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Consumer Behavior in Clothing Industry and Its Relationship with Open Innovation Dynamics during the COVID-19 Pandemic

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Abstract: The fashion industry has been continuously growing over the years, yet it is an industry that was greatly affected by the COVID-19 pandemic. The purpose of the study was to determine the factors affecting buying behavior of Filipinos towards clothing apparel during the COVID-19 pandemic. There were 457 respondents who voluntarily participated and answered an online questionnaire. Structural equation modeling (SEM) indicated that marketing mix was found to have significant effects on attitude, subjective norm, and perceived behavioral control, which subsequently led to purchase intention. In addition, the COVID-19 latent variable was found to have significant effects on self-perceived severity and self-efficacy, which subsequently led to attitude and purchase intention. Interestingly, marketing mix was found to have the highest effect on actual purchase, which indicated that innovation dynamics are the keys for the buying behavior. The SEM construct can be applied to determine the clothing apparel buying behaviors of consumers in other countries, particularly during the COVID-19 pandemic. Finally, advertisements, promotions, sales, and health safety should be considered as innovation dynamics during the COVID-19 pandemic.

Keywords: consumer behavior; fashion industry; COVID-19 pandemic; protection motivation theory; theory of planned behavior; innovation dynamic; open innovation

1. Introduction

The fashion industry has gained interest and attention due to the continued growth of the global fashion industry and brands [1]. The global clothing and apparel market has achieved \$758.4 billion with a Compound Annual Growth Rate (CAGR) of 7.5% from 2014. In addition, it is projected to increase up to 11% with \$1182.9 billion by 2022 [2]. Having the fashion brands that consistently ranked in the top 100 brands in the world, the fashion industry has experienced changes over the years that led to significant success [3].

The significant success of the fashion industry has been the result of an assortment of two products: basic items and fashion items [4]. Basic items correspond in being efficient to meet the market's demands, while fashion items are the response to the current trend in the market [5]. One of the global brands that have successfully responded to the market

demands and fashion trends were Zara and H&M [5]. These brands became one of the drivers of the fast fashion industry that has successfully encouraged the market to spend.

Population figures can be associated with consumer spending [6]. The Philippines, as the second-most populous country in Southeast Asia, has a high consumer spending (Figure 1) [7]. The Philippines has encouraged new firms in the fashion industry to enter its country over the past decade. This led to the continuous growth of the market and has been eyed by the wholesale and retail industry [8]. The retail industry's growth in the country can also be associated with the Filipinos attachments to the malls, which have resulted in the continuous mall expansion and openings [8]. This drew great interest among foreign brands that have been entering the country over the years [7]. With that, Figure 2 presents the clothing shares, both import and export, in countries around the world. This highlights how the consumer's spending can be immensely evident even in different countries from products across the world.

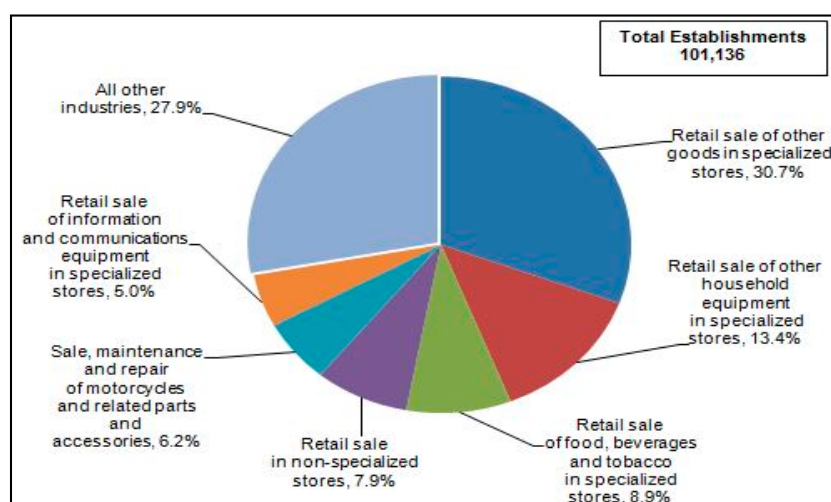


Figure 1. Percentage distribution of establishments for wholesale and retail trade: Philippines, 2017 [7].

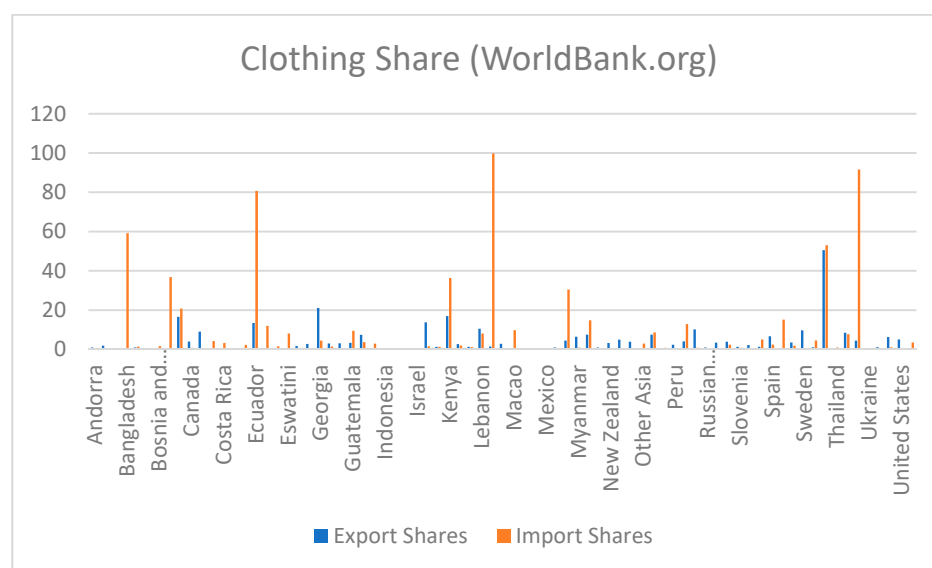


Figure 2. Export and import clothing shares across the world [9].

Several studies about the fashion industry have discussed consumer buying behavior, which can be classified into compulsive buying and impulsive buying. Compulsive buying is a buying habit that is exposed by excessive buying—with a lack of impulse control [10].

On the other hand, impulsive buying is spontaneous and immediate without any pre-shopping intentions [11]. It positively affects personality variables such as shopping enjoyment, which urge consumers to purchase. This shopping enjoyment is categorized as a hedonic motive, a type of motivation characterized by a desire for pleasure and fun. Aside from the hedonic motive, another type of shopping motive is the utilitarian motive. It is a type of motivation that considers shopping as a functional or practical task. Hedonic and utilitarian motives are considered as types of shop browsing [12].

