

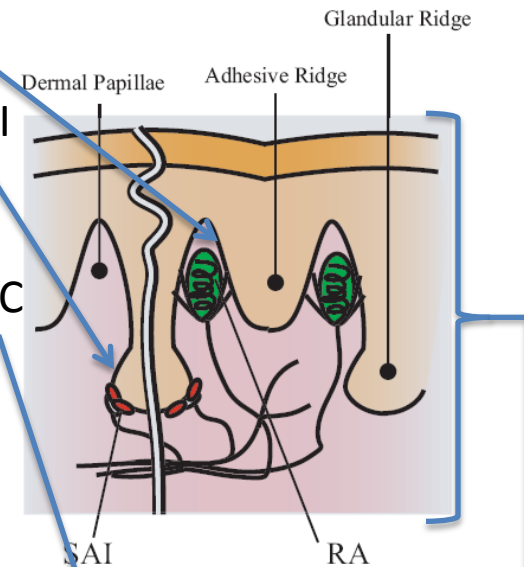
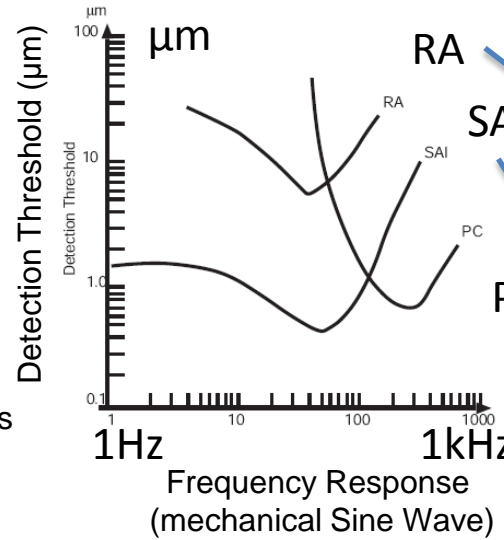
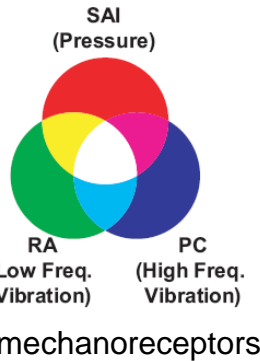
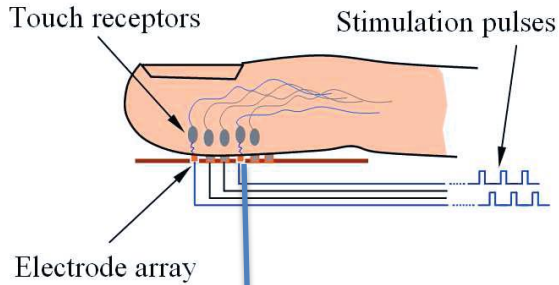
Real-time Identification of Human Finger Electrical impedance for haptic Rendering

Authors: John Gregory,
Dr. Yantao Shen and Dr. Ning Xi

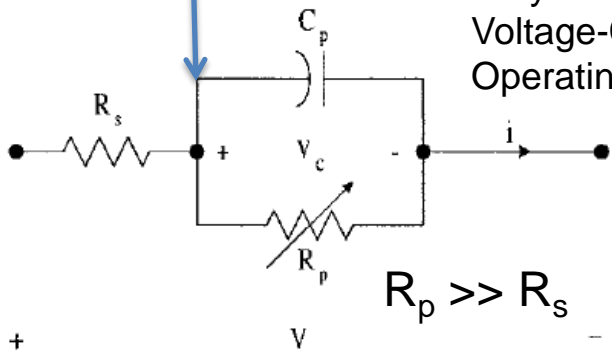
Presented for 2010 Inverse Problems Symposium
June 6th to 8th, 2010 Michigan State University

Principle of Electronic Tactile Sensing

Illustration of fingertip electro-tactile through electrode arrays.



Only valid at a set Voltage-Current Operating Point

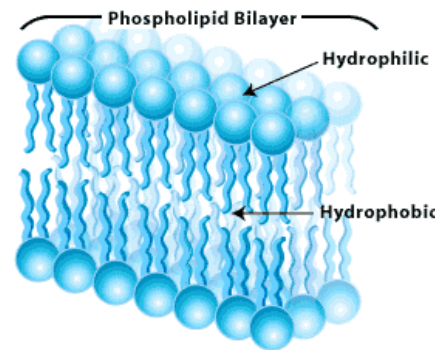


Simple Electronic Impedance Model of Skin

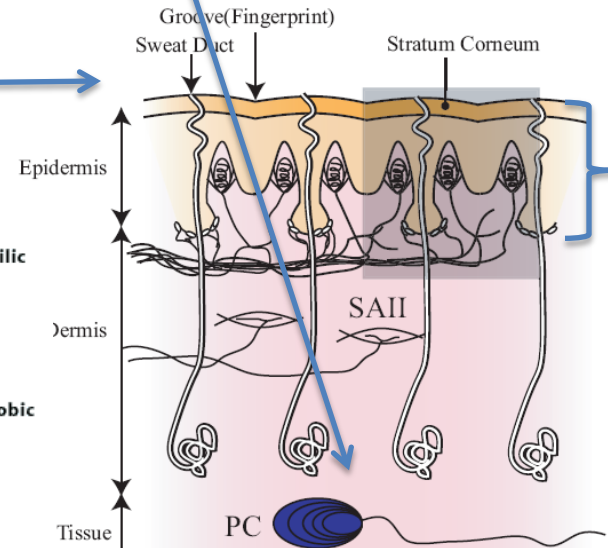
R_s = models deep tissue

R_p = S.C. Impedance (non linear and time variant)

C_p = cell membrane (lipid bylayer that polarizes from a voltage potential)



