

1 Digital health technologies to strengthen patient-centred outcome
2 assessment in clinical trials in inflammatory arthritis

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65 Summary

66 Common to all of the inflammatory arthritides, namely rheumatoid arthritis, psoriatic
67 arthritis, ankylosing spondylitis, and juvenile idiopathic arthritis, is a potential for reduced
68 mobility manifested through joint pain, swelling, stiffness and ultimately joint damage.
69 Across these conditions, consensus has been reached on the need to capture outcomes related
70 to mobility, such as functional capacity and physical activity as core domains in randomised
71 controlled trials. Existing endpoints in use within these core domains rely wholly on self-
72 reported questionnaires capturing patients' perception of their symptoms and activities which
73 are subjective, inherently vulnerable to recall bias and fail to capture the granularity of
74 fluctuations over time. There have been several early adopters integrating sensor-based
75 digital health technology-derived endpoints to measure physical function and activity in
76 randomised controlled trials for conditions including Parkinson's disease, Duchenne's
77 muscular dystrophy, chronic obstructive pulmonary disease, and heart failure. Despite these
78 applications, there have been no sensor-based digital health technology-derived endpoints in
79 clinical trials recruiting patients with inflammatory arthritis. Borrowing from case studies
80 across medicine, we outline the untapped opportunities and challenges in developing novel
81 sensor-based digital health technology-derived endpoints which capture the symptoms and
82 disease manifestations most relevant to patients with inflammatory arthritis.

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87 Introduction

88 The inflammatory arthritides (IA) represent a collection of autoimmune rheumatic diseases
89 including rheumatoid arthritis (RA), ankylosing spondyloarthritis (AS), psoriatic arthritis
90 (PsA), and juvenile idiopathic arthritis (JIA).¹ These conditions are characterized by intra-
91 articular inflammation which causes joint pain and stiffness as well as fatigue which can
92 follow a fluctuating course between episodic clinical encounters. Over time, particularly if
93 inflammation is not adequately controlled, patients with IA can progress to develop
94 irreversible articular and peri-articular damage contributing to impaired joint mobility and
95 limitations on individuals' ability to perform daily activities of living. This process can
96 contribute to the marked overall functional and psychosocial disability experienced by people
97 with IA.

98 The past 30 years have marked a substantial shift in the treatment and management of
99 inflammatory arthritis (IA), profoundly improving the outlook for patients. With a growing
100 number of treatments available that demonstrate the ability to improve disease activity and
101 reduce inflammation, clinical trials are increasingly focused on the effect of different
102 treatments on patient-relevant aspects such as mobility and physical function over time. This
103 is also reflected in the emergence of the universal goal of treatment across IA to preserve
104 physical function and social participation as a means of maximising health-related quality of
105 life (HRQoL).^{2,3} Despite this priority for both trials and clinical practice, assessment of such
106 aspects of disease is currently reliant on self-reported questionnaires, which may lack
107 sensitivity, are subject to recall bias and are burdensome for participants if used frequently.
108 The over-reliance on such tools raises concern that some treatments may be incorrectly
109 assessed as ineffective because of the deficiencies in, and poor performance of, current
110 assessment instruments.

111 Across medicine, sensor-based digital health technologies (sDHTs) are emerging as a
112 solution for the more objective assessment of patient-specific domains related to physical
113 function and mobility, and its downstream relative, physical activity. sDHT-derived
114 endpoints can capture a range of patient-centred concepts of interest and may broadly include
115 data generated; a) while a patient performs a specific task designed to measure key
116 symptoms, otherwise known as *active tasks* or b) through passive data acquisition monitoring
117 human behaviours, such as physical activity or sleep, or physiological signs, such as heart
118 rate or body temperature.⁴ Active tasks can be administered using any sDHT, however, this is
119 easiest on technologies with a screen for instructions, such as tablets and smartphones which
120 have in-built gyroscopes and accelerometers measuring angular motion and velocity whilst
121 passive data can be collected using a wide range of devices such as smartphones,
122 smartwatches, wearables, and smart patches. In the context of patients with IA, active tasks
123 could be deployed to capture individual joint range of motion tasks (i.e. a digitisation of the
124 modified Schober's test in ankylosing spondylitis) or overall functional capacity through tests
125 such as sit-to-stand or 6-Minute Walk Test. Passive wearable sensors, such as accelerometers,
126 can directly measure movement behaviours, such as physical activity which is implicitly
127 connected to musculoskeletal function. Sensor-based measurements can be collected
128 remotely over a continuous period or at higher frequency and granularity, rather than
129 restricting assessments only to intermittent clinic visits or through some remote electronic
130 patient-reported outcome measures (ePROs).

131 We aim to present the current landscape of outcome assessment in IA trials before moving on
132 to present the feasibility and validity of sensor-based digital health technologies in IA
133 populations from existing observational research. We will examine several case studies of
134 early adopters of sensor-based digital endpoints in trials across other areas of medicine to
135 explore both the validity for inflammatory arthritis and regulatory steps towards digital

136 endpoint adoption in inflammatory arthritis trials. The final section will serve as a call to
137 action and roadmap for the breadth of the rheumatology community to act in developing and
138 integrating meaningful, effective and equitable sDHT-derived endpoints into RCTs.

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140 Current deployment of patient-centred outcome measures in inflammatory
141 arthritis trials

142 A significant contributor to the successful trial landscape in rheumatology has been the
143 development of an extensive array of outcome measures. The Outcome Measures in
144 Rheumatology (OMERACT) initiative has established international consensus on the core
145 outcomes sets required in IA RCTs. Broadly, these core outcome sets reflect the need to
146 capture either the underlying pathophysiological process (disease activity) or the
147 manifestations of the condition on the individual (disease impact). In traditional efficacy
148 trials, the gold standard objective disease activity outcome has been demonstrating
149 differences in radiographic progression between placebo and treatment arms. There are
150 ethical concerns about using radiographic progression with evidence to suggest irreversible
151 joint damage and resulting functional limitation even from minor levels of radiographic
152 progression.^{5,6} In efforts to capture more of the heterogeneity of IA, OMERACT has endorsed
153 several disease-specific composite disease activity measures.⁷ There is emerging evidence
154 suggesting rheumatology RCTs are at high risk of unblinding and there is often reported
155 discordance between the physician and patient global domains, with patients reporting these
156 measures do not accurately reflect their disease state.^{8,9}

