Social media for cultural communication: A critical investigation of museums’ Instagram practices

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Social media for cultural communication: A critical investigation of museums’ Instagram practices

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Abstract

**Purpose**: The purpose of the study is to investigate the use of Instagram by museums in the Greek cultural scene. Specifically, the study focuses on examining the use of Instagram by museum communication professionals and aims at carrying out a twofold investigation: Firstly, if and how the Instagram is used to reach out to their visitors and secondly, the public response to this type of communication.

**Methods**: A list of all archaeological museums in Greece was obtained and related Instagram accounts were retrieved. The dataset structure was enhanced by eleven variables, which were measured and visualized by a descriptive statistics analysis. Inter-variable correlations, normality and equality tests were also performed. Moreover, a linear predictive model for the number of museum tags was investigated.

**Results**: Only one museum in Greece maintains an Instagram account. Visitors usually tag museum exhibits or people and exhibits on the photographs they upload on their personal accounts. T-tests and Mann-Whitney U tests revealed equal distributions for all variables between central and peripheral museums.

**Implications**: Museum officials have not seized the opportunity offered by social media and especially Instagram today. Their importance seems to be underestimated. With respect to the linear model derived, results suggest that more features should be surveyed; this could be the subject of future research studies.

**Keywords**: social media, communication, digital marketing, Instagram, multiple regression

**JEL Classification**: N7, L82, D83

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1 INTRODUCTION

Social media platforms seem to be popular with consumers and businesses as they provide new opportunities for interactivity and connectivity for both. Social media are described as “content that has been created by its audience” (Comm, 2009) and they are characterized by user-generated content (DeNardis, 2014). Social media are defined as “a group of internet-based applications that build on the ideological and technical foundations of Web 2.0 that allow the creation and exchange of user generated content” (Kaplan and Haenlein, 2010:61). They are used for the development of social connections, using highly available and scalable publishing methods (Sajid, 2016) changing the ways society consumes and contributes to the creation of information (Hays, Pages and Buhalis, 2013).

Interactivity, connectivity and sharing are the most important features offered by social media. “Interactivity is defined in terms of the immediacy of the responsiveness and the degree to which the communication resembles human discourse” (Liu and Shrum, 2002:54). Social media interactivity can take place via desktop computer, laptop, mobile phone or a personal digital assistant of the iPod (Grover and Stewart, 2010).

Connectivity is also one of the characteristics of social media. “Social media have increased individuals’ connectivity and enabled users’ direct participation” (Chou, Hunt, Beckjord, Moser and Hessee, 2009:2). Organizations can maintain “an ongoing dialogue with customers by exploiting the
connectivity, content creation and sharing functionalities of social media” (Cabiddu, Carlo and Piccoli, 2014:187). This application gives the opportunity to users to be online and to be always connected. That means that they can share each other’s content.

Sharing is another application that helps companies and organisations to gain publicity (Kietzman, Hermkens, McCarthy and Silvestre, 2011). It is common for users to share posts on their wall. As Kirtis and Karahan (2011) claim, the social media user can create and comment on blogs, share contents or communicate with friends via social networking sites like Facebook or the earlier MySpace. On the other hand, other media like YouTube give users the opportunity to make and share different videos with different sets of friends (Lange, 2007).

More than two billion users all over the world use social media. In Table 1 (redrawn from https://www.statista.com/statistics/272014/ there is a list showing the number of registered users for each platform.

<table>
<thead>
<tr>
<th>Platform</th>
<th>Registered users (in millions, May 2020)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Facebook</td>
<td>2,498</td>
</tr>
<tr>
<td>WeChat</td>
<td>1,165</td>
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<tr>
<td>QZone</td>
<td>517</td>
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<td>Snapchat</td>
<td>398</td>
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<td>YouTube</td>
<td>2,000</td>
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<tr>
<td>Instagram</td>
<td>1,000</td>
</tr>
<tr>
<td>Sina Weibo</td>
<td>516</td>
</tr>
<tr>
<td>Twitter</td>
<td>386</td>
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<tr>
<td>WhatsApp</td>
<td>2,000</td>
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<tr>
<td>TikTok</td>
<td>800</td>
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<tr>
<td>Reddit</td>
<td>430</td>
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<td>Pinterest</td>
<td>366</td>
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<tr>
<td>Kuaishou</td>
<td>400</td>
</tr>
<tr>
<td>QQ</td>
<td>731</td>
</tr>
<tr>
<td>FB Messenger</td>
<td>1,300</td>
</tr>
</tbody>
</table>

Social media has become a key factor in the way that cultural organisations communicate with their public, as they have the opportunity to support the marketing of performing arts organizations. It continues to gain prominence in communication campaigns due to the high levels of public usage and public involvement with organizations on social media sites (Waters and Jones, 2011). Many museums have started using different forms of social media, such as social networking, podcasting, and blogging (Russo and Peacock, 2009), as they allow museums to move their activities into the digital space while also reinforcing their reputation in terms of cultural authority and authenticity (Dearolph, 2014).