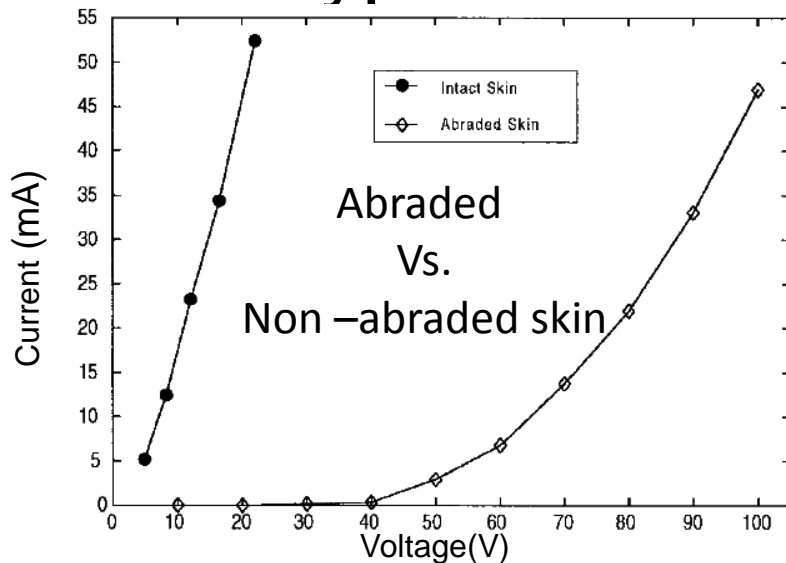
Lipid-Bi-Layer
Cellular Membrane



Adaptations from:

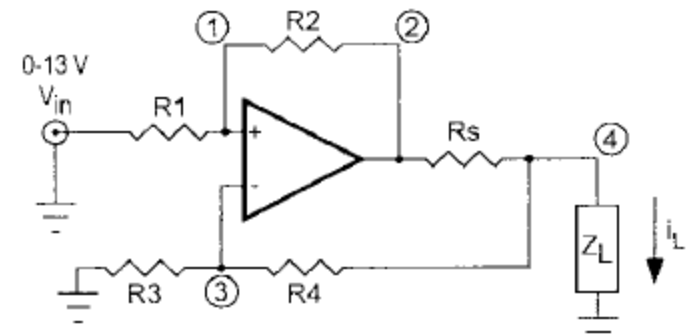
Hiroyuki Kajimoto, et al., "Electro-Tactile Display with Tactile Primary Color Approach", ISR 2004, 35th International Symposium on Robotics, 23-26 March, Paris France

Two Types of Electro-Tactile Stimulators



S.S. I-V characteristic for Abraded and Non-Abraded Skin Measured using a CCD

The Two Classes of Tactile Display Drivers

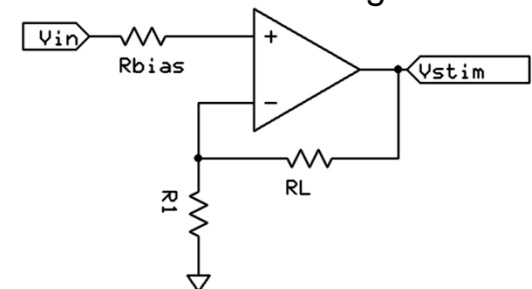


Howard Current Pump Tactile Driver

Constant Current

VS

Constant Voltage



Non-Inverting OpAmp (Our Driver)

Removal of the S.C to make system appear LTI
Invasive because removal of S.C. or use of
subcutaneous Implants needed to achieve stimulation

Both have been shown to mimic pressure or mechanical stimulation :

Constant Current Driver (CCD) (popular method)

➤ Way of neglecting a model of skin electrical impedance (just set current level)

Constant Voltage Driver (CVD) (our preferred method) :

Attempt to set voltage pulse waveform at an operating point

➤ Cannot be used in open loop system (cannot control painful or harmful current)

Ultimately the **skin is an obstacle** that must be overcome to stimulate the **axons of the mechanoreceptors** underneath so that they produce an action potential signal to the CNS

Why Using a CCD is not an optimal approach

Table VI

Identification of Parameters for 10 different subjects; Dry and Damp Finger Case. Under square wave stimulation
Operating point held at 75 volts or below sensation threshold

subject	sex	age	finger condition	R _p (kΩ)	C _p (nF)	R _s (kΩ)	RRSE %
1	M	30	DAMP	20.17	0.845	4.07	0.030
2	M	29	DAMP	44.49	0.379	90.29	0.033
3	M	26	DAMP	65.84	0.445	7.11	0.023
4	M	26	DAMP	23.21	0.847	3.93	0.042
5	M	28	DAMP	12.49	0.201	15.81	0.026
6	F	late 30's	DAMP	14.53	0.165	19.44	0.052
7	M	27	DAMP	16.48	0.160	19.81	0.022
8	F	28	DAMP	38.30	0.291	13.17	0.038
9	M	35	DAMP	30.96	0.344	11.30	0.056
10	M	30's	DAMP	61.89	0.382	8.410	0.020
1	M	30	DRY	194.00	0.165	19.00	0.031
2	M	29	DRY	272.67	0.104	30.45	0.024
3	M	26	DRY	258.19	0.085	38.39	0.032
4	M	26	DRY	195.04	0.012	26.18	0.025
5	M	28	DRY	236.34	0.115	27.55	0.022
6	F	late 30's	DRY	615.58	0.045	69.83	0.062
7	M	27	DRY	265.91	0.082	39.54	0.020
8	F	28	DRY	235.55	0.102	31.86	0.030
9	M	35	DRY	271.97	0.140	22.17	0.043
10	M	30's	DRY	258.80	0.075	44.72	0.028