157 In efforts to increase patient-centricity in trial endpoints, OMERACT has unequivocally
158 endorsed the use of measures that quantify the impact of treatment on physical function and
159 health-related quality of life (HrQoL) across RCTs in IA. The only validated means of

160 measuring these core domains is through self-reported questionnaires, which are subject to
161 several limitations, including notable floor and ceiling effects and a lack of sensitivity to
162 change despite patients reporting disease improvement across various other domains.¹⁰⁻¹²

163 In patients with ankylosing spondylitis, there is only a moderate correlation between the
164 OMERACT-endorsed physical function outcome, the Bath Ankylosing Spondylitis
165 Functional Index (BASFI) and objective tests of physical function with patients appearing to
166 incorporate exertion and pain levels in the self-assessment of physical function on the BASFI
167 questionnaire.¹³ The limitations of self-reported questionnaires of physical function have
168 stimulated the development of more objective performance measures of physical function in
169 AS.^{14,15} The AS Performance Index (ASPI) is a clinic-based assessment which measures the
170 time to perform three daily activities (bending to pick 6 pencils from the floor, putting on
171 socks and standing up from the floor). The ASPI is still undergoing further validation prior to
172 deployment in RCTs in AS; however, it has shown adequate to excellent intra-rater test-retest
173 reliability, good responsiveness to treatment following TNF inhibition, and successfully
174 measures different aspects of physical function when compared to the BASFI
175 questionnaire.^{14,15} These findings point to a benefit in measuring physical function through
176 tests over simply collecting self-reported data. However, there are limitations with clinic-
177 based performance measures with evidence showing that they capture maximal effort and are
178 strongly influenced by participant motivation.¹⁶

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180 What is the feasibility and validity of sensor-based digital endpoints in
181 inflammatory arthritis?

182 Despite the lack of sensor-based digital health technology-derived endpoints in IA trials,
183 there is a significant body of observational research investigating the use of sDHTs in a

184 variety of settings for patients with IA. These studies have investigated a range of sDHTs for
185 measuring physical function and mobility, either deploying through active tasks completed by
186 the participant (e.g. using a smartphone gyroscope which captures angular motion to measure
187 specific joint range-of-motion or specific sit-to-stand or walking tasks to measure overall
188 physical function), through passive wearable data (e.g. to measure physical activity level in a
189 free-living environment), or a combination of these approaches and can serve as early
190 milestones highlighting the potential validity of sDHT-derived endpoints in IA trials.

191 *Active tasks*

192 The PARADE pilot study utilised inbuilt gyroscopes in a smartphone to assess a range of
193 active tasks related to a wrist range-of-motion task and a gait task to mobility at a multitude
194 of affected joints in RA.¹⁷ The authors found a relationship between reduced joint velocity in
195 the range-of-motion task with higher pain intensity levels.¹⁷ The ongoing Psorcast study,
196 developed by Sage Bionetworks and the Psoriasis and Psoriatic Arthritis Clinics Multi-center
197 Advancement Network (PPACMAN), is a prospective, observational cohort study recruiting
198 participants with PsA to complete a range of active tasks on the Psorcast mobile application.
199 This app leverages smartphone gyroscope and accelerometer-based active tasks in the form of
200 the “Digital Jar Open” and 30-second walking task to monitor upper extremity function and
201 overall physical function, respectively.¹⁸ The “Digital Jar Open” provides a continuous
202 variable outcome measure of overall upper limb angular rotation and as a single feature was
203 able to distinguish participants with and without physician-assessed arm joint tenderness and/
204 or enthesitis and has the potential to remotely capture granular changes of functional
205 impairment over time. There are other useful functional tests in inflammatory arthritis, such
206 as the modified Schober’s test or hand grip strength, which could be digitised to form a
207 remotely measurable active task, however, these are yet to be investigated. Active tasks can
208 also include the use of electronic patient-reported outcomes (ePROs) which are often

209 digitisations of existing outcome measures. There is a large body of evidence supporting the
210 feasibility, acceptability, and value of ePROs from observational studies across
211 rheumatology, notably, highlighting the benefit of increasing the frequency of capturing
212 symptoms over traditional measurements.¹⁹ Most of this research is generated in the realm of
213 remote monitoring vs standard clinical care and has been reviewed elsewhere in detail.¹⁹

214 *Passively collected wearables data*

215 Several other studies have investigated the relationship between accelerometer-assessed
216 measures of physical activity with IA-specific factors including disease activity, physical
217 function, and flare. Across PsA, two studies demonstrated an inverse relationship between
218 disease activity and objectively measured physical activity volumes.^{20,21} Hernandez-
219 Hernandez found physical activity to be longitudinally responsive to fluctuations in disease
220 activity across study visits.²⁰ In work by Prioreshi et al., treatment-naive RA patients who
221 responded to DMARD therapy showed statistically significant improvements in physical
222 activity volumes, with less sedentary and more light activity per day compared to non-
223 responders.²² In the ActConnect study, the authors leveraged physical activity data (steps per
224 minute) to identify flares assessed with weekly self-reports collected from 20 healthy controls
225 and 157 patients (83 with RA and 74 axial spondyloarthritis) over 3 months.²³ This study
226 demonstrated the association between self-reported flare onset and modest reductions in
227 physical activity volumes. The authors could predict a flare from accelerometer data alone
228 with a sensitivity and specificity of 96% and 97%, respectively.²³ Collectively, these findings
229 suggest that physical activity is closely related to the fluctuating course of disease activity in
230 IA.

