## A systematic review of tailored eHealth interventions for weight loss

### Kathleen Ryan (), Samantha Dockray and Conor Linehan

### Abstract

**Objective:** The aim of this study is to review the evidence for tailored eHealth weight-loss interventions, describing in detail: 1. how tailoring was implemented in these studies and 2. whether these tailored approaches were effective in producing weight loss compared with generic or inactive controls.

**Methods:** A systematic review was carried out. Five databases were searched up until 15 March, 2018, including: EBSCO, Science Direct, Pubmed, EMBASE and Web of Science, using combinations of the concepts 'tailoring', 'eHealth' and 'overweight'.

**Results:** Eight articles relating to six interventions were accepted. Tailoring was carried out in a number of ways, based on, for example, anthropometric data, health-related behaviours (e.g. dietary intake, physical activity), goals (e.g. weight goal), theoretical determinants (e.g. confidence/willingness to change behaviours), psychosocial factors (e.g. social support) and participant location. Systems acquired data using strategies that ranged from online questionnaire administration, to the dynamic gathering of data from web-based diaries, websites, mobile applications and SMS messaging. Tailored interventions were more effective in supporting weight loss than generic or waitlist controls in four of the six articles. Effect sizes were very small to moderate, with evidence for fluctuations in effect sizes and differences of effect between tailoring and non-tailoring interventions, and between tailoring types, over time.

**Conclusions:** We contribute an enhanced understanding of the variety of methods used for the tailoring of eHealth interventions for weight loss and propose a model for categorising tailoring approaches.

### **Keywords**

Tailoring, weight loss, eHealth, obesity, health behaviour change, systematic review, engagement, personalisation

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### Introduction

There is an urgent need to identify