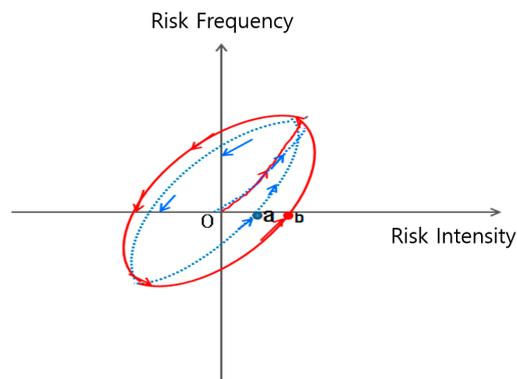


Figure 9. Conceptual illustration of hysteresis effect.

Table 5. Steepness and intercept of clusters.

No.	Cluster	Steepness (a)	Intercept (b)
0	Pharmaceutical investment	0.2	0.7
1	Clinical infrastructure	0.2	0.9
2	Bio-technology development	0.2	0.9
3	Health food development	0.2	0.6
4	R&D investment	0.1	1.4
5	Expanding commercialization	0.1	2.2
6	Development of medical equipment	0.1	1.5
7	New drug infrastructure	0.2	0.8
8	Healthcare development	0.2	0.8
9	Fostering new industries	0.2	1
10	Technology development	0.2	0.9
11	Medical complex development	0.2	1.1

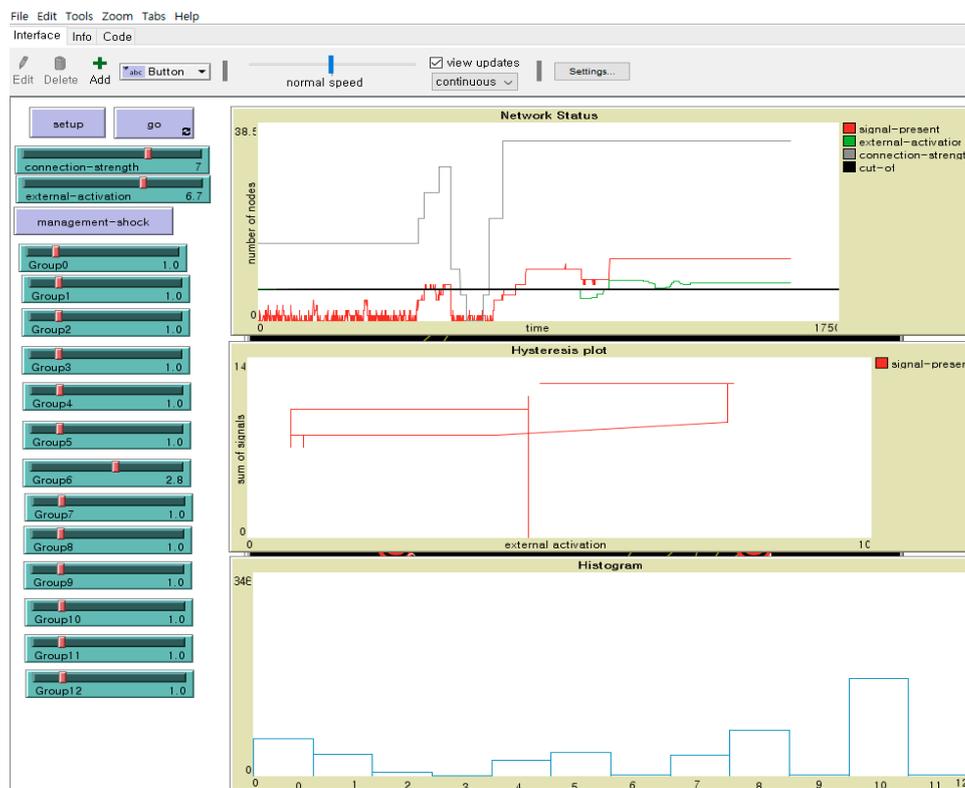


Figure 10. Screen of simulation model developed by ABM tool NetLogo.

Scenario 1

When the intensity of connection between elements is changed, the demand frequency does not change even if the inter-element connection strength is increased above the demand excess threshold (cutoff: 8). Conversely, when the intensity of the connection below the threshold value is lowered, the demand frequency is relatively changed. Only the 7th factor (new drug Infrastructure) and the 8th factor (healthcare development) have survived (Figure 11).

