#### LEMBAR

## HASIL PENILAIAN SEJAWAT SEBIDANG ATAU PEER REVIEW

#### KARYA ILMIAH: JURNAL ILMIAH

Judul Karya Ilmiah (Paper)	: The Prediction of Stiffness Reduction Non-Lin Element Method (FEM) and Artificial Neural	ear Phase in Bamboo Reinforced Concrete Beam Using the Finite Networks (ANNs)			
Jumlah Penulis	: 1 Orang (1. Muhtar, 2, 3, 4	)			
Status Pengusul	: Penulis pertama / <del>penulis ke</del> / penulis korespondensi**				
Identitians Jurnal/Prosiding	: a. Nama Jurnal/ Prosiding	:Forests			
	b. ISSN/ISBN	: 1999-4907			
	c. Tahun Terbit, (tempat pelaksanaan jika prosiding)	: 2020, 11(12), 1313			
	d. Penerbit/Organiser	: MDPI Multidisciplinary Digital Publishing Institute			
	e. Alamat repository PT/ Web	: http://repository.unmuhjember.ac.id/8741/			
	f. Terindek di (jika ada)	: SCOPUS (Q1 SJR 0.68)			
Kategori Publikasi Artikel Ilmiah	: V Jurnal Internasonal Bereputasi				
(Beri √ pada kategori yang tepat)	Jurnal Internasional				
	Jurnal Nasional Terakreditasi				
	Jurnal Nasioanl Tidak Terakreditasi				
	Jurnal Nasional Terindeks DOAJ dll				
	Prosidng Forum Ilmiah Internasional				
	Prosiding Forum Ilmiah Nasioanl				

## Hasil Penialain Peer Review:

	N	Nilai Maksimal Jurnal Ilmiah (Isikan di kolom yang sesuai)				
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Kelengkapan unsur dan kualitas penerbit (30%)	12					12
Total 100%	40					39
Kontribusi Pengusul (Penulis Pertama/ Anggota Utama)						100 % x 39 = 39
Komentar Peer Review	<ol> <li>Tentang Keler</li> <li>Tentang Ruar</li> <li>Kecukupan, I</li> <li>Kelengkapan</li> <li>Indikasi Plag</li> <li>Kesesuian bi</li> </ol>	ngkapan dan kese ng Lingkup dan ke Kemutakhiran dati unsur kualitas pe jasi: + dole dang ilmu: Lin	suain unsur: Sa edalaman pembah a serta metodelogi nerbit: Junna ava tot wer lan	ugat lingtop Lingtop 1 asan: Fembrah BRC. An i: Sangat less boldsfanden L. Q. D. Ku at Similar Sangat &	that little tent about mendalam alua la FEM o saliap Bori Reti alitos ponellit de rety 17 90 enai Dengan	tang beton bertulang - am n. Novelty berupa nibal to dan ANN untul Validaci Essenti meral Qualifiet ata lulu dan ACI, ditulis of b Langat baile, kalenghow en bidang ilmu penulis

Malang, 5 Moret 2021 Reviewer 1

(Prof. Dr. Ir. Sri Murni Dewi, MS.) NPK/NIP. 195112111981032001 Unit kerja: Teknik Sipil UB Malang Jafung: Guru Besar Bidang Ilmu: Teknik Sipil

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(Prof. Dr. Ir. Rudy Sornoko, M.Eng.Sc.) NPK/NIP. 194909111984031001 Unit kerja: Teknik Mesin UB Malang Jafung: Guru Besar Bidang Ilmu: Teknik Mesin

Editor Decision

Decision Accept in current form Decision Date 25 November 2020

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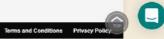
#### Previously Published Papers

Muhtar; Gunasti, A.; Suhardi; Nursaid; Irawati; Dewi, I.C.; Dasuki, M.; Arlyani, S.; Fitriana; Mahmudi, I.; Abadi, T.; Rahman, M.; Hidayatuliah, S.; Nilogiri, A.; Desta Galuh, S.; Eko Wardoyo, A.; Budi Hamduwibawa, R. The Prediction of Stiffness of Bamboo-Reinforced Concrete Beams Using Experiment Data and Artificial Neural Networks (ANNs). Cyprates 2020, 10, 757. doi: 10.3390/cypst10090791