Apart from the different behaviors and motives, unusual purchasing behavior has been observed during the COVID-19 pandemic. COVID-19 pandemic is a pandemic caused by a new strain of coronavirus, which is an infectious disease that was unknown before an outbreak started in Wuhan, China, in December 2019 [13]. With the health threat that the COVID-19 has and the limited mobility of consumers due to quarantine protocols, retailers are experiencing unusual buying behavior from the consumers [14]. The unusual consumer behavior during the initial stages of the pandemic was noted, wherein consumers were hoarding essential goods and foods to be able to self-isolate due to cyberchondria [14]. Cyberchondria is characterized as an excessive or repetitive online search for health-related data, pressured by the need to relieve health-related distress or anxiety, but it results in its worsening instead [15]. Using the different information gathered online and offline by the consumers, it led to anxiety, and behavioral response that intends to isolate that made the unusual purchase. Previous research has also found the behavioral change in the consumer during outbreaks is due to internal motivation and government policy and restrictions [16].

With the different changes that are caused by the COVID-19 pandemic, the common literature found is regarding consumers buying behavior towards panic buying of essential goods that are consists of foods, medicines, and other household needs [17–19], whereas there is limited available literature for other industries affected by the pandemic, such as the fashion industry. The study would like to address the gap in the present literature in order to help the fashion industry adapt during unprecedented situations, specifically, the COVID-19 pandemic.

The purpose of the study was to determine factors affecting buying behavior of consumers in the Philippines towards clothing apparel during the COVID-19 pandemic. Extending the previous studies [12,14,20–22], the study integrated protection motivation theory and extended theory of planned behavior by utilizing the structural equation modeling (SEM) approach. This study is one of the first studies that explore the buying behaviors of consumers, particularly related to the fashion industry during the COVID-19 pandemic. The SEM construct can be a theoretical foundation that will be beneficial for academicians, retailers, and even the government. Finally, the SEM construct can be applied and extended to analyze the buying behaviors of consumers in other countries, particularly during the COVID-19 pandemic.

2. Conceptual Framework and Innovation Dynamics

Figure 3 represents the conceptual framework and innovation dynamics of the study with five exogenous latent variables (COVID-19, marketing mix—product, and macro-environmental factors—economic, technological, and political) and seven endogenous latent variables (self-efficacy, attitude towards the behavior, perceived severity, subjective norm, purchase intention, perceived behavioral control, and actual purchase).

2.1. Protection Motivation Theory

In order to identify the effect of COVID-19 on buying behavior, protection motivation theory was integrated with the theory of planned behavior (TPB). Protection motivation theory (PMT) is a theory of social cognition designed to explain how individuals respond to health threats such as COVID-19 [23], wherein perceived severity and self-efficacy predict adoption of individual prevention measures. Perceived severity refers to the seriousness of the whole situation (e.g., COVID-19 can lead to death) [14,23,24]. Self-efficacy refers to

when an individual is assured that they have the ability to perform a defensive act (e.g., I consider the clothes as sanitized before purchase) [24,25].

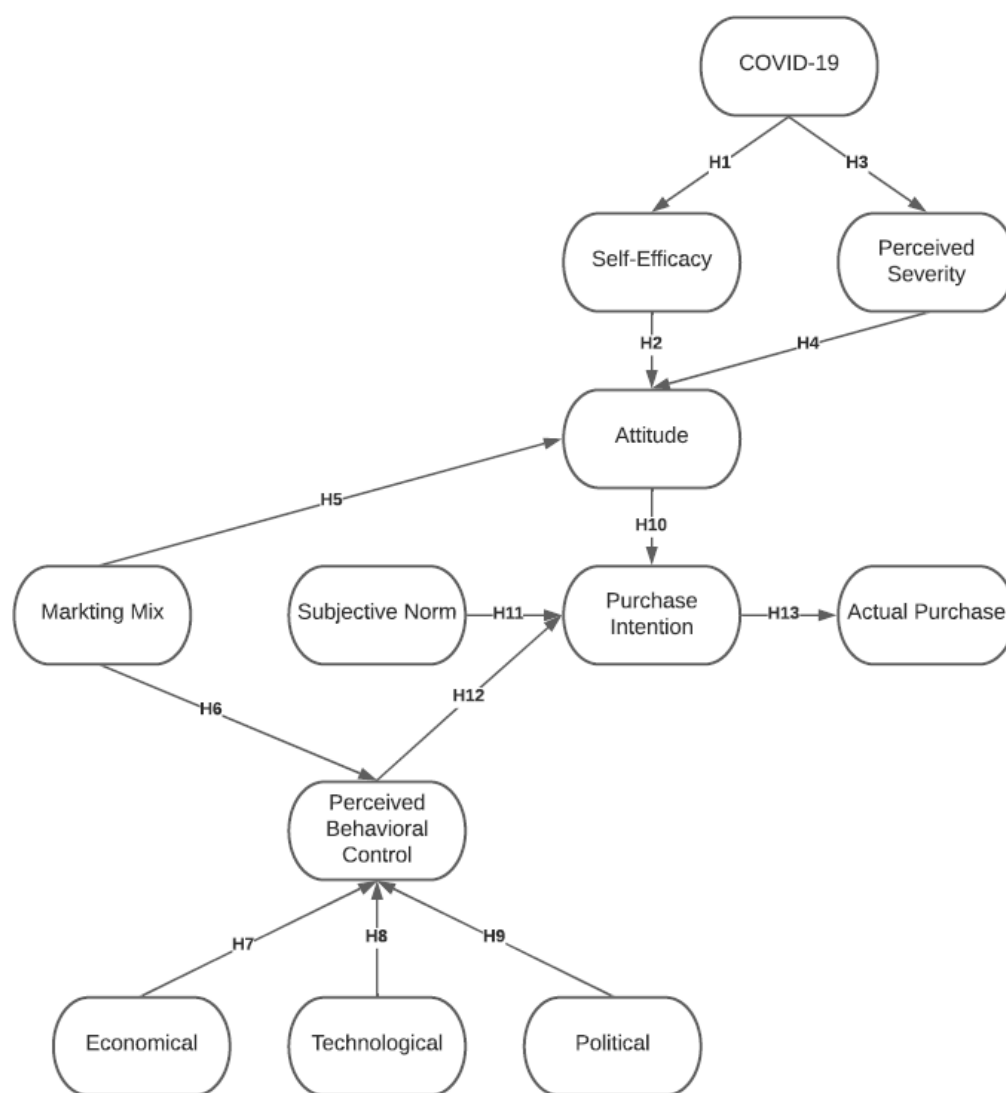


Figure 3. Conceptual framework.

Protection motivation theory in the study is composed of self-efficacy and perceived severity. Self-efficacy refers to the individual's ability to successfully perform the preventative behavior [26]. O'Leary stated that an individual need to feel assured in their ability and knowledge to perform preventive measures [26]. Studies found that when an individual's knowledge towards a threat is high, then a higher self-efficacy is expected [24,27]. In order to validate if the current pandemic affects self-efficacy and if self-efficacy can affect an individual's behavior, we hypothesized that:

Hypothesis 1 (H1). *COVID-19 has a significant effect on self-efficacy.*

Hypothesis 2 (H2). *Self-efficacy has a significant effect on attitude towards behavior.*

Perceived severity is defined as predictive of actual behavioral intention [28]. When an individual who had greater perceived severity to a health threat has already adopted the preventive measure, the individual would likely assume that they are less vulnerable to a health threat. On the other hand, Lau et al. found that perceived severity related to a pandemic was not significant [28]. Based on the different findings, this study would

want to determine if an individual would take action to prevent COVID-19. Thus, we hypothesized that:

Hypothesis 3 (H3). *COVID-19 has a significant effect on perceived severity.*

Hypothesis 4 (H4). *Perceived severity has a significant effect on attitude towards the behavior.*

2.2. Marketing Mix

Apart from the different factors that could affect behavioral intentions, retailers must have strategic ways on how to survive. To compete effectively and efficiently, retailers must focus on the different factors that affect the consumer's purchasing power [29]. In order to do this, using the different micro- (marketing mix) and macro-environmental factors were utilized as an instrument. For the micro factors, the 7Ps model was used. As compared to 4Ps, the 7Ps model offers a more systematic and informative method for evaluating the marketing mix of a service product [30]. With macro-environmental forces, it depends on the landscape of the industry, the evolving retail landscape that made e-commerce grew rapidly because of its benefits [31], and the present crisis that the world is facing caused by the pandemic and having limited mobility of consumers due to quarantine protocols has caused global recession [32].

