

Review

Domotics, Smart Homes, and Parkinson's Disease

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Abstract. Technology has an increasing presence and role in the management of Parkinson's disease. Whether embraced or rebuffed by patients and clinicians, this is an undoubtedly growing area. Wearable sensors have received most of the attention so far. This review will focus on technology integrated into the home setting; from fixed sensors to automated appliances, which are able to capture information and have the potential to respond in an unsupervised manner. Domotics also have the potential to provide 'real world' context to kinematic data and therapeutic opportunities to tackle challenging motor and non-motor symptoms. Together with wearable technology, domotics have the ability to gather long-term data and record discrete events, changing the model of the cross-sectional outpatient assessment. As clinicians, our ultimate goal is to maximise quality of life, promote autonomy, and personalisation of care. In these respects, domotics may play an essential role in the coming years.

Keywords: Domotics, smart home, technology, unsupervised monitoring, Parkinson's disease

BACKGROUND

Parkinson's disease (PD) is a neurodegenerative condition with widespread social and economic implications [1]. As with all chronic diseases, effective, patient-centred, and equitable systems for monitoring and management are desirable [2].

Chronic neurological conditions have experienced a digital revolution over the last decade [3]. Several aspects make PD an excellent candidate for the integration of technology into routine clinical care [4]. First, there is a lack of validated diagnostic and disease progression biomarkers for PD, and hence

there is a reliance on clinical assessment. Second, the heterogenous clinical manifestations of PD demand a personalised approach to care. Finally, although PD is a generally progressive disorder, daily variation of symptoms is a norm experienced by many patients. The timing of medication, dietary choices, and psychological factors can influence the clinical examination findings. Gross motor fluctuations, which occur in many patients, are a source of even greater variability during the disease course. As such, it is difficult to get an accurate picture of a patient's current status from a single outpatient consultation [5]. For these reasons, unsupervised evaluation of patients over longer periods of time, ideally in their home environment, could help us to better understand the complexity, diversity, and true functional implications of PD [6].

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Although there has been substantial progress with regards to digital technology in PD, the focus to date has been on wearable devices and smart phone apps, or sophisticated sensors in dedicated laboratories [7]. Better use of digital technology could be implemented at home to support day to day management, and this need has never been greater than it has been during the coronavirus pandemic [8]. Even when there are no restrictions on attendance to hospital, home assessment can be used to supplement traditional face-to-face visits, or provide information on vulnerable patients or those that have busy working lives [9].

A Movement Disorders Society (MDS) Task Force on Technology recently published a roadmap to facilitate the integration of digital technologies in healthcare systems [10]. Their strategy was based on four areas: target domains, means of assessment, open and integrated display platforms, and regulated commercialisation. A lot of progress has been made in the design and development of home integrated tools. Now is time to study their potential applications for the care of patients with PD. As has been the case for wearables and apps, technological evolution risks outpacing clinical testing and implementation. To our knowledge, apart from the guidance by the MDS Task Force on Technology, there are no validated standards of assessment for domestic technology. Creation of such guidance for regulation and clinical use is necessary [7]. This review will focus on domestic integrated devices connected to the internet, otherwise known as 'domotics'. We will summarise the potential applications, current challenges, and future directions.

DEFINITIONS

The term domotics comes originally from the Latin 'domus' which means house and 'tics' which includes robotics, telematics, and computational science. Domotics are not new; the first 'smart house' was designed by the French engineer Pierre Sarda in 1974 (<https://youtu.be/cqPsIIYBSgc>).

Domotics, smart homes, and home automation are often used as interchangeable terms and describe the integration of technology and appliances to maximise well-being and function in the home environment [11]. From a healthcare perspective, they are not only designed for increasing comfort, security, and autonomy of patients, but can also be a rich source of continuous data [3, 4]. While domotics were

originally created for automating tasks, the range of possibilities, alongside internet connectivity, could hugely improve understanding and management of PD, leading to optimised clinical decision making [3].