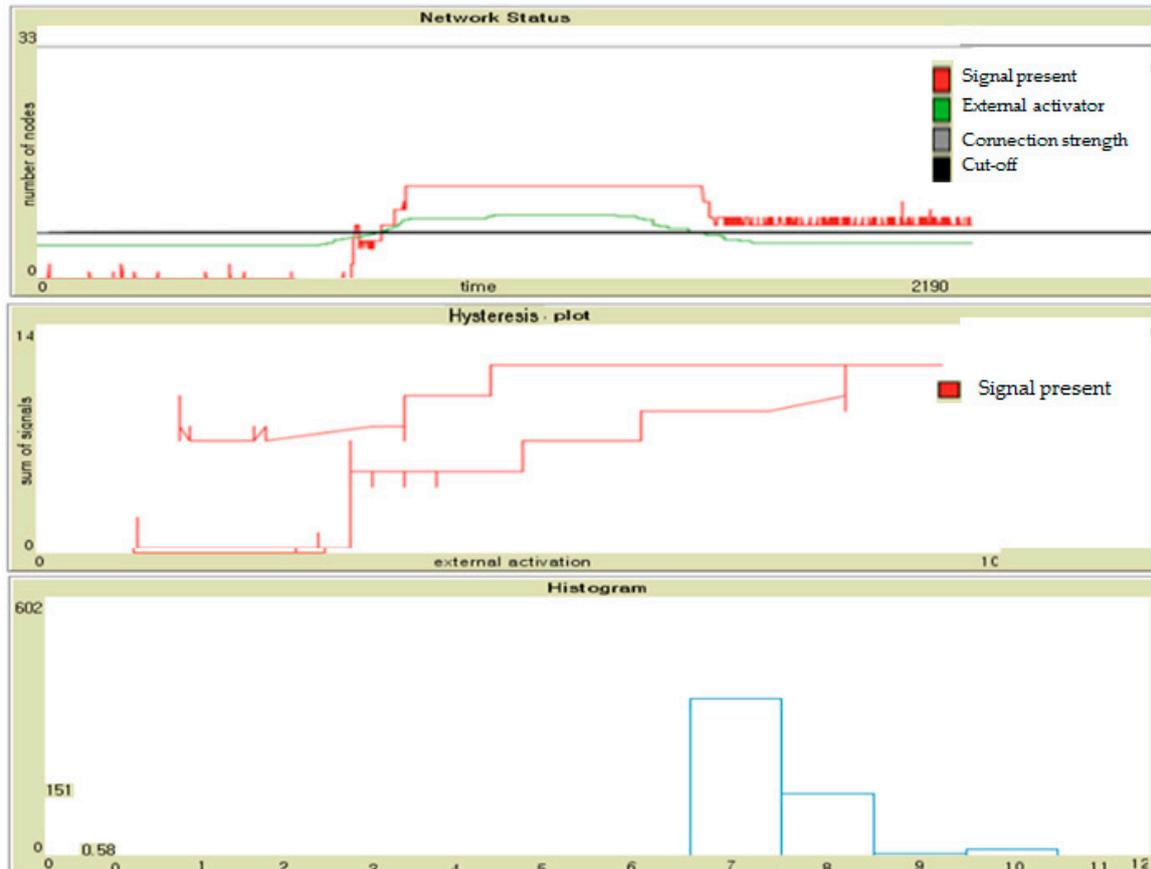


Figure 11. The case of changing the connection strength between the factors.

Scenario 2

In this model, we try to simulate the characteristics of the factors depending on the external impact strength. First, when external shocks are strong, the most frequent factors are 0th (pharmaceutical investment), 4th (R&D investment), and 5th (expanding commercialization) (Figure 12). On the other hand, when the external shock was weak, no element was activated (Figure 13).

As described above, we found the following five of 12 major topics through hysteresis analysis: new drug infrastructure, healthcare development, pharmaceutical investment, R&D investment, and expanding commercialization. By analyzing the common areas of these five topics, we selected the pharmaceutical market for our target market. As shown in Figure 14, the pharmaceutical market plays an important role in R&D, commercialization, and infrastructure of healthcare systems in South Korea.

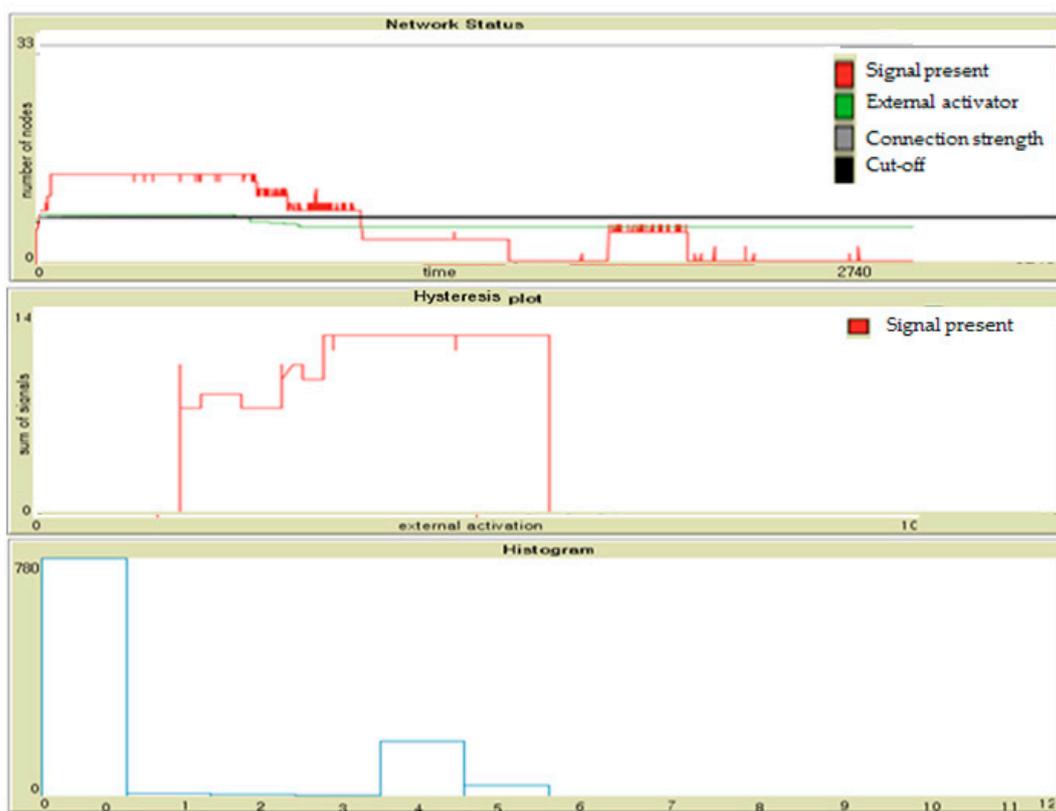


Figure 12. Comparison of hysteresis between factors when the external impact is weak.