The marketing mix is a comprehensive plan to create various approaches on the target market, which is identified based on the company's capabilities and goals on how to transform consumers time, money, and energy into sales [33]. The marketing mix in the study is composed of the six factors from the 7Ps (product, price, place, promotion, people, and process). The study of Nugroho and Irena considered marketing mix towards consumer's purchasing intention in Surabaya [34]. However, their study focused on the Halal cosmetic market. Anjani et al. considered the 7P's marketing mix and the relationship with customer loyalty in traditional markets of Indonesia [35]. Lastly, Kusumawati et al. considered the 7P's marketing mix in buying intention of music products in Indonesia [36]. It could be seen that studies have linked marketing mix with people's behavior. Thus, this study incorporated a marketing mix to extend the theory of planned behavior. Thus, the following were hypothesized:

Hypothesis 5 (H5). *Marketing mix factors have a significant effect on attitude towards the behavior.*

Hypothesis 6 (H6). *Marketing mix factors have a significant effect on perceived behavioral control.*

According to Venkatesh et al., PBC is defined as an easy way to perform a behavior and overcome the constraints with the different health protocols and prevention such as lockdowns, which resulted in economic consequences [37]. According to the study of International et al., there is a recession in the economic aspect of people globally during the COVID-19 [32]. With that, their funds or money are said to be only spent on the needs of a person. However, the perception of needs would depend on an individual's point of view [14,38]. With the recession, political factors such as being on lockdown and strict implementation of limited people able to go out during the COVID-19 pandemic may also be a factor that affects buying behavior. Moreover, the study of Chen et al. and Peng et al. showed how technological aspects such as access to different online stores and the user's response are affected by their shopping behavior [39,40]. Thus, the following were hypothesized:

Hypothesis 7 (H7). *Economic factors have a significant effect on perceived behavioral control.*

Hypothesis 8 (H8). *Technological factors have a significant effect on perceived behavioral control.*

Hypothesis 9 (H9). *Political factors have a significant effect on perceived behavioral control.*

2.3. Theory of Planned Behavior

To measure the buying behavior holistically, TPB was integrated. TPB is one of the most commonly studied models by social psychologists for predicting behavioral intention [41,42]. It indicates that purchasing intention is a very powerful predictor of actual purchase [41,42]. Thus, this study uses purchasing intention to represent the buying behavior of the consumer. Moreover, the TPB in this study utilized the variables such as attitude towards the act of behavior, subjective norm, and perceived behavioral control. The attitude towards the act of behavior is when an individual's intention is impacted by the person's attitude toward an act (e.g., purchasing clothing apparel is a good idea) [43]. Subjective norm is the attitude towards the behavior along with the impact of the relevant reference people (e.g., people around me influence my purchasing behavior), and the perceived behavioral control is defined as the ease or difficulty to perform a behavior (e.g., I intend to purchase clothing apparel in my next purchase) [44].

The study would want to validate how society affects or influence the purchase intention of an individual. Attitude is one of the most addressed variables in the field of research aiming to understand the reasons why people choose to develop such behavioral intention [45]. Tommasetti et al. identified attitude as a concept of belief of an individual [45]. Likewise, Kayagil considered attitude to be a state of mind capable of exerting a direct or indirect effect on the individual's reaction to all objects and circumstances associated with it [46]. On the other hand, subjective norm (SN) is identified as an individual's belief that people who are significant to an individual should (should not) carry out a particular action [47]. According to Kim et al., subjective norm plays an important part in influencing behavioral intentions [48]. This can affect actual behavior, not only directly but can also convey personal norms. Studies have verified that attitude and subjective norm have a strong ability to distinguish the individuals' behavioral intention [49,50]. However, in certain instances, the effect of the two variables has not proved to be adequate [51]. Thus, the following were hypothesized:

Hypothesis 10 (H10). *Attitude towards the behavior has a significant effect on purchase intention.*

Hypothesis 11 (H11). *Subjective norm has a significant effect on purchase intention.*

Perceived behavioral control (PBC) is intended to affect both intent and behavior [49,51]. According to Armitage and Conner, PBC is more likely to hinder than influence the actual intention of the behavior [49]. Robinson and Smith have the same result where it stated that consumers are not highly confident in their ability to purchase [50]. Hence, it was hypothesized that:

Hypothesis 12 (H12). *Perceived behavioral control has a significant effect on purchase intention.*

According to Tommasetti et al., behavioral intention is a variable that influences the actual behavior taken, but can at the same time be dependent on other factors (attitude, subjective norms, and perceived behavior) [45]. Ajzen also stated that behavioral intention is a predictor of an individual's action because it is aligned with the different motivational factors [51]. These factors carry out the intention if an opportunity arises. Therefore, it was hypothesized that:

Hypothesis 13 (H13). *Purchase intention has a significant effect on actual purchase.*

3. Methodology

3.1. Participants

The study included a total of 457 respondents that had an interest in clothing apparel seen in Table 1. The convenience sampling method was done in choosing the respondents who voluntarily answered an online questionnaire distributed through social media

platforms. The survey was made available on Facebook in a public setting. Aside from posting as status, the survey was also posted in different Facebook groups to obtain more respondents and was made available from December 2020 to February 2021.

Most of the participants were within the age range of 23–30 (47%), followed by 15–22 (31%). About 8% of the respondents were 31–38 years of age, 6% were aged between 39–46, 4% were of ages 47–54, while 55–62 and 63–70 were 3% and 1%, respectively. For the educational level of the respondents, the majority were University/College graduates (70%), while Graduate School Graduates are about 15%, High School and Vocational Graduates were 14% and 1%, respectively. As for the employment status, 49% of the respondents were employed, 30% were students, 12% were self-employed, about 8% were unemployed, and 1% were already retired. For the monthly family income, 23% of the respondents has between PhP 21,914 and PhP 43,828, 19% were between PhP 43,828 and PhP 76,699, 18% were between PhP 10,957 and PhP 21,914, 14% were between PhP 76,699 and PhP 131,484, at least PhP 219,140 were 11%, 9% and 6% were less than PhP 10,957 per month and between PhP 131,483 and PhP 219,140, respectively. For the region, most of the respondents are from Metro Manila/NCR with 74%, followed by CALABRAZON with 17%, Central Luzon at 4%, and the rest were from different regions in the Philippines.

