

Digital Biomarker Deployment for Gait Analysis in Multiple Sclerosis

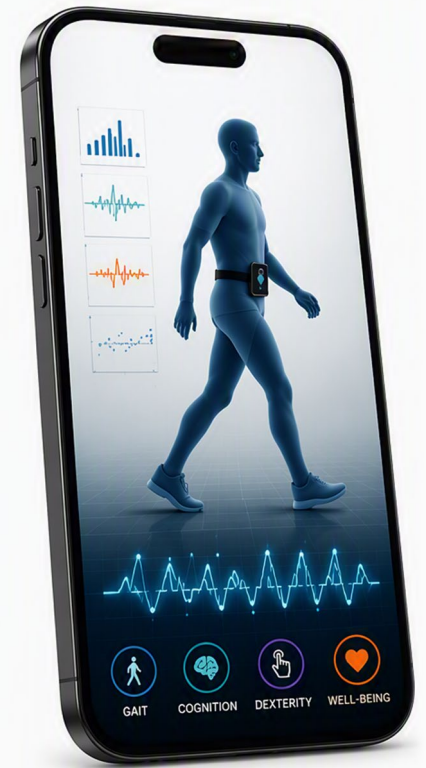
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Agenda

- | | | | |
|---|---|----|--|
| 1 |  MS Gait Characteristics | 6 |  Analytical Validation |
| 2 |  Digital Endpoints | 7 |  Clinical Validation |
| 3 | V3+ Qualification Framework | 8 |  Longitudinal Sensitivity |
| 4 |  Verification | 9 |  Clinical Interpretability (MCID) |
| 5 |  Usability & Adherence | 10 |  Conclusion |

Multiple Sclerosis Gait Characteristics

“Gait impairment represents one of the most common and disabling symptom of multiple sclerosis.”

Lizrova Preiningeroova et. al, J. Neuroeng Rehabil. 2015

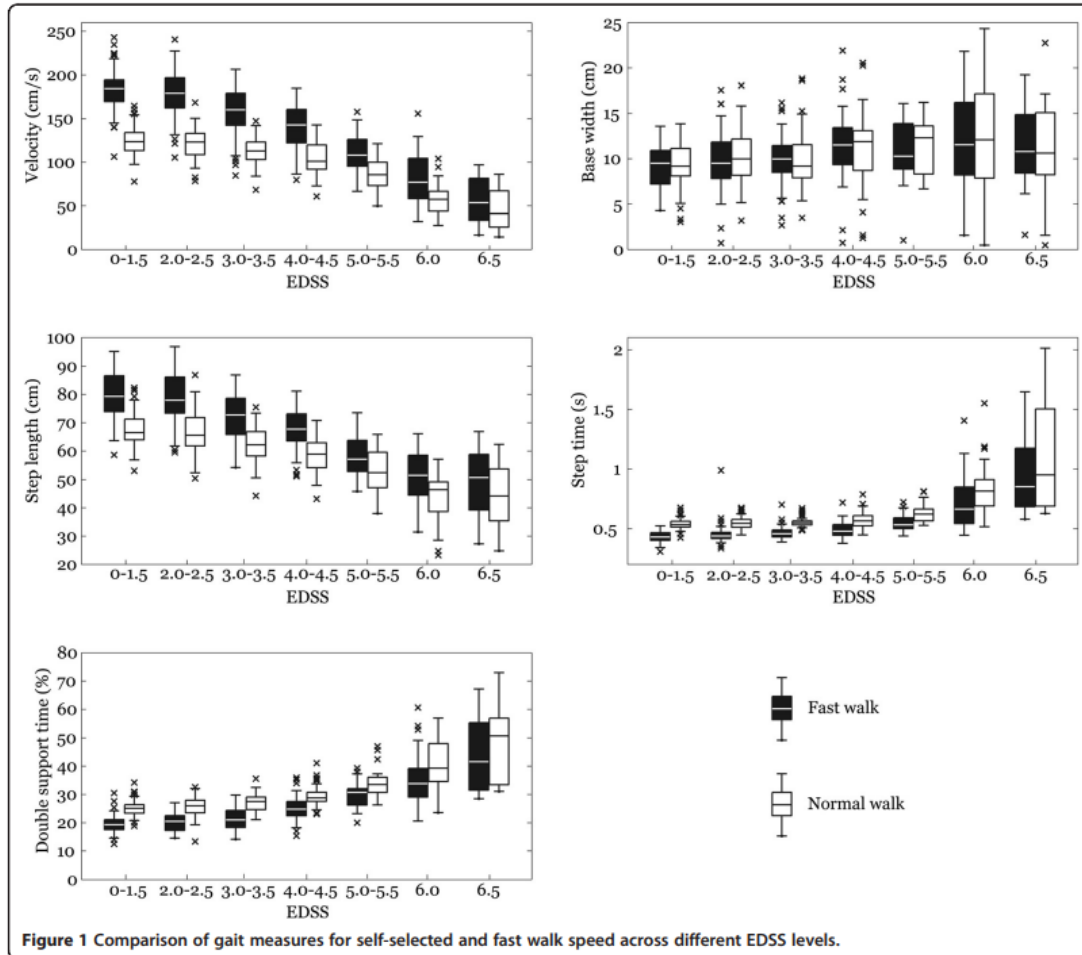
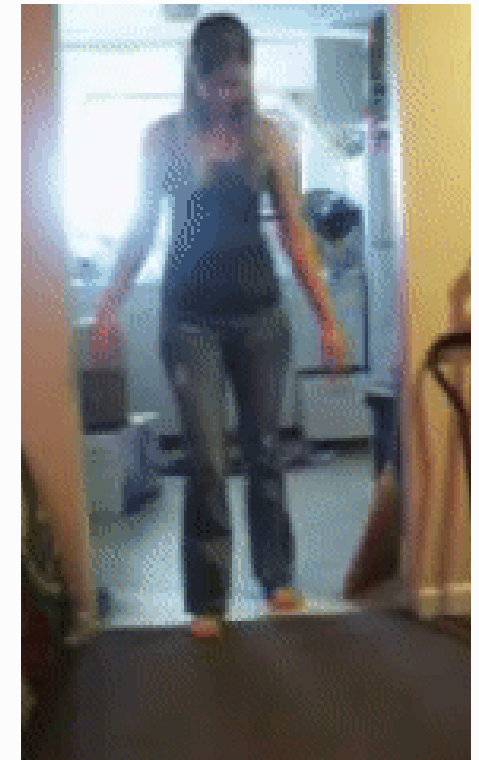
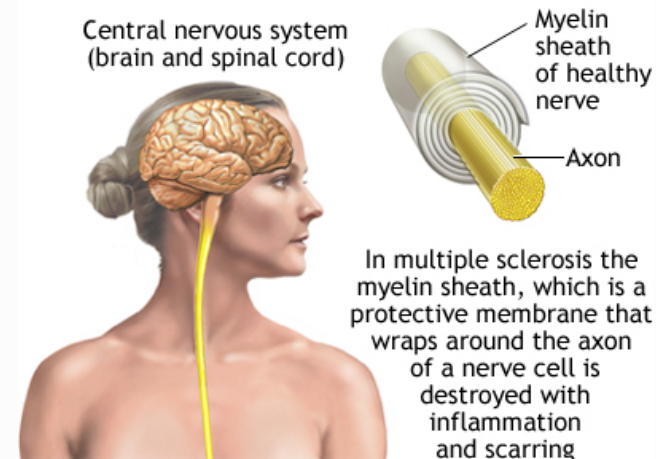


Figure 1 Comparison of gait measures for self-selected and fast walk speed across different EDSS levels.

Spatial and temporal characteristics of gait as outcome measures in multiple sclerosis (EDSS 0 to 6.5) – PMC
Gait impairment in persons with MS (PwMS):

- ↓ Slower gait speed
- ↓ Reduced cadence
- ↓ Shorter stride/step length
- ↑ Increased stride/step time
- ↑ Increased double support time



