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Conversion of a digital camera into a non-contact colorimeter for use in stone cultural heritage: the application case to Spanish granites

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1 Abstract

2 In this study, a digital CMOS camera was calibrated for use as a non-contact 3 colorimeter for measuring the color of granite artworks. The low chroma values of the 4 granite, which yield similar stimulation of the three color channels of the camera, 5 proved to be the most challenging aspect of the task. The appropriate parameters for converting the device-dependent RGB color space into a device-independent color 6 7 space were established. For this purpose, the color of a large number of Munsell 8 samples (corresponding to the previously defined color gamut of granite) was measured 9 with a digital camera and with a spectrophotomer (reference instrument). The color data 10 were then compared using the CIELAB color formulae. The best correlations between 11 measurements were obtained when the camera works to 10-bits and the 12 spectrophotometric measures in SCI mode. Finally, the calibrated instrument was used 13 successfully to measure the color of six commercial varieties of Spanish granite.

14

15 Keywords: CIELAB system; CMOS camera; color calibration; cultural heritage; 16 granite; ornamental stone; RGB values.

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25 **1. Introduction**

26

27 Color is one of the most important visual properties of ornamental and monumental 28 stone. Color changes caused by weathering and decay greatly influence the aesthetic 29 value of stone. Extent of such change can be quantified by contact-type color measuring 30 devices (colorimeters and spectrophotometers [1-6]) and analyzed in a device-31 independent color space, such as CIE-XYZ or CIE-L*a*b*. But these devices present 32 some limitations: (1) sometimes is not possible to reach the target object with the 33 instrument, (2) they are more expensive and complex than other non-dedicated color 34 measuring devices (digital cameras, scanners and even mobile-phone cameras) and (3) 35 as the field of view of contact-type color devices is limited, measurement of 36 heterogeneous surfaces produces unrealistic color values. To overcome these 37 limitations, digital cameras can be used because (1) the field of view is only limited by 38 the size of the appropriately illuminated area, (2) contact with the target object is not 39 required, and (3) they encode each point of the entire surface simultaneously, thus quantifying surface characteristics and defects. 40

41 Digital cameras only detect changes in light intensity, not color. To encode color, they 42 require three different filters in addition to the sensors. These filters usually have 43 spectral bands in the red (R), green (G) and blue (B) regions, and therefore the encoded 44 values are RGB digital values. RGB is a device-dependent color space as the filters and 45 other parameters are specific to individual cameras and can be changed with camera 46 settings such as the spectral exposure level, white balance and the dynamic range. As 47 RGB values cannot be transformed to XYZ or L*a*b* values directly by using a 48 standard formula, a transformation that defines the mapping between RGB digital 49 values and a device independent color space is necessary. This process is known as 50 camera characterization [7]. Several camera characterization techniques have been used 51 with the aim of developing a model (and estimating its parameters) for obtaining 52 L*a*b* color measurements from RGB digital values (e.g. [7-15]). In general, these 53 techniques can be divided into two categories: (1) spectral characterization, which 54 measures the three spectral-sensitivity functions for the red-green-blue (RGB) channels 55 and requires a monochromator and a radiance meter [16]; and (2) colorimetric 56 characterization, which involves mathematical transformations that yield the tristimulus 57 values from the digital values and which require use of a reference target that contains a 58 certain number of color samples. In the present study, we used the latter color target-59 based approach, which only requires a certain number of color samples and is, therefore, 60 a more practical method [7]. We chose target-based characterization procedure 61 described by Hong et al. [7], which is based on polynomial modeling. This calibration 62 model has been used successfully in nearly two hundred scientific papers with different 63 objectives, e.g., to determine how facial skin coloration affects perceived health of 64 human faces [17, 18] and for use in dental color matching [19].