As Russo, Watkins and Smith (2009:153) argue, “social media offer young people agency previously unavailable in informal learning environments in order to explore complex responses to and participation with cultural content”. Museums mostly use social media to promote event listings, reminders and to reach larger or newer audiences (Fletcher and Lee, 2012) and for engaging their audiences (Zafiropoulos et al 2015). With social media, museums are facilitated in distributing information to their visitors and creating opportunities for them to learn about and interact with each other (Chung, Marckett and Fiore 2014). Stuedahl and Smordal (2011) believe that social media have been adopted by museums to help visitors interact with museum exhibits themselves. For museums, this could mean that they have to cater for the needs of an informed, potentially demanding and more difficult to satisfy public.

The goal of this research is to describe how museums utilize social media and especially Instagram as a platform, so as to identify the most effective tools for public engagement. As there is limited research about the use of Instagram in the Greek cultural environment, this research attempts to fill the gap regarding the use of social media as a communication and marketing tool by cultural organizations such as the Greek archaeological museums.

2 LITERATURE REVIEW

Photos are a means of communication and as Miranda (2013) claims, most of the museums have removed the barriers to taking photographs in their buildings. Most cultural organizations take photographs of collections, as professionals believe it is important to engage visitors with photographs (Edwards and Sigrid Lien, 2014). Photographs posted by museum visitors with positive attitudes to the museum help raise the museum’s popularity and accessibility to its future visitors (Stylianou-Lambert, 2017). Brooklyn Museum, realising the importance of enhancing visitor experience through photos, uses photos, blogs and videos in order to be connected with its online visitors (Black, 2012). In their study about the use of Flickr in cultural institutions, Beaudoin and Bosshard (2012) showed that most of the images were posted by individual users and by cultural organisations. The images posted included recent events of the institutions or images of the exhibits of the museum. Facebook and Instagram now dominate the social media use landscape. Instagram is one of the fastest growing social networks globally among young people. In terms of what users can do with the use of Instagram in museums is to communicate their experiences through both choice of photo subjects and ways they choose to modify and present them (Weilenmann, Hillman and Jungseelius, 2013).

As Instagram is a relatively new form of communication (Hu, Manikonda and Kambhampati, 2014) it gives users the opportunity to communicate their experiences through photos (Weilenmann, Hillman and Jungseelius, 2013). Instagram is an online photo-album that was launched in 2010 (Hochman and Manovich, 2013); it enables users to take photos and videos with their mobile devices, perform some basic processing by applying digital filters and share them on a variety of social media platforms. As every moment counts, or at least so it seems through the eyes of social media users (Hochman and Schwartz, 2012). Instagram is ‘an application that combines and makes use of smartphones with cameras, offers the possibility of constant access to social media, thus enabling easy sharing of images of people’s lives’ (Araújo, Corrêa, da Silva, Prates and Meira, 2014:20).

As a social network, Instagram allows users to create a personal profile and build relationships by following other users, thus creating asymmetrical relationships (Zappavigna, 2016). The larger the number of followers an individual has, the greater their perceived social influence (Jin and Phua, 2014). A key idea behind the application is to share photographs and videos by using a hashtag (#) so that other users can find the photographs (Sheldon and Bryant, 2016). Users can also choose to tag an image or video during the publishing process, either independently as they wish to describe it or by following a predetermined dictionary of tags (Schwartz and Halegoua, 2014). They can also use the @ symbol, when adding captions thereby mentioning other users. This effectively links their posts to the referenced user’s account, before posting them (Hu, Manikonda and Kambhampati, 2014).
Instagram expands the array of aspects of everyday life that become organized in relation to flows of images. Navigation seems to be natural. Users interact with the screen of a smartphone and can see the photographs uploaded (Carah and Shaul, 2016). The online photography facility, such as uploading photos, as provided by Instagram, is a symbol of the reality that the user experiences as a mobile digital citizen. In some social media, photographs have become a central element to the post that is uploaded but in Facebook and Instagram, photographs are used as a point of reference as users can also make comments, likes and sometimes sharing (Budge and Barnes, 2017).