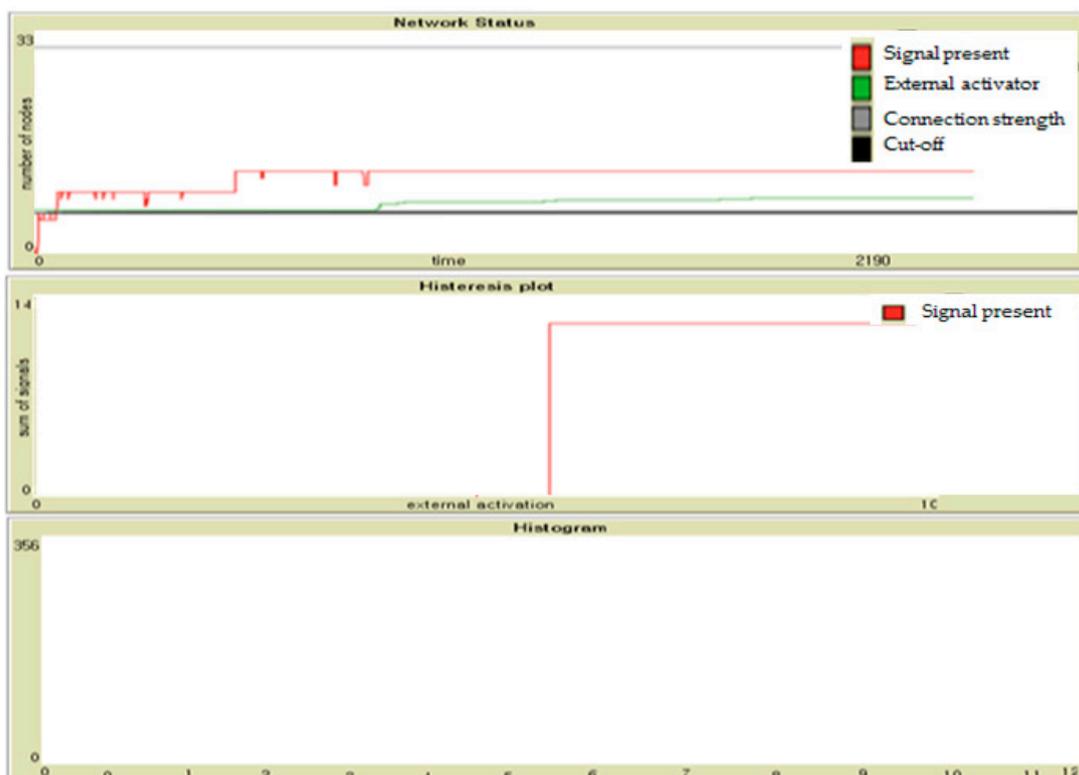


Figure 13. Comparison of hysteresis between factors when the external impact is strong.

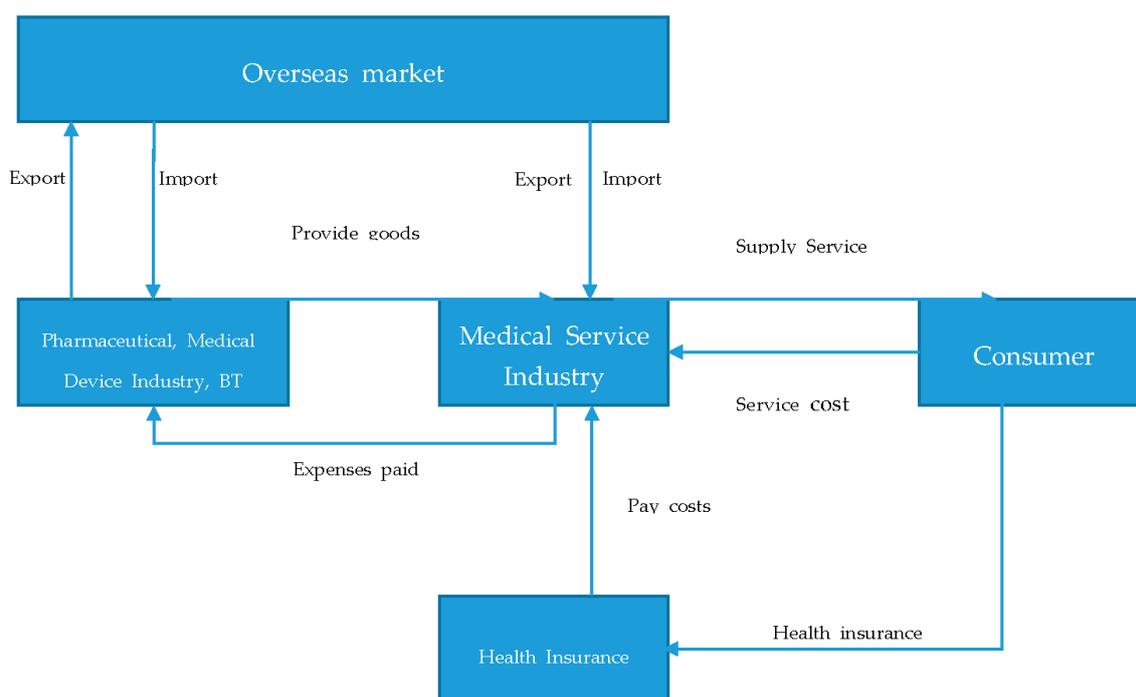


Figure 14. Healthcare systems in South Korea.