66 In the field of lithology, the image captured by the camera is usually processed by 67 different segmentation strategies. For example, one innovative application focuses on 68 the segmentation of decay zones from images of stone materials [20, 21]. Another 69 strategy enables improvement and semi-automatization of the study of chemical decay 70 causing visible changes in color of some regions [22]. A portable stereo active vision 71 system (AVS) has also been specifically designed to perform on-site processing of the 72 data acquired in the field of cultural heritage conservation [23]. Moreover, the digital 73 decorrelation of RGB images by Principal Components Analysis (PCA) enables 74 contrast enhancement of minority elements apparently absent from the initial RGB 75 digital image [24-27]. Camera characterization has been used in very few studies, 76 including that of Chorro et al. [28], who used the sRGB model to predict the CIE-XYZ 77 tristimulus values depending on the RGB digital data, with the final aim of quantifying 78 color changes in the appearance of a paving stone (marble) in relation to the viewing 79 distance. More recently, Concha-Lozano et al. [29] used spectroradiometric 80 measurements to calibrate a camera in order to establish the color ranges within which 81 replacement of biodetritic limestone in medieval walls will be imperceptible. 82 Nevertheless, to our knowledge, camera characterization has not previously been 83 reported for granite. Measurement of the color of granite is complicated by the low 84 chroma and spatially heterogeneous color, which is formed by the different colors of the 85 constituent minerals. There is great interest in measuring the color of granite because, 86 amongst other reasons, granite is one of the most commonly used types of igneous rock 87 owing to its abundance and great variety of color and textures, and because it is a major 88 construction material in European historical buildings and monuments [30].

89

90 The present study focused on developing a method of RGB digital camera colorimetric 91 characterization for studying stone, specifically granite. The nearly neutral colors of 92 granite yield similar stimulation of the three color channels of the camera (red, green 93 and blue), which makes the task in hand particularly challenging. For the first time, the 94 settings of a digital camera have been adjusted to obtain the camera response closest to 95 that of the reference instrument (spectrophotometer) for granite color measurement 96 using the CIELAB system. The developed method was successfully used to measure the 97 color of granite samples. This is of particular interest in the field of stone conservation, 98 in which innovative non-invasive tools for monitoring the aesthetic changes in stone 99 surfaces are required.

- 100 2. Experimental
- 101 2.1.Fine-tuning of the camera calibration method
- 102

103 The methodology developed for estimating the RGB \rightarrow L*a*b* transformation consisted 104 of two parts. In the first part, we determined the appropriate settings and working 105 conditions of the acquisition system (camera) and reference instrument 106 (spectrophotometer). In the second part, we selected a large set of Munsell matte and 107 glossy samples corresponding to the previously defined color gamut of granite [31]. The 108 colors of these samples were measured using both devices under the conditions

109	indicated in the first part. The digital images were obtained with the following image
110	acquisition system (Figure 1):
111	
112	• PixeLINK PL-A782 color digital camera, 2008 (suitable for industrial use), with
113	CMOS sensor architecture, 6.6 Mega Pixels of resolution and a user-selectable 8
114	or 10-bit output. The camera was placed vertically at a distance of 112 cm from
115	the sample. The angle between the axis of the camera and the source of
116	illumination was approximately 45°. Thus, following the CIE nomenclature [32],
117	the measurement geometry was 45°x90° or 45/0, which is very common in
118	industrial applications in order to avoid specular reflection.
119	• Camera lens: Fujinon CF50HA-1, 50 mm focal length, 1", designed to be used
120	with high resolution cameras with images up to 1.5 Megapixels, with manual iris
121	and focus.
122	• Lighting was achieved with Kaiser RB-5004-HF high frequency daylight copy
123	light set with four Oxram Dulux L 36W/954 fluorescent light tubes (41.5 cm in
124	length), with a correlated color temperature of 5400 K (natural daylight) and a
125	color rendering index (R_a) close to 90%.
126	• The room where images were taken was totally dark and a black cloth was
127	placed on the floor under the table used as the sample stand, to minimize
128	background light.
129	• The size of the captured images was 240 pixels (width) by 192 pixels (height).
130	The pixel size was 347 x 375 μ m ² . The images were stored in uncompressed
131	tagged image format files (TIFF).
132	• The camera settings used in the present study are summarized in Figure 2. The
133	main purpose of this step was to maintain constant any software camera control
134	(white balance, exposure time, gain, etc) to obtain a stable, reliable and
135	reproducible RGB color space, although this would presumably limit the
136	dynamic range of luminance of the camera [33].
137	
138	The lighting level, and its uniformity, is critical for image acquisition, so that the
139	camera can deliver meaningful, repeatable data [34]. Therefore, the lighting map for the
140	reference target needs to be as spatially and temporally uniform as possible. The
141	uniformity of light intensity was tested using a radiometer (DHD 2302.0, HERTER)
142	(<i>Figure 3</i>).
143	
144	However, the combination of lens aperture size and exposure time determines the
145	amount of light reaching the CMOS sensor of the camera. Obviously, the signals
146	generated by the CMOS sensor vary with the amount of light reaching CMOS sensor.
147	Therefore, both aperture size $(f/4)$ and exposure time (99.537 ms, milliseconds) were
148	fixed during the period of image acquisition. We also totally occluded the camera-lens
149	aperture for the black reference, and we captured a standard white reference plate for