A key Instagram function, launched in 2016, is Instagram Stories. With Instagram Stories, users have the ability to create a digital photo story with all their daily highlights. This shared image and video experience can also be enhanced by e.g. using drawing tools and embedding stickers and emoticons in order to make it more appealing (Veissi, 2017). Instagram Stories last 24 hours before they disappear from a users’ profile. In terms of Instagram use in the museums, Instagram is used as a medium for the visitors in order to communicate their experiences in museums. (Weilenmann, Hillman and Jungselsiu, 2013).

As Instagram is quite a new medium in the context of cultural organizations there is not much published work about the use of the medium in the museums. A few studies appear in Flickr too. Weilenmann, Hillman and Jungselsiu (2013) focused on the Gothenburg natural history museum and its visitors’ experience. They extracted 222 Instagram posts and performed 14 interviews with the visiting Instagrammers, highlighting the reasons contributing to their postings. Lazaridou, Vrana and Paschaloudis, in their work in 2017 about the use of Instagram in museums and galleries, suggest the use of the application by museums so that they can enhance their internet presence and reach further to more potential visitors (Lazaridou, Vrana and Paschaloudis, 2017). In 2017, Budge and Burness analyzed visual and textual posts by Instagrammers, related to their experience. They argue that museum visitors post on Instagram guided by their experience and draw attention to exhibition objects (Budge and Barnes 2017). The following year, Budge and Burness (2018) expanded their research in the Museum of Contemporary Art in Sydney, Australia. Their study aimed at investigating the visitor’s Instagram posts using the museum geotag. The results showed that visitors were keenly engaged with the objects while they were communicating them through their photos.

Suess (2018) referred to the use of Instagram by visitors to the Gerhard Richter exhibition at the Queenslands Gallery of Modern Art. The results of his work showed that Instagram lets visitors to surpass the physical space and enhance their aesthetic experience. Villarspesa and Wowokowych (2020:11) “examined people’s behaviours using Snapchat and Instagram stories. Social ephemeral photography is often motivated by capturing a feeling, an aesthetically pleasing museum object, sharing an experience, and building self-identity”.

The dataset was constructed by initially visiting the Hellenic Ministry of Culture webpage and extracting the names of all archaeological museums in Greece, yielding a list of 125 museums. The next step involved searching Instagram for the museum accounts and their web presence, either on their own websites, if any, (owned) or hosted under The Ministry of Culture, the respective Archaeological Ephorate or Municipality (hosted). The process revealed that only the Acropolis Museum in Athens maintains an active Instagram account. At this point however, each museum’s type of website (owned or hosted) was recorded as well as the number of different social media platforms they are engaged in, creating two dataset variables. Website variable values were coded as 0 for no presence, 1 for a hosted website and 2 for a dedicated, owned website. For the social media presence variable, we encountered cases with 0-3 different platforms. Three additional dataset variables were introduced for the geographical region the museums belong to (coded with numbers 1 to 13 in ascending order of population), the actual region population (in thousands) and a binary variable that characterizes a museum as being central (Athens or Thessaloniki, coded as 1) or peripheral (rest of Greece, coded as 0). This research methodology was developed for the purpose of this study and it can be justified as it is expected that generating and measuring variables related to web presence and other social media engagement, as well as region related variables will provide a basis for quantifying each museum’s impact to the public and resulting user interaction.

As stated above, the Acropolis Museum is the only archaeological museum maintaining an Instagram account. This does not mean however that the museums do not appear at all on Instagram-mediated dialogue. In fact, for quite a few of them the name-search on Instagram returned location and hashtag results. There were 72 museums with location results and 53 with no location results, which were discarded from the dataset and the rest of the analysis. For the 72 museums, a note of the occurrences where the museum is tagged was taken, distinguishing between five different categories with respect to the image content: exhibits, people, people and exhibits, outdoor scenes and other. Overall, the dataset consists of 72 observations and 11 variables, tagged and the related five content categories, website and social media and the three region related variables.

4 FINDINGS AND DISCUSSION

A statistical analysis of the constructed dataset was performed leveraging Python (Anaconda distribution) in an interactive Jupyter notebook. The .csv file was read into a Pandas dataframe object (first five rows are shown on Table 2), enabling rapid calculation of descriptive statistics measures and execution of more advanced statistical tests. It is noticed at this point, that there are museums that returned location results with no tagging occurrences though.

3 METHODOLOGY OF THE STUDY
Understanding the dataset and gaining insight on its structure is crucial and has to be established (aided by descriptive statistics and various visualizations) before proceeding with hypothesis testing and more advanced methods. As an example, Table 3 depicts the pairwise Pearson correlation table, where it is noticed that all correlations are positive. This is of course partially due to the specific coding scheme chosen, e.g. for region.

