Group 7404

THE EFFECT OF LIMB POSITION ON MYOELECTRIC PROSTHETIC CONTROL USING LINEAR REGRESSION



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INTRODUCTION

Electromyography (EMG) is widely used for controlling functional prosthetics. However, EMG signals for the same movements change with variations in limb position and lowers the accuracy in control schemes [1]. Most previous studies have utilized classification for pattern recognition when changing limb position, with a negative effect in performance. Linear regression is a newer method in control of myoelectric prosthetics, which has proven to yield robust simultaneous and proportional control [2]. This study investigated the effect of limb position in a linear regression-based control scheme, when using the commonly used Mean Absolute Value (MAV) and Logarithmic Variance (LogVar) feature, where the latter has shown linear properties [2].

METHODS AND MATERIAL

Surface EMG (sEMG) data was collected from seven able-bodied subjects. The subjects were instructed to performed four different hand gestures (Figure 1), in three limb positions (down the side, lifted to the side, lifted forward). sEMG signals were recorded with a Myo armband (eight channels, 200Hz sample rate), positioned on the dominant forearm of the subjects.

The regression models (regressors) were calculated as following:

$$Y = \alpha + \beta X \tag{1}$$

One regressor was build for each wrist movement for each test subject: four for each feature. The offline regressor accuracy was tested qualitatively through superimposition of the expected and calculated output.



Figure 1: A myoelectric transradial prosthetic.

The regressors were tested quantitatively by calculating the Root Mean Square Error (RMSE) between expected and calculated output for both training and test data. Testing of the regressors were done online in a virtual environment, where the time to complete a target-reaching task of 16 targets was measured.

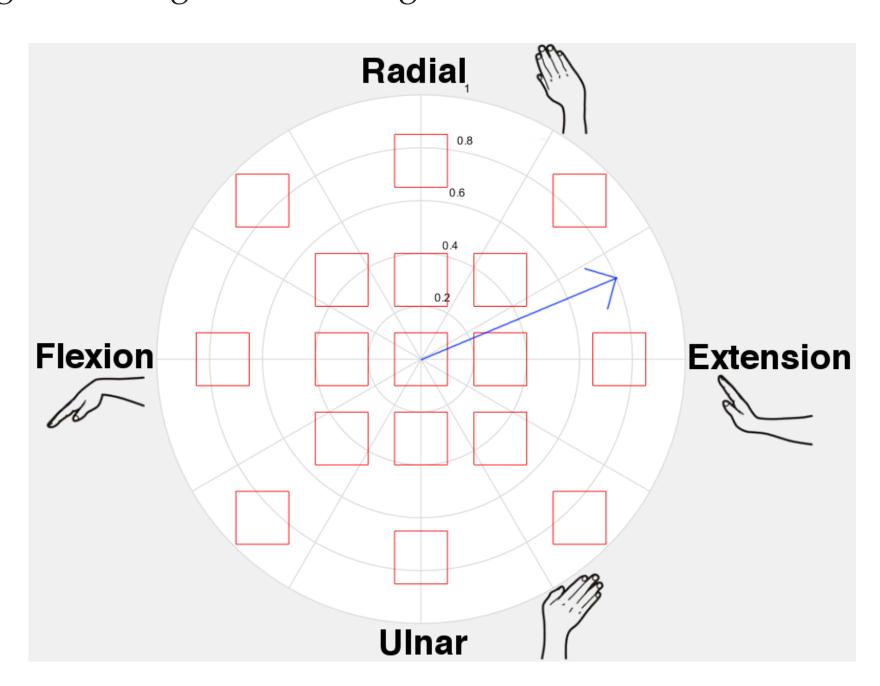


Figure 1: The interface for the target reaching task. During testing an arrow originating from origin would show direction and intensity of movement

The performance (time per reached target) of the online test was compared between the different limb positions of the same feature and between all limb positions of the two features through statistical analysis.

AIM

The aim for this project is expressed in the following hypothesis:

• It is possible to yield similar performance in different limb positions in myoelectric prosthetic control using a linear regression based control scheme.

ONLINE RESULTS

Results for the online test of regressor accuracy and control. The test is performed in a modified Fitts' Law test of reaching targets. The score is calculated as time per reached targets.

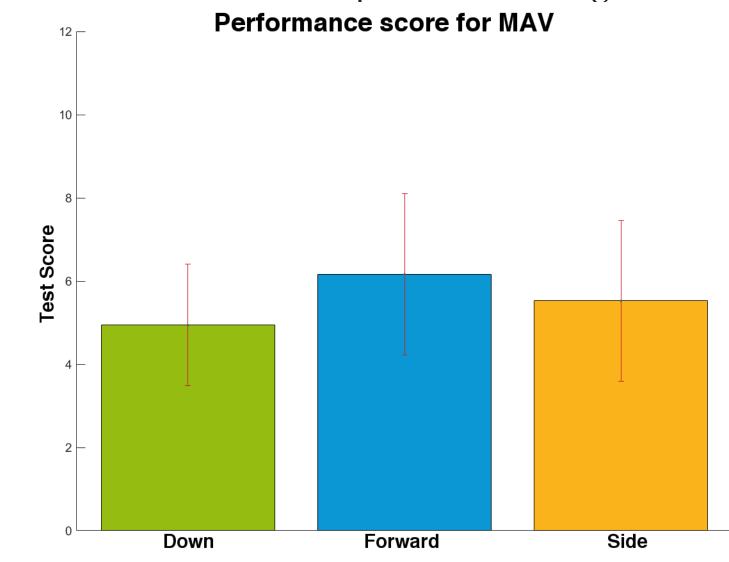


Figure 5: The mean test score among seven subjects when using regressors trained with MAV