Table 1. Demographic profiles of the respondents ($N = 457$).

	Category	N	%
Age	15–22	143	31%
	23–30	216	47%
	31–38	38	8%
	39–46	26	6%
	47–54	19	4%
	55–62	12	3%
	63–70	3	1%
Education level	High school	63	14%
	Vocational	5	1%
	University/college	320	70%
	Graduate school	69	15%
Marital status	Single	384	84%
	Married	67	15%
	Cohabitant	3	1%
	Widowed	2	0%
	Separated	1	0%
Employment status	Employed	224	49%
	Unemployed	37	8%
	Student	138	30%
	Self-employed	53	12%
	Retired	5	1%

Table 1. *Cont.*

	Category	N	%
Range of monthly family income	Less than PhP 10,957 per month	41	9%
	Between PhP 10,957 and PhP 21,914 per month	84	18%
	Between PhP 21,914 and PhP 43,828 per month	107	23%
	Between PhP 43,828 and PhP 76,699 per month	85	19%
	Between PhP 76,699 and PhP 131,484 per month	63	14%
	Between PhP 131,483 and PhP 219,140	27	6%
	At least PhP 219,140	50	11%
Region	National Capital Region (NCR) or Metro Manila	337	74%
	CALABARZON	79	17%
	Northern Mindanao	3	1%
	MIMAROPA	3	1%
	Cagayan Valley	2	0%
	Bicol Region	3	1%
	Central Luzon	19	4%
	Western Visayas	1	0%
	CARAGA	1	0%
	Eastern Visayas	2	0%
	Central Visayas	4	1%
	Ilocos Region	1	0%
	Cordillera Administrative Region (CAR)	1	0%
	Zamboanga Peninsula	1	0%

3.2. Questionnaire

A self-administered questionnaire was designed for this study following the conceptual framework (Table 2). With the advantage of convenient distribution and convenient filling process, an online survey was done via Google Form. The questionnaire was composed of 2 parts: demographics and structural equation model (SEM) constructs. The first part of the questionnaire was demographic questions of the participants: age, education level, marital status, and employment status, followed by SEM questionnaires.

Structural equation modeling (SEM) was used to check, both empirically and simultaneously, the relationship between the factors of the developed framework [45]. The questionnaire for SEM was created with 4 sections—marketing mix, macro factors, protection motivation theory, and theory of planned behavior, wherein participants rated the questionnaire via a 5-point Likert scale (from 1 = ‘Strongly Disagree’ to 5 = ‘Strongly Agree’). For marketing mix, 7Ps marketing mix framework was used in order to help marketers make decisions regarding segmentation, positioning, and differentiation. With this, marketers may enhance a product’s marketing mix for an improved sales results [52]. The macro-environmental factors were used in order to consider other factors that would affect the retailers and the consumers buying behavior. It is an indicator that reflects the significance and attractiveness of the industry that would help create competitive strategies [51]. The theory of planned behavior was developed to point out and understand the specific behaviors that the consumer possess in purchasing [45]. Protection motivation theory was adapted in order to determine how purchasing behavior affects behavioral change during a global pandemic [14].

Table 2. The questionnaire construct in the current study.

Construct	Items	Measures	Reference
Marketing mix	PRODUCT	MM1 Clothes offered must be the latest trends.	[53]
		MM2 Clothes offered must be high quality.	[20]
		MM3 The clothes that I need are always available.	[54]
		MM4 Clothes offered in different varieties or colors influence my buying decision.	[55]
	PRICE	MM5 Retail price can influence purchase intention.	[42]
		MM6 I compare store prices when shopping.	[40]
		MM7 Charging lower prices than competitors is a must.	[56]
	PLACE	MM8 Need for touch is necessary when purchasing clothes.	[12]
		MM9 I prefer shopping for clothes in actual stores pre-COVID-19.	
		MM10 I prefer shopping for clothes in actual stores during COVID-19.	
		MM11 I prefer shopping for clothes online pre-COVID-19.	
		MM12 I prefer shopping for clothes online during COVID-19.	
	PROMOTION	MM13 Brand image influences purchase intention.	[57]
		MM14 Brand endorser/s influence buying behavior.	[58]
		MM15 Social media posts can influence buying behavior.	[59]
		MM16 Social media posts can influence brand image.	[60]
		MM17 Sales/Promos influence buying decision.	[61]
	PEOPLE	MM18 Salespeople create a positive impact with store/brand image.	[62]
		MM19 Salespeople's recommendations influence buying decisions.	[63]
		MM20 Salespeople have an impact on customer satisfaction.	[63]
	PROCESS	MM21 Maintaining stock availability influences buying decision.	[64]
		MM22 Maintaining stock availability has an impact on customer satisfaction.	[65]
		MM23 Store/website design influences brand loyalty.	[56]
		MM24 Merchandise displays inside the store influence buying decisions.	[55]

Table 2. Cont.

Construct	Items	Measures	Reference
Macro environmental factors	ECONOMIC	MFA1 COVID-19 has caused recession.	[32]
		MFA2 The recession has affected my household.	[38]
		MFA3 My purchase spending was reduced due to COVID-19.	[14]
		MFA4 I prefer shopping for clothes pre-COVID-19.	
		MFA5 I prefer shopping for clothes due to COVID-19.	
	TECHNOLOGICAL	MFA6 I prefer online shopping for clothes.	[39]
		MFA7 I obtain more information about clothes when online shopping.	[40]
		MFA8 I save more time when online shopping.	[40]
	POLITICAL	MFA9 COVID-19 protocol prevention affects my buying behavior.	
		MFA10 Community quarantine declarations affect my buying decisions.	
		MFA11 Different preventive measures discourage me from shopping for clothing apparel.	
Protection motivation theory	COVID-19	PMT1 I do understand the health risk from COVID-19.	[66]
		PMT2 I do understand possible transmission of COVID-19.	[67]
		PMT3 I am aware of the symptoms of COVID-19.	[68]
		PMT4 I do understand health protocols for COVID-19	[68]
	PERCEIVED SEVERITY	PMT5 I can be infected with COVID-19 when going to malls.	
		PMT6 I can be infected with COVID-19 when buying online.	
		PMT7 COVID-19 can lead to death.	[14]
	SELF-EFFICACY	PMT8 I consider the clothes as sanitized before purchase.	[25]
		PMT9 I can use the face mask as a preventive measure for COVID-19 when shopping.	[69]
		PMT10 Disinfecting my purchase can prevent COVID-19	[70]

Table 2. Cont.

