Wearables and Application Solutions for Parkinson's Disease: An Overview

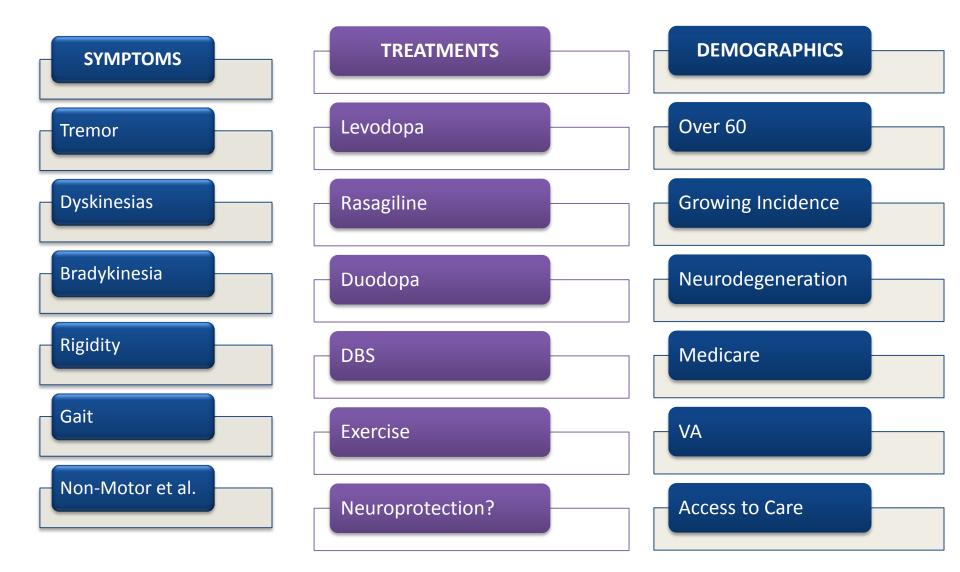


Joseph P. Giuffrida, PhD President & Principal Investigator

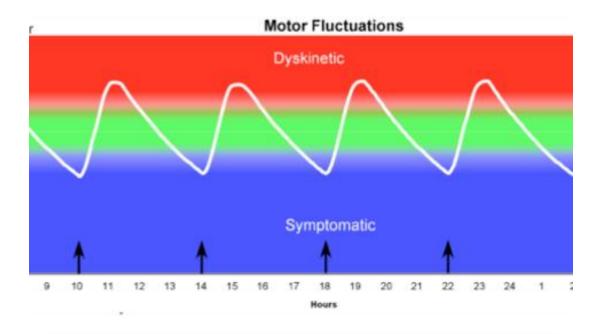
Parkinson's Disease: Challenges and Opportunities	5 min
Wearable Technology Space & Big Data	
Wearables Road Map	10 min
Clinical Validation of Wearables	
Targeted Applications within Targeted Applications	
Closing the Clinical Workflow	10 min
Closing the Business Case	
Closing the Patient Perspective	

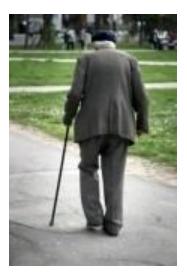
Parkinson's Disease: Challenges and Opportunities

Parkinson's is Very Difficult



Advanced Parkinson's





Wearable Technology Space & Big Data



Intel, Michael J. Fox and Big Data: Fighting Parkinson's Disease

By Virginia Backaitis | Aug 13, 2014 5 Follow 908 followers

Featured Webinar: Why You Can't Afford Your Homegrown CMS



You've been there. In the doctor's office that is. You're not feeling well and you want to tell the doc all about it, but he wants to ask you questions like: How would you rate the pain on a scale of 1 to 10? When did this start? How long does it last? How would you rate your sleep 1 to 10?

You answer the questions with what is, at best, a guess. And the doctor

makes assessments based upon your answers. But is what he calls an "8" the same thing you call an 8? And what does "sleeping well" actually mean? (And, yes, we know there's information like heart rate, blood pressure, lab work data to consider, but we're putting that aside for the moment.)

Now forget about yourself and think of a Parkinson's patient. Michael J. Fox or Intel's Andy Grove may be the ones we "know" best, unless there's someone in our personal lives who has been affected. Their doctors probably include some

Apple ResearchKit Turns iPhones Into Medical Diagnostic Devices

Posted Mar 9, 2015 by Josh Constine (@joshconstine)





Medical research is plagued by small sample sizes and inconsistent dat



Parkinson mPower study app

By Sage Bionetworks, a Not-For-Profit Research Organization

Open iTunes to buy and download apps.



Seller: Sage Bionetworks, a Not-For-Profit Research Organization © 2015, Sage Bionetworks

You must be at least 17 years old to download this app.

Medical/Treatment Information

Infrequent/Mild Alcohol,

Tobacco, or Drug Use or

References Frequent/Intense

Free

Category: Medical Released: Mar 09, 2015

Version: 1.0

Size: 69.6 MB Language: English

Description

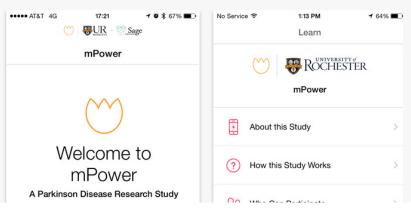
How can we together better manage the symptoms of Parkinson disease?

Living with Parkinson disease means coping with symptoms that change daily. Yet these daily changes are not

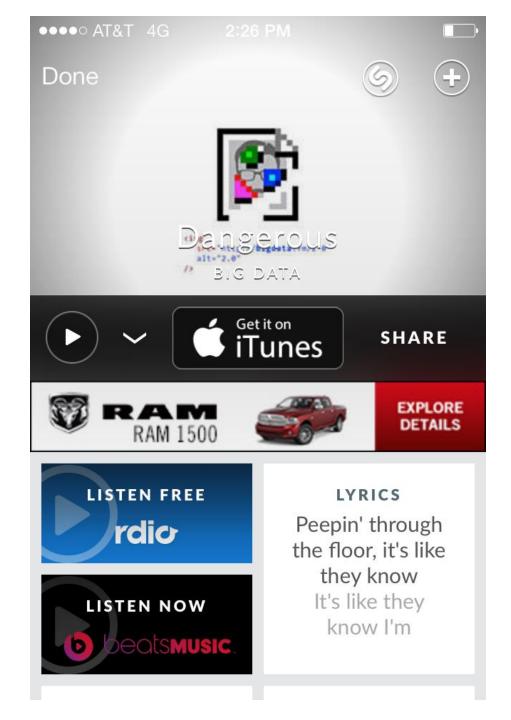
Sage Bionetworks, a Not-For-Profit Research Organization Web Site > Parkinson mPower studyMore app Support >

View More by This Developer

iPhone Screenshot







Can Parkinson's Wearables Learn from MS?

Biogen idec explores use of wearables to track MS patient activity

March 10, 2015 8:29 pm by Stephanie Baum | 10 Comments



As part of an effort by Biogen idec to explore ways to use wearables with MS patients to help physicians quantify patient activity, it recently completed a study of 250 patients in collaboration with PatientsLikeMe, Naomi Fried, vice president of medial information and innovation at Biogen, referenced the study as part of a keynote presentation on digital health at the MidAmerica Healthcare Venture Forum in Chicago this week.

