buying tendency makes an impulsive purchase and later realizes that it was unnecessary or too expensive, they may experience feelings of regret or guilt. These negative emotions and feelings can reinforce their awareness of the potential negative consequences of impulsive buying behavior and make them more cautious about engaging in such behavior in the future.

CHAPTER V

CONCLUSION AND MANAGERIAL IMPLICATION

This chapter covers to the conclusion drawn from the research results, which will be presented and elaborated upon. Additionally, the final section of this chapter offers recommendations based on the study's outcomes.

5.1 Conclusion

Impulse buying in social commerce environment is a recent and vital topic, but still little known particularly in the context of newly emerged in social media. Based on the result of the analysis, there are no significant effect on all moderating hypothesis toward the Urge To Buy Impulsively. This means that Hypothesis H1b, H1c, H1d, H4b, H4c, and H4d all rejected

In this research, the positive significant effect in this research is only the direct effect, even though not all the direct effect has positive significant effect. Only Impulse Buying tendency toward Urge To buy Impulsively, Positive Sentiment toward Urge To Buy Impulsively, and Observational learning toward Urge To Buy Impulsively that has positive significant effect. The rest direct effect such as Negative Sentiment toward Urge To Buy Impulsively and Authenticity toward Urge To Buy Impulsively don't have significant effect.

5.2 Managerial Implication

This study provides valuable insights for practitioners, particularly those in the Social Media Food Influencer and restaurant sectors, as it is one of the few studies that examines the role of social media Food Influencer and interactive factors. First in the Impulse Buying Tendency toward Urge to Buy Impulsively, the impulse buying tendency has a direct and positive significant effect on the urge to buy impulsively. This means that individuals who have a higher tendency for impulse buying are more likely to experience the urge to buy impulsively. In other words, their propensity to make impulsive purchases is a predictor of their likelihood to experience the urge to buy impulsively. The company can make the selling process or transaction process to be as simple as it can, or could make it quickest, in that way, it can facilitate the user or potential buying to fulfil it impulsive buying tendency.

Second is the connection between Observational Learning toward the Urge To Buy Impulsively. Observational learning has a direct and positive significant effect on the urge to buy impulsively. This means that individuals who observe others making impulsive purchases are more likely to experience the urge to buy impulsively themselves. Observational learning refers to the process of acquiring new behaviours or attitudes by observing others and the consequences of their actions. In the context of impulse buying, it means that individuals are more likely to engage in impulsive purchases if they observe others doing so. Therefore, the more an individual observes impulsive buying behaviour, the stronger their urge to buy impulsively becomes. The business owner can improve the relationship between the business, in this context is the food influencer and its community, such as replying comment in the comment section, our using hashtag to improve the spreading of the content in order to cover wider range of the potential buyer.

Lastly, is the effect between Positive Sentiment toward the Urge to Buy Impulsively. Positive sentiment also has a direct and positive significant effect on the urge to buy impulsively. This means that when individuals are exposed to positive information, such as positive reviews or comments, they are more likely to experience the urge to buy impulsively. Positive sentiment refers to the overall positive emotional tone of a message, which can influence an individual's behavior. In the context of

impulse buying, positive sentiment may include positive reviews, endorsements, or recommendations. Therefore, the more positive sentiment an individual is exposed to, the stronger their urge to buy impulsively becomes.

5.3 Research Limitation

While this study provides valuable insights into the phenomenon of impulse buying in social commerce environments, it is important to acknowledge its limitations. First, the study focuses on a specific group of participants and may not be generalizable to other populations. The sample consists of individuals who have engaged in social commerce activities and may not represent the broader population. Therefore, caution must be exercised when generalizing the findings to other contexts or populations.

Second, the study relies on self-reported measures, which may be subject to response biases such as social desirability bias or inaccurate reporting. Participants may have provided answers that they believed were expected of them or may have misunderstood the questions. Future research could employ more objective measures, such as physiological or behavioural measures, to provide a more accurate representation of participants' experiences.

Third, the study focuses primarily on Instagram as a social commerce platform and does not consider other platforms such as Facebook or Twitter. Each platform has its unique features and may elicit different consumer behaviours. Therefore, future research could explore impulse buying behaviour in other social commerce platforms to understand whether the findings are platform-specific or generalize across different platforms.