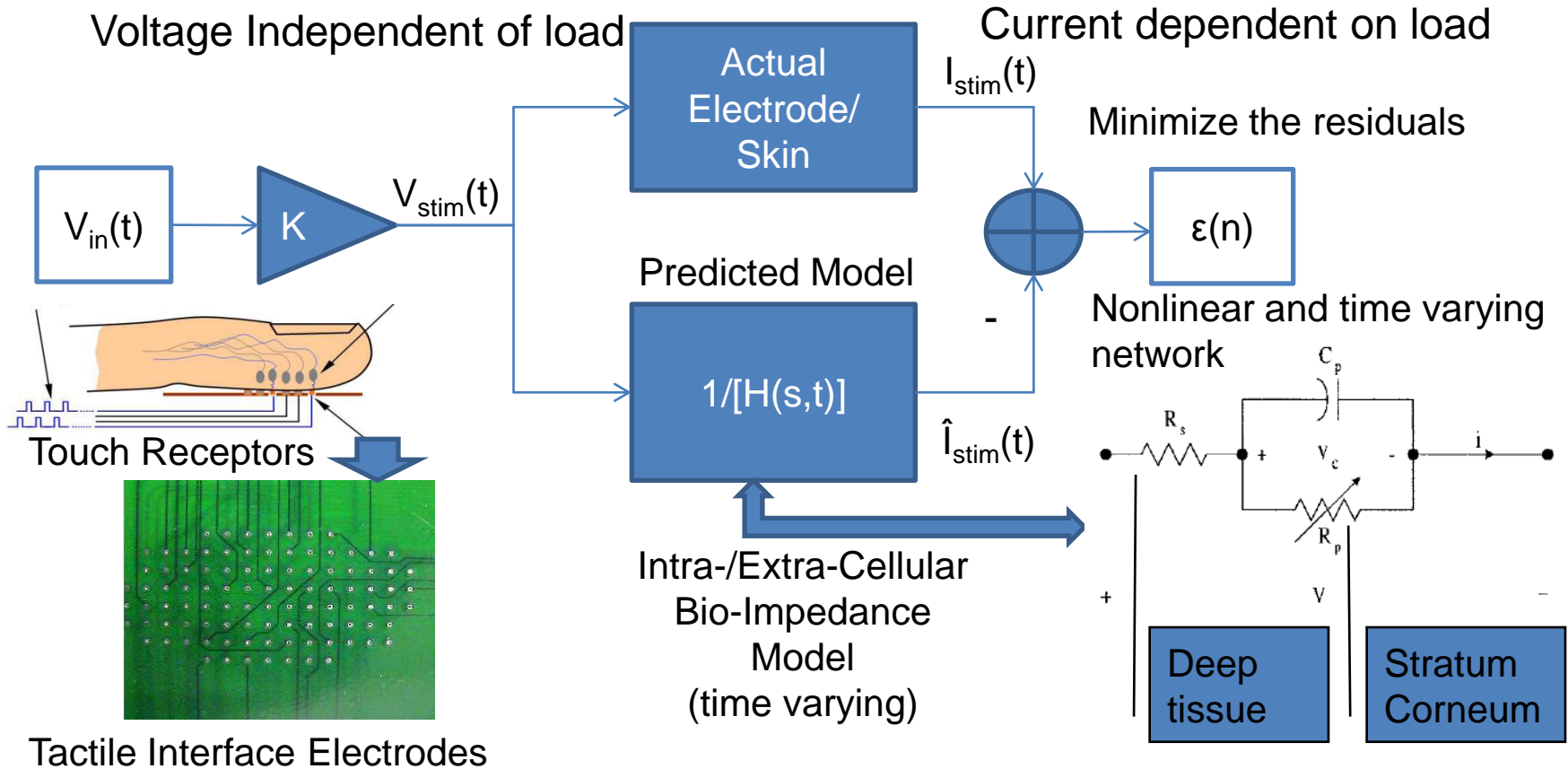
Remarks

Incorporate ease of design and efficiency of CVD introduce and identify parameters of human bio-impedance in real-time to set current levels

Key:

Blue: largest R_p
 Red: Smallest R_p
 Orange: Largest C_p
 Green: Smallest C_p

Overview: Using CVD with Parameter Identification to set stimulation current intelligently



Future Goal

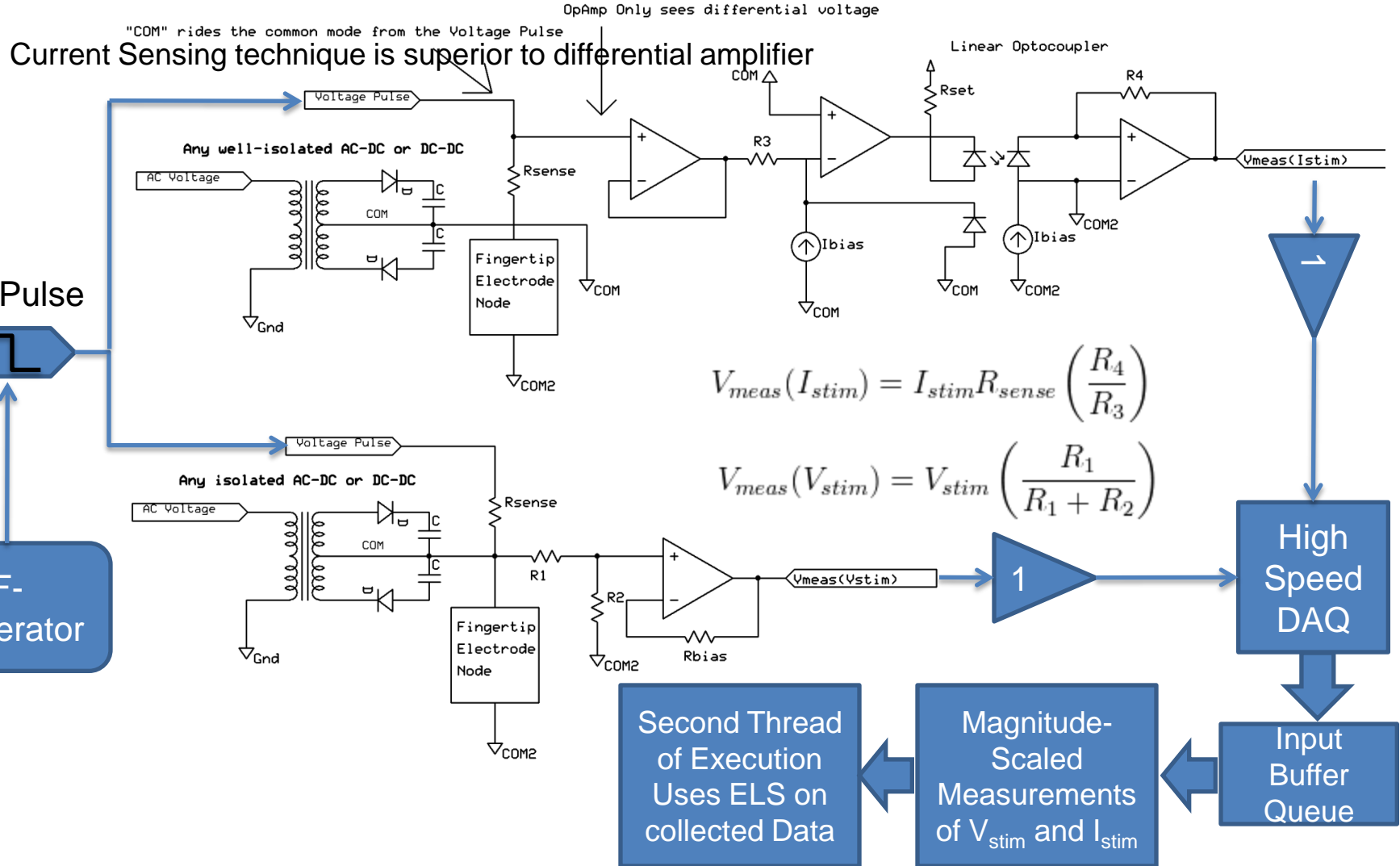
Model-Based Adaptive Control: use predicted $H(s,t)$ to set V_{in} to match set I_{stim} for user

