

Prctica 1: Prediction with Back-Propagation and Linear Regression

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1 Introduction

The goal of any supervised learning algorithm is to find a function that best maps a set of inputs to its correct output. Back-propagation is a common method for training a neural network. The goal and motivation for developing the backpropagation algorithm was to find a way to train a multi-layered neural network such that it can learn the appropriate internal representations to allow it learn any arbitrary mapping of input to output. Alongside Multiple linear regression attempts to model the relationship between two or more explanatory variables and a response variable by fitting a linear equation to observed data. Every value of the independent variable x is associated with a value of the dependent y .

2 Objective

Objective: Prediction of the power of the turbine of a hydro electrical plant, using the following algorithms:

1. Back-Propagation (BP)
2. Multiple Linear Regression (MLR)

3 Back-Propagation (BP)

Backpropagation is a supervised learning algorithm and is mainly used by Multi-Layer-Perceptrons to change the weights connected to the net's hidden neuron layer(s). The back-propagation algorithm uses a computed output error to change the weight values in backward direction. To get this net error, a forward-propagation phase must have been done before. While propagating in forward direction, the neurons are being activated using

the sigmoid activation function. The formula of sigmoid activation is: $f(x) = 1/(1+e^{-x})$. The algorithm works as follows:

```
Randomly initialize weights and thresholds
for epoch 1 to total number of epoch
    for pattern 1 to total number of pattern
        choose random pattern.
        feed-forward
        back-propagation of error
        calculation of weights and thresholds updates
        update weights and thresholds
    end
end
```

Implementation decisions:

In this current implementation we can choose number of hidden layers and number of neurons per layer. For our data it is better to use two hidden layers. On the other hand to choose I have chosen thirteen nodes in first hidden layer and 5 at second hidden layer. Because this network architecture gives best prediction with error rate 0.57%

Along side number of epoch and learning rate plays an important role in the prediction. In this case I preferred epoch = 10000000 and Learning rate = 0.1

Language: Java

Tool: Eclipse JavaEE mars

Result:

| real | prediction | error | total error |
|---------|------------|-------|-------------|
| 3698.29 | 3705.107 | 6.82 | 0.57% |
| 1935.13 | 1947.488 | 12.36 | |
| 4509.67 | 4509.252 | 0.42 | |
| 5708.99 | 5701.963 | 7.03 | |
| 2024.05 | 2030.317 | 6.27 | |
| 3981.74 | 3990.353 | 8.61 | |
| 6171.22 | 6180.054 | 8.83 | |
| 4427.36 | 4420.862 | 6.5 | |
| 2117.9 | 2098.754 | 19.15 | |
| 2692.87 | 2689.051 | 3.82 | |
| 3237.11 | 3260.684 | 23.57 | |
| 3558.33 | 3590.829 | 32.5 | |
| 2327.54 | 2313.536 | 14 | |
| 4063.35 | 4089.213 | 25.86 | |
| 3838.09 | 3852.186 | 14.1 | |
| 6098.9 | 6114.126 | 15.23 | |
| 3432.48 | 3454.15 | 21.67 | |
| 2042.36 | 2038.665 | 3.69 | |
| 2841.89 | 2841.965 | 0.08 | |
| 4572.62 | 4540.121 | 32.5 | |
| 4384.97 | 4377.646 | 7.32 | |