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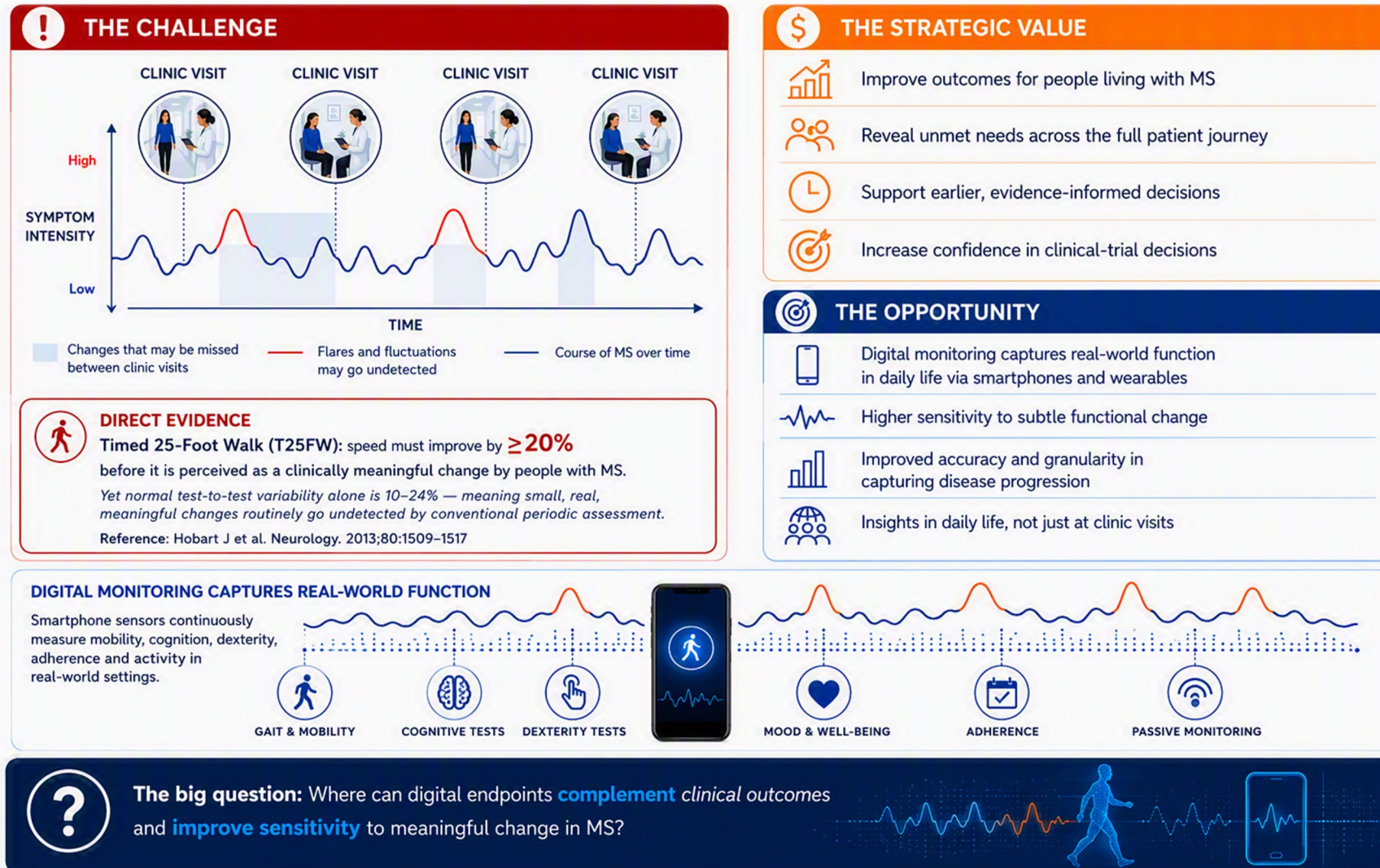
8  Longitudinal Sensitivity

9  Clinical Interpretability (MCID)

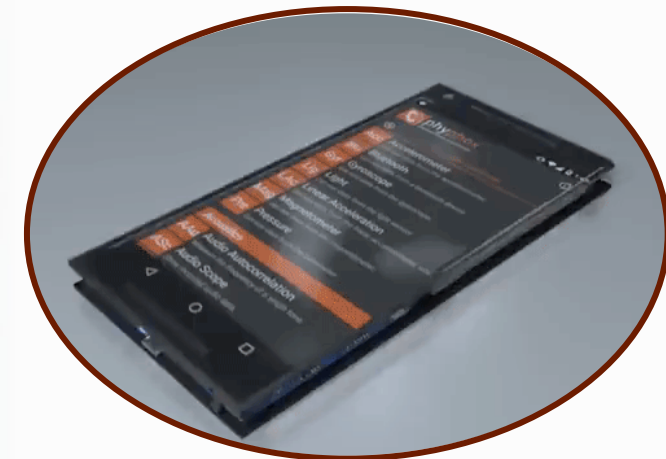
10  Conclusion

Why Digital Endpoints matter in Multiple Sclerosis

Better insights. Faster decisions, Better outcomes for people living with MS.



Commercial smartphone sensors



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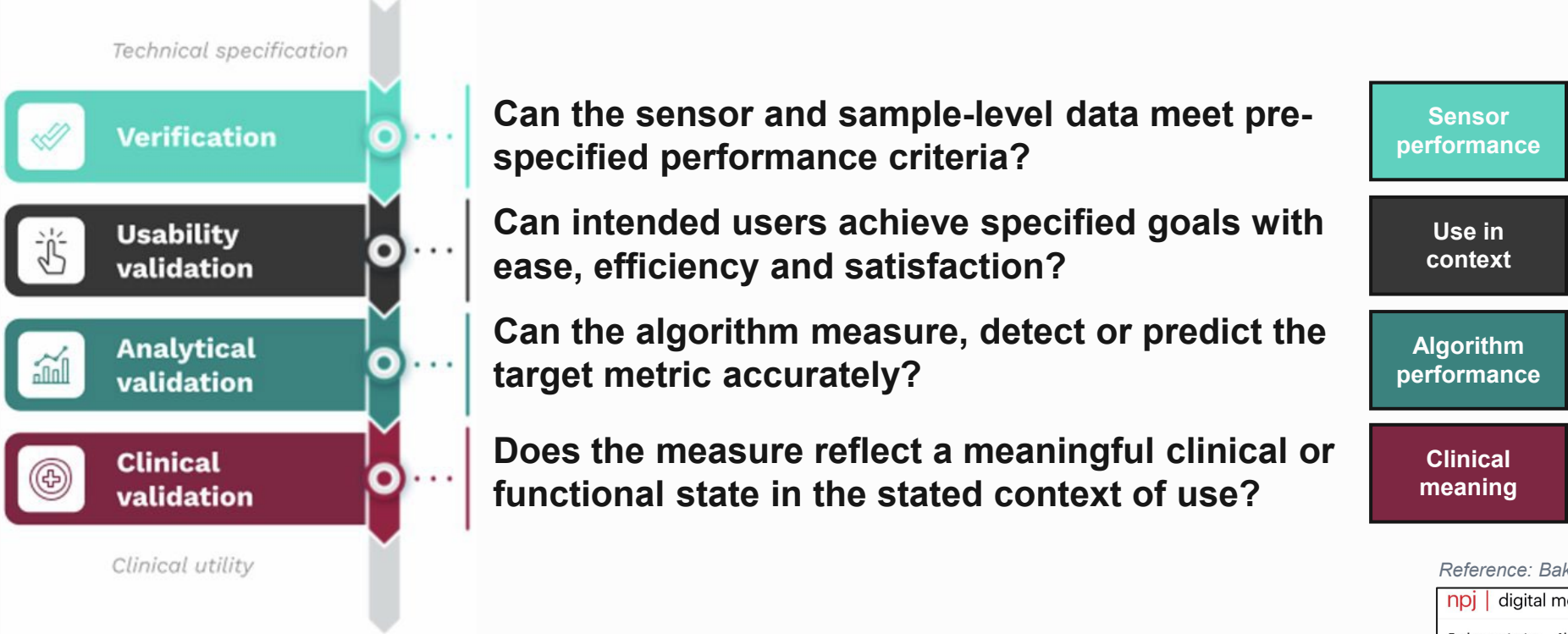
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What evidence makes a Digital Health Technology fit for purpose?

V3+ separates four complementary evidence questions before a digital measure is used to evaluate an intervention.



For this talk, longitudinal sensitivity and MCID extend the evidence story beyond the four formal V3+ components.

FIT FOR PURPOSE = sufficient evidence for the specified context of use

Reference: Bakker JP et al. *npj Digital Medicine*. 2025;8:51

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nature > npj.digital.medicine > perspectives > article

Perspective | [Open access](#) | Published: 24 January 2025

V3+ extends the V3 framework to ensure user-centricity and scalability of sensor-based digital health technologies

[Jessie P. Bakker](#) , [Roland Barge](#), [Jacob Centra](#), [Bryan Cobb](#), [Chas Cota](#), [Christine C. Guo](#), [Bert Hartog](#), [Nathalie Horowicz-Mehler](#), [Elena S. Izmailova](#), [Nikolay V. Manyakov](#), [Samantha McClenahan](#), [Stéphane Motola](#), [Smit Patel](#), [Oana Paun](#), [Marian Schoone](#), [Emre Sezgin](#), [Thomas Switzer](#), [Animesh Tandon](#), [Willem van den Brink](#), [Srinivasan Vairavan](#), [Benjamin Vandendriessche](#), [Bernard Vrijens](#) & [Jennifer C. Goldsack](#)

[npj Digital Medicine](#) **8**, Article number: 51 (2025) | [Cite this article](#)

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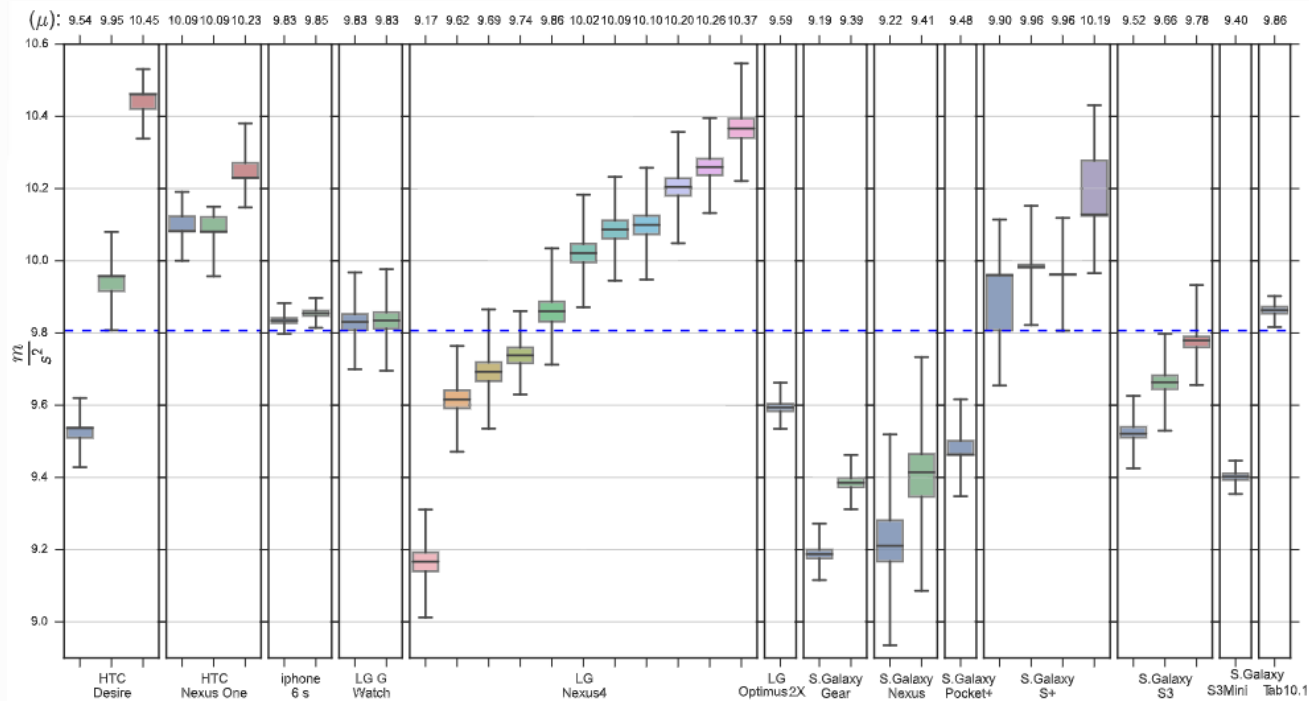
9  Clinical Interpretability (MCID)

