

Effects of wearable device training on upper limb motor function in patients with stroke: a systematic review and meta-analysis Journal of International Medical Research 2024, Vol. 52(10) 1–21 © The Author(s) 2024 Article reuse guidelines: sagepub.com/journals-permissions DOI: 10.1177/03000605241285858 journals.sagepub.com/home/imr



Qianqian Song¹, Qin Qin¹, Lorna Kwai Ping Suen², Guangmei Liang¹, Haixia Qin¹ and Lingling Zhang¹

Abstract

Objective: To evaluate the effect of wearable device training on improving upper limb motor function in patients who experienced strokes.

Methods: The PubMed, Embase, Cochrane Library, Web of Science, MEDLINE, SCOPUS, China National Knowledge Infrastructure, WanFang, and VIP databases were searched for randomized controlled trials (RCTs) that assessed the effectiveness of wearable device training in improving upper limb motor function in patients with stroke. Two investigators independently screened studies by their titles and abstracts and cross-checked, downloaded, and evaluated the results. Disagreements were resolved by a third highly experienced researcher. Risk of bias was evaluated using the Cochrane risk-of-bias tool. This meta-analysis was registered in PROSPERO (registration No. CRD42023421633).

Results: This study comprised 508 patients from 14 RCTs. The experimental group assessed various wearable devices, including 3D-printed dynamic orthoses, inertial measurement unit (IMU) sensors, electrical stimulation devices, and virtual reality (VR) devices for virtual interactive training. The control group received traditional rehabilitation therapies, including physical and conventional rehabilitation. The experimental group scored better on the Fugl–Meyer Assessment (FMA-UE) scale (standardized mean difference [SMD] 0.26, 95% confidence interval [CI] 0.07, 0.45) and Box and Block Test (BBT) (SMD 0.43, 95% CI 0.17, 0.69) versus controls.

¹ Guangxi University of Chinese Medicine, Nanning,	Corresponding authors: Qin Qin, Guangxi University of Chinese Medicine, 13 Wuhe Avenue, Qingxiu District, Nanning, Guangxi 530023, China. Email: 2057829736@qq.com
Guangxi, China	Lorna Kwai-Ping Suen, School of Nursing, Tung Wah
² Faculty of Nursing, Tung Wah College, Homantin,	College, 31 Wylie Road, Homantin, Hong Kong.
Hong Kong	Email: Iornasuen@twc.edu.hk

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Conclusions: Wearable device training was found to significantly improve upper limb motor function in patients with stroke, particularly for large-range movements. Improvements in FMA-UE and BBT scores reflected reduced impairment and enhanced manual dexterity, respectively. However, the training had no significant effect on hand movement frequency, fine motor skills, or spasticity. Among the different wearable devices tested, orthoses produced the most effective results.

Keywords

Stroke, wearable devices, upper limbs, motor function, meta-analysis

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Introduction

Stroke, also known as cerebral infarction or cerebrovascular accident, is an acute cerebrovascular disorder characterized by the loss of neurological function in specific areas of the brain due to ischemia or hemorrhage.¹ Stroke is distinguished by its high mortality and morbidity rates, affecting approximately 70-80% of patients exhibiting physical impairments.² In particular, 55-75% of these patients continue to experience upper limb motor dysfunction 3-6 months post-stroke, with symptoms such as reduced motor function, decreased muscle strength, increased muscle tone, joint contractures, and impaired fine motor coordination.³ Due to the crucial role of upper limbs in performing basic daily tasks and complex and precise activities, upper limb motor dysfunction affects patients' self-care abilities, including writing, typing, dressing, toothbrushing, and combing hair.⁴⁻⁶ Moreover, upper limb motor dysfunction extends beyond self-care and affects social interactions, thereby increasing the risk of depression among stroke survivors.4,7 Loss of upper limb function may limit participation in social activities, exacerbating social isolation, which is closely associated with depression.⁸ Consequently, improving upper limb motor function in patients who have experienced strokes has become a primary focus in the clinical and rehabilitation field.⁹

Modern rehabilitation techniques, particularly those utilizing wearable devices, have gained significant attention due to the portability, real-time monitoring capabilities, and personalization features of the devices.^{10,11} Various types of wearable devices have been used for poststroke upper limb motor function rehabilitation, such as wearable sensor devices that integrate advanced technological features including physiological monitoring, pressure sensors, and inertial sensors, allowing for real-time capture and analysis of patients' physiological parameters and movement states.^{12,13} Wearable orthoses are devices designed to support, enhance, or replace the functions of body parts, such as the arms, legs, and spine.^{14,15} These devices feature passive driving mechanisms that, for example, offer functional, strengthening, and taskspecific hand training, enabling the affected limbs to be used repetitively and extensively in daily activities.¹⁶ Wearable dynamic orthoses are adjustable, offering to manage stiffness and provide assistance levels based on the specific needs of the patient. By assisting with correct upper limb movements, the orthosis receives positive physiological and tactile feedback, which reinforces normal movement patterns and supports the maintenance and enhancement of functional neural pathways in the cerebral cortex.¹⁷

Wearable virtual reality (VR) technology offers patients with stroke a novel form of training by allowing them to interact with virtual or augmented reality environments. This immersive experience can contribute to improving limb function. High-quality, high-frequency feedback from the technology enhances synaptic efficacy, thereby improving the flow of information between the cerebral cortex and subcortical structures and establishing normal sensory feedback. Consequently, this enhances hand coordination and limb function.^{18,19} Electrical stimulation devices use auditory. visual, and tactile feedback to enhance proprioception in patients, thereby correcting their posture and movement patterns.²⁰

Wearable functional electrical stimulation (FES) devices are associated with the recovery of gross hand coordination by activating neural circuits and facilitating functional reorganization in healthy neural tissue surrounding damaged areas.²¹ By utilizing vibratory tactile stimulation, wearable FES devices can reduce the incidence of spastic hypertonia.²² Compared with traditional devices, wearable electrical stimulators may be used in daily life, overcoming limitations related to portability and operational complexity. Their compact and integrated design allows patients to receive prolonged repetitive stimulation during daily activities.²³ Additionally, these devices assist rehabilitation specialists in developing personalized training programs to enhance treatment efficacy and safety. Their data-driven incentive mechanisms have been shown to improve patient compliance and effectively promote the recovery of upper limb function.²⁴

Wearable device-based rehabilitation training has become increasingly common among patients who have experienced strokes. Early studies suggest that wearable devices can positively impact post-stroke upper limb motor function.^{25,26} However, current research on these devices highlights significant methodological variability, including the types of intervention used, their content, and their duration. These discrepancies result in inconsistent treatment effects, limiting comprehensive evaluation of the overall efficacy of these interventions. Despite the extensive use of various types of wearable devices in clinical practice, such as sensor devices, FES units, VR systems, and wearable orthoses, systematic research comparing the effectiveness of different device types on the improvement of post-stroke upper limb motor function is limited. Consequently, clinicians face challenges in selecting the most effective interventions for enhancing rehabilitation.