4.3. Focus of Consumer Demands

As mentioned above, we targeted the pharmaceutical sector, in which the factors of both growth and welfare coexist and are highly associated. Market intervention was carried out by the government through policy change. Policies were focused on new technologies for drug development (to increase the competitiveness of products) and those that improved the national protection and welfare of the country.

As mentioned earlier, the number of patients living with long-term chronic diseases is increasing rapidly due to the extension of human life. Furthermore, there have been social crises caused by sudden illnesses (i.e., swine flu) throughout the world. As such, policies on appropriate technology development and pricing have a powerful influence on increasing the sales of related companies, improving the quality of life of patients, and diffusing social unrest. In consideration of risks such as disasters (e.g., the sinking of the Sewol ferry in South Korea and the Middle East Respiratory syndrome (MERS)), the variable impact factors from both sociocultural and subjective perspectives, apart from a technical and scientific point-of-view, have become increasingly important [15].

We utilized the Agent-Based Model to study the relationship between happiness and maximum sales in a society and included corporations, consumers, and potential consumers in the investigation. This model is based on an individual's decision and allows for the measurement of changes in the system. We used different scenarios for creating policies and then evaluated them in relation to business success and the satisfaction of human needs. Ultimately, we analyzed the possibilities for enhancing both sustainability and social soundness. We did this by comparing policies focused on improving the success of businesses and consumer happiness.

4.3.1. Model

Human decisions, such as a purchasing decision, can be explained by analyzing a combination of uncertainties and risks with the Bayesian Probability Theory of multi-attribute utility. Through this process, the maximized effectiveness of various alternatives can be evaluated. This cognitive decision-making process can be divided into the following groups: the recognition step,

information-gathering stage, alternative evaluation stage, purchasing, and post-purchasing stage [16]. In addition, when considering various alternatives, the purchasing decision is based on maximizing the value of goods by considering a number of the product's properties [17].

However, in the real market, these rational decisions are affected by both personal traits and socioeconomic interactions. Every individual has different motivations and buying behaviors based on external influences such as changes in technological innovation [18].

The purchasing decision is the sum of personal utility values for each product attribute. These values constitute an individual's rational desire for a product. The post-purchase evaluation of a product's attributes affects the personal utility of the product. In terms of the desirability of a product, an individual is affected by the people who are using that product, as manifested in social interactions. The motivation to buy a particular product is affected by a risk assessment in relation to social issues, such as epidemics of infectious diseases and the side effects of medications (Figure 15).

A consumer's purchasing decision can be expressed as the sum of values that are provided to the consumer by the utility of specific properties. Consumer i 's utility for product j that has the n -numbered attribute can be expressed as follows:

$$Utility_{i,j} = \sum_n A_n \times product\ attribute_n.$$

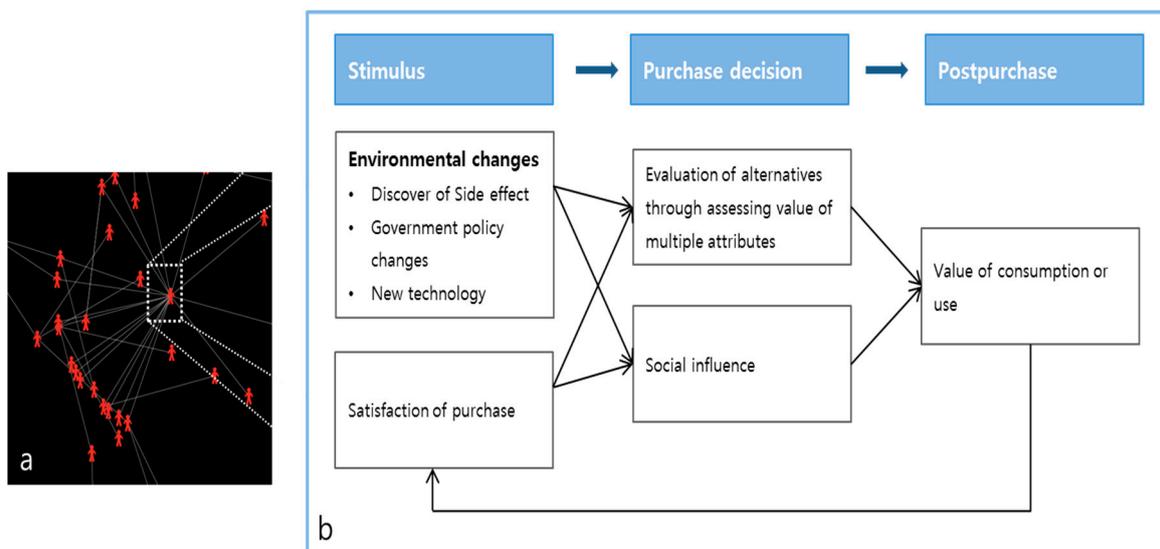


Figure 15. A consumer's purchasing decision process. (a) The model environment within which individual's purchasing decisions are made in social activities; (b) at the individual level, each customer has personal sensitivity to each attribute (price, brand, drug efficiency, convenience, etc.). Purchasing decisions and purchasing behavior were influenced by these individual characteristics and other people's stimuli within their social network system.

A product attribute refers to the n -properties of the product, such as price and brand. A is a coefficients of these independent properties of the product. The weights of major independent attributes, as well as attributes that were important in the decision-making process, were calculated using the conjoint analysis results in a pharmaceutical market (Table 6) [19]. Finally, the personal utility value was randomly distributed to provide diversity in the ABM model.