POTENTIAL APPLICATIONS

Motor symptoms

Technology can provide real-world information that is difficult to obtain from a brief clinical consultation [6, 7]. Most of the current research in motor symptoms, including the cardinal signs and motor complications, has been centred on using body-worn sensors (for use either in free-living settings or dedicated movement laboratories), smartphone apps or other domestic hardware, such as measuring typing patterns using computer keyboards [12–17].

There are potential advantages to be gained through combining wearable technology with fixed sensors integrated in the home (such as video cameras, or sensors of movement, temperature, and pressure) to contextualise patterns of movement in the home environment. This helps capture the global clinical picture and provide feedback to users, caregivers, and clinicians about patient-relevant endpoints [16–18]. Additionally, voice-controlled lights, automated electrical appliances, and smart beds, may offer tangible benefits to patients with disabling symptoms [6] (see Fig. 1).

Falls detection along with the identification of precipitating factors such as sudden OFF periods and freezing of gait (FOG), are potential examples for how domotics may be used. Falls are one of the most challenging aspects of PD to treat, with limited responsiveness to medication. They are frequently encountered during the course of PD progression, and the cause of falls may be obscure; ranging from postural hypotension, gait impairment (including freezing) and postural instability [19]. Currently, falls that do not lead to hospital attendance, tend to be tracked by patients and caregivers in diaries. However, diaries are often not reliable, with a tendency for under-reporting, and a lack of clarity about fall mechanisms. Technology has gone some way to address these limitations mainly through wearable sensors and smartphone technology, but most research has been centred on describing patterns of movements in PD rather than exploring potential therapeutic interventions and preventive measures [16, 17, 20, 21]. As a detection system, domotics might help to interpret kinetic data from wearable sensors and ambulation