Several systematic reviews and metaanalyses have evaluated the effectiveness of wearable device training in improving upper limb motor function impairments following stroke. Through their narrative review, Parker et al.25 concluded that the impact of wearable device training on limb motor function impairments was not significant, possibly due to small sample sizes and the complexity of the interventions. Conversely, Maceira-Elvira et al.²⁷ explored the effects of different types of wearable devices on patients with upper limb motor function impairment and concluded that wearable devices have a positive impact on upper limb rehabilitation. Previous studies primarily provided have narrative overviews with limited emphasis on the quantitative integration of data.

The novel aim of the present study was to systematically compare the effects of different types of wearable devices on poststroke upper limb motor function through a network meta-analysis, to not only integrate and quantify existing research data, but also analyse the relative effectiveness of different device types, providing clearer guidance for clinicians. By quantitatively evaluating the effects of various wearable devices, this study aimed to advance the clinical application of personalized and precise rehabilitation.

Therefore, in the present meta-analysis, the impact of wearable device technology was compared with conventional rehabilitation on limb motor dysfunction, hand dexterity, arm and hand function, and daily arm use. The primary outcome measures were the Fugl-Meyer assessment-upper extremity (FMA-UE) and Box and Block Test (BBT) scores, while the secondary outcome measures included the Action Research Arm Test (ARAT), motor activity log (MAL), and modified Ashworth scale (MAS) scores. Furthermore, the extent of improvement in limb motor function provided by different types of wearable devices was assessed, and the overall effects of various interventions were comprehensively compared.

Materials and methods

This study adhered to the Preferred Reporting Items for Systematic Reviews Meta-Analyses (PRISMA) guideand lines,²⁸ was registered and on the PROSPERO platform (https://www.crd. vork.ac.uk/prospero/) under the registration CRD42023421633. No. Comprehensive searches in PROSPERO and other relevant databases revealed no ongoing or published systematic reviews or meta-analyses on the same intervention method. Ethical approval and patient consent were not required as the meta-analysis and bioinformatic analyses were based on published research and public databases.

Literature search strategy

The PubMed, Embase, Cochrane Library, Web of Science, MEDLINE, SCOPUS, China National Knowledge Infrastructure, Wanfang Data, and VIP databases were searched for studies published in Chinese and/or English from database inception until 1 July 2024. Search keywords included 'stroke, hemiplegia, apoplexy, cerebrovascular accident, CVA, brain vascular accident, cerebral vascular accident, wearable electronic devices, wearable technology, wearable devices, skin electronic, upper extremity, upper limb, arm, hand, motion capture devices, activity monitor, and inertial measurement unit.' The specific search strategy, exemplified by the PubMed database, was as follows:

#1 "stroke"[MeSH] OR hemiplegia OR apoplexy OR cerebrovascular accident OR CVA OR brain vascular accident OR cerebral vascular accident

#2 "Wearable Electronic Devices" [MeSH] OR Wearable Technology OR Wearable Devices OR Skin Electronic OR Motion Capture Devices OR Inertial Measurement Unit OR Wearable orthotic OR Wearable accelerometer OR Wearable activity tracker

#3 "upper extremity" [MeSH] OR upper limb OR arm* OR hand*

#4 #1 AND #2 AND #3

Inclusion and exclusion criteria

The study inclusion criteria were: (1) Study population–studies comprising patients diagnosed with stroke,^{29,30} confirmed through CT and/or MRI, with no restrictions on age or sex; (2) Intervention measures–studies wherein the experimental

group received wearable-device training either alone or in combination with conventional rehabilitation treatments; (3) Control measures-studies that adopted basic treatments, including conventional rehabilitation treatments, such as physical and occupational therapy; (4) Outcome indicators-studies that included at least one of the following outcome measures: FMA-UE, ARAT, BBT, MAL, and MAS; (5) Research method-randomized controlled trials (RCTs); and (6) Languages-studies published in Chinese and/or English.

Exclusion criteria comprised the following: (1) Conference papers, academic theses; (2) original data that was incomplete or could not be extracted, with inadequate outcome data even after unsuccessful attempts to contact the authors; (3) duplicate publications; and/or (4) animal experiments and systematic reviews.

Literature screening and data collection

Following the removal of duplicates, the titles and abstracts of studies identified during the database search were independently screened by two investigators (Song and Liang). The results were cross-checked, downloaded, and the full texts were then independently evaluated. Disagreements were resolved by a third highly experienced researcher (Qin).

The following data were extracted from the remaining articles: (1) basic information, including the first author and year of publication; (2) details of the study design, encompassing the sample size, randomization method, allocation concealment (participant blinding and evaluator blinding), and data analysis; (3) participant data, such as age, disease history, and disease stage; (4) details of interventions used for the experimental group and traditional rehabilitation therapies for the control group, such as physical and conventional rehabilitation therapy; and (5) relevant outcome measurements.

Quality assessment of included studies

The quality of included studies was assessed by two researchers (Song and Liang) using the Cochrane Handbook for Systematic Reviews, version 5.1.0 (https://www.coch rane.org/). Assessments focused on several key areas: selection bias, including the generation of random sequences; performance bias, specifically the adequacy of blinding; attrition bias, such as incomplete outcome data; reporting bias, including the selective reporting of results; and other biases. Each study was reviewed for risk of bias based on these criteria.