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Verification: Are Accelerometer Measurements Consistent Across Devices?

Device and sensor heterogeneity can introduce systematic error before the gait algorithm is applied.

Figure 1: Accelerometer biases of mobile devices in six different orientations under static conditions.

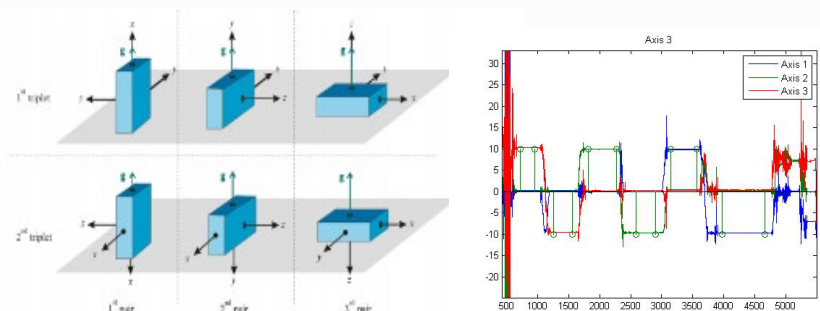


(a) Box plots of acceleration magnitudes and respective means (top) for investigated devices when lying on the back, with the whiskers indicating min/max values.

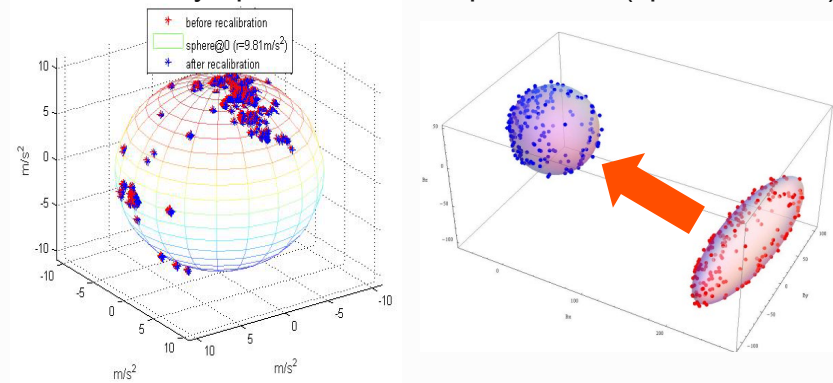
Reference: Stisen A et al. Smart Devices are Different. Proc. ACM SenSys. 2015;127–140.

Accelerometer calibration techniques

Controlled six-position calibration:
Requires access to the device and accurate positioning



Ellipsoid fitting: Does not require known sensor orientations
Offset update = fitted ellipsoid center
Sensitivity update = fitted ellipsoid radii (sphere radius)



Ellipsoid or sphere fitting for sensor calibration: ST Microelectronics

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Artios study overview & adherence

ARTIOS is a **single arm**, prospective, multicentre and open-label, **96-week study** to **evaluate the treatment effectiveness of ofatumumab** in subjects with **relapsing multiple sclerosis** who had switched from commonly used oral disease-modifying therapies.

Bove R et al. J Neurol 273, 434 (2026).

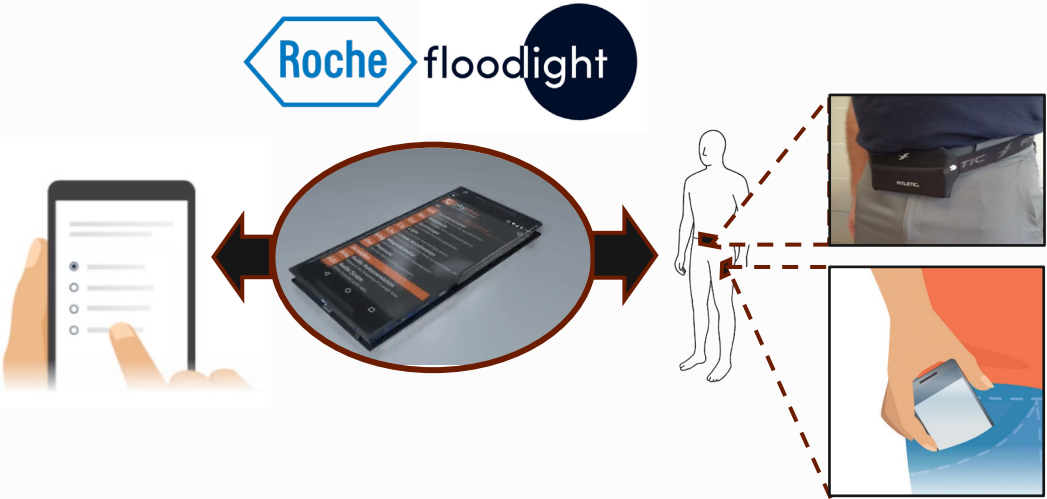


Figure 1: Overview of assessment suite for Roche FLOODLIGHT mobile app

Test type	Active tests									Passive monitoring	
	Experience sampling			Cognition	Hand & arm		Gait & posture			Gait & posture	
Test name	Daily Mood Question (DMQ)	Symptom Tracker (ST)	Multiple Sclerosis Impact Scale (MSIS-29)	Information Processing Speed (IPS)	Pinching Test	Draw a Shape Test	Static Balance Test (SBT)	Five-U-Turn Test (5UTT)	Two Minute Walk Test (2MWT)	Gait Behaviour	Mobility Pattern
Frequency	Daily	Fortnightly & ad-hoc	Fortnightly	Weekly	Daily	Daily	Daily	Daily	Daily	Continuous	Continuous

Conduct assessments...

- In a quiet space with no distractions
- While seated

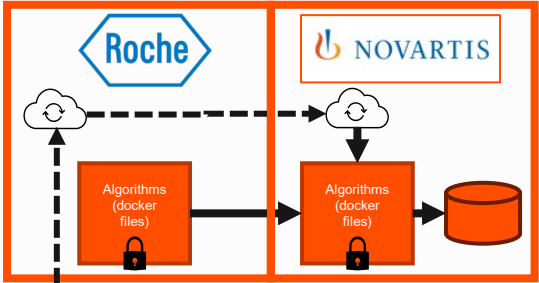
Conduct assessments...

- In a quiet space with no distractions
- While standing

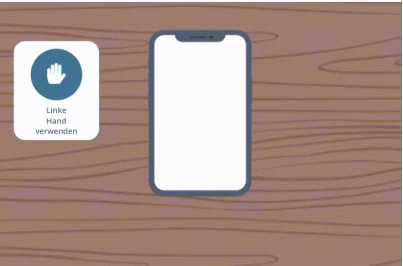
Conduct assessments...

- Outdoors, e.g. in a nearby park

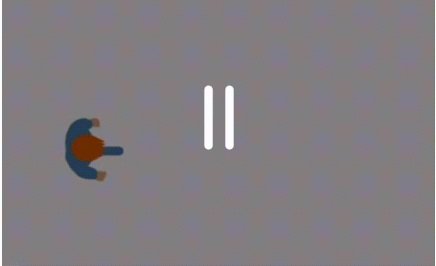
Reference: Bourke AK et al. Quantifying adherence to smartphone-based tests over 96 weeks in the ARTIOS study: monitoring motor function, dexterity, cognition and patient-reported outcomes. ECTRIMS 2025; Poster P1952.