➤ $H(s,t)$ is the Q-point-linearized, time varying finger impedance model calculated by iterative extended L.S. (with forgetting factor λ)

➤ Measurements of $V_{stim}(t)$ and $I_{stim}(t)$ are acquired from tactile interface electrodes to get $\hat{I}_{stim}(t)$

Driving & Measuring the Voltage and Current: First Thread of Execution

Current Sensing by riding the Common Mode Voltage using Isolation Transformer
Voltage Divider (high-Z) w/ high bandwidth precision opAmp



Real Time Identification Algorithm: Second Thread of Execution

What we want Identified (at an operating point)

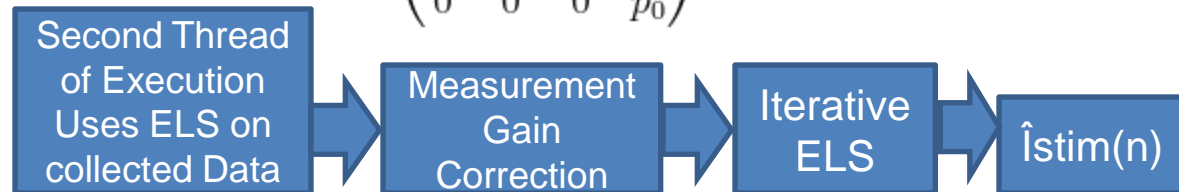
$$\hat{I}_{stim}(t) = \frac{\beta_2 + \beta_1 z^{-1}}{1 + \alpha_1 z^{-1}} V_{stim}(t) + \frac{c_1}{1 + \alpha_1 z^{-1}} e(t)$$

$$\hat{\theta}(t) = (\alpha_1, \beta_1, \beta_2, c_1)^T$$

Initial Conditions:

$$P_0 = \begin{pmatrix} p_0 & 0 & 0 & 0 \\ 0 & p_0 & 0 & 0 \\ 0 & 0 & p_0 & 0 \\ 0 & 0 & 0 & p_0 \end{pmatrix} > 0 \quad \hat{\theta}_0 = (0, 0, 0, 0)^T$$

Adjustment Model Algorithm (ELS)



For, $t = 1 \dots k$, where k is a subset of the input buffer queue length for storing $V_{meas}(V_{stim})$ and $V_{meas}(I_{stim})$

$$P(t) = \frac{1}{\lambda} P(t-1) [1 - \varphi(t-1) \varphi^T(t-1) P(t-1)] [\lambda + \varphi^T(t-1) P(t-1) \varphi(t-1)]^{-1}$$

$$K(t) = P(t) \varphi(t-1) \quad \epsilon_a(t) = I_{stim}(t) - \hat{I}_{stim}(t)$$

$$\hat{\Theta}(t) = \hat{\Theta}(t-1) + K(t) \epsilon_a(t)$$

MSE goodness of fit criteria

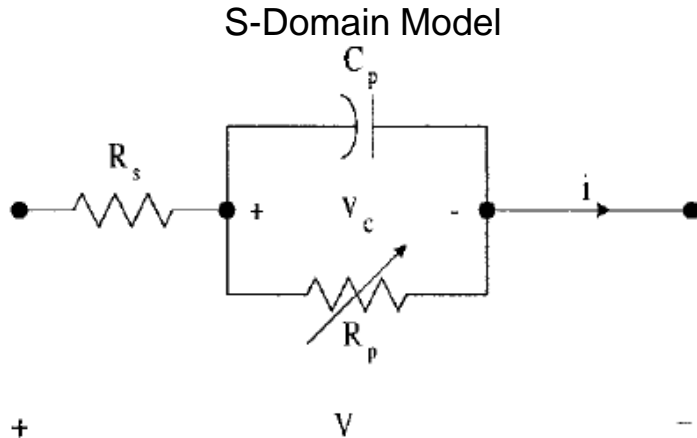
$$E_{n+1}^2 = (\epsilon_{a(n)}^2 + E_n^2 \text{mod}(n, N)) (\text{mod}(n, N) + 1)^{-1}$$

Repeat for $n = j*k + 1 \dots (j+1)*k$, for $j=1, \dots$ (user stop)

Note: forgetting factor weight λ is included in the ELS minimization to account for the time varying $R_p(t)$

$$N \approx \frac{2}{1 - \lambda}$$

Extracting Physical Model from Z-Domain Model



$$\hat{Z}_A(t) = \frac{R_s + R_p(t) + R_p(t)R_s C_p s}{1 + R_p(t)C_p s}$$

$$V_{stim} \gg e(t) \Rightarrow \frac{1}{\hat{Z}_D(z)} \approx \frac{\hat{I}_{stim}}{V_{stim}} = \frac{\beta_2 + \beta_1 z^{-1}}{1 + \alpha_1 z^{-1}}$$

We use the Tustin Transform to take the above identified model in to the S domain:

$$z \approx \frac{1 + sT_s/2}{1 - sT_s/2}$$

Assuming:

- no frequency warping from sufficiently high sampling rate.
- First corner frequency \ll zero

$$R_p(t)R_s C_p \gg R_p(t)C_p \Rightarrow \omega_c = \frac{1}{R_p(t)C_p}$$

$$\omega_D \ll \frac{2}{T_s} \Rightarrow \omega_D \approx \omega_A$$

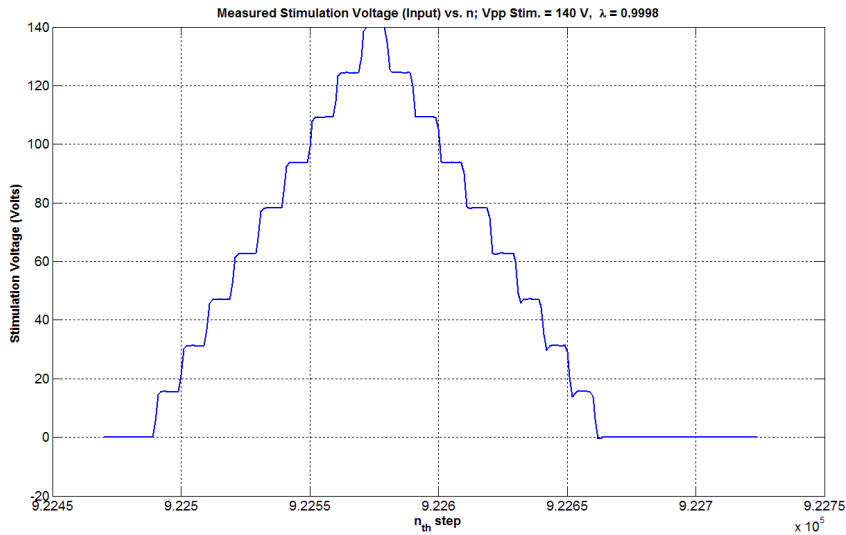
The physical bio-Impedance parameters can be Identified:

$$R_s = \frac{1 - \alpha_1}{\beta_2 - \beta_1}$$

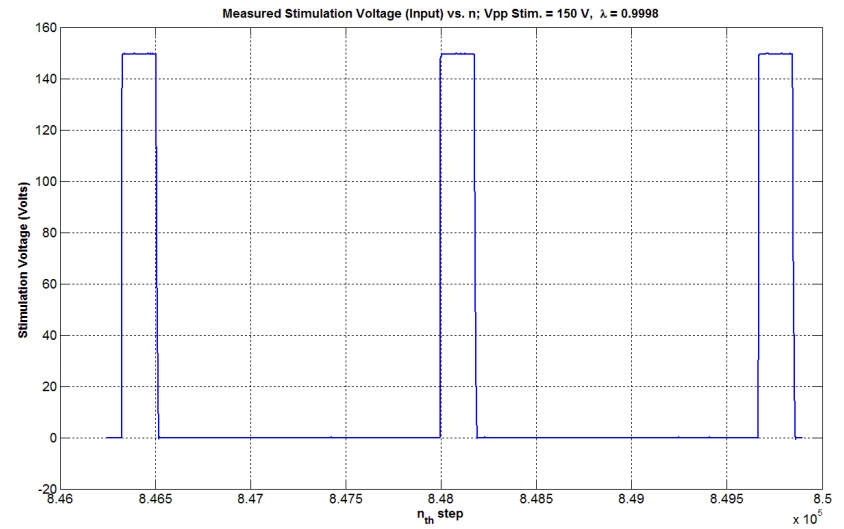
$$R_p = \frac{1 + \alpha_1}{\beta_1 + \beta_2} - R_s$$

$$C_p = \frac{T_s}{2} \left(\frac{\beta_2 - \beta_1}{\beta_2 + \beta_1} \right) \frac{1}{R_p(t)}$$

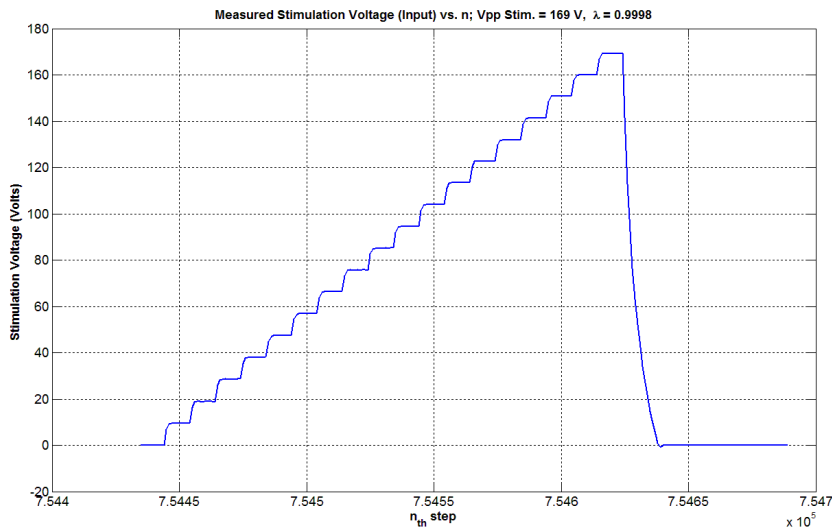
Stimulating Voltage Signal Primitives:



Triangle Waveform



Square Waveform



Saw Tooth Waveform

Configurations :Triangle, Saw Tooth or Square Wave Shape

V_{pp} : 50V – 250V

Frequency : 50 to 100 Hz

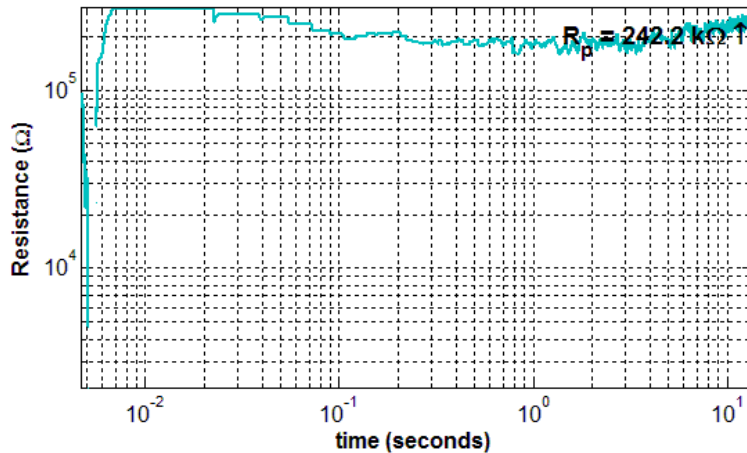
Duty Cycle : 5% - 11%

Features: Intra-Duty Cycle Modulation

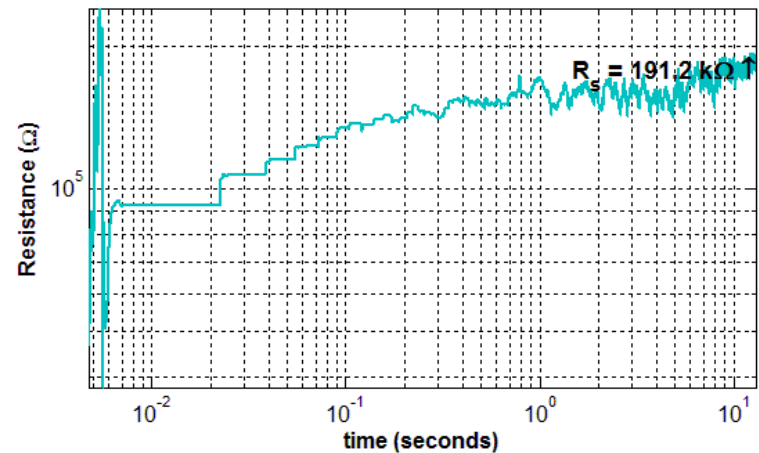
Used for both stimulation and identification

Results: ELS and Convergence (Typical)

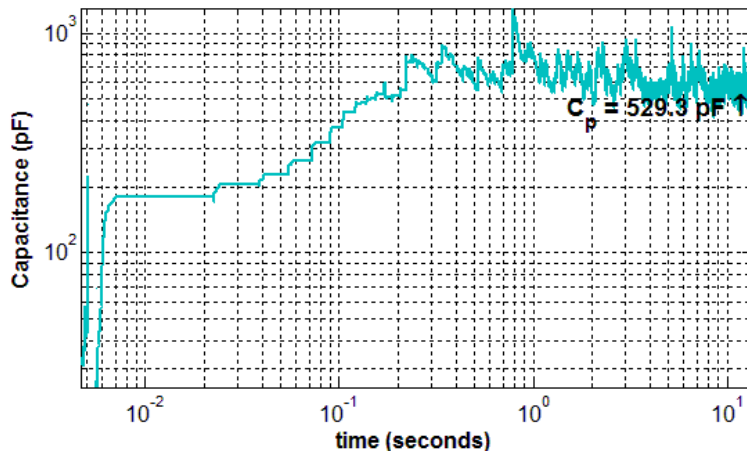
Extra-Cellular Resistance: R_p vs. Time; V_{pp} Stim. = 130 V, $\lambda = 0.9998$



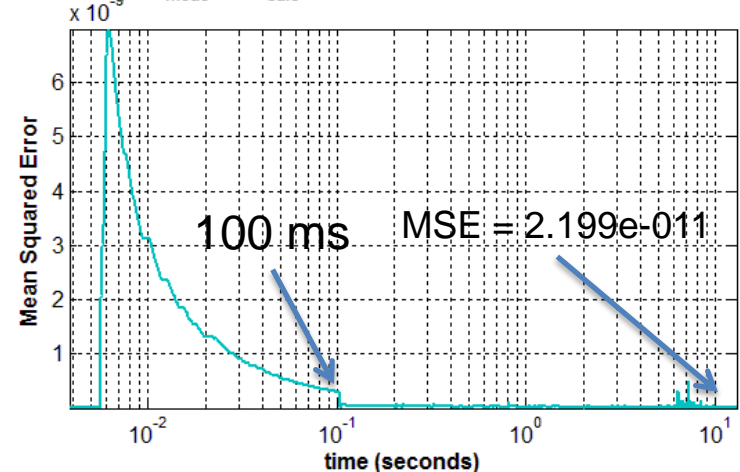
Intra-Cellular Resistance: R_s vs. Time; V_{pp} Stim. = 130 V, $\lambda = 0.9998$



Cellular Membrane Capacitance: C_p vs. Time; V_{pp} Stim. = 130 V, $\lambda = 0.9998$



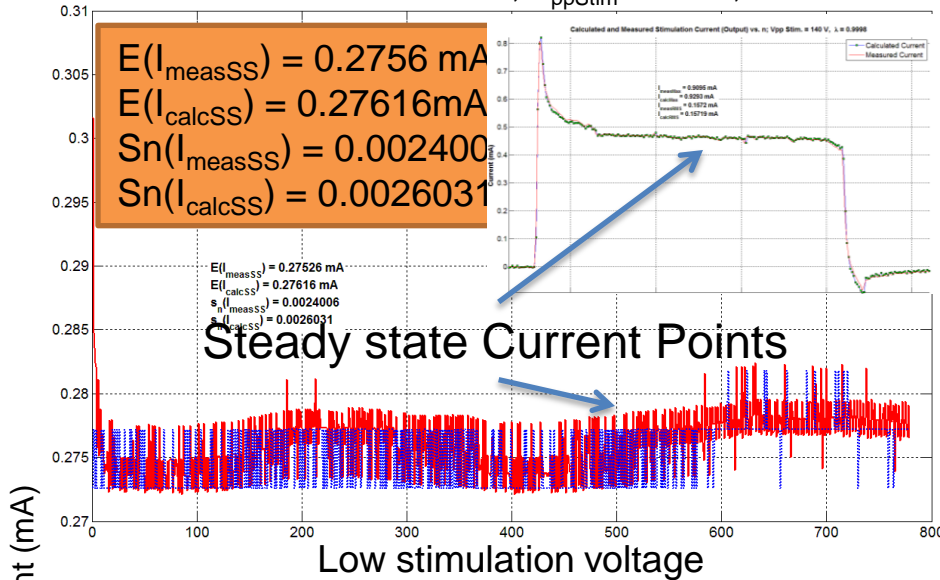
MSE: I_{meas} and I_{calc} vs. Time; V_{pp} Stim. = 130 V, $\lambda = 0.9998$