This is the problem that Biogen idec wants to solve. Impaired mobility affects more than 90 percent of people with MS, but the quantified assessment of their walking ability tends to be limited to clinical settings. Sensors could give physicians a more accurate assessment of the level of activity of these patients if they were willing to wear activity trackers between appointments.

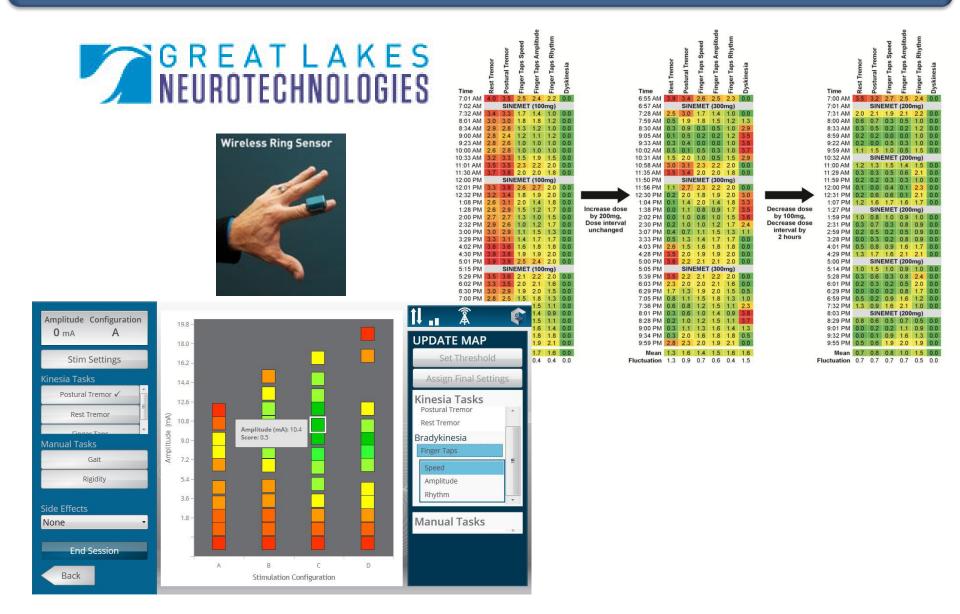
The study of 250 people with MS sought an answer to the questions: Would patients actually use wearables as part of their daily lives and be willing to share that information with physicians?

The initial takeaway from the study is that it needs to use devices with more sophisticated sensors to quantify movement accurately and consistently. "Current technology is not built to provide consistent and validated data in MS. We are early in the process, but hope to have progress in the coming months," according to an emailed statement about the study from Biogen. It also noted that it was encouraged by the "overwhelming positive participation" from MS patients. It took the response as a sign of encouragement as it explores using wearables in the future.

Big Enough Data



Agile Development: Systems, Algorithms, and Applications



Clinical Validation and Publications

NIH Funding

Over \$25 Million in SBIR Funding

- National Institute of Health, US
- State of Ohio Commercialization Programs

Collaborators

Over 30 Collaborating Institutions

- Clinical Testing
- Research Collaboration
- Commercialization Partners
- Controlled Database with over 1,000 Patients

Publications

Over 75 Peer-Reviewed & Presentations

- Tremor, Bradykinesia & Dyskinesia Assessment
- Parkinson's Telemedicine
- Deep Brain Stimulation Programming
- Clinical Trials

Quality and Regulatory

FDA Clearance to Market

- 510k Clearance to Market
 - Intended Use
 - Kinesia is intended to monitor physical motion and muscle activity to quantify kinematics of movement disorder symptoms such as tremor and assess activity in any instance where quantifiable analysis of motion and muscle activity is desired.

ISO, CE Mark, Health Canada, and TGA

- ISO 13485:2003
- European Medical Device Directive 93/42/EEC
- Canadian Medical Device Conformity Assessment System
- EMERGO EUROPE: Authorized Agent

Standards and Testing

- Tested to IEC 60601 Standards
- Complies with FCC Part 15 Rules
- HIPAA Compliant



Wearables Roadmap

Context



Environment



Hello My Name Is...

Would I Use Wearables?

I Want My Doctor to See...

A Road Map for Parkinson's Wearables



What Symptoms are You Trying to Measure?

Detection or Severity? Are Symptoms Voluntary or Involuntary?

Context of Daily Life?

Patient Environment and Confounding Factors?

Clinical Validation of Wearables

Clinical Validation Workflow

Start with Controlled Environment

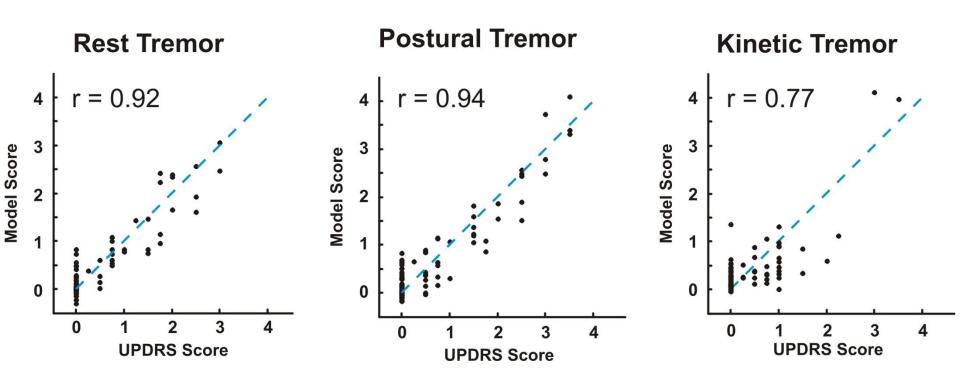
Include Broad Range of Severities

Compare Versus Traditional Gold Standard (Video)