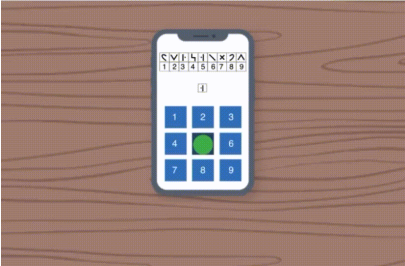
Subjects with DHTs



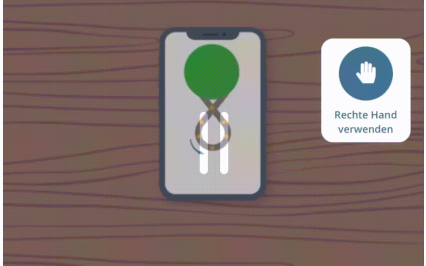
Pinching test



5 u-turn test

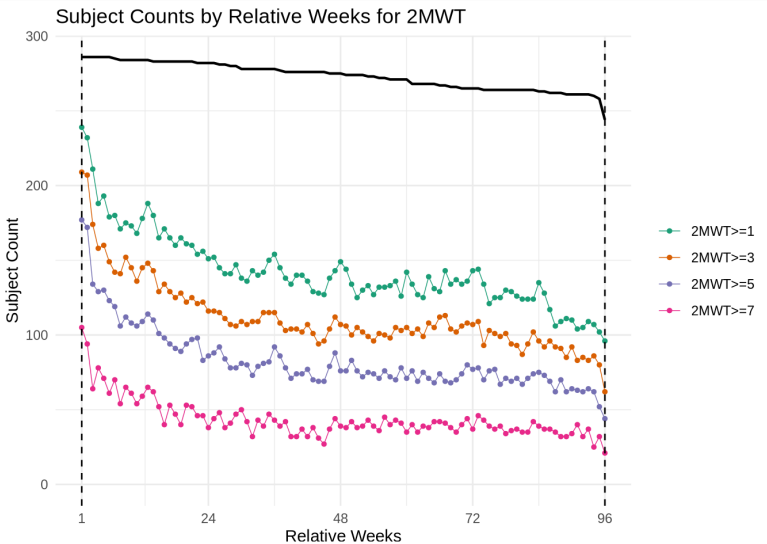
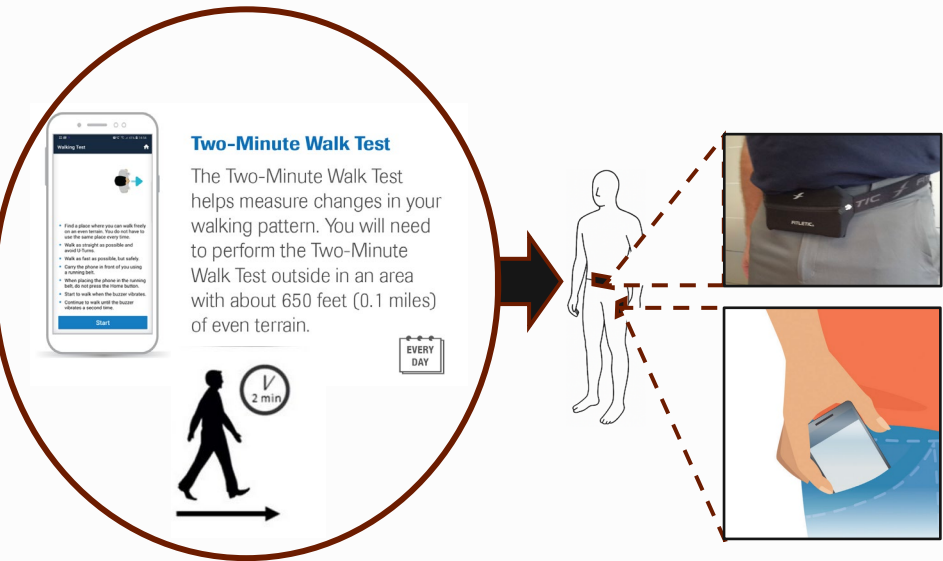


eSDMT
(Information processing speed)



Draw-a-shape test

Artios 2MWT summary adherence statistics



Characteristic	N = 454,696 ¹
Total ACTIVE TESTS	454,696 (100%)
CATEGORY	
ACTIMETRY	434,915 (96%)
COGNITIVE TEST	19,781 (4.4%)
TEST	
Draw-a-shape	95,912 (21%)
2MWT	74,963 (16%)
eSDMT	19,781 (4.4%)
Pinching test	98,334 (22%)
static balance test	86,528 (19%)
5 U-Turn test	79,178 (17%)

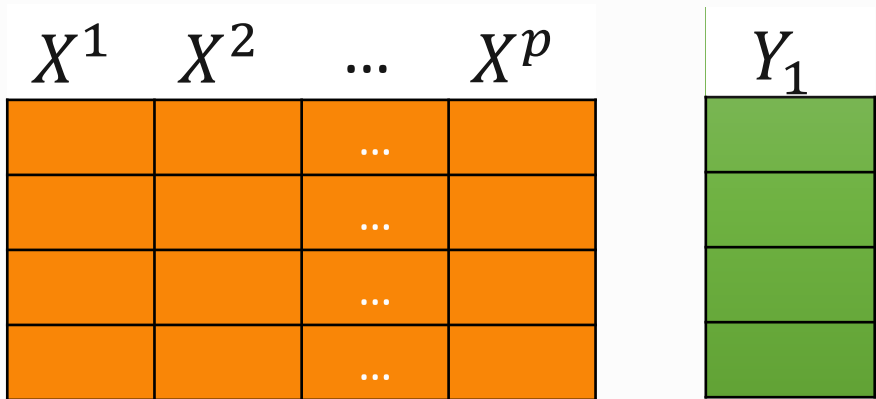
Retention rates for daily tests

Threshold (number of tests) ≥1			
PRO/Test	Week 1: Participant, n	Week 96: Participant, n	Retention rate (%)
DMQ	257	127	49.42
Draw-a-shape	254	120	47.24
Pinching test	254	120	47.24
SBT	250	113	45.2
5UTT	246	105	42.68
2MWT	239	96	40.17

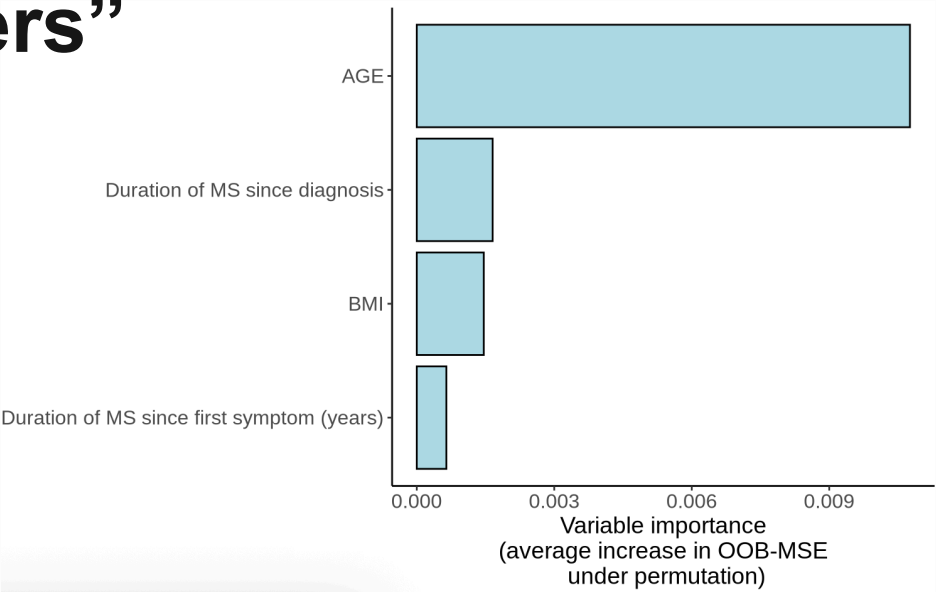
Threshold (number of tests) ≥3			
PRO/Test	Week 1: Participant, n	Week 96: Participant, n	Retention rate (%)
DMQ	243	84	34.57
Draw-a-shape	239	77	32.22
Pinching test	240	77	32.08
SBT	232	69	29.74
5UTT	228	60	26.32
2MWT	209	62	29.67

Threshold (number of tests) ≥7			
PRO/Test	Week 1: Participant, n	Week 96: Participant, n	Retention rate (%)
DMQ	170	33	19.41
Draw-a-shape	161	27	16.77
Pinching test	162	28	17.28
SBT	141	26	18.44
5 U-Turn Test	129	22	17.05
2MWT	105	21	20

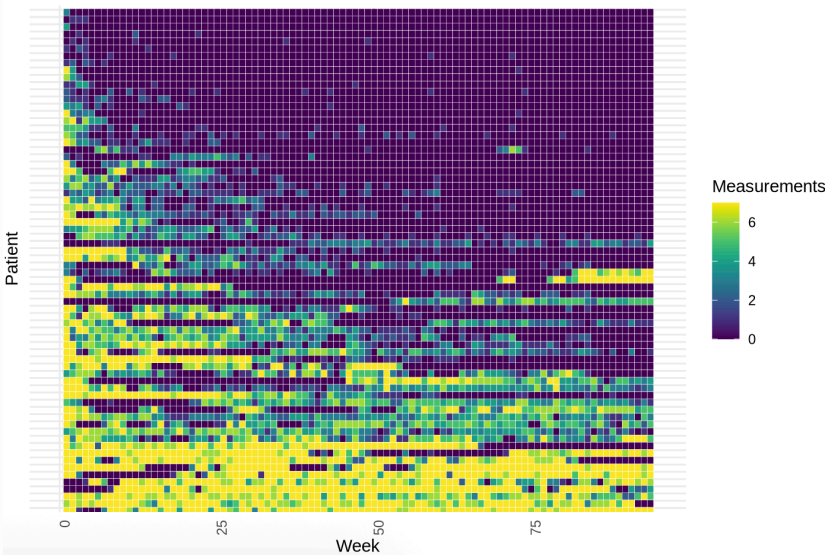
Adherence of 2MWT: ML modelling to understand “drivers”



