The Utilization of Pearson’s Method to Analyze Piezoresistive Effect in Self-Sensing Cement Composite with Graphite

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Structural health monitoring (SHM) techniques aim to detect and prevent failures in constructions, although their use may require many sensors, which makes this technique expensive and laborious. In this sense, the use of self-sensing cementitious composites based on the piezoresistivity effect could be a solution to some monitoring problems. Thus, the evaluation of the piezoresistive effect is commonly performed by analyzing the linearity between mechanical forces and the variation of electrical resistivity, through the coefficient of determination (R²). However, this work has been used to perform the analysis through Pearson’s correlation in samples of self-sensing cementitious composites with graphite addition. The results obtained have shown that Pearson’s correlation has the potential to be used for the evaluation of the correlation between electrical resistivity and mechanical forces to verify the piezoresistive effect in the cases studied.

Keywords: Self-sensing composite, piezoresistivity, cement, graphite, Pearson’s correlation.

1. Introduction

Structural health monitoring (SHM) techniques can provide increasing service life and construction safety, from early detection of strain, failures and damages1-2. However, continuous and wide-ranging monitoring may require high-cost investments due to the need for numerous sensors throughout the structure. So, using materials to be able to self-monitoring can be a solution to some problems with the use of sensors3-5. These materials can be created from a variety of matrices, such as polymers, asphalts, and cements6-9. The built-in sensors can make heterogeneity in the structure, which may cause some damage to the structure. While self-sensing composites make homogeneity to the structure2.

For the monitoring without using coupled or embedded sensors, the concrete structures need self-sensing cementitious composites in their manufacture to be able to detect changes in mechanical stress. Several electrical effects are the object of study for the application of self-sensing in materials, such as piezoresistivity. The piezoresistive effect allows the detection of mechanical strain and stress changes, as a result of the monitored electrical resistance change10-11. Further on the piezoresistivity, piezoelectric effect, capacitance and impedance are also used in self-sensing materials and sensor’s manufacture and mechanism12-16. Also used in sensors and sensing platforms are nanomaterials, which are another large field of research that is growing17-20.

Regular concrete and mortar exhibit a few piezoresistive responses when under loads, that is not enough to consider as self-sensing materials21. At least up to a certain point, piezoresistive composites offer a certain electrical conductivity for concretes and mortars. This conductivity cannot be too low to be considered the composite almost insulating, or too high to the offer almost no electrical resistivity, as this would impede the measurement of the piezoresistive effect22. In other words, for the composite to be able to self-monitoring, a minimum conductive filler needs to be used, but not without prejudice to the highest possible electrical conductivity. This limit is called the “percolation threshold” as shown in Figure 123.

As shown in Figure 2, as the self-sensing composite is compressed, the electrical resistivity decreases, as well as it is tensioned, it increases. In other words, this effect is what characterizes piezoresistivity, as a relationship between mechanical forces and the change in resistivity of the material. This change happens because conductive fillers move closer together or further apart, with compression or tension, thus altering the conductive paths. The change occurs linearly in the elastic regime of the composite, that is, up to a certain point in this relationship, as shown in Figure 324, and the reversible strain happens in the elastic regime. As cracks appear after the elastic regime, the continuous paths are interrupted and the resistivity increases again2.

However, challenges must be overcome so that it is possible to use self-sensing composite to monitoring real structures, for example, to mitigate and reduce the influence of external environmental issues, which affect the performance of self-sensing25. In this way, it could be useful to use Pearson’s correlation analysis (r) to see how the piezoresistive effect works in a cementitious composite.

Pearson’s correlation analysis (r) makes it possible to quantify the degree of association between two linear variables in a sample. This association has been described as a level of assessment of intensity and direction between the two compared variables. The correlation is great when the
The coefficient is 1 (one) or -1 (negative one) and unsatisfactory when it approaches zero. This method of correlation, which today has the name of its author, was developed in 1985, by Karl Pearson, and is obtained by Equation 1, where “x” and “y” are the variables, “$x_\bar{}$” and “$y_\bar{}$” the means, and “n” the number of variable pairs:

$$r = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^{n} (x_i - \bar{x})^2 \sum_{i=1}^{n} (y_i - \bar{y})^2}}$$

(1)

There are some limitations to Pearson’s method, such as the need for analyzed data to have a bivariate normal distribution and a linear relationship between them. Although the samples analyzed are significantly large in relation to the population, no more than two decimal places of the coefficient should be considered if the sample size <500, not even more than one decimal place, if the value of the coefficient is very small. For three decimal places, should be considered more than 100,000 the sample size.

