Online dating apps as a marketing channel: a generational approach

Paulo Rita
NOVA Information Management School (NOVA IMS), Universidade NOVA de Lisboa, Portugal
Ricardo Filipe Ramos
Instituto Universitário de Lisboa (ISCTE-IUL), ISTAR-IUL, Lisboa, Portugal and CICEE, Universidade Autonoma de Lisboa, Lisboa, Portugal
Sérgio Moro
Instituto Universitário de Lisboa (ISCTE-IUL), ISTAR-IUL, Lisboa, Portugal, and Marta Mealha and Lucian Radu
Departamento de Ciências e Tecnologias da Universidade Autónoma de Lisboa, Universidade Autonoma de Lisboa, Lisboa, Portugal

Abstract

Purpose – This study aims to understand if an online dating app is considered an acceptable channel to conduct advertising activities and understand the differences between Generations X, Y and Z for such acceptance.

Design/methodology/approach – A total of 411 Tinder users’ reactions were obtained and analyzed using text mining to compute the sentiment score of each response, and a Kruskal–Wallis H test to verify if there are statistical differences between each generation.

Findings – The results showed positive acceptability toward the marketing campaign on Tinder, especially Z Generation. Nevertheless, the statistical analysis revealed that the differences between each generation are not statistically significant.

Research limitations/implications – The main limitation relates to the fact that the participants, during the data collection, revealed their identification, perhaps leading to acquiescence bias. In addition, the study mainly covered the male population. A balanced sample would be positive to examine any possible differences between gender.

Practical implications – Results provide an essential indication for companies regarding their marketing activities conducted on Tinder to fully exploit the possibility of using Tinder as an alternative and valuable channel to conduct marketing activities.

Originality/value – Up until now, no studies tried to understand the effect of a marketing activity online on an online dating app.

Keywords Social media, Marketing communication channel, Online dating apps, Tinder, Text mining, Sentiment analysis

Paper type Research paper

1. Introduction

In a world driven by electronic word-of-mouth based on Social Media (SM) platforms, marketers have taken it into an advantage to procure new relationships between brands, potential customers and developing existing ones (Litterio et al., 2017). On this less tangible and sophisticated era, individuals have changed their roles as consumers, since they take an
autonomous part in seeking information about products/services of their interest, making it possible to acquire nearly everything with a simple “click” (Ramos et al., 2019). On a global scale, there are 2.25 billion users of SM daily (Statista, 2017). The average adult (18+) spent about 2 h and 25 min a day in navigating on SM in 2017, with Generation Y taking the most significant portion of that daily usage, with around 3 h and 72 min (Statista, 2017). This shift in user behavior entails companies to reconsider their marketing strategies inside the digital world, by turning their head to relationship-based interactions with their target market, to improve marketing engagement (Stojanovic et al., 2018). As a result, marketers have transformed their roles. In this paradigm, there are two leading SM platforms marketers tend to focus on when they decide to invest in digital marketing: Facebook and Instagram (Voorveld et al., 2018).

Nevertheless, the increasing number of media channels turned the media effectiveness challenging. Time and attention have turned into a rare assent for users, and the number of alternative ways of communication influence the quantity and type of communication (Hartemo, 2016). Although advertising has a positive effect and visibility on brand equity (Abril and Rodriguez-Cánovas, 2016), it has become less efficient. In a situation of high quantity, there are too many ads concentrated in one channel claiming for the users’ attention, putting in jeopardy the effects of memory, behaviors toward the ad, attitude, advertised product and even the channel itself where the communication occurs (Rejón-Guardia and Martínez-López, 2014).

The main reason to download a dating app is due to its mass marketing popularity and peer influence (LeFebvre, 2017). The game-like swiping characteristic of these apps makes its navigation almost addictive, resulting in placing them into the entertaining app category (Sumter et al., 2017). The age group of 25–34 holds a substantial share of online dating activity with a value of 42.2%, with a total of 279.2 million worldwide users in 2017, and a forecast to reach 331.3 users by 2022 (Statista, 2018).

Although a few marketing campaigns have been executed on dating apps, academia has not developed efforts to understand the receptivity of a dating app as a marketing tool. Therefore, the purpose of this investigation is to explore the willingness to receive intrusive marketing communications through a dating app, and if there are statistical differences between generations for its acceptance. To achieve this objective, 411 user reactions were collected from Tinder after the launch of a marketing campaign, providing the corpus for analysis. Sentiment analysis was adopted to rank the responses, thus enabling us to understand the types of user feedback toward marketing communication and a Kruskal–Wallis H test to comprehend if there are statistical differences between each generation. Although the creative aspect of marketing communication is not considered, from a marketer’s perspective, it is crucial to understand how dating apps users react facing marketing communication, considering that the channel influences its effect.

Using generational cohorts permit an additional understanding of the users’ response toward a marketing campaign, as each group is perfectly delimited by a particular period and involves people who were born within a specific period, with similar values, experiences and priorities (Bento et al., 2018). The period when the audience was born can provide indicators of target groups and interests and particular insights of each generational microculture.

By uncovering the users’ reactions, this paper intends to bring valuable insights for scientific literature and marketers by understanding if an online dating app is an appropriate channel to conduct marketing activities since there is a positive correlation between channel acceptance and ad attitude (Bakr et al., 2019), and if audience generation is a determinant for its acceptance.

2. Literature review

2.1 Social media advertising

The reason marketers have begun to consider SM as one of the most valuable marketing channels is their inexpensive characteristic to engage and communicate with a worldwide
audience (Ashley and Tuten, 2015; Jaakonmaki et al., 2017). Firms who take advantage and invest in SM marketing consider SM an ongoing corporate communication channel more effective than company-sponsored messages (Clark et al., 2017). SM aims to generate content engaging enough to lure social network users into interacting with them to create digital exposure (Ramos et al., 2020). Hence, the opportunity to connect, create and bring customer value became higher and much easier to obtain (Vrontis et al., 2017).

SM networks have become indispensable, bringing up a revolution in how SM affects peoples’ views and participation in political and civil life, in terms of marketing campaigns surrounding political and social causes (Boulianne, 2015). Promotion via SM channels has become a technique pursued by marketers, by a single share of a picture or by creating a contest to win a prize or a reward (Királová and Pavlíček, 2015). Due to humans’ relationship toward their own unconscious emotions, marketing campaigns whose backbone is to reach consumers’ vulnerability and their emotional side, turn almost immediately into success, since such feelings related to advertising get effortlessly retained on the audience’s mind (Hudson et al., 2015; Schivinski and Dabrowski, 2016). Such content influences consumers’ willingness to recommend and comment on experiences with the advertising to their peers (Hudson et al., 2015).