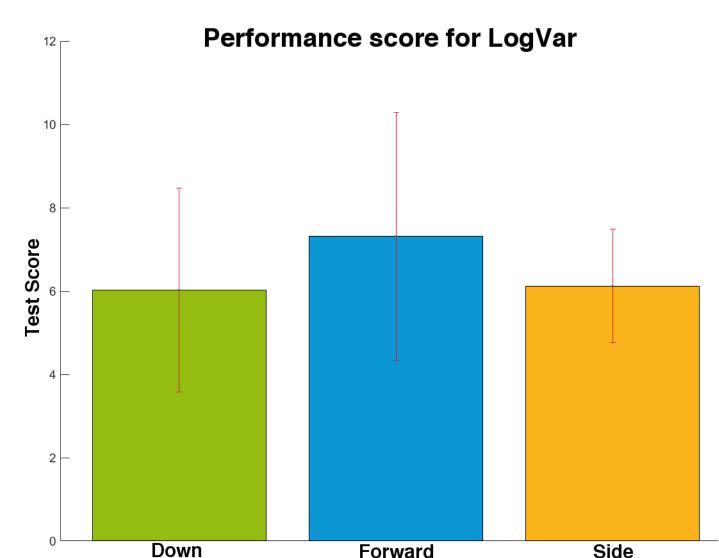


Figure 6: The mean test score among seven subjects when using regressors trained with LogVar

Using a Friedman's test the performance scores between the three limb positions prove not to be significantly different (p = 0.1561), when applying the MAV trained regressors in the online test. For the LogVar trained regressors the performance score between all limb positions cannot be proven significantly different either (p = 0.5647). There was no significant difference in the time to reach the targets across the two features (MAV: $5.5 \, \text{s}$, LogVar: $6.5 \, \text{s}$; p = 0.13).

OFFLINE RESULTS

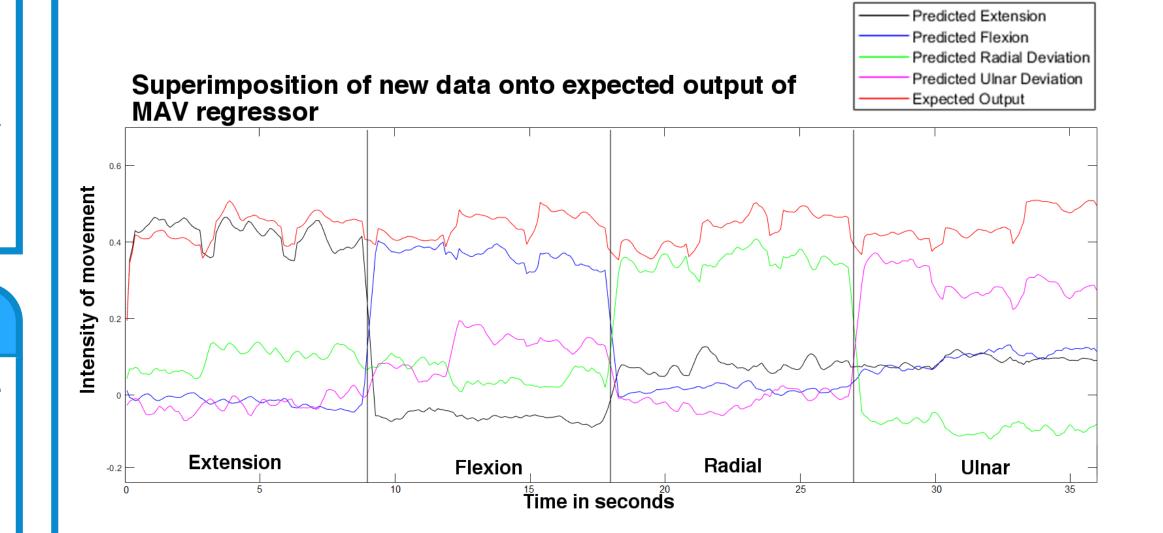


Figure 2: Superimposition of test data and the regressor output for the MAV feature.