231 *Integrated multi-modal approaches*

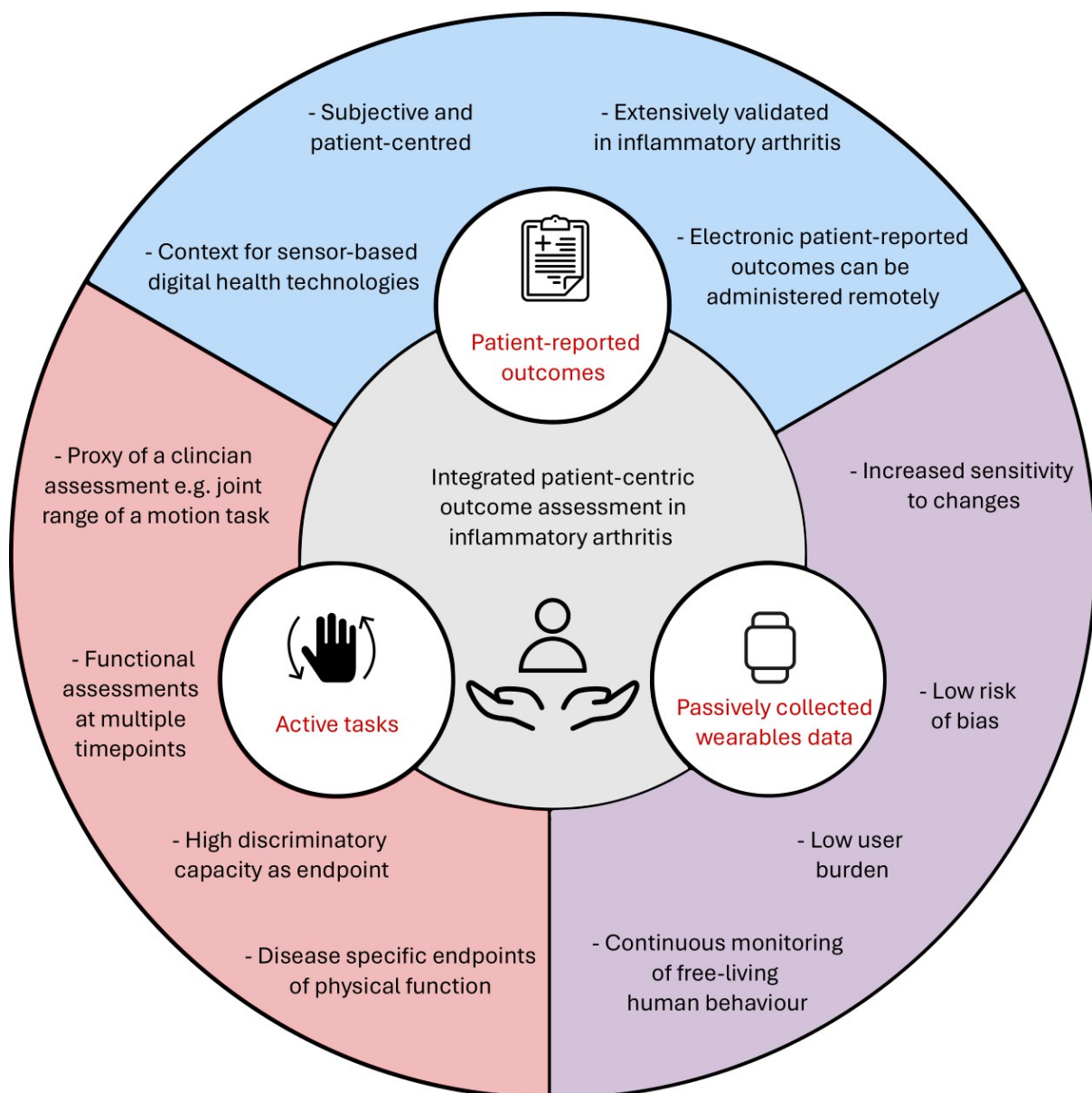
232 The previously outlined approaches have tended towards testing a singular measure derived
233 either from passively collected wearables data or active task data. An inherent limitation of
234 any singular domain endpoint is the risk of confounding from factors outside the condition of
235 interest. Overall physical activity may be affected both by the underlying pathophysiological
236 process in inflammatory arthritis but also by acute intercurrent illness or comorbid diseases
237 such as osteoarthritis, depression and anxiety and cardiometabolic diseases such as ischaemic
238 heart disease or diabetes, all of which are associated with lower physical activity levels.²⁴
239 There is evidence emerging from the development of digital endpoints in COPD and
240 Parkinson's disease around the utility of combining multiple data sources into a multi-modal
241 composite digital endpoint to reduce the risk of confounding when compared to a singular
242 measure.²⁵⁻²⁸ To this end, the data garnered from an sDHT approach are placed synergistically
243 alongside other patient-reported data to provide important context to the digital signal (as
244 demonstrated in the context of IA in figure 1). There has been one data-intensive
245 observational study testing the validity of a multi-modal DHT-based approach in capturing
246 disease activity in rheumatology. The WeaRAble-PRO study investigated how standard
247 PROs could be augmented with remote DHT sensor data, collected passively from
248 smartwatches and via active tasks deployed on smartphones in patients with RA.²⁹ These
249 tasks included ePROs as well as lying-to-standing mobility tasks using inbuilt gyroscopes to
250 collect data on the longitudinal course of RA, accurately identifying participants in low and
251 high disease activity states, as assessed by the RAPID-3 questionnaire.²⁹ A key insight from
252 the WeaRAble-PRO study is the potential for sDHT-derived endpoints to complement and
253 extend traditional PROs, offering a more nuanced view that encompasses both the subjective
254 symptoms perceived by patients and the objective measurements captured by sDHT.²⁹ This
255 perspective is crucial for patients with IA, where the impact extends beyond measurable

256 clinical signs to include fluctuating symptoms and higher comorbidity burden that can affect
257 daily life and well-being

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262 **Figure 1** - Summary of key advantages of sensor-based digital health technology-derived
263 endpoints and their integrated place alongside traditional patient-reported outcomes in the
264 comprehensive assessment of patient-centred domains in inflammatory arthritis trials.

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267 What is the uptake of digital endpoints outside of rheumatology?