Consumers’ emotions respecting SM advertising can differ, as seen in Knoll’s (2016) review of several studies regarding users’ attitudes toward SM advertising, which emphasized some of the most valuable conclusions obtained. In essence, advertising can be bothersome, especially when disrupting an activity that was being executed on a SM platform. The participants noted that advertising, which features nothing more than a simple link, is often not engaging; thus, these authors recommended that advertising on SM should be mostly concerned about interaction, enjoyment and co-creation between companies and the SM users (Sashittal et al., 2012; Soares and Pinho, 2014). Moreover, studies such as Yang’s (2014) and McCoy et al.’s (2017) regarding contrary acceptance toward advertising on social platforms, concluded that ads invasive and distractive play a negative influence regarding users’ attitudes toward them. However, such a consequence takes a much smaller impact compared to the positive influence resulted from advertising’s entertainment aspect. Expanding with time, SM users tend to be more accepting of the vast amount of advertising, coupled with the need to create credible, entertaining, interactive, trust-worthy and personalized marketing activities (Alalwan et al., 2017; Knoll, 2016).

Furthermore, several studies determined the highly positive value of entertainment’s influence on positive attitudes toward SM advertising: when comparing to information, entertainment took up four times more influence (Knoll and Matthes, 2017; Saxena and Khanna, 2013). Set on Killian et al. (2015) research, several senior managers responsible for the digital planning of their respective companies were asked to recognize the four key customer engagement strategies regarding SM networks’ purposes. Once more, entertainment was claimed to be the most critical factor, as entertaining SM activity quickly enhances users’ engagement and curiosity.

Marketers must invest their time in being attentive to feedback and emotions are given by users on their SM platforms, since marketing and advertising should be a result of a constant co-creation process and firms must be conscious of their audiences’ deliberation toward their marketing activities (Bernabé-Moreno et al., 2015; Hartmann et al., 2018).

2.2 Online marketing campaign effect on different generations

For understanding the effectiveness of online advertising, it is relevant to understand the users’ responses toward them. In this context, consumer motivation plays an essential role in the online environment since it affects how advertising is perceived on social media (Lin and Kim, 2016). Need, drive and desire are psychological states that reveal the way a user processes information, makes decisions and is involved. User response toward advertising is
associated with the motivation that drives a user to find information or good deals on the internet and make a purchase (Zhang and Mao, 2016). For instance, the perceived usefulness of a Facebook ad revealed to be a significant predictor of attitude toward advertising and product purchase intention (Lin and Kim, 2016).

Similarly, the effectiveness of an ad depends on the possibility of engagement before the action is taken. An entertaining ad will have a positive reaction toward an online ad (Zhang and Mao, 2016). For that, it is essential that ads are interactive, appealing and permit a direct virtual experience. The ad personalization also plays a vital role in response to an ad. Moderated personalized ads have increased the click-through intention and click-through rate on online ads when compared to a nonpersonalized ad. Click-through rates also increase when an ad meets user interests (Boerman et al., 2017). Brand consciousness also has a relevant impact on an online ad, influencing users’ attitudes toward it, affecting their behavioral response (Boateng and Okoe, 2015). Clearly, there is a positive relationship between users’ attitudes toward an online ad and their behavioral responses.

Generation X includes those who were born between 1965 and 1980 (Kitchen and Proctor, 2015) with specific characteristics that have an impact on the perception of online marketing campaigns. Although this generation is not acquainted with the new technologies, they tend to be more responsible when using SM platforms, but, on the other hand, are more effectively targeted by marketers through advertisements. This generation is not susceptible to many factors within marketing and tends to ignore online advertising due to the lack of interest, they are not significantly influenced, and their purchasing behavior is not determined by them (Slootweg and Rowson, 2018). Nevertheless, this generation frequently uses the internet as a source of information and tend in general to be participative in online activities. A significant majority uses online social media on a daily basis (Kitchen and Proctor, 2015).

Generation Y (born between 1981 and 1995) (Kitchen and Proctor, 2015), also known as millennials, have a particular way to respond to online advertisements. Since they are digital natives, their interaction with SM is more natural and intuitive, and they share and consume content actively on SM. They are both consumers and producers of information and are more likely to spread marketing messages than Generation X (Bento et al., 2018). This cohort is exceptionally tolerant, give especially attention to social responsibility campaigns and promotion campaigns. This generation spends a considerable amount of time on SM platforms and considers the online environment more trustworthy and safer, making them easy to target using online marketing strategies. However, they are suspicious regarding marketing tactics (Lissitsa and Kol, 2016). Nevertheless, there are few ways in which this generation is consistently affected by marketing since they highly value opinions from others online. Their preference for online advertising is spots and clips broadcast on YouTube, game advertising and pop-up ads from websites (Smith, 2011).

Generation Z is concerned with a target group of people born after 1995 (Kitchen and Proctor, 2015). This generation is more SM savvy and with high levels of swapping online information and conversation. They do not know the world without the internet and are the most educated and connected users among all generations (Chaney et al., 2017). As Generation Y, they are highly tolerant, have a positive attitude toward SM advertising, and prefer online advertising formats that offer control. For these reasons, this generation finds SM advertisements more informative than other generations (Southgate, 2017). By facing these assumptions, it is expected that the reaction toward the marketing campaign on Tinder will differ.

2.3 Users’ reaction to ads

The theory of psychological reactance refers to intrusiveness as a threat that exposes the lack of freedom and autonomy (Quick et al., 2015). A reaction occurs when an individual’s freedom
of choice happens, usually creating a motivation to regain the lost freedom (Wotrich et al., 2018). Contextualizing this concept to the SM advertising, when a user is confronted with a highly intrusive ad, a reactance occurs, leading the user to advertising evasion. Intrusiveness is considered a critical factor in explaining the avoidance of a consumer toward and advertising (Riedel et al., 2018).