Outcome indicators

The primary outcome indicators were the FMA-UE and BBT scores. The FMA-UE is a stroke-specific performance-based impairment index designed to assess motor function, balance, sensation, and joint functioning in patients with post-stroke hemiplegia,³¹ and has been widely used in clinical settings and research to determine a patient's motor capacity and track changes over time. The BBT measures manual dexterity and is used to evaluate motor improvements over the course of rehabilitation. During the test, a patient moves blocks from one compartment of a box to another; this action helps quantify their grasp, transport, and release of the objects, reflecting their hand function.^{32,33} Secondary outcome indicators were the ARAT, MAL, and MAS scores. The ARAT is used to assess specific motor arm functions and the ability to handle objects of varying sizes, weights, and shapes, providing a detailed measure of upper extremity motor function.34,35 The MAL measures how often and how well a patient uses their affected arm and hand for activities of daily living, reflecting the

functional use of the impaired limb in everyday tasks.³⁶ The MAS is a measure of muscle spasticity in patients with neurological conditions that assesses resistance during passive soft tissue stretching and is used to evaluate changes in muscle tone over time.

Statistical analyses

To determine the overall effect of wearable devices, a quantitative analysis of extracted data from the included studies was performed using Review Manager (RevMan) software, version 5.4 (The Cochrane Collaboration, 2020). Continuous results with the same units were analysed using mean difference (MD), while standardized mean difference (SMD) was used to analyse inconsistent scale scores. Uncertainty was determined using a 95% confidence interval (CI). Effect sizes were categorized as: small (SMD < 0.2), medium (0.2 < SMD < 0.5), or large (SMD > 0.5). Heterogeneity was assessed using I^2 , where $I^2 < 50\%$ and P > 0.1 indicated minimal heterogeneity and a fixed-effects model was used. For $I^2 > 50\%$ and P < 0.1, a random-effects model was employed. The significance level was set at $\alpha = 0.05$. To compare the effects of different wearable devices. R software was used to plot a network relationship diagram comparing the efficacy of different types of wearable devices. A network meta-analysis with FMA-UE as the outcome indicator was performed using Aggregate Data Drug Information System (ADDIS) software, version 1.16.8 (https:// drugis.org/software/addis). In ADDIS, the potential scale reduction factor (PSRF) assessed the model convergence, and a PSRF value between 1 and 1.05 indicated good convergence performance, suggesting that the analysis model was relatively stable. Ranking probability plots were used to determine the intervention method with the highest probability of being the most effective. The probabilities of each intervention were ranked from Rank 1 to Rank 5, with Rank 1 indicating the best effect and Rank 5 indicating the worst effect. Higher probabilities for an intervention indicated better treatment effects. Sensitivity was analysed using one-by-one elimination. If the statistical results did not change, it indicated that a single study was not the main reason for the heterogeneity. Funnel plots were used to assess publication bias; a symmetrical distribution on both sides indicated a low likelihood of publication bias, while an asymmetrical plot suggested potential publication bias.

Results

Literature search and process results

Results of the literature search and screening process are summarised in Figure 1. A total of 14 articles were included, $^{37-50}$ involving 528 participants with a mean age >50 years.

Study characteristics

Among the included studies, four investigated wearable sensor devices,^{37–39,41} three investigated wearable VR technology devices,^{40,42,49} three investigated wearable FES devices, 45,48,50 and one investigated a wearable orthotic device.47 The intervention frequency ranged from 30-60 min per session, 2-5 times a week, for a duration of 3-12 weeks. For the experimental group, interventions included either wearable device training, such as 3D-printed dynamic orthoses, IMU sensors that prompt movement, FES devices, or VR devices, alone or in combination with conventional rehabilitation training. The control group received conventional rehabilitation training, including physical and occupational therapy. Among the included studies, 10 articles used the FMA-UE scale to assess the recovery of upper limb function in patients who experienced strokes.^{37–42,45,47–49} Eight studies

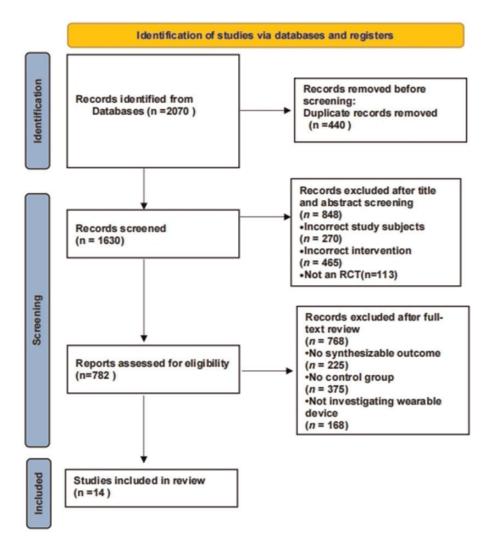


Figure 1. PRISMA flowchart diagram of the search process and results.

used the BBT to assess hand dexterity.^{38-40,42-} ^{45,48} Four reports used the ARAT to assess fine motor functions of the arm and hand.38,39,42,43 Two studies used the MAL to assess daily arm use.^{38,39} Two studies used the MAS to assess muscle tone and spasticity of the patient's hand.^{46,47} In two studies.40,42 the 'control group' received conventional rehabilitation alone, whereas the 'control group 1' received mirror therapy combined with conventional rehabilitation in one study,⁴⁰ and isometric grip training

combined with conventional rehabilitation in the other study.⁴² For these two studies, the 'control group' and 'control group 1' were merged for meta-analysis, according to the Cochrane Handbook for Systematic Reviews, version 5.1.0 formula. One study had an unmatched intervention duration and frequency between the control and intervention groups.⁴⁴ The control group received conventional rehabilitation, including group therapy three times a week (45 min per session) and individual therapy at least four times a week (45 min per session). The intervention group received SaeboFlex device training at least once daily (45 min per session), 5 days a week, for 8 weeks. The remaining 13 studies had consistent intervention durations and frequencies, with doses generally matching. Basic information regarding the 14 included studies is summarised in Table 1.

Bias risk of the included studies

According to the revised Cochrane risk of bias tool (RoB 2) for risk of bias assessment, among the included studies, two did not specify the randomization methods used,^{38,48} raising concerns. For deviations from intended interventions, only three studies were assessed as low risk, 37,45,48 while the remaining 12 studies were evaluated as having some concerns, 37-45,47,49,50 primarily due to unclear blinding of participants and intervention providers. Regarding missing outcome data, all 14 studies were assessed as low risk.37-50 To measure the outcome and select the reported results, all studies were assessed as low risk. Notably, two studies were evaluated as having a low risk across all bias categories. The RoB 2 results for each study are shown in Figure 2.