Demonstrate Correlations, Sensitivity, and Test-Retest Reliability

Move to Unconstrained Environment and Tasks That Mask or Mimic Symptoms

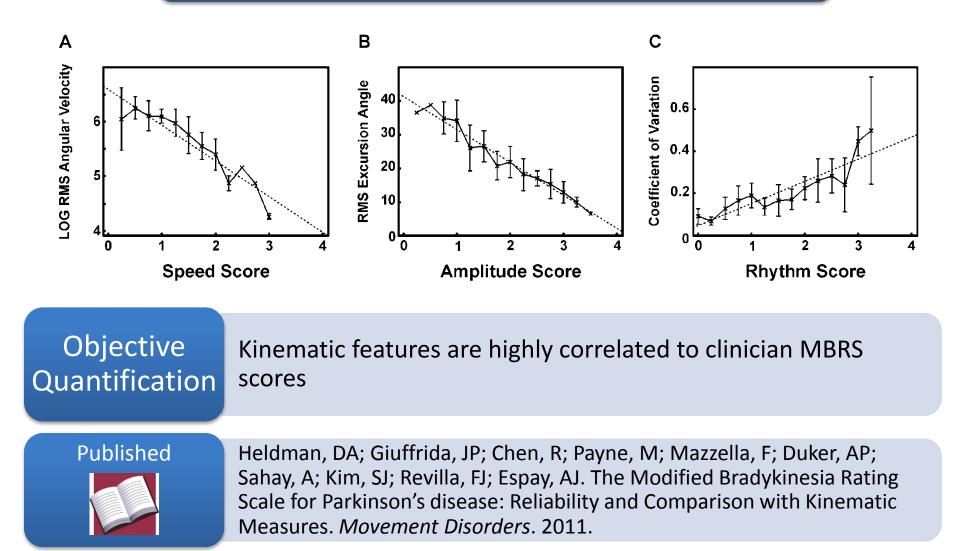
Discrete Tremor Assessment



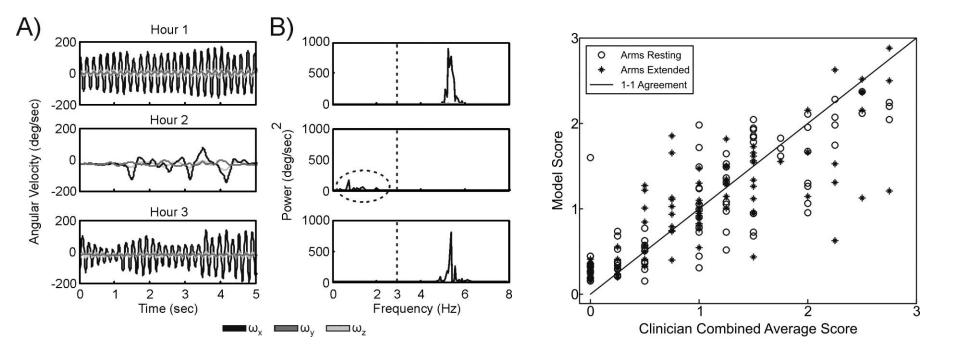


Giuffrida, J. P., Riley, D, Maddux, B, and Heldman, D.A. Clinically deployable Kinesia technology for automated tremor assessment. *Movement Disorders* 24 (5): 723-730, 2009.

Discrete Bradykinesia Assessment



Discrete Dyskinesia Assessment

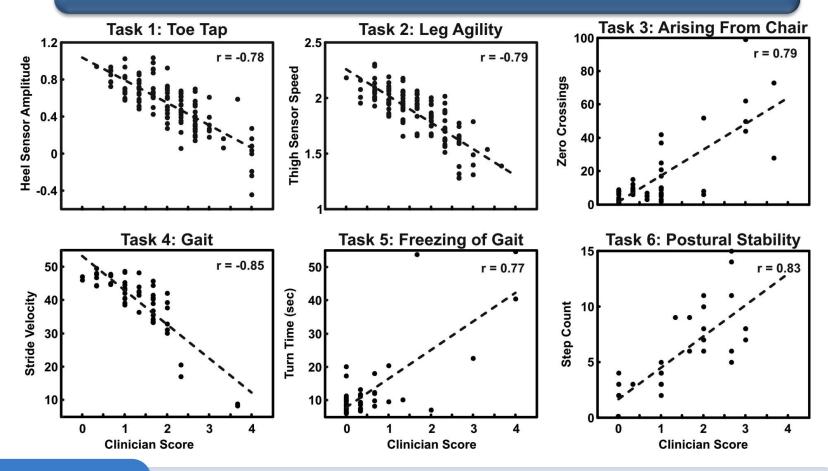


Published



Mera, TO, Burack, MA, and Giuffrida, JP. "Quantitative Assessment of Levodopa Induced Dyskinesia Using Automated Motion Sensing Technology", IEEE-EMBS Proceedings 2012.

Discrete Gait and Balance Assessment

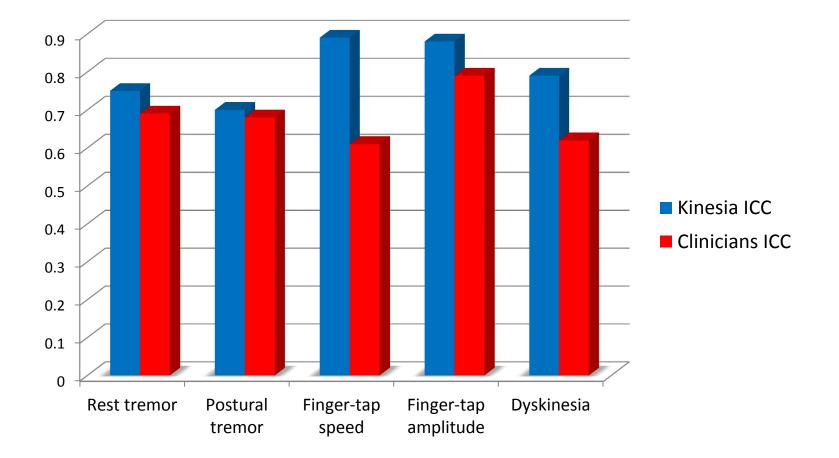


Recently Published

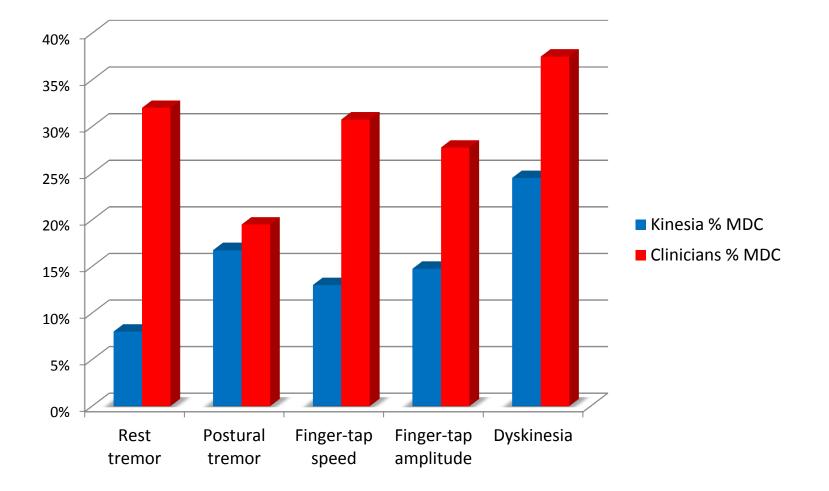


TO. Mera, DE. Filipkowski, DE. Riley, Christina M. Whitney, Benjamin L. Walter, Steven A. Gunzler, Joseph P. Giuffrida, "Quantitative analysis of gait and balance response to deep brain stimulation in Parkinson's disease "<u>Gait & Posture</u> - 07 December 2012

