Authentic Science Research Program

Byram Hills HS, Armonk NY

## USING FORCE SENSORS TO EFFECTIVELY CONTROL A BELOW-ELBOW INTELLIGENT PROSTHETIC DEVICE

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Introduction

- US troops in Iraq required limb amputations at twice the rate of past wars (The Boston Globe, 2004)
- 795 amputations for soldiers in Iraq and Afghanistan since December 2001



Figure 1: Wounded Soldier (thememoryhole.org, 2004)

(The Seattle Times, 2007)

Myoelectric Prosthetic Control



Figure 2: i-Limb Hand (Touch Bionics, Inc., 2007)

#### Myoelectric Control

- Current applied state-of-the-art
- Residual muscle contractions control hand
- Myoelectric signals (MES) measured from muscles (Muzumdar, 2004)



Figure 3: Variation of Myoelectric Signal with Contraction Level (D.F. Lovely, 2004)

Limitations of Pattern-Recognition Myoelectric Control

### Only 95% Accurate

• Advanced models predict accurately 95% of the time (L. J. Hargrove et al., 2007)

### Surgical Implantation Required for Pattern Recognition

Electrodes may become infected or fall out of place (Muzumdar, 2004)

### Signal Processing Required

- No processing technique is perfect (L. P. J. Kenney et al., 1999)
- Difficult to customize hand (A.S. Poulton et al., 2002)

### High Cost

• Each electrode can cost \$2000 (P. Kyberd, 2007)

Force Sensors – A More Accurate Control Method

#### • Force sensors

- Applied Pressure  $\rightarrow$  Change in output voltage
- Can measure muscle contraction
- Never before used in multi-sensor, pattern recognition setup with the purpose of controlling a prosthesis (L.J. Kenney et al., 1999)

## SO WHY USE FORCE SENSORS?

Force Sensors – A More Accurate Control Method

# LESS EXPENSIVE

Force Sensors – A More Accurate Control Method

# NO SURGERY NEEDED

Force Sensors – A More Accurate Control Method

# POTENTIAL TO ELIMINATE OR REDUCE NEED FOR SIGNAL PROCESSING

Force Sensors – A More Accurate Control Method

# Are Force Sensors as accurate as myoelectrodes?

### Hypothesis



### **Objectives**



### Build proof-of-concept hand



• Develop computer interface and acquire raw data



 Write and implement input analysis program

#### Objective 1 - Prosthetic Prototype









Figure 6: Vibration Sensor Circuit (Kyberd/Blum, 2007)

*Objective 2 - Computer Interfaced Force Sensor circuit* 





Figure 7: Lateral Arm Aspect (Blum, 2007)

#2-Extensor Carpi Ulnaris #4-Supinator #6-Extensor Carpi Radialis Brevis



Figure 8: Medial Arm Aspect (Blum, 2007) #1-Pronator Teres#3-Flexor DigitorumSuperficialis#5-Flexor Carpi Ulnaris

Table 1: Six Forearm Muscles Tested (Green Muscles had sensors) (Blum, 2007)Flexor Digitorum Superficialis - finger flexionExtensor Carpi Ulnaris - finger extensionFlexor Carpi Ulnaris - wrist flexionExtensor Carpi Radialis Brevis - wrist<br/>extensionPronator Teres - wrist pronationSupinator - wrist supination

Objective 2 – Computer Interfaced Force Sensor circuit





Figure 9: DAQ and amplification circuit board with four force sensors connected (Blum, 2007)



Figure 10: TLC2274 Operational Amplifier (adapted from Texas Instruments, 2007)

#### Objective 2 - Computer Interfaced Force Sensor circuit





Figure 11: Internally mounted force sensors (Blum, 2007)



Figure 12: Cast fully equipped with sensors mounted (Blum, 2007)



Figure 13: Cast worn by Blum, Connected to DAQ (Blum, 2007)

**Objective 2 - Computer Interfaced Force Sensor circuit** 





Objective 3 – Computer input analysis program

**SVEN Function – linear discriminant analysis** (Torunn Midtgaard, 2006)  $F(x) = Wx + w_0$ 

> F(x) > 0 ACTIVATION  $F(x) \le 0$  NO ACTIVATION

When calibration muscles matches activation muscle, F(x) should be > 0



*Objective 1 – Prosthetic Prototype* 





Video 1: Prosthetic Prototype Demonstration (Blum, 2007)

Hand opens and closes upon force input Force Approximation Hand can grip objects Slip circuit successfully arrests slip

\*Low cost (force sensors are 100 times less expensive than myoelectrodes)

\*no surgery required

**Objective 2 – Matlab Data Acquisition** 

#### **Muscles Relaxed**



Figure 14 (a & b): Inactive Myoelectrode and Force Sensor Calibration Data (Blum, 2007)