According to Cohen’s standard, inter-variable correlations can be interpreted as weak [0.1–0.29], intermediate [0.3–0.49] and strong [0.5–1]. Following this scheme, different levels of positive correlations between the variables can be identified, e.g. tagged is in strong positive relation with all content categories (exhibits 0.922, people_exhibits 0.905, people 0.825, outdoor 0.808, other 0.703) which is somewhat expected as tagged is the sum of the five content categories. This ordering of the coefficients is however, indicative of people’s tagging behavior and preferences, i.e. exhibits rather than outdoor scenes. Figure 1 displays the tagged categories as a percentage of the total number of tags that the museums have received. Out of the 1561 total tags, 50% of them are exhibits, while the other four categories account for the 17%, 15%, 14% and 4% of the total tags respectively.

Variable tagged is also in weak (to moderate) positive relation with website, social media and population. The correlations in Table 3 were calculated with the pandas “corr” function passing “pearson” as the method to use. Some further experimentation with the non-parametric (more details on normality given later) tests Spearman and Kendall revealed that the variables tagged and population are uncorrelated (retain H0) at 0.95 significance level (Spearman coefficient 0.222, p=0.061, Kendall coefficient 0.164, p=0.063). The mean, standard deviation and Tukey 5-number summary for all variables are shown on Table 4 below. The museum of Thessaloniki is the one mostly tagged (350), followed by the museum of Piraeus (196), the museum of Heraklion (139) and the Acropolis museum in Athens (122). These are all large Greek cities and one might expect some degree of correlation with the region population. This however has not been verified as seen previously, at least to a significantly large extent.

The skew of the variable distributions has also been calculated. All of the variables have positive skewed distributions (ranging from 6.713 for people_exhibits variable to 0.174 for region variable) apart from the social_media variable (-0.685). This has also been verified by also looking at the histogram and density plots (not shown here). The presence of outliers is apparent especially with the box-whisker plots (again not shown here). This fact, combined with the small sample size, may have some undesired effects in the statistical analysis further on but for the scope of this contribution, no outlier elimination process was employed.

At this point, normality tests were conducted for the variable distributions, employing the Shapiro-Wilk Test and the D’Agostino K2 Test. In some cases, a power transform like Box-Cox was utilized. Results showed that tagged, exhibits, social_media, region and population follow a Gaussian-like distribution, whereas the rest of the variables are non-Gaussian distributed. Following these normality tests, independent T-tests were employed for the Gaussian variables and Mann-Whitney U tests for the non-Gaussian ones to argue about the similarity or the difference between central and peripheral museums. All tests have revealed similar distributions. Results at the 0.95 level are summarized on Table 5.

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A popular effect size measure for quantifying the difference between groups is the Cohen’s d measure. In this case, the central and peripheral groups have been shown to be similar. As an example, the Cohen’s d measure was calculated for the case of tagged variable and found a score of 1.827, which can be interpreted as very large (according to initial interpretations by Cohen, subsequently expanded by

<table>
<thead>
<tr>
<th>variable</th>
<th>test type</th>
<th>distributions</th>
<th>statistic</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
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<td>T-test ind.</td>
<td>Same (retain H0)</td>
<td>1.644</td>
<td>0.151</td>
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<tr>
<td>exhibits</td>
<td>T-test ind.</td>
<td>Same (retain H0)</td>
<td>1.373</td>
<td>0.218</td>
</tr>
<tr>
<td>social_media</td>
<td>T-test ind.</td>
<td>Same (retain H0)</td>
<td>1.074</td>
<td>0.315</td>
</tr>
<tr>
<td>people_exhibits</td>
<td>Mann-Whitney U</td>
<td>Same (retain H0)</td>
<td>224.500</td>
<td>0.480</td>
</tr>
<tr>
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<td>Mann-Whitney U</td>
<td>Same (retain H0)</td>
<td>198.000</td>
<td>0.277</td>
</tr>
<tr>
<td>outdoor</td>
<td>Mann-Whitney U</td>
<td>Same (retain H0)</td>
<td>198.000</td>
<td>0.280</td>
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<tr>
<td>other</td>
<td>Mann-Whitney U</td>
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<td>196.000</td>
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<td>website</td>
<td>Mann-Whitney U</td>
<td>Same (retain H0)</td>
<td>212.500</td>
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</table>

Table 2. First five rows of the dataset

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<th>people</th>
<th>exhibits</th>
<th>people</th>
<th>outdoor</th>
<th>other</th>
<th>social_media</th>
<th>region</th>
<th>population</th>
<th>central</th>
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</thead>
<tbody>
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<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
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<td>0</td>
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<td>0</td>
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</tbody>
</table>