Perceived intrusion measures the user’s distraction during the conduction of a task. Ads that appear without the user’s permission can be perceived as an invasion into an individual’s private subjects (Rejón-Guardia and Martínez-López, 2014). Users consider ads intrusive if they are not expecting them or, if not, find them familiar. The intrusion or nondesired ads could cause the user to perceive them as adverse. In a situation where the user sees self in a case of ad intrusion or without permission, the reaction can be annoyance and negative, leading to a possible ad evasion to complete their planned tasks. Therefore, users can develop negative feelings toward the ad, the advertised brand and the channel itself (Varnali, 2014). The perception and attitude toward the channel can be damaged by the perceived intrusion or absence of permission.

One other emotional reaction related to intrusiveness is irritation. Irritation occurs when a user is unable to close the unwanted ad, being forced to view it, requiring a tremendous cognitive effort, triggering an adverse emotional reaction toward the ad, leading to avoidance behavior (Heinonen and Strandvik, 2007).

Channel acceptance/disturbance refers to the extent to which users accept/reject a particular channel as an advertising channel. It portrays the communication context, meaning that includes when, where and how the user accesses the information (Heinonen and Strandvik, 2007).

The channel can be perceived as acceptable or disturbing, influencing user responsiveness (Boateng and Okoe, 2015). If a channel is perceived as convenient, it will intensify the acceptance of marketing communication. However, if considered as disturbing, it will influence the attention of the user toward the message, revealing a feeling of irritation and avoidance behavior, compromising the efficacy of the communication (Bakr et al., 2019; Boerman et al., 2017). This means that the acceptance of a specific channel is a prerequisite for a positive ad attitude and that when a user understands the channel as disturbing, the negative emotions will not accept the ad regardless of its relevance or usefulness.

3. Methodology

For this empirical research, the reactions toward a marketing campaign promoting a hypothetical clothing website on the dating app Tinder were collected.

The choice for collecting data from Tinder is due to its great success among the digital dating world: users swipe right and left about 1.5 billion times per day (March et al., 2017). An average Tinder user logs onto the app 9 to 11 times per day (LeFebvre, 2017). Males make up 62% of users, females 38% and 85% of the total number of users are aged between 18 and 34, with an average user age of 27 years (Smith and Anderson, 2018).

For the campaign, we adopted a familiar clothing website to give the research a digital backbone and credibility. Adopting an intrusive marketing communication, and with the adoption of an unstructured questionnaire to understand the reaction of users toward the marketing campaign, a total of 411 users’ reactions were collected at the end of the campaign execution.

For setting up the campaign, two Tinder profiles were created, a woman and a man, who were the faces of the campaign. These two individuals were fictional characters, whose Tinder profile pictures were taken from a free stock photograph website, for commercial purposes. Carlos and Maria, respectively the male and female accounts, had their location setting set to its maximum (160 km), the age gap for potential matches from 18 to 55+ years.
old, and their sexual preference as heterosexual. The type of communication used toward the dating app users was formal, along with some nuances of classic Portuguese, for it is highly uncommon and entertaining for the users in a scenery that is usually quite ordinary. For it to be possible, a script was drawn (Table 1).

The developed script was implemented for both accounts, and the dialog was kept considering the users’ answers, never allowing the conversation slide away from the friendly and more professional side (Wang et al., 2017). The creation of the script was based on the need for it to be entertaining, credible, interactive and personalized (Alalwan et al., 2017; Knoll, 2016) to capture the users’ attention since the first interaction, as throughout the entire conversation. Since the objective of this unstructured survey was to acquire the reactions of users after letting them know the conversation was part of a marketing campaign (step 6 of Table 1), it was requested their permission to be part of it. After their acceptance, the capture of responses was proceeded, along with retaining the age and gender of each user. In Table 2, it is possible to find the gender characterization of the sample.

Ages of respondents ranged from 18 to 58 years old ($M = 30.00$ years; 23.3% were from 18 to 24 years old, 64.5% from 25 to 39 years old and 12.2% from 40 to 58 years old). From a total of 411 respondents, 91.5% were male, and 8.5% were female. Such discrepancy of genders exists since Tinder users are generally males (Ward, 2017), and therefore, most reactions obtained were from the female account used, for heterosexual men have no hesitation in engaging and initiating conversations on the app.

| Step | Script |
|------|--------|
| 1st  | [If the user is smiling in any of their profile pictures] It was pointed out to me your striking and endless smile, and therefore, I succumbed to the urge to contact you. How have you been? [If the user is not smiling] Greetings, dear gentleman/madam. How have you been? or [If a user starts a conversation first without a “how are you?”] Greetings, dear gentleman/madam. How have you been? [If a user starts a conversation with a “how are you?”] Greetings, dear gentleman/madam. I am afraid to declare that my heart has lived better days. How have you been? |
| 2nd  | [When the user answers to how are they] I’m quite glad to hear so [When the user asks “how are you?”] I am afraid to declare that my heart has lived better days [Reason:] This untriumphing struggle to find love with fashion sense has been leaving me breathless and unhopeful |
| 3rd  | [When the user comments and says reassuring things] Thank you for your words [If the user asks to meet, we skip to the 4th question.] [If the user asks to contact via Instagram or WhatsApp] I’d prefer it if we talked via this platform for now. [Then proceeding to the 4th question.] |
| 4th  | May I ask what brings you to this quite modern application? |
| 5th  | [The user answers the question and then we tell them our reason:] The pursuit of such human being of exquisite taste and special fashion sense |
| 6th  | [The user will say something and regardless, we reply:] What would you tell me if right now you were being part of a marketing campaign? I’m currently studying Tinder’s users’ receptivity toward a marketing campaign |

| Generation | Age | Male | Female | Total |
|------------|-----|------|--------|-------|
| Z          | 18–24 | 82   | 14     | 96    |
| Y          | 25–39 | 245  | 20     | 265   |
| X          | +40   | 49   | 1      | 50    |
| Total      |       | 376  | 35     | 411   |

Table 2.
Sample characterization
The collected dataset was analyzed using the R statistical software, with the developed script for the text mining being implemented through the RStudio interactive environment. The R tool is a free and open software for data analysis benefitting from a large online community, including packages such as the “sentiments”, which computes a sentence’s sentiment score (Cortez, 2014).

Text mining is a multidisciplinary field to extract information from a significant portion of nonstructured textual data, eliminating irrelevant text to find pertinent details and to uncover patterns of relevant knowledge (Brochado et al., 2019; Moro et al., 2015). Text mining tools are well suited to automate, refine and transform business intelligence activities that are traditionally conducted employing intensive work of manual literature revision in the search for patterns among the data. Text mining has been used in competitive intelligence, customer management, research, among others.