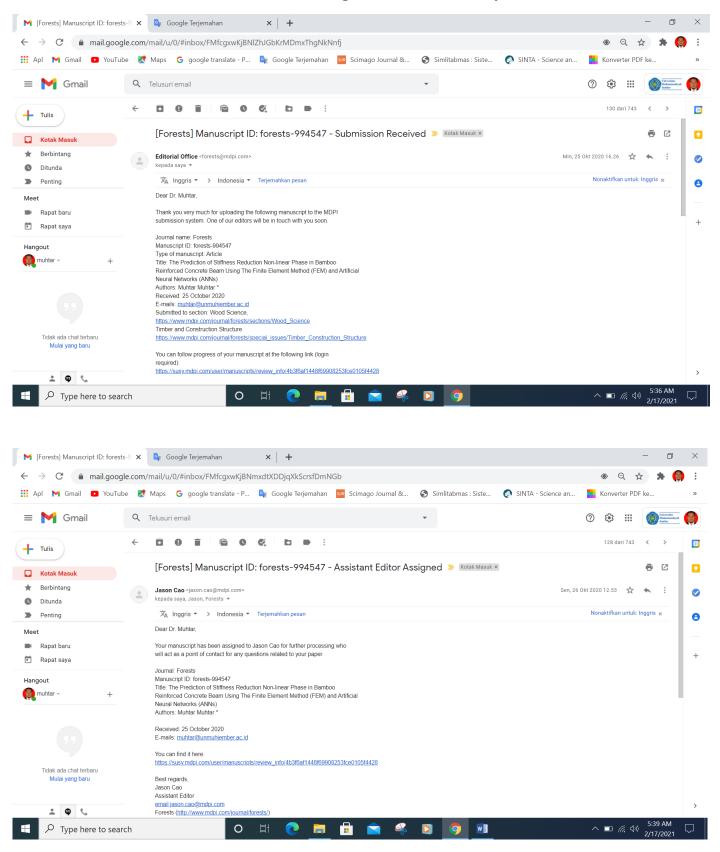
Muhtar. Precast Bridges of Bamboo Reinforced Concrete in Disadvantaged Village Areas in Indonesia, Appl. Sci. 2020, 10, 7158. doi: 10.3390/app10207158

