ANN Robot Energy Modeling

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Abstract: This paper proposes energy modeling for robot based on real measurements data. First, the paper proposes six preliminary ANN Models on both carpet and hard floor. These models' Inputs: Theoretical Time, Theoretical Velocity and Output: The Current; then with Inputs: Theoretical Time, Theoretical Velocity and Output: The Voltage; and finally with Inputs: Time, Real Linear Velocity, Rotational Velocity and Output: The Voltage. Second, a global ANN model with time and speed as inputs and current, voltage, linear speed, rotational speed on carpet, along with current, voltage, linear speed, rotational speed on carpet, along with current, voltage, linear speed, rotational speed on the form of Simulink model after care selection of number in neurons in hidden layer. This model has the capability to predict and simulate the robot energy characteristics under different conditions. This real data measurements on both hard floor and carpet are presented to be used as training data for Neural Network. All the ANN models are checked in the form of minimum error, accuracy, good regression constants and comparisons between real and predicted data. ANN with feed forward back-propagation technique is used to implement the models. It is adopted to make benefits from its ability of interpolation. ANN models with Back - Propagation (BP) technique is created with suitable numbers of layers and neurons. The last model will be used with the aid of Genetic Algorithm to improve and optimize the energy efficiency of robots.

Keywords: Energy, Modeling, Simulink, Artificial Neural Network, Estimation.

I. Introduction

It is a very important issue to model Mobile robots in order to make complete performance analysis and optimize this performance. The purpose of proper modeling and optimization is to solve the problem of limited amount of robots energy. Many important contributions from other researchers related to robots modeling are reviewed with advantages and drawbacks before starting this work [1], [2], [3], [4], [5]. Robots design, control, and implementation work are done using ANN like [6], [7]. Due to the enormous development in the field of robots; robots and especially autonomous mobile robots have found their use in a lot of applications. These applications include search and rescue, security, rehabilitation, cleaning and delivery. Autonomous mobile robots most of the time rely on batteries as their energy source and batteries have very limited energy capacity. This finite amount of energy can make the robot work only for a limited time and this is why the use of these robots in complicated missions is not feasible. Although these robots could be refueled while they are operating and their time of operation can be increased; the cost of replacement of their energy source is too high to be realistic. So, our global ANN model will become a very important one especially for optimization analysis. Especially, this model accomplishes all important performance characteristics and based on real experimental measurements.

II. Experimental Data

The experimental data are drawn from the robot and associated measurements devices which used in [1] for the same authors as shown in the following samples figures. The used robot was built at the first author's laboratory by his previous students as shown in Fig. 1 to measure required data.



Fig. 1 The used Robot for measuring the training data [1]

The current is measured by an ACS712 current sensor as shown in Fig. 2. Current data is fed into an Arduino Mega 2560 during the robot's motion. The sampling time was set at 26 ms because our robot's maximum speed is 0.58 m/s



Fig. 2. ACS712 current sensor module [1]

This data introduced in the following figures and more is used as training data for ANN models.







III. Ann Models

Artificial Neural Network with feed forward back-propagation technique is used to implement the various ANN models [8]-[11]. All the models consist of one hidden contains log-sigmoid function and other is output layer contains pure-line function. Carpet Model 1: Inputs: Time, Velocity, Output: Current; with 7 neurons and 1 neuron. (R= 0.99105). 2nd Carpet Model: The same i/ps as 1st model, O/ps: Current, Voltage; 8 neurons and 2 neurons. (R= 0.9766). 3rd Carpet Model: I/ps: Time, Linear Velocity, Rotational Velocity; O/ps: Current, Voltage; 7 neurons and 2 neurons. (R= 0.99511). Hard floor: Model 1: I/ps: Time, Velocity; O/p: Current; 6 neurons, 1 neuron. (R= 0.98189). Model 2: same i/ps; O/p: Current, Voltage; 7 neurons. (R= 0.98188). Model 3: I/ps: Time, Linear Velocity, Rotational Velocity, 7 neurons, 2 neurons. (R= 0.97472).

A. Carpet Models

ANN Model 1 for the carpet:

Inputs: Theoretical Time, and Theoretical Velocity Output: The Current

Neural Network consists of two layers one hidden contains log-sigmoid function with seven neurons and the other is the output layer contains pure-line function with one neuron.