Test-Retest Reliability: Intra Class Correlations



Sensitivity: Minimum Detectable Changes, % of Full Scale



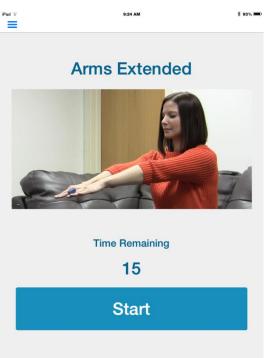
Targeted Applications

Targeted Applications

Objective Sensor



Mobile App



Web Portal and Reports

Time	Rest Tremor	Postural Tremor	Finger Taps Speed	Finger Taps Amplitude	Finger Taps Rhythm	Dyskinesia
6:55 AM	3.9	3.4	2.6	2.5	2.3	0.0
6:57 AM			EMET	(300		64.005
7:28 AM	2.5	3.0	1.7	1.4	1.0	0.0
7:59 AM	0.5	1.9	1.8	1.5	1.2	1.3
8:30 AM	0.3	0.9	0.3	0.5	1.0	2.9
9:05 AM	0.1	0.5	0.2	0.2	1.2	3.5
9:33 AM	0.3	0.4	0.0	0.0	1.0	3.8
10:02 AM	0.5	0.1	0.5	0.3	1.0	3.7
10:31 AM	1.5	2.0	1,0	0.5	1.5	2.9
10:58 AM	3.0	3.1	2.3	2.2	2.0	0.0
11:35 AM	35	3.4	2.0	2.0	1.8	0.0
11:50 PM		SIN	EMET	(300	mg)	Same
11:56 PM	1.1	2.7	2.3	2.2	2.0	0.0
12:30 PM	0.2	2.0	1.8	1.9	2.0	3.0
1:04 PM	0.1	1.4	2.0	1.4	1.8	3.0 3.3 3.5 3.6
1:38 PM	0.0	1.1	0.8	0.9	1.7	3.5
2:02 PM	0.0	1.0	0,6	1.0	1.5	
2:30 PM	0.2	1.0	1.0	1.2	1.7	2.4
3:07 PM	0.4	0.7	1.1	1.5	1.3	1.1
3:33 PM	0.5	1.3	1.4	1.7	1.7	0.0
4:03 PM	2.6	1.5	1.6	1.8	1.8	0.0
4:28 PM	3.5	2.0	1.9	1.9	2.0	0.0
5:00 PM	3.8	2.2	2.1	2.1	2.0	0.0



Task-Based Motor Assessments

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Time	Rest Tremor	Postural Tremor	Finger Taps Speed	Finger Taps Amplitude	Finger Taps Rhythm	Dyskinesia	
7:01 AM	4.0	3.5	2.5	2.4	2.2	0.0	
7:02 AM				(100		0.0	
7:32 AM	3.4	3.3	1.7	1.4	1.0	0.0	
8:01 AM	3.0	3.0	1.8	1.8	1.2	0.0	
8:34 AM	2.9	2.8	1.3	1.2	1.0	0.0	
9:00 AM	2.8	2.4	1.2	1.1	1.2	0.0	
9:23 AM	2.8	2.6	1.0	1.0	1.0	0.0	
10:00 AM	2.6	2.8	1.0	1.0	1.0	0.0	2
10:33 AM	3.2	3.3	1.5	1.9	1.5	0.0	
11:01 AM	3.5	3.5	2.3	2.2	2.0	0.0	
11:30 AM	3.1	3.8	2.0	2.0	1.8	0.0	
12:00 PM	2.2	-	EMET	STATISTICS.	and the second second	0.0	
12:01 PM	3.3	3.8	2.6	2.7	2.0	0.0	
12:32 PM	3.2	3.4	1.8	1.9	2.0	0.0	
1:08 PM	2.6	3.1	2.0	1.4	1.8	0.0	
1:28 PM	2.6	2.9	1.5	1.2	1.7	0.0	
2:00 PM	2.7	2.7	1.3	1.0	1.5	0.0	
2:32 PM	2.9	2.6	1.0	1.2	1.7	0.0	
3:00 PM	3.0	2.9	1.1	1.5	1.3	0.0	
3:29 PM	3.3	3.1	1.4	1.7	1.7	0.0	
4:02 PM	3.8	3.6	1.6	1.8	1.8	0.0	
4:30 PM	3.9	3.8	1.9	1.9	2.0	0.0	
5:01 PM	3.9	3.9	2.5	2.4	2.0	0.0	
5:15 PM	3.5		EMET		State of the second second	0.0	
5:29 PM	3.5	3.6	2.1	2.2	2.0	0.0	
6:02 PM	3.3	3.5	2.0	2.1	1.6	0.0	
6:30 PM	3.0	2.9	1.9	2.0	1.5	0.0	
7:00 PM 7:33 PM	2.8	2.5	1.5	1.8	1.3	0.0	
(2) 2.344 (2017) 2.01 (2017) 2.01	2.6	2.6	1.2	1.5	1.1	0.0	
8:04 PM	2.6	2.6	1.0	1.4	0.9	0.0	
8:30 PM	2.9	2.8	1.2	1.5	1.1	0.0	
9:02 PM	3.3	3.2	1.3	1.6	1.4	0.0	
9:33 PM	3.5	3.6	1.6	1.8	1.8	0.0	
10:00 PM		CONTRACTOR OF	2.0	1.9	2.1	0.0	
Mean	3.2	3.2	1.6	1.7	1.6	0.0	
Fluctuation	0.4	0.5	0.5	0.4	0.4	0.0	

Increase dose

by 200mg,

Dose interval

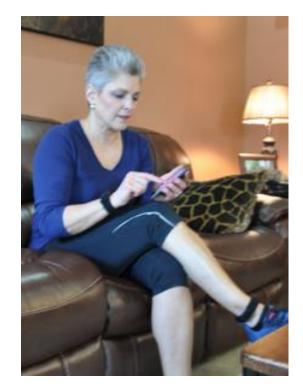
unchanged

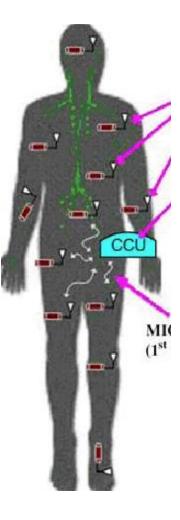
	Time	Rest Tremor	Postural Tremor	Finger Taps Speed	Finger Taps Amplitude	Finger Taps Rhythm	Dyskinesia	
	6:55 AM	3.9	3.4	2.6	2.5	2.3	0.0	
	6:57 AM					mg)		
	7:28 AM	2.5	3.0	1.7	1.4	1.0	0.0	
	7:59 AM	0.5	1.9	1.8	1.5	1.2	1.3	
	8:30 AM	0.3	0.9	0.3	0.5	1.0	2.9	
	9:05 AM	0.1	0.5	0.2	0.2	1.2	3.5	
	9:33 AM	0.3	0.4	0.0	0.0	1.0	3.8	
	10:02 AM	0.5	0.1	0.5	0.3	1.0	3.7	
	10:31 AM	1.5	2.0	1.0	0.5	1.5	2.9	
	10:58 AM	3.0	3.1	2.3	2.2	2.0	0.0	
	11:35 AM	3.5	3.4	2.0	2.0	1.8	0.0	
	11:50 PM		SIN	EMET	(300			
	11:56 PM	1.1	2.7	2.3	2.2	2.0	0.0	
	12:30 PM	0.2	2.0	1.8	1.9	2.0	3.0	
	1:04 PM	0.1	1.4	2.0	1.4	1.8	3.3	
e	1:38 PM	0.0	1.1	0.8	0.9	1.7	3.5	
10	2:02 PM	0.0	1.0	0.6	1.0	1.5	3.6	
L	2:30 PM	0.2	1.0	1.0	1.2	1.7	2.4	
	3:07 PM	0.4	0.7	1.1	1.5	1.3	1.1	
	3:33 PM	0.5	1.3	1.4	1.7	1.7	0.0	
	4:03 PM	2.6	1.5	1.6	1.8	1.8	0.0	
	4:28 PM	3.5	2.0	1.9	1.9	2.0	0.0	
	5:00 PM	3.8	2.2	2.1	2.1	2.0	0.0	
	5:05 PM		SINI	EMET				
	5:39 PM	3.5	2.2	2.1	2.2	2.0	0.0	
	6:03 PM	2.3	2.0	2.0	2.1	1.6	0.0	
	6:29 PM	1.7	1.3	1.9	2.0	1.5	0.5	
	7:05 PM	0.8	1.1	1.5	1.8	1.3	1.0	
	7:36 PM	0.6	0.8	1.2	1.5	1.1	2.3	
	8:01 PM	0.3	0.6	1.0	1.4	0.9	3.8	
	8:28 PM	0.2	1.0	1.2	1.5	1.1	3.7	
	9:00 PM	0.3	1.1	1.3	1.6	1.4	1.3	
	9:34 PM	0.3	2.0	1.6	1.8	1.8	0.5	
	9:59 PM	2.8	2.3	2.0	1.9	2.1	0.0	
	Mean	1.3	1.6	1.4	1.5	1.6	1.6	
F	luctuation	1.3	0.9	0.7	0.6	0.4	1.5	
r	nuctuation	1.5	0.5	0.1	0.0	0.4	1.5	