Fourth, the study relies on a relatively small sample size, which may limit the generalizability of the findings. Future research could employ larger sample sizes to increase the statistical power and robustness of the findings.

Finally, the study is limited by its geographic focus on a specific country. Social commerce environments may differ across different countries and cultures, and therefore, the findings may not apply universally. Future research could explore crosscultural differences in impulse buying behaviour in social commerce environments to provide a more nuanced understanding of the phenomenon.

5.4 Future Research Suggestions

While this study provides valuable insights into impulse buying in social commerce environments, there are several Suggestion for future research that could further advance the understanding of the phenomenon.

First, future research could explore the influence of different types of social media Food Influencer on impulse buying behaviour. The present study focuses on newly emerged social media Food Influencer, but there are other types of Food Influencer, such as traditional celebrities or micro-influencers, who may have different effects on consumer behaviour. Investigating these different types of celebrities could provide a more nuanced understanding of their impact on impulse buying.

Second, future research could examine the role of trust and credibility in impulse buying behaviour. Social commerce platforms rely heavily on user-generated content and recommendations, which can be influenced by fake reviews or sponsored content. Investigating how consumers evaluate the trustworthiness and credibility of these sources of information could shed light on how they make impulse buying decisions.

Third, future research could investigate the effects of different types of social cues, such as scarcity or social proof, on impulse buying behaviour. Social commerce platforms often use these cues to create a sense of urgency or to influence consumer behaviour. Understanding how these cues affect impulse buying behaviour could provide insights into how marketers can effectively use them to increase sales.

Fourth, future research could explore the moderating effects of individual differences, such as personality traits or cultural values, on impulse buying behaviour. Different individuals may be more or less susceptible to impulse buying, depending on their personality traits or cultural backgrounds. Investigating these individual differences could help identify subgroups of consumers who are particularly vulnerable to impulse buying and develop targeted interventions to help them make more informed purchasing decisions.

Finally, future research could explore the ethical implications of impulse buying in social commerce environments. While impulse buying can be beneficial for businesses, it can also lead to negative consequences for consumers, such as overspending or buyer's remorse. Investigating the ethical considerations of impulse buying in social commerce environments could help develop guidelines and policies that promote responsible marketing practices and protect consumers' interests.

BIBLIOGRAPHY

- A. U. Zafar, J. Qiu, Y. Li, J. Wang, M. Shahzad (2019). The impact of social media celebrities' posts and contextual interactions on impulse buying in social commerce
- Ahmed, A. (2016). 'Foodies R Us' brings revolution in taste. Retrieved October 27, 2019.
- Akman, I., & Mishra, A. (2017). Factors influencing consumer intention in social commerce adoption. Information Technology & People.
- Akram, U., Hui, P., Khan, M., Yan, C., & Akram, Z. (2018). Factors affecting online impulse buying: Evidence from Chinese social commerce environment. *Sustainability*, 10(2), 352.
- Alexander P. Schouten, Loes Janssen & Maegan Verspaget (2020). Celebrity vs.

 Influencer endorsements in advertising: the role of identification, credibility,
 and Product-Endorser fit.
- Bilgin (2020). The Effect of Social Media Marketing Activities On Brand Awarness, Brand Image, and Brand Loyalty.

- Bressolles, G., Durrieu, F., & Giraud, M. (2007). The impact of electronic service quality's dimensions on customer satisfaction and buying impulse. *Journal of Customer Behaviour*, 6(1), 37–56.
- Brown, S., Kozinets, R. V., Jr, J. F. S., Meaning, B., Bruhn, M., Eggers, F., et al. (2014). The quest for authenticity in consumption: Consumers' purposive choice of authentic cues to shape experienced outcomes. *Journal of Business Research*, 31(3), 342–363.
- Bryman, A. (2007). Barriers to integrating quantitative and qualitative research. *Journal of Mixed Methods Research*, 1(1), 8–22. https://doi.org/10.1177/2345678906290531.
- Bryman, A., & Bell. (2011). *Business research methods* (3rd ed.). New York: Oxford University Press.
- Busalim, A. H., & Hussin, A. R. C. (2016). Understanding social commerce: A systematic literature review and directions for further research. *International Journal of Information Management*, 36(6), 1075–1088.
- Chan, T. K. H., Cheung, C. M. K., & Lee, Z. W. Y. (2017). The state of online impulse-buying research: A literature analysis. *Information & Management*, *54*(2), 204–217.
- Chen, Y., Lu, Y., Wang, B., & Pan, Z. (2019). How do product recommendations affect impulse buying? An empirical study on WeChat social commerce. *Information & Management*, 56(2), 236–248.
- Chen, J. V., Su, B. C., & Widjaja, A. E. (2016). Facebook C2C social commerce: A study of online impulse buying. *Decision Support Systems*, 83, 57–69.