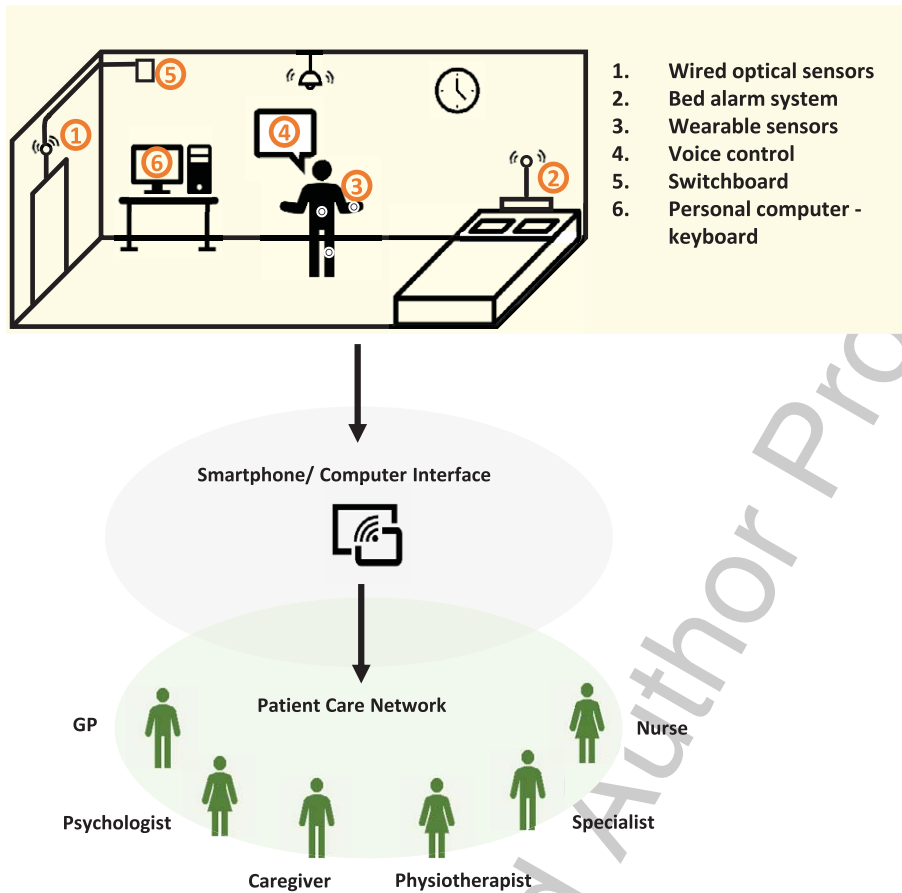


Fig. 1. Multi-sensor system integrated at home connecting people with PD with their health care network. 1) Wired optical sensors able to detect patient interaction with home environment and request switchboard access to emergency contacts. 2) Bed alarm system connected to a pressure sensor able to detect vigorous movements during sleep (RBD), time spent in bed (apathy/depression marker), wandering at night. 3) Wearable sensors interconnected with (1), (2), (5), and (6). 4) Voice control to home appliances. 5) Switchboard when fall is detected by (1) or voice operated (4). 6) Patient interaction with computer: typing (bradykinesia) and internet browsing or shopping (ICD).

148 monitoring devices to quantify and characterise falls
149 or FOG in the home environment [16]. Imagine a
150 smart home capable of tracking movement which
151 could immediately assist patients during FOG and
152 release an external cue when it occurred, such as shin-
153 ing a light on the floor or playing music at a given
154 tempo [21, 22]. Similarly, for sudden OFF periods,
155 when patients are alone, speech recognition systems
156 designed for controlling household devices and auto-
157 mated connection to the internet could be reassuring
158 [23]. Patients would be able to contact caregivers,
159 clinicians or emergency support. The net effect of
160 this increased connectivity is that patients could feel
161 more secure at home, while simultaneously relieving
162 caregiver burden.

163 Domestic entertainment appliances and virtual
164 reality could be used in home physiotherapy

165 programmes to improve balance and gait perfor-
166 mance in people with PD. For example, in a
167 study using a Nintendo® Wii the authors demon-
168 strated that 20 sessions of balance training for 5
169 days a week improved balance and gait perfor-
170 mance [24]. The authors suggested that continuous
171 visual feedback may facilitate movement execution
172 and maintain focused attention. The fact that it
173 was self-administered in the home facilitated long
174 term compliance. Another clinical trial with a ran-
175 domised, controlled design measured the feasibility
176 of home-based training using a smartphone app
177 (CuPiD-system) which provided real-time feedback
178 to patients. The investigators studied the effects on
179 gait in people with PD, finding that it was well
180 tolerated and easy to use. Despite a limited follow-
181 up period, patients experienced a positive effect on

182 their balance and quality of life [21]. A separate
 183 double-blind randomised controlled trial found that
 184 'gamifying' exercises using virtual reality had bene-
 185 fits on mobility [25].

186 *Non-motor and neuropsychiatric symptoms*

187 Cognitive impairment is a milestone of disease pro-
 188 gression in PD [26]. It has a huge impact on the extent
 189 of disability and caregiver burden [27]. An important
 190 consequence of cognitive impairment in PD is that the
 191 therapeutic window narrows; most drugs used to treat
 192 neuropsychiatric symptoms of PD can worsen motor
 193 symptoms. To-date most remote technology tools for
 194 dementia have been studied in the Alzheimer's dis-
 195 ease field, but the potential benefits for PD are also
 196 clear [28]. Beyond tracking of movement, motion
 197 sensors in smart houses could be used to analyse
 198 behavioural patterns. Episodes of disorientation and
 199 confusion, patients wandering or leaving the house
 200 at unusual times of the day, and the amount of time
 201 spent in bed are all examples of information that could
 202 be extracted from combining domotics and wearable
 203 sensors [29]. Collateral information from relatives
 204 and caregivers is crucial for understanding the needs
 205 of patients with dementia, but at early stages and for
 206 patients living alone this information can be difficult
 207 to obtain. By monitoring domestic tasks, early detec-
 208 tion of cognitive impairment or behaviour change
 209 could be possible, even before symptoms are noticed
 210 by others.

