

# Digital Technologies and Big Data in CNS Diseases

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# Twin Revolutions in Neurobiology and Mobile Computing



## A decade for psychiatric disorders

There are many ways in which the understanding and treatment of conditions such as schizophrenia are ripe for a revolution.

MENTAL HEALTH  
A JAMA THEME ISSUE

## Rethinking Mental Illness



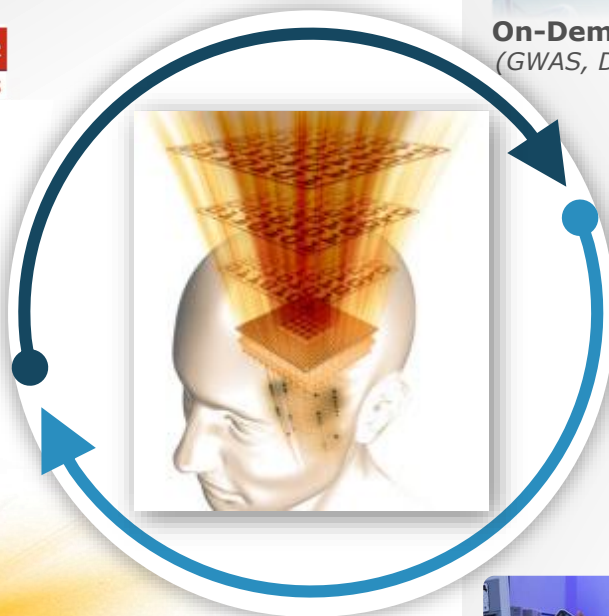
## The Future of Psychiatric Research: Genomes and Neural Circuits



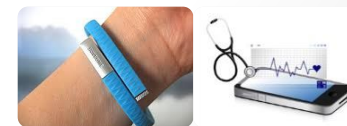
Method of the year, 2011



Recent advances referred to as  
**"the biggest breakthroughs in Psychiatry in 40 yrs"**  
 – NIMH Director



## On-Demand Genomics (GWAS, Deep Sequencing)



## On-Body Physiology, Behavior (EEG, HRV, GSR, Sleep, Speech, Physical/Social Activity, Cognition)



Quantified Self

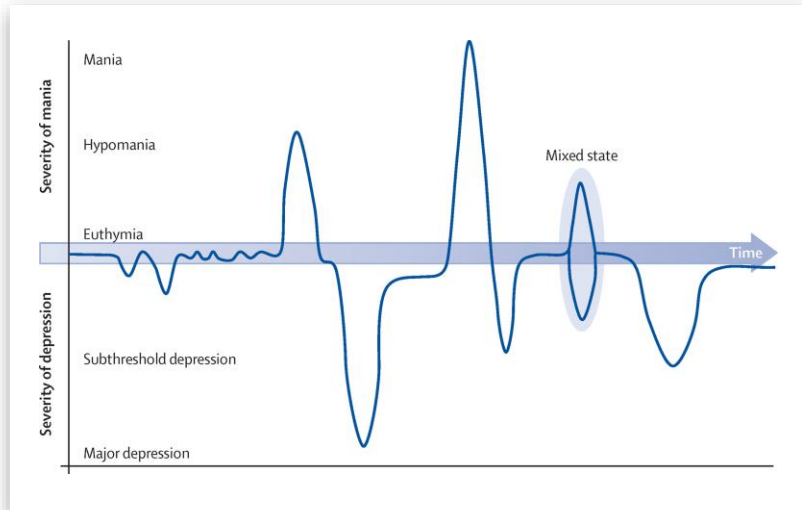
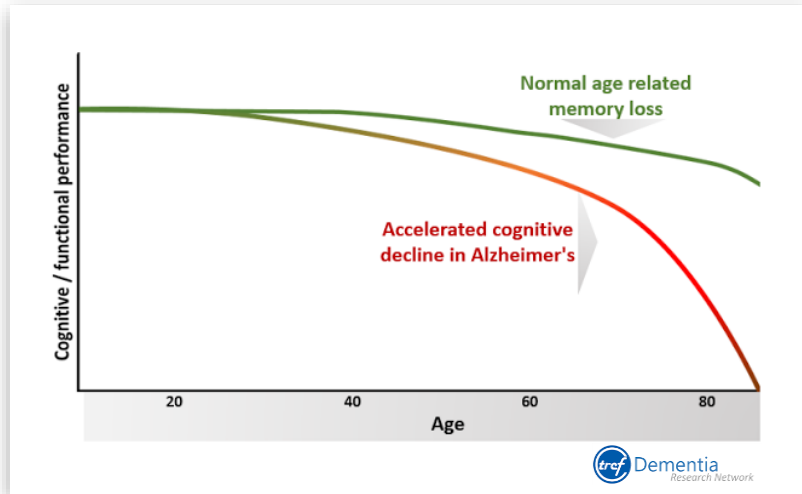