Table 6. The weight of attributes for multi-attribute utility decisions.

Attribute	Definition	Weight
Brand	Brand awareness	1.3
Price	The money required to pay for a single dose of pills	21.7
Efficiency	Medical efficacy to decrease main symptoms of disease	9.5
Convenience	Improve convenience, including administration method, number, etc.	7.1
Side Effects	Unwanted non-therapeutic effects caused by a drug	12.8
Marketing	Advertisement to increase product awareness	6.7

Consumers, as actors interacting within a social network structure, are affected by the products selected by the people around them. The influence of surrounding people regarding product j on consumers can be expressed as follows:

$$\text{Social effect}(\text{Influence of surrounding people}_{i,j}) = \frac{\sum \text{Neighbor}_{i,j}}{\sum \text{Neighbor}_j + \sum \text{Neighbor}_{j'}}.$$

The purchasing decisions of consumers can be expressed as the sum of individual utility value and social effects:

$$\text{Purchase decision}_{i,j} = \alpha * \text{Utility}_{i,j} + \beta * \text{Social effect}_{i,j}.$$

We simulated a market model from a political perspective instead of a corporate marketing perspective. As such, the products were classified as market leading products, low-cost products, and new technological products rather than specific products from a particular company (Table 7).

Table 7. Classification and characterization of products.

Product	Price (Won)	Efficiency (%)	Dosing Convenience	Side Effects	Brand	Marketing
Market leading product (A)	500~1000	100	No	Yes or No	High	High
Low-cost product (B)	500~600	100	No	Yes or No	Middle or Low	Middle
New-technology product (C)	600~1000	100~200	Yes or No	Yes or No	High, Middle, Low	High or Middle

4.3.2. Simulation of Agent-Based Model

In this paper, we targeted the prescription drug market for high blood pressure for two reasons. First, high blood pressure increases significantly with the aging of the global population. Secondly, a large number of people in the elderly population have been greatly influenced by the development of new technologies for treating high blood pressure. Since this is a huge market, pharmaceutical companies continue to develop new products with unique competitive advantages. In fact, there are a number of products on the market at the same time with a variety of prices and various utility values. However, there are significant development costs associated with bringing new technologies to the point of clinical trials for proving safety and efficacy. Because of these barriers in bringing new technologies to market, governments support a variety of policies to foster their development. In addition, changes in the external environment, e.g., the emergence of new-technology products or a crisis that affects specific factors, cause changes in the value of products to consumers. Thus, both satisfaction and desire welfare are closely related to appropriate technological policy.

Scenario 1

When a global company has a monopoly with their market-leading product and it develops a new, low-price product, its brand and financial powers are relatively weak. To what extent can the market share of this low-price product grow? What is the best price point for sales and profits? If a

policy on price support was introduced, what price point would be optimal for maximizing national happiness on a limited budget (Table 8)?

Table 8. Characterization of product in Scenario 1.

Product	Price (Won)	Efficiency (%)	Dosing Convenience	Side Effects	Brand	Marketing
Market-leading product (A)	800	100	No	No	Very High	High
Low-cost product (B)	500~800	100	No	No	Low	Middle

As the price of the product decreased, the market penetration of the low-price product quickly increased over time (Figure 16A,B). Both the market share and sales of product B increased as the price was lowered. Interestingly, when calculating profit in consideration of its initial development costs, the best price point for sale was not the lowest price point. When we assumed a cost of 400 won, the best price for Product B for business success was 600 won (red star in Figure 16D).

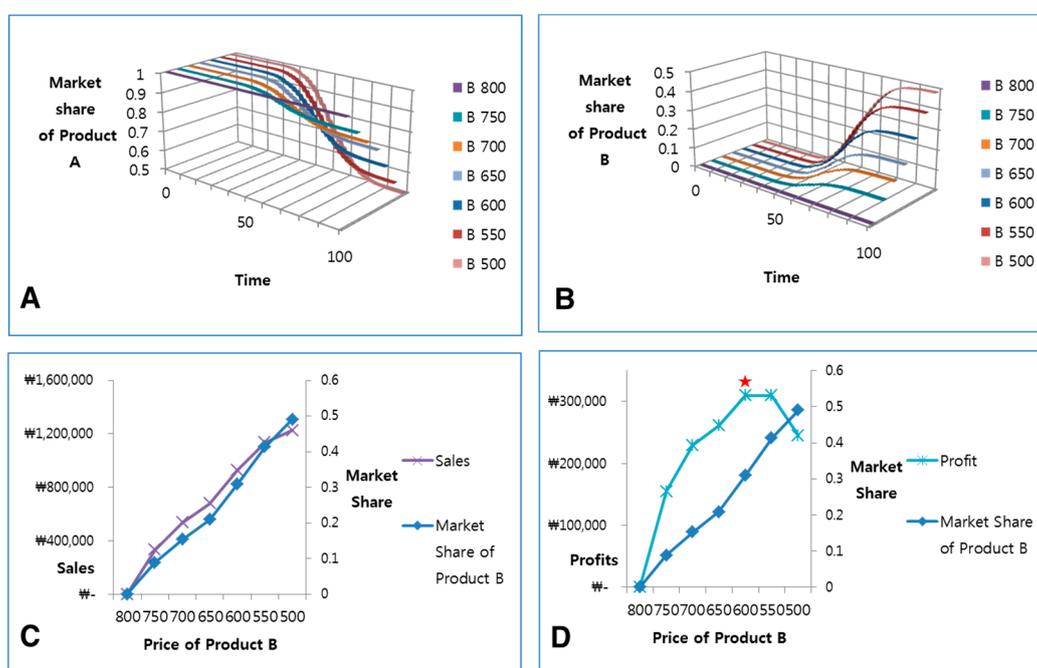


Figure 16. Changes of market share of Product A (A) and Product B (B) according to the price changes of Product B (low-price product, which entered market). Changes of market share, sales (C), and profits (D) of Product B in several conditions of Product B.