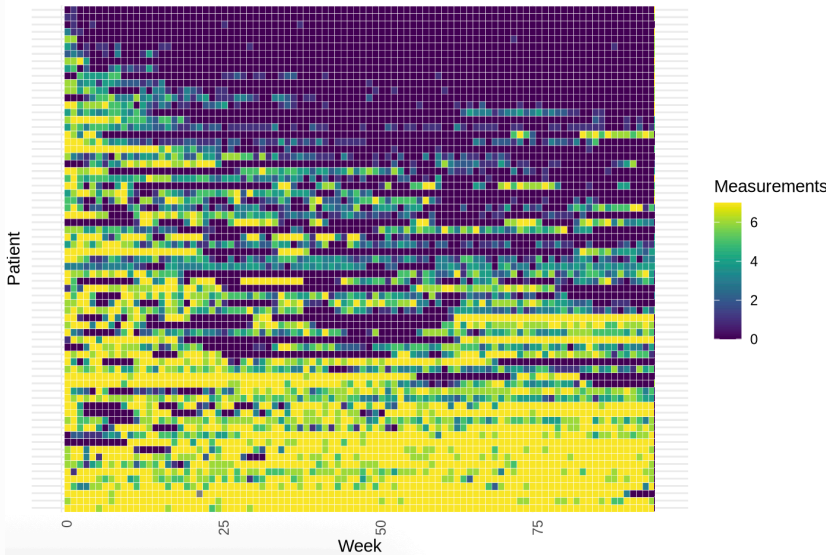
Baseline covariates



Younger
participants
(>28 years old)



Older
participants
(>45 years old)



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From Raw Sensor Data to Digital Gait Measures



sensors



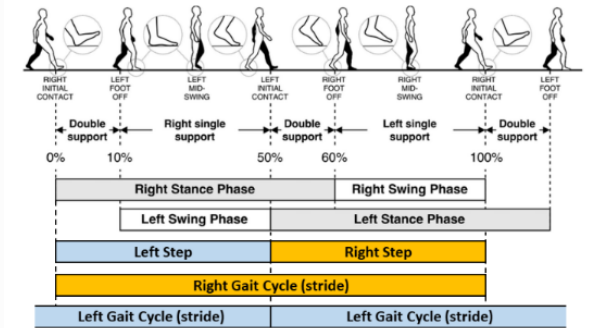
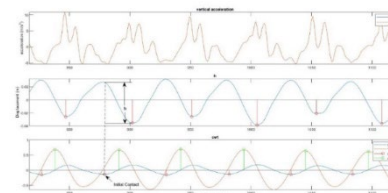
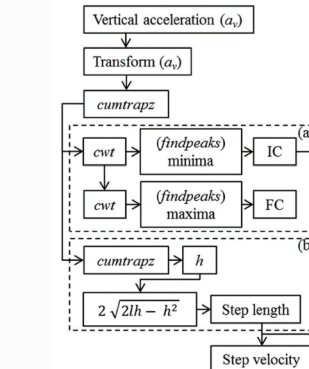
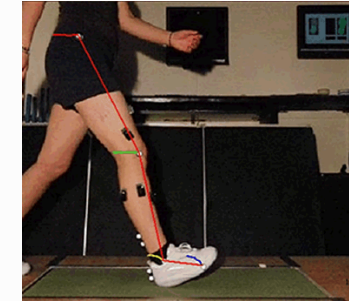
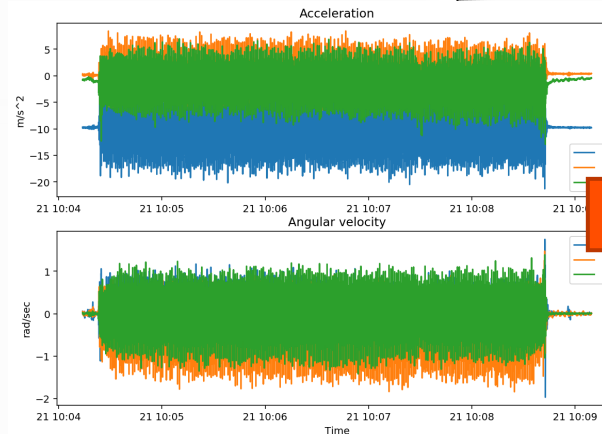
Article

Gait Characteristics Harvested during a Smartphone-Based Self-Administered 2-Minute Walk Test in People with Multiple Sclerosis: Test-Retest Reliability and Minimum Detectable Change

Alan K. Bourke ¹, Alf Scotland ², Florian Lipsmeier ¹, Christian Gossens ^{1,*} and Michael Lindemann ¹

¹ Roche Pharma Research and Early Development, pRED Informatics, Roche Innovation Center Basel, F Hoffmann–La Roche Ltd., Grenzacherstrasse 124, 4070 Basel, Switzerland; alan.bourke@roche.com (A.K.B.); florian.lipsmeier@roche.com (F.L.); michael.lindemann@roche.com (M.L.)
² Inovigate, Aeschenvorstadt 55, 4051 Basel, Switzerland; alf.scotland@inovigate.com
 * Correspondence: christian.gossens@roche.com; Tel.: +41-61-687-5113

- Pre-processing steps applied to perform appropriate signal processing techniques (filtering, resampling, sensor frame alignment)
- Straight line gait bout extracted, initial and termination steps & turns removed
- Initial and final contact points extracted and pendulum h delta
- Temporal, spatial and spatiotemporal gait characteristics exported



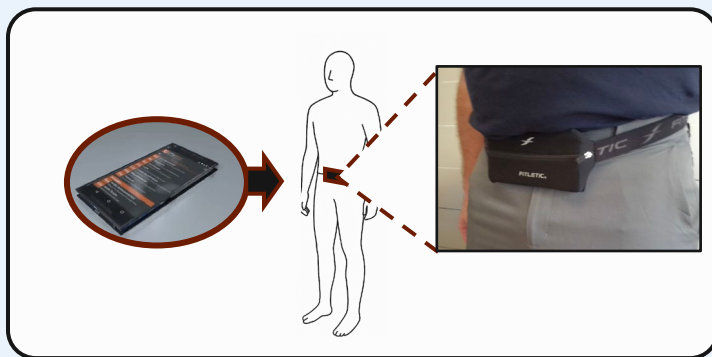
Step ID	Step length	Index	Step time	Step velocity	Step length
0000-01-01	0.0000	00	0.00	0.0000	0.00
0000-01-02	0.0000	01	0.00	0.0000	0.00
0000-01-03	0.0000	02	0.00	0.0000	0.00
0000-01-04	0.0000	03	0.00	0.0000	0.00
0000-01-05	0.0000	04	0.00	0.0000	0.00
0000-01-06	0.0000	05	0.00	0.0000	0.00
0000-01-07	0.0000	06	0.00	0.0000	0.00
0000-01-08	0.0000	07	0.00	0.0000	0.00
0000-01-09	0.0000	08	0.00	0.0000	0.00
0000-01-10	0.0000	09	0.00	0.0000	0.00
0000-01-11	0.0000	10	0.00	0.0000	0.00
0000-01-12	0.0000	11	0.00	0.0000	0.00
0000-01-13	0.0000	12	0.00	0.0000	0.00
0000-01-14	0.0000	13	0.00	0.0000	0.00
0000-01-15	0.0000	14	0.00	0.0000	0.00
0000-01-16	0.0000	15	0.00	0.0000	0.00
0000-01-17	0.0000	16	0.00	0.0000	0.00
0000-01-18	0.0000	17	0.00	0.0000	0.00
0000-01-19	0.0000	18	0.00	0.0000	0.00
0000-01-20	0.0000	19	0.00	0.0000	0.00

Analytical Validation: Reliability and Minimum detectable change

Remote smartphone-derived gait characteristics compared with established clinic-based walking tests

REMOTE SMARTPHONE 2MWT

Repeated, self-administered assessment at home



58 / 92

gait features had
excellent (ICC>0.9) or
good-to-excellent (ICC>0.75)
test-retest reliability

Temporal 33/48 | Spatial 11/22 | Spatiotemporal 14/22

46 / 64*

=> MDC% <20%

13 / 64*

=> MDC% <11%

* Distribution features = percentiles (e.g. 95th percentile), min, mean and max

- Most precise features (<11% MDC): Step time, Stride time, Swing time, Stance time.
- Notable features with MDC <20%: Stride velocity, Step velocity, Step length, Stride length.

CLINIC-BASED WALKING TESTS

Periodic standardized assessment



T25FW **36%** MDC = 2.7 s

6MWT **20%** MDC = 88 m

6MWT, EDSS <4 **11%** MDC = 58 m

6MWT, EDSS 4–6.5 **39%** MDC = 120 m

Learmonth et al. 2013; n=82, assessments 6 months apart

Smartphone 2MWT characteristics were repeatable, and many had detectable-change thresholds below published walking-test benchmarks.

Analytical Validation

Smartphone-based assessment of gait with Floodlight in multiple sclerosis shows high operational robustness and analytical validity against reference tool.

Napiorkowski N et al. ECTRIMS 2024; Poster P781.



Key Question

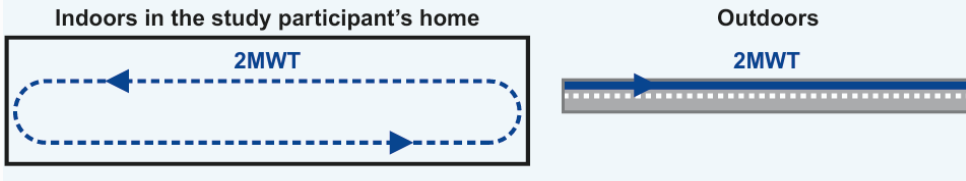
Can a waist-worn smartphone measure gait speed accurately and reproducibly?