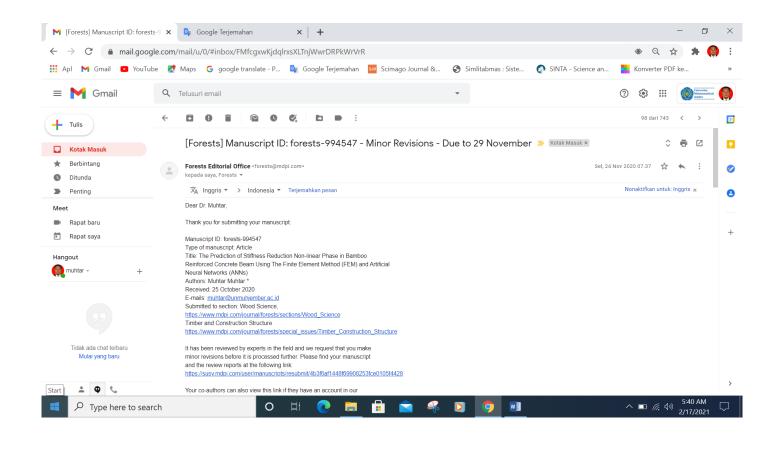
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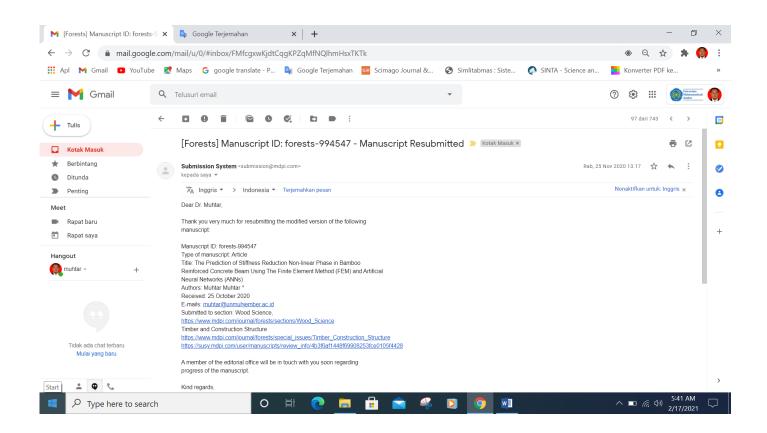
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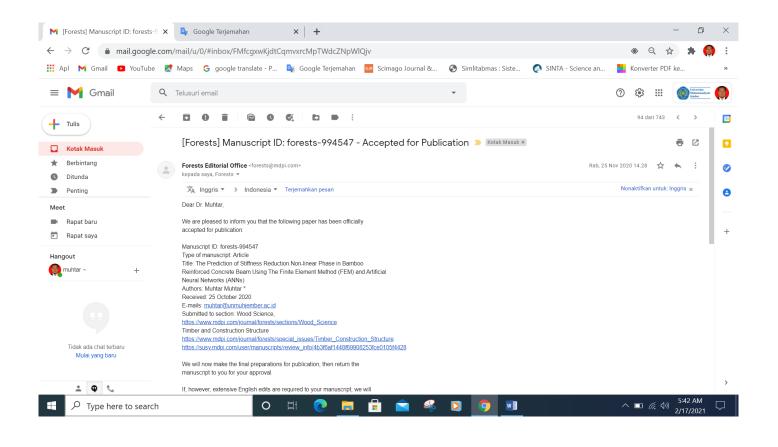


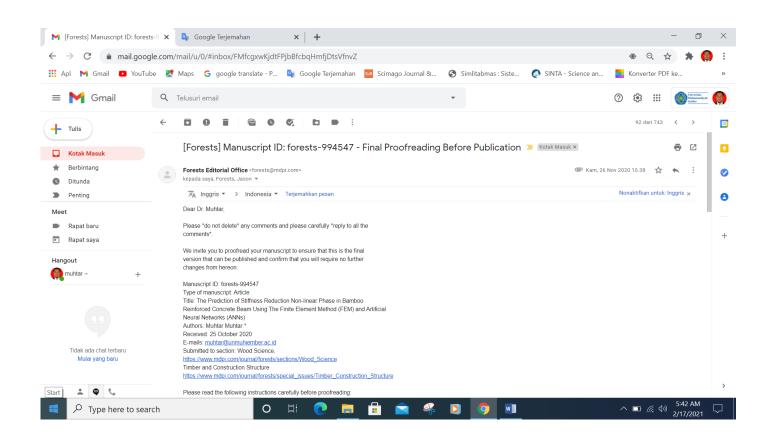
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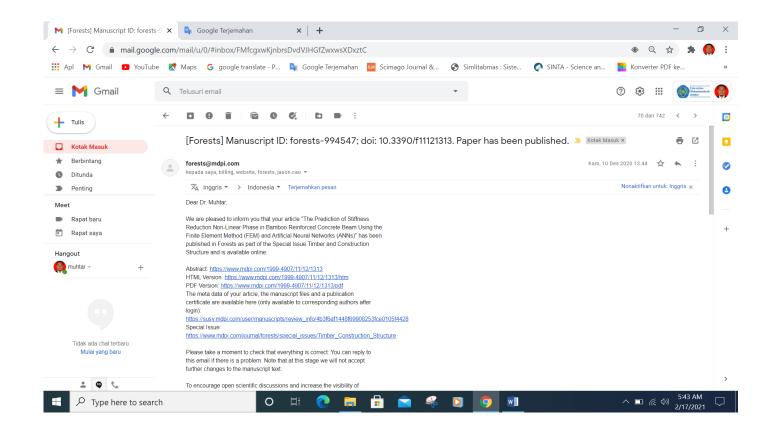