	Time	Rest Tremor	Postural Tremor	Finger Taps Speed	Finger Taps Amplitude	Finger Taps Rhythm	Dyskinesia
	7:00 AM	3.5	3.2	2.7	2.5	2.4	0.0
	7:01 AM			EMET			
	7:31 AM	2.0	2.1	1.9	2.1	2.2	0.0
	8:00 AM	0.6	0.7	0.3	0.5	1.0	0.0
	8:33 AM	0.3	0.5	0.2	0.2	1.2	0.0
	8:59 AM	0.2	0.2	0.0	0.0	1.0	0.0
	9:22 AM	0.2	0.0	0.5	0.3	1.0	0.0
	9:59 AM	1.1	1.5	1.0	0.5	1.5	0.0
	10:32 AM	1.0		EMET	and the second second		
	11:00 AM	1.2	1.3	1.5	1.4	1.5	0.0
	11:29 AM	0.3	0.3	0.5	0.6	2.1	0.0
	11:59 PM 12:00 PM	0.2	0.2	0.3	0.3	1.0	0.0
	12:00 PM	0.1	0.0	0.4	0.1	2.3	0.0
	1:07 PM	0.2	0.6	0.6	0.1	1.7	0.0
Decrease dose	1:27 PM	1.2	1.6	1.7 EMET	1.6		0.0
by 100mg,	1:59 PM	1.0	0.8	1.0	0.9	1.0	0.0
Decrease dose	2:31 PM	0.3	0.7	0.3	0.8	0.9	0.0
interval by	2:59 PM	0.2	0.5	0.2	0.5	0.9	0.0
2 hours	3:28 PM	0.0	0.3	0.2	0.8	0.9	0.0
	4:01 PM	0.5	0.8	0.9	1.6	1.7	0.0
	4:29 PM	1.3	1.7	1.6	2.1	2.1	0.0
	5:00 PM		mg)				
	5:14 PM	1.0	1.5	1.0	0.9	1.0	0.0
	5:28 PM	0.3	0.6	0.3	0.8	2.4	0.0
	6:01 PM	0.2	0.3	0.2	0.5	2.0	0.0
	6:29 PM	0.0	0.0	0.2	0.8	1.7	0.0
	6:59 PM	0.5	0.2	0.9	1.6	1.2	0.0
	7:32 PM	1.3	0.9	1.6	2.1	1.0	0.0
	8:03 PM	SINEMET (200mg)				mg)	
	8:29 PM	0.8	0.6	0.5	0.7	0.5	0.0
	9:01 PM	0.0	0.2	0.2	1.1	0.9	0.0
	9:32 PM	0.0	0.1	0.9	1.6	1.3	0.0
	9:55 PM	0.5	0.6	1.9	2.0	1.9	0.0
	Mean	0.7	0.8	0.8	1.0	1.5	0.0
	Fluctuation	0.7	0.7	0.7	0.7	0.5	0.0

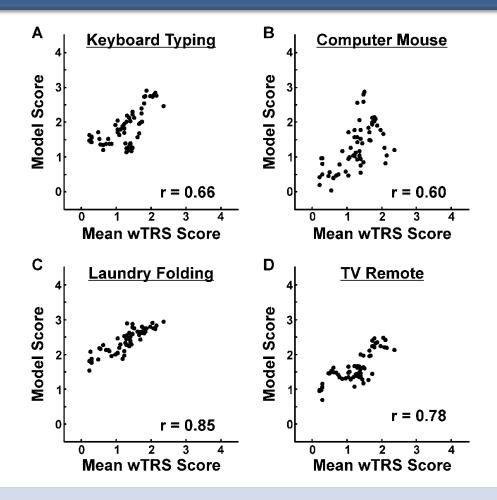
Sensitivity of Data Versus Patient Burden







Continuous Tremor Assessment

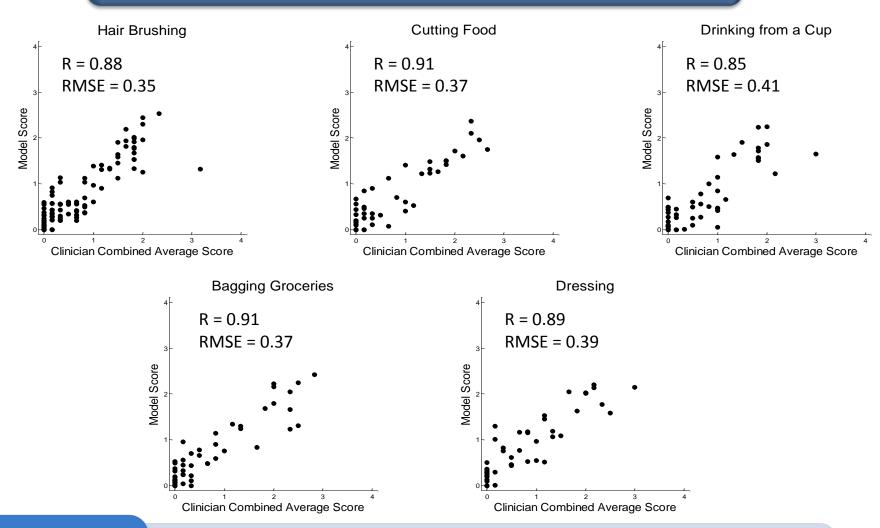


Published



D. A. Heldman, J. Jankovic, D. E. Vaillancourt, J. Prodoehl, R. J. Elble, and J. P. Giuffrida. Essential tremor quantification during activities of daily living. *Parkinsonism & Related Disorders*, 2011.