Key Findings:

- ✓ **High agreement with the GaitUp reference system**
 - ICC(3,1)=0.91–0.96 across locations
- ✓ **Minimal systematic bias**
 - Remote front-belt bias: 0.03 m/s
- ✓ **Excellent reliability across six smartphone locations**
 - Belt: Front, Back | Pocket: Right front, Left front, Right back, Left back
- ✓ **Reliability improves with aggregation**
 - ICC increased from 0.88 (single test) to 0.98 (6 aggregated tests)
- ✓ **Supports robust unsupervised remote assessment**
 - High analytical validity in both laboratory and remote settings



Remote setting: Floodlight 2MWT performed remotely every day in a home environment for a 14-day period



Floodlight 2MWT

- Front belt

Reference system (Gait Up)

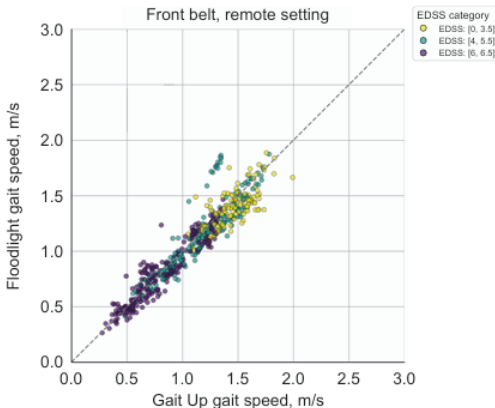
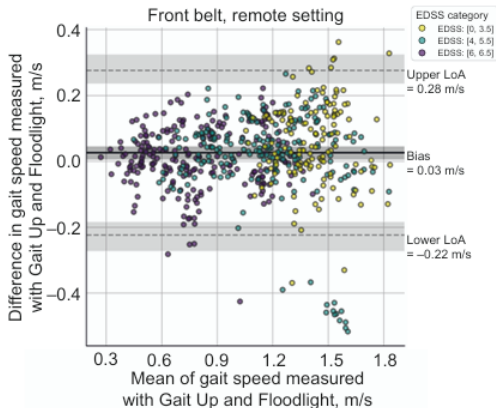
- Right shoe
- Left shoe



n=74, EDSS 4.8 ± 1.4

What is the Agreement Between Gait Speed Measured with the Floodlight 2MWT and Gait Speed Measured with Gait Up IMU Sensors in the Remote Setting?

- In the remote setting, the agreement between gait speed measured with the 2MWT and simultaneously with Gait Up IMU sensors remains high with minimal systematic bias



Analytical validity in the remote setting ^a	
Floodlight gait speed, m/s, median (IQR)	1.12 (0.54)
Gait Up gait speed, m/s, median (IQR)	1.15 (0.58)
ICC(3,1) (95% CI)	0.94 (0.91; 0.96)
Bias (95% CI)	0.03 (0.00; 0.05)
Lower LoA (95% CI)	-0.22 (-0.27; -0.18)
Upper LoA (95% CI)	0.28 (0.24; 0.33)

^aAnalytical validity was assessed for the front belt smartphone wear location only. 2MWT, Two-Minute Walk Test; EDSS, Expanded Disability Status Scale; ICC, intraclass correlation coefficient; IMU, inertial measurement unit; IQR, interquartile range; LoA, limits of agreement.

High level of absolute agreement and minimal systematic bias in the remote setting

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Key Question

Can smartphone-derived gait features measure meaningful disability and gait impairment in multiple sclerosis?

Key Findings:

✔ Strong correlation with:

- T25FW speed
- EDSS
- Ambulation Score

✔ Differentiates disability severity groups



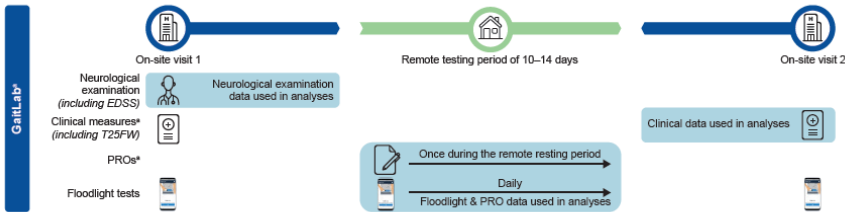
Smartphone-Based Assessment of Gait with Floodlight Shows Association with Clinical Measures of Disability and Gait Impairment in Multiple Sclerosis.

Festanti A et al. ECTRIMS 2024; Poster P786.

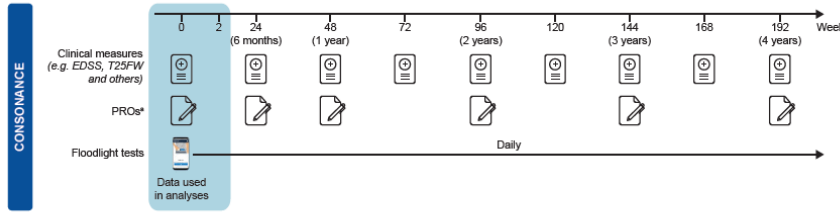


	GaitLab (n=87)	CONSONANCE (n=363)
Female, n (%)	56 (64)	191 (53)
Age, years, mean (SD)	54.3 (12.8)	48.5 (9.4)
EDSS, mean (SD)	4.8 (1.5)	5.0 (1.3)
MSWS-12 transformed total score, mean (SD)	47.2 (31.7)	62.7 (25.2)
T25FW time, s, mean (SD)	7.3 (5.2)	14.5 (16.4)

What is the Schedule of Assessments in the GaitLab and CONSONANCE Studies?



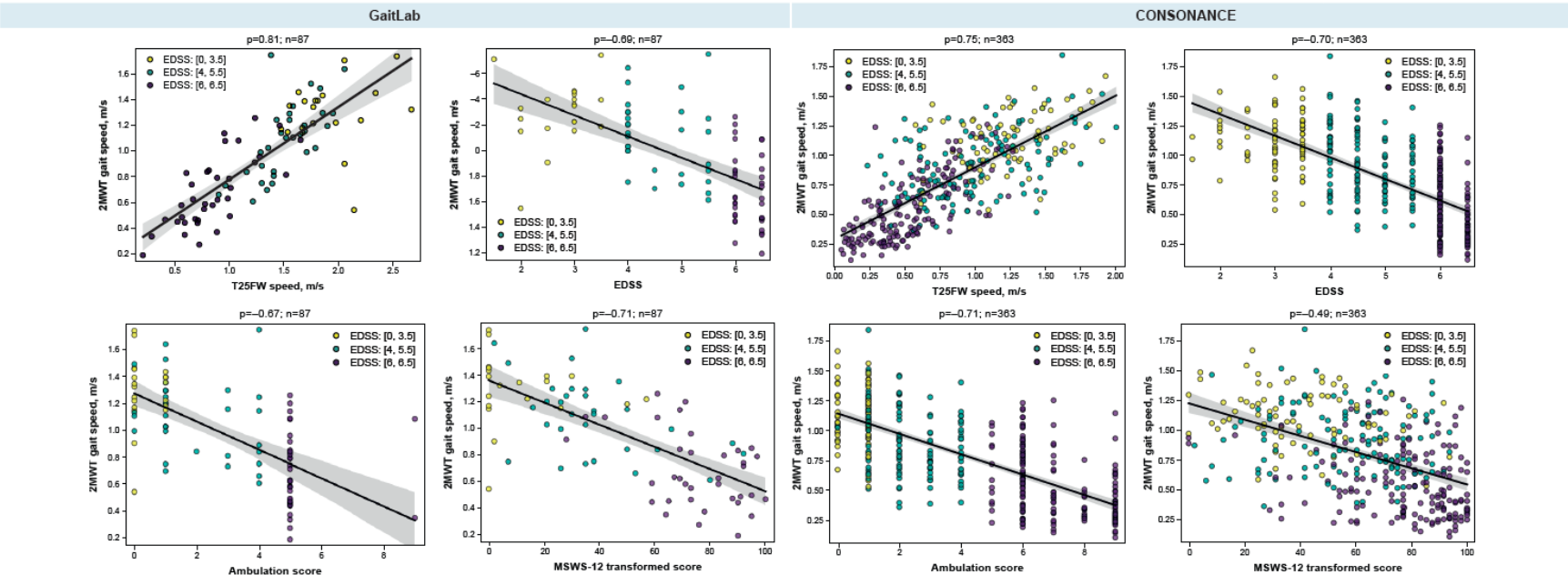
*Includes only assessments on gait; assessments of other domains, including hand motor function and cognition, are not shown.



*Includes only assessments on gait; assessments of other domains, including hand motor function and cognition, are not shown.

What Are the Correlations Between the Floodlight 2MWT and Clinical Measures of MS-Related Disability and Gait Impairment?

- In both studies, cross-sectional correlations between gait speed measured during remote 2MWTs and in-clinic T25FW speed, EDSS, MSWS-12 transformed score, as well as Ambulation score were moderate to strong, and in the expected direction



The strength of the correlations were defined as follows: |p| : 0.80–1.00 = very strong; |p| : 0.60–0.79 = strong; |p| : 0.40–0.59 = moderate; |p| : 0.20–0.39 = weak; |p| : 0.00–0.19 = very weak.

Clinical Validation

Key Question

Can smartphone-derived gait speed differentiate clinically meaningful levels of disability and gait impairment?

Key Findings:

✓ Distinguishes EDSS disability groups in GaitLab

- EDSS 0–3.5 | EDSS 4–5.5 | EDSS 6–6.5

✓ Distinguishes Ambulation groups in GaitLab

- No gait imp. | Mod. gait imp. | Sev gait imp.