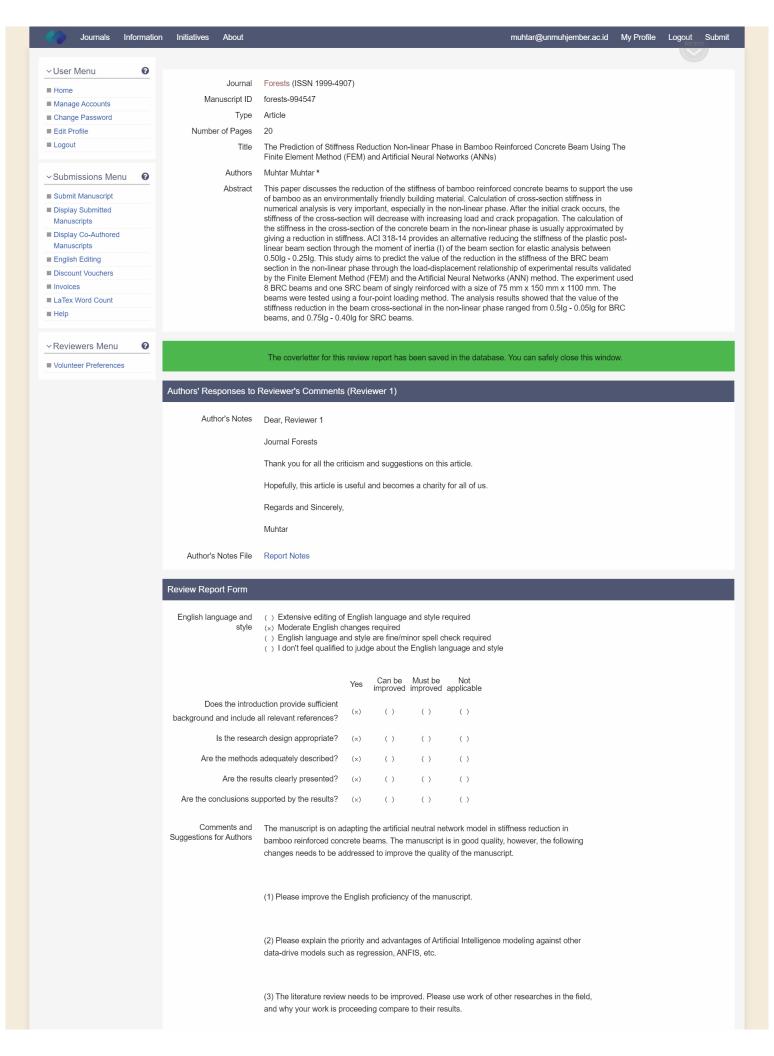












(4) Please explain the content of your artificial neutral network in more detail. You may want to refer to the following two manuscripts:

(A)Khademi, F., Akbari, M., Jamal, S. M., & Nikoo, M. (2017). Multiple linear regression, artificial neural network, and fuzzy logic prediction of 28 days compressive strength of concrete. *Frontiers of Structural and Civil Engineering*, 11(1), 90-99.

(B)Li, X., Khademi, F., Liu, Y., Akbari, M., Wang, C., Bond, P. L., ... & Jiang, G. (2019). Evaluation of data-driven models for predicting the service life of concrete sewer pipes subjected to corrosion. *Journal of environmental management*, 234, 431-439.

(5) Please explain your conclusion in more details.

The overall quality of the manuscript is in good condition, and my suggestion is minor revision. After these revisions are made, this manuscript has a high chance of getting published in this journal.

Submission Date 25 October 2020

Date of this review 23 Nov 2020 20:39:56

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# Response to Reviewer 1 Comments

**Comments:** The manuscript is on adapting the artificial neutral network model in stiffness reduction in bamboo reinforced concrete beams. The manuscript is in good quality, however, the following changes needs to be addressed to improve the quality of the manuscript.

Thank you very much to the reviewers who have reviewed for the sake of perfection this paper.

Point 1: (1) Please improve the English proficiency of the manuscript

#### **English Editing uses "Language Editing" from MDPI**

**Point 2:** (2) Please explain the priority and advantages of Artificial Intelligence modeling against other data-drive models such as regression, ANFIS, etc.