The normalized inputs eq.n are:

$$Time_n = (Time - 1.5000) / (0.9092)$$

Velocity_n = (Velocity - 0.1070) / (0.0649) (1)

Equation (1) presents the normalized input for the power and the following equations lead to the required derived equation.

n: Subscript denotes normalized parameters

Ei: Sum of input with input weight and input bias for each node in hidden layer in neural network Fi: Output from each node in hidden layer to output layer according to transfer function here is logsig E1= -0.0206 Time_n + 6.2084 Velocity_n + 0.8792

F1=1/(1 + exp(-E1))E2= - 41.9445 Time_n - 36.2368 Velocity_n + 107.7541 F2=1 / (1 + exp (- E2)) 1.8870 $Time_n \ + \ 4.3406 \ Velocity_n \ + \ 1.3116$ E3= F3=1 / (1 + exp (- E3)) $Time_n + 7.9388$ Velocity_n - 2.9239 E4= 12.2653 (2)F4=1 / (1 + exp (- E4)) Time_n - 58.0546 Velocity_n + 71.4314 E5= - 51.7383 F5=1 / (1 + exp (-E5))Time_n - 68.1760 Velocity_n + 180.9475 E6= - 63.1920 F6=1 / (1 + exp (- E6))Time_n - 2.7997 Velocity_n - 1.7873 E7= - 3.7685 F7=1/(1 + exp(-E7))The normalized current relation from ANN: $I_n = -171.7834 \ F1 - 114.3126 \ F2 + 318.7558 \ F3 + 4.4619 \ F4 - 1.4824 \ F5 + 113.6865 \ F6 + 150.6382$ (3)

(4)

F7 -149.7478

The un- normalized output (Current) on Carpet $I = 0.5977 * I_n + 1.2486$









Fig. 14 Regression for ANN Model

2nd ANN Model for the carpet:

Inputs: Theoretical Time, and Theoretical Velocity

Output: The Current, The Voltage

Neural Network consists of two layers one hidden contains log-sigmoid function with eight neurons and the other is the output layer contains pure-line function with two neurons.

Equation (1) presents the normalized input for the power and the following equations lead to the required derived equation.









Fig. 17 Regression for ANN Model

3rd ANN Model for the carpet:

Inputs: Time, Linear Velocity, and Rotational Velocity

Output: The Current, The Voltage

Neural Network consists of two layers one hidden contains log-sigmoid function with seven neurons and the other is the output layer contains pure-line function with two neurons.

The normalized inputs eq.n are: Time_n = (Time - 1.5000) / (0.9092) Linear_Velocity_n = (Linear_Velocity - 0.0617) / (0.0681) Rotational_Velocity = (Rotational_Velocity - 0.0094) / (0.0206) (10)Equation (10) presents the normalized input for the power and the following equations lead to the required derived equation. E1= -15.5297 Time_n - 5.3634 Linear_Velocity_n + 2.8217 Rotational_Velocity_n + 29.0958 $F_{1=1} / (1 + \exp(-E_{1}))$ E2= -9.7170 Time_n + 25.6755 Linear_Velocity_n + 6.7923 Rotational_Velocity_n + 5.5822 F2=1 / (1 + exp (- E2))E3= - 185.6356 Time_n + 77.7530 Linear_Velocity_n + 77.9753 Rotational_Velocity_n - 38.3748 F3=1/(1 + exp(-E3))3.2609 Time_n - 99.2405 Linear_Velocity_n - 112.4696 Rotational_Velocity_n + 97.7252 E4=F4=1/(1 + exp(-E4))17.2092 Time_n - 10.0285 Linear_Velocity_n - 2.0355 Rotational_Velocity_n - 4.6875 (11) E5= F5=1/(1 + exp(-E5))E6= -0.1548 Time_n + 1.3294 Linear_Velocity_n - 0.7397 Rotational_Velocity_n - 1.0455 F6=1 / (1 + exp (- E6))E7= -0.1677 Time_n + 1.3318 Linear_Velocity_n - 0.7223 Rotational_Velocity_n - 1.0656 F7=1 / (1 + exp (-E7))The normalized current relation from ANN: $I_n = 9.6341 \; F1 - 7.3508 \; F2 - 0.0096 \; F3 - 9.4918 \; F4 + 4.7189 \; F5 + 449.9340 \; F6 - 444.7533 \; F7 - 2.7859 \; F3 - 1.0096 \; F3 - 1.00$ (12) $V_n = -2.6704 \ F1 + 0.8331 \ F2 + 1.6853 \ F3 + 0.2252 \ F4 - 8.1039 \ F5 + 869.4779 \ F6 - 869.6830 \ F7 + 0.5736 \ (13)$ The un- normalized outputs (Current & Voltage) for carpet $I = 0.5977 * I_n + 1.2486$ (14)(15)







Fig. 20 Regression for ANN Model

B. Hard floor Models

ANN Model 1 for the hard floor: Inputs: Theoretical Time, and Theoretical Velocity Output: The Current Neural Network consists of two layers one hidden of

Neural Network consists of two layers one hidden contains log-sigmoid function with six neurons and the other is the output layer contains pure-line function with one neuron.