211 There are other neuropsychiatric symptoms which
 212 are under-reported by patients. Impulse control dis-
 213 orders (ICD) and apathy might be detectable based
 214 on abnormal day and night-time behavioural patterns
 215 such as spending long hours in front of the com-
 216 puter, performing repetitive tasks or staying in bed
 217 during the daytime. This could be used to detect ICD
 218 in patients or monitor treatment response for apathy
 219 and depression.

220 Sleep quality has mainly been studied using accel-
 221 erometers and gyroscope worn at the wrists or on the
 222 trunk [30, 31]. Patients with REM sleep Behaviour
 223 Disorder (RBD) act out their dreams due to a lack of
 224 muscle atonia during REM sleep. Physical safeguards
 225 may be employed, but technology could support diag-
 226 nosis or offer a therapeutic intervention for RBD.
 227 For example, Howell and colleagues designed a bed
 228 sensory-alarm system to prevent sleep related injuries
 229 in medically refractory RBD patients. They found
 230 their method to be an effective measure to pre-
 231 vent injuries in RBD as an alternative for medically

refractory patients or those who did not tolerate med-
 232 ication [32].
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Patient empowerment

234 We have provided some examples of how domotics
 235 might support patients as disease milestones loom.
 236 Overall, what technology, particularly domestic tech-
 237 nology, may offer is reassurance and empowerment
 238 of patients. Mobile technologies including wear-
 239 able sensors, smartphones, and domestic-integrated
 240 devices can work together to provide patients with
 241 feedback about their symptoms [5]. This digital
 242 health pathway could integrate patients, caregivers,
 243 and clinicians in a network model centred on person-
 244 alised care in which patients have a proactive role
 245 in decision making and feel more confident with the
 246 management of their symptoms [33]. Having an inte-
 247 grated model also offers the possibility of connecting
 248 automatically or through voice command with care-
 249 givers and emergency services if an unexpected event
 250 occurs. This offers further reassurance to the care-
 251 givers of more vulnerable people and a greater sense
 252 of security [28].
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254 The concept of health literacy is emerging and
 255 comprises the process of patient education regarding
 256 their condition [34]. Internet and home-environment
 257 monitored data can be an important source of infor-
 258 mation to enable effective self-management which
 259 will hopefully be demonstrable through improved
 260 quality of life [9].

CHALLENGES

261 There are several limitations to consider when
 262 gathering and interpreting digital health data which
 263 we have summarised here [5, 6].
 264

Data privacy and ownership

265 The nature of recording aspects of daily life brings
 266 legal and ethical issues [35]. Although domotics have
 267 a potential role in helping to understand the needs and
 268 functional status of the most vulnerable patients, the
 269 amount of data and the images that result from record-
 270 ing can threaten individual privacy [36]. Data sharing
 271 is necessary for the cross-validation and interpreta-
 272 tion of data from technology-based tools. Whether
 273 gathered for research or clinical care, data about
 274 patients in their home environment must be treated
 275 in the same way as other confidential information
 276 and governed by data protection laws. Issues around
 277

Table 1
Ethical issues and possible solutions

Ethical concerns	Approach
Privacy issues: intrusive surveillance sensors, unwanted image data, third parties involvement	Data encryption (blur, pixelating, silhouettes, skeleton, 3D avatar) to protect identity [36]
Loss of autonomy: feeling of lack of data control involving private life content	<ul style="list-style-type: none"> • Written consent after detailed information disclosure • To inform about rights: to view and delete unwanted images, temporarily pause image recording whenever they wish • Cognitively impaired individuals: consent given by people with decision-making authority anticipating benefits and risks • Participants to ask third parties for consent
Security issues: full reliance on technology, sensor failure to detect a dangerous situation, software hacking	<ul style="list-style-type: none"> • Technology demystification • Glitches detection • Trained investigators
Data ownership: right of self-management of personal data	<ul style="list-style-type: none"> • Support regulatory bodies • Testable quality standards certification [10]

consent and not infringing on autonomy, even when intentions are good, are important considerations and we must be vigilant about conflicts of interest [37].