- Chen, Y., Wang, Q. I., & Xie, J. (2011). Online social interactions: A natural experiment on word of mouth versus observational learning. *Journal of Marketing Research, XLVIII*(April 2011), 238–254.
- Chen, C. C., & Yao, J. Y. (2018). What drives impulse buying behaviors in a mobile auction? The perspective of the stimulus-organism-response model. *Telematics and Informatics*, (February), 0–1.
- Huang, L. T. (2016). Flow and social capital theory in online impulse buying. *Journal of Business Research*, 69(6), 2277–2283.
- Schwarz, A., Rizzuto, T., Carraher-Wolverton, C., Roldan, J. L., & Barrera-Barrera, R. (2017). Examining the impact and detection of the "urban legend" of common method bias. *The DATA BASE for Advances in Information Systems*, 48(1), 93–119.
- Sheldon, P., & Bryant, K. (2016). Instagram: Motives for its use and relationship to narcissism and contextual age. *Computers in Human Behavior*, 58, 89–97.
- Johnson, M. A. (2018). Understanding Impulse Buying: A Literature Review
- Islam, T., Wei, J., Sheikh, Z., Hameed, Z., & Azam, R. I. (2017). Determinants of compulsive buying behavior among young adults: The mediating role of materialism. *Journal of Adolescence*, 61(October), 117–130.
- Javadi, Dolatabadi, Nourbakhsh, Poursaeedi, and Asadollahi. (2011). An Analysis of Factors Affecting on Online Shopping Behavior of Consumers.
- Jeong, E. H., & Jang, S. C. S. (2011). Restaurant experiences triggering positive electronic word-of-mouth (eWOM) motivations. *International Journal of Hospitality Management*, 30(2), 356–366.

- Kaplan, A. M., & Haenlein, M. (2012). The Britney Spears universe: Social media and viral marketing at its best.
- Kim, E. E. K., Mattila, A. S., & Baloglu, S. (2011). Effects of gender and expertise on consumers' motivation to read online hotel reviews. *Cornell Hospitality Quarterly*, 52 (4), 399–406.
- Kim, C., & Yang, S.-U. (2017). Like, comment, and share on Facebook: How each behaviour differs from the other. Public Relations Review.
- Kock, N. (2015). Common method bias in PLS-SEM: A full collinearity assessment approach. *International Journal of e-Collaboration*, 11(4), 1–10.
- Kowalczyk, C. M., & Pounders, K. R. (2016). Transforming celebrities through social media: The role of authenticity and emotional attachment. *The Journal of Product and Brand Management*, 25(4), 345.
- Kumar, V., & Mirchandani, R. (2012). Increasing the ROI of social media marketing.

 MIT Sloan Management Review, 54(1), 55.
- Lee, J. H., Jung, S. H., & Park, J. H. (2017). The role of entropy of review text sentiments on online WOM and movie box office sales. *Electronic Commerce Research and Applications*, 22, 42–52.
- Lee, S. Y. (2020). Online Impulse Buying: Factors Influencing Consumer Behavior in E-commerce.
- Mason, Narcum & Mason (2020). Social media marketing gains importance after Covid-19
- Pine, Bradley J. (2018) "The Role of Authenticity in Consumer Behavior: A Literature Review"

- Poeczea, Ebster, Strauss (2018). Social media metrics and sentiment analysis to evaluate the effectiveness of social media posts.
- Rungsrisawat, Joemsittiprasert, Jermsittiparsert (2018). Factors Determining Consumer Buying Behaviour in Online Shopping.
- Veronika (2013). Motivation of Online Buyer Behavior.
- Verplanken, Bas, (2018). "Understanding Impulse Buying: A Literature Review".
- Yi-Shih Loa, Linb, Hsuc (2016). Motivation for online impulse buying: A two-factor theory perspective.
- Zulfikar, Aprianti, Rachmawati (2018). Digital Marketing and Brand Image to Increase Consumer Purchase Interest.