Construct		Items	Measures	Reference
Theory of planned behavior	ATTITUDE TOWARDS BEHAVIOR	TPB1	Purchasing clothing apparel is a good idea.	[44,45]
		TPB2	Purchasing clothing apparel is a wise idea.	[44,45]
		TPB3	Purchasing clothing apparel would be pleasant.	[44,45]
	SUBJECTIVE NORM	TPB4	People around me influence my purchasing behavior.	[45]
		TPB5	My family and friends expect me to purchase clothing apparel.	[44]
		TPB6	I value the opinions and feelings of my family and friends towards clothing apparel.	[44]
	PERCEIVED BEHAVIORAL CONTROL	TPB7	I have the resources to purchase clothing apparel.	[44]
		TPB8	I can participate in the decision-making process of purchasing clothing apparel.	[44]
		TPB9	I am free to choose when purchasing clothing apparel	[44]
	PURCHASE INTENTION	TPB10	I intend to purchase clothing apparel in my next purchase.	[44,71]
		TPB11	I would like to purchase a clothing apparel.	[44]
		TPB12	I would like to recommend to others to purchase clothing apparel.	[44]
		TPB13	There are plenty of opportunities for me to buy a clothing apparel.	[71]

3.3. Statistical Analysis

For the statistical analysis, SEM, a powerful tool that measures the causal relationship of the constructs, was utilized. The first step in the analysis would be to check, empirically and simultaneously, the relationships among the factors of the developed framework using SEM through SPSS Statistics 23 and AMOS version 26 [45]. SEM is broadly acknowledged as an empirical method in the marketing and consumer behavior disciplines for theoretical research and expansion [3]. For the assessment of structural model fit, Incremental Fit Index (IFI), Tucker Lewis Index (TLI), Comparative Index (CFI), Goodness-of-Fit Index (GFI), and Adjusted Goodness-of-Fit Index (AGFI), and Root Mean Square Error of Approximation (RMSEA) were used [72].

4. Results

Structural Equation Model

Figure 4 represents the initial SEM evaluating the Filipino consumers buying behavior during COVID-19. However, the initial SEM shows non-significance in some of the factors, thus leading to a revised SEM. From the figure, it could be seen that there are constructs that are deemed not significant following the suggestion of Hair [45]. Therefore, the initial model was revised, removing the non-significant latent and constructs. Table 3 presents the initial and final descriptive statistics of the factor loading. Following the suggestion of [73] and Hair [45], constructs that had values less than 0.5, together with the non-significant latent (p -value < 0.05), could be removed to enhance the model fit of the study. With that, Figure 5 represents the final SEM evaluating the Filipino consumers buying behavior during the COVID-19 pandemic.

Table 3. Descriptive statistics of the constructs.

Variable	Item	Mean	StD	Factor Loading	
				Initial	Final
Product	MM1	3.24	1.039	0.306	-
	MM2	4.51	0.800	0.380	-
	MM3	3.76	0.996	0.189	-
	MM4	4.19	0.945	0.449	-
Price	MM5	4.51	0.698	0.335	-
	MM6	4.32	0.996	0.324	-
	MM7	3.84	1.032	0.259	-
Place	MM8	4.12	0.940	0.126	-
	MM9	1.40	0.910	0.223	-
	MM10	2.73	1.335	0.052	-
	MM11	2.61	1.171	0.154	-
	MM12	3.45	1.283	0.287	-
Promotion	MM13	3.93	0.918	0.544	0.786
	MM14	3.09	1.173	0.530	0.673
	MM15	3.85	1.001	0.561	0.688
	MM16	4.11	0.930	0.648	0.701
	MM17	4.48	0.749	0.505	0.698
People	MM18	4.11	0.887	0.566	0.715
	MM19	3.69	0.988	0.584	0.705
	MM20	4.16	0.876	0.536	0.703
Process	MM21	4.24	0.814	0.500	0.686
	MM22	4.26	0.865	0.532	0.779
	MM23	4.00	0.893	0.570	0.718
	MM24	4.25	0.792	0.590	0.661
Economic	MFA1	4.49	0.738	0.391	-
	MFA2	3.74	1.067	0.396	-
	MFA3	3.97	1.168	0.585	-
	MFA4	3.87	1.133	0.388	-
	MFA5	2.19	1.083	-0.178	-
Technological	MFA6	2.79	1.163	0.750	-
	MFA7	3.28	1.210	0.550	-
	MFA8	3.65	1.103	0.419	-
Political	MFA9	4.19	1.013	0.835	-
	MFA10	4.15	1.073	0.825	-
	MFA11	3.84	0.487	0.527	-
COVID-19	PMT1	4.87	0.509	0.840	0.840
	PMT2	4.89	0.433	0.840	0.841
	PMT3	4.88	0.452	0.800	0.800
	PMT4	2.86	1.375	0.821	0.819
Perceived severity	PMT5	2.90	1.156	0.733	0.763
	PMT6	4.14	1.087	0.366	-
	PMT7	2.81	1.126	0.610	0.712
Self-efficacy	PMT8	4.12	0.924	0.237	-
	PMT9	2.84	1.260	0.660	0.728
	PMT10	4.49	0.854	0.741	0.756
Attitude	TPB1	3.42	0.952	0.861	0.857
	TPB2	3.10	0.972	0.760	0.762
	TPB3	3.66	0.901	0.651	0.655
Subjective norm	TPB4	3.50	1.149	0.536	0.714
	TPB5	2.63	1.174	0.672	0.742
	TPB6	3.36	1.135	0.536	0.708

Table 3. Cont.

Variable	Item	Mean	StD	Factor Loading	
				Initial	Final
Perceived behavioral control	TPB7	3.75	1.035	0.626	0.635
	TPB8	3.97	0.901	0.789	0.798
	TPB9	4.40	0.786	0.580	0.702
Purchasing intention	TPB10	3.18	1.155	0.818	0.821
	TPB11	3.43	1.159	0.788	0.787
	TPB12	3.29	1.117	0.774	0.775
	TPB13	3.68	1.051	0.406	-