Sentiment analysis enables us to understand how the users express themselves in text, revealing a positive or negative reaction (Calheiros et al., 2017; Guerreiro and Rita, 2019). A large amount of studies has used sentiment analysis. For instance, Calheiros et al. (2017) applied sentiment analysis to characterize a given hospitality issue. Lee et al. (2017) used text mining techniques specifically to sentiment classification analysis to understand the relationship between the entropy of review text sentiment and the online word of mouth effects. Pathak and Pathak-Shelat (2017) used sentiment analysis to explain the negative sentiments expressed by virtual tribes. Therefore, the use of sentiment analysis to conduct unstructured text data has been used in different contexts revealing fascinating results. The reactions obtained from the users function perfectly as feedback/reviews, therefore the adequate way to analyze the collected data.

The sentiment analysis developed script was run to score the 411 user responses via a scale of sentiments: sentiments can be negative, neutral, or positive. In the present research, values above zero count as positive, values below zero are negative, and the values equal to zero are neutral.

To understand if the reaction toward the marketing campaign on each generation is different, we have used generational cohorts as Generation X (born between 1965 and 1980), Generation Y (born between 1981 and 1995), and Generation Z (born after 1995), following the approach of Kitchen and Proctor (2015). A normality Shapiro–Wilk test was conducted first to test the assumptions for a One-way ANOVA analysis. As those assumptions failed, we then did a Kruskal–Wallis H test considering a significance level at \( p < 0.05 \). All inferential statistical calculations were performed using SPSS (26.0).

4. Results and discussion

First, it was proceeded to examine the frequency of words from all user responses to be able to get more insights and scrutinize the vast information that was obtained in text format. As it can be found in Table 3, the ten most frequently used objectives are displayed. From the ones that are considered negative, there were only two found – “bother”, with a frequency of 0.79% and “bad” with 0.28%. Also, we can see “good”, “well”, “interesting”, “luck”, “interest”, “funny”, “great” and “nice” as positive words, the most relevant ones with the frequencies of 2.31% (“good”) and 1.15% (“well”). Positive influence is a crucial driver to SM users/consumer’s reactions by commenting their opinion on the subject, in opposition to negative feelings, giving the audience a lack of motivation to comment and merely leading them to ignore such experience (Berger, 2014; Boateng and Okoe, 2015). Remarkably, these results acknowledge a generally positive reaction toward the possibility of receiving marketing information through Tinder.

For understanding the sentiment of users behind their responses, data were analyzed through text mining and the sentiment scale. For the analyzed dataset of reactions, the most
negative one scored –0.76 of sentiment, while the most positive scored 1.34. For an exhaustive analysis, a sentiment scale was created using SentiWordNet (Ahmed and Danti, 2016) (Table 4).

A descriptive analysis of the overall results is presented in Table 5.

From the 411 user’s perspective, 34.31% declare a fragile positive sentiment toward the perspective of receiving marketing communications through Tinder. On the negative side, 25.30% report a fragile negative sentiment. These results indicate that Tinder users have a generally positive attitude toward advertising through this platform and less irritated when approached. A positive feeling affects the users’ response and attitude toward advertising positively and legitimates Tinder as a communication channel to conduct marketing activities (Rejón-Guardia and Martínez-López, 2014).

| Word     | Frequency | Percentage |
|----------|-----------|------------|
| good     | 100       | 2.31       |
| well     | 50        | 1.15       |
| interesting | 39       | 0.9        |
| bother   | 34        | 0.79       |
| luck     | 31        | 0.72       |
| interest | 22        | 0.51       |
| funny    | 17        | 0.39       |
| great    | 16        | 0.37       |
| bad      | 12        | 0.28       |
| nice     | 11        | 0.25       |
| Total    | 4,330     | 100.00     |

| Sentiment          | Scale          |
|--------------------|----------------|
| Solid positive     | ≥0.60          |
| Regular positive   | [0.30, 0.59]   |
| Fragile positive   | [0.01, 0.29]   |
| Neutral            | 0              |
| Fragile negative   | [–0.29, –0.01] |
| Regular negative   | [–0.59, –0.30] |
| Solid negative     | ≤–0.60         |

| Absolute values | Relative values | Mean | Std. deviation |
|-----------------|-----------------|------|----------------|
| Solid Positive  | 34              | 8.3% | 0.77           | 0.16 |
| Regular Positive| 89              | 21.7%| 0.43           | 0.08 |
| Fragile Positive| 144             | 35.0%| 0.16           | 0.08 |
| Neutral         | 23              | 5.6% | 0             | 0    |
| Fragile Negative| 104             | 25.3%| –0.12          | 0.08 |
| Regular Negative| 13              | 3.2% | –0.36          | 0.06 |
| Solid Negative  | 4               | 1.0% | –0.69          | 0.05 |
| Total Positive  | 267             | 65.0%| 0.33           | 0.23 |
| Total Negative  | 121             | 29.4%| –0.17          | 0.14 |
| Total           | 411             | 100% | 0.16           | 0.30 |

Table 3. Top 10 more frequent words
Table 4. Sentiment scale
Table 5. Descriptive analysis (Absolute, relative, mean and standard deviation)
It is significant to reference the variance between both positive and negative means: the mean of positive value corresponds to a regular positive in the sentiment scale, while the mean of negative values fits on the fragile negative scale. The absolute mean shows a general fragile positive reaction toward the possibility of users to receive marketing information through Tinder, confirming the results obtained in Table 3. The relevance of these results is the recognition of Tinder as a potential marketing channel, ready to be explored. Companies need to understand the potential of a channel before invest in it with marketing campaigns (Verbraken et al., 2014).

Neutral sentiments toward the campaign (values equal to zero on the scale) were 5.60%, and they can be considered quite optimistic, as only very few people felt “nothing” toward the marketing campaign. The percentage of positive reactions and negative reactions were of 64.96% and 29.44%, respectively. More than half of Tinder users’ emotions toward the campaign were positive.

Results were transformed into a scatter plot for a visual interpretation and analysis of the dispersion of the results.

On the scatter plot from Figure 1, it can be noticed that it is considerably more condensed above zero value, the positive scale of sentiments’ area, as expected, since 65.9% of the values obtained were above zero.

Most of the negative emotions’ values stand between 0.01 and −0.29 in the scale of sentiments, which is meaningfully low. Many Tinder users are receptive while coming across a marketing campaign, widening the range of opportunities that can be taken when creating advertising.