ATTACHMENT

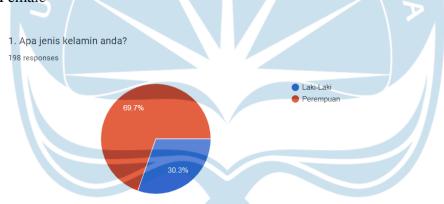
Research Questionnaire

Filter Questions

- 1. Do you use and have an account on social media Instagram?
 - a. Yes
 - b. No
- 2. Do you follow (become the follower) one or more of these Food influencers on Instagram below: @Mgdalenaf, @nexcarlos, @farida.nurhan, @anakjajan, @Dyodoran
 - a. Yes
 - b. No
- 3. Have you ever seen a post from one of the Food influencer accounts on Instagram in the list above:
 - a. Yes
 - b. No

Respondent Demographic

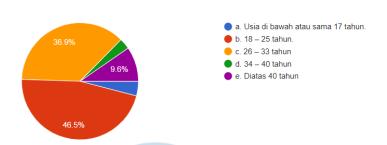
- 1. What is your gender?
 - a. Male
 - b. Female



- 2. How old are you?
 - a. Age under or equal to 17 years
 - b. 18 25 years old
 - c. 26 33 years old
 - d. 34 40 years old
 - e. Over 40 years old

2. Berapa usia anda?

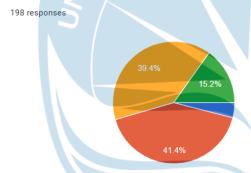
198 responses

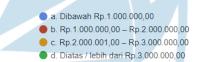


- 3. How much is your income / pocket money within 1 month?
 - a. Under Rp1.000.000,00
 - b. Rp1.000.000,00 Rp2.000.000,00
 - c. Rp2.000.001,00 Rp3.000.000,00
 - d. Above Rp3.000.000,00
 - 3. Berapa pendapatan anda / uang saku dalam kurun waktu 1 bulan?

Сору

□ Сору





- 4. How long do you use or access Instagram in 1 day?
 - a. Less than 30 minutes
 - b. 31-60 minutes
 - c. 61 120 minutes
 - d. More than 120 minutes



Research question

Researchers use a scale of 1 - 5. Value 1 to strongly disagree, while a value of 5 if the respondent feels strongly agree with the statements in the questionnaire. provide your response below based on your experience on the Instagram food influencer account that you follow.

Strongly Disagree	disagree	neutral	agree	Strongly Agree
1	2	3	4	5

	GL 4			option	1	
no	Statement	1	2	3	4	5
1	I take Food Influencers' suggestions on social media					
_	Instagram to be serious.					
2	Food influencers' posts on Instagram about various places					
	to eat seem real to me.					
3	The opinions of food influencers on Instagram are reliable					
4	The comments column on food influencer posts is great.					
5	Reviews of various products featured on food influencer					
3	posts are good.					
6	The food influencer's comments column is amazing.					
7	Reviews from Food Influencers are noteworthy.					
8	Reviews of various posts generated by food influencers are					
0	negative and unsatisfactory					
9	Feedback about the restaurant is unflattering and negative.					
10	The comment column on food influencer posts is					
10	unsatisfactory.					
11	The quality of the review content on food influencer posts					
11	is unsatisfactory.					

10	It is easy for me to observe that food influencer posts are												
12	liked by many people.												
13	I observed that the number of likes on food influencer posts												
13	is good.												
14	I observed that the number of likes associated with various												
14	products is large.												
15	It is easy to observe that food Influencer posts are reviewed												
13	by a large number of people.												
16	I see a lot of comments on various food influencer posts.												
17	I see tons of reviews from Food Influencers for various												
1/	products.												
18	When I use Instagram, I often buy things spontaneously.												
19	When I use Instagram, I often buy things online without a												
19	second thought.												
20	"I saw it, I bought it" describes how I buy goods/products.												
21	When using Instagram, sometimes I am a bit reckless about												
21	what I buy.	上											
22	When I see a Food Influencer post, I feel a sudden urge to	Ty.											
22	order or visit a restaurant that I had never planned before	\ \											
23	When I see a Food Influencer post, I have the urge to buy		50										
23	goods/products that are irrelevant to my shopping goals.	\nearrow											
24	When I see a Food Influencer post, I have a tendency to buy		P										
4	beyond my shopping list.												