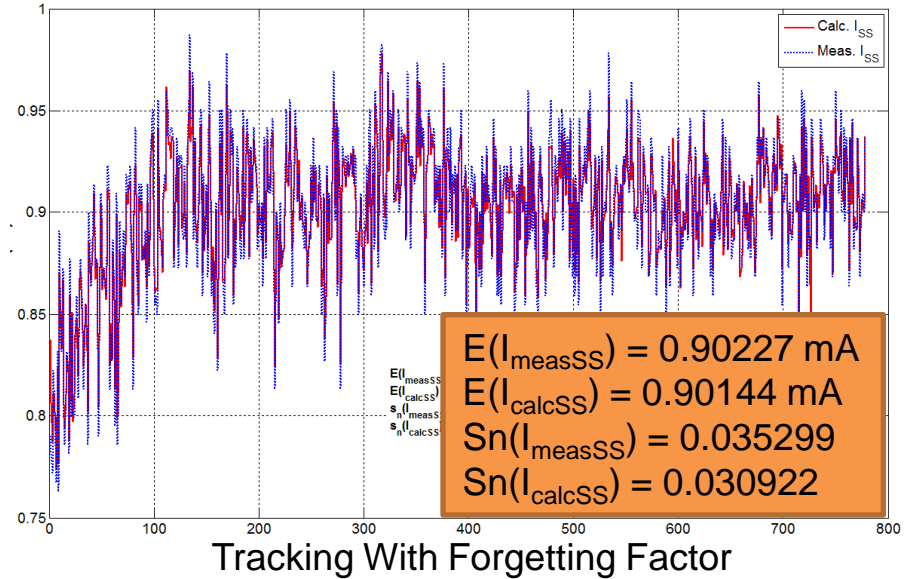
Subject: Male, **Age:** 31, **Finger Condition:** Dry, **Finger Tested:** Index; **Finger Location:** kept const.
Test Condition: One-Subject Z_L at 130 Volts; successive measurement (no time lapse); no repeatability testing; const. pressing force;
 Δ time: 0-13 sec. , **waveform:** square wave, no P.E., **D.C.** = 11.0%, **Freq.** = 60 Hz $\lambda = 0.9998$, **Pressing Force:** 700grams

Results: Algorithm's ability to track with time variation

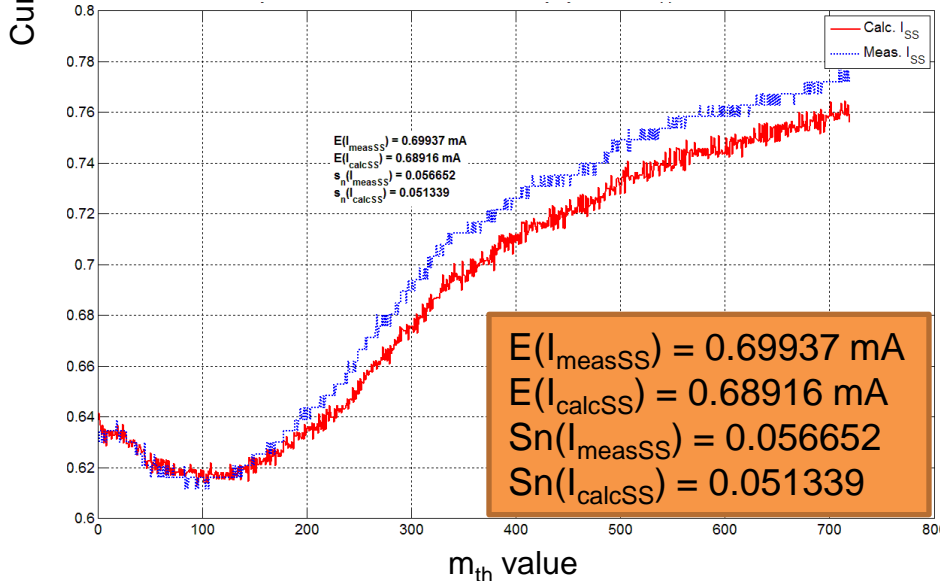
S.S Stim. Current all Active D.Cs; $V_{ppStim} = 110 \text{ V}$, $\lambda = 0.9998$



S.S Stim. Current all Active D.Cs; $V_{ppStim} = 160 \text{ V}$, $\lambda = 0.9998$



