Review of Person-Generated Health Data.

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What data are generated from using the system?		
Reference	Description	
Performance Data (N=25)		
[22,36-39,46-48,53,57,59,60,69,70]	Generated data in the form of game scores, i.e.,	
	how much of a task was completed well by the	
	user.	
[52]	Recorded the number of incorrect positions	
	executed by the user.	
[62,64,67]	Recorded both correct and incorrect movement	
	executions.	
	Additionally: [64] also recorded exercise history	
	and progress; [67] also recorded patient	
	compensation movements.	
[66]	Generated task counts that recorded progress,	
	posture duration, and errors committed.	
[42,65]	Counted the number of blocks moved from one	
	side to the other.	
	Additionally: [65] also recorded the hand	
	movement accuracy and speed.	
[40]	Used functional electrical stimulation in a platform	
	that produced arm and shoulder reach and flexion	
	data; and time to complete tasks.	
[57]	Measured the shoulder and elbow flexion range	
	from the Qualisys Motion Capture System.	
Variable Data (N=20)		
[23,30,49,56,61]	Collected kinematic variables, which are data used	
	for motion tracking. These data included hand	
	velocity, acceleration, and jerkiness.	
	Additionally: [23] kept video recordings of users,	
	movement distance and predicted stroke recovery	
	duration; [56] recorded hand orientation, grasp,	
	and haptic interaction; and [61] recorded gaze	
	direction and objects selected, and motor imagery	
	Dased on EEG signals.	
	Generated timed-up and go test variables, i.e.,	
	standing peak trunk nexion angle, standing angular	
	speed turn time, and total TUC time, including the	
	variables' standard error of measurement (SEM)	
	and minimum detectable change (MDC)	
	Recorded 3D skeletal position data (ankles and	
[LTT]	shoulder centre) gait velocity centre of pressure	
	nath velocity force distribution (lower limbs)	
[55]	Collected range of movement position and	
	rotation values of users	

Table 13. Data generated from systems.

[39]	Collected shoulder flexion and abduction, elbow
	flexion, and compensatory movements.
[63]	Recorded the number of user attempts and an
	error summary.
[53]	Monitored brain signals from a brain computer
	interface and electroencephalographic data.
[58]	Monitored electrocardiogram and
	photoplethysmograph data as well as blood
	pressure.
[67]	Used wearble sensors to monitor inertial and
	acceleration data.
[60]	Used a smart glove they developed to produce a
	detailed hand skeleton image that includes finger
	movement.
[71]	Study protocol: will use a wearable sensor to
	monitor energy expenditure.
[35]	Used weight and accelerometre sensors at the
	bottom of tea saucers and jugs to track errors and
	error patterns.
[54]	Used a training platform that consisted of the
	Nintendo Wii Balance Board, Microsoft Kinect,
	TEREFES electrosimulator, and a computer. Data
	generated from the platform included frequency
	stimulation and muscle synergies.
[68]	Color image, 3D depth image, and voice signal
[69]	Biosignals for muscle activity
[50]	Cortical activation pattern, blood-oxygen-level
	dependent signal volume

Table 14. Patient access to PGHD.

Did patients have acccess to their PGHD?		
Reference	Description	
Guidance (N=19)		
[44,45,49,52,53,60-62,64,67,70-72]	Guided patients in performing a task through a visual interface. Additionally: [67,70-72] also provided game scores, i.e., number/percent of tasks performed correctly; and [67] provided the compensatory movements patients made.	
[30]	Provided auditory feedbacks.	
[23,35,54,56,66]	Provided both visual and auditory guidance. Additionally: [23] also provided game scores; [66] showed patients the duration they were able to hold a certain position; [54] provided patients their centre of pressure information as they stand on a balance platform; and [56] also provided tactile feedback.	
Progress (N=4)		
[42,65]	Technologised the box-and-blocks test. Progress was tracked through the number of blocks patients needed to move from one side to the other.	
[63]	Showed red and green balloons to show progression for uncompleted and completed tasks, respectively. Also showed	

	users the number of attempts they took to do an exercise, as
	well as their errors.
[46]	Informed patients when they have met exercise goals.
Task Scores (N=10)	
[22,37-39,47,50,55,57,59,69]	All the papers simply provided scores at the end of a task
	execution.
	Additionally: [39] showed patients the compensatory
	movements they made; [47] informed patients whether the
	images they selected were correct or not.

Table 15. PGHD utilisation.

How was PGHD Utilised?		
Reference	Description	
Patient (N=19)		
[46,48,70]	Using data for therapists to prescribe or tailor-fit rehabilitation to individual patient	
	needs through calibrating game intensity or duration.	
[23,49,65,66,69]	Using data for therapists to prescribe or tailor-fit rehabilitation to individual patient	
	needs through prescribing appropriate exercises.	
[47,58,67,72]	Using data for therapists to prescribe or tailor-fit rehabilitation to individual patient	
	needs through monitoring and evaluation of patient progress.	
[35,39,49,52,56,61,	Utilised PGHD primarily to guide patients as they perform rehabilitation exercises.	
64,68]	Additionally: [39] utilised data for evaluation of patient progress.	
Comparison (N=2)		
[57]	Utilised the data to compare performance between four different groups of participants	
	(20 patients - 10 left, 10 right hemiparesis; 20 healthy - 10 left, 10 right trained) on a set	
	of exercises.	
[72]	Will compare system-generated data with clinical outcome measures, e.g., Fugl-Meyer	
	scores and BBT.	
Kinect-based Systems	s (N=13)	
[22,55]	Utilised data to assess their Kinect-based systems on their effectiveness of assessing	
	patient improvement.	
[42,45]	Utilised data to assess their Kinect-based systems on their effectiveness through	
	evaluation of their system's performance as compared with traditional rehabilitation.	
[50]	Utilised data to assess their Kinect-based systems on their effectiveness to activate	
	significant brain regions.	
[23]	Assessed the feasibility of their system to predict the duration of patients to recovery.	
[71]	Assessed both feasibility and effectiveness of their system.	
[30,41,44,56]	Utilised PGHD to assess the reliability of their systems in tracking movement, and motor	
	function of patients. [56] assessed reliability of interaction between their exercise	
	games and intelligent objects.	
[54,61]	Assessed their platform's accuracy, i.e., timing and synchronisation.	
Other Technologies (I	N=2)	
[40]	Used PGHD to observe the effects of applying functional electrical stimulation to patient	
	muscles as they attempt to complete functional tasks.	
[35]	Used PGHD produced by weight and accelerometre sensors to record the errors	
	patients make, and observe their error patterns.	