Raw Data

											О	О	О	О	О	О	I	I	I	I	U	U	U
			P	P	P	P	N	N	N	N	L	L	L	L	L	L	В	В	В	В	T	T	T
A	A	A	S	S	S	S	S	S	S	S	L	L	L	C	C	C	T	T	T	T	В	В	В
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Appendix A. Result of Data Processing

	Outer loadings
A1 <- aunthenticity	0.936
A2 <- aunthenticity	0.917
A3 <- aunthenticity	0.924
IBT1 <- Impulse Buying Tendency	0.927
IBT2 <- Impulse Buying Tendency	0.935
IBT3 <- Impulse Buying Tendency	0.930
IBT4 <- Impulse Buying Tendency	0.939
NS1 <- Negative Sentiment	0.905
NS2 <- Negative Sentiment	0.916
NS3 <- Negative Sentiment NS4 <- Negative Sentiment OLC1 <- Observational learning OLC2 <- Observational learning OLC3 <- Observational learning OLL1 <- Observational learning OLL2 <- Observational learning	0.892
NS4 <- Negative Sentiment	0.920
OLC1 <- Observational learning	0.933
OLC2 <- Observational learning	0.936
OLC3 <- Observational learning	0.908
OLL1 <- Observational learning	0.921
OLL2 <- Observational learning	0.939
OLL3 <- Observational learning	0.925
PS1 <- Positive Sentiment	0.933
PS2 <- Positive Sentiment	0.906
PS3 <- Positive Sentiment	0.920
PS4 <- Positive Sentiment	0.925
UTBI1 <- Urge To Buy Impulsively	0.917
UTBI2 <- Urge To Buy Impulsively	0.937
UTBI3 <- Urge To Buy Impulsively	0.937
aunthenticity x Positive Sentiment -> aunthenticity x Positive Sentiment	1.000
Impulse Buying Tendency x Observational learning -> Impulse Buying Tendency x Observational learning	1.000
Impulse Buying Tendency x Negative Sentiment -> Impulse Buying Tendency x Negative Sentiment	1.000
aunthenticity x Negative Sentiment -> aunthenticity x Negative Sentiment	1.000
Impulse Buying Tendency x Positive Sentiment -> Impulse Buying Tendency x Positive Sentiment	1.000
aunthenticity x Observational learning -> aunthenticity x Observational learning	1.000

Attachment figure 1. Outer loading

	Cronbach's alpha	Composite reliability (rho_a)	Composite reliability (rho_c)	Average variance extracted (AVE)
Impulse Buying Tendency	0.950	0.950	0.964	0.869
Negative Sentiment	0.929	0.930	0.950	0.825
Observational learning	0.967	0.968	0.973	0.859
Positive Sentiment	0.941	0.942	0.957	0.849
Urge To Buy Impulsively	0.922	0.923	0.951	0.865
aunthenticity	0.916	0.917	0.947	0.857