## Point of Need Biochemistry (CRP, Cortisol, Cytokines)

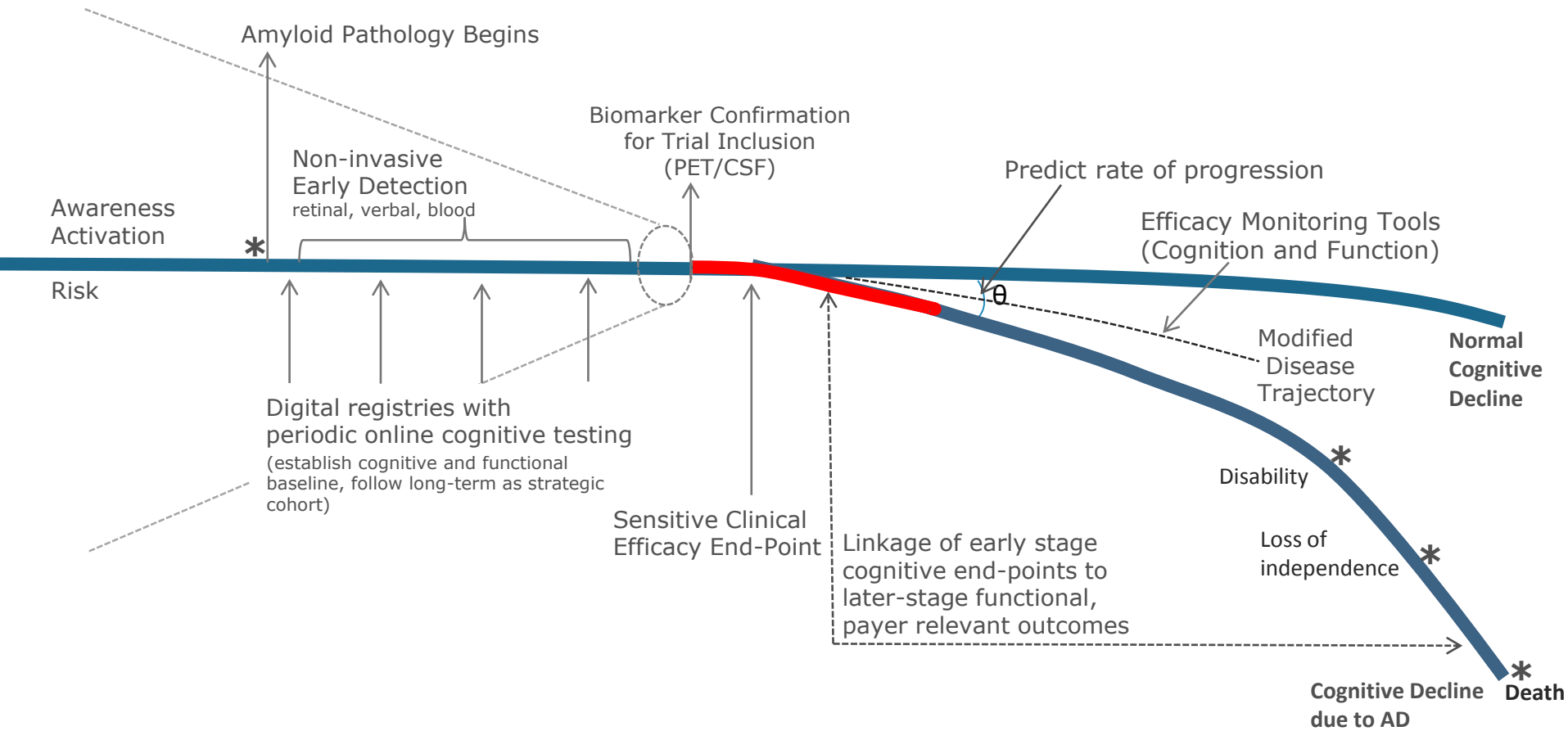


## Computer based cognitive and behavioral therapies

# Neurodegenerative and Neuropsychiatric Changes: Different Time Scales



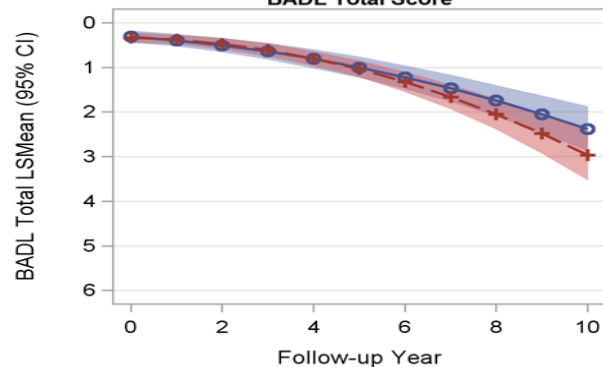
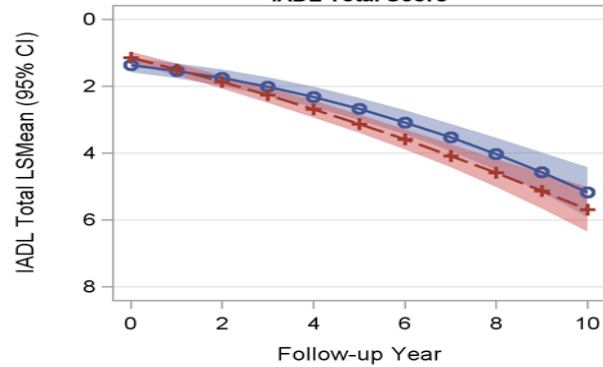
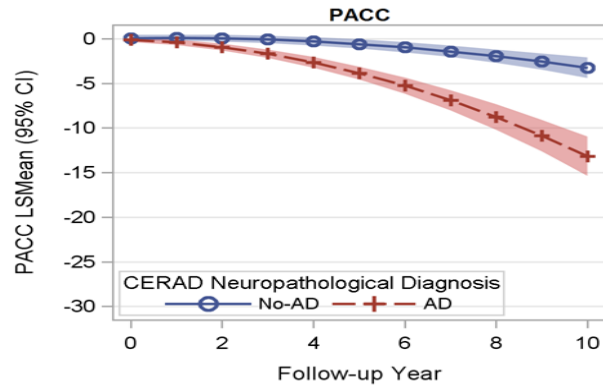
# ALZHEIMER'S DISEASE: COGNITION AND FUNCTION



