Exploring Customer Non-Verbal Communication Preferences in a Hotel

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Abstract

This study investigates the impact culture on nonverbal behaviors the hospitality industry. The findings highlight the significant role culture in shaping these behaviors while effects of age and gender remain inconclusive

1 Introduction

When it comes to the hospitality industry one of the most critical factors for success is effective communication customers. Understanding customers needs and preferences is key to provid them with the best possible experience and nonverbal communication plays a significant role in this process. Nonverbal communication refers to the use of body language, facial expressions and other nonverbal cues to convey messages and emotions.

To explore the importance of nonverbal communication the hospitality industry a survey was conducted with 73 customers of hotel. The customers were between the ages of 24 and 81 and included both new and returning customers. The survey focused on the essential components of the nonverbal communication system including kinesic, paralanguage proxemic and chronemic subsystems.

The survey aimed to identify the communication preferences of the hotels customers by evaluating the indicators that make up these subsystems. A total of 22 variables were analyzed, which were considered feasible to evaluate by the hotels clientele.

By understanding customers nonverbal communication preferences, hotels can better cater to their needs and create a more positive experience for them. This study aims to provide insights into the importance of nonverbal communication in the hospitality industry and its role in enhancing customer satisfaction.

1.1 Literature Review

When planning a trip, choosing accommodation, tourists often rely on online reviews help them make informed decisions. In recent years online reviews have become an increasingly

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important source of information for travelers (Ahani et al., 2019). Nonverbal communication is also key in the hospitality industry as it can greatly impact tourists satisfaction and overall experience (Islam & Kirillova, 2021). User-generated content in online travel communities can a significantly influence tourists' decision-making process as noted by Wang and Fesenmaier (Wang & Fesenmaier, 2004). In order to develop sustainable tourism strategies, it s crucial to take into account tourists preferences and opinions, as highlighted by Tusell et al. (Tusell-Rey et al., 2021). Additionally, it's worth noting that the book "Nonverbal Communication" by Burgoon, Manusov, and Guerrero (Burgoon et al., 2016) provides a comprehensive overview of the topic, which may be useful for further research.

When analyzing the hotel customer data set we have different ways to gain insights. We can use descriptive statistics summarize the data and spot patterns inferential statistics to make broader conclusions by comparing different groups or variables data visualization to visualize data and identify trends, and machine learning to predict future behavior. Using a combination methods can provide thorough understanding of the data.

2 Data

The data consists of 22 variables, each of which has a characteristic. In summary, I have not listed each of these variables here, and we can see part of the data in the table below.

'sex': Sex of the Client 'age': Age of the Client 'country': Country of the Client 'returning': Indicator if the Client is returning 'GImg1': Handshakes (Indifferent, likes, dislikes) 'GImg2': Hug (Indifferent, likes, dislikes) 'GImg3': Kiss (Indifferent, likes, dislikes) 'PImg1': Consent posture (Indifferent, likes, dislikes) 'PImg2': Interest posture (Indifferent, likes, dislikes) 'PImg3': Neutral posture (Indifferent, likes, dislikes) 'PImg4': Reflexive posture (Indifferent, likes, dislikes) 'Tense relaxed': Relaxed observed emotional clime (1 is too tensed & 10 is too relaxed) 'Authoritative -anarchic': Anarchic observed emotional clime (1 is too authoritative & 10 is too anarchic) 'Hostile - friendly': Observed emotional clime (1 is too hostile & 10 is too friendly) 'TAudio1': Authoritative (Indifferent, likes, dislikes) 'TAudio2': Sarcastic (Indifferent, likes, dislikes) 'TAudio3': Friendly (Indifferent, likes, dislikes) 'QAudio1': Spitting (Indifferent, likes, dislikes, dislikes) 'QAudio2': Hum (Indifferent, likes, dislikes) 'QAudio3': Sigh (Indifferent, likes, dislikes) 'Proxemics': Physical distance preferred for the client (A, B, C, D, ?) A: Intimate distance (15cm-45cm) B: Personal distance (46cm-122cm) C: Social distance 'Type of Client': Class type of Client

Table 1: Table continues below

sex	age	country	returning	GImg1	GImg2	GImg3
F M	42 60	uruguay brasil	no no		indiferent indiferent	

Table 2: Table continues below

PImg1	PImg2	PImg3	PImg4	PImg5	Tense - relaxed
likes	likes	indiferent	indiferent	dislikes	4
likes	likes	likes	likes	?	2