For a detailed examination of the results, data were segmented by generation (Table 6).

The only generation that revealed a solid negative result was Generation Y (1.5%), showing high resistance to receive marketing communications through Tinder. This information is validated by a user, 27 years old, stated, “I would say it would be annoying” after being asked for his opinion. Generation Z (4.2%), followed by Generation X (4.0%), revealed the highest results within the regular negative effects. In the fragile negative sentiment scale, Generation Y declared the highest results (28.3%). A user, 19 years old, expressed his opinion by saying: “Seriously? Very strange”, while a user, 29 years old revealed that: “I would say that everything is enough to get the desired message to the consumer, it is unheard of on this platform, but it does not surprise me”.

Users with the best results within the fragile positive sentiment scale were those from Generation X (46.0%) and Generation Z (40.6%), while on the regular positive results, those who showed more receptivity was Generation Y (25.7%). A user, 47 years old stated, “it does
|          | Solid negative | Regular negative | Fragile negative | Neutral | Fragile positive | Regular positive | Solid positive | Total negative | Total positive |
|----------|----------------|------------------|------------------|---------|------------------|------------------|----------------|----------------|----------------|
| Gen Z ($n = 82$) | 0    | 4.2              | 18.8             | 6.3     | 40.6             | 17.7             | 12.5           | 22.9           | 70.8           |
| Gen Y ($n = 265$) | 1.5  | 2.6              | 28.3             | 5.3     | 29.8             | 25.7             | 6.8            | 32.5           | 62.3           |
| Gen X ($n = 50$)  | 0    | 4.0              | 22.0             | 6.0     | 46.0             | 14.0             | 8.0            | 26.0           | 68.0           |
not bother me at all, but I do not think it is a big hit either”, showing a fragile positive sentiment. A user, 20 years old, revealed his regular positive reaction by saying, "I already know the store! It is funny”.

Among all generations, Generation Z revealed to be one with the highest acceptability to receive marketing information through Tinder (12.5%).

By observing general results, the age gap with the most favorable results were those from Generation Z (70.8%), followed by users from Generation X (68.0%). A user, 24 years old, revealed his opinion by saying, “I would give sincere congratulations for the creativity”, confirming this result.

On the other hand, Generation Y, revealed more resistance to receive marketing communications through Tinder (32.5%), confirming previous studies that highlight that this generation is considered to be skeptical and distrustful toward marketing campaigns (Lissitsa and Kol, 2016; Valentine and Powers, 2013). Negative results will reveal negative emotions, such as irritation and annoyance (Varnali, 2014).

To understand if the reaction toward the marketing campaign on each generation is different, parametric and nonparametric analyses were undertaken, considering the descriptive statistics of sentiment scores grouped by generation (Table 7).

The result obtained by the parametric test (one-way ANOVA) was unfeasible due to the nonnormal distribution of outcomes ($W_Z(96) = 0.97, p = 0.020$; $W_Y(265) = 0.99, p = 0.017$; $W_X(50) = 0.97, p = 0.184$).

Sentiment scores of Generation Z (Median = 0.17) were higher than those of Generation Y (Median = 0.12), and Generation X (Median = 0.12). A Kruskal–Wallis test showed that the differences were not statistically significant ($H(2) = 1.099, p = 0.577$).

Kruskal–Wallis test showed that generation does not significantly affect the sentiment of Tinder users. Nevertheless, the descriptive statistics revealed that the generation with more willingness to receive marketing campaigns on Tinder could be Generation Z. Every age group has a positive mean, although the results are considered fragile positive. Generation Z (0.20) is the one showing the most positive reaction to receiving marketing communications through Tinder, confirming that this generation has a great willingness and positive attitude toward SM advertising (Southgate, 2017). On the other hand, Generations Y and Z showed less propensity (0.15) to receive marketing communications through Tinder, although the results are considered positive (fragile).

5. Conclusions, limitations and future research
This study aimed to analyze online dating apps users’ responsiveness toward a marketing campaign promoting a clothing website, making use of a text mining analysis using a sentiment scale, and a Kruskal–Wallis test to understand the statistical differences between each generation. In general, results showed that online dating apps users have positive feelings toward the marketing campaign, revealing to be an acceptable channel to conduct intrusive marketing activities. First, not only the percentage of positive reactions was of

| Generation | Frequency | Sentiment Mean (SD) | Mdn (IQR) | Mean rank | df | Statistics  | Kruskal–Wallis H | p    |
|------------|-----------|---------------------|-----------|-----------|----|-------------|------------------|------|
| Z          | 96        | 0.20 (0.31)         | 0.17 (0.0;0.39) | 217.11    | 2  | 1.099       | 0.577            |
| Y          | 265       | 0.15 (0.31)         | 0.12 (-0.07;0.36) | 202.46    |    |             |                  |
| X          | 50        | 0.15 (0.24)         | 0.12 (-0.01;0.29) | 203.42    |    |             |                  |

| Note(s): SD = Standard Deviation; Mdn = Median; IQR = Interquartile Range; df = Degrees of freedom; p = p value |

Table 7. Descriptive statistics of sentiment score grouped by generation
65.94% and the negative ones of 30.17%, but the difference of the range of values of both poles was significant: the mean of the positive reactions was 0.32, much higher than the negative reactions mean of \(-0.16\), which lead us to conclude the general positive feeling of users to accept a marketing campaign through an online dating app. The conduction of an intrusive marketing campaign through an online dating app will have a positive influence on user’s responsiveness and intensify the acceptance of marketing communication (Bakr et al., 2019).

Generation Z were the ones who revealed more receptivity toward the campaign, highlighting that this is the generation that shows a propensity to engage with an online dating app campaign. Companies with an audience according to these criteria should bet in an online dating app to conduct marketing campaigns. Nevertheless, the statistical analysis revealed that the differences between each generation are not statistically significant.

For academia, this research contributes to the literature by revealing the acceptance of an online dating app as a marketing channel and particularly those who are more receptive to a marketing campaign on this type of SM platform. From a managerial standpoint, companies can benefit from the exposure obtained by the number of active users present on Tinder and its possible segmentations. Firms can engage with Tinder users, as they are open to communicate with everyone and seek. Additionally, this research reinforced the need to be interactive with users so that their acceptance turns positive, and develop and create a relationship to become a long-term continuous relationship (Gummesson, 2017), especially with consumers’ age gap that makes the most use of SM.