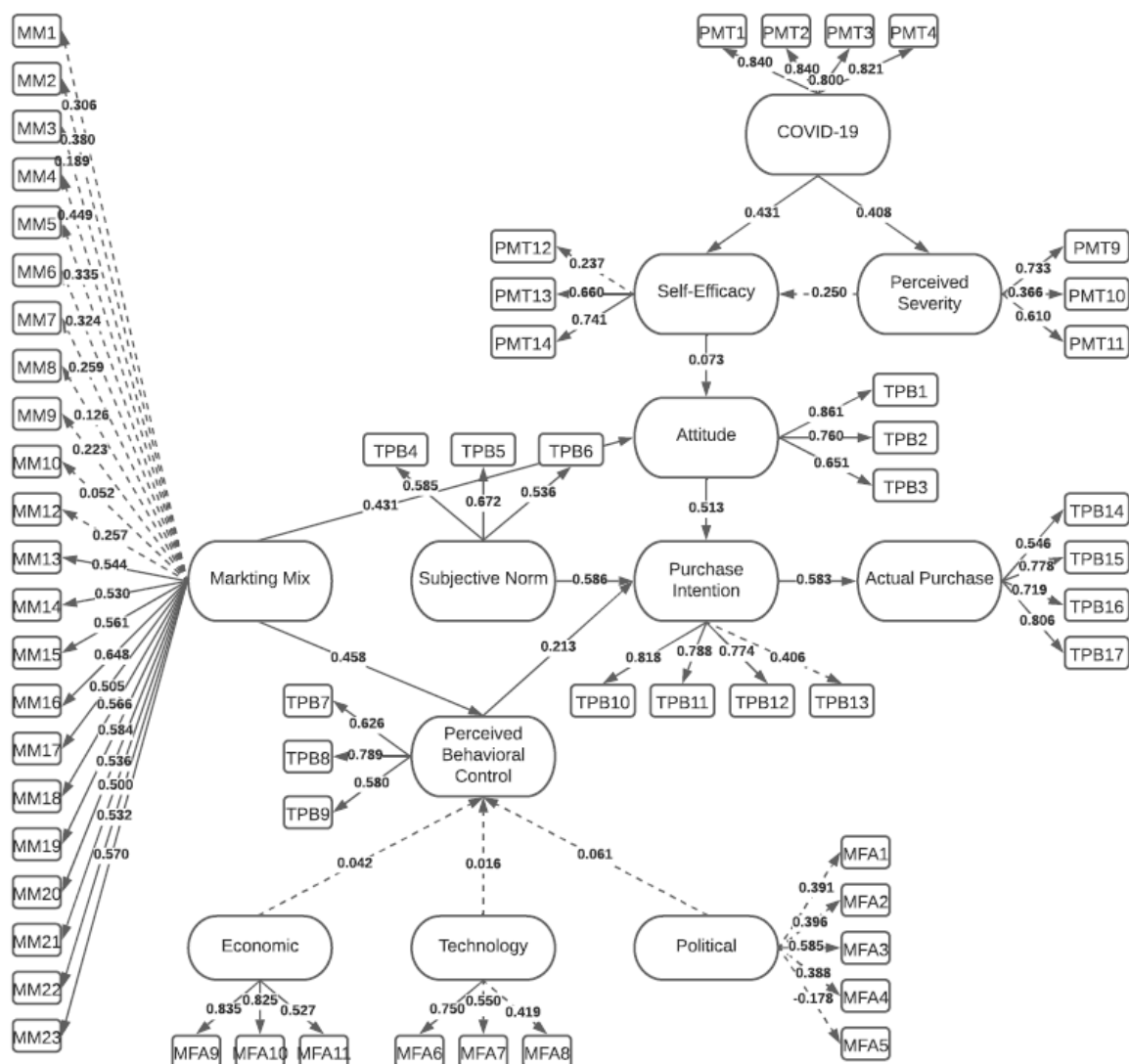


Figure 4. Initial SEM with indicators.

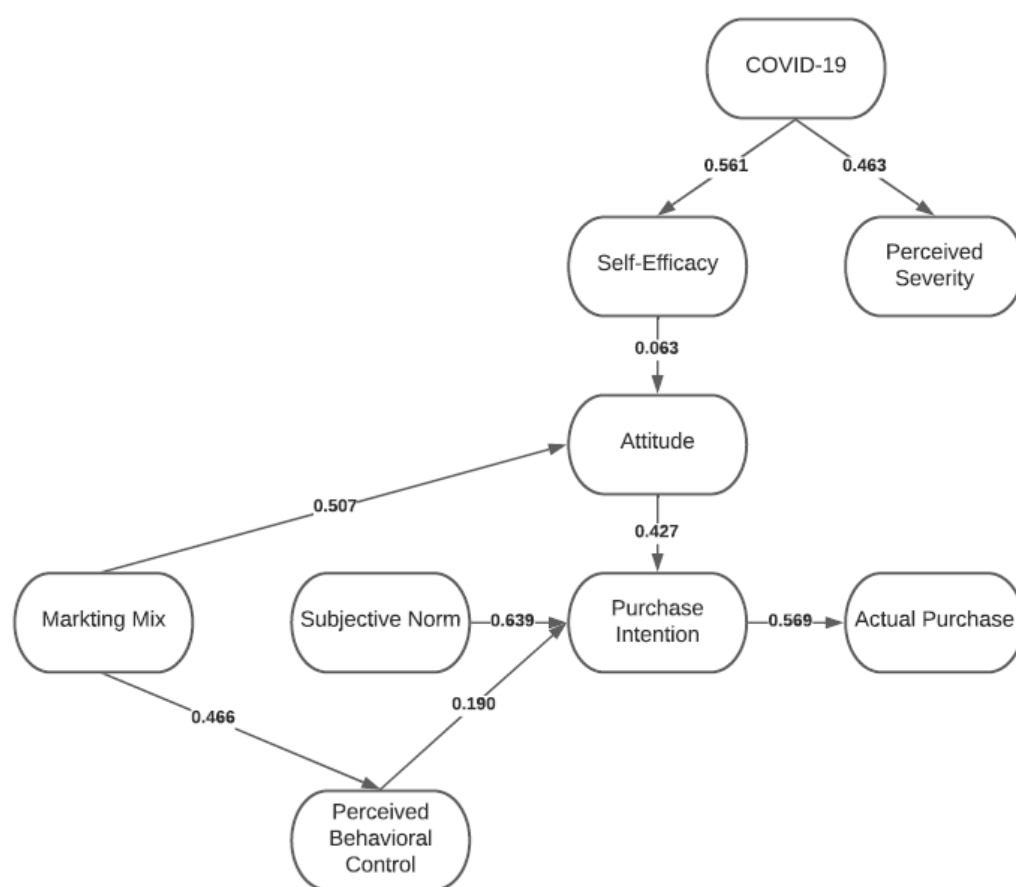


Figure 5. Final SEM for identifying consumer buying behavior.

Presented in Table 4 is the model fit of this study. It can be seen that the Incremental Fit Index (IFI) has a parameter estimate of 0.808, Tucker Lewis Index (TLI) has 0.888, while Comparative Fit Index (CFI) with 0.807, Goodness-of-Fit Index (GFI) has 0.868, and Adjusted Goodness-of-Fit Index (AGFI) has 0.857. The values of the indices were greater than 0.800, which is closer to 1 that are considered adequate values according to [73,74]. Moreover, the Root Mean Square Error of Approximation (RMSEA) was 0.063, which is acceptable since it is lower than the recommended value, 0.07 [74]. The fitness of the model allowed us to accept and reject the hypothesis, as seen in the final SEM in Figure 4.

Table 4. SEM goodness-of-fit measurements.

Goodness-of-Fit Measures of the SEM	Parameter Estimates	Minimum Cutoff	Suggested by
Incremental Fit Index (IFI)	0.808	>0.80	[73,74]
Tucker Lewis Index (TLI)	0.888	>0.80	[74,75]
Comparative Fit Index (CFI)	0.807	>0.80	[73,74]
Goodness-of-Fit Index (GFI)	0.868	>0.80	[75,76]
Adjusted Goodness-of-Fit Index (AGFI)	0.857	>0.80	[75,76]
Root Mean Square Error of Approximation (RMSEA)	0.063	<0.07	[76]

Table 5 presents the validity of the constructs. Following the suggestion of Hair [74], average variance extracted (AVE) will be considered acceptable if the values are greater than 0.5. In addition, Cronbach's alpha and composite reliability (CR) should have a value

greater than 0.70. With that, it could be seen that all constructs showed internal consistency and validity.

Table 5. Composite reliability.

