experts, biologists, and dermatologists will be required to expand the scope of this research.

The promise of an AI revolution in dermatology also comes with an accompanying fear of black boxes and a concern for how this may affect patient care and patient perceptions of care. Similarly, there is a prevailing fear among physicians that machines will largely replace clinicians in dermatology, as well as in radiology and pathology. ¹³ It is our view that ML will not replace dermatologists. ¹⁴ Rather, these tools will enable dermatologists to provide a higher quality of care to their patients. ¹⁵ In fact, we believe that ML tools, such as downloadable local programs on personal computers, open-source online webservers, or mobile applications on smartphones, will be tightly integrated into the daily clinical practice of the dermatologist in the near future.

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Interactive skin self-examination digital platforms for the prevention of skin cancer: A narrative literature review



To the Editor: Skin self-examination (SSE) is an important secondary prevention strategy to reduce melanoma deaths. Noninteractive teaching and facilitating aids (eg, brochures, handouts) to promote SSE behaviors have been recently complemented by interactive digital platforms, including mobile health (mHealth) apps. Although digital platforms may provide increased engagement and broader access, their safety and utility for improving health outcomes are unclear. We sought to identify and describe the methodology, teaching and facilitating aids, and outcomes of published SSE intervention studies using interactive platforms.

Table I. Characteristics of included studies*

						Sample			Types of aid	ls included
	Study, country	Study design	Key inclusion criteria	Setting	Study time period	-	Age, y	Female sex, %	Teaching	Facilitating
Nonmobile digital platform	Study 1, USA	Type: Randomized controlled trial, 3 mo Intervention group: Skin safe program administered in a single sitting on a laptop computer SSE role play tutorial telecommunication monthly reminders to perform SSE (text, email, phone call, or letter), and SSE brochure Control group: SSE brochure only	Patients, family members, caregivers, friends at dermatology clinic	Outpatient dermatology clinics	2010	210	53 (mean)	61	Integrated in desktop application via modules: • Educational content on SSE within intervention (module) • Personalized content, feedback, advice • Risk calculator • Quiz to test knowledge gained from educational content	-
	Studies 2-5, USA	Type: Randomized controlled trial, 12 mo Intervention group Website with melanoma prevention messaging and information, per sonal risk for developing melanoma, a chat room interactive section for family members, and telecommunication reminders (e-mail every 3 months) to view website Control group: not specified	Internet access Family triads: melanoma case, first-degree relative, relative with child age 18 or younger	SEER Registry at Cancer Center and Regional Sit of the Cancer Genetics Network		311 family triads	s 56 (mean)	56	intervention:	Through participants' devices: • E-mail reminders
	Study 6, England	Type: Cluster- randomized controlled trial, 6 mo	3 r	5 pairs of family practices (rural, 1 urban, 3 suburban)	1 1998	589	38 (mean)	80	Integrated in desktop application via modules: • Educational content on SSE within intervention (module) • Personalized content, feedback, advice • Risk calculator	None

Integrated into experimental

Studies 7-8, USA	 Type: Randomized controlled trial with 3 groups, 3 mo Control group: Not specified Control website intervention group: Provided Skin Cancer Foundation's website, which included educational topics on prevention, educational information on skin cancer, news, and other relevant topics. Received telecommunication reminders (e-mail) Experimental website intervention group: Tailored interactive multimedia internet intervention program (UV4.me), which used multiple media formats (text, audio/video, images). Website had educational modules on risk or prevention behaviors with algorithms using data from baseline survey to personalize content and a goal setting section. Received 	or on	200	zz (mean)	00	intervention: website intervention: • Educational content on • Goal setting SSE within intervention • Personalized content, feedback, advice
Studies 9-11, USA	telecommunication (e-mail) reminders • Type: Randomized controlled trial with 3 groups, 24 mo treatment of AJCC star		5 494	55 (mean)	51	Integrated into web Integrated into web intervention delivered via intervention delivered via tablet:
	Control group: Treatment of No history of ocular mucosal melanoma Intervention groups: Provided laminated ABCDE	or				Seducational content on Goal setting SSE within intervention Quiz to test knowledge gained from educational content Diary Body maps Ruler Lighted magnifying lens
	were given skill quizzes					Continued

2014

965

22 (mean)