Meta-analysis

Upper limb motor impairment score. Eleven studies used the FMA-UE scale to assess the severity of upper limb motor impairment in 448 patients who experienced stroke.^{37–42,45,47–49} No heterogeneity was observed between the studies ($I^2 = 0\%$, P = 0.62). Using a fixed-effects model, the FMA-UE score was higher in the experimental group than in the control group who received traditional rehabilitation training including physical and conventional rehabilitation therapy (SMD 0.26, 95% CI 0.07, 0.45, P < 0.05; Figure 3). Hand coordination ability score. Eight studies assessed hand coordination using the BBT in a combined total of 253 patients who experienced stroke.^{38–40,42–45,48} No heterogeneity was found among the included studies ($I^2 = 23\%$, P = 0.26). Using a fixed-effects model, the BBT score was significantly higher in the experimental group than in the control group that employed traditional rehabilitation training, including physical and conventional rehabilitation therapies (SMD 0.43, 95% CI 0.17, 0.69, P < 0.05; Figure 4).

Hand fine motor skill score. Five studies,^{38,39,42,43,50} comprising 110 patients who experienced stroke, used the ARAT to assess the ability of patients to handle fine movements of the hand. No heterogeneity was observed between the studies ($I^2 = 3\%$, P = 0.39). Furthermore, using a fixed-effects model, no statistically significant differences were observed in ARAT scores between the experimental group and control group (SMD 0.2, 95% CI -0.15, 0.55, P = 0.27; Figure 5).

Upper limb activity frequency score. Three studies, 37,49,50 comprising 79 patients who experienced stroked, used the MAL to assess the frequency and quality of hand activities. No heterogeneity was observed between the studies (I²=0%, P=0.48). Additionally, using a fixed-effects model, no statistically significant differences in MAL score were observed between the experimental group and control group (SMD 0.05, 95% CI -0.35, 0.44, P=0.82; Figure 6).

Upper limb muscle spasticity score. Two studies assessed a combined total of 57 patients who experienced stroke,^{46,47} using the MAS for muscle tone and spasticity. Moderate heterogeneity was observed among the studies ($I^2 = 47\%$, P = 0.17). Using a fixed-effects model, no statistically significant difference was observed in terms

Table I. Su	immary of	Table 1. Summary of 14 studies included		atic review and meta-an	in the systematic review and meta-analysis of patients who experienced stroke.	rienced stroke.			
	Study sample,	é,			Intervention method				
Study	total <i>n</i> (exp/ctrl)	Age, years	Clinical course, Y, M, or D	Upper limb impairment degree	Experimental group	Control group	- Duration, weeks	Frequency	Outcome measure
Alon, 2008 ⁴⁸	26 (13/13)	Exp 63.15±10.91 23. Ctrl 66.46±15.22	I 23.79 ± 8.8 D 2	FM-UE score (2–10) indicating moderate to severe impairment	Wearable orthotic with functional Conventional exercise electrical stimulation (intensity and mode of electrical stimu- lation controlled by preset program), plus conventional	Conventional exercise	12	30 min, twice daily	FMA-UE, BBT
Barry, 2012 ⁴³	(6/01) 61	Exp 52.7 \pm 16.1 Ctrl 50.3 \pm 7.5	Exp 4.l ± 2.3 Y Ctrl 5.l ± 5.7 Y	Severe limb impairment	exercise Dynamic wrist-hand orthosis (SaeboFlex) for mechanical assisted treatment	Manual-assisted therapy from therapist	6	l h/week + home BBT, ARAT exercise 4 dave/week	BBT, ARAT
Friedman, 2014 ⁴² 12 (4/8)	⁴² 12 (4/8)	Total 57 \pm 30.5	I	FMA score (34–62) indicating mild to moderate upper limb	le musical	lsometric exercises and traditional desktop exercises	9	6 h, 3 times/week FMA-UE	FMA-UE
Lannin, 2016 ⁴⁴	9 (5/4)	Exp 63 ± 10.4 Ctrl 51 + 21 1	Exp 2.5±1.7 M Ctrl 4 7+61 M		Wearable SaeboFlex orthotic	Conventional therapy	ω	45 min/day, 5 times/week	BBT
Knutson, 2016 ⁴⁵ 72 (40/32)	5 72 (40/32)	Exp 55.4 (17.0) Ctrl 56.3 (12.7)	Exp 1.8 (2.5) M Ctrl 1.6 (4.9) M	FM-UE: 34 points indicating moderate impairment	Wearable glove functional elec- trical stimulation (FES)	Cyclic neuromuscular electrical stimulation (cNMFS)	12	60 min/time	FMA-UE, BBT
Lin, 2018 ⁴¹	18 (9/9)	Exp 52.2 \pm 10.2 Ctrl 62.6 \pm 7.1	Exp 1115.7 D Ctrl 510.9 D	Brunnstrom stages 3–5	Wearable device based on inertial Conventional therapy measurement units (IMUs) and wireless sensor networks (VVSNs) + rehabilitation	Conventional therapy	Ω.	15 times/day, 3 days/week	FMA-UE
Villafane, 2018 ⁴⁶	⁵ 32 (16/16)	Exp 67 ± 11	I	Ashworth scale $<$ 3	exercise Wearable glove device (Clorcha) DT and OT	Traditional rehabilitation	e	30 min/day, 3 doue/work	MAS
Wei, 2019 ³⁹	59 (32/27)	Curl 70 ± 12 Exp 59.19 ± 11.2! Ctrl 63.11 ± 10.2	Curl 70 ± 12 Exp 59:19 ± 11:25 Exp 47.75 ± 21:93 D Ctrl 63:11 ± 10:27 Ctrl 53:67 ± 41:16 D	Hemiplegic upper limb) functional test /EM LEV > 3	(Glorenia) + r 1 and Cl Wearable vibration alert watch + regular exercise	uter apy + r 1 and C1 Standard care, sports	4	3 h/day	FMA-UE, BBT, ARAT, MAL
Yang, 2021 ⁴⁷	25 (13/12)	Exp 44.4 ± 10.2 Ctrl 47.1 ± 9.2	Exp 22.5 \pm 11.9 M Ctrl 20.0 \pm 10.4 M	MAS score I-3	Wearable dynamic 3D-printed wrist splint + standard	Standard rehabilitation	ę	6 h/day	FMA-UE, MAS
Park, 2021 ⁴⁹	44 (22/22)	Exp 60.59 ± 18.1; Ctrl 62.29 ± 13.9	Exp 60.59 ± 18.12 Exp 17.73 ± 5.98 M Ctrl 62.29 ± 13.97 Ctrl 16.82 ± 7.28 M	Ashworth scale $>$ 3	Wearable glove device for game- based training equipped with multiple sensors + conven- tional PT	Conventional PT	4	30 min/time, 5 times/week	FMA-UE