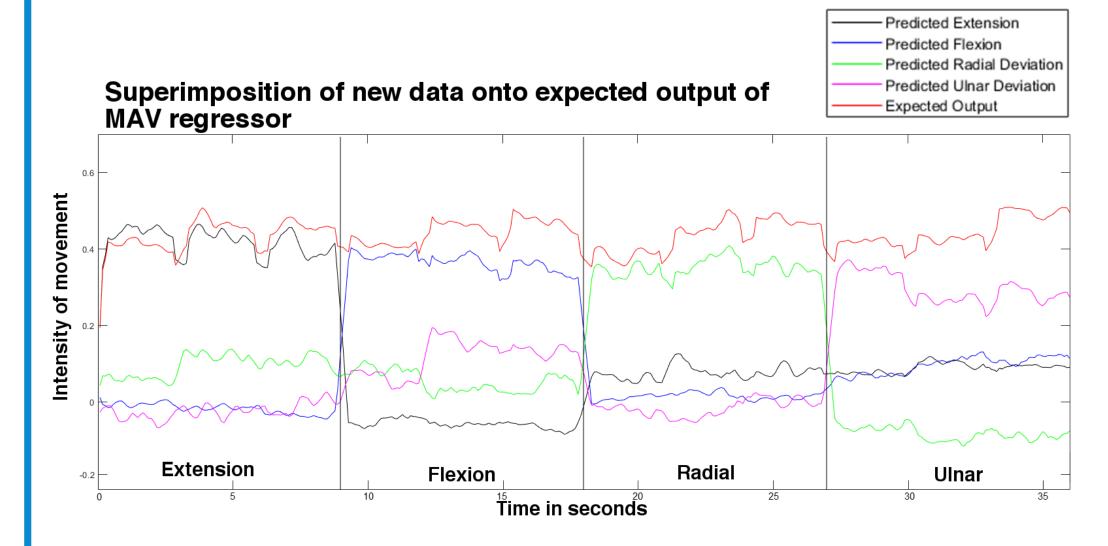


Figure 3: Superimposition of test data and the regressor output for the LogVar feature.

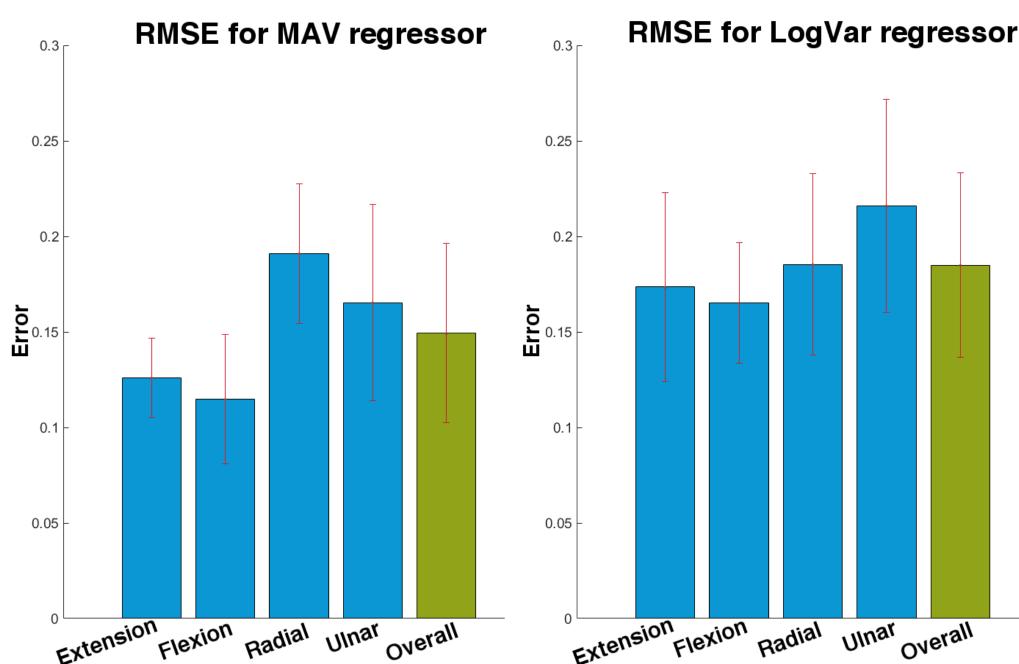


Figure 4: RMSE when using test data

Feature	Overall mean error	Standard deviation
MAV	0.1493	± 0.0469
LogVar	0.1850	± 0.0484

Table 1: Overall mean RMSE when using test data as input in the regression model

Comparing the RMSE of MAV and LogVar when using test data again indicated significant difference (p = 0.0082), where LogVar again has the higher mean. The RMSE of the MAV when using training data proves significantly smaller than the RMSE when using test data (p = 0.0002). Similar results are obtained for the LogVar regressor (p = 0.0002).

DISCUSSION

- The online test results indicate that linear regression can be implemented as control scheme in myoelectric prosthetic control to yield similar performance across variations of limb position. This is opposed to previous studies using classification as control scheme.
- It was found that offline and online performance of the implemented regressors is not necessarily correlated. The offline tests showed overfitting of the regression models, but the online test yielded robust control of wrist movements in all limb positions, which could be a result of the subjects' ability to compensate for a poorer fitting model, when visual feedback is provided.
- These findings agree with a recently published study by Hwang et al. [3], who tested other limb postions with the RMS feature.
- Even though LogVar shows linear properties in previous studies, the online test show no significant difference in performance when compared to MAV.
- Linear regression has the potential to be used in future control schemes for myoelectric prosthetics for use in daily life tasks.

REFERENCES

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