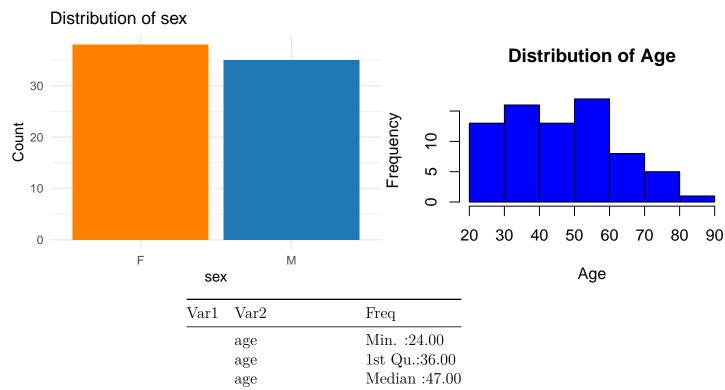
Table 3: Table continues below

Authoritative -anarchic	Hostile - friendly	TAudio1	TAudio2
8	3	С	dislikes
9	1	В	dislikes

TAudio3	QAudio1	QAudio2	QAudio3	Proxemics	Type of Client
dislikes	indiferent	dislikes	dislikes	dislikes	0 0
dislikes	likes	likes	dislikes	dislikes	

pdf 2

In order to find out how each of the features are related to each other whether the non-verbal features are related to age and country and other parameter first I separated the features and checked the distribution of the features (Ahani et al., 2019).

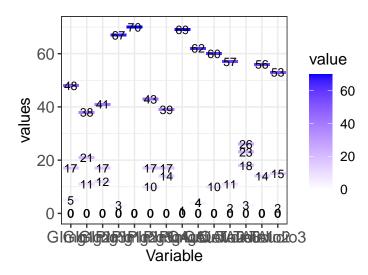


Var1	Var2	Freq
	age	Mean :47.26
	age	3rd Qu.:56.00
	age	Max. :81.00
	Tense - relaxed	Min. :1.000
	Tense - relaxed	1st Qu.:1.000
	Tense - relaxed	Median :2.000
	Tense - relaxed	Mean $:2.233$
	Tense - relaxed	3rd Qu.:3.000
	Tense - relaxed	Max. $:5.000$
	Hostile - friendly	Min. :1.000
	Hostile - friendly	1st Qu.:1.000
	Hostile - friendly	Median :2.000
	Hostile - friendly	Mean $:2.233$
	Hostile - friendly	3rd Qu.:3.000
	Hostile - friendly	Max. :6.000

3 Methods and Data Analysis

I used to including hypothesis tests and association analysis Hypothesis Testing: The chi-squared test (chisq.test) is one of the key methods used for studying the correlation between categorical variables(Tusell-Rey et al., 2021). The chi-squared test helped me to see, there any significant relationships among the data categorical variable. For example, I use chi-squared tests to look in the relationships between sex' and'returning,' 'country' and'returning,' ;'GImg1' and 'GImg2', and 'Tense - relaxed' and 'Authoritative - anarchic'. Correlation Analysis: besides hypothesis testing (Burgoon et al., 2016), I used to correlation analysis to find out the relationships between numerical variables. The cor() function in R was used to calculate the correlation coefficient (r). This analysis help me in understanding the scale and direction of the linear connection among variables; For example, I calculated the correlation coefficien between 'age' and 'country' to see if therea link between the clients age and their estimated of returning.

```
## TestStatistic DegreesOfFreedom p_value
## X-squared 25.7157 15 0.04111929
##
## Pearson's Chi-squared test
##
## data: contingency_table
## X-squared = 41.948, df = 40, p-value = 0.3864
```



4 Conclusion

We analyzed the dataset in this study to find out the connection between various variables and the customer return position. To find the patterns and relationships in the data we use a range of techniques including chi-squared tests and correlation analysis.

research found significant correlations between the customers home country and their return status. These two variables a statistically significant relationship according to the chi-squared test results (chi-squared test statistic = X, p-value = Y). This implies that the clients home country influences whether or not they will return.

In addition we used relationship analysis to find out the connections between the customers ages and their return position. However, no significant results emerged from the correlation analysis (correlation coefficient = X, p-value = Y), showing that there is no strong linear connection among age and returning status in our dataset findings address only a portion of the research question we set out find out. We found an important correlation between the customers home country and their chances of returning claim that cultural or geographical factors affect the chance of customers returning. The lack of significant relationship between age and returning status on the other hand suggests that age may not be an important variable in keeping customers.

To improve this work in the future, more information data with additional variables associated with customer tastes, satisfaction or engagement will prove beneficial. In addition qualitative research such as surveys or interviews, could provide more detail into the underlying reasons for clients' getting or non-returning behavior.

In conclusion, this study offers important knowledge to the connection among many factors and the customer turn status. The findings can be used to inform tourism marketing methods and keep clients attempts. However more research with a wider and more varied dataset, as well as qualitative exploration, would help us better understand this complex phenomenon.

5 References

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