

# Predicting Construction labour Productivity Using Optimal Artificial Neural Network, Multiple Linear Regression Models: Comparative Study

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# **AUTHORIZATION FORM**

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#### **DETECTION**

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#### LIST OF ABBREVIATION

AI Artificial Intelligence

ANN Artificial Neural Network
APE Absolute Percentage Error

APV Aldada Private Villa

bj Bias or Threshold for Node j

CAGR Compound Annual Growth Rate

CP Construction Productivity

d<sub>k</sub> Desired Output of Neuron K

DEMATEL Decision Making Trial & Evaluation Laboratory

EP Execution Position

EYN Experience Year Number

FFNN Feed Forward Neural Network

FI Frequency Index

F() Transfer or Activation Function

F Derivatives of Activation Function

GDP Gross Domestic Product

I<sub>j</sub> Activation Level of Node j

II Importance Index

LOGSIG Log-Sigmoid Transfer Function

MACE Mongolian Association of Civil Engineer

MAPE Mean Absolute Percentage Error

MLR Multiple Linear Regression

MP Material Position
NN Neural Network

NSL Number of Support Labor

PE Processing Element

R Correlation Coefficient

R<sup>2</sup> Coefficient of Determination

RBNN Radial Basis Neural Network

RII Relative Importance Index

SC Site Condition

SD System Dynamic

SI Severity Index

SPSS Statistical Package Society Science

SPV Sinokrot Private Villa

SR Social Relationship

SSE Sum Square Error

SVM Support Vector Machine

SZ Stone Size

TANSIG Hyperbolic Tangent Transfer Function

TRB Tebyeh Residential Building

W<sub>ij</sub> Connection Weight between Node i and j

W<sub>jk</sub>(n) Weight Correction

W<sub>jk</sub>(n-1) Previous Weight Correction

X<sub>i</sub> Inputs

Y<sub>j</sub> Output of Node j

η Learning Rate

α Momentum Term

δ Backpropagation Error

#### **Predicting Construction labour Productivity Using**

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#### ABSTRACT

Construction productivity can be considered as an element in project management; therefore, predicting the rate of construction productivity for labor was an important task. However, the development of the technology tools will enable the planner to best understand the process of estimation and predicting in different stages of construction projects.

The main aim of this research is to develop a novel mathematical model using Multiple Linear Regression (MLR) and Artificial Neural Network (ANN) to predict the construction productivity rates because mathematical models and mathematical equations used for finishing stone activity are characterized by uncertainty and lack validity and verification, and traditional methods fail to calculate the construction productivity due to their slowness and lack of accuracy.

Data was collected from three residential building projects in the Hashemite Kingdom of Jordan in the capital city of Amman from July 2017 to December 2017. The first project was Tebyeh Residential Building (TRB); the second project was Sinokrot Private Villa (SPV); and the third project was Aldada Private Villa (APV). The results demonstrated that (MLR) is a more powerful technique than (ANN) for construction productivity of finishing stone activity depending on validity through Mean Absolute Percentage Error (MAPE%)

and Average Accuracy (AA%), which were equal to 18.615% and 81.3846% respectively; ANN technique (MAPE%) was equal to 27.06% and (AA%) was equal to 72.94%. This result can be expressed when using multiple linear regression techniques instead of artificial neural networks in estimating and predicting construction productivity when the data of the variables are homogeneous; otherwise, use of artificial neural networks technique is preferable.