- 150 the white reference.
- 151

152 The camera is capable of both 8-bit depth and 10-bit depth linear data acquisition; both were used in the present study. 8-bit data can hold $2^8 = 256$ possible values ranging 153 from 0 to 255. For an RGB image in which the values are 8-bit unsigned integers, 0 0 0 154 155 represents black, 255 255 255 represents white, 255 0 0 represent red, 0 255 0 represents green, and 0 0 255 represents blue. 10-bit data yields $2^{10} = 1024$ possible 156 157 values, ranging from 0 to 1023. For an RGB image in which the values are 10-bit 158 unsigned integers, 0 0 0 represents black, 1023 1023 1023 represents white, 1023 0 0 159 represents red, 0 1023 0 represents green, and 0 0 1023 represents blue. Special attention was paid to setting the exposure to avoid any "color clipping" for the white 160 161 reference, i.e., saturation of one or more of the three RGB channels, obtaining R, G or B 162 values above 255 with 8-bit data and 1023 with 10-bit data [7, 13].

163

The spectrophotometer used was a portable spectrophotometer (Konica Minolta CM-164 700d) equipped with CM-S100w (SpectraMagicTM NX) software. The measuring 165 conditions were illuminant D65, observer 2° and a 3-mm diameter viewing area. 166 167 Measurements were made in both specular component included (SCI) and specular 168 component excluded (SCE) modes to determine which mode approximates better to the 169 camera vision. The SCI mode, in which the gloss trap of the spectrophotometer is 170 closed, includes the total reflectance (considering both specular and diffuse reflections); 171 the SCE mode, in which the gloss trap is open, includes the diffuse reflectance and 172 excludes most of the specular component and is therefore more sensitive to color 173 differences due to differences in surface roughness [31, 35]. It is generally accepted in 174 the field of color science that the SCE mode approximates the view with the naked eye 175 and the SCI mode is adequate for analyzing the intrinsic color of objects [31, 36].

176

In attempting to adjust the camera settings to make the camera response more similar to 177 178 the reference-instrument response in the CIELAB system, standard color targets 179 consisting of an assortment of color patches are commonly applied. The Gretag 180 Macbeth color-checker color rendition chart [37] is one of the most commonly used, although it consists of only 24 patches. In some cases, as in the second part of our 181 182 camera characterization method, a customized characterization target, consisting of a 183 large number of patches, was designed and applied. Thus, a set of samples (212 Munsell 184 color charts, 125 from the glossy and 87 from the matte collection), corresponding to 185 the three-dimensional color area of the CIELAB space, in which the color of the 186 ornamental granites is defined [31] was selected. In each of the 212 color samples, the 187 $L^*a^*b^*$ color values were measured using the portable spectrophotometer under the 188 measuring conditions described above. One reading was taken per sample. An RGB 189 digital image was also taken of each Munsell sample/chip. The digital camera was 190 placed orthogonally to the Munsell sample. The field of view of the camera was fully occupied by a single Munsell chip. Thus, 212 RGB measurements, i.e., R, G and B 191 192 color values, were obtained. Note that the granite color is located in the nearly neutral 193 region of CIELAB color space, far from the highly saturated colors like intense or pure 194 yellows, reds and greens. This makes it difficult for the instrument to measure the color, 195 as the nearly neutral colors yield similar stimulation of the three color channels (red,