Integrated into website

Studies 7-8, USA

• Type: Randomized

Age: 18-25 y

Online

Table I. Cont'd

					Sample			Types of aids included	
Study, country	Study design	Key inclusion criteria	Setting	Study time period	_	Age, y	Female sex, %	Teaching	Facilitatin
	• In-person intervention	n							
	group: A scripted in	n-							
	person presentatio	n							
	delivered by researc	h							
	team member								
	 Take-home bookle 	et							
	intervention group	p:							
	Read workbook during	g							
	baseline visit and	d							
	brought it home								
	 Tablet intervention 	n							
	group: Electronic web	D-							
	based interactive inter	r-							
	vention delivered on	a							
	tablet in clinic only								
Study 12, Australia	 Type: Randomized 	• Age: ≥18 years 3	urban and 1 rural gene	ral 2016	272	46 (mean)	72	Via web-based application: No	ne
	controlled trial, 6 wk	 No personal history of 	medical practices					 Personalized content, 	
	• Control group: Bookle	et melanoma						feedback, advice	
	with generic melanoma	a- • Not overburdened by						 Risk calculator 	
	prevention information	n disease							
	on risk factors and pre	e-							
	vention measures								
	 Intervention group 	p:							
	Booklet plus real-time	e,							
	on-screen and printe	d							
	model-generated mela	a-							
	noma risk assessmen	nt							
	with tailored melanoma	a-							
	prevention advice								
Studies 13-16, Australia	 Type: Randomized 	 No personal history of 	ommunity (Queensland	2012-2013	546	32 (mean)	67	Via texting: No	ne
	controlled trial with 3	melanoma	electoral roll or Medic	are				 Educational content on 	
	groups, 12 mo	 Ownership of mobile 	register)					SSE within intervention	
	 All groups: Received 2 	1 phone						 Personalized content, 	
	text messages about as	s-						feedback, advice	
	signed topic. Message	25							
	personalized by partic	i-							
	pant's name, sex, base	e-							
	line skin cancer ris	sk							
	factors, and dat	ta							
	collected during baselin	ie							
	interview								
	 Randomized to text mes 								
	sages on (1) physical ac								
	tivity (control), (2) su	n							
	protection, or (3) SSE								

Mobile digital Studies 17-20, Australia platform	controlled trial, 5 mo Control group: SSE by naked eye examination alone. Web-based instructions on SSE performance. Telecommunication reminders (e-mail, text, phone call) if SSEs	 Age: ≥18 y ≥ 2 risk factors for skir cancer History of skin excisions or numerous nevi Smartphone compatible with dermoscope Partner, relative, or friend available to assist No history of melanoma in past 5 y 	, ,	2017-2018	199	41 (mean)	71	Accessed via web (not mHealth app): • Educational content or SSE within intervention	Through participants' devices: • E-mail reminders • Phone call reminders • Text reminders Nondigital aids: • Dermoscope
		 Age: ≥18 y ≥1 criterion for having high risk of melanoma 	Community	2013	49	Min 50; max 64	51	None	Integrated in mHealth app: • Photo storage/skin history
	moscopy (FotoFinder handyscope app) without specific SSE instructions • Group 2: Mobile teledermoscopy (FotoFinder handyscope app) with 10-step guide on how	nigii iisk üi meiaiionia							Ability to write note on photo in app Send photo from app to specialist
Study 23, USA	 to conduct thorough SSE Type: Randomized trial with 4 groups, 6 mo 	• Age: ≥18 y	Pigmented lesion clinic	2015-2017	69	54 (mean)	64	Integrated in mHealth app:	Integrated in mHealth app:

Table I. Cont'd

iPad in MelaWatch app plus standard of care

					Sample			Types of aids included		
Study, country	Study design	Key inclusion criteria	a Setting	Study time period	-	Age, y	Female sex, %	Teaching	Facilita	ting
-	App-only intervention	n • Ownership of iPhone of	or	-			-	Educational content or	n • Photo	storage/skin
	group: Mobile app or	n iPad						SSE within intervention	history	
	iPhone/iPad (MelaSight	e) • Already in possession of	of					 Personalized content 	t, • Side-by-side	compari-
	loaded with patient's to	total body photographs	i					feedback, advice	sons of pho	tos
	tal body photograph	S							 Ability to w 	rite note on
	and instructions on SSI	E							photo in ap	0
	performance plus stan	ı -							 Ability to 	flag/mark
	dard of care (printed	d							changes or	photo for
	photos and CD contain	ı -							follow-up w	th clinician
	ing digital total body	y							 Skin exam 	metrics in
	photograph images)								monthly per	formance re-
	 Reminders intervention 	n							ports	
	group: Mobile app plu	S							Through	participants'
	monthly telecommuni	-							devices:	
	cation reminders (e-mai	il							 Text remind 	
	and text) plus standard	d							 E-mail remir 	ders
	of care									
	 Accountability partne 	r								
	intervention group: Mo	-								
	bile app plus account	÷								
	ability partner received	d								
	monthly report of their	r								
	skin exam progress plu	S								
	standard of care									
	 Reminders-account- 									
	ability partner interven	I -								
	tion group: Mobile app	p								
	plus monthly reminder	S								
	(e-mail and text) plu	S								
	accountability partne	r								
	plus standard of care									
Study 24, USA	 Type: Randomized 	 Patients new to total 	Pigmented lesion clinic	2015-2016	71	47 (mean)	59	None	Integrated in mi	Health app:
	controlled trial, 7 mo	body photography							Photo	storage/skin
	 Control group: Standard 	d							history	
	of care, including printed	d								
	photos and CD contain	-								
	ing digital copies of tota	al								
	body photographs									
	 Intervention group: Total 	ıl								
	body photograph	S								
	loaded onto iPhone	/								

• Coi car • Int Ski par rec ten to	ntrolled trial, 12 mo on trol group: Usual e at family practice ervention group: MynPal app loaded onto otticipant phone, eived verbal and writtinstructions on how use app and monthly	Age: 18-75 y Identified as increased risk of melanoma based on MelaTools Q risk assessment tool Ownership of a smartphone	l :	2016-2017	238	55 (median)	55	Integrated in mHealth app: Photo storage/skin history Body map Gamification Through participants' devices: Text reminders	•
(te: app Studies 27-29, Scotland • Typ stu • Rec trai	oe: Nonrandomized	 Age: ≥18 y Treated for melanoma in past 5 y Currently receiving hospital-based follow-up 	ı	NR	20	Min 37, max 83	50	Educational content or SSE within intervention	o: Integrated in mHealth app: n • Photo storage/skin history t, • Checklist for SSE • Ability to write note on
det ver me cor one	ICA app in addition to a tailed manuals. Inter- ntion was experi- ental and implementary to their going care. Tracked anges in app to send	 No nodal involvement or metastases 							photo in app Ability to flag/mark changes on photo for follow-up with clinician Body map
Rec cat tex	oort to study staff. ceived telecommuni- ion reminders (e-mail, t, or phone call) to rform TSSE								Through participants' devices: Text reminders Phone call from overseeing specialist nurse to reassure patient or schedule in-person assessment for flagged changes E-mail reminders
nes Per car ph abl Ser ger dat Sep Ap		 Informed consent only 	Online	2016-2017	3326	32.7 (mean)	60	•	Integrated in mHealth app: • Gamification (e.g. ability to send challenges to

Table I. Cont'd

					Sample		_	Types of aids included		
Study, country	Study design	Key inclusion criteria	Setting	Study time period	size, No.	Age, y	Female sex, %	Teaching	Facilitating	
	level/score of cancer pro	e-			_					
	vention using weighte	ed								
	values for different be	e-								
	haviors (physical activit	y,								
	fruit/vegetable intak	e,								
	tobacco and alcoh-	ol								
	consumption, boo	ły								
	mass index, sunbur	n,								
	SSE, Papanicolaou tes	st,								
	breast self-exam, testi	c-								
	ular self-exam, huma	ın								
	papilloma virus vaccina	a-								
	tion). App feature	es								
	included daily question	ns								
	at bedtime with recalcu	u-								
	lation of behavioral pro	e-								
	vention score, tailore	ed								
	messages via push not	ti-								
	fications, integration	on								
	with social networking	ng								
	(Facebook), challenge	es								
	designed to give use	rs								
	goals, and milestones/re	e-								
	wards (gamification)									

ABCDE, Asymmetry, border, color, diameter, elevation; AJCC, American Joint Committee on Cancer; ASICA, Achieving Self-Directed Integrated Cancer Aftercare; NR, not reported; SEER, Surveillance, Epidemiology, and End Results; SSE, skin self-examination; TSSE, thorough skin self-examination; USA, United States of America.

*Citations of included studies not included due to journal citation constraints and are available from the authors upon request.