S.S Stim. Current all Active D.Cs; $V_{ppStim} = 150 \text{ V}$, $\lambda = 1$

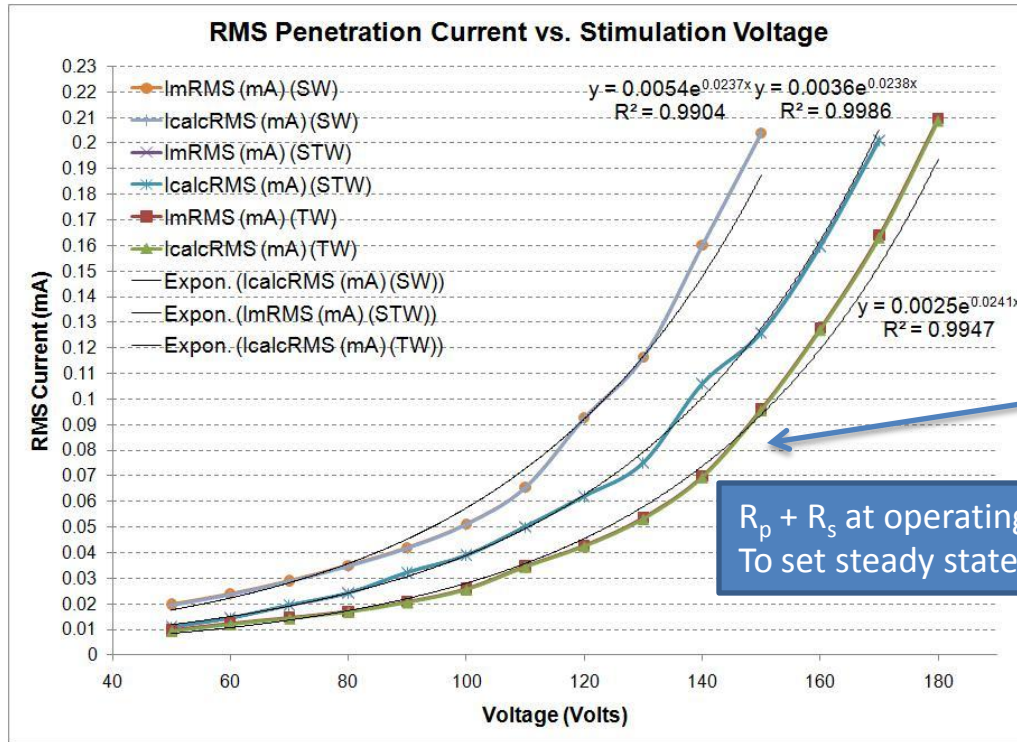


Tracking steady state I_{pp} of square wave voltage Stimulation from $t = 0 \dots 13 \text{ sec}$:

In General :

- The standard deviation of measured steady state penetration current increases as stimulation voltage increases past a certain threshold
- requires fast tracking

Results: Identified R_p Dynamic Range



Preliminary Remarks:

The steady state skin penetration current can be set by the identified R_p (and R_s) on-line

This is supported by:

The ELS convergence time being less than the skins time variation (as seen in previous slide)

Algorithms ability to track the IV DC characteristic accurately

I-V DC characteristic with electrical skin impedance

Subject: Male, **Age:** 31, **Finger Condition:** Dry, **Finger Tested:** Index; **Finger Location:** kept const.

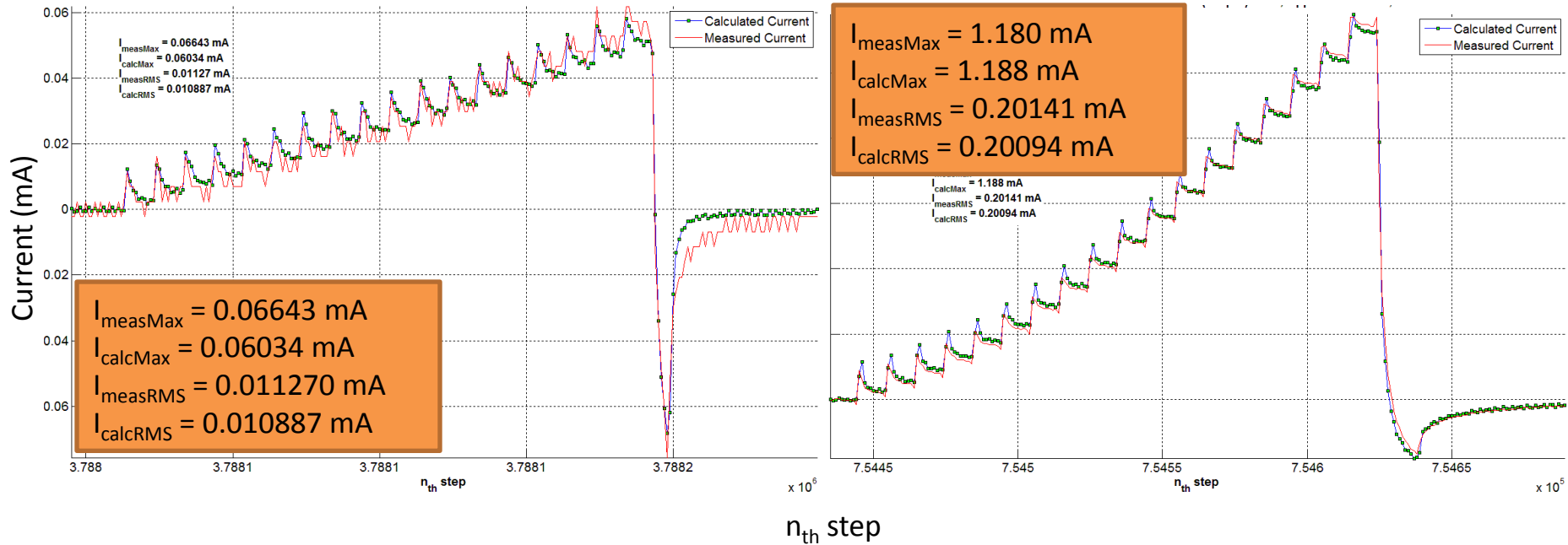
Test Condition: One-Subject Z_L at multiple Q-Points; successive measurement (no time lapse); no repeatability testing; const. pressing force;

Δtime: 3 sec. assumed steady state interval, **waveform:** all no P.E., **D.C.** = 11.0%, **Freq.** = 60 Hz λ = 0.9998,

Pressing Force: 300 grams

Results: Advantage using ELS with slow Stimulation Voltage rise times

Calculated and Measured Current (Output) vs. n ; $V_{ppStim} = 50$ V (left) and $V_{ppStim} = 169$ V (right), $\lambda = 0.9998$



Saw Tooth Wave:

$$\hat{I}_{stim}(t) = \frac{\beta_2 + \beta_1 z^{-1}}{1 + \alpha_1 z^{-1}} V_{stim}(t) + \frac{c_1}{1 + \alpha_1 z^{-1}} e(t)$$

- As driving voltage Increases a notable exponential increase in skin penetration current is observed as a function of quantization level.
- By using the Extended least Squares, the regressor c_1 (error correlation term) characterizes the unknown signal nonlinearities as to enhance tracking at each delta step

Conclusion & Future Work

- Built versatile CVD, measurement and output current tracking system
- Plan to develop a feedback loop to set I_{pp} current levels based on identified model
- Developed high bandwidth current measuring technique
- Demonstrated ELS algorithm with forgetting factor can track non-linear and time varying current in real-time
- Future work will use the completed haptic render to investigate mimicking tactile sensation
- Currently Developing 200 point electro-tactile stimulator using arrays of multiplexed CVDs