Attachment figure 2. Cronbach's alpha and AVE

	Impulse Buying Tendency	Negative Sentiment	Observational learning	Positive Sentiment	Urge To Buy Impulsively	aunthenticity
A1	0.769	0.755	0.647	0.799	0.654	0.936
A2	0.753	0.730	0.693	0.821	0.703	0.917
A3	0.770	0.754	0.665	0.771	0.703	0.924
IBT1	0.927	0.762	0.558	0.764	0.758	0.741
IBT2	0.935	0.775	0.561	0.746	0.782	0.768
IBT3	0.930	0.760	0.600	0.782	0.803	0.780
IBT4	0.939	0.790	0.618	0.759	0.798	0.789
NS1	0.759	0.905	0.628	0.768	0.656	0.750
NS2	0.745	0.916	0.624	0.731	0.670	0.721
NS3	0.727	0.892	0.658	0.745	0.699	0.732
NS4	0.775	0.920	0.621	0.741	0.711	0.729
OLC1	0.563	0.639	0.933	0.715	0.714	0.661
OLC2	0.586	0.638	0.936	0.718	0.702	0.673
OLC3	0.595	0.664	0.908	0.719	0.697	0.669
OLL1	0.522	0.608	0.921	0.710	0.669	0.639
OLL2	0.595	0.643	0.939	0.746	0.709	0.695
OLL3	0.623	0.682	0.925	0.736	0.733	0.684
PS1	0.772	0.759	0.717	0.933	0.682	0.799
PS2	0.780	0.756	0.704	0.906	0.711	0.832
PS3	0.755	0.771	0.702	0.920	0.642	0.778
P\$4	0.703	0.740	0.757	0.925	0.651	0.761
UTBI1	0.728	0.626	0.732	0.670	0.917	0.675
UTBI2	0.800	0.727	0.705	0.685	0.937	0.693
итвіз	0.821	0.747	0.686	0.683	0.937	0.705
aunthenticity x Positive Sentiment	-0.405	-0.428	-0.477	-0.536	-0.360	-0.465
Impulse Buying Tendency x Observational learning	-0.269	-0.299	-0.489	-0.431	-0.299	-0.303
Impulse Buying Tendency x Negative Sentiment	-0.502	-0.429	-0.348	-0.477	-0.426	-0.400
aunthenticity x Negative Sentiment	-0.382	-0.367	-0.365	-0.471	-0.329	-0.425
Impulse Buying Tendency x Positive Sentiment	-0.451	-0.450	-0.474	-0.554	-0.406	-0.441
aunthenticity x Observational learning	-0.282	-0.306	-0.496	-0.440	-0.326	-0.294

Attachment figure 3. Discriminant Reliability

	sso	SSE	Q² (=1-SSE/SSO)
Impulse Buying Tendency	792.000	792.000	0.000
Negative Sentiment	792.000	792.000	0.000
Observational learning	1,188.000	1,188.000	0.000
Positive Sentiment	792.000	792.000	0.000
Urge To Buy Impulsively	594.000	184.836	0.689
aunthenticity	594.000	594.000	0.000

Attachment figure 4. Construct cross-validated redundancy

	f-square
Impulse Buying Tendency -> Urge To Buy Impulsively	0.448
Negative Sentiment → Urge To Buy Impulsively	0.002
Observational learning -> Urge To Buy Impulsively	0.461
Positive Sentiment -> Urge To Buy Impulsively	0.070
aunthenticity -> Urge To Buy Impulsively	0.000

Attachment figure 5. F-Square

	R-square	R-square adjusted
Urge To Buy Impulsively	0.827	0.817

Attachment figure 6. R- Square

	Original sample (O)	Sample mean (M)	Standard deviation (STDEV)	T statistics (O/STDEV)	P values
Impulse Buying Tendency -> Urge To Buy Impulsively	0.682	0.651	0.127	5.354	0.000
Negative Sentiment → Urge To Buy Impulsively	0.036	0.056	0.074	0.487	0.627
Observational learning -> Urge To Buy Impulsively	0.546	0.540	0.075	7.304	0.000
Positive Sentiment -> Urge To Buy Impulsively	-0.282	-0.234	0.106	2.656	0.008
aunthenticity -> Urge To Buy Impulsively	0.020	-0.014	0.106	0.187	0.851
Impulse Buying Tendency x Observational learning -> Urge To Buy Impulsively	0.159	0.156	0.184	0.860	0.390
aunthenticity x Negative Sentiment -> Urge To Buy Impulsively	-0.104	-0.084	0.178	0.580	0.562
aunthenticity x Positive Sentiment -> Urge To Buy Impulsively	0.224	0.221	0.178	1.254	0.210
Impulse Buying Tendency x Negative Sentiment -> Urge To Buy Impulsively	-0.036	-0.047	0.154	0.236	0.814
aunthenticity x Observational learning -> Urge To Buy Impulsively	-0.126	-0.138	0.158	0.797	0.425
Impulse Buying Tendency x Positive Sentiment → Urge To Buy Impulsively	-0.110	-0.100	0.193	0.569	0.569

Attachment figure 7. Mean, stdev, P values, and T values