**Regulators:** will only provide conditional approval, without function

**Payers:** want to see impact on function which drives cost and care-giver burden

# Longitudinal Changes in Function

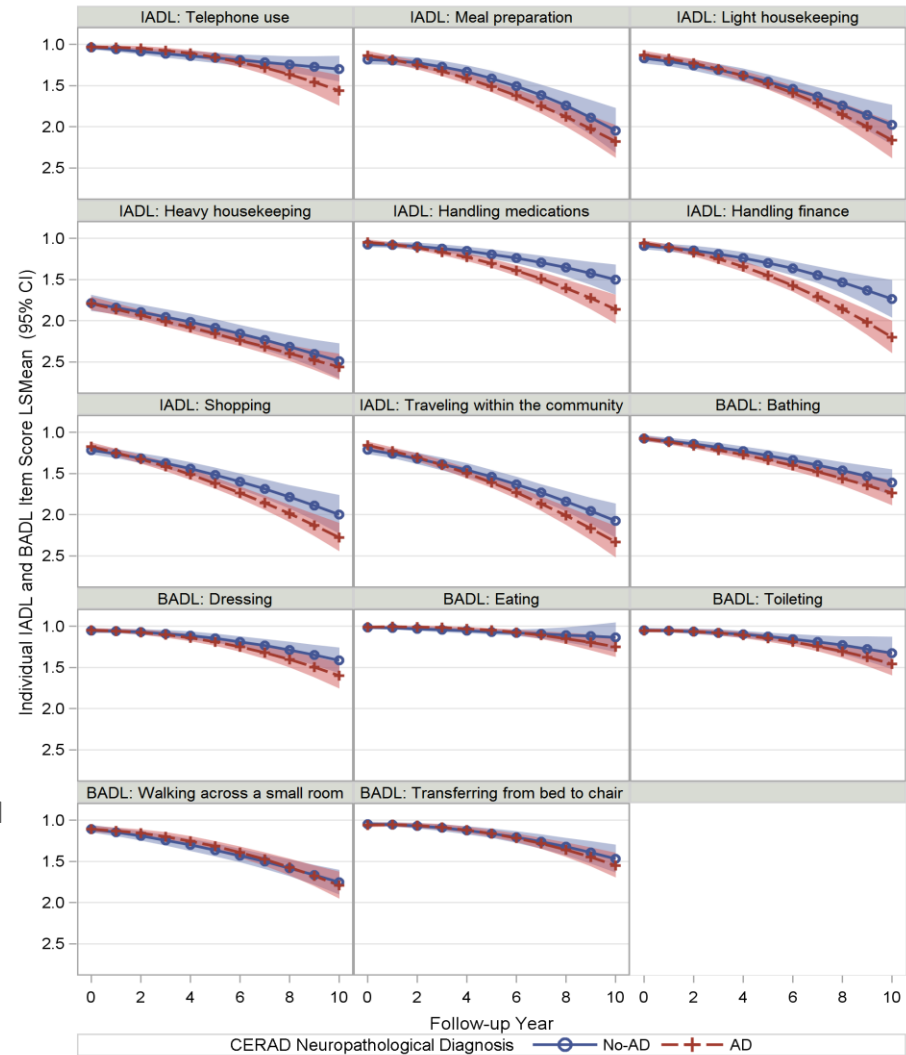


## IADL

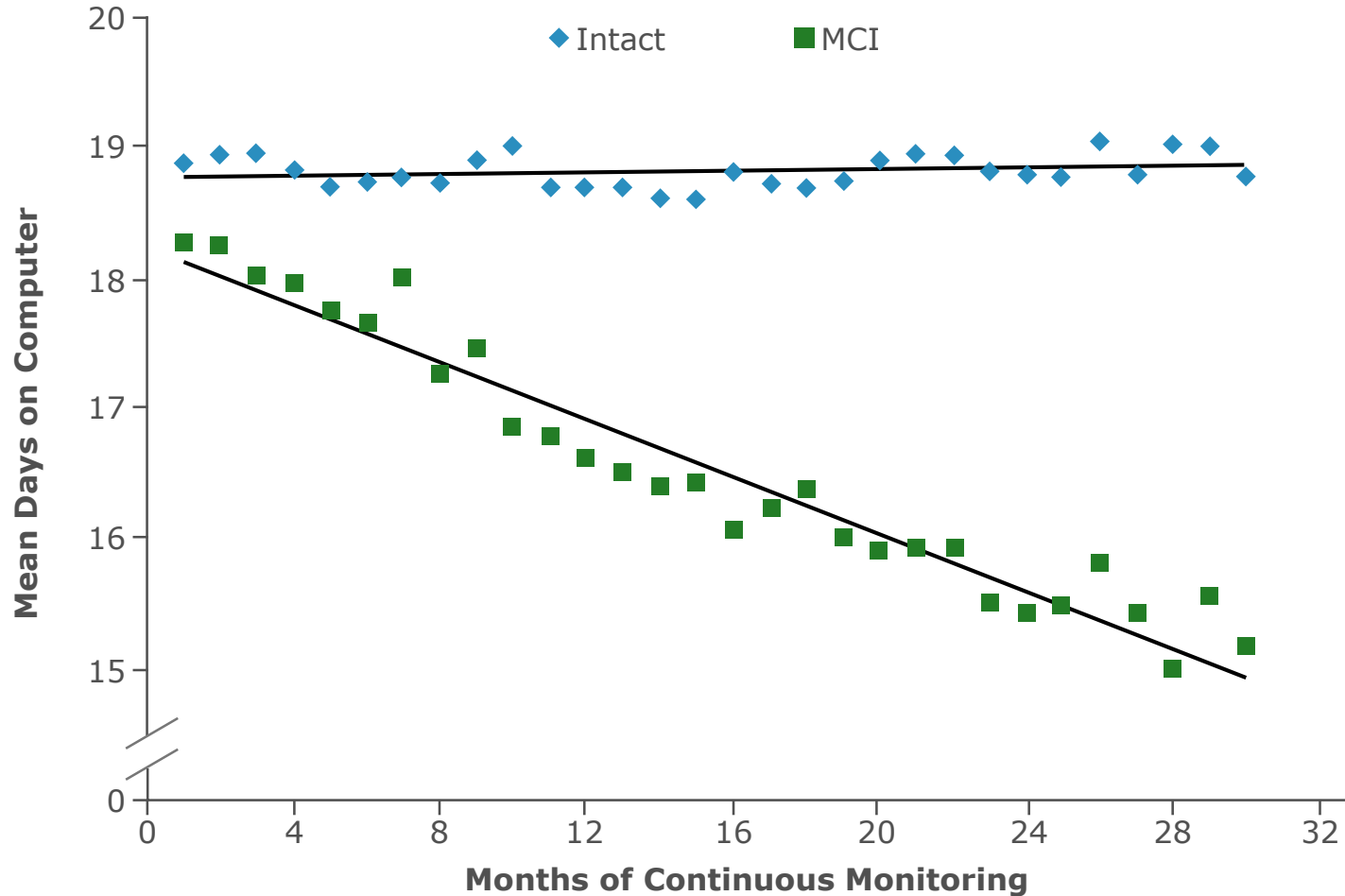
- Managing finances
- Handling transportation (driving or navigating public transit)
- Shopping
- Preparing meals
- Using the telephone and other communication devices
- Managing medications
- Housework and basic home maintenance

## BADL

typically restricted to activities involving functional mobility (ambulation, wheelchair mobility, bed mobility and transfers) and personal care (feeding, hygiene, toileting, bathing and dressing)



# Unobtrusive Measurement of Daily Computer Use to Detect Mild Cognitive Impairment



Kaye J., Mattek N., Dodge H. H., Campbell I., Hayes T., Austin D., et al. (2014). Unobtrusive measurement of daily computer use to detect mild cognitive impairment. *Alzheimers Dement.* 10, 10–17.10.1016/j.jalz.2013

# The Motor Signature of Mild Cognitive Impairment: Results From the Gait and Brain Study

Manuel Montero-Odasso,<sup>1,2,3</sup> Afua Oteng-Amoako,<sup>1</sup> Mark Speechley,<sup>1,2,3</sup> Karen Gopaul,<sup>1</sup> Olivier Beauchet,<sup>4</sup> Cedric Annweiler,<sup>4</sup> and Susan W. Muir-Hunter<sup>1,2</sup>

<sup>1</sup>Gait and Brain Lab, Parkwood Hospital, Lawson Health Research Institute,

<sup>2</sup>Schulich School of Medicine and Dentistry, Department of Medicine and Division of Geriatric Medicine and

<sup>3</sup>Department of Epidemiology and Biostatistics, University of Western Ontario, London, Ontario, Canada.

<sup>4</sup>Department of Neuroscience, Division of Geriatric Medicine, Angers University Hospital, Angers, France.