Table 3. Pairwise Pearson correlation table

<table>
<thead>
<tr>
<th>variable</th>
<th>exhibits</th>
<th>people</th>
<th>exhibits</th>
<th>people</th>
<th>outdoor</th>
<th>other</th>
<th>social_media</th>
<th>region</th>
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</thead>
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<td>0.905</td>
<td>0.825</td>
<td>0.808</td>
<td>0.703</td>
<td>0.286</td>
<td>0.397</td>
<td>0.783</td>
<td>1</td>
</tr>
<tr>
<td>exhibits</td>
<td>0.922</td>
<td>1</td>
<td>0.905</td>
<td>0.825</td>
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<td>0.703</td>
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<td>0.397</td>
<td>0.783</td>
<td>1</td>
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<td>0.905</td>
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<td>0.286</td>
<td>0.397</td>
<td>0.783</td>
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<tr>
<td>outdoor</td>
<td>0.825</td>
<td>0.825</td>
<td>0.825</td>
<td>1</td>
<td>0.744</td>
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<tr>
<td>other</td>
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<td>0.808</td>
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<tr>
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<td>0.783</td>
<td>1</td>
<td>1</td>
</tr>
</tbody>
</table>

Table 4. Mean, standard deviation and 5-number summary for dataset variables
Sawilowsky, 2009). Thus, it can be argued that with a very large effect size measure, the T-test result holds firm.

The final objective under investigation is to model the number of tags each museum receives, based on region population, social media presence and website status as predictors. Multiple linear regression was carried out in RStudio (within Anaconda environment as well). After loading the .csv formatted dataset the linear model was built and its coefficients displayed (Table 6).

<table>
<thead>
<tr>
<th>Coefficients:</th>
</tr>
</thead>
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<td>Estimate</td>
</tr>
<tr>
<td>Intercept</td>
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<tr>
<td>website</td>
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<tr>
<td>population</td>
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Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 48.91 on 69 degrees of freedom
Multiple R-squared: 0.2114, Adjusted R-squared: 0.1751

Thus, the model equation can be written as:

\[ \text{tagged} = -19.649 + 24.417 \times \text{website} + 0.013 \times \text{population} \]

The model confidence intervals could have been calculated with the R function confint(model) but are not shown here as the website predictor is retained as well. In multiple linear regression, R2 represents the correlation coefficient between the observed values of the outcome variable (y) and the fitted (i.e., predicted) values of y. A problem with the R2 is that, when more variables are introduced to the model, R2 always increases. This happens even in cases when these predictors have a weak association with the response (Gareth, Witten, Hastie and Tibshirani, 2014:79-80). A solution is to adjust the R2 (Adjusted R Square) by taking into account the number of predictor variables. Here, the adjusted R2 value suggests that only 11% of the variance in the measure of tagged can be predicted by website status and region population. Therefore, as a conclusion, the model is rather unsatisfactory and other features should be surveyed.

The findings of the study are in accordance with the relative literature which shows that visitors are primarily engaged with the objects as most of the visitors of our sample gave emphasis to photographs of exhibits. The paper tried to explore the role of subjectification in the choice of the visitors and examine how visual choices could affect the visitor’s impression of the museum. The study seeks to alert museums to the importance of using social media as a means of promoting museums and their exhibits. It is evident that Instagram, largely based around image, constitutes a very important visual information provision channel.

5 CONCLUSIONS

The findings of this research indicate that most of the museums use at least one to three social media platforms for communication and marketing reasons. That means that with the use of social media, museums have the choice to illuminate and explore some tensions, such as, to highlight knotty juxtapositions, to share public experiences and private importance seems to be underestimated. As a conclusion, the model is rather unsatisfactory and other features should be surveyed.

Even though the study of Draskalaki et al (2020) found that in an Archeological Museum there is limited interaction between exhibits and visitors, in our study most of the uploaded photos include museum exhibits. This, according to Budge and Burness (2017), means that attention to exhibition content is specifically drawn by objects. Some people though,
upload photos of themselves along with the exhibits – those people could be influencers. As influencer marketing is partly viral marketing (Chatzigeorgiou 2017), museums can exploit these posts in order to reach and attract more visitors.

Finally, as the statistical analysis has shown, there is no significant difference related to the locations of the museums, as both major archeological museums in Athens and Thessaloniki, as well as museums in the Greek province have similar distributions in all the dataset variables.

REFERENCES


Kyratis, P., Chatzigeorgiou, C., Tsagkaris, A. & Dinopoulou, V. (2010). Failure mode and effect analysis (FMEA)