(continued)

	Study sample,	e,	į		Intervention method				(
Study	total <i>n</i> (exp/ctrl)	Age, years	Clinical course, Y, M, or D	Upper limb impairment degree	Experimental group	Control group	 Duration, weeks 	ı, Frequency	Outcome measure
Schwerz de Lucena, 2022 ³⁸	20 (10/10)	Exp 56 ± 17 Ctrl 58 ± 12	Exp 32 ± 14 D Ctrl 48 ± 45 D	1	real-time feedback including the meter device' + regular	Standard watch	m	10 h/day	FMA-UE, BBT, ARAT, MAL
Kim, 2022 ⁴⁰	36 (12/24)	Exp 59.58 ± 5.65 Ctrl 62.88 ± 4.71	Exp 7.00 \pm 2.45 M Ctrl 6.71 \pm 2.44 M	FMA score (26–56) indicating Wearable reflective device mild to moderate motor mirror therapy +	sports Wearable reflective device mirror therapy +	Traditional mirror + conventional	4	30 min, 5 times/week	FMA-UE, BBT
Guo, 2023 ³⁷	120 (60/60)	Exp 56.25 \pm 9.83 Ctrl 55.82 \pm 9.68	120 (60/60) Exp 56.25 ± 9.83 Exp 46.22 ± 36.63 D Ctrl 55.82 ± 9.68 Ctrl 61.2 ± 46.67 D	Вц	conventional reliabilitation including Conventional PT Wearable rehabilitation including Conventional PT IMUs and flex sensors, con- ventional clinical PT training	Conventional PT	m	30 min/day, 2 days per week	FMA-UE
Maeda, 2024 ⁵⁰ 20 (10/10)	20 (10/10)	Exp 71.6 ± 14.1 Ctrl 74.1 ± 9.0	Exp 52.7 ± 21.8 D Ctrl 50.9 ± 22.4 D	Classification of finger function test, Class 1 C or above, or Level 2 and above	and standard drug uterapy Standard rehabilitation nursing measures + 'MIVES' electrical stimulation device	Standard rehabilitation nursing measures + IVES	4	8 h/day	FMA-UE, ARAT, MAL
Data presente	ed as <i>n</i> preva	Data presented as n prevalence or mean \pm SD,	SD, or median (interquartile range).	quartile range).					

Exp, experimental group; Ctrl, control group; Y, years; M, months; D, days; PT, physical therapy; OT, occupational therapy; IVES, integrated volitional control electrical stimulation; WIVES, wearable integrated volitional control electrical stimulation.

Table I. Continued.

(a)	🐱 Randonization process	• Deviations from intended	 Missing outcome data 	 Measurement of the outcomestication 	 Selection of the reported 	- Overall	B Low risk	(b)					
2 Schwerz de 2022	(?)	?	•	•	•	(Some concerns	Percentage of Risk of Bias Items	in the	Includ	ed Stu	dies.	
3 Maeda 2024	•	?	•	•	•	!	🕐 High risk						
4 Kim 2023	•	3	•	•	•	()							
5 Knutson 2016	•	?	•	•	•	1		Overall Bias					
6 Barry 2012	•	?	•	•	•	(Selection of the reported result					
7 Villafañe 2018	•	•	•	•	•	•		Measurement of the outcome					
8 lin 2018	•	?	•	•	•	1		Mising outcome data					
9 Guo 2023	•	•	•	•	•	•		Deviations from intended					
10 Lannin 2016	•	?	•	•	•	1		Randomization process				-	
11 Friedman 2014	•	?	•	•	•	•		nandernization process					
12 Park 2021	•	?	•	•	•	1		0	20	40	60	80	100
13 Yang 2021	•	?	•	•	•	1							
14 Vayne 2018	•	?	•	•	•	1		Low risk Some conc	cerns	Hi Hi	gh risk	<	

Figure 2. Risk of bias assessment results in the 14 included studies, showing: (a) risk of bias summary; and (b) percentages of risk of bias items.

	Exp	eriment	tal	0	ontrol			Std. Mean Difference	Std. Mean Difference
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Fixed, 95% CI	IV, Fixed, 95% CI
Alon G, 2008	24.2	12.7	13	14.5	10.3	13	5.4%	0.81 [0.01, 1.62]	
Friedman 2014	0.875	3.19	4	1.29	2.17	8	2.4%	-0.15 [-1.35, 1.05]	
Guo 2023	36.76	18.65	60	31.13	21.67	60	27.2%	0.28 [-0.08, 0.64]	
Kim 2023	32.33	3.8	12	29.83	3.91	24	7.0%	0.63 [-0.08, 1.34]	
Knuton 2016	2.4	5.27	32	2.2	4.69	40	16.3%	0.04 [-0.42, 0.50]	+
Lin 2018	30.7	15.2	9	35.6	20.5	9	4.1%	-0.26 [-1.19, 0.67]	
Maeda 2024	51.36	7.4	10	51.36	14.4	10	4.6%	0.00 [-0.88, 0.88]	
Park 2021	87.95	14.16	22	86	15.97	22	10.1%	0.13 [-0.46, 0.72]	
Schwerz de 2022	4.4	2.3	10	3	2.5	10	4.4%	0.56 [-0.34, 1.46]	the second se
Wei 2019	57.19	12.29	32	54.48	15.38	27	13.4%	0.19 [-0.32, 0.71]	
Yang 2021	36.8	11.2	13	26	14.7	12	5.2%	0.80 [-0.02, 1.63]	
Total (95% CI)			217			235	100.0%	0.26 [0.07, 0.45]	•
Heterogeneity: Chi ² =	8.08, df	= 10 (P	= 0.62); $I^2 = 09$	6				
Test for overall effect	Z= 2.72	2 (P = 0.	006)						-4 -2 U 2 4 Favours [control] Favours [experimental]

Figure 3. Comparison of Fugl-Meyer assessment-upper extremity (FMA-UE) scores between patients who experienced stroke and received wearable device training (experimental group) and controls who received conventional treatment.