The present research contains various limitations that can be imperative for future research, as well as for a deeper understanding of this study. A limitation is the fact that the participants, during the data collection, revealed their identification, perhaps leading to acquiescence bias. This usually happens when the respondents tend to agree to agree-disagree questions (Kam and Zhou, 2015). For future research, it would be relevant to collect data from users that do not reveal their name or face, to prevent bias.

The difference between the number of answers of both genders must do with the different behaviors they pursue while navigating on dating applications. For example, male users have minimal criteria when it comes to “swiping right”, while female users are very particular and demanding while doing so (Ward, 2017). Nevertheless, for future research, a more balanced sample in terms of gender would be positive to examine any possible differences of emotions toward the campaign. A balanced generation sample would be relevant since the Generation Z and Generation X were smaller compared to Generation Y. Another limitation is the generalization of the obtained results since the marketing campaign was applied only for the clothing business.

Additionally, the data analysis process, while using the text mining analysis and respective sentiment classification to analyze the reactions has the limitations of not being able to detect sarcasm, which was sometimes present in a few users’ responses. However, these possible limitations do not inflict the results of the present study regarding the potential found on Tinder for future successful marketing campaigns.

Furthermore, it is critical to have in consideration the type of product/service that is going to be advertised on Tinder, to verify if it somehow connects with the concept of the app, trying to avoid incongruency. It would be interesting to apply a similar marketing campaign to not only fashion related but also to other areas of business such as cosmetics. Hence, the acceptance toward the marketing campaign on the dating app might be affected by different cultural perspectives, revealing the pertinence to be validated in a future study.
References
Abril, C. and Rodríguez-Cánovas, B. (2016), “Marketing mix effects on private labels brand equity”, European Journal of Management and Business Economics, Vol. 25 No. 3, pp. 168-175, doi: 10.1016/j.redeen.2016.09.003.

Ahmed, S. and Danti, A. (2016), “Effective sentimental analysis and opinion mining of web reviews using rule based classifiers”, in Behera, H.S. and Mohapatra, D.P. (Eds), Computational Intelligence in Data Mining—Volume 1, Springer India, New Delhi, pp. 171-179.

Alalwan, A.A., Rana, N.P., Dwivedi, Y.K. and Algharabat, R. (2017), “Social media in marketing: a review and analysis of the existing literature”, Telematics and Informatics, Vol. 34 No. 7, pp. 1177-1190, doi: 10.1016/j.tele.2017.05.008.

Ashley, C. and Tuten, T. (2015), “Creative strategies in social media marketing: an exploratory study of branded social content and consumer engagement”, Psychology and Marketing, Vol. 32 No. 1, pp. 15-27, doi: 10.1002/mar.20761.

Bakr, Y., Tolba, A. and Meshreki, H. (2019), “Drivers of SMS advertising acceptance: a mixed-methods approach”, The Journal of Research in Indian Medicine, Vol. 13 No. 1, pp. 96-118, doi: 10.1108/JRIM-02-2018-0033.

Bento, M., Martinez, L.M. and Martinez, L.F. (2018), “Brand engagement and search for brands on social media: comparing Generations X and Y in Portugal”, Journal of Retailing and Consumer Services, Vol. 43 January 2017, pp. 234-241, doi: 10.1016/j.jretconser.2018.04.019.

Berger, J. (2014), “Word of mouth and interpersonal communication: a review and directions for future research”, Journal of Consumer Psychology, Vol. 24 No. 4, pp. 586-607, doi: 10.1016/j.jcps.2015.05.002.

Bernabé-Moreno, J., Tejeda-Lorente, A., Porcel, C., Fujita, H. and Herrera-Viedma, E. (2015), “CARESOME: a system to enrich marketing customers acquisition and retention campaigns using social media information”, Knowledge-Based Systems, Vol. 80, pp. 163-179, doi: 10.1016/J.KNOSYS.2014.12.033.

Boateng, H. and Okoe, A.F. (2015), “Consumers’ attitude towards social media advertising and their behavioural response”, The Journal of Research in Indian Medicine, Vol. 9 No. 4, pp. 299-312, doi: 10.1108/JRIM-01-2015-0012.

Boerman, S., Kruikemeier, S. and Borgesius, F.J.Z. (2017), “Online behavioral Advertising: a literature review and research agenda”, Journal of Advertising, Vol. 46 No. 3, pp. 363-376, doi: 10.1080/00913367.2017.1339347.

Boulianne, S. (2015), “Social media use and participation: a meta-analysis of current research”, Information, Communication and Society, Vol. 18 No. 5, pp. 524-538, doi: 10.1080/1369118X.2015.1008542.

Brochado, A., Rita, P., Oliveira, C. and Oliveira, F. (2019), “Airline passengers’ perceptions of service quality: themes in online reviews”, International Journal of Contemporary Hospitality Management, Vol. 31 No. 2, pp. 855-873, doi: 10.1108/IJCHM-09-2017-0672.

Calheiros, A.C., Moro, S. and Rita, P. (2017), “Sentiment classification of consumer-generated online reviews using topic modeling”, Journal of Hospitality Marketing and Management, Vol. 26 No. 7, pp. 675-693.

Chaney, D., Touzani, M. and Ben Slimane, K. (2017), “Marketing to the (new) generations: summary and perspectives”, Journal of Strategic Marketing, Vol. 25 No. 3, pp. 179-189, doi: 10.1080/0965254X.2017.1291173.

Clark, M., Black, H.G. and Judson, K. (2017), “Brand community integration and satisfaction with social media sites: a comparative study”, The Journal of Research in Indian Medicine, Vol. 11 No. 1, pp. 39-55, doi: 10.1108/JRIM-07-2015-0047.

Cortez, P. (2014), Modern Optimization with R, Springer.

Guerreiro, J. and Rita, P. (2019), “How to predict explicit recommendations in online reviews using text mining and sentiment analysis”, Journal of Hospitality and Tourism Management No. July, pp. 1-4, doi: 10.1016/j.jhtm.2019.07.001.
Gummesson, E. (2017), “From relationship marketing to total relationship marketing and beyond”, Journal of Services Marketing, Vol. 31 No. 1, pp. 16-19, doi: 10.1108/JSM-11-2016-0398.

Hartemo, M. (2016), “Email marketing in the era of the empowered consumer”, The Journal of Research in Indian Medicine, Vol. 10 No. 3, pp. 212-230, doi: 10.1108/JRIM-06-2015-0040.