The ANN method is currently very popular with researchers in predicting and evaluating the behavior of structures in the field of civil engineering, this is because the ANN method has an advantage in the nonlinear correlation between the input variables presented is better. Khademi et al. (2017) [40] predicts the compressive strength of concrete at 28 days of age by considering the experimental results, three different models of multiple linear regression (MLR), artificial neural networks (ANN), and adaptive neuro-fuzzy inference system (ANFIS). The results of his research concluded that the ANN and ANFIS models can predict the 28-day concrete compressive strength more accurately and the ANN model can perform better than the ANFIS model in terms of R². The ANN and ANFIS models are preferred because the nonlinear correlation between the input variables presented is better. The ANN and ANFIS models have higher accuracy requirements than the multiple linear regression (MLR) model. The accuracy of the prediction is very much dependent on the number of input variables, the greater the number of input parameters, the more accurate the results of the predictor model will be.

Xuan Li et al. (2019) [41] predicts the service life of corroded concrete sewer pipes using three data-driven models, namely multiple linear regression (MLR), artificial neural networks (ANN), and adaptive neuro-fuzzy inference system (ANFIS). The one conclusion suggests that the ANN and ANFIS models perform better than the MLR models for corrosion prediction, with or without considering the interactions between environmental factors.

already written on line 259 - 276

**Point 3:** (3) The literature review needs to be improved. Please use work of other researches in the field, and why your work is proceeding compare to their results.

Figure 19 shows that the artificial neural networks (ANN) model has a higher  $R^2$  value when compared to the  $R^2$  value of the multiple linear regression model (MLR). ANN analysis has better predictive accuracy. This is the same as the conclusion of 2 researchers, namely Khademi et al. (2017) [40], which concluded that The ANN model has higher accuracy than the multiple linear regression (MLR) model, and Xuan Li et al. (2019) [41] concluded that the ANN model performs better than the MLR models with or without considering the interactions between factors. The accuracy of the

prediction is very much dependent on the number of input variables, the greater the number of input parameters, the more accurate the results of the predicted model.

Already written on line 392 - 399

- **Point 4:** Please explain the content of your artificial neutral network in more detail. You may want to refer to the following two manuscripts:
  - (A) Khademi, F., Akbari, M., Jamal, S. M., & Nikoo, M. (2017). Multiple linear regression, artificial neural network, and fuzzy logic prediction of 28 days compressive strength of concrete. Frontiers of Structural and Civil Engineering, 11(1), 90-99.
  - (B) Li, X., Khademi, F., Liu, Y., Akbari, M., Wang, C., Bond, P. L., ... & Jiang, G. (2019). Evaluation of data-driven models for predicting the service life of concrete sewer pipes subjected to corrosion. Journal of environmental management, 234, 431-439.

The ANN data is divided into three different subsets [40], namely: (1) Training, at this stage, the subset is trained and studied as occurs in the human brain, where the number of epochs is repeated until an acceptable model accuracy is obtained; (2) Validation, at this stage, the subset shows how well the model is trained, and estimates model properties such as misclassification, mean error for numerical predictors; and (3) Test, at this stage, the subset verifies the performance of the training subset built into the ANN model.

Already written on line 277 – 282

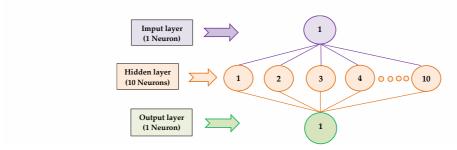


Figure 5. Schematic of ANN model architecture for BRC beam and SRC beam

The process of implementing input data in the ANN model architecture consists of (1) Input layer; consisting of 1 neuron, namely displacement data variable of experimental results; (2) Hidden layer, consisting of 10 neurons. At this stage, the input layer will forward the data to the hidden layer or the output layer through a set of weights. This weight is a link from each neuron to other neurons in the next layer which will help adjust the ANN structure to the given displacement data pattern using learning. In the learning process, the weights will be updated continuously until one of the numbers of iterations, errors, and processing time has been reached. This is done to adjust the ANN structure to the desired pattern based on certain problems that will be solved using ANN. Weight or what is known as the independent parameter. During the training process, the weights will be modified to improve the accuracy of the results; and (3) Output layer, consisting of 1 neuron which is the expected output target, error, and weight. Error is the error rate of the displacement data node of the process carried out, while weight is the weight of the displacement data node with a value ranging between -1 and 1. Then the displacement data resulting from the training process is processed into a graphic image of the load vs. displacement relationship.