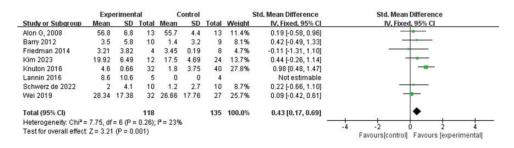


Figure 4. Comparison of Box and Block test (BBT) scores between patients who experienced stroke and received wearable device training (experimental group) and controls who received conventional treatment.

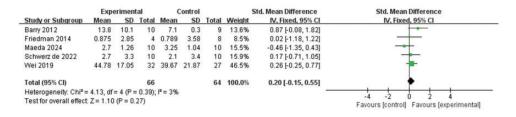


Figure 5. Comparison of Action Research Arm Test (ARAT) scores between patients who experienced stroke and received wearable device training (experimental group) and controls who received conventional treatment.

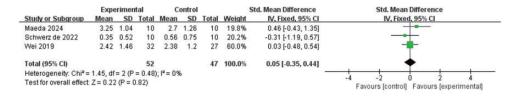


Figure 6. Comparison of motor activity log (MAL) scores between patients who experienced stroke and received wearable device training (experimental group) and controls who received conventional treatment.

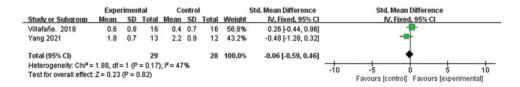


Figure 7. Comparison of modified Ashworth scale (MAS) scores between patients who experienced stroke and received wearable device training (experimental group) and controls who received conventional treatment.

of MAS score between the experimental group and control group (MD -0.06, 95% CI -0.59, 0.46, P = 0.82; Figure 7).

Network meta-analysis. Eleven studies were included in the network meta-analysis,^{37–42,45,47–50} using the FMA-UE score as an indicator. The network relationship comparing the effects of different wearable devices, with a PSRF value of 1, is shown in Figure 8. The probability ranking is displayed in Table 2.

The best probability rankings of different types of wearable devices for FMA-UE scores in stroke patients were as follows: the wearable orthotic device ranked the highest, with a score of 0.39, followed by the FES device with a score of 0.25, the sensor device with a score of 0.2, and lastly the VR device with a score of 0.16.

Assessment of publication bias risk. Funnel plots were drawn for outcome indicators with ≥ 10 articles (all upper limb scores).

The funnel plots were not perfectly symmetrical, indicating a potential publication bias (Figure 9).

Discussion

The present meta-analysis of 14 studies was conducted to evaluate the impact of wearable devices on upper limb motor rehabilitation in patients who had experienced stroke. Overall, compared with conventional rehabilitation therapy alone, such as physical and occupational therapy, the use of wearable device training combined with conventional rehabilitation therapy improved two outcome measures of upper



Figure 8. Network relationship diagram comparing the efficacy of different types of wearable devices in patients who experienced stroke and received wearable device training and controls who received conventional treatment: (a) wearable electrical stimulation device; (b) wearable sensor device; (c) wearable sensor technology; (d) wearable upper limb orthosis; and (e) control group.

limb motor function, namely FMA-UE and BBT. Improvement effects measured with the ARAT, MAS, and MAL were not statistically significant. This study revealed that wearable device training is a promising approach for improving upper limb motor function in post-stroke individuals. In particular, the results indicate that wearable device training has a positive effect on enhancing upper limb motor function and gross hand coordination in individuals with upper limb motor impairment following stroke. The probability ranking showed that wearable orthotic devices yielded optimal results. Although improvements in fine motor coordination and the frequency and quality of hand activities were not prominent, this finding presents novel possibilities for stroke rehabilitation.

The present meta-analysis results indicated that wearable device training has a significant positive effect on upper limb motor function. The FMA scores were significantly higher in the experimental group than in the control group. Among the included studies, four utilized sensor-based wearable devices, 37-39,41 and three of these studies showed significant improvement in limb motor function. IMUs, components of wearable sensors, integrate accelerometer readings and gyroscope angular rate detection. This integration enables accurate tracking and analysis of complex upper limb movement patterns, thereby providing precise data support for motor rehabilitation.⁵¹⁻⁵³

 Table 2. Probability ranking of the effects of different interventions in 14 studies of patients who experienced stroke.

Intervention	Rank I	Rank 2	Rank 3	Rank 4	Rank 5
Wearable sensor device	0.23	0.28	0.29	0.14	0.07
Wearable electrical stimulation device	0.21	0.24	0.22	0.21	0.12
Wearable orthotic device	0.45	0.11	0.08	0.07	0.29
Wearable virtual reality device	0.11	0.35	0.25	0.18	0.12
Control group	0.00	0.03	0.16	0.40	0.41

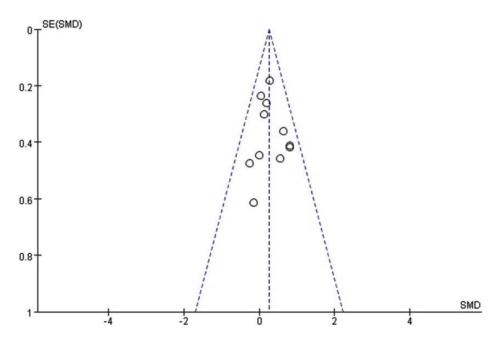


Figure 9. Funnel plot for Fugl-Meyer assessment-upper extremity scores in the published studies. SMD, standardised mean difference.