Continuous Dyskinesia Assessment

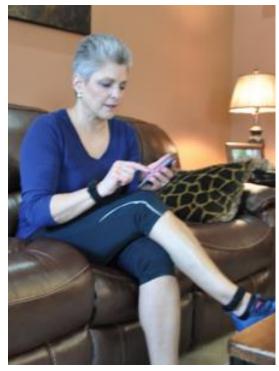


Published



Christopher L. Pulliam, Michelle A. Burack, Dustin A. Heldman, Joseph P. Giuffrida and Thomas O. Mera. Motion Sensor Dyskinesia Assessment During Activities of Daily Living. *Journal of Parkinson's Disease 2014*.

Wearable Sensors

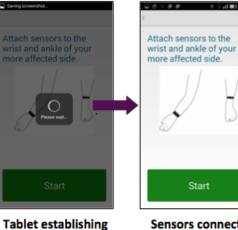






Sensor bands on charge pad, press Next

Mobile App



connection

Sensors connected. Press Start

8 ⊊ al ∎0 400



Continuous Parkinson's Monitoring

Summary Data

Total Wear Time: 9:52 HR

Tremor Summary Data

	Time	Percent Wear Time
Total Time TremorDetect=0	2:46	28
Total Time TremorDetect=1	7:06	72
Time Rnd (TremorScore) = 0	2:46	28
Time Rnd (TremorScore) = 1	1:42	17.2
Time Rnd (TremorScore) = 2	5:24	54.7
Time Rnd (TremorScore) = 3	0:00	0
Time Rnd (TremorScore) = 4	0:00	0
Average (Tremor Score)	0.88	
Standard Dev (Tremor Score)	0.69	

Dyskinesia Data Summary

	Time	Percent Wear Time
Total Time DysDetect=0	8:20	84.5
Total Time DysDetect=1	1:32	15.5
Time Rnd (DysProb) = 0	7:42	78
Time Rnd (DysProb) = .5	0:38	6.4
Time Rnd (DysProb) = 1	1:32	15.5

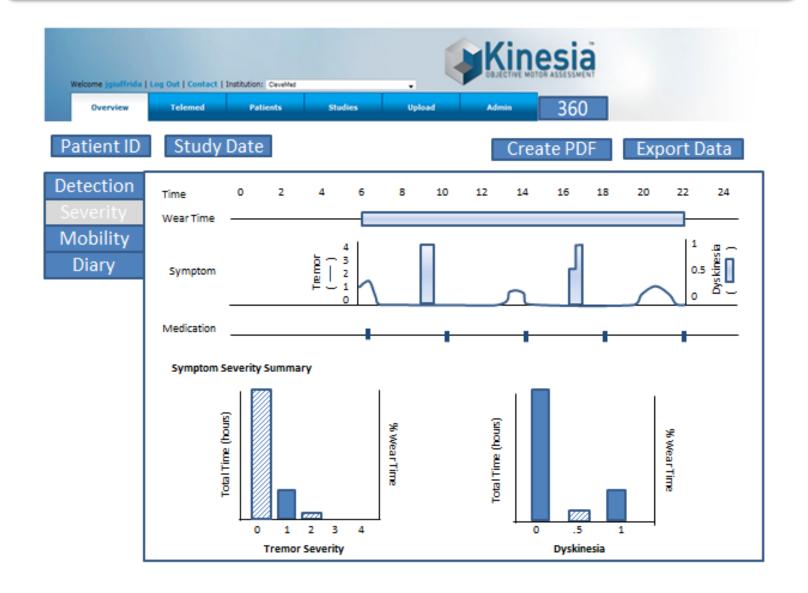
Mobility Data Summary

	Time	Percent Wear Time
Total Time Rest	0:25	4.3
Total Time Active (Non-Gait)	2:14	22.7
Total Time Gait	7:12	73
Total Active Time	9:26	95.7
Wear Time Steps	549	
% Arm Swing During Gait	92.1	

Subject Reported Diary

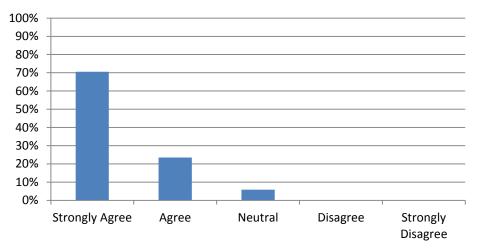
	Time	Percent Day	Percent Waking Day
Sleeping	9:30	39.6	
OFF	8:30	35.4	58.6
ON	1:30	6.3	10.3
ON with Non-Troublesome Dys	2:00	8.3	13.8
ON with Troublesome Dys	2:30	10.4	17.2