196 green, and blue) of the camera, and the differences between these colors constitute small197 variations in a high nearly constant background signal [10].

- 198
- 2.2. Performance testing and verification of the resulting calibration
- 199 200

201 To confirm selection of the camera working conditions, the method described in Section 2.1 was applied to the color characterization of granite samples. Six commercial 202 203 varieties of granite (Aldán, Blanco Cristal, Grissal, Monte Enxa, Rosa Porriño and *Silvestre*) were considered. Data on the origin, geochemistry and textural and mineral 204 205 characteristics of each type of granite are shown in Table 1. Five square specimens (25 or 36cm²) of each type of granite were prepared with a honed surface finish. An image 206 207 of each specimen was taken using the image acquisition system described in Section 2.1. During the process, each of the samples was placed on the marked area of a light 208 table. The measurement area in the specimens was approximately 6.25 cm² (width, 25 209 210 mm and length, 25 mm). The color of granite samples was then measured with a 211 portable spectrophotometer (Konica Minolta CM-700d) equipped with CM-S100w 212 (SpectraMagicTM NX) software, following the working methodology designed by 213 Prieto et al. [38]. The measuring conditions and specular component modes were the 214 same as those used to measure the Munsell samples (see Section 2.1).

215 **3. Results and discussion**

The stability of the light source was evaluated prior to establishing the color measurement protocol for the study. Figure 3 shows the light levels (in millilux = 10^{-3} lux or lx) across the table top. The light was not completely homogeneous and varied from 1560 ± 20 lx at the upper center to 2400 ± 20 lx to the right and left of the middlecenter. An area of the table where the percentage of light level (in millilux) did not vary by more than 3% was marked. The average level of lighting was 1780 ± 20 lx within this area, which is where the images were captured.

223

The color of the 212 test samples (Section 2.1) was measured using both devices. Hong et al. [7] noted that better results can be achieved if more terms (e.g. R^2 , G^2 , B^2 , etc) are included to the matrix derived by the characterization process of the digital camera. In the present study, a third order polynomial (matrix with 20 terms) was used. This can be expressed as follows [9]:

229

$$D = \begin{bmatrix} L_1^* & a_1^* & b_1^* \\ \vdots & \vdots & \vdots \\ L_n^* & a_n^* & b_n^* \end{bmatrix}$$
(1)

233
$$V = \begin{bmatrix} 1 & R_1 & G_1 & B_1 & R_1G_1 & R_1B_1 & \cdots & R_1^3 & G_1^3 & B_1^3 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots & \vdots \\ 1 & R_n & G_n & B_n & R_nG_n & R_nB_n & \cdots & R_n^3 & G_n^3 & B_n^3 \end{bmatrix}$$
(2)
234

$$235 D = M \cdot V (3)$$

236

237
$$M = [m_{ij}]_{3x20} = [(V^t \cdot V)^{-1} V^t \cdot D]^t$$
(4)

238

where M is the matrix with the transformation polynomial coefficients characterizing the camera, { R_n , G_n , B_n } are the digital levels of the training color patches (i.e. 212 color charts-Section 2.1) measured by the camera and { L^*_n , a^*_n , b^*_n } are the CIE-L*a*b* values of the training set (i.e. 212 color charts-Section 2.1) measured by the spectrophotometer. Finally, transformation of the RGB values was achieved by using the following equation:

245
$$\begin{bmatrix} L^* \\ a^* \\ b^* \end{bmatrix} = \begin{bmatrix} m_{ij} \end{bmatrix}_{3x20} \cdot \begin{bmatrix} 1 \\ R \\ G \\ B \\ RG \\ \vdots \\ B^3 \end{bmatrix}_{20x1}$$
(5)

247

248 The absolute color measurement by the camera, calculated using Eq. (5), in both 8 and 249 10-bits of color depth, was compared with the external reference provided by the 250 spectrophotometer, in both specular component included (SCI) and excluded (SCE) 251 modes. The CIELAB coordinates of each measured chip, obtained separately by the 252 camera and the reference instrument were compared, taking into account the classical 253 CIELAB formula (ΔE^*_{ab}) and the other color difference formulae based on CIELAB 254 space (ΔE_{94} , ΔE_{00} and CMC). The results obtained are shown in Table 2, which 255 includes the average, maximal and minimal values of the computed total color 256 differences, viz. ΔE_{ab}^* , ΔE_{94} (2:1:1), ΔE_{94} (1:1:1), ΔE_{00} (2:1:1), ΔE_{00} (1:1:1), CMC (2:1) 257 and CMC (1:1). No equivalence of scale factor was found in the values calculated using 258 the different formulae considered, as reported by Prieto et al. [38] on comparing the 259 results obtained by measuring granite samples with the different reading areas (or 260 measuring head sizes) of a spectrophotometer and a colorimeter. It is difficult to specify 261 admissible color differences between devices, because most recommendations on color 262 differences refer to situations in which the colors of different objects are measured 263 under the same illuminant, unlike in the present study (Section 2.1). The color-tolerance 264 concept is based on color discrimination, which largely depends on observational 265 conditions. In this case, we should take into account that the color of the sample was 266 viewed with different illumination, leading to greater color tolerance. For instance, 267 analysis of the color differences in both natural and artificial objects over one day 268 revealed values exceeding 3 CIELAB units when the color of the objects under the 269 maximum solar elevation was compared with that at twilight [39]. Based on these 270 findings, consideration of 1 CIELAB unit as the visual color difference threshold or just 271 noticeable difference (jnd), which constitutes the lower limit of perception in an 272 individual with normal color vision [35, 40] appears too strict in this case. Likewise 273 1.75 CIELAB units, considered as the suprathreshold color-difference [41]. Thus, we 274 decided to consider for evaluation of the results perceptual limits starting from 3 275 CIELAB units and taking into account the following established thresholds: (1) the 276 normal color tolerance, specified by Lozano [42] as being between 2.8 and 5.6 CIELAB

277 units (according to the usual conversion factors between color-difference units [40]); (2) 278 the acceptable color tolerance of 3 CIELAB units [43, 44]; (3) the normal limit of 279 perception in industrial or technical applications of 5 CIELAB units [45-47], and (4) the 280 perceptible but acceptable difference in color of 6 CIELAB units considered by Hardeberg [48]. We found that the average total color differences obtained, ranging 281 282 from 1.9 to 1.1 CIELAB units (**Table 2**), are nearly undetectable to the untrained eye. 283 The maximal total color differences, with values ranging from 3.7 to 6.9 CIELAB units, 284 must be considered virtually acceptable for most industrial applications. Furthermore, 285 the color difference formulae based on CIELAB space include three parametric factors, 286 k_L , k_C and k_H , which are correction terms for the variation in experimental conditions. 287 Under reference conditions, these are all set at 1 [32]. However, in the present study, the 288 illumination conditions were not reference conditions and the samples were not 289 homogeneously colored. For textured samples, it is not clear which values should be 290 used for the parametric factors [49-51]. Considering an increase in the relative 291 contribution of the lightness term (k_L parametric factor 2, instead of 1) in the color 292 difference formulae, the maximum value decreased greatly by between 2 and 3 293 CIELAB units, and only reached values of between 3.7 and 3.9 CIELAB units (Table 294 2). So, depending on the used color difference formula, the better setting could be 295 different. However, as it can be observed in the **Table 2**, if we adopt the maximal value 296 of the total color differences as the criterion of choice, in all cases except $\Delta E00$ (2:1:1), 297 these differences are lesser when the spectrophotometer on specular component 298 included (SCI) mode and the digital camera with 10-bits data acquisition were used. 299 Likewise, although with other combinations, the average total color differences were 300 slightly lower (maximum 0.4 CIELAB units lower with respect the conditions just 301 cited), in those cases were also achieved the biggest maximal color differences (up to 302 9.5 CIELAB units of difference with respect to the above cited conditions). 303 Consequently and considering the lowest maximal value of the total color differences as the selection criterion, although also comparing its results with the average and 304 maximum values of total color differences, the digital camera 10-bit depth linear data 305 306 acquisition is the best for our purpose and should be compared with SCI 307 spectrophotometric data.