Already written on line 285 – 299, and references from reviewers' suggestions are included in References 40 and 41 as below:

40. Khademi, F.; Akbari, M.; Mohammadmehdi, S.; Nikoo, M. Multiple linear regression, artificial neural network, and fuzzy logic prediction of 28 days compressive strength of concrete. *Frontiers of Structural and Civil Engineering*. 2017, 1190–99.

41. Li, X.; Liu, Y.; Akbari, M.; Wang, C.; Bond, P. L.; et al. Evaluation of data-driven models for predicting the service life of concrete sewer pipes subjected to corrosion. *Journal of Environmental Management*. 2019, 234, 431–439.

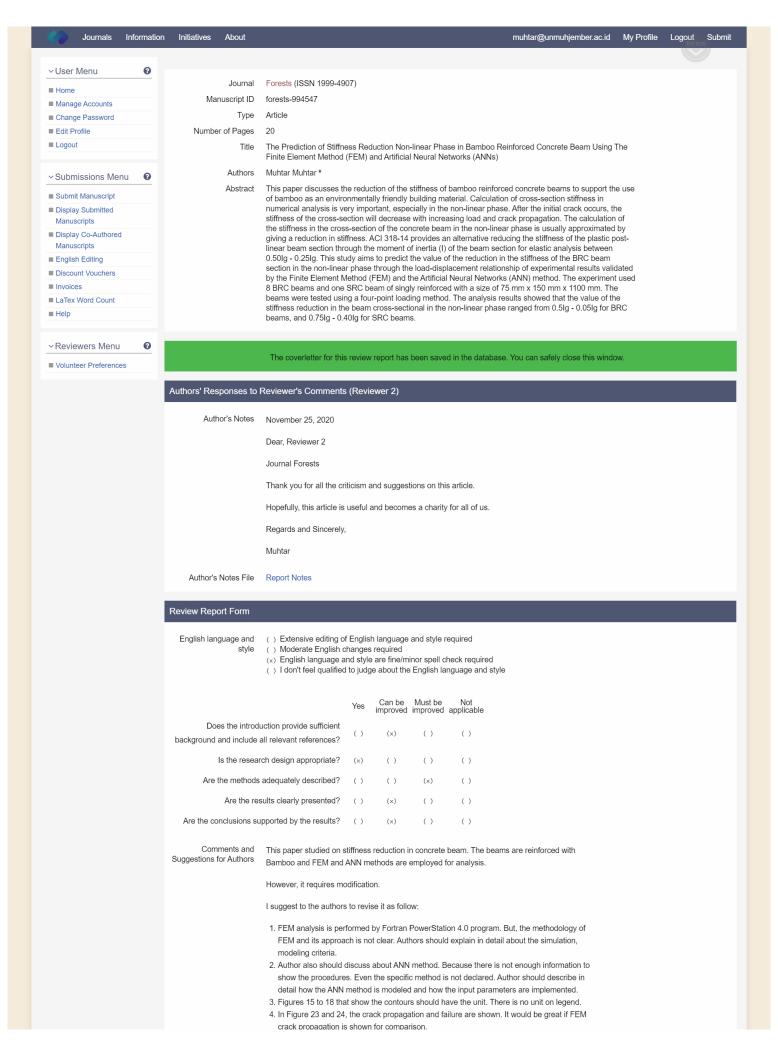
**Point 5:** (5) Please explain your conclusion in more details.