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Table 3. Dual-Task Gait Cost (%) Differences in a-MCI and na-MCI

Walking Test Condition [Mean ( $\pm$ SD)]	Full MCI Sample (n = 64)	na-MCI (n = 22)	a-MCI (n = 42)	p Value*
<b>Velocity dual-task cost (%)</b>				
Counting gait	7.60 ( $\pm$ 11.63)	3.44 ( $\pm$ 9.28)	9.79 ( $\pm$ 12.24)	<b>.04</b>
Naming animals gait	18.34 (15.43)	12.32 ( $\pm$ 13.39)	21.50 ( $\pm$ 15.63)	<b>.03</b>
Serial sevens gait	20.10 ( $\pm$ 16.52)	14.15 ( $\pm$ 15.59)	23.21 ( $\pm$ 16.31)	<b>.03</b>
<b>Stride time variability dual-task cost (%)</b>				
Counting gait	59.91 ( $\pm$ 92.03)	47.36 ( $\pm$ 77.98)	66.49 ( $\pm$ 98.84)	.43
Naming animals gait	61.09 ( $\pm$ 82.60)	72.11 ( $\pm$ 74.89)	54.88 ( $\pm$ 86.98)	.63
Serial sevens gait	63.46 ( $\pm$ 88.08)	59.03 ( $\pm$ 73.22)	65.74 ( $\pm$ 95.68)	.89

Notes: CI = confidence intervals; MCI = mild cognitive impairment; SD = standard deviation.

\*Linear regression modeling adjusted for age, sex, Trail Making Test B, physical activity level, and comorbidities; significant values are in bold.

Table 4. Association Between Dual-Task Gait Cost (%) and Episodic Memory and Learning Performance (RAVLT) in the MCI Group

Dual-Task Cost (%) (Predictor)	Change in RAVLT (Unadjusted $\beta$ Coefficient)	t Statistic	p Value*	95% Confidence Interval	
<b>Velocity</b>					
Counting gait	.079	2.556	<b>.013</b>	.017	.140
Naming animals gait	.058	2.437	<b>.018</b>	.010	.105
Serial sevens gait	.053	2.422	<b>.019</b>	.009	.096
<b>Stride time variability</b>					
Counting gait	.003	.779	.439	-.005	.012
Naming animals gait	.001	.795	.430	-.002	.005
Serial sevens gait	-.005	-1.014	.316	-.015	.005

Notes: MCI = mild cognitive impairment; RAVLT = Rey Auditory Verbal Learning Test; changes shown above represent 1% increase in dual-task cost.

\*Linear regression adjusted for age, sex, Trail Making Test B, physical activity level, and comorbidities; significant values are in bold.



# Detecting Declining Financial Skills in Preclinical Alzheimer's Disease: The Financial Capacity Instrument--Short Form

Denise C. Mason, PhD, JD <sup>1,2</sup>, Kristen L. Trisbel, PsyD <sup>1,2</sup>, Adam Gerstenecker, PhD <sup>1</sup>, Roy C. Martin, PhD <sup>1,2</sup>,  
Kelly Edwards, MA <sup>3</sup>, Vernon S. Pankratz, PhD <sup>3</sup>, Dana Swenson-Davis, MA <sup>4,5</sup>, and Ronald C. Petersen, MD, PhD <sup>4,5</sup>



<sup>1</sup> Department of Neurology, University of Alabama at Birmingham [UAB], Birmingham, AL; <sup>2</sup> Alzheimer's Disease Center, UAB, Birmingham, AL; <sup>3</sup> Division of Biomedical Statistics and Informatics, Mayo Clinic and Foundation, Rochester, MN; <sup>4</sup> Department of Neurology, Mayo Clinic and Foundation, Rochester, MN; <sup>5</sup> Mayo Clinic Alzheimer's Disease Research Center, Mayo Clinic and Foundation, Rochester, MN