308

309 In the cultural heritage field, most colorimetric measurements are used to estimate color 310 differences (e.g. [6, 52]). Therefore, to calibrate a digital camera as a colorimeter for use in this field, it is advisable to explore the discriminatory capacity of the camera and its 311 reliability for measuring small differences between very similar colors. A certain 312 313 number of color differences between pairs of nearest-neighbor chips were calculated separately by both the camera and the reference instrument, according to the classical 314 315 CIELAB formula (ΔE^*_{ab}) and other color difference formulae based on the CIELAB 316 space (ΔE_{94} , ΔE_{00} and CMC). Comparison of the results obtained with the camera and 317 the reference instrument indicated the discrepancy between the two devices. This 318 discrepancy was used to test the reliability of the camera performance and was 319 compared with the precision and tolerance of the devices (Tables 3 and 4). More than 320 half of the absolute discrepancies exceed the suprathreshold value for visual 321 discrimination of 0.887 CIELAB units [41]. Nonetheless, the values of the relative discrepancy were very low and although the absolute discrepancy exceeded the 322 uncertainty or precision of both devices, it remained within the camera tolerance $(1.32 \pm$ 323 324 1.06 vs. 2.4 CIELAB units). Thus, the camera and reference instrument showed a high degree of consistency in the estimation of small color differences, and therefore the camera performed well [53].

327

328 The selected camera working conditions were then used to characterize the color of six 329 commercial varieties of granite (Aldán, Blanco Cristal, Grissal, Monte Enxa, Rosa 330 Porriño and Silvestre). The results obtained (Figure 4) appeared sufficiently accurate 331 and reliable: considering the set of samples, regardless of type of granite and granite 332 sample, the total color difference (ΔE^*_{ab}) between the measured granite color (using the 333 spectrophotometer in specular component included (SCI) mode) and the estimated 334 granite color (using the digital camera with 10-bits data acquisition) was generally 335 below 6 CIELAB units. Specifically, the ΔE^*_{ab} values ranged between 2.7 and 5.5 336 CIELAB units for Grissal and 3.4 and 5.1 CIELAB units for Blanco Cristal, indicating that, with the measurement method used, the best results were obtained with achromatic 337 338 rocks. The values of ΔE^*_{ab} for *Monte Enxa* and *Rosa Porriño* ranged from 4.4 to 6.6 339 and from 4.3 to 7.0, respectively. These were the largest color differences reached in the 340 study and corresponded to those types of granite in which the color is farthest from the achromatic area. Intermediate values of ΔE^*_{ab} were obtained for Aldán, with values 341 342 within the range of 3.0 - 6.5 CIELAB units, and for Silvestre, with values within the 343 range 3.3 - 5.5 CIELAB units. In this case, differences of 6 CIELAB units cannot be 344 considered high as two different devices with different lighting conditions were used. 345 For granite color measurements, differences of nearly 3 CIELAB units are obtained, 346 even when using the same device with different measuring heads [38]. Moreover, the 347 limits of perception are usually calculated for homogeneous samples (in terms of color 348 and texture) (for further details, see, e.g. [54]), unlike the granite samples that were the 349 target of the present study.