There are additional conclusions as follows:

The relationship pattern of load vs. displacement reflects the stiffness pattern of structural elements. The properties and characteristics of the material in the reinforcing concrete elements have a dominant influence on the relationship pattern of the load vs. displacement of reinforced concrete elements. Bamboo reinforced concrete beams (BRC) have a different load vs. displacement relationship pattern when compared to steel reinforced concrete beams (SRC). BRC beams have elastic properties and high resilience properties that can accept high impact loads without causing over stress at the elastic limit, even though displacement has occurred. While SRC beams have high stiffness and toughness so that SRC beams are not subject to excessive displacement or deformation at service load ranges or elastic conditions.

Results of the validation of the relationship pattern of the load vs. displacement of the BRC beams shows that the ANN model has a higher R<sup>2</sup> value when compared to the R<sup>2</sup> value of the MLR model. ANN analysis has a higher prediction accuracy. The accuracy of the prediction depends very much on the number of input variables, the greater the number of input parameters, the more accurate the prediction model results.

And already written on line 496 – 509



Submission Date 25 October 2020

Date of this review 15 Nov 2020 02:52:33

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# Response to Reviewer 2 Comments

**Point 1:** (1) FEM analysis is performed by Fortran PowerStation 4.0 program. But, the methodology of FEM and its approach is not clear. Authors should explain in detail about the simulation, modeling criteria.

The simulation and steps for preparing a FEM analysis with the Fortran PowerStation 4.0 program [32] are summarized as follows:

Step 1: Discretization of BRC and SRC beam planes with the discretization of triangular elements, the numbering of triangular elements, and the numbering of nodal points as shown in Figure 3 and Figure 4.

Step 2: Calculation and collection of geometry and material data, such as the modulus of elasticity of the material (E), Poisson's ratio (v), etc.

Step 3: Writing a programming language for triangular elements using the Fortran PowerStation 4.0 program according to the constitutive relationships and FEM modeling as shown in the following link: http://bit.ly/2F17w8F.

Step 4: Open the Fortran PowerStation 4.0 program. An example is shown at the following link: http://bit.ly/2MTh22j.

Step 5: Write programming language data (Step 3) in the Fortran PowerStation 4.0 program. Examples can be seen at the following link: http://bit.ly/2ZvZWMU.

Step 6: Input DATA.DAT of BRC beam and SRC beam in the Fortran PowerStation 4.0 program. Input data is displayed at the following link: <a href="http://bit.ly/351FPqU">http://bit.ly/351FPqU</a> and <a href="http://bit.ly/2MBqas9">http://bit.ly/2MBqas9</a>. An example of displaying input data is shown on the following link: <a href="http://bit.ly/2u2K2xR">http://bit.ly/2u2K2xR</a>.

Step 7: Analyze the program until there are no warnings and errors. If there are warnings and errors, check and correct program data and input data.

Step 8: Download stress data. The stress data are shown at the following link: <a href="http://bit.ly/2rDPeal">http://bit.ly/2rDPeal</a> for the stress of BRC beam, and <a href="http://bit.ly/2Q4lhc1">http://bit.ly/2Q4lhc1</a>. for the stress of SRC beam. An example of displaying stress data from the Fortran PowerStation 4.0 program is shown at the following link: <a href="http://bit.ly/2ZybLCd">http://bit.ly/2ZybLCd</a>.

Step 9: Download displacement data. An example of displaying data displacement from the Fortran PowerStation 4.0 program is shown on the following link: http://bit.ly/2Q7j2Wp.

Step 10: Enter stress and displacement data into the Surfer program to obtain contour image data of stress and displacement. Stress and displacement contour image data are shown in Figures 15-18.

Already written on line 216 – 242

**Point 2:** (2) Author also should discuss about ANN method. Because there is not enough information to show the procedures. Even the specific method is not declared. Author should describe in detail how the ANN method is modeled and how the input parameters are implemented.

The ANN data is divided into three different subsets [40], namely: (1) Training, at this stage, the subset is trained and studied as occurs in the human brain, where the number of epochs is repeated until an acceptable model accuracy is obtained; (2) Validation, at this stage, the subset shows how well the model is trained, and estimates model properties such as misclassification, mean error for numerical predictors; and (3) Test, at this stage, the subset verifies the performance of the training subset built into the ANN model.

