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Emotional Maintenance: A Digital Model to Support Maintenance Decisions in Buildings' Coatings

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Abstract: Maintenance decisions at the end of building components' service life are frequently11driven by subjective motivations, that can arise from various sources, including the building12owner's personal preferences, sentimental attachments, aesthetic considerations and individual/col-13lective preferences or sense of taste. This study support decision-making regarding maintenance14actions by combining objective indicators of building degradation and the subjective user perceptions to prioritize areas of focus, determine appropriate maintenance strategies, and allocate resources effectively.17

Keywords: Maintenance; Digital model; Automation of inspection; dwellers' feelings; Rendered facades 19

1. Introduction

Maintenance decisions at the end of building components' service life are frequently 22 driven by subjective motivations [1]. Subjective motivations can arise from various 23 sources, including the building owner's personal preferences, sentimental attachments, 24 aesthetic considerations and individual/collective preferences or sense of taste [2]. It is 25 important to acknowledge that subjective motivations for maintenance decisions, beyond 26 technical requirements, hold value as they reflect the human element and emotional con-27 nections associated with buildings. Balancing these subjective motivations with technical 28 considerations and financial constraints is essential for making well-informed decisions 29 that address both practical needs and emotional connections. In this study, a digital model 30 is proposed to assess the degradation state of buildings' coatings, based on in situ inspec-31 tions (using robotized platforms and a high-resolution camera), while engaging the build-32 ing's occupants, tenants, or users to understand their perceptions, experiences, and con-33 cerns regarding the building's condition. A stains' detection algorithm, using multispec-34 tral analysis, is developed to identify the area affected by stains (a visible sign of degra-35 dation), which affect the facade's aesthetic appearance but also contribute to the worsen-36 ing of other anomalies and consequent degradation and loss of properties of the coating. 37

2. Case study - Bairro de Alvalade, in Lisbon, Portugal

The methodology was applied in Bairro de Alvalade, in Lisbon, Portugal, a neighborhood located at north part of the city. Alvalade is a neighborhood with a population of 30.000 people and an area of 5.34 square kilometers. The study was focused in three identical buildings, with different conditions, named building 4, 7, and 9 [3].

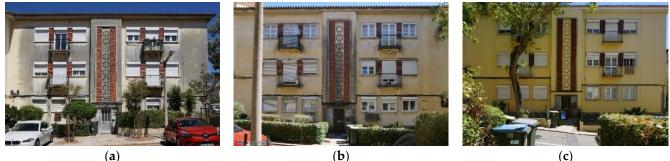
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(a)

Figure 1. Façades' analyzed: (a) Buildings 4; (b) Buildings 7; (c) Buildings 9 (adapted from [1]).

3. Final Digital Model

The methodology proposed includes 3 main steps: Automatic evaluation of façades; Dwellers' feelings survey; Final digital model. In the following sub-section each of the steps is described and the results presented.

3.1. Automatic evaluation of façades

First, a collection of images of the façades were acquired, tacking advantages of using terrestrial robotic platforms to perform a planned survey [3]. Then, a mosaic image is computed to from the set of images acquired, Finally, an automatic identification of anomalies in facades, through image classification, is applied. In this case, a supervised classification performed in MultiSpec [4], a multispectral image data analysis software. The images were loaded, and the users creates classes by manually selecting specific areas on the image. Then, all of building facades is classified into the different classes created at pixel level [5,6].

3.2. Dwellers' feelings

An opinion survey was conducted among the residents and neighbors of the Alva-16 lade neighborhood to assess their perceptions of the degradation of façades. Twenty-five 17 respondents, comprising fifteen men and ten women, ranging in age from 17 to 80 years 18 (with an average age of 48 years), provided their impressions of the facade's condition of 19 the three buildings (Figure 2). 20

An analysis of these perceptions, based on the Vader analysis [7], was carried out to 21 create acceptance profiles (Table 1). VADER gathers the sentiment scores linked to words 22 from a lexicon and aggregates them to determine the score for sentences. The compound 23 score is computed by summing the valence scores of individual words in the lexicon, with 24 certain adjustments, and falls within a range from -33 (indicating highly negative senti-25 ment) to +33 (reflecting strongly positive sentiment) [7,8]. 26

3.3. Digital model

A digital model is assembled, combining the quantitative and qualitative data col-28 lected, reflecting the physical degradation condition and the dwellers' perception, respec-29 tively. This model can be used to perform a comprehensive diagnosis and, therefore, sup-30 port maintenance decisions. The information on each façade processed, automatic inspec-31 tion and dwellers' feeling, was merged in a 3D model and can be visualized, as exempli-32 fied in Table 2. 33

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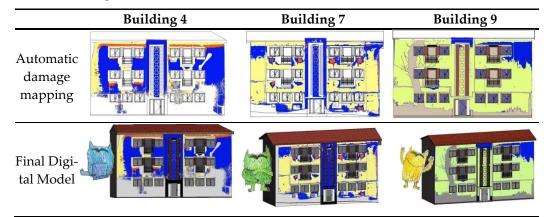
Building 4				Building 7				Building 9			
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	Negative	-1,333 -1,818 -3,333 -4,494 -4,878 -5,263 -6,452 11,111 12,500	64,820		Negative	0,000 -1,250 -1,786 -2,128 -2,759 -3,333 -5,556 -7,143 -8,571	-65,858		Negative	0,000 0,000 0,000 -0,962 -1,818 -2,128 -6,061 -6,667	-27,635

-33,333

Table 1. Dwellers' feelings.

Table 2. Final Digital Model.

-13,636



4. Discussion of the Results

Building 4 exhibits the most extensive physical degradation on the façade's coating 4 (Figure 1 and Table 2). This building received more negative and neutral reactions from 5 the respondents regarding their assessment of the degradation state, in comparison with 6 the other two buildings. However, the differences between Building 4 and Building 9 are 7 not as significant. Building 9, which is in better condition, received an overall less negative 8 response, albeit with more neutral sentiments and less overall positive feelings. The re-9 sults indicate that respondents can clearly identify the building in the worst condition, not 10 having such a clear idea of the condition of the other two buildings. 11

5. Final Remarks

The methodology presented enables an emotional maintenance procedure for buildings', through the development of a digital model that support maintenance decisions in buildings' façades. The digital model allows identifying patterns, correlations, and discrepancies between objective degradation indicators and subjective user feedback. The dwellers' feelings are modelled based on onsite interviews to collect qualitative data on their observations and satisfaction levels. This study support decision-making regarding 18

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maintenance actions by combining objective indicators of building degradation and the subjective user perceptions to prioritize areas of focus, determine appropriate maintenance strategies, and allocate resources effectively. 3

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Conflicts of Interest: The authors declare no conflict of interest.

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