There are several considerations which can be divided into privacy and confidentiality, threats to autonomy, safety issues ('do not harm' principle), and the boundaries of data ownership. Table 1 summaries the most relevant issues with examples and possible solutions based on two ethical guidelines designed for digital health research [35] and home environment technology for people with dementia [28]. The main principle is to focus on the interests of the patient above the interests of research and industry. In the research setting, IRB (Institutional Review Board) approval is mandatory for any clinical study involving patients and provides important safeguards. Whilst the guiding principles of data confidentiality are ubiquitous in many countries, the interpretation of such guidance varies and must be considered. Country-specific evaluation will be required for devices before regulatory approval is granted and this is an important aspect of implementing new technology [28].

Motor considerations

Hyperkinetic movements, such as tremor or dyskinesia, have characteristic patterns in accelerometer data, but other features such as bradykinesia can be misinterpreted through unsupervised assessments [5]. When motion sensors detect slowness or lack of movement it is not necessarily due to bradykinesia, but may also be seen with fatigue, pain, and apathy. Fixed sensors, as part of a domotic setup, could help to contextualise movement patterns suggested from accelerometry data.

Spontaneous physical activity captured by remote, unsupervised devices involves a great amount of

background noise and high variability between individuals [38]. Coexisting factors such as performing multiple tasks simultaneously, interference from other people, and domestic obstacles can confound data interpretation. Again, this limitation could partially be addressed by combining domotic devices with wearable sensors.

Uptake and implementation

The technology era has not been embraced by all and a substantial proportion of patients are reluctant to adopt new technology. The coronavirus pandemic has helped to increase the acceptability of technology as an alternative means of providing clinical information. Further research and consideration of the utility of domotics has never been timelier.

Setting up domotics into a private environment like someone's home could be considered intrusive for many and may be a limitation compared to wearable technology and apps. There is also more setup time involved given the need to take account of room layout, furniture configuration and individual requirements. One might expect that over time patients will be increasingly comfortable with technology compared with the current elderly population, and as such, acceptability will improve gradually (Raghunath et al., unpublished data).

Feasibility and usability studies are essential to understand compliance and comfort. The SENSE-PARK study assessed a quantitative assessment (wearable sensor, app, balance board, and computer software) of PD symptoms. As a primary outcome the number of dropouts were quantified. Secondly, feedback from participants regarding usability was evaluated using a Post-Study System Usability Questionnaire (PSSUQ) [15]. All patients completed the

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348 12-week study, provided good feedback and high-
 349 lighted the user-friendly design. A study on long
 350 term feasibility of wearable sensors in PD sug-
 351 gested that having a 'helpdesk' improved adherence
 352 of participants which would be worth considering for
 353 designing further digital-health protocols [39].

354 *Clinical validation and relevance*

355 Unsupervised outcomes need to be validated
 356 against standard measures, such as disease severity
 357 rating scales or diaries. However, inter-rater variabil-
 358 ity in case of disease rating scales and self-reporting
 359 biases related to diaries need to be considered when
 360 these are used as a 'gold' standard measures [40].
 361 Thus, using test-retest repeatability and accuracy
 362 could be a better way for validating information
 363 from domotic setups. However, the validation of a
 364 home sensor system is challenging on its own and is
 365 subject to patient factors such as variation in symp-
 366 toms and awareness of being constantly observed
 367 (known as 'The Hawthorne' effect [41]), and environ-
 368 ment factors, such as the variability of home layouts.
 369 Distinct context (supervised vs unsupervised) and
 370 different raw data (accelerometry vs video images)
 371 will demand the creation of validation standards to be
 372 used across different studies. Increasing the number
 373 of participants and raters, including assessment bat-
 374 tery with diaries, telephone calls and the use of other
 375 devices with data filters could be possible solutions
 376 to overcome these issues and improve the quality of
 377 validation studies and ensure results are not device
 378 dependent [42]. Another way to potentially improve
 379 the power of the study is expanding the amount of data
 380 collected through continuous monitoring [10]. It is
 381 important to bear in mind that large quantities of data
 382 or "big data", does not necessarily mean "good data".
 383 Although there is expansion in the use of sophisticated
 384 artificial intelligence and deep learning algorithms,
 385 these in themselves generate challenges and depend
 386 on the quality of the underlying data [43, 44].