268 There have been several early adopters of sensor-based digital health technology-derived
269 endpoints in clinical trials across other areas of medicine. At the time of writing, the DiMe
270 Library of Digital Endpoints reports 430 unique digital endpoints being used in industry-
271 sponsored trials of new medical products. However, none have been deployed in any trial
272 recruiting participants with a rheumatological condition.³⁰ Recent work by Masannek et al.
273 presented the growth trends of digital endpoints in use in neurology trials over the past
274 decade.³¹ The authors found that the relative frequency of clinical trials in neurology
275 implementing digital health technology-derived endpoints has increased from 0.7% in 2010
276 to 11.4% in 2020 with the majority of these endpoints measuring motor function.³¹ Across
277 neurology, the poor performance of subjective questionnaire-based assessment tools for
278 movement disorders has driven the development of digital endpoints for monitoring disease
279 progression in Parkinson's disease and Duchenne's muscular dystrophy (DMD).^{25,26,32} The
280 first European Medicines Agency (EMA) approval of a digital endpoint for *stride velocity*
281 *95th centile* (SV95C) in patients with DMD was motivated by challenges presented by
282 assessing physical function in clinic.³² DMD is a progressive neuromuscular disease
283 characterised by proximal muscle weakness and fatigability with patients often experiencing
284 significant burden to get to the clinic contributing to higher levels of fatigue confounding
285 assessment hence the need for capturing the measure in a free-living environment. In rare

286 neurodegenerative conditions, such as Friedrich's Ataxia and amyotrophic lateral sclerosis,
287 challenges around recruitment of sufficient sample sizes for RCTs has stimulated innovation
288 in developing DHT-derived physical function endpoints which are more sensitive at detecting
289 a minimal clinically important difference (MCID).^{33,34} Akin to these neurological conditions,
290 for patients with IA where mobility is so closely tied to functional capacity, these approaches
291 have significant validity.

292 Beyond neurology, RCTs assessing interventions for heart failure have traditionally utilised
293 efficacy endpoints that focus on mortality or hospitalisation. The availability of effective
294 treatments has also driven an earlier shift towards head-to-head active comparator study
295 designs when compared to rheumatology. As a result, this necessitates the design of
296 extremely large and costly trials (median sample size of the 25 RCTs supporting the chronic
297 heart failure guidelines was 2331 participants).³⁵ Furthermore, these endpoints give no
298 information on how a patient's ability to function may change in response to an intervention.
299 In response to this need to incorporate patient-centred perspectives on the symptoms and
300 functional ability of patients in heart failure trials, The Heart Failure Collaboratory, a public-
301 private partnership between regulators (US FDA), academia, payers, sponsors, and patients
302 with HF was formed with a focus for integrating accelerometry-derived endpoints in HF
303 trials.³⁶ A recent review by the Heart Failure Collaboratory highlights the 8 published RCTs
304 using accelerometry for primary or secondary endpoints and a further six ongoing trials
305 utilising this technology.³⁷ These trials have primarily focused on measuring change in daily
306 step counts or global physical activity levels in response to treatment. The external validity of
307 these studies to an IA population could be expanded due to the inherent relationship between
308 joint pain and stiffness causing mobility limitations and subsequent reduced physical activity
309 levels.

310

311 What are the regulatory considerations?

312 The development of digital endpoints is complex and includes several important steps leading
313 to validation and approval. Progress towards implementation of sensor-based digital health
314 technology-derived endpoints in clinical studies has been further reflected in recent
315 regulatory trends, as seen in the first qualification of a wearable-derived digital endpoint by
316 the EMA in the form of SV95C in patients with DMD,³² whilst the FDA has finalized
317 Guidance for industry regarding the use DHTs for remote data acquisition in clinical trials.³⁸
318 As these tools are being deployed to collect patient data to evaluate the safety and/or efficacy
319 of a medicinal product in trials, regulatory bodies such as the EMA and FDA must ensure
320 that measurements from these digital tools are clinically meaningful, representative, robust,
321 and valid.³⁹ These regulatory bodies are both encouraging the formation of and actively
322 participating in multistakeholder consortia with far-reaching domain expertise including
323 patient perspective, clinicians, engineers, computer scientists, information governance, and
324 industry to expedite DHT-derived endpoints that are valid and patient-centric.⁴⁰ Whilst some
325 inconsistencies exist between the EMA and FDA regulation process, both bodies have
326 endorsed the V3 Framework as a development process for any novel digital endpoint.^{38,41,42}
327 This first starts by framing key activities patients would wish to be able to complete but that
328 their condition limits. In the context of patients with inflammatory arthritis, physical function
329 or ability to perform daily ambulatory activities, is a key priority (Figure 2). From this,
330 stakeholders can define a measurable outcome and identify a suitable DHT to capture this
331 outcome. Once the need for a DHT is established, it is essential to demonstrate it is fit-for-
332 purpose for gathering data in a clinical trial, for example showing that it is well tolerated by
333 research and clinical volunteers.⁴² Verification evaluates the performance of a sensor
334 technology against pre-specified criteria to demonstrate the sample-level data generated is
335 accurate e.g. raw data from accelerometer at specific measurement frequency, consistent over