We next looked at the effect of introducing a pricing policy that supported the national welfare of elderly people through access to effective, affordable medications. The highest utility value was demonstrated when the price was dropped (with support from the budget) from 750 to 700 won (Figure 17). As such, 700 won is considered the optimal price point. By using this method of evaluation, the second optimal price point was determined to be 600 won. When the price of Product B was changed by policy within a society, the changes in utility values for Product B were calculated as follows:

$$\text{Changed Utility Values for Product C B} = \frac{\sum \text{Utility Value for Product } B_n - \sum \text{Utility Value for Product } B_{n-1}}{\text{Price of } B_n - \text{Price of } B_{n-1}}$$

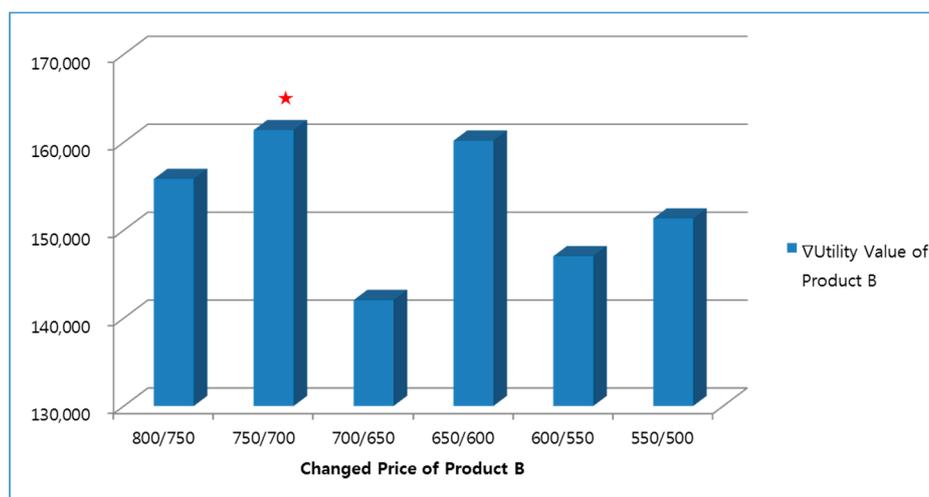


Figure 17. The sum of utility values for Product B in a society, according to the changes in price of Product B by introducing a pricing policy. When the price was dropped from 750 won to 700 won, the increase of utility values for members of society was highest (red star).

Scenario 2

Following the successful market penetration of a low-price product, a new-technology product was developed. Given several different conditions, what is the best technology policy to help this product enter the market or help existing companies retain their position in the market? The price of the low-price product was set at 600 won to maximize profits, while that of the market-leading product was set at 800 won (Table 9). Product B has 70% of the market share, while Product A has the remaining 30%.

Table 9. Characterization of products in Scenario 2.

Product	Price (won)	Efficiency (%)	Dosing Convenience	Side Effects	Brand	Marketing
Market-leading product (A)	800	100	No	No	Very High	High
Low-cost product (B)	600	100	No	No	Low	Middle
New-technology product (C)	600~1000	150	No or Yes	No	Low	Middle

The reaction of the market to pricing changes of Product C shows two inflection points (the dotted lines) (Figure 18A). When the price of Product C is greater than 900 won, it is difficult for new-technology products to enter the market (Figure 18B); furthermore, low-price products dominate the market in this scenario. When the price of Product C is between 750 and 850 won, Product B is the most sensitive to changes in price of Product C (Figure 18C). This means that Product B is the main competitor within the price range of Product C. Interestingly, the rate of decrease in market share of Product A slightly slowed. When the price of Product C fell below 700 won, the market share of Product C rose steeply, while that of Product B was not impacted. Also, the rate and speed of decline of market share of Product A were sensitive to the changes in price of Product C. In this scenario, Product C rapidly replaced Product A in terms of market share.