✓ Replicated in CONSONANCE

- Same pattern observed in larger progressive-MS cohort

n = 363 PwPMS at baseline

✓ Known-groups validity demonstrated

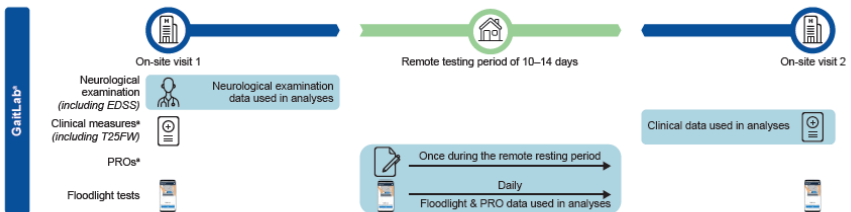
- Lower gait speed consistently associated with higher disability and greater walking impairment

Smartphone-Based Assessment of Gait with Floodlight Shows Association with Clinical Measures of Disability and Gait Impairment in Multiple Sclerosis.

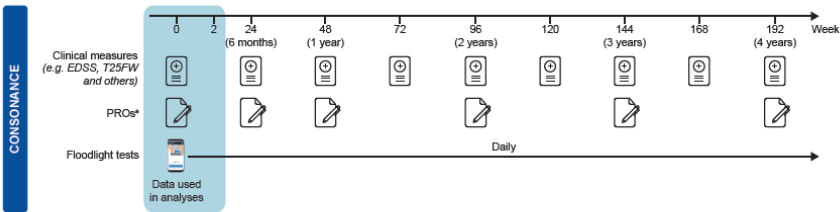
Festanti A et al. ECTRIMS 2024; Poster P786.



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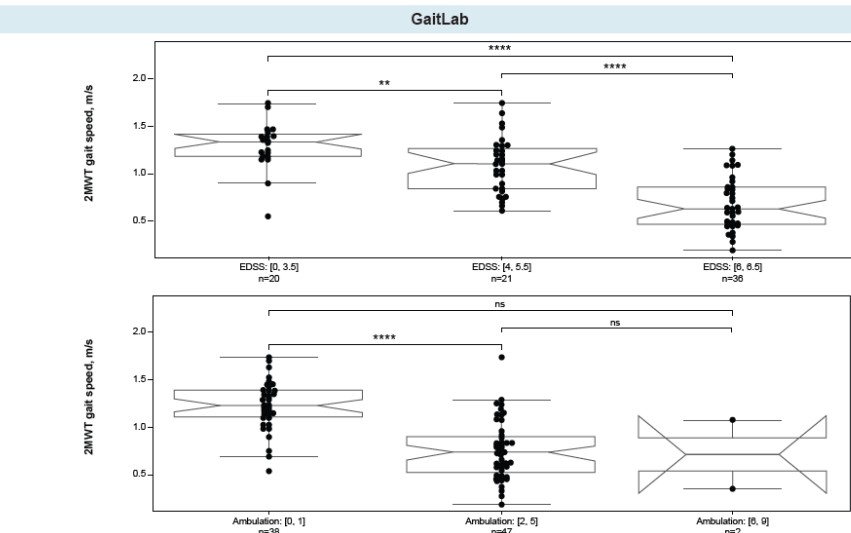
*Includes only assessments on gait; assessments of other domains, including hand motor function and cognition, are not shown.



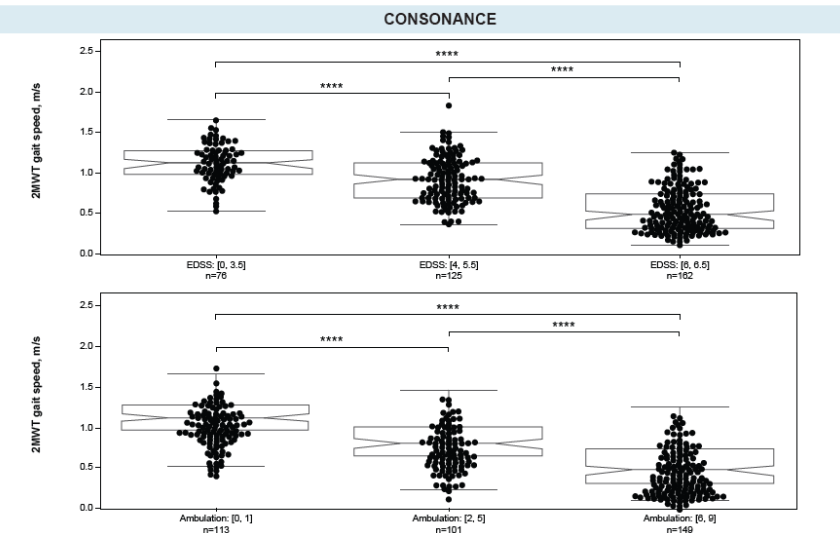
*Includes only assessments on gait; assessments of other domains, including hand motor function and cognition, are not shown.

Can the Floodlight 2MWT Differentiate Between PwMS with Different Levels of MS-Related Disability and Gait Impairment?

- In GaitLab, gait speed measured during remote 2MWTs showed differentiation between PwMS subgroups defined by EDSS or Ambulation scores
 - The small sample size of PwMS with an Ambulation score between 6–9 (n=2) likely explains the lack of a statistical significance to difference between these subgroups and the other Ambulation score subgroups
- The CONSONANCE study confirms the discrimination of gait speed



** p<0.01; *** p<0.001; **** p<0.0001.



Agenda

1  MS Gait Characteristics

2  Digital Endpoints

3 **V3+** Qualification Framework

4  Verification

5  Usability & Adherence

6  Analytical Validation

7  Clinical Validation

8  Longitudinal Sensitivity

9  Clinical Interpretability (MCID)

10  Conclusion

Sensitivity to Longitudinal Change

Smartphone-Based Assessment of Gait with Floodlight in Multiple Sclerosis Shows Sensitivity to Change Over a Follow-Up Period of 3 Years.

Rodrigues J et al. ECTRIMS 2024; Poster P1559.



Key Question

Can remote smartphone-derived 2MWT gait measures distinguish groups with and without EDSS progression compared to the In-Clinic T25FW?

Key Findings:

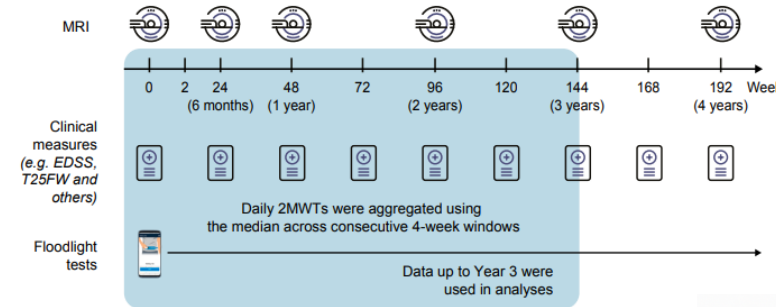
- ✓ Comparable discrimination to the in-clinic T25FW
- ✓ Captures changes in ambulation beyond gait speed
- ✓ Detects subtle worsening among EDSS non-progressors
- Year 3: 2MWT gait measures **-3% to -13%**
- T25FW speed: **<5% change**
- ✓ May be sensitive to change in PwMS with stable EDSS

Future work:

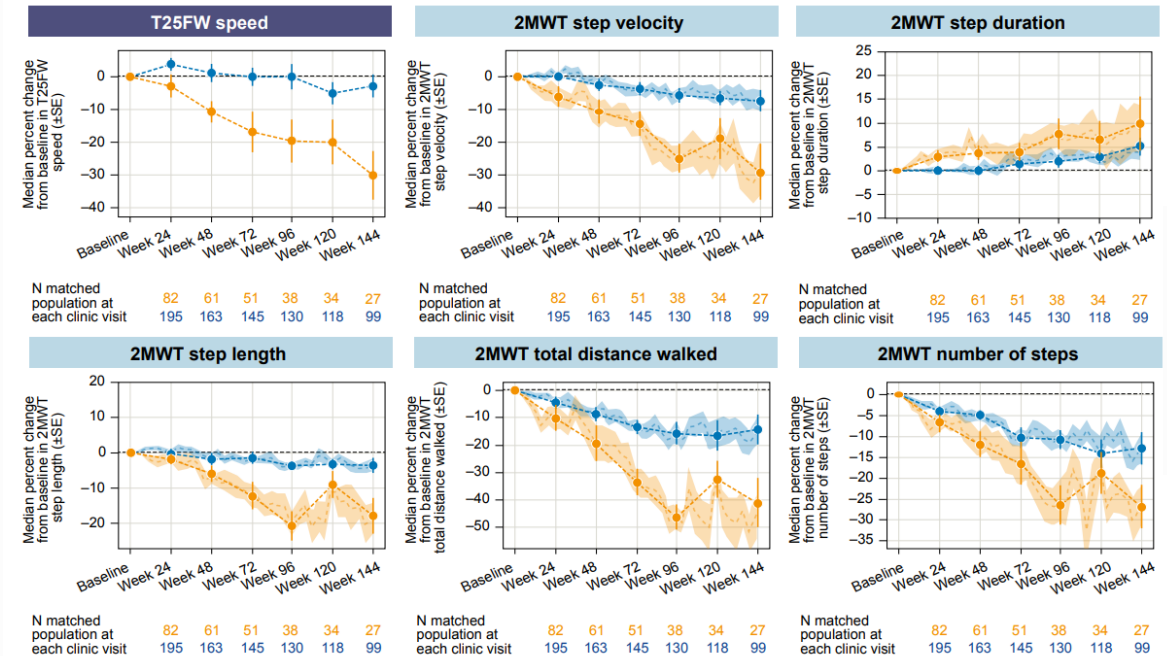
Supports further evaluation as a progressive-MS trial outcome

	Matched population at baseline n=368
Female, n (%)	193 (52)
Age, mean (SD), years	48.5 (9.5)
Diagnosis, n (%)	
PPMS	175 (48)
SPMS	193 (52)
EDSS score, mean (SD)	5.0 (1.3)
MSWS-12 transformed total score, mean (SD)	62.7 (25.2)
T25FW time, mean (SD), s	14.5 (16.3)

What Is the Schedule of Assessments?