The process of implementing input data in the ANN model architecture consists of (1) Input layer; consisting of 1 neuron, namely displacement data variable of experimental results; (2) Hidden layer, consisting of 10 neurons. At this stage, the input layer will forward the data to the hidden layer or the output layer through a set of weights. This weight is a link from each neuron to other neurons in the next layer which will help adjust the ANN structure to the given displacement data pattern using learning. In the learning process, the weights will be updated continuously until one of the numbers of iterations, errors, and processing time has been reached. This is done to adjust the ANN structure to the desired pattern based on certain problems that will be solved using ANN. Weight or what is known as the independent parameter. During the training process, the weights will be modified to improve the accuracy of the results; and (3) Output layer, consisting of 1 neuron which is the expected output target, error, and weight. Error is the error rate of the displacement data node of the process carried out, while weight is the weight of the displacement data node with a value ranging between -1 and 1. Then the displacement data resulting from the training process is processed into a graphic image of the load vs displacement relationship.

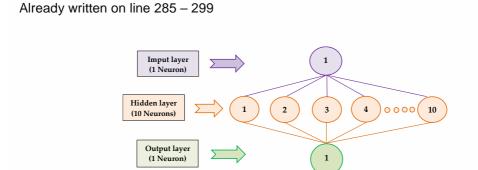
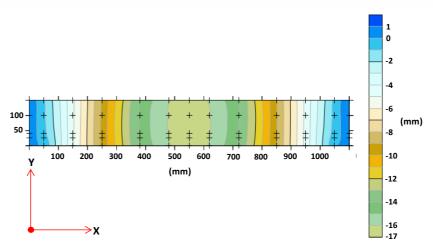


Figure 5. Schematic of ANN model architecture for BRC beam and SRC beam

**Point 3:** (3) Figures 15 to 18 that show the contours should have the unit. There is no unit on legend.

The addition of units in Figures 15-18 has been carried out as shown in the figure below.



**Figure 15.** The displacement contour of Y-direction of BRC beam

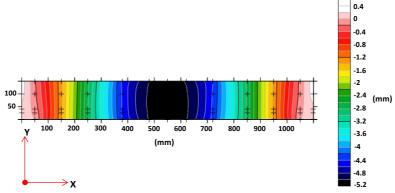


Figure 16. The displacement contour of Y-direction of SRC beam

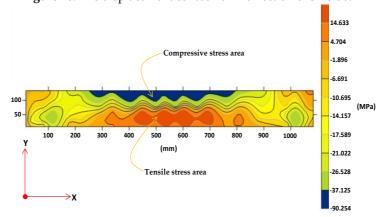


Figure 17. The stress contour of X-direction of BRC beam

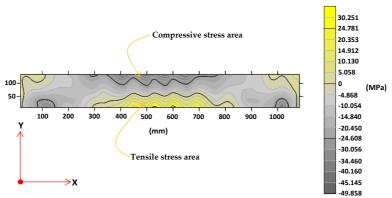


Figure 18. The stress contour of X-direction of SRC beam

The revised figure above has been included in the paper in Figures 15-18

**Point 4: (4)** In Figure 23 and 24, the crack propagation and failure are shown. It would be great if FEM crack propagation is shown for comparison.

The output of the FEM analysis using the Fortran PowerStation 4.0 program is stress and displacement so that the crack pattern of the experimental results can only be compared with the tensile stress zone that causes cracks in the beam, as shown in Figure 23-24 below.

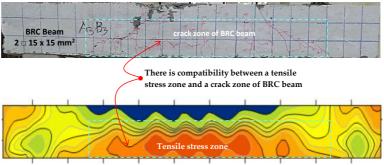


Figure 23. The crack pattern and tensile stress zone of BRC beam

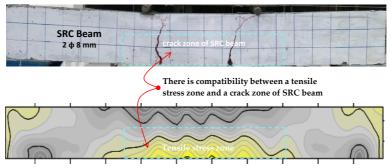


Figure 24. The crack pattern and tensile stress zone of SRC beam

The revised figure above has been included in the paper in Figures 23-24