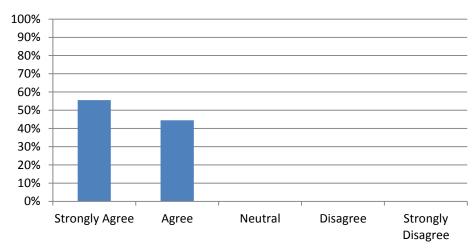
Continuous Parkinson's Monitoring



Patient Perspective

The sensor was easy to put on

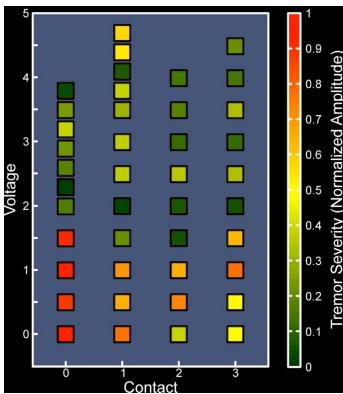


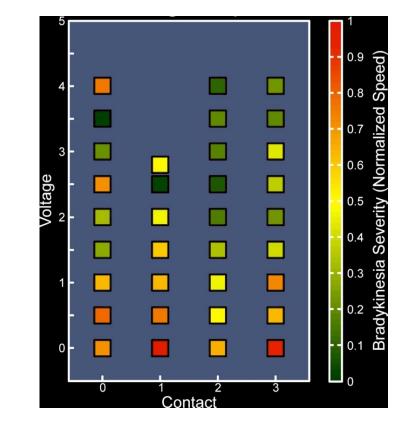


The sensor is comfortable

Deep Brain Stimulation: Outpatient Programming



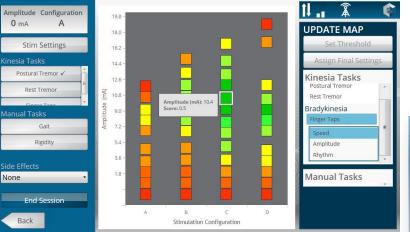




Objective Sensor



DBS Programming Map



Web Portal

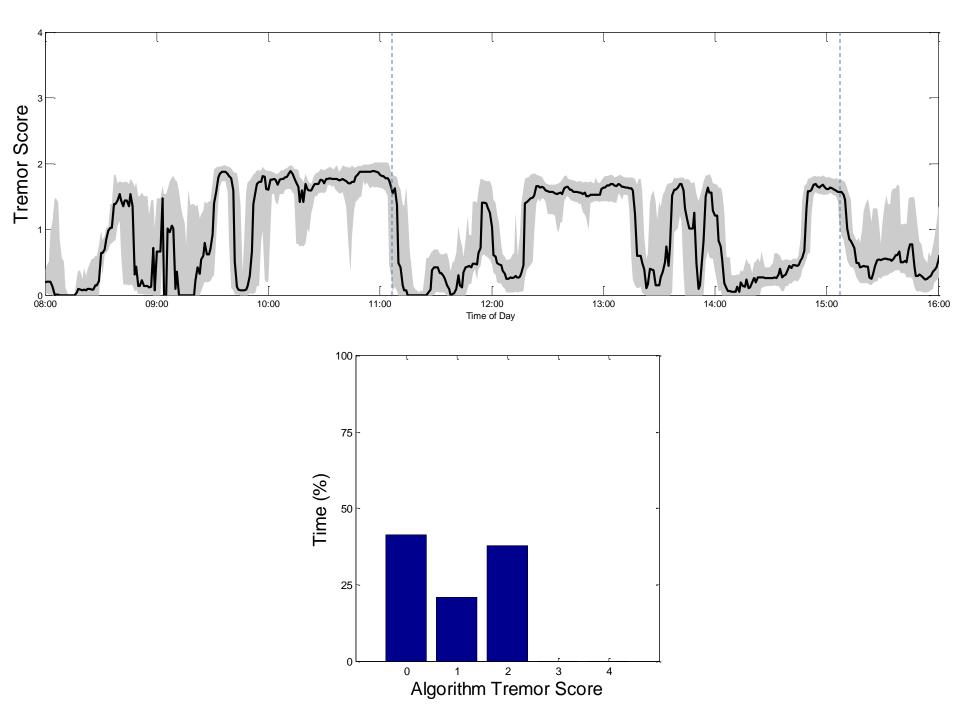
Change the way you see PD

Kinesia HomeView

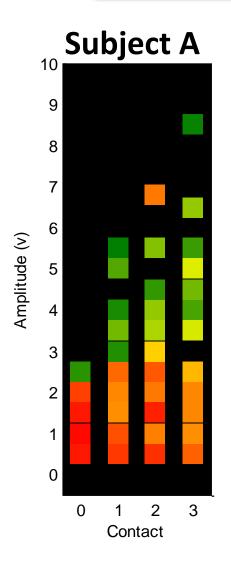
- CAR	Crea Stud			
🖥 Setup Study 🌅 Pending Upload 🌅	Scoring In Process 🗖 New i	teport Ready <mark></mark> Archived R	eport	Results per page: 10 Y
Status	Start Date	Last Name	First Name	Last Hodified 🐨
Pending Upload	04/06/2011	Doe	John	04/06/2011 7:54:29 AM
Archived Report		Smith	Jim	02/04/2011 3:19:46 PM
Archived Report		Smith	Jim	02/04/2011 3:19:40 PM
Archived Report	02/28/2012	Smith	Jim	10/07/2010 11:39:47 AM
Archived Report	10/07/2010	Doe	John	10/07/2010 11:36:08 AM
Archived Report		James	Georgia	10/07/2010 11:21:11 AM
Archived Report		Smith	Jim	10/07/2010 11:07:16 AM
Archived Report		Doe	John	10/07/2010 11:02:35 AM
Pending Upload	10/07/2010	Doe	John	10/07/2010 10:20:44 AM
Archived Report	10/09/2011	James	Georgia	10/07/2010 9:43:00 AM

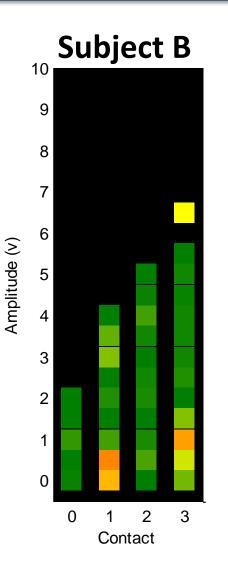


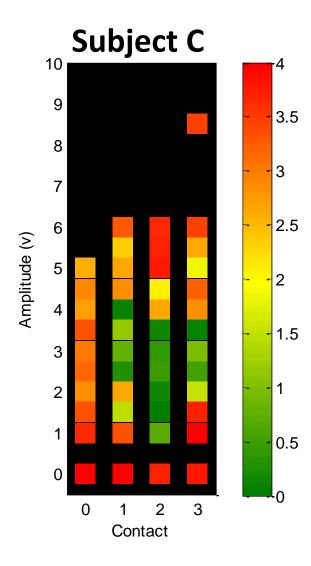
Closing The Clinical Workflow



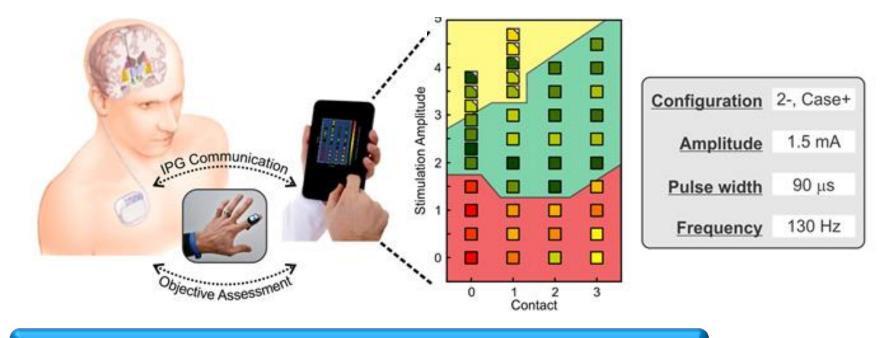
Rest Tremor DBS Tuning Maps







Potential Solution for DBS Programming

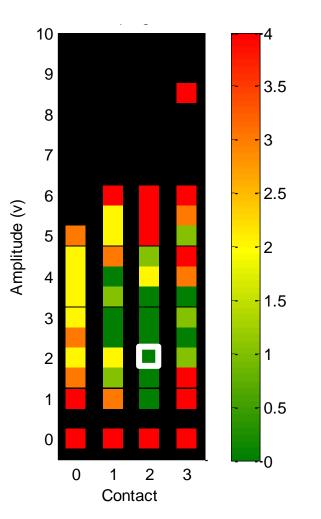


Can a Computerized Sensing System...

- 1. Find DBS Settings that Improve Motor Outcomes Compared to Clinicians?
- 2. Find DBS Settings that Provide Similar Motor Outcomes at Lower Amplitude?
- 3. Automatically Guide a Programming Session to Improve Motor Outcomes?