Researchers from different areas of science usually give names to the levels of Pearson’s coefficient intervals used to rank their results, such as “moderate correlation” between 0.6 and 0.7. However, like the classification for this range,
it is also called otherwise by other researchers who consider it strong and not moderate. That is to say, the interpretation of the coefficients varies substantially between the different areas of scientific research, and thus the use of classification nomenclature should be avoided, given the lack of absolute rules for interpreting the correlation between the variables.

Even though researchers have been working to analyze self-sensing behavior in cement composites, they have used only the coefficient of determination ($R^2$) or graphic comparisons. However, this work does not aim to evaluate the mechanical properties of the cementitious composite but to use the Pearson’s correlation to analyze the piezoresistive effect of self-sensing cement, which is why only graphite was used in the composite. Carbon-based materials have good electrical conductivity, so they are the materials most commonly used nowadays to make self-sensing cement composites. Graphite is used, in a wide range of composites and materials, including smart structures, electromagnetic shielding, building, electric power, road engineering, aircraft parts, and thermal management. Graphite powder was chosen because it is a low-cost material when compared with nanomaterials.

2. Materials and Methods

2.1. Materials

Following the literature, four mortar specimens were prepared for the piezoresistive test, in cubic shape samples with a 40 mm edge, dimensions recommended in NBR 16868-2 for compression test. The mortar mix rate was made in the following proportions of cement, sand, and graphite: T1C=1:4:0; T2C=1:4:0.125; T3C=1:4:0.250; and T4C=1:4:0.375, in proportions to the cement mass, and these graphite proportions are within the range (0-0.4) used by others. The water/cement ratio (w/c) of 0.50 used in the T1C mix could not be maintained in the other mixes, in which the following w/c ratios were used: T2C=0.75; T3C=1.05; and T4C=1.40, to get the consistency for molding the samples.

The specifications of the materials, according to the manufacturers, are: Portland cement, marketed by the company Votorantim Cimentos – Classification CP-II-F-32, 75% - 89% by mass of clinker + gypsum; 11% - 25% carbonate material; The quartz sand has a Fineness Modulus 2,03 according to NBR 7211. The powdered graphite, marketed by Wonder - Carbon (Loss to Fire) >72%; Ash a maximum of 28%; maximum moisture 0.5%. In the Figure 4, the granulometric of sand and graphite are shown.

During the molding of the sample’s specimens, 4 electrodes (aluminum plates) with a size of 35 x 20 x 0.1 mm were inserted, spaced 10 mm between each other and with 25 mm embedded in the mortar, as shown in Figure 5.

Although graphite improves the electrical conductivity of the cementitious composite, it generally causes a reduction in the composite’s compressive strength. Much research reports the use of graphite with other fillers, such as polymer fibers or carbon, to improve the mechanical strength.

2.2. Methods

The samples were tested by compression in a manual press, for three repetitions of each sample, with a maximum load of nearly 2 kN, since this load would not cause rupture of the samples, because it is a non-destructive test. The voltage (U) changes and force (compression) applied were recorded synchronously at a sampling rate of 10 Hz, using the scheme in Figure 6, which had a data acquisition.
system (DAQ) model 8000-8-SM (Micro-Measurements), a load cell, and a laptop.