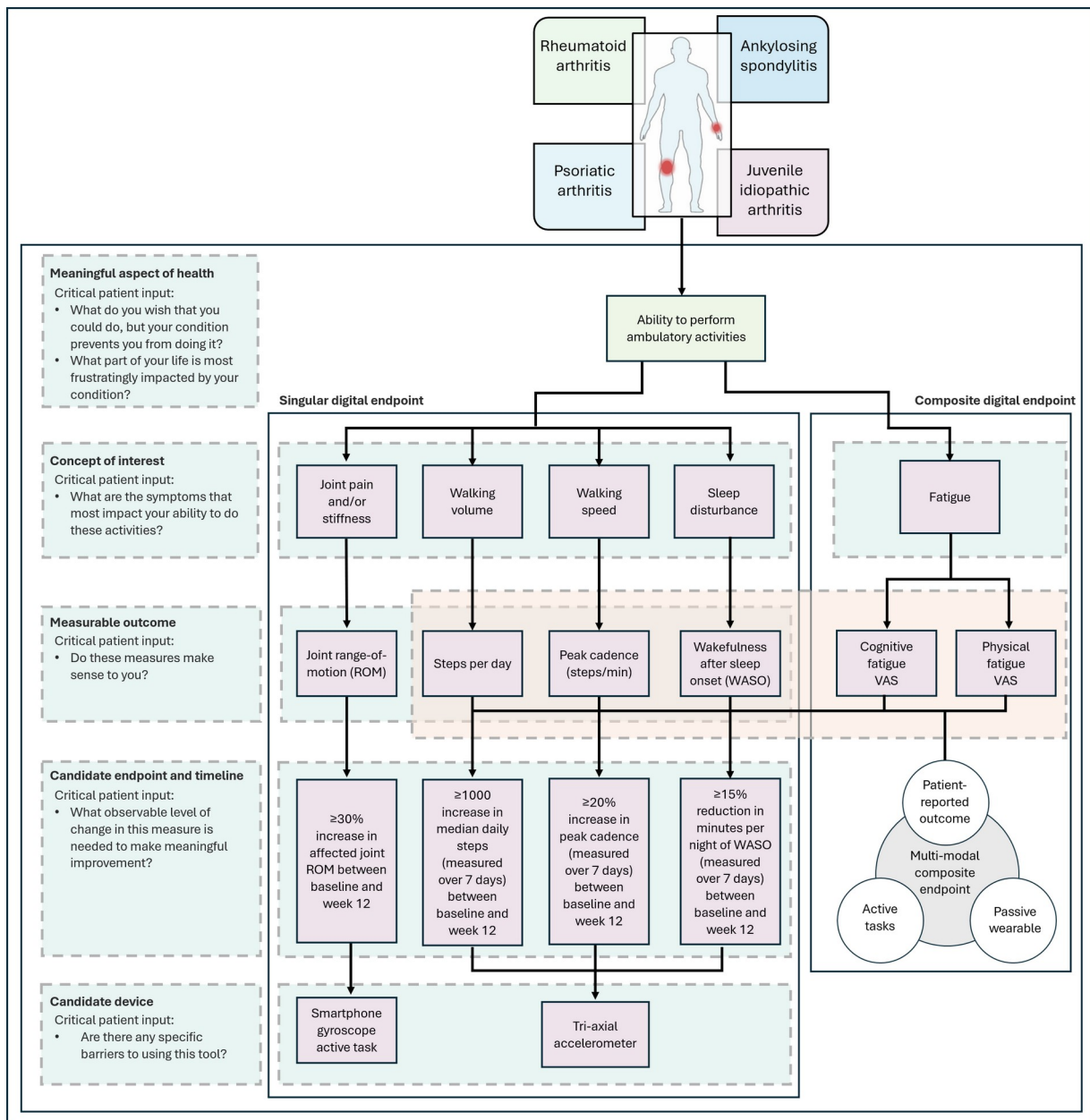
336 time, and uniform across different environmental bench testing conditions. Analytical
337 validation evaluates the performance of the sensor technology in human subjects, and its
338 ability to measure, detect, or predict physiological or behavioural metrics against the pre-
339 defined concept of interest. This step is agnostic of a specific clinical population of interest
340 and can recruit healthy volunteers. The aim is to prove that the outcome of interest, for
341 example, daily step count, is accurately captured when compared to a gold-standard measure,
342 in this case, observer confirmed steps. Finally, clinical validation tests a digital
343 measurement's predictive power in the context of its intended clinical use.⁴² This step
344 involves determining the optimal measurement window and frequency of sampling (e.g. days
345 vs weeks) for a measure considering peak sensitivity, specificity, and reliability of a measure
346 alongside user compliance and burden. This pipeline is an iterative, multi-study approach
347 integrating patients, engineers, clinicians, industry partners, and regulatory bodies. This
348 process can take several years to execute highlighting the need for timely engagement from
349 the rheumatology community to realise the potential of these tools to capture meaningful
350 measures for patients with IA in rheumatology RCTs.

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356 **Figure 2 - A pipeline for the development patient-centric sensor-based digital endpoints**
 357 **for patients with inflammatory arthritis. (Adapted from Manta et al).⁴³** Measurable
 358 outcomes and resulting candidate endpoints outlined are hypothetical and purely illustrative
 359 of workflow. Specific validation studies are required to define the observable level of change
 360 as anchored to existing measures of physical function or health-related quality of life.
 361 Importantly, each step requires iterative patient input before progressing to the next step
 362 within a multi-study development pipeline. VAS; visual analogue scale.

364 Considerations regarding validity of digital endpoints in rheumatology

365 Whilst the implementation of sensor-based digital health technologies offers a means of
 366 obtaining objective, high-resolution data from patients captured during their daily lives, their

367 use also poses some challenges. Large and diverse validation datasets are needed to avoid
368 algorithmic biases and in turn the inaccurate collection and interpretation of digital health
369 data from certain demographic groups.⁴⁴ Furthermore, clinical validation studies for sDHT-
370 endpoints in IA anchored to existing disease activity and physical function measures need to
371 recruit groups of participants that are diverse and representative. Other important issues
372 include variable data quality, often stemming from improper wearing and/or a lack of
373 compliance with DHT hardware, as well as the psychological impact of wearable devices.
374 Sleep data from sensory devices can skew patients' perception of their quality of sleep, and
375 in turn, reinforce sleep-related anxiety (i.e. orthosomnia).⁴⁵ In the context of inflammatory
376 arthritis where individuals may have disease manifestations affecting their fine-motor
377 movements of the hands, it is imperative to ensure that any adjustments to the sensor or
378 display are reasonably developed with these considerations so as to reduce the risk of digital
379 exclusion. In pursuit of developing robust and valid digital endpoints in IA, special attention
380 should be paid to whether candidate measurements are specific IA alone rather than being
381 global indicators of illness. In developing sDHT-endpoints in juvenile idiopathic arthritis,
382 understanding the important confounding influence of age on physical activity level needs to
383 be considered and any endpoint needs to be validated within specific age subgroups to
384 account for changes in stride length and velocity. Altogether, the concerns outlined exemplify
385 the importance of well-designed technical and clinical validation studies for a given DHT,
386 involving patients with diverse demographic characteristics. Trials in inflammatory arthritis
387 have systemically underrepresented and underrecruited patients from racial and ethnic
388 minority groups.⁴⁶ Whilst not explored in detail in this review, these endpoints offer
389 opportunities for remote data capture and a shift towards decentralised trial designs.
390 Following a patient-centric approach to endpoint development and integrating patient
391 partners at each stage of the development pipeline (figure 2) offers the opportunity to reduce

392 biases and increase equity in trials for people with inflammatory arthritis. This process can
393 also be formalised through specific Health Equity Impact Assessment (HEIA) processes
394 which aim to identify and address any risk of digital exclusion.^{47,48}

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396 What role does machine learning play?