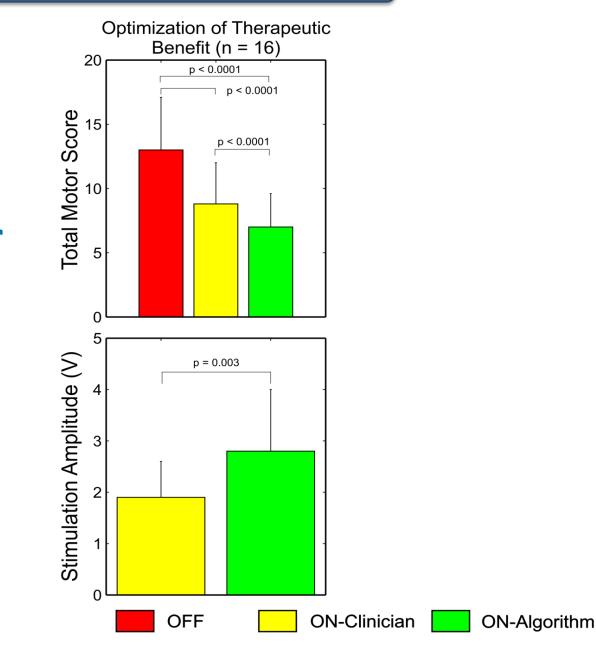
Post Hoc Comparisons to Clinicians

DBS Parameter Selection



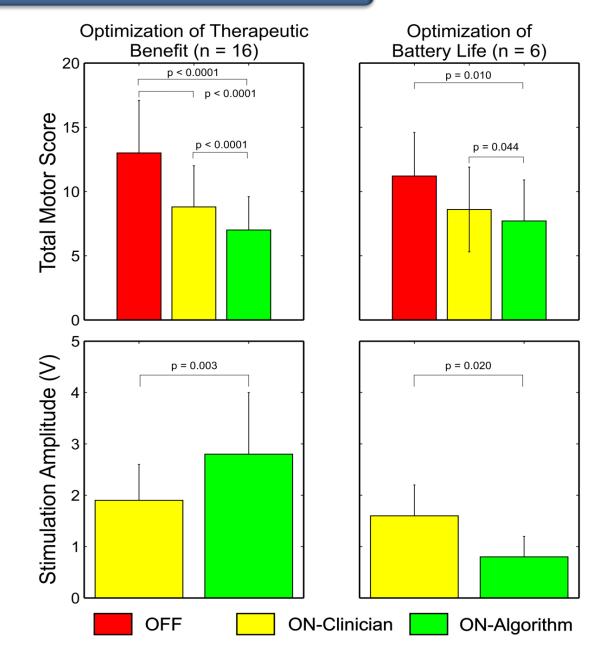
Post Hoc Comparisons to Clinicians

Optimization Algorithms for Tremor & Bradykinesia



Post Hoc Comparisons to Clinicians

Optimization Algorithms for Tremor & Bradykinesia



Real-Time Computer Guided Programming

Algorithm Settings

Subject	Contact / Polarity	Amplitude	Pulse Width (μs)	Frequency (Hz)	Kinesia Score Off	Kinesia Score On After Functional Mapping	Percent Improvement
1	0-/C+	1.8 mA	90	130	0.5	0.4	14.5%
2	1-/C+	0.5 mA	90	130	0.9	0.5	47.9%
3	0-/C+	1.2 mA	90	130	1.8	1.5	13.7%
4	2-/C+	3.9 mA	90	130	2.8	2.0	29.6%
6	1-/C+	1.5 mA	90	130	1.5	0.9	38.1%
7	1-/C+	2.4 mA	90	130	2.8	0.5	82.7%
Average					1.7	1.0	37.8%*

*p = 0.01

Closing The Business Case

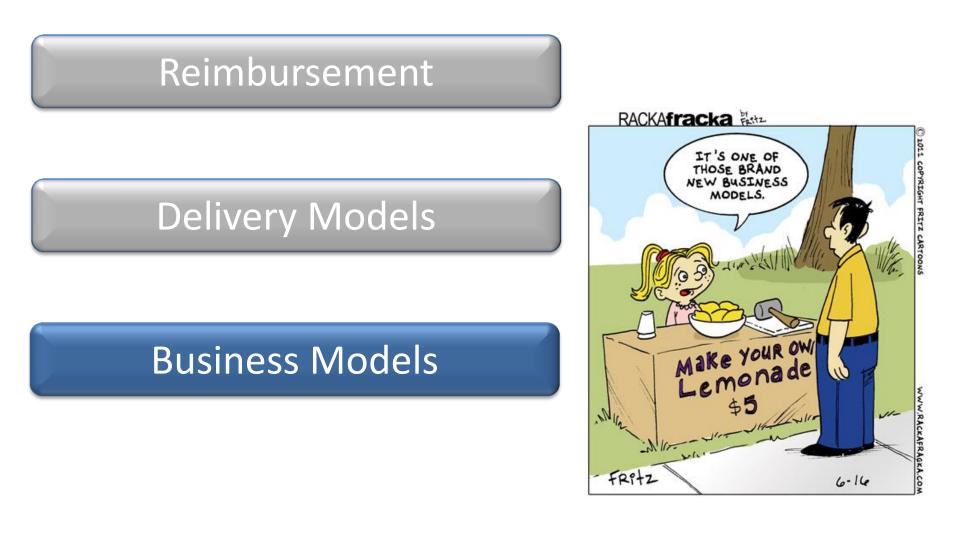
Reimbursement



Reimbursement

Delivery Models





Reimbursement

Delivery Models

Business Models

EHR Integration



"All this talk about EMRs and EHRs is just a fad - like the Internet thing."

Closing The Patient Perspective

I, Myself, Want to See...

Wearables are Important Because...

Wearables Roadmap For the Future?

Acknowledgements: Clinical Collaborators

Site	Location	Collaborator(s)	Research Area	
Baylor College of Medicine	Houston, TX	Dr. Joseph Jankovic	Tremor Assessment, Telemedicine	
Cleveland Clinic	Cleveland, OH	Dr. Hubert Fernandez;	Bradykinesia Assessment	
Henry Ford Health System	Detroit, MI	Dr. Peter LeWitt	Quantitative PD Assessment, Neuroprotection	
Kent State University	Kent, OH	Angie Ridgel, PhD	PD Exercise Therapy	
Johns Hopkins University	Baltimore, MD	Dr. Zoltan Mari	Quantitative PD Assessment	
University of Rochester	Rochester, NY	Dr. Ray Dorsey	Telemedicine	
NIH NINDS	Bethesda, MD	Dr. Mark Hallett	tDCS for Parkinson's	
Rush University	Chicago, IL	Dr. Christopher Goetz	Patient Home Evaluations, Telemedicine	
Southern Illinois University School of Medicine	Springfield, IL	Dr. Rodger Elble	Tremor Assessment	
Greenville Neuromodulation Center	Greenville, PA	Dr. Erwin Montgomery	DBS, Rigidity, Telemedicine	
University of Cincinnati College of Medicine	Cincinnati, OH	Dr. Espay, Dr. Revilla, and Dr. Duker	Quantitative Assessments, Home Monitoring, Deep Brain Stimulation	
University Hospitals Case Medical Center	Cleveland, OH	Dr. David Riley and Dr. Ben Walter	Tremor, Bradykinesia, Gait, Balance, and DBS	
University of Florida	Gainesville, FL	David Vallaincourt, PhD	Tremor	
University of Minnesota	Minneapolis, MN	Dr. Jerrold Vitek	Deep Brain Stimulation	
University of Rochester	Rochester, NY	Dr. Michelle Burack	Dyskinesias	
University of Florida	Gainesville, FL	Dr. Michael Okun	Patient Home Evaluations	
Hospital Universitario de Burgos	B urgos, Spain	Dr. Esther Cubo	Telemedicine	

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Questions?



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