**Table 2: FCI-SF Performance Scores by PiB Status**

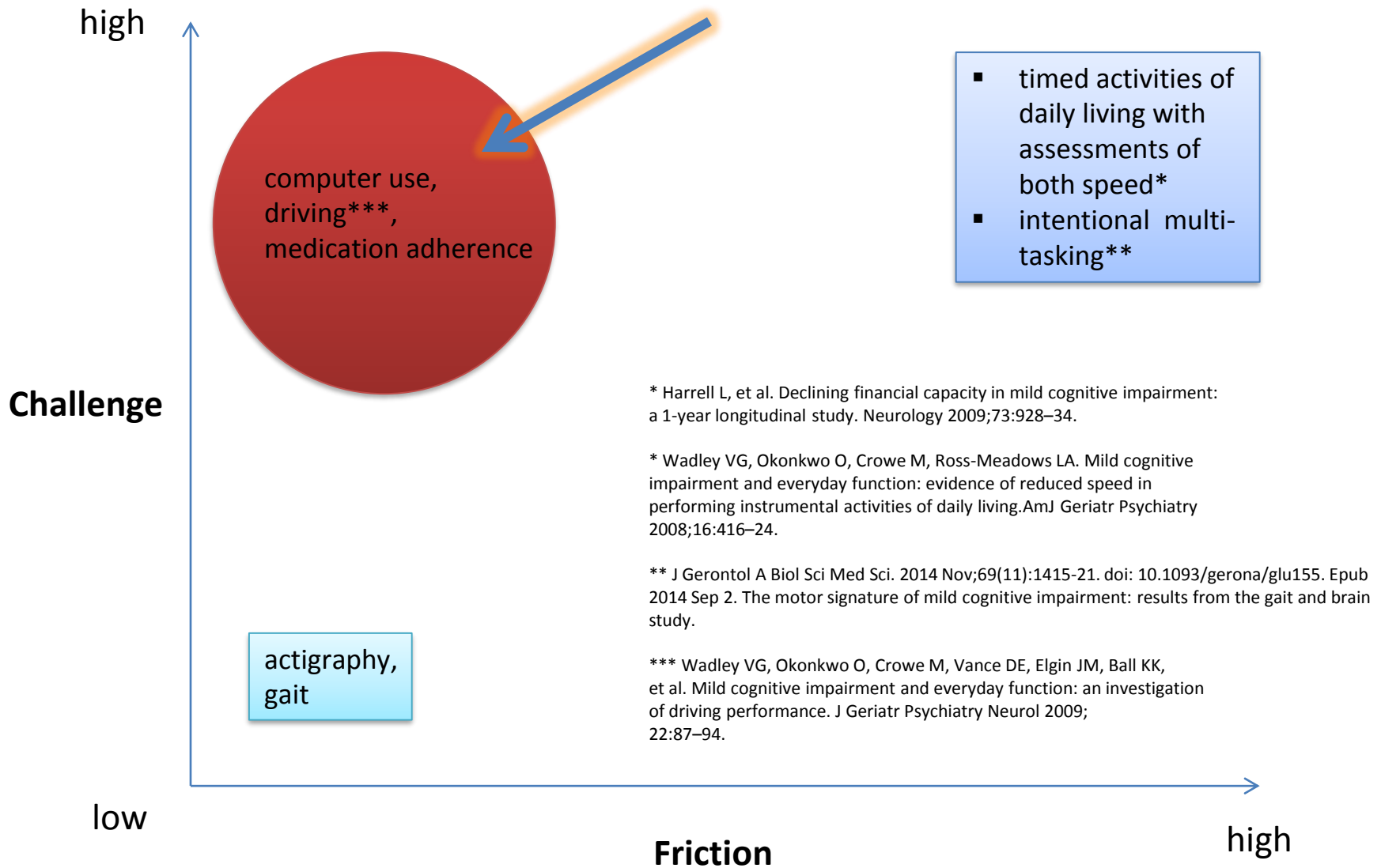
	Max Score	Amyloid Negative N = 120	Amyloid Positive N = 66	p
<b>Total Score</b>				
FCI-SF Total Score	0-74	64.2 (8.3, 33-74)	62.1 (7.9, 44-74)	.094
<b>Component Performance Scores</b>				
Bank Statement Management	0-14	11.0 (2.2, 2-13)	10.4 (2.5, 3-13)	.080
Mental calculation problems	0-4	3.6 (0.9, 0-4)	3.4 (0.9, 2-4)	.125
Check/Register/Deposit Transactions	0-28	23.1 (4.9, 4-28)	22.2 (5.3, 8-28)	.230
Financial Conceptual Knowledge	0-8	7.0 (1.3, 1-8)	6.8 (1.4, 4-8)	.319
Single Check Transactions/Register	0-20	18.8 (1.9, 10-20)	18.8 (1.7, 14-20)	.987
<b>Select FCI-SF Test Items</b>				
How many quarters in \$	0 or 2	1.9 (0.4)	1.7 (0.7)	.011
Correct amount of deposit	0 or 2	1.6 (0.8)	1.3 (1.0)	.022
Checks cleared in bank statement	0 or 2	1.3 (1.0)	1.0 (1.0)	.028

**Table 3: FCI-SF Task Completion Times by PiB Status**

	Max score/time	Amyloid Negative N = 120	Amyloid Positive N = 66	p
<b>Time Variables</b>				
Medical deductible problem (sec)	90 sec	17.2 (23.4)	18.9 (24.4)	.637
Income tax problem (sec)	90 sec	7.1 (10.1)	10.2 (11.0)	.062
Check/register task (sec)	240 sec	112.4 (34.6)	125.4 (36.2)	.017
Check/register/deposit task (sec)	300 sec	225.1 (62.8)	250.1 (54.9)	.007
2 check tasks composite time (sec)	540 sec	337.7 (87.1)	375.5 (77.4)	.004
Overall composite time (sec)	720 sec	359.6 (95.6)	404.6 (87.2)	.002

- Functional change in the form of subtle financial skill decline occurs in preclinical AD and is detectable using a brief performance measure of financial skills. Both slower financial task completion times, and diminished performance of complex financial tasks, characterized the A+ group and represent measurable very early functional changes in preclinical AD.
- Performance measures of complex everyday function may outperform cognition variables as predictors of preclinical AD. In a logistic model of participant amyloid status, only FCI-SF predictors (composite time and two complex performance items) entered the model. Age and cognitive variables were not retained in the model.
- Brief performance based measures of complex everyday function like the FCI-SF represent a promising new addition to AD clinical trial outcomes. Such measures appear to be sensitive to early functional changes across preclinical, prodromal, and clinical dementia stages of AD, and have face and ecological validity for consumers, families, and clinicians and researchers. Future studies will be needed to confirm our present findings.
- Measures like the FCI-SF should be considered for AD prevention and clinical trials.