397 The field of digital endpoint adoption in clinical trials is rapidly evolving. Recent regulatory
398 guidance from the FDA on the use of DHTs in endpoint development did not address the role
399 of machine learning. Despite this, machine learning is becoming ever more widely used to
400 interpret the highly granular data collected using sensor-based digital health technologies.
401 Through machine learning, it becomes possible to discern intricate patterns and relationships
402 within the data, which might remain undetected with conventional statistical approaches. By
403 learning from large volumes of data, foundation models have been shown to improve the
404 reliability of digital measures across populations, devices, and modalities, critical for the
405 deployment of DHTs in clinical trials.⁴⁹ Particularly relevant is the value of machine learning
406 in multi-modal sDHT approaches that enable the development of novel digital markers,^{33,34}
407 and more precise monitoring of disease progression.²⁹ Machine learning techniques allow for
408 the integration and analysis of diverse data sources, including active tasks, accelerometer-
409 measured physical activity, and patient-reported outcomes. This could be particularly
410 important in the context of developing composite sDHT-derived endpoints, akin to a digital
411 endpoint of fatigue proposed in figure 2. Whilst machine learning holds immense promise in
412 this field, there are important considerations to ensure its effective implementation. Key
413 considerations include promoting model explainability, ensuring transparency and
414 reproducibility, establishing robust governance infrastructures, training models using relevant
415 and representative clinical validation datasets and addressing bias.^{50,51}

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417 What is the overall set of actions we need to enact as a community to realise this
418 vision in an IA population?

419 Through OMERACT there exists an infrastructure that integrates academics, industry,
420 regulators, and patient partners to develop and endorse core outcomes for implementation in
421 rheumatology RCTs. A specific OMERACT Digital Working Group with the view to
422 considering meaningful aspects of health across IA that could be measured using sensor-
423 based digital health technologies is imperative to integrate these outcomes in IA trials. There
424 have been strides taken to create core outcome sets related to digital measures of physical
425 activity and mobility which would serve as a focus for such a group.⁵² The development of
426 sDHT-derived endpoints is a lengthy and intricate process, underscoring the urgency for
427 immediate action to initiate this vital journey toward revolutionizing patient-centric outcomes
428 in clinical trials in inflammatory arthritis.

429 In the pursuit of developing digital endpoints in rheumatology, the model set out by the
430 ambitious IDEA-FAST and MOBILISE-D consortia should be followed closely.^{53,54} These
431 consortia consist of patient partners, academia, and industry stakeholders aiming to develop
432 cross-disease endpoints based on meaningful aspects of health which are conserved across a
433 multitude of diseases, for example, fatigue being relevant in neurodegenerative diseases such
434 as Parkinson's disease as well as immune-mediated inflammatory diseases such as RA or
435 Crohn's disease.⁵³ This cross-disease approach aims to accelerate device verification and
436 endpoint validation steps with both IDEA-FAST and MOBILISE-D having received positive
437 support for this approach from the EMA.^{53,55} Further to this cross-disease approach, recent
438 work by Bertha et al. proposes a framework building on the V3 verification and validation
439 pipeline which permits leveraging verification evidence from a prior related disease area and

440 application to a new related use case.⁵⁶ The work from IDEA-FAST, MOBILISE-D, and
441 Bertha et al. have significant relevance for the rheumatology community given there is a high
442 degree of harmony in patient-centred concepts of interest across each of the IA, potentially
443 enabling cross-disease validation studies of DHT-derived endpoints which could accelerate
444 the development process.

445

446 Conclusion

447 The integration of sensor-based digital health technology-derived endpoints holds immense
448 promise for rheumatology, offering the potential to revolutionize assessment in IA trials.
449 Whilst the journey towards digital endpoint development in rheumatology may be
450 challenging, it is paved with a wealth of valuable case studies from across medicine. Drawing
451 from these exemplars, there is a robust foundation to build upon, offering a clear pathway for
452 the successful integration of digital endpoints supporting shorter, more cost-effective, and
453 patient-centred RCTs that result in new therapies that address the symptoms most important
454 to all individuals with IA.

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459 Search strategy and selection criteria

460 This narrative review was supported by a literature search in PubMed and Google Scholar,
461 focusing on publications in English from 2010 to 2024. Search terms included “inflammatory
462 arthritis”, “rheumatoid arthritis”, “ankylosing spondylitis”, “psoriatic arthritis”, “juvenile

463 idiopathic arthritis”, “outcome measures”, “patient-reported outcome measures”,
464 “OMERACT”, “digital health technology”, “digital endpoint”, “digital health” and
465 “randomized controlled trials”. Additional publications were identified from references using
466 a pearl-growing approach.

467 Conflicts of interest statement

468 HY has received a personal payment via University of Oxford for software licenses he has
469 developed. APC; this work was conducted while APC was an employee of University of
470 Oxford. During this time, APC was partially funded by GSK plc. and has worked on projects
471 related to GSK. APC has never had a contract of employment, owned any stocks or shares,
472 equity, or had a named position on a company board associated with GSK. GSK has no
473 involvement or affiliation with this review work, in any kind. On submission of this work,
474 APC is currently an employee and shareholder of Sanome Ltd. Sanome has no involvement
475 or affiliation with this review work, in any kind. APC was involved in initial idea generation,
476 conceptual work of this review, while he was still employed by the University of Oxford,
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517 References

518 Ledingham J, Snowden N, Ide Z. Diagnosis and early management of inflammatory arthritis.

519 *BMJ (Online)* 2017; **358**. DOI:10.1136/bmj.j3248.

520 Smolen JS, Aletaha D, Bijlsma JWJ, *et al.* Treating rheumatoid arthritis to target:

521 Recommendations of an international task force. *Ann Rheum Dis* 2010; **69**.

522 DOI:10.1136/ard.2009.123919.

523 Smolen JS, Schöls M, Braun J, *et al.* Treating axial spondyloarthritis and peripheral

524 spondyloarthritis, especially psoriatic arthritis, to target: 2017 update of recommendations by

525 an international task force. *Ann Rheum Dis* 2018; **77**: 3–17.

526 Taylor KI, Staunton H, Lipsmeier F, Nobbs D, Lindemann M. Outcome measures based on

527 digital health technology sensor data: data- and patient-centric approaches. *NPJ Digit Med*

528 2020; **3**: 97.

529 van der Heijde D, Landewé R, van Vollenhoven R, Fatenejad S, Klareskog L. Level of

530 radiographic damage and radiographic progression are determinants of physical function: a

531 longitudinal analysis of the TEMPO trial. *Ann Rheum Dis* 2008; **67**: 1267–70.

532 van der Heijde D, Landewé R. Should radiographic progression still be used as outcome in
533 RA? *Clinical Immunology* 2018; **186**: 79–81.