The electrical resistance was obtained from Ohm’s law and, through the electrical circuit in Figure 7, it was possible to determine the electrical resistance ($R_s$) of the composite by Equation 2 (adapted from\textsuperscript{53}). A DAQ system was used to obtain the voltages in the sample ($U_s$) and in the circuit supply ($U_{in}$), with the use of a reference resistor ($R_{ref}$) 1000 $\Omega$\textsuperscript{34,54,55}. The data from voltages in the sample ($U_s$) corresponds to the wires green and yellow in Figure 6.

$$R_s = R_{ref} \frac{U_s}{U_{in} - U_s}$$  \hspace{1cm} (2)

The electrical resistivity ($\rho$) was calculated by Equation 3\textsuperscript{35,56}, where the distance (L) between the central electrodes and their contact area (A) with the composite are known values, while the fractional change in resistivity (FCR) was calculated using Equation 4\textsuperscript{2}.

$$\rho = \frac{RA}{L}$$  \hspace{1cm} (3)

$$\text{FCR} = \frac{\Delta \rho}{\rho}$$  \hspace{1cm} (4)

\textbf{Figure 5.} Cubic specimens with electrodes

\textbf{Figure 6.} Data acquisition scheme of the experiment.

\textbf{Figure 7.} Electrical circuit for measuring electrical resistance (adapted from\textsuperscript{53}).
\[
\rho = \frac{R_{s} \cdot A}{L}
\]  
(3)

FCR\% = \frac{\Delta \rho}{\rho} \times 100

(4)

Among the ways indicated in the literature, the sensitivity of the piezoresistive composite was determined from the fractional change of resistivity per unit stress (σ), using Equation 5.

\[
S = \frac{FCR\%}{\sigma}
\]  
(5)

In other words, the sensitivity is the relationship between mechanical and electrical effects for the evaluation of piezoresistivity, when the resistivity can be affected by stress amplitude.

Pearson’s correlation (r), coefficient of determination (R²), and other statistical parameters were determined using OriginLab Pro software, version 2021b.

3. Results and Discussion

According to literature, the applied force (compression) values and the fractional change in resistivity (FCR) obtained in the experiment are compared in the graphs of Figure 8, for each of the samples and repetitions, and a visual correlation analysis would be very subjective for qualifying the results, which is why a preliminary sensitivity analysis of the piezoresistive effect was performed.

The sensitivity analysis was made from the relationship between the peak values of the compression force and the FCR, using Equation 5, as shown in Table 1.

![Figure 8](image_url)

**Figure 8.** Compressive force values and the FCR in each mix and repetition test for T1C (a, b, c), T2C (d, e, f), T3C (g, h, i) and T4C (j, k, l).
Through the sensitivity relationship, it was possible to verify that the T2C and T3C samples had better results compared to the other samples. This is even clearer when presented graphically in Figure 9, with shows the mean values and their standard deviations (SD).

Even though the sensitivity relationship analysis shows a better response of the T2C and T3C samples, it also makes evident the accentuated variations between repetitions. This is especially true for the T3C sample, with has a high standard deviation compared to the other samples.

In a preliminary linearity analysis, the graphs of Figure 10 were plotted, which visually show better results for the T2C and T3C samples, but also possible to the subjectivity of their interpretation.

The Pearson’s correlation values (r), based on the data in the graphs in Figure 8, and the coefficients of determination (R²) based on the data in the graphs in Figure 10, are shown in the graph in Figure 11. Although both coefficients (r and R²) have been plotted on the same graph, the direct comparison between them cannot be adequate because the first express values from -1 to 1, and the second from 0 (zero) to 1.

The means and standard deviations of each coefficient (r and R²) for the samples tested are shown in the graphs in Figure 12, where again the best results for samples T2C and T3C are shown.

The R² values obtained for the T2C and T3C samples are close to those found in the literature available⁴⁶, while the low correlation of the T1C sample is justified by the

### Table 1. Sensitivity of samples based on FCR/Stress relationship.

| Sample | Repetition | Peak Value | Sensitivity (Equation 5) |
|--------|------------|------------|-------------------------|
|        |            | Stress* (MPa) | FCR (%) |
| T1C    | 1st        | -1.30     | -19.86 | 15.33 |
|        | 2nd        | -1.33     | -21.97 | 16.46 |
|        | 3rd        | -1.27     | -25.62 | 20.12 |
| T2C    | 1st        | -1.27     | -140.79 | 111.20 |
|        | 2nd        | -1.31     | -118.99 | 90.84 |
|        | 3rd        | -1.26     | -114.24 | 90.92 |
| T3C    | 1st        | -1.29     | -186.47 | 144.53 |
|        | 2nd        | -1.32     | -140.54 | 106.49 |
|        | 3rd        | -1.28     | -129.67 | 101.56 |
| T4C    | 1st        | -1.26     | -38.36 | 30.39 |
|        | 2nd        | -1.37     | -36.65 | 26.81 |
|        | 3rd        | -1.32     | -33.01 | 25.01 |