| | | |
|---------|----------|--------|
| 4384.97 | 4377.646 | 7.32 |
| 6297.76 | 6345.418 | 47.66 |
| 2589.43 | 2585.458 | 3.97 |
| 2569.39 | 2563.981 | 5.41 |
| 4192.01 | 4187.853 | 4.16 |
| 3692.94 | 3702.912 | 9.97 |
| 2081.66 | 2090.112 | 8.45 |
| 4591.21 | 4731.523 | 140.31 |
| 2590.7 | 2614.435 | 23.74 |
| 4102.62 | 4126.668 | 24.05 |
| 2865.92 | 2877.225 | 11.3 |
| 3090.96 | 3142.397 | 51.44 |
| 4096.06 | 4023.985 | 72.08 |
| 1938.47 | 1941.835 | 3.37 |
| 4801.54 | 4758.522 | 43.02 |
| 3299.93 | 3317.742 | 17.81 |
| 3670.99 | 3693.869 | 22.88 |
| 3743.56 | 3741.261 | 2.3 |
| 3128.97 | 3147.344 | 18.37 |
| 1776.96 | 1839.691 | 62.73 |
| 5153.46 | 5160.462 | 7 |
| 5016.03 | 4947.968 | 68.06 |

| | | |
|---------|----------|-------|
| 5016.03 | 4947.968 | 68.06 |
| 3618.87 | 3636.821 | 17.95 |
| 5632.13 | 5647.897 | 15.77 |
| 5283.22 | 5282.003 | 1.22 |
| 2337.68 | 2311.808 | 25.87 |
| 6451.24 | 6505.081 | 53.84 |
| 4806.83 | 4814.235 | 7.4 |
| 4450.58 | 4478.64 | 28.06 |
| 2893.53 | 2905.953 | 12.42 |

Evaluation of the predictions(Error and Plot):
Error: 0.57%

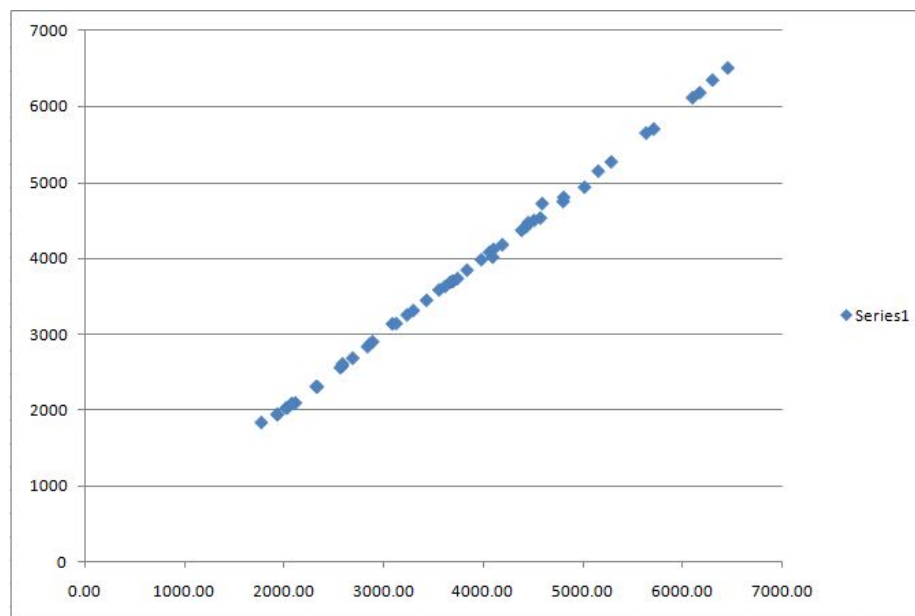


Figure1: Back-propagation prediction plot

4 Multiple Linear Regression

Multiple linear regression attempts to model the relationship between two or more explanatory variables and a response variable by fitting a linear equation to observed data. Every value of the independent variable x is associated with a value of the dependent variable y . Formally, the model for multiple linear regression, given n observations, is $y = b_0 + b_1x_1 + b_2x_2 + \dots + b_nx_n$ for $i = 1, 2, \dots, n$

Implementation decisions:

Language: Using software

Tool: Microsoft office Excel 2007

Result:

| Real | Prediction | Error | Total Error |
|---------|-------------|------------|-------------|
| 3698.29 | 3810.2275 | 111.9375 | 4.97% |
| 1935.13 | 1552.841207 | 382.288793 | |
| 4509.67 | 4649.645717 | 139.975717 | |
| 5708.99 | 5585.837291 | 123.152709 | |
| 2024.05 | 1739.008274 | 285.041726 | |
| 3981.74 | 3993.76927 | 12.02927 | |
| 6171.22 | 5913.775515 | 257.444485 | |
| 4427.36 | 4540.332976 | 112.972976 | |
| 2117.9 | 2037.832114 | 80.067886 | |
| 2692.87 | 2640.237134 | 52.632866 | |
| 3237.11 | 3132.144469 | 104.965531 | |
| 3558.33 | 3912.385969 | 354.055969 | |
| 2327.54 | 2183.41351 | 144.12649 | |
| 4063.35 | 4132.439131 | 69.089131 | |
| 3838.09 | 3829.800158 | 8.289842 | |
| 6098.9 | 6083.748464 | 15.151536 | |
| 3432.48 | 3693.760487 | 261.280487 | |
| 2042.36 | 2288.678167 | 246.318167 | |
| 2841.89 | 3015.686084 | 173.796084 | |
| 4572.62 | 4973.168076 | 400.548076 | |
| 4384.97 | 4485.676605 | 100.706605 | |

| | | |
|---------|-------------|------------|
| 4102.62 | 4058.612081 | 44.007919 |
| 2865.92 | 2729.977216 | 135.942784 |
| 3090.96 | 3562.24979 | 471.28979 |
| 4096.06 | 4505.587007 | 409.527007 |
| 1938.47 | 1960.739944 | 22.269944 |
| 4801.54 | 5082.480817 | 280.940817 |
| 3299.93 | 3475.135005 | 175.205005 |
| 3670.99 | 3703.774163 | 32.784163 |
| 3743.56 | 3913.813648 | 170.253648 |
| 3128.97 | 3142.330909 | 13.360909 |
| 1776.96 | 1578.14535 | 198.81465 |
| 5153.46 | 5591.841129 | 438.381129 |
| 5016.03 | 5288.229172 | 272.199172 |
| 3618.87 | 3524.925503 | 93.944497 |
| 5632.13 | 5670.823766 | 38.693766 |
| 5283.22 | 5312.555438 | 29.335438 |
| 2337.68 | 2099.404912 | 238.275088 |
| 6451.24 | 6302.373946 | 148.866054 |
| 4806.83 | 5039.273585 | 232.443585 |
| 4450.58 | 4401.429853 | 49.150147 |
| 2893.53 | 2873.914928 | 19.615072 |

Evaluation of the predictions(Error and Plot): Error: 4.97%

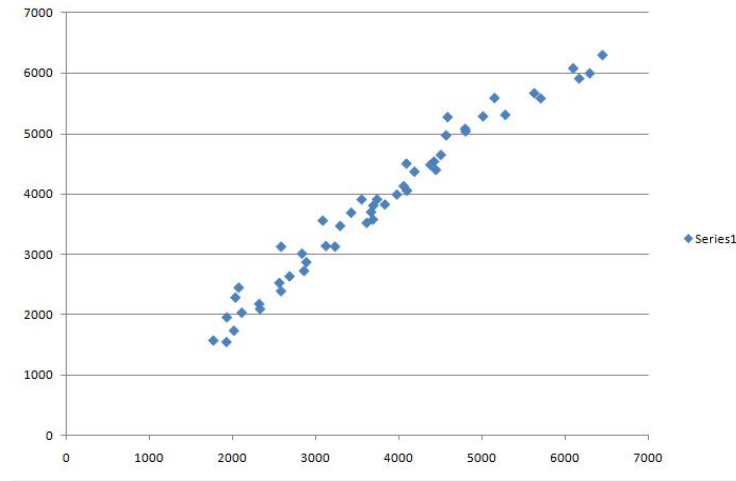


Figure2: Multi linear regression prediction plot

5 Conclusion

Our experiment suggests that ANN-BP technology holds great promise for the prediction of turbine data set. This research has provided some evidence that is very promising from ANN-BP rather than multi-linear regression.

References

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