534 Day J, Antony A, Tillett W, Coates LC. The state of the art—psoriatic arthritis outcome
535 assessment in clinical trials and daily practice. *Lancet Rheumatol* 2022; **4**: e220–8.

836 Bruggemeyer C, Nepal D, Putman M. Unintentional unblinding in rheumatic disease trials.
537 *Lancet Rheumatol* 2023; **5**: e633–6.

938 Desthieux C, Hermet A, Granger B, Fautrel B, Gossec L. Patient-Physician Discordance in
539 Global Assessment in Rheumatoid Arthritis: A Systematic Literature Review With Meta-
540 Analysis. *Arthritis Care Res (Hoboken)* 2016; **68**: 1767–73.

501 Maska L, Anderson J, Michaud K. Measures of functional status and quality of life in
542 rheumatoid arthritis: Health Assessment Questionnaire Disability Index (HAQ), Modified
543 Health Assessment Questionnaire (MHAQ), Multidimensional Health Assessment
544 Questionnaire (MDHAQ), Health Assessment Questionnaire II (HAQ-II), Improved Health
545 Assessment Questionnaire (Improved HAQ), and Rheumatoid Arthritis Quality of Life
546 (RAQoL). *Arthritis Care Res (Hoboken)* 2011; **63 Suppl 11**: S4-13.

547 Orbai A-M, Ogdie A. Patient-Reported Outcomes in Psoriatic Arthritis. *Rheum Dis Clin*
548 *North Am* 2016; **42**: 265–83.

529 Højgaard P, Klokke L, Orbai A-M, *et al*. A systematic review of measurement properties of
550 patient reported outcome measures in psoriatic arthritis: A GRAPPA-OMERACT initiative.
551 *Semin Arthritis Rheum* 2018; **47**: 654–65.

532 van Weely SFE, van Denderen JC, Steultjens MPM, *et al*. Moving instead of asking?
553 Performance-based tests and BASFI-questionnaire measure different aspects of physical
554 function in ankylosing spondylitis. *Arthritis Res Ther* 2012; **14**. DOI:10.1186/ar3765.

545 van Weely SFE, Dekker J, Steultjens MPM, *et al.* Objective evaluation of physical
556 functioning after tumor necrosis factor inhibitory therapy in patients with ankylosing
557 spondylitis: a selection of 3 feasible performance-based tests. *J Rheumatol* 2015; **42**: 623–9.

558 van Bentum RE, Ibáñez Vodnizza SE, Poblete de la Fuente MP, *et al.* The Ankylosing
559 Spondylitis Performance Index: Reliability and Feasibility of an Objective Test for Physical
560 Functioning. *J Rheumatol* 2020; **47**: 1475–82.

561 Alfano LN, Lowes LP, Berry KM, *et al.* T.P.1: Pilot study evaluating motivation on the
562 performance of timed walking in boys with Duchenne muscular dystrophy. *Neuromuscular*
563 *Disorders* 2014; **24**: 860.

564 Hamy V, Garcia-Gancedo L, Pollard A, *et al.* Developing Smartphone-Based Objective
565 Assessments of Physical Function in Rheumatoid Arthritis Patients: The PARADE Study.
566 *Digit Biomark* 2020; **4**: 26–44.

567 Webster DE, Haberman RH, Perez Chada LM, *et al.* Clinical validation of digital biomarkers
568 and machine learning models for remote measurement of psoriasis and psoriatic arthritis.
569 *medRxiv* 2022.

570 Marques A, Bosch P, de Thurah A, *et al.* Effectiveness of remote care interventions: a
571 systematic review informing the 2022 EULAR Points to Consider for remote care in
572 rheumatic and musculoskeletal diseases. *RMD Open* 2022; **8**. DOI:10.1136/rmdopen-2022-
573 002290.

574 Hernández-Hernández MV, Sánchez-Pérez H, Luna-Gómez C, Ferraz-Amaro I, Díaz-
575 González F. Impact of Disease Activity on Physical Activity in Patients With Psoriatic
576 Arthritis. *Arthritis Care Res (Hoboken)* 2021; **73**. DOI:10.1002/acr.24422.

577 McGagh D, McGowan N, Hinds C, Saunders KEA, Coates LC. Actigraphy-derived physical
578 activity levels and circadian rhythm parameters in patients with psoriatic arthritis:

579 relationship with disease activity, mood, age and BMI. *Ther Adv Musculoskelet Dis* 2023; **15**:
580 1759720X231174989.

581 Prioreshi A, Hodkinson B, Tikly M, Mcveigh JA. Changes in physical activity measured by
582 accelerometry following initiation of dmard therapy in rheumatoid arthritis. *Rheumatology*
583 (*United Kingdom*) 2014; **53**. DOI:10.1093/rheumatology/ket457.

584 Gossec L, Guyard F, Leroy D, *et al*. Detection of Flares by Decrease in Physical Activity,
585 Collected Using Wearable Activity Trackers in Rheumatoid Arthritis or Axial
586 Spondyloarthritis: An Application of Machine Learning Analyses in Rheumatology. *Arthritis*
587 *Care Res (Hoboken)* 2019; **71**. DOI:10.1002/acr.23768.

588 Barker J, Smith Byrne K, Doherty A, *et al*. Physical activity of UK adults with chronic
589 disease: cross-sectional analysis of accelerometer-measured physical activity in 96 706 UK
590 Biobank participants. *Int J Epidemiol* 2019; **48**: 1167–74.

591 Jha A, Menozzi E, Oyekan R, *et al*. The CloudUPDRS smartphone software in Parkinson’s
592 study: cross-validation against blinded human raters. *NPJ Parkinsons Dis* 2020; **6**: 36.

593 Lipsmeier F, Taylor KI, Postuma RB, *et al*. Reliability and validity of the Roche PD Mobile
594 Application for remote monitoring of early Parkinson’s disease. *Sci Rep* 2022; **12**: 12081.

595 Jha A, Espay AJ, Lees AJ. Digital Biomarkers in Parkinson’s Disease: Missing the Forest for
596 the Trees? *Mov Disord Clin Pract* 2023; **10**: S68–72.

597 Garcia-Aymerich J, Puhan MA, Corriol-Rohou S, *et al*. Validity and responsiveness of the
598 Daily- and Clinical visit-PROactive Physical Activity in COPD (D-PPAC and C-PPAC)
599 instruments. *Thorax* 2021; **76**: 228–38.