Factor	Cronbach's α	Average Variance Extracted (AVE)	Composite Reliability (CR)
Promotion	0.749	0.505	0.834
People	0.745	0.501	0.835
Process	0.720	0.507	0.804
COVID-19	0.894	0.681	0.895
Perceived severity	0.702	0.545	0.705
Self-efficacy	0.707	0.551	0.710
Attitude	0.799	0.581	0.805
Subjective norm	0.735	0.521	0.765
PBC	0.702	0.511	0.757
Purchasing intention	0.842	0.631	0.837

Table 6 presents the path analysis of the model. From the results, it could be seen that subjective norm (SN) had the highest significant direct effect on purchasing intention (PI) (β : 0.512; p = 0.012) with an indirect effect in actual purchase (β : 0.048; p = 0.012). This supports hypothesis 11. Marketing mix (MM) was seen to have a direct effect in attitude (β : 0.370; p = 0.006; hypothesis 5) and perceived behavioral control (PBC) (β : 0.299; p = 0.012; hypothesis 4). Moreover, it was seen that MM had an indirect effect to PI (β : 0.412; p = 0.002) and actual purchase (AP) (β : 0.226; p = 0.002), following which is the PI being significant and directly affects AP (β : 0.469; p = 0.019; hypothesis 13). In addition, COVID-19 risks and knowledge had a significant and direct effect to self-efficacy (SE) (β : 0.343; p = 0.010) and perceived severity (β : 0.306; p = 0.007). This justifies Hypotheses 1 and 3. It could also be seen that PBC had a direct significant effect to PI (β : 0.087; p = 0.030) and an indirect effect in AP (β : 0.048; p = 0.028), both of which justified hypothesis 12. Interestingly, SE showed a negative direct significant effect in attitude (β : -0.020; p = 0.049), justifying hypothesis 2. This shows that even with the COVID-19 pandemic, customers would still tend to purchase what they want. If they have preventive measures, such as wearing a face mask and following protocols, people will continue with their daily living [69] and even have a positive behavior towards purchasing [29,44]. Interestingly, economic, technological, and political factors were not significant to PBC. This shows that actual products are more preferred by customers when purchasing clothing apparel and that they have the means to shop, even during the COVID-19 pandemic (Hypotheses 7–9). Lastly, attitude was not seen to be a significant factor in PBC (hypothesis 10). The result showed that self-efficacy had the highest contributing factor, and PBC was the main factor leading to the actual purchase of clothing apparel.

Table 6. Path statistics.

No	Variable	Direct Effect	p -Value	Indirect Effect	p -Value	Total Effect	p -Value
1	SN→PI	0.512	0.012	-	-	0.512	0.012
2	PI→AP	0.469	0.019	-	-	0.469	0.019
3	MM→A	0.370	0.006	-	-	0.370	0.006
4	COV→SE	0.343	0.010	-	-	0.343	0.010
5	COV→PS	0.306	0.007	-	-	0.306	0.007
6	MM→PBC	0.299	0.012	-	-	0.299	0.012
7	PBC→PI	0.087	0.030	-	-	0.087	0.030

Table 6. Cont.

No	Variable	Direct Effect	<i>p</i> -Value	Indirect Effect	<i>p</i> -Value	Total Effect	<i>p</i> -Value
8	SE→A	−0.020	0.049	-	-	−0.020	0.049
9	MM→PI	-	-	0.412	0.002	0.412	0.002
10	MM→AP	-	-	0.226	0.002	0.226	0.002
11	PBC→AP	-	-	0.048	0.028	0.048	0.028
12	SN→AP	-	-	0.270	0.012	0.270	0.012

5. Discussion

SEM was utilized to assess the causal relationship between the constructs. Based on the results, consumers are aware and understand the risks and protocols for COVID-19, wherein this could develop different perceptions and behaviors during a pandemic. From the results, it could be seen that subjective norm (SN) had the highest direct significant effect in purchase intention (PI) (β : 0.512; p = 0.012). The customers do feel the impact of parents and friends. It could be seen from the constructs that people around them influence their buying behavior since they are expected to have new clothing apparel. This led to the indirect effect of SN on actual purchase (β : 0.048; p = 0.012).

It could be deduced from the results that the brand image, endorsement, sales/promos, and social media advertisements greatly affect the promotion side of clothing apparel. This is a great finding for different brands of clothing apparel since advertising promotes the different products they have. In addition, people such as the sales personnel showed great effect towards enticing the customers to buy clothing apparels. Companies should take into consideration how enticing the customer would lead to positive buying behavior. Lastly, customers saw that the processes such as maintenance of stock, availability, merchandise displays, even store or website design greatly affects customer's positive buying behavior through the subjective norm. Thus, the indirect effect of MM led to high effect in buying intention (β : 0.412; p = 0.002) and actual purchase (β : 0.226; p = 0.002).

Armitage et al. showed that purchase intention is primarily driven by attitude and subjective norms, which is reflected in the current study's results [49]. This shows how a consumer's intention of purchasing is driven by the influence of its own belief and environment. Thus, it confirms how intentions are developed with consumer's cognition and emotion which results in actual purchase [22]. Moreover, this proves that the result of the current study showed purchase intentions significantly affect actual purchase (β : 0.469; p = 0.019) as the third-highest factor. The indicators showed that consumers are continuously purchasing clothing apparel even during COVID-19. However, it is not the customer's top priority, unlike pre-COVID-19.

The result also showed that COVID-19 had a significant direct effect to self-efficacy (SE) (β : 0.343; p = 0.010) and perceived severity (β : 0.306; p = 0.007). This provides insight into how customers purchase. From the constructs, customers believe that the clothing apparel is sanitized every time, they are protected against COVID-19 when shopping (following the protocol set by the government), and they disinfect the purchases. Even with the COVID-19 pandemic, people still consider clothing apparel as essential, though not the priority. From the constructs, it could be stated that if the customers feel safe, they will continue to purchase clothing apparel. The previous study has shown to significantly predict perceived severity in a pandemic [77]. The indicators under perceived severity show that customers are aware that COVID-19 could lead to death and can be transmitted when going to malls or in public places [15]. On the other hand, the study shows that there is no perceived severity in getting infected when buying online. This can be connected to self-efficacy indicator PMT14, wherein consumers believe that disinfecting purchase could prevent COVID-19 [70].

In line with the result, the infection of COVID-19 led to the negative direct effect of SE on attitude (β : −0.020; p = 0.049). Though customers continue to buy clothing apparels, it is not their top priority during the COVID-19 pandemic. They do not believe that buying more apparel during the COVID-19 pandemic is a wise idea. In contrast to the constructs

of SE, it is interesting that people believe that as long as they are wearing a face mask and the different available preventive measures to the virus [69] are considered, consumers still consider purchasing as a positive behavioral response [44,45].

Marketing mix had a direct significant effect on perceived behavioral control (β : 299; $p = 0.012$), and PBC was seen to have a direct significant effect on purchasing intention (β : 0.087; $p = 0.030$). Lastly, PBC had an indirect effect to actual purchase (β : 0.048; $p = 0.028$). This is an interesting finding in line with the overall results. Based on the constructs, if people have the means and resources to buy clothing apparel, participate in buying clothing apparel, and are free to choose which one to buy, then they have a positive effect on the actual purchase of different clothing apparel. Customers believe in buying clothing apparels, recommend buying clothing apparel, and they have plenty of opportunities to buy clothing apparel.