Progressor versus non-progressor trajectories over 3 years



*The non-matched population includes all PwMS for whom 2MWT data at baseline period and at the respective clinic visit period are available, regardless whether or not T25FW data are available.

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Clinical Interpretability: From Detectable Change to Meaningful Change

Establishing the Minimal Clinically Important Difference (MCID) for Smartphone-Derived Gait Measures in Multiple Sclerosis.

Rinderknecht MD et al. arXiv. 2026



Anchor	Threshold
T25FW	≥20% worsening
Ambulation Score	≥1 point
EDSS	≥1 / 0.5 point
MSWS-12	≥8 points

Digital gait measure	MCID for worsening <i>Rinderknecht,2026,arXiv</i>	MDC _{50th} magnitude <i>Bourke,2020,Sensors</i>	Relationship
Step Velocity	-0.16 m/s	0.152 m/s	MDC < MCID
Step Velocity Scaled to Walk Time	-0.18 m/s	0.152 m/s	MDC < MCID
Step Duration / Time	+0.06 s	0.045 s	MDC < MCID
Step Length	-0.07 m	0.082 m	MDC > MCID
Total Number of Steps	-28	-	
Total Distance Walked	-24 m	-	

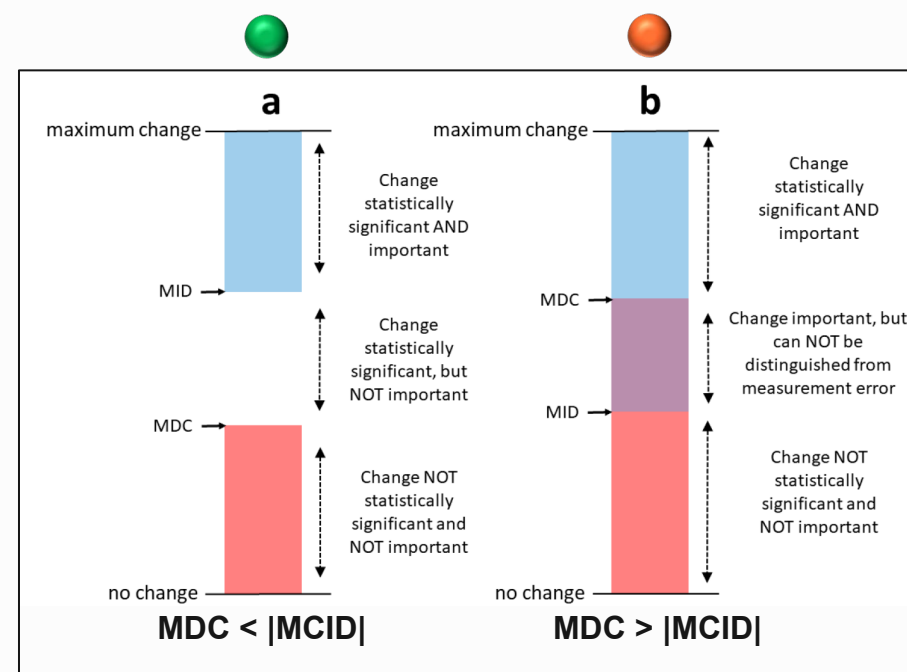
MCID signs indicate the direction of worsening; MDC is an unsigned measurement-error threshold. Comparisons therefore use absolute magnitudes.

Algorithm code: [GitHub - matt002/GaitPy](#)



The **MDC** is the minimum magnitude of change required to be 95% confident that the observed change between measurements is **true change** and **not measurement error**

The **MCID (MID)** is defined as the smallest change in the score of the construct to be measured, which is perceived as important by patients, clinicians, or relevant others



References: Bourke AK et al. Sensors. 2020;20(20):

Take-home message

Bourke,2020,Sensors quantified detectable change (MDC).
Rinderknecht,2026,arXiv quantified meaningful change (MCID). Together they provide the foundation for clinically interpretable digital gait endpoints.

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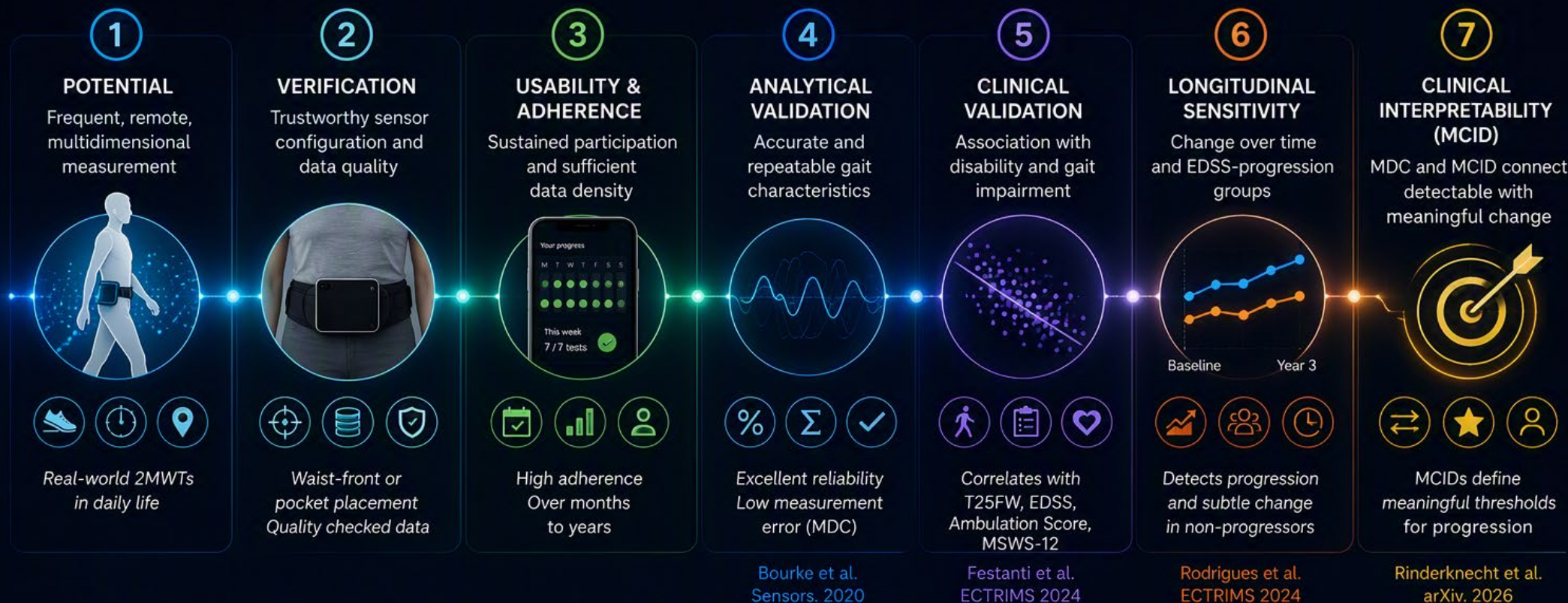
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From measurement to meaning.

A V3+-informed evidence pathway for smartphone-derived gait measures in MS



The evidence shows that smartphone-derived gait measures can be **usable**, **reliable**, **clinically valid**, **sensitive to progression** and **interpretable**.



Waist-front or pocket placement enables frequent, real-world measurement.

Less noise.
More signal.
Better decisions.

KEY ENABLERS



Secure & scalable infrastructure



Patient-centered design



Robust analytics & QA



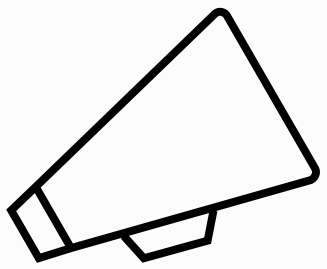
Regulatory-aligned frameworks (V3+)



Cross-sector collaboration



Advancing digital gait endpoints together for people living with MS



Thank you



A shout-out to the wider digital health technology community in Basel

CONSORTIUM

Mobilise-D

Developing and validating real-world digital mobility outcomes across conditions, including MS, with a pathway toward regulatory and clinical endorsement.

CONTINUED COLLABORATION

RC2NB

For continued collaboration in developing and validating digital biomarkers to characterize MS progression and support clinical translation.

BASEL ECOSYSTEM

Indivi

A Basel-based technology partner providing smartphone measurement platforms for neuroscience research and clinical development.

FOUNDATIONAL COLLABORATION

Roche - Digital Biomarkers

For Floodlight, gait algorithms, validation studies and the scientific foundation supporting this evidence journey.

CURRENT COLLABORATION

Novartis – AQS Neuroscience

For advancing analytics and quantitative science, and translating digital-measure evidence into fit-for-purpose drug-development applications.

Advancing digital biomarkers requires collaboration across academia, healthcare, technology and biopharma. Thank you to everyone helping move the field forward.

Alan Bourke | alan.bourke@novartis.com

Digital Biomarker Deployment for Gait Analysis in Multiple Sclerosis

Backup: Key References

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Bakker JP et al. V3+ extends the V3 framework to ensure user-centricity and scalability of sensor-based digital health technologies. *npj Digital Medicine*. 2025;8:51

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Rinderknecht MD et al. An observational study to assess validity and reliability of smartphone sensor-based gait and balance assessments in multiple sclerosis: Floodlight GaitLab protocol. *Digital Health*. 2023;9

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