600 Creagh AP, Hamy V, Yuan H, *et al*. Digital health technologies and machine learning
601 augment patient reported outcomes to remotely characterise rheumatoid arthritis. *NPJ Digit*
602 *Med* 2024; **7**: 33.

803 Digital Medicine Society (DiMe). Digital Medicine Society (DiMe) Library of Digital
604 Endpoints. <https://dimesociety.org/get-involved/library-of-digital-endpoints/>. 2023; published
605 online Aug.

806 Masannek L, Gieseler P, Gordon WJ, Meuth SG, Stern AD. Evidence from
607 ClinicalTrials.gov on the growth of Digital Health Technologies in neurology trials. *NPJ*
608 *Digit Med* 2023; **6**. DOI:10.1038/s41746-023-00767-1.

809 Servais L, Camino E, Clement A, *et al*. First Regulatory Qualification of a Novel Digital
610 Endpoint in Duchenne Muscular Dystrophy: A Multi-Stakeholder Perspective on the Impact
611 for Patients and for Drug Development in Neuromuscular Diseases. *Digit Biomark* 2021; **5**:
612 183–90.

833 Gupta AS, Patel S, Premasiri A, Vieira F. At-home wearables and machine learning
614 sensitively capture disease progression in amyotrophic lateral sclerosis. *Nat Commun* 2023;
615 **14**: 5080.

846 Ricotti V, Kadirvelu B, Selby V, *et al*. Wearable full-body motion tracking of activities of
617 daily living predicts disease trajectory in Duchenne muscular dystrophy. *Nat Med* 2023; **29**.
618 DOI:10.1038/s41591-022-02045-1.

859 Docherty KF, Campbell RT, Jhund PS, Petrie MC, McMurray JJ V. How robust are clinical
620 trials in heart failure? *Eur Heart J* 2017; **38**: 338–45.

881 Psotka MA, Abraham WT, Fiuzat M, *et al*. Functional and Symptomatic
622 Clinical Trial Endpoints: The HFC-ARC Scientific Expert Panel. *JACC Heart Fail* 2022; **10**:
623 889–901.

824 Anouché K, Elharram M, Oulousian E, *et al*. Use of Actigraphy (Wearable Digital Sensors
625 to Monitor Activity) in Heart Failure Randomized Clinical Trials: A Scoping Review. *Can J*
626 *Cardiol* 2021; **37**: 1438–49.

627 U.S. Food and Drug Administration (FDA). Digital Health Technologies for Remote Data
628 Acquisition in Clinical Investigations. 2023.

629 Colloud S, Metcalfe T, Askin S, *et al.* Evolving regulatory perspectives on digital health
630 technologies for medicinal product development. *NPJ Digit Med* 2023; **6**: 56.

631 Stephenson D, Alexander R, Aggarwal V, *et al.* Precompetitive Consensus Building to
632 Facilitate the Use of Digital Health Technologies to Support Parkinson Disease Drug
633 Development through Regulatory Science. *Digit Biomark* 2020; **4**: 28–49.

634 Clinical Trials Transformation Initiative (CTTI). Developing Novel Endpoints Generated by
635 Digital Health Technology for Use in Clinical Trials. 2022.

636 Goldsack JC, Coravos A, Bakker JP, *et al.* Verification, analytical validation, and clinical
637 validation (V3): the foundation of determining fit-for-purpose for Biometric Monitoring
638 Technologies (BioMeTs). *NPJ Digit Med* 2020; **3**: 55.

639 Manta C, Patrick-Lake B, Goldsack JC. Digital Measures That Matter to Patients: A
640 Framework to Guide the Selection and Development of Digital Measures of Health. *Digit
641 Biomark*. 2020; **4**. DOI:10.1159/000509725.

642 Zinzuwadia A, Singh JP. Wearable devices-addressing bias and inequity. *Lancet Digit Health*
643 2022; **4**: e856–7.

644 Baron KG, Abbott S, Jao N, Manalo N, Mullen R. Orthosomnia: Are Some Patients Taking
645 the Quantified Self Too Far? *J Clin Sleep Med* 2017; **13**: 351–4.

646 Strait A, Castillo F, Choden S, *et al.* Demographic Characteristics of Participants in
647 Rheumatoid Arthritis Randomized Clinical Trials: A Systematic Review. *JAMA Netw Open*
648 2019; **2**: e1914745.

649 Were MC, Sinha C, Catalani C. A systematic approach to equity assessment for digital health
650 interventions: case example of mobile personal health records. *J Am Med Inform Assoc* 2019;
651 **26**: 884–90.

482 Richardson S, Lawrence K, Schoenthaler AM, Mann D. A framework for digital health
653 equity. *NPJ Digit Med* 2022; **5**: 119.

494 Yuan H, Chan S, Creagh AP, *et al.* Self-supervised learning for human activity recognition
655 using 700,000 person-days of wearable data. *NPJ Digit Med* 2024; **7**: 91.

506 U.S. Food and Drug Administration. Good Machine Learning Practice for Medical Device
657 Development: Guiding Principles. 2021.

518 European Medicines Agency (EMA). Reflection paper on the use of Artificial Intelligence
659 (AI) in the medicinal product lifecycle. 2023.

520 Digital Medicine Society (DiMe). The core set of digital measures of physical activity.
661 <https://datacc.dimesociety.org/digital-measures-physical-activity/all-core-measures/>. 2024;
662 published online Feb 15.

533 Nobbs D, Piwko W, Bull C, *et al.* Regulatory Qualification of a Cross-Disease Digital
664 Measure: Benefits and Challenges from the Perspective of IMI Consortium IDEA-FAST.
665 *Digit Biomark* 2023; **7**: 132–8.

546 Mikolaizak AS, Rochester L, Maetzler W, *et al.* Connecting real-world digital mobility
667 assessment to clinical outcomes for regulatory and clinical endorsement-the Mobilise-D study
668 protocol. *PLoS One* 2022; **17**: e0269615.

559 European Medicines Agency (EMA). Letter of support for Mobilise-D digital mobility
670 outcomes as monitoring biomarkers. 2020.

571 Bertha A, Alaj R, Bousnina I, *et al.* Incorporating digitally derived endpoints within clinical
672 development programs by leveraging prior work. *NPJ Digit Med* 2023; **6**: 139.
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