In building the indicators, the study determined that SN and attitude had the highest factor affecting purchase intention. Moreover, customers' positive perceived behavioral control greatly affected the actual purchase of clothing apparel. Other studies indicated that customer satisfaction has an effect on consumers buying behavior since previous studies showed customer satisfaction is a way to predict consumer's retention that affects consumer's purchases [31,78,79]. However, the findings of the study measured actual purchase behavior from different customers. It is evident that even during the COVID-19 pandemic, people would still purchase and consider clothing apparel as essential but not a priority. Proper advertisement, sales/promotions, and health safety are being considered by customers when purchasing clothing apparel. This could be taken into consideration when creating a marketing strategy in engaging people to continue purchasing clothing apparel despite the COVID-19 pandemic.

The Relation between Consumer Behavior and Open Innovation Dynamics in Clothing Industry

This study has a great impact on business model development in the clothing industry. The COVID-19 pandemic became a challenge for business, especially in clothing industries, wherein the perceived severity and self-efficacy affected attitude leading to buying apparel. The environmental factors [80,81], such as the current state and subjective norms like public and self-consciousness [82,83], have a contributing effect on consumers' behavior. This greatly contributes towards open innovation with regards to the marketing strategies during the COVID-19 pandemic. With the different brands continuously investing in operating in the Philippines, the highlight of their branding as a marketing strategy may be their way of promotion. Similar to the study of Lee and Workman [84], brand loyalty and attachment are attributes that engage consumers to buy a product in the clothing industry. Moreover, Le et al. [85] indicated that the local brands might also capitalize on this as a marketing strategy, looking for ways to promote their own brands in the local market and eventually internationally.

The recommendation to consider the framework created from the study of Tsakalidis and Vergidis [86] may be utilized as a roadmap to redesigning the business process during this unprecedented time. In addition, the results from the study of Patricio et al. [87] may be considered for managing clothing industries. From their results, knowledge accumulation, integration, utilization, reconfiguration, sensing, and seizing may be applied to the clothing industries as well to highlight how they can manage the business during the COVID-19 pandemic. Lastly, the study of Rodionov et al. [88] has considered changes in the business for the digital economy. These studies could be beneficial for redesigning the current business state of the clothing industry.

With the changes that may be made in the clothing industry during the COVID-19 pandemic, the consumers may have a better experience and would lead to the intention to purchase. This would also highlight how a company would be customer-experience-driven from open innovation, similar to the study of Kokins et al. [89]. This could be an advantage for building brand personality in the new normal [90], which could be a pillar of open management during the COVID-19 pandemic. This may also be extended to consider

new product development [91]. Moreover, the marketing difference may also be studied upon by upcoming emerging business or those who plans to take a path in business [92].

6. Conclusions

The continued growth of the global fashion industry and brands is associated with consumers spending, wherein several studies have discussed consumers buying behavior towards clothing apparel with different methods and approaches from different countries. However, clothing apparel buying behavior is not widely discussed during a pandemic despite being one of the industries that are greatly affected. Hence, this led to the purpose of the study—to determine the factors affecting buying behavior of consumers in the Philippines towards clothing apparel during the COVID-19 pandemic. The study consisted of 457 respondents that voluntarily participated and answered an online questionnaire. From the SEM, the results indicated how COVID-19 directly affects perceived severity and self-efficacy while attitude towards behavior is driven by self-efficacy, which shows how changes in current events and environment affect consumer's buying behavior. Moreover, the framework shows that purchase intention is mainly driven by subjective norms and attitudes towards the behavior, which lead to actual purchases. Integrating the PMT and TBP could help practitioners and researchers identify the buying behaviors of Filipino consumers during a pandemic. As discussed, it is evident that during the COVID-19 pandemic, people would still purchase and consider clothing apparel as essential but not a priority. Proper advertisement, sales/promotions, and health safety are being considered by customers when purchasing clothing apparels. This could be taken into consideration when creating a marketing strategy in engaging people to continue purchasing clothing apparel despite the COVID-19 pandemic.

6.1. Theoretical Contributions

The integrated framework of PMT and TPB holistically measured the purchasing behavior of customers for clothing apparel. With the PMT, it was seen to measure the consideration of people of the COVID-19 virus and how they move with it. It could be seen that safety is still the priority of the people as measured in the PMT. Moreover, knowing how to be safe led to positive behavior towards customer's intention. With the extension of TPB with marketing mix (MM), it was a great contribution on how to engage people in buying behavior, purchasing intentions, and actual purchase. Therefore, the inclusion of MM was seen to be a way to create strategies. It could be deduced that the integrated framework did not only consider the behavior but also showed great contribution in creating marketing strategies among clothing apparel. The integrated framework could then be utilized to measure and create strategies for other products and even services.

6.2. Practical Applications with Open Innovation Dynamics

The result of the study is particularly useful for retail brand management. The study highlighted the factors affecting the purchasing behavior of consumers during unprecedented times and identify which attributes to focus on in order to achieve consumer's trust. The framework and findings of the study should be used to guide management and start-up businesses in identifying the positioning and differentiation that the market prefers. This is due to the unique research setting, which identified not only the important attributes that consumers prefer but also the relationship between factors in reaching actual purchase.

For the open innovation dynamics, the SEM showed the decision-making process of consumers behavior that would help marketing managers in creating a marketing strategy to trigger consumer's willingness to purchase. In addition, the study showed how the attitude towards the behavior and subjective norm becomes the driving forces of purchase intention. This can be applied in the planning element of attracting customers. Marketing materials and activities of the company/brand must give emphasis on where the consumers and its environment are focused and invested.

6.3. Limitations and Future Research

The study can be applied and extended for evaluating the fashion industry in other countries. However, it has several limitations. First, regarding the demographic factors, the geographical sampling was not diverse since most of the respondents are from the Luzon region, specifically Metro Manila, while there is only a limited number of respondents from the provinces, especially in the Visayas and Mindanao regions. Additionally, the study was not able to capture the gender identity of the respondents wherein it could have created a huge impact on the understanding of the consumer's buying behavior. Future researchers could improve this by having a more diverse sampling process and making sure that different demographic factors are included in future studies. Moreover, the study was not able to consider and differentiate between online and in-person retail purchases. This would be a great extension of this study by considering the purchasing intention of a person. It is suggested to utilize conjoint analysis [93,94] and clustering like K-Means to clearly show how a person considers purchasing intention. The marketing mix could be highlighted [81,95,96], specifically, the 7Ps, especially the person to clearly determine if online and in-person retail affects the buying behavior or purchase intention.

Second, the study has 14 hypotheses, but only 10 were seen to be significant. Future studies may further investigate these and identify new findings. Furthermore, the study was not able to relate the demographic characteristics of the respondent to its behavior which would have created a deeper explanation of the consumers' buying behavior. Thus, future researchers are encouraged to produce insights on it. Moreover, the creation of questionnaires and chosen attributes are based on the researchers' experience as practitioners in the fashion retail industry and can be improved by future researchers. Practitioners and researchers may further use the findings, questionnaires, and metrics in future research and apply new methods and different theories in order to produce new insights and understanding regarding consumer's buying behavior. Lastly, future researchers may explore consumers' buying behavior after the pandemic or once the vaccines are available to identify how the behavior of the consumers will shift.

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