The normalized inputs eq.n are: $Time_n = (Time - 1.5000) / (0.9092)$ $Velocity_n = (Velocity - 0.1070) / (0.0649)$

(16)

Equation (16) presents the normalized input for the power and the following equations lead to the required derived equation.

N: Subscript denotes normalized parameters

Ei: Sum of input with input weight and input bias for each node in hidden layer in neural network Fi: Output from each node in hidden layer to output layer according to transfer function here is logsig E1= 3.5120 Time_n - 2.1097 Velocity_n - 5.2000F1=1 / (1 + exp (- E1)) E2= -4.5842 Time_n - 1.4050 Velocity_n + 6.5328F2=1 / (1 + exp (- E2))


Fig. 21 Output VS Target for ANN Model testing data





Fig. 23 Regression for ANN Model





3rd ANN Model for the HARD FLOOR: Inputs: Time, Linear Velocity, and Rotational Velocity Output: The Current, The Voltage Neural Network consists of two layers one hidden contains log-sigmoid function with seven neurons and the other is the output layer contains pure-line function with two neurons. The normalized inputs eq.n are: $Time_n = (Time - 1.5000) / (0.9092)$ Linear_Velocity_n = (Linear_Velocity - 0.0590) / (0.0636) Rotational_Velocity_n = (Rotational_Velocity + 0.0107) / (0.0239) (26)Equation (26) presents the normalized input for the power and the following equations lead to the required derived equation. E1= 6.1011 Time_n - 4.3725 Linear_Velocity_n - 0.5366 Rotational_Velocity_n - 7.8921 F1=1/(1 + exp(-E1))E2= 7.4497 Time_n - 24.2664 Linear_Velocity_n + 2.0769 Rotational_Velocity_n + 1.1058 F2=1 / (1 + exp (- E2))E3= -10.4953 Time_n + 15.7299 Linear_Velocity_n + 3.3590 Rotational_Velocity_n - 3.4718 F3=1/(1 + exp(-E3))E4= 7.1575 Time_n - 62.6552 Linear_Velocity_n + 10.0593 Rotational_Velocity_n + 8.0336 (27) F4=1/(1 + exp(-E4))E5=- 6.5531 Time_n + 9.4656 Linear_Velocity_n + 1.4800 Rotational_Velocity_n - 3.4261 F5=1/(1 + exp(-E5))E6= -45.3300 Time_n + 22.5199 Linear_Velocity_n - 23.0080 Rotational_Velocity_n - 22.0751 F6=1 / (1 + exp (- E6))E7= 138.7388 Time_n - 49.9340 Linear_Velocity_n + 10.4912 Rotational_Velocity_n + 38.7297 F7=1 / (1 + exp (-E7))The normalized current relation from ANN: $I_n = -31.3817 F1 - 580.7507 F2 - 1.5190 F3 + 580.0468 F4 + 2.8268 F5 - 0.2776 F6 + 0.6847 F7 - 0.1780 (28)$ $V_n = -\ 551.2813\ F1 + 106.8142\ F2 - 3.2684\ F3 - 106.6242\ F4 + 2.5206\ F5 + 1.6734\ F6 + 0.7518\ F7 + 0.1575$ (29)

The un- normalized outputs (Current & Voltage) for carpet	
$I = 0.5977 * I_n + 1.2486$	(30) (31)
$V = 0.0808 * V_n + 7.6220$	







Fig. 29 Regression for ANN Model

Global model: ANN model with time and speed as inputs and current, voltage, linear speed, rotational speed on carpet, with current, voltage, linear speed, rotational speed on hard floor.





Fig. 31 Full detailed ANN Simulink Model





IV. Conclusion

Using the Artificial Neural Network (ANN), with feed forward back-propagation technique to introduce Robot model. This is done to use the ability of neural network for interpolation. It can predict the characteristics and performance of this mobile robot properly. The model are checked and verified by comparing actual and predicted ANN values, with good error and excellent regression factor. Finally, the algebraic equations could be deduced to use them without training the neural unit in each time or using the Simulink model to use them for optimization purposes for future work.

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