*Compression

Figure 9. Mean sensitivity and standard deviations (SD) of samples.
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Figure 10. Graphical relationship between FCR and compression force, with linear regression (line in red) in each mix and repetition test for T1C (a, b, c), T2C (d, e, f), T3C (g, h, i) and T4C (j, k, l).

Figure 11. Pearson (r) and determination ($R^2$) coefficients per sample.

absence of graphite and that of the T4C sample by the excess of graphite, which is also valid to justify the results obtained with the Pearson analysis.

For the final visualization of the results, each sample was ranked for each of the analyzes performed according to the table in Figure 13a, and each sample was spatialized...
according to the values of each parameter obtained, as shown in the graph in Figure 13b.

With the ranking shown (Figure 13a), the proportionality between the Pearson and determination coefficients becomes even more evident. Although they have different purposes, there is a quadratic relationship between them. This relationship is shown in Figure 14, which exposes the analytical values of Pearson starting in 0.01 and moving forward in centesimal fractions to 1.00. Then, the quadratic value of Pearson that corresponds to $R^2$, for each centesimal fraction, is plotted (Pearson versus $R^2$). The experimental values of Pearson and $R^2$ of tested samples are plotted in the same graph (Figure 14), and are the same as the quadratic relationship analytical.

Figure 12. Mean and standard deviation per sample for each type of coefficient ($r$ and $R^2$).

Figure 13. Ranking table of the results per sample for each analysis parameter (a); spatialization of the results of each parameter in each sample (b).

Figure 14. Analytical values (continuous line) and tested samples’ values (dashed line) of Pearson versus $R^2$ and the difference between them; maximum difference (red line).
On the other hand, the graphic spatialization (Figure 13b) visually supported the best results obtained for T2C and T3C.

4. Conclusions

Based on the results obtained, it can be considered that for the conditions and cases analyzed, the Pearson’s coefficient has the potential to be used in the analysis of the correlation between changes in electrical resistivity and mechanical forces (or stress and strain) to verify the piezoresistive effect in self-sensing cementitious composites, although more studies are required to extrapolate these considerations to other cases and conditions.

We also saw that the higher the sensitivity, the standard deviation was higher too, which indicates that the repeatability was affected. But, in contrast to this, the correlation from Pearson (r) and determination (R²) coefficients did not show that the repeatability was affected, because the higher the coefficients, the smaller the standard deviation.

It is also possible to pronounce that:
1. The evaluation of piezoresistive effect should not be limited only to graphical analysis of the values of electrical and mechanical parameters. Although it is important, this may lead to subjective or even mistaken interpretations, when not compared to numerical analysis;
2. The correlation analysis by Pearson must be limited to the linear behavior range of the study otherwise, other methods must be used;
3. The use of the Pearson correlation method should not exempt analysis by the coefficient of determination (R²) method, because they have different purposes. But, the first does not have the influence of a quadratic relationship in the results. The quadratic relationship makes the difference between the coefficients be minimum when approaching 0 (zero) or 1 (one) and maximum, when approaching 0.5 (half), as shown in Figure 14.

Determination of sensitivity, a relationship between FCR and stress, is also an important way of evaluating the self-sensing composite, although it has been observed (Figure 13) that few the sensitivity of the T4C-1 sample did not directly impact the bivariate correlation, as this sample obtained the 3rd position in the ranking of the Pearson and determination (R²) coefficients.

Lastly, both graphic and numerical evaluations (sensitivity and correlation coefficient) must be present in the analysis of the piezoresistive effect in self-sensing cementitious composites because, together, they allow a better interpretation of this effect.

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