**Objective 2 – Matlab Data Acquisition** 



#### Finger Flexion (Flexor Digitorum Superficialis)

#### (b) Active Calibration Data for **Force Sensors** Dataset1-Contraction Active (Calibration) Finger Flexion Dataset1-Contraction Active (Calibration) Finger Flexion 5 5 Flexor Digitorum Superficialis Flexor Digitorum Superficialis 4.5 4.5 Extensor Carpi Ulnaris Extensor Carpi Ulnaris Flexor Carpi Ulnaris Flexor Carpi Ulnaris 4 4 Extensor Carpi Radialis Brevis Extensor Carpi Radialis Brevis 3.5 3.5 3 3 Signal(Volts) Signal(Volts) 2.5 2.5 2 2 1.5 1.5 1 0.5 0.5 A STORE STALL DOOR 0 1000 2000 3000 4000 5000 6000 7000 8000 9000 10000 2000 3000 4000 5000 6000 7000 8000 9000 10000 Samples Samples

Figure 15 (a & b): Comparison of Finger Flexion Raw Calibration Data (Blum, 2007)

#### (a) Active Calibration Data for Myoelectrodes

**Objective 2 – Matlab Data Acquisition** 



(b) Active Calibration Data for **Force Sensors** 

#### Finger Extension *(Extensor Carpi Ulnaris)*

#### (a) Active Calibration Data for Myoelectrodes

Signal(Volts)

#### Dataset1-Contraction Active (Calibration) **Finger Extension** Dataset1-Contraction Active (Calibration) **Finger Extension** 5 5 Flexor Digitorum Superficialis Flexor Digitorum Superficialis 4.5 4.5 Extensor Carpi Ulnaris Extensor Carpi Ulnaris Flexor Carpi Ulnaris Flexor Carpi Ulnaris 4 4 Extensor Carpi Radialis Brevis Extensor Carpi Radialis Brevis 3.5 3.5 3 3 Signal(Volts) 2.5 2.5 2 2 1.5 1.5 0.5 0.5 0 0 Π 1000 2000 3000 4000 5000 6000 7000 8000 9000 10000 0 1000 2000 3000 4000 5000 6000 7000 8000 9000 10000 Samples Samples

Figure 16 (a & b): Comparison of Finger Extension Raw Calibration Data (Blum, 2007)

Objective 2 – Matlab Data Acquisition



(b) Active Calibration Data for **Force Sensors** 

#### Wrist Flexion (Flexor Carpi Ulnaris)

#### (a) Active Calibration Data for Myoelectrodes

#### Dataset1-Contraction Active (Calibration) Wrist Flexion Dataset1-Contraction Active (Calibration) Wrist Flexion 5 5 Flexor Digitorum Superficialis Flexor Digitorum Superficialis 4.5 4.5 Extensor Carpi Ulnaris Extensor Carpi Ulnaris Flexor Carpi Ulnaris Flexor Carpi Ulnaris 4 4 Extensor Carpi Radialis Brevis Extensor Carpi Radialis Brevis 3.5 3.5 Signal(Volts) Signal(Volts) 2.5 2.5 2 2 1.5 1.5 0.5 0.5 1000 2000 3000 4000 5000 6000 7000 8000 10000 2000 3000 4000 5000 6000 7000 8000 9000 10000 n Samples Samples

Figure 17 (a & b): Comparison of Wrist Flexion Raw Calibration Data (Blum, 2007)

**Objective 2 – Matlab Data Acquisition** 



#### Wrist Extension (Extensor Carpi Radialis Brevis)



Figure 18 (a & b): Comparison of Wrist Extension Raw Calibration Data (Blum, 2007)

Objective 3 – Matlab Analysis of Accuracy with SVEN



#### Finger Extension *(Extensor Carpi Ulnaris)*



Figure 19 (a & b): Myoelectric and Force Sensor Finger Extension Action Data for SVEN Function (Blum, 2007)

**Objective 3 – MATLAB Comparison of Accuracies** 

#### <u>Finger Flexion</u> ≠ Finger Extension (Should not exceed zero)



Figure 20 (a & b): Myoelectric and Force Sensor Finger Extension Smoothed SVEN Function (Blum, 2007)

**Objective 3 – MATLAB Comparison of Accuracies** 



#### **Finger Extension = Finger Extension (Should exceed zero)**



Figure 21 (a & b): Myoelectric and Force Sensor Finger Extension Smoothed SVEN Function (Blum, 2007)

**Objective 3 – MATLAB Comparison of Accuracies** 

#### <u>Wrist Flexion</u> ≠ Finger Extension (Should not exceed zero)



Figure 22 (a & b): Myoelectric and Force Sensor Finger Extension Smoothed SVEN Function (Blum, 2007)



**Objective 3 – MATLAB Comparison of Accuracies** 



#### <u>Wrist Extension</u> ≠ Finger Extension (Should not exceed zero)



Figure 23 (a & b): Myoelectric and Force Sensor Finger Extension Smoothed SVEN Function (Blum, 2007)

### Conclusions

- Low cost  $\rightarrow$  open to new socioeconomic group
- No implantation  $\rightarrow$  no risk of infection or sensor movement
- Prosthesis can be easily removed
- Little interference + low cross-talk = high accuracy rates
- Pattern Recognition (SVEN function) mostly works, but is not perfect for use with force sensors
- Sufficient voltage separation → eliminate post processing (voltage boundary can be measured using a comparator circuit with a fixed reference voltage)

With additional research, force sensor technology can be used in future prosthetic devices

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