Hartmann, N.N., Wieland, H. and Vargo, S.L. (2018), “Converging on a new theoretical foundation for selling”, Journal of Marketing, Vol. 82 No. 2, pp. 1-18, doi: 10.1509/jm.16.0268.

Heinonen, K. and Strandvik, T. (2007), “Consumer responsiveness to mobile marketing”, International Journal of Mobile Communications, Vol. 5 No. 6, p. 603, doi: 10.1504/IJMC.2007.014177.

Hudson, S., Roth, M.S., Madden, T.J. and Hudson, R. (2015), “The effects of social media on emotions, brand relationship quality, and word of mouth: an empirical study of music festival attendees”, Tourism Management, Vol. 47, pp. 68-76.

Jaakonmaki, R., Muller, O. and Vom Brocke, J. (2017), “The impact of content, context, and creator on user engagement in social media marketing”, Hawaii International Conference on System Sciences, HICSS-50, pp. 1152-1160, (November 2016), doi:10.24251/HICSS.2017.136.

Kam, C.C.S. and Zhou, M. (2015), “Does acquiescence affect individual items consistently?”, Educational and Psychological Measurement, Vol. 75 No. 5, pp. 764-784, doi: 10.1177/0013164414560817.

Killian, G. and McManus, K. (2015), “A marketing communications approach for the digital era: managerial guidelines for social media integration”, Business Horizons, Vol. 58 No. 5, pp. 539-549, doi: 10.1016/j.bushor.2015.05.006.

Kitchen, P.J. and Proctor, T. (2015), “Marketing communications in a post-modern world”, Journal of Business Strategy, Vol. 36 No. 5, pp. 34-42, doi: 10.1108/JBS-06-2014-0070.

Knoll, J. (2016), “Advertising in social media: a review of empirical evidence”, International Journal of Advertising, Vol. 35 No. 2, pp. 266-300, doi: 10.1080/02650487.2015.1021898.

Knoll, J. and Matthes, J. (2017), “The effectiveness of celebrity endorsements: a meta-analysis”, Journal of the Academy of Marketing Science, Vol. 45 No. 1, pp. 55-75, doi: 10.1007/s11747-016-0503-8.

Lee, J.H., Jung, S.H. and 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, Vol. 22, pp. 42-52, doi: 10.1016/j.elerap.2017.03.001.

LeFebvre, L.E. (2017), “Swiping me off my feet”, Journal of Social and Personal Relationships, Vol. 35 No. 9, 026540751770641, doi: 10.1177/0265407517706419.

Lin, C.A. and Kim, T. (2016), “Predicting user response to sponsored advertising on social media via the technology acceptance model”, Computers in Human Behavior, Vol. 64, pp. 710-718, doi: 10.1016/j.chb.2016.07.027.

Lissitsa, S. and Kol, O. (2016), “Generation X vs. Generation Y - a decade of online shopping”, Journal of Retailing and Consumer Services, Vol. 31, pp. 304-312, doi: 10.1016/j.jretconser.2016.04.015.

Litterio, A.M., Nantes, E.A., Larrosa, J.M. and Gómez, L.J. (2017), “Marketing and social networks: a criterion for detecting opinion leaders”, European Journal of Management and Business Economics, Vol. 26 No. 3, pp. 347-366, doi: 10.1108/EJMBE-10-2017-020.

March, E., Grieve, R., Marrington, J. and Jonason, P.K. (2017), “Trolling on Tinder® (and other dating apps): examining the role of the Dark Tetrad and impulsivity”, Personality and Individual Differences, Vol. 110, pp. 139-143, doi: 10.1016/j.paid.2017.01.025.

McCoy, S., Everard, A., Galletta, D.F. and Moody, G.D. (2017), “Here we go again! the impact of website ad repetition on recall, intrusiveness, attitudes, and site revisit intentions”, Information and Management, Vol. 54 No. 1, pp. 14-24, doi: 10.1016/j.im.2016.03.005.
Moro, S., Cortez, P. and Rita, P. (2015), “Business intelligence in banking: a literature analysis from 2002 to 2013 using text mining and latent Dirichlet allocation”, Expert Systems with Applications, Vol. 42 No. 3, pp. 1314-1324.

Pathak, X. and Pathak-Shelat, M. (2017), “Sentiment analysis of virtual brand communities for effective tribal marketing”, The Journal of Research in Indian Medicine, Vol. 11 No. 1, pp. 16-38, doi: 10.1111/jrim.2015.0069.

Quick, B.L., Kam, J.A., Morgan, S.E., Montero Liberona, C.A. and Smith, R.A. (2015), “Prospect theory, discrete emotions, and freedom threats: an extension of psychological reactance theory”, Journal of Communication, Vol. 65 No. 1, pp. 40-61, doi: 10.1111/jcom.12134.

Ramos, R.F., Rita, P. and Moro, S. (2019), “From institutional websites to social media and mobile applications: a usability perspective”, European Research on Management and Business Economics, Vol. 25 No. 3, pp. 138-143, doi:10.1016/j.iedeen.2019.07.001.

Ramos, R.F., Rita, P. and Moro, S. (2020), “Is this the beginning of the end for retail websites? A professional perspective”, International Journal of Internet Marketing and Advertising.

Rejón-Guardia, F. and Martínez-López, F.J. (2014), “Online advertising intrusiveness and consumers’ avoidance behaviors”, in Martínez-López, F.J. (Ed.), Berlin, Heidelberg, Springer Berlin Heidelberg, pp. 556-586, doi: 10.1007/978-3-642-39747-9_23.

Riedel, A.S., Weeks, C.S. and Beatson, A.T. (2018), “Am I intruding? Developing a conceptualisation of advertising intrusiveness”, Journal of Marketing Management, Vol. 34 Nos 9-10, pp. 750-774, doi: 10.1080/0267257X.2018.1496130.

Sashittal, H.C., Sriramachandramurthy, R. and Hodis, M. (2012), “Targeting college students on Facebook? How to stop wasting your money”, Business Horizons, Vol. 55 No. 5, pp. 495-507, doi: 10.1016/j.bushor.2012.05.006.

Saxena, A. and Khanna, U. (2013), “Advertising on social network sites: a structural equation modelling approach”, Vision: The Journal of Business Perspective, Vol. 17 No. 1, pp. 17-25, doi:10.1177/0972262912469560.