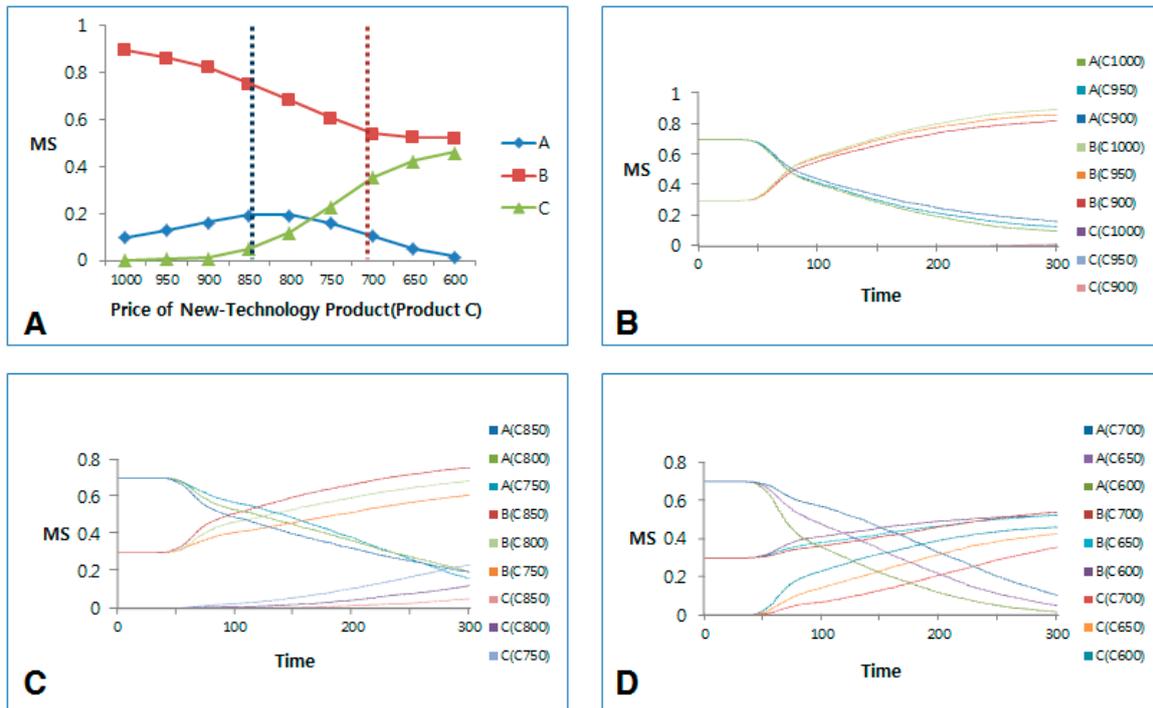


Figure 18. Changes of market share of Product A, Product B, and Product C according to the price changes of Product C (new-technology product, which entered market) (A). Changes of market share when the price of Product C is greater than 900 won (B); when the price of Product C is between 750 and 850 won (C); and when the price of Product C is 700 won or less (D).

Assuming that the development of a new-technology product costs 500 won per product, the price point to obtain the best profit is 700 won (Figure 19).

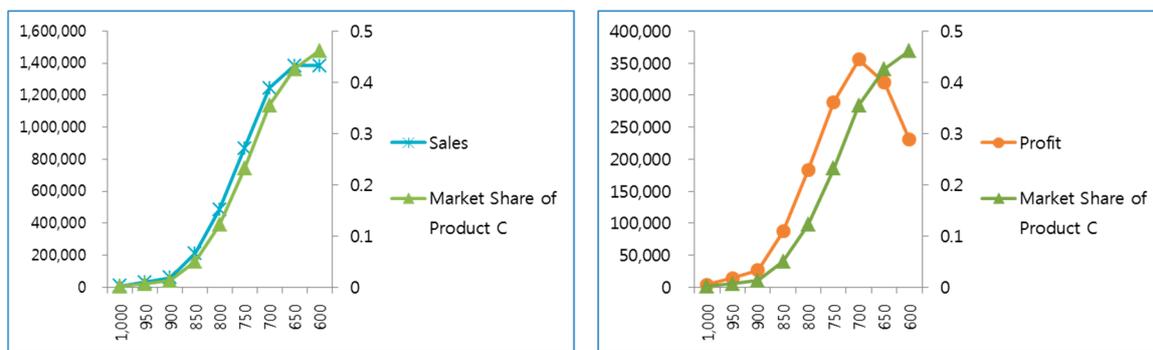


Figure 19. Changes of market share, sales (A), and profits (B) of Product C according to the price changes of Product C.

Furthermore, if there is support for the price of the new-technology product, the optimal point for cost-effectiveness is 850 won (Figure 20). When the price of Product C was changed by a policy that focused on welfare or on the development of new technologies in society, the changes in utility values were calculated as follows:

$$\text{Changed Utility Values for Product C} = \frac{\sum \text{Utility Value for Product } C_n - \sum \text{Utility Value for Product } C_{n-1}}{\text{Price of } C_n - \text{Price of } C_{n-1}}$$

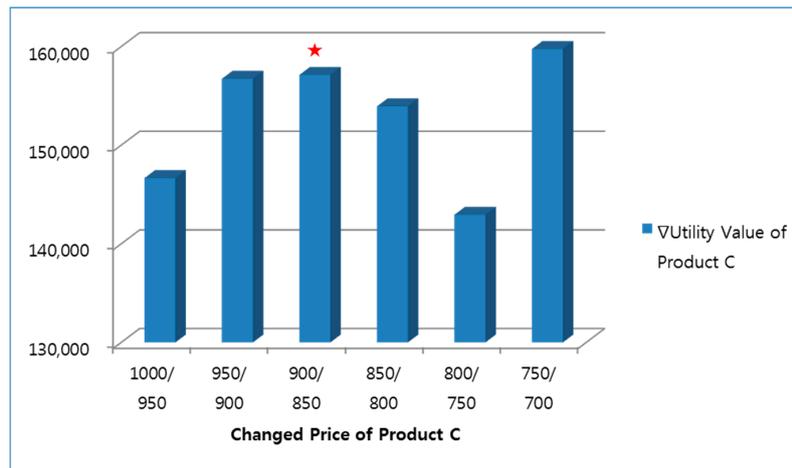


Figure 20. The changes of utility values for Product C, according to changes in the price of Product C.

If a company develops new technologies that meet a number of utilities, what is the impact on the market? We used the scenario of new technologies being developed for efficiency and dosing convenience.

Although two features were improved by new technological development, there were no significant differences in the pattern of market share compared to that of the previous technology product (Figures 21 and 22A). This demonstrates the importance of developing new technologies based on consumer needs in the market. Assuming that the initial cost is greater than the previous new-technology product, the cost must be at least 600 won (Figure 22B). In this case, the profit is approximately half that of the previous new-technology product and the appropriate price point is 750 won. However, if the initial cost is the same as the previous new-technology product (500 won), it is possible to have an optimal price of 700 won and slightly higher profits.