# Digital Technologies, Big Data and AD Function



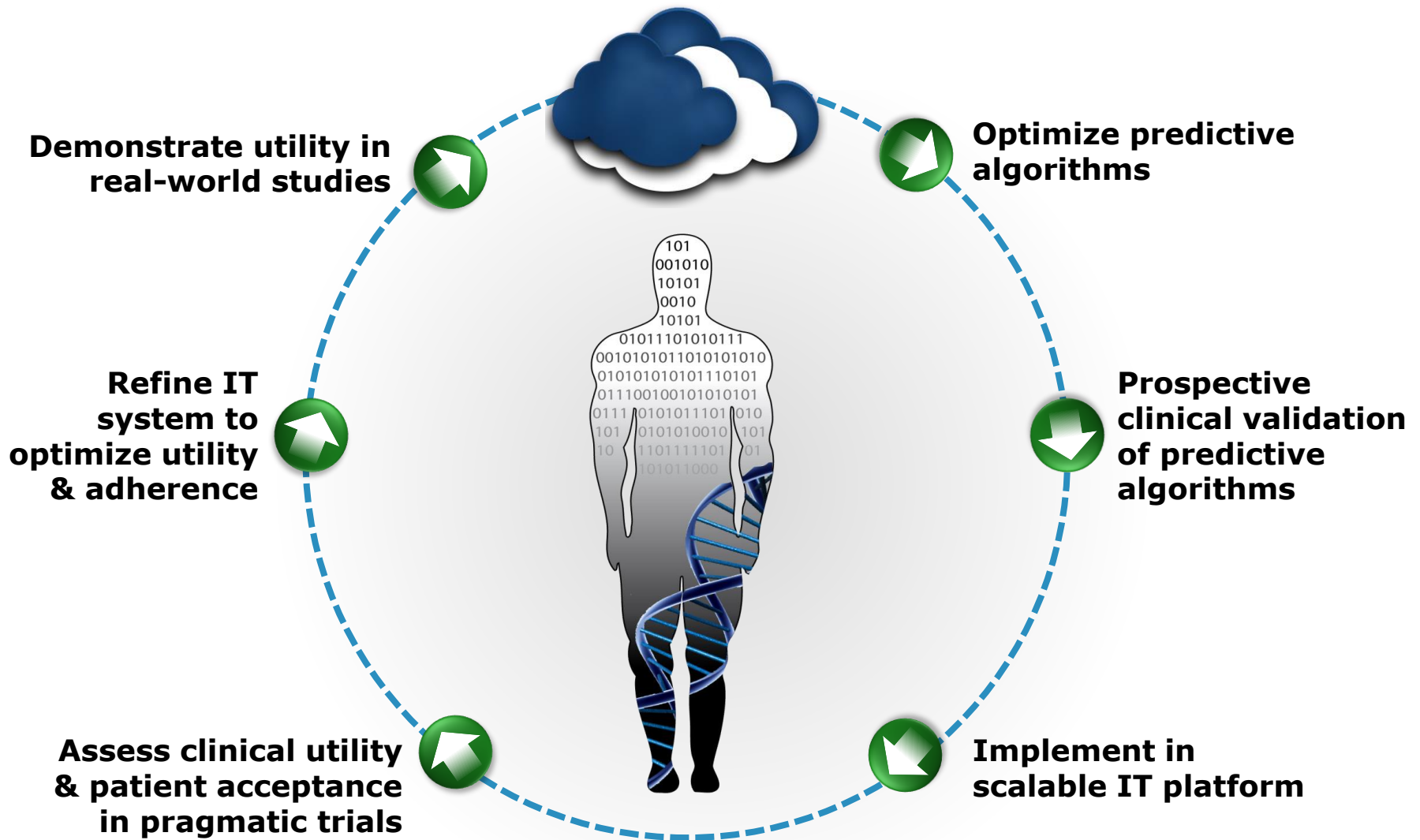


# Biosensor Outcome Measures/Assessments of Clinical Signs: Uses in Clinical Trials



Context-of-Use	Definition	CDISC Standards	Evidentiary Standards
Internal Decisions	Commercial devices that is not a medical device (e.g., Smartphones, Fitbits™, Home-sensor monitoring, etc.).	No	Arbitrary
Exploratory Measures	Commercial devices that is not a medical device for potential patient stratification (e.g., Smartphones, Home-sensor monitoring, etc.,)	No	
Prognostic Assessment	Medical devices for use in in trial enrichment to predict rate of disease progression	YES	<div style="border: 1px solid black; padding: 5px; display: inline-block;">Regulatory Endorsement</div>  <b>Robust; Multi-trial; Longitudinal</b>
Predictive Assessment	Medical devices for trial enrichment to identify individuals who are more likely than similar patients without the outcome to experience a favorable or unfavorable effect from a specific intervention/ exposure.	YES	
Susceptibility/ Risk Assessment	Medical devices for use in in trial enrichment to indicate the potential for developing a disease or medical condition or sensitivity to an exposure in an individual without clinically apparent disease or medical condition.	YES	
Surrogate Endpoint	Medical devices in clinical trials as a substitute for a direct measure of how a patient feels, functions, or survives. A surrogate endpoint does not measure the clinical benefit of primary interest in and of itself, but rather is expected to predict that clinical benefit or harm based on epidemiologic, therapeutic, pathophysiologic, or other scientific evidence.	YES	
Clinical Endpoint	Medical devices measuring precisely defined variable(s) intended to reflect an outcome of interest that is statistically analyzed to address a particular research question. A precise definition of an endpoint typically specifies the type of assessments made, the timing of those assessments, the assessment tools used, and possibly other details, as applicable, such as how multiple assessments within an individual are to be combined.	YES	