387 **FUTURE PERSPECTIVE: PD CARE AND** 388 **RESEARCH**

389 In contrast to wearable sensors and smartphone
 390 applications, clinical trials of home-based technol-
 391 ogy, especially for people with PD, are limited. More
 392 feasibility and acceptability studies are needed to
 393 identify patient-relevant endpoints which will guide
 394 the design of clinical trials. Home-based sensors offer
 395 the opportunity to analyse a wide range of outcomes:

Take-home messages

- Domestic integrated devices connected to the internet (domotics) go beyond portable sensors, providing context to real-time and highly granular information.
- Integrated multisensory systems at home can be used to assist and prevent falls. They can also be used as a source of automate cueing delivery to treat FOG.
- The study of behavioural patterns in a home environment is a promising area of research with potential applications in early detection of dementia and monitoring ICDs.
- Digital medicine in combination with traditional medical care can help to empower patients and relieve caregiver burden.
- There are several limitations to tackle in the future: privacy implications, heavy and complex data (unsupervised, heterogenous, subject to external interferences), and restricted applicability in non-technology literate users.

disease progression markers, therapeutic interven- 396
 tions for a specific symptom (freezing, falls), and 397
 monitoring of treatment response and side effects 398
 which initially could only be used as a surrogate 399
 markers, but in the future might be even used as 400
 primary outcomes [10]. 401

We can learn from similar studies done in demen- 402
 tia and aging. The Oregon Centre for Aging & 403
 Technology (ORCATECH) is a multi-disciplinary 404
 organization focused on developing cutting-edge 405
 technologies to measure real-life data (<https://www.ohsu.edu/oregon-center-for-aging-and-technology>). 406
 The Collaborative Aging (in Place) Research Using 407
 Technology (CART) is an initiative which is part of 408
 ORCATECH platform and has a decade of experi- 409
 ence in technology for aging and Alzheimer's disease 410
 [45]. Data was gathered from multiple sources of 411
 information such as sensors in the home, in the car, 412
 and worn on the person. This system was iteratively 413
 tested and embedded into to 232 homes across the 414
 USA for 3.5 years. Cognitive performance, physical 415
 mobility, sleep duration, and social interaction 416
 were used as outcome measures. Another example 417
 is The HomeAssist project which developed an 418
 assisted living platform at the home of the elderly. 419
 A multi-disciplinary approach (geriatrics, psychol- 420
 ogists, caregivers, and users) was essential to 421
 identify user needs from a variety of perspectives. 422
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424 Thirty-two dyads were monitored for 6 months:
 425 half of them were equipped with the HomeAssist
 426 platform and the other half did not. Overall, their
 427 findings showed potential applications in home
 428 support and reducing burden on caregivers [46].

429 CONCLUSIONS

430 Future directions will be centred on develop-
 431 ing multi-disciplinary digital platforms, connecting
 432 patients, carers, and clinicians [18]. More research
 433 is necessary and there is a need to share and com-
 434 bine data on a large scale to train recognition systems
 435 and classification methods to identify a wide range of
 436 movement signatures [47].

437 Domotics have the ability to increase autonomy,
 438 self-management, and provide security, whilst pro-
 439 viding data about functional status over time. These
 440 are crucial aspects of the shift towards precision and
 441 personalised care for PD patients.

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 448 has not been published before nor is being considered
 449 for publication in another journal.

450 CONFLICT OF INTEREST

451 The authors have no conflict of interest to report.

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