Among these studies, one showed that the FMA-UE scores of both the experimental and control groups significantly increased over time, suggesting an improvement in upper limb motor function.³⁸ However, there was no significant between-group difference in FMA-UE score improvement, which was attributed to the 3-week intervention period, suggesting that extending the intervention duration might result in a more pronounced improvement in the experimental group than in the control group. The BBT scores revealed a nonlinear relationship between hand usage intensity and BBT scores, whereby hand usage intensity increased significantly only when BBT scores exceeded 28.38 Younger participants, with better arm motor abilities, demonstrated a relatively high hand usage intensity even with lower BBT scores. Another study demonstrated a significantly higher improvement in FMA-UE scores in the experimental group versus controls

receiving conventional clinical rehabilitation training.³⁷ The average upper limb score of the experimental group increased by 11.28 points, while that of the control group increased by 7.45 points (P < 0.05). In a study that used a 'Reminder to Move' (RTM) wearable device as a treatment intervention for upper limb recovery in patients who experienced a subacute stroke, the experimental group showed significantly higher upper limb motor function scores than the control group.³⁹ Both groups showed improvement in BBT scores; however, the amount of movement recorded by the RTM device was significantly higher in the experimental group than the control group (12.52% and 9.41%, respectively). Lin et al.,41 demonstrated that wearable device training based on IMU modules (3-axis accelerometers and gyroscopes) resulted in a significant improvement in the experimental group's FMA scores pre- and post-intervention, from 27.2 ± 13.3 to 30.7 ± 15.2 (P < 0.05). Using IMUs for objective functional motor assessment in comprehensive upper limb training post-stroke provides immediate feedback, helping correct maladaptive rehabilitation postures.^{54,55} Tactile feedback and sensors in wearable devices can correct local abnormal movement patterns and facilitate global neural network reorganization.⁵⁵ This process is closely related to 'motor relearning', which involves repairing or optimizing damaged neural pathways through repetitive practice and positive feedback, thereby improving upper limb motor function.⁵⁶

In the present meta-analysis, three studies used wearable VR devices.40,42,49 In the study by Kim et al.⁴⁰ the experimental group received mirror therapy using a video augmented wearable reflection device. Their FMA-UE scores improved significantly from 27.67 ± 4.81 to $32.33 \pm$ 3.80, and the BBT scores increased from 15.83 ± 5.95 to 19.92 ± 6.49 blocks. Both the FMA-UE and BBT scores showed significant improvements versus controls, enhancing patients' visual feedback and engagement. In another study, а MusicGlove device with an engaging game interface was demonstrated to promote functional grasp movements.⁴² High levels of engagement and motivation led to significant improvements in upper limb motor function and hand dexterity, and the participants' FMA-UE scores indicated mild to moderate improvement in upper limb function. Those using MusicGlove experienced a mean increase of 3.21 ± 3.82 blocks in the BBT, which was significantly higher than the control group.⁴² Finally, the impact of the RAPAEL smart glove on FMA-UE scores was investigated in patients with acute stroke.⁴⁹ In the experimental group, the FMA-UE scores increased significantly from a baseline of 66.50 ± 24.43 to $87.95 \pm$ 14.16, while those of the control group increased from 62.95 ± 28.81 to $86.00 \pm$

15.97. No significant interaction between group and time was observed (F = 0.123, P = 0.728), indicating that both groups improved over time at similar rates. A longer follow-up period may reveal the potential long-term benefits of wearable devices for rehabilitation. Related studies indicate that wearable VR technology offers a novel training method for patients with stroke, by optimizing neural signal transmission through enhanced multisensory feedback. High-quality, high-frequency feedback may improve synaptic efficacy, facilitating information flow between cortical and subcortical structures. Consequently. this process enhances normal sensory feedback and improves hand coordination and limb function.^{18,19}

Three studies employed wearable FES devices.45,48,50 One study investigated contralaterally controlled functional electrical stimulation (CCFES) and found significant improvements in BBT scores in the CCFES group compared with the cyclic neuromuscular electrical stimulation (cNMES) group after 6 months.⁴⁵ The CCFES group showed an average improvement of 4.6 blocks. whereas the cNMES group improved by 1.8 blocks, with an intergroup difference of 2.8 blocks (P = 0.045). For participants with moderate hand impairment and a post-stroke duration of less than 2 years, the CCFES group showed greater improvement (9.6 blocks) than the cNMES group (4.1 blocks), with an intergroup difference of 5.5 blocks (P = 0.023). The greatest improvements in BBT scores were observed in patients who experienced a stroke less than 2 years previously and had moderate hand function impairment. No significant difference in the FMA-UE scores was observed between the groups (P=0.888), although both improved.⁴⁵ Alon et al.48 studied a Bioness H-200 device (Bioness Inc., Valencia, CA, USA) that features a set of surface electrodes for complete finger and thumb flexion and

extension. Using an 11 KHz carrier frequency and 36 Hz time-modulated current, patients underwent 30-minute FES training sessions twice daily for 12 weeks, combined with specific task practice. The FES group showed significantly higher FMA-UE scores (24.2 ± 13.7) than those of the congroup (14.5 ± 10.3) . Initially, trol all patients were unable to transfer a block within 60 seconds. However, by the end of the training, the FES group averaged 10.5 ± 12.0 blocks, while the control group averaged 2.5 ± 4.9 blocks. Eight patients in the FES group and three in the control group regained the ability to transfer five or more blocks. Although the BBT scores did not reach statistical significance, FES may aid in recovering functional hand use, with the small sample size potentially contributing to the lack of statistical significance.⁴⁸ Another study demonstrated that both integrated volitional control electrical stimulation (IVES) and wearable IVES (WIVES) significantly improved upper limb function in patients who experienced an early subacute stroke.⁵⁰ The mean betweengroup difference in FMA-UE scores was 1.3 points, with the upper limit of the 95% CI not exceeding the non-inferiority margin, indicating that WIVES was not inferior to IVES in terms of improving motor function. The WIVES device, which is smaller and lighter $(98 \text{ mm} \times 49 \text{ mm} \times 14.5 \text{ mm}; 55 \text{ g})$, is designed for ease of use and features electrode pads with magnetic plugs, allowing for optimal electrode positioning without removal. Additionally, wearable FES devices aid in hand coordination recovery by specifically activating neural circuits and reticulospinal pathways, promoting the functional remodelling of healthy neural tissue that surrounds damaged areas.⁵⁷ Choudhury et al.⁵⁸ observed improvements in gross hand coordination and flexibility, such as grasping ability, with wearable FES devices; however, fine motor enhancements were not observed, possibly because of reticulospinal rather than corticospinal activation.