350 **4.** Conclusions

351 A calibration procedure was developed for granite color measurement using a non-352 contact device (a CMOS digital camera). Working conditions for the reference 353 instrument (spectrophotometer) and the digital camera were examined to ensure the best 354 possible correlation between both devices. An improvement was obtained by quantizing 355 the camera RGB values to 10-bits relative to those recorded in 8-bits. Likewise, better 356 results were achieved with the specular component included (SCI) mode than with the 357 specular component excluded (SCE) mode in the reference instrument 358 (spectrophotometer).

359

The resulting calibration was successfully applied to six commercial varieties of granite, and the differences between data obtained with the reference instrument and with the camera calibrated as colorimeter were no higher than 6 CIELAB units.

363

This method, which enables RGB data to be expressed as device independent L*a*b* data, without introducing a noticeable amount of error, is sufficiently adaptable to be transposed to any computer vision system that can produce consistent RGB source data. The method can be used in many industrial applications using textured colored

materials and products. Apart from the fact that contact is not required for the color measurement, the other main advantage is the flexibility afforded by the choice of the size of the area to be characterized, which can range from small areas (347×375 pixel size μ m2) to areas as large as allowed by the lens size.

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583 system, (b) PL-A782 CMOS digital camera.

- 584 Figure 2. Screenshot of the camera settings conditions, showing the exposure time,
- 585 saturation, white balance and image format file (TIFF).
- Figure 3. Camera-light configuration. (a) Light sources aligned 45° with the camera's 586
- optical axis. Control for lighting or illuminance level was performed with a radiometer. 587
- (b) Results achieved with the radiometer (in millilux, 10^{-3} lx) appear across the table. 588
- White numbers indicate the area where the percentage of light level did not vary by 589
- 590 more than 3% and because of that, the images were taken inside that area.
- 591 **Figure 4.** Total color difference (ΔE^*_{ab}) between data obtained with spectrophotometer
- 592 and camera for the six commercial varieties of granite. Five specimens (represented by
- 593 different bars) were measured for each variety of granite.





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Acceleration



Granite specimen 1 Granite specimen 2 Granite specimen 3 Granite specimen 4 Granite specimen 5

Table 1. Mineralogical and petrographic features of the types of granite under study.

Granite name	Location of Quarry	Macroscopic Aspect / Classification and Geochemistry	Textural Characteristics	Mineral Composition
Aldán	Area of Morrazo (Pontevedra, Spain)	Yellow-white, medium-, coarse- grained / Micaceous calcalkaline granite	Granoblastic heterogranular of coarse grain	Quartz (35%), Feldespar-K (21%), Plagioclases (23%), Biotite (10%), Moscovite (10%)
Blanco Cristal	Cadalso de los Vidrios pluton (Madrid, Spain)	White, medium- grained / Biotite adamellitic granite	Heterogranular- panallatriomorphic of medium grain	Quartz (26%), Feldespar-K (29%), Plagioclases (27.5%), Biotite (9%), Moscovite (2%), Clorite (4.5%)
Grissal	Rivadavia pluton (Ourense, Spain)	Grey coarse-grained / Alkaline granite	Porphyritic- panallatriomorphic of coarse grain	Quartz (30.5%), Feldespar-K (34.5%), Plagioclases (17.5%), Biotite (0.6%), Moscovite (0.5%), Clorite (3.5%)
Monte Enxa	Area of Barbanza (A Coruña, Spain)	White, medium-, coarse-grained / Two mica granite	Heterogranular- allatriomorphic of medium-, coarse- grain	Quartz (45%), Feldespar-K (18%), Plagioclases (12%), Biotite (7%), Moscovite (17%)
Rosa Porriño	Porriño pluton (Pontevedra, Spain)	Pinkish, coarse- grained granite / Biotite adamellitic	Porphyritic- panallatriomorphic of coarse grain	Quartz (30%), Feldespar-K (33%),