# The Data Driven **LEARNING** Engine





Is his medication working?  
 Is he at a higher risk of relapse?  
 Is he taking his medication?

physician



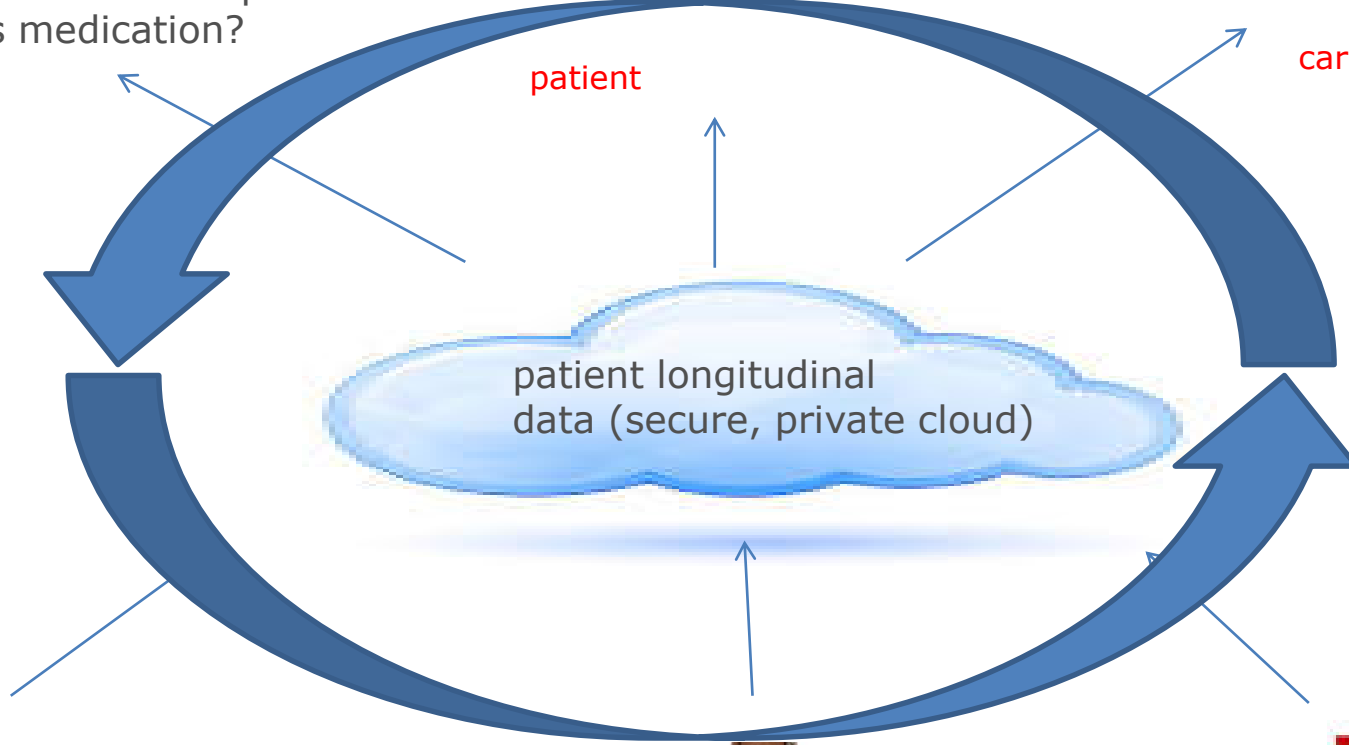
Is my medication working?  
 Am I doing better?

patient



Is he doing well?  
 Does he need help?

care-giver



physician



clinical assessments  
 point-of-care adherence test  
 (EHR)



patient



self-report symptoms  
 passive sleep, activity  
 (social, physical) data

care-giver



independent  
 informer

# From Diagnose and Treat to Predict and Preempt: A Look Ahead

## Pre-Diagnosis

## Early Disease

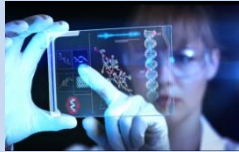
## Chronic Disease



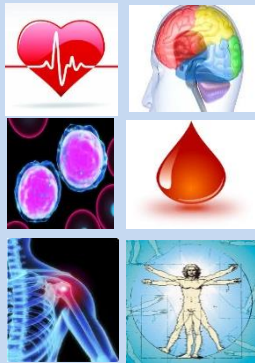
awareness  
education



risk



health & wellness  
tracking



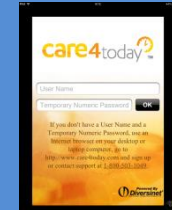
organ channels



early diagnosis  
disease progression monitoring



non-pharmacological  
interventions



adherence management, efficacy  
monitoring, relapse prediction



low risk

high risk

early disease

managed disease

← continuous, longitudinal engagement →

*THANK YOU*

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