Table II. Behavioral, engagement, and psychosocial outcomes of teaching and facilitating aids for skin self-examination in studies using digital platforms

			Beh	avioral outco	mes [†]		En	gagement o	itcomes [†]		Psychosocial outcomes*			
Study type	Study*	SSE frequency [‡]	TSSE frequency [‡]	SSE or TSSE ever performed [§]	Body areas examined	Skin cancer detection (diagnostic accuracy)	Satisfaction with intervention	Feasibility	Usability	Engagement	SSE confidence	Risk perception	Intention to perform SSE	
Nonmobile	Study 1			/							✓	*		
digital	Studies 2-5				?									
	Study 6						?					Χ		
	Studies 7-8													
	Studies 9-11	1							?					
	Study 12							?				Χ	Χ	
	Studies 13-16		Χ		?		?			?				
Mobile	Studies 17-20	Χ		Χ	?	Χ	?		?		?	?	?	
digital	Studies 21-22					?		?	?		?		?	
platform	Study 23	?					?							
	Study 24	Χ												
	Studies 25-26 [§]										Χ	?	Χ	
	Studies 27-29		?		?			?			?		?	
	Studies 30-32			?				?	?	?				

SSE, Skin self-examination; TSSE, thorough skin self-examination; M, outcome was assessed and a positive statistically significant improvement was reported compared with a control group; X, outcome was assessed, and no statistically significant improvement was reported compared with a control group; ?, outcome was assessed, but results were not reported or not relevant, significance of findings were not reported, results were descriptively reported, results were not compared with a control group (ie, before-after intervention study), or final study results were not yet published; blank cell, outcome was not assessed.

^{*}Citations of included studies not included due to journal citation constraints and are available from the authors upon request.

[†]Not all outcomes were included because only the most frequently overlapping behavioral, engagement, and psychosocial outcomes related to interactive teaching and facilitating aids for skin self-examination are shown.

[‡]Measured by asking participants to report how many times they performed a skin self-examination during the study period.

[§]Measured in a yes/no format.

Although this is a clinical outcome, we categorized it as under behavioral outcomes for simplicity.

This study did not require ethical approval. A PubMed/MEDLINE comprehensive of search was conducted on March 13, 2020, without date/language restrictions (Supplemental Methods via Mendeley at https://doi.org/10.17632/9fhmfgx8y5.1). We considered a platform to be interactive if it responded to a user's input or if it delivered personalized information. Mobile digital platforms (MDP) were defined as applications on smartphones or tablets (ie, mHealth apps). Nonmobile digital platforms (NMDP) were defined as web browsers, desktop applications, or text messages. We identified 487 articles, and 32 were included. These 32 articles described 14 unique studies; 12 were randomized and 2 were nonrandomized (Table I and Supplemental Table I). Seven studies used apps on MDPs.

There were 4 teaching and 16 facilitating digital aids identified (Table I and Supplemental Table I). Studies with NMDP primarily used teaching aids, such as educational content on SSE within an intervention (n=10). Conversely, facilitating aids were primarily found in studies with MDPs; the most common included photo storage (n=5), texts (n=5), and e-mails (n=5). There was significant heterogeneity in the behavioral, psychosocial, and engagement outcomes reported (Table II). Seven studies (53%) demonstrated a statistically significant improvement in ≥ 1 outcome; most (n=6) were in studies using NMDPs. Only 1 study demonstrated an improvement in melanoma detection compared with a control group.

Despite the proliferation of interactive digital tools to aid SSE, there is limited literature to gauge the quality and impact of published interventions. Researchers are shifting away from web-based and print-based materials to aids located on devices. With 81% of United States adults owning a smartphone in 2018, mHealth apps are a promising resource to encourage preventive health behaviors. However, because smartphone ownership varies by age, income, education, and community, alternative strategies must be simultaneously developed to prevent health care disparities. At this juncture, mHealth apps have used a greater diversity of teaching and facilitating aids than NMDPs.

Despite these possible benefits, our review identified only 1 study using an mHealth app that showed an improvement in an SSE outcome compared with a control group. This finding echoes concerns that there is a mismatch between the promise of mHealth apps and the evidence supporting their validity and utility.^{3,4}

To personalize the delivery of health care to patients, future studies should also aim to better assess the usability and functionality of mHealth apps. Given the variability in the study interventions, it was not possible to determine the effectiveness of

different interventions and compare studies that use non-overlapping measures in different contexts. However, our findings may help standardize future research efforts in this domain.

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