		granite		Plagioclases (21%), Biotite (9%), Clorite (3.5%)
Silvestre	Area of Vigo (Pontevedra, Spain)	White medium- grained with some ochre spots due to biotite weathering / Two mica adamellitic granite	Equigranular- panallatriomorphic of medium grain	Quartz (29%), Feldespar-K (26%), Plagioclases (24%), Biotite (8%), Moscovite (8%), Clorite (3.5%)

.scribed in Petrographic characteristics and mineral composition described in [31, 55, 56]. 609

612 Table 2. Average, maximum and minimum total color differences between the 613 measured and the estimated CIELAB color stimuli, of the 212 color patches from the 614 glossy and matte Munsell collection.

- 615
- 616

		S	CI	SC	CE
		8-bits	10-bits	8-bits	10-bits
	Average	2.0	1.9	1.5	1.7
$\Delta E^*{}_{ab}$	Maximum	8.7	6.9	9.9	10.5
	Minimum	0.1	0.2	0.2	0.1
	Average	1.2	1.2	1.1	1.2
$\Delta E_{94}(2:1:1)$	Maximum	4.4	3.9	5.0	5.4
	Minimum	0.1	0.1	0.2	0.1
	Average	1.8	1.8	1.4	1.5
$\Delta E_{94}(1:1:1)$	Maximum	8.7	6.9	9.9	10.5
	Minimum	0.1	0.2	0.2	0.1
	Average	1.7	1.7	1.4	1.5
$\Delta E_{00}(2:1:1)$	Maximum	3.8	4.1	4.5	4.6
	Minimum	0.1	0.1	0.1	0.1
	Average	1.7	1.7	1.4	1.5
$\Delta E_{00}(1:1:1)$	Maximum	6.5	5.8	6.3	6.7
	Minimum	0.1	0.2	0.2	0.1
	Average	1.1	1.1	1.1	1.3
CMC (2:1)	Maximum	3.9	3.7	8.0	8.4
	Minimum	0.1	0.1	0.1	0.1
	Average	1.5	1.5	1.4	1.5
CMC (1:1)	Maximum	7.8	6.3	15.8	16.4
	Minimum	0.1	0.2	0.2	0.1

617 In order to select the optimal camera conditions, the lowest maximal value of the total color differences

618 was used as a selection criterion. The optimal camera conditions, according to this criterion, for 8- and

619 10-bit data, are highlighted in bold in the table for each color difference formulae calculated using SCI or SCE data.

620

- 622 **Table 3.** Absolute and relative discrepancies between the spectrophotometer and the
- 623 digital camera in the measurement of ΔE^*_{ab} total color difference.
- 624
- 625

Average ± SD Maximum Minimum 626	1.32 ± 1.06 6.41 0.00	0.06 ± 0.08 0.84 0.00
626	6.41 0.00	0.84 0.00
626		
626		
	P MA	
C C		
G		

Table 4. Summary table of precision and tolerance (in CIELAB units) of the628 instrumental devices used.

050	1	
	$n\Delta E_{ab}^{*}$ (Precision)	Instrumental tolerance
Spectrophotometer	0.01	0.1
Digital camera	0.24	2.4
631	·	
632		
	· ·	
	*	
Y		

633 **Highlights:**

634

635 • We develop the fine-tuning of a method for the remote color measurement of granite. 636

It is reported the description of a affordable methodology with digital camera. 637 •

- We estimate the effect of uncertainty on the measurement result. 638 •
- 639 Choice combination of camera and spectrophotomer minimizes uncertainty of •
- 640 measurement.

- The calibrated camera was successfully used on granite stones. 641 •
- 642
- 643
- 644