Among the studies included in the present meta-analysis, one utilized a wearable orthotic device.⁴² Although only one study investigated wearable orthotics, a network meta-analysis systematically compared different wearable devices for improving post-stroke upper limb motor function, providing scientific references for clinical decision making. The analysis indicated that all four major types of wearable devices were more effective than interventions used in the control groups, with probability rankings suggesting that wearable orthotics were the most effective. Yang et al.47 investigated 3D-printed dynamic wrist splints worn for at least 6 hours daily for 6 weeks, and showed significant improvements in FMA-UE scores compared with controls. This wearable orthotic employed a four-bar linkage mechanism to adapt to finger trajectories, achieving customization and low cost through 3D printing, and it was comfortable and easy to wear independently. In another study,¹⁷ wearable orthotics were shown to provide positive physiological and tactile feedback by facilitating correct upper limb movements. This feedback mechanism reinforces normal movement patterns and promotes the maintenance and enhancement of functional neural pathways in the cortex, which are crucial for early-stage upper limb-function recovery.⁵⁹ Various factors influence and constrain the efficacy of limb-function rehabilitation. A comparison of the basic information of studies using the FMA-UE score as an outcome measure revealed inconsistent intervention durations, ranging from 3 to 12 weeks. Future research should further verify whether the intervention duration significantly affects improvement of post-stroke upper limb motor function.

Potential reasons for the lack of significant improvement in fine motor coordination, and the frequency and quality of hand activities, in patients who have experienced stroke when using wearable devices after stroke are as follows: In contrast to the present findings, previous studies have reported improvements in fine hand movements and spasticity with wearable device training,^{58,60} possibly because: (1) The literature reviewed in the present study primarily focused on FES devices designed to improve gross hand coordination. In addition, the mechanical construction and resolution of IMU sensor devices may be insufficient for capturing fine hand movements, rendering them less effective for assessing dexterity, grip exercises, applied forces, and muscle states.⁶¹ Although studies have demonstrated that surface electromyography (sEMG) sensors are effective in improving fine motor skills of the upper limb,²⁷ the present review did not include studies on sEMG sensors due to their lack of randomization. Instead, this review mainly included studies that used accelerometers (n=4), which may not offer detailed feedback on fine motor movements. (2) The duration and frequency of rehabilitation interventions are crucial for achieving maximal neuroplasticity.⁶² In the present study, the interventions ranged from 3 to 12 weeks, with the timing and frequency tailored to individual patient needs. High-intensity somatosensory stimulation induces changes in motor cortex excitability, thereby affecting limb function.⁶³ (3) Patient baseline data and compliance: the stroke condition and injury location vary among patients, which may directly affect rehabilitation.⁶⁴ effectiveness the of Therefore, individualization of rehabilitation training and patient compliance are factors that need to be considered. (4) Technical limitations of wearable devices: issues with data processing and synchronization may affect real-time feedback and efficacy assessments during training. Slow patient movements may result in underreporting or omission devices.²⁰ wearable of data by the

Therefore, the effectiveness and sensitivity of the device should be evaluated before use. Despite including literature on wearable devices, no statistically significant differences in outcomes were observed in the present study.

Future research directions

Traditional wearable VR devices typically rely on monitors or tablets, which limit user portability and flexibility,⁶⁵ however, the portability of smartphones provides users with the flexibility to train at any location.⁶⁶ Future research should explore the use of smartphones as visual display devices for VR training. This innovative approach could increase the acceptability and convenience of rehabilitation training and potentially reduce costs and technical barriers. Leveraging smartphones may render VR training more accessible and more easily integrated into daily life, enabling individuals to benefit from personalized rehabilitation programs at home or in other personal spaces. Future research should prioritize dose matching to ensure consistency and comparability of interventions across studies. This includes standardizing the frequency, duration, and intensity of wearable sensor device use. Clear definitions and reporting of these parameters are essential for meaningful comparisons and metaanalyses. Furthermore, continuous advancements in wearable technology should be leveraged to improve the accuracy, usability, and affordability of wearable sensors. Advanced technologies, such as machine learning and artificial intelligence, should be integrated to enhance device functionality and data analysis.

Limitations

The results of the present study may be limited by several factors. First, differences in the intervention duration and conventional rehabilitation measures might limit the

ability to judge the results. Secondly, the data were collected post-intervention, and the long-term therapeutic effects remain unclear. Therefore, standardized, highquality RCTs with large sample sizes are required to verify the effectiveness of wearable device rehabilitation training. Thirdly, the present discussion of the clinical evidence of wearable devices for improving upper limb motor functions after stroke primarily focused on experimental studies, namely RCTs. Future research should include more diverse study designs, including RCTs and observational studies, to explore the research question from different perspectives, thereby providing a more comprehensive understanding.

Conclusion

Meta-analysis and network meta-analysis were conducted to assess the therapeutic efficacy of wearable device-based rehabilitation training for improving upper limb motor function in patients with stroke. The results indicate that wearable device-based rehabilitation training has a positive impact on the enhancement of upper limb motor function and gross hand coordination in these patients. Furthermore, probability ranking revealed that wearable orthotic devices vielded the most optimal outcomes. However, no improvement in fine hand movements, muscle spasticity, or movement frequency was observed in the present study population. Different types of wearable devices vary in convenience and rehabilitation mechanisms. Therefore, individual factors, rehabilitation needs, convenience, and rehabilitation goals should be considered when selecting a suitable wearable device for a patient.

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Author contributions

Qin Qin and Lorna Kwai-Ping Suen conceived the study idea and designed the meta-analysis. Qianqian Song and Guangmei Liang contributed to data acquisition and analysis. Qianqian Song drafted the manuscript and interpreted the data. Haixia Qin and Lingling Zhang critically revised the manuscript for important intellectual content. All authors approved the final manuscript.

Data availability statement

Data are available from the corresponding author upon reasonable request.

Declaration of conflicting interest

The authors declare that there are no conflicts of interest.

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ORCID iD

Qianqian Song D https://orcid.org/0009-0001-3449-8359

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