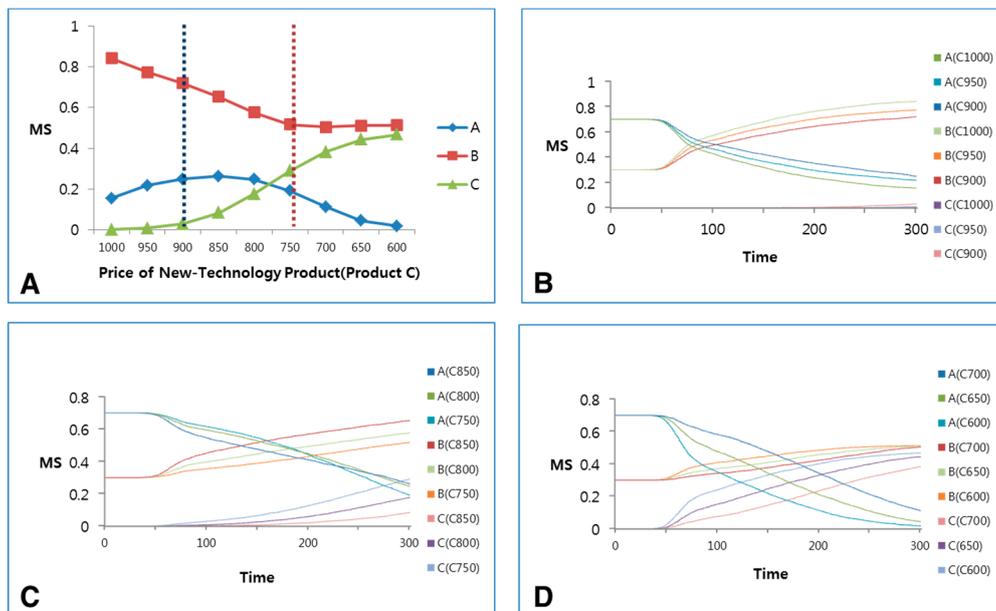


Figure 21. Changes of market share of Product A, Product B, and Product C according to the changes in price of Product C (new-technology product with improvements to two of its features) (A). Changes of market share when the price of product C is over 900 won (B); when the price of Product C is from 750 won to 850 won (C); and when the price of Product C is 700 won or less (D).

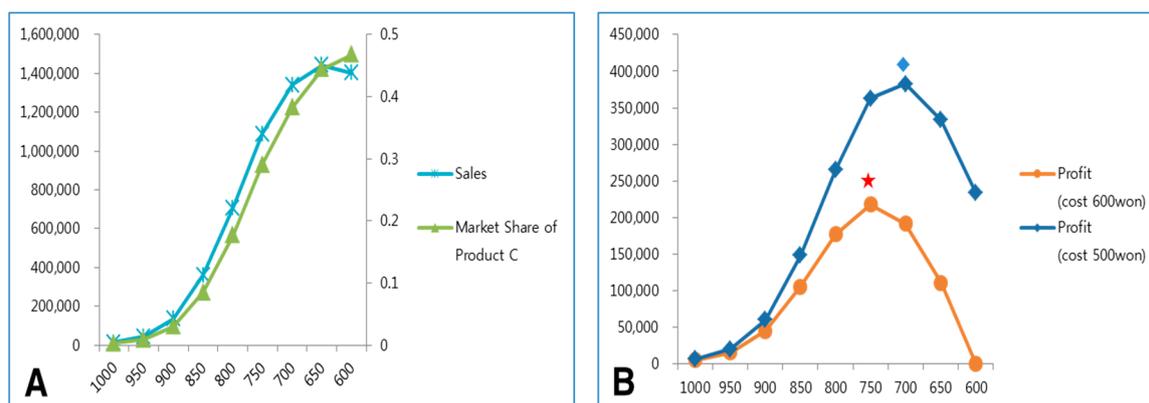


Figure 22. Changes of market share, sales (A), and profits (B) of Product C (new-technology product developed in two selected utility areas) according to the price changes of Product C.

5. Conclusions

In the existing literature, there were no examples of how human needs were reflected by the market and how technological and political policies affect the market share as well as the satisfaction of human desires and consequent happiness. In this paper, we have shown the dynamics of consumer decision-making processes in relation to technology products in the market. Also, we present a new marketing model based on human needs, wants, and demands, and focused on both holistic and social perspectives. We used the pharmaceutical market as an example to obtain a better national policy with regard to how human-happiness-based policy can impact the growth and sustainability of an industry.

We have shown that human-based policy dynamics and sustainable human happiness can be realized by stimulating national policies for consumer happiness in the human-needs-based sector (e.g., the healthcare industry). The overall impact of national policy on human happiness is stronger in these industries, where the gap between owned and currently offered products quickly shrinks.

This study has important implications for policymakers. The results of human happiness in the healthcare industry, as a barometer, may not only influence the sustainability of consumer satisfaction but also the creation of industrial policies. Forecasting the impact of national policy on human happiness and consumer satisfaction is crucial for designing sustainable R&D and for creating the right national vision.

A government's support for life-sciences research and innovation not only improves health outcomes domestically but also generates spillover effects, especially among a country's trade partners. However, it is difficult to uniformly apply the results of this study because people's needs differ according to the degree of economic growth of each country [20]. Therefore, future research should explore needs considering each country's political and economic context.

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