Schivinski, B. and Dabrowski, D. (2016), “The effect of social media communication on consumer perceptions of brands”, Journal of Marketing Communications, Vol. 22 No. 2, pp. 189-214, doi: 10.1080/13527266.2013.871323.

Slootweg, E. and Rowson, B. (2018), “My generation: a review of marketing strategies on different age groups”, Research in Hospitality Management, Vol. 8 No. 2, pp. 85-92, doi: 10.1080/22243534.2018.1553369.

Smith (2011), “Digital marketing strategies that Millennials find appealing, motivating, or just annoying”, Journal of Strategic Marketing, Vol. 19 No. 6, pp. 489-499, doi: 10.1080/09696484.2011.581383.

Smith, A. and Anderson, M. (2018), Social Media Use 2018: Demographics and Statistics, Pew Research Center.

Soares, A.M. and Pinho, J.C. (2014), “Advertising in online social networks: the role of perceived enjoyment and social influence”, The Journal of Research in Indian Medicine, Vol. 8 No. 3, doi:10.1108/JRIM-01-2013-0004.

Southgate, D. (2017), “The emergence of generation Z and its impact in advertising”, Journal of Advertising Research, Vol. 57 No. 2, pp. 227-235, doi: 10.2501/JAR-2017-028.

Statista (2017), Global Time Spent on Social Media Daily, available at: https://www.statista.com/statistics/433871/daily-social-media-usage-worldwide/, (accessed 26 July 2018).

Statista (2018), Online Dating Worldwide, available at: https://www.statista.com/outlook/372/100/online-dating/worldwide/#market-age, (accessed 26 July 2018).

Stojanovic, I., Andreu, L. and Curras-Perez, R. (2018), “Effects of the intensity of use of social media on brand equity”, European Journal of Management and Business Economics, Vol. 27 No. 1, pp. 83-100, doi: 10.1108/EJMBE-11-2017-0049.
Sumter, S.R., Vandenbosch, L. and Ligtenberg, L. (2017), “Love me Tinder: untangling emerging adults’ motivations for using the dating application Tinder”, Telematics and Informatics, Vol. 34 No. 1, pp. 67-78, doi: 10.1016/J.TELE.2016.04.009.

Valentine, D.B. and Powers, T.L. (2013), “Generation Y values and lifestyle segments”, Journal of Consumer Marketing, Vol. 30 No. 7, pp. 597-606, doi: 10.1108/JCM-07-2013-0650.

Varnali, K. (2014), “SMS advertising: how message relevance is linked to the attitude toward the brand?”, Journal of Marketing Communications, Vol. 20 No. 5, pp. 339-351, doi: 10.1080/13527266.2012.699457.

Verbraken, T., Goethals, F., Verbeke, W. and Baesens, B. (2014), “Predicting online channel acceptance with social network data”, Decision Support Systems, Vol. 63, pp. 104-114, doi: 10.1016/j.dss.2013.08.011.

Voorveld, H.A.M., van Noort, G., Muntinga, D.G. and Bronner, F. (2018), “Engagement with social media and social media advertising: the differentiating role of platform type”, Journal of Advertising, Vol. 47 No. 1, pp. 38-54, doi: 10.1080/00913367.2017.1405754.

Vrontis, D., Thrassou, A. and Amirkhanpour, M. (2017), “B2C smart retailing: a consumer-focused value-based analysis of interactions and synergies”, Technological Forecasting and Social Change, Vol. 124, pp. 271-282, doi: 10.1016/j.techfore.2016.10.064.

Wang, S., Kirillova, K. and Lehto, X. (2017), “Reconciling unsatisfying tourism experiences: message type effectiveness and the role of counterfactual thinking”, Tourism Management, Vol. 60, pp. 233-243, doi: 10.1016/j.tourman.2016.12.008.

Ward, J. (2017), “What are you doing on Tinder? Impression management on a matchmaking mobile app”, Information, Communication and Society, Vol. 20 No. 11, pp. 1644-1659, doi: 10.1080/1369118X.2016.1252412.

Wottrich, V.M., van Reijmersdal, E.A. and Smit, E.G. (2018), “The privacy trade-off for mobile app downloads: the roles of app value, intrusiveness, and privacy concerns”, Decision Support Systems, Vol. 106, pp. 44-52, doi: 10.1016/j.dss.2017.12.003.

Yang, H.C. and Liu, H. (2014), “Prior negative experience of online disclosure, privacy concerns, and regulatory support in Chinese social media”, Chinese Journal of Communication, Vol. 7 No. 1, pp. 40-59, doi: 10.1080/17544750.2013.816756.

Zhang, J. and Mao, E. (2016), “From online motivations to ad clicks and to behavioral intentions: an empirical study of consumer response to social media advertising”, Psychology and Marketing, Vol. 33 No. 3, pp. 155-164, doi: 10.1002/mar.20862.

Further reading

Duguay, S. (2017), “Dressing up Tinderella: interrogating authenticity claims on the mobile dating app Tinder”, Information, Communication and Society, Vol. 20 No. 3, pp. 351-367, doi: 10.1080/1369118X.2016.1168471.

Guerreiro, J. and Moro, S. (2017), “Are Yelp’s tips helpful in building influential consumers?”, Tourism Management Perspectives, Vol. 24, pp. 151-154, doi: 10.1016/j.tmp.2017.08.006.

Morente-Molinera, J.A., Kou, G., Peng, Y., Torres-Albero, C. and Herrera-Viedma, E. (2018), “Analysing discussions in social networks using group decision making methods and sentiment analysis”, Information Sciences, Vol. 447, pp. 157-168, doi: 10.1016/j.ins.2018.03.020.

Moro, S., Pires, G., Rita, P. and Cortez, P. (2018), “A cross-cultural case study of consumers’ communications about a new technological product”, Journal of Business Research, doi: 10.1016/j.jbusres.2018.08.009.

Moro, S., Ramos, P., Esmerado, J. and Jalali, S.M.J. (2019), “Can we trace back hotel online reviews’ characteristics using gamification features?”, International Journal of Information Management, Vol. 44, pp. 88-95, doi: 10.1016/j.ijinfomgt.2018.09.015.
Corresponding author
Ricardo Filipe Ramos can be contacted at: ricardo Ramos@iscte.pt

Regan, P. (2017), The Mating Game: A Primer on Love, Sex, and Marriage, 3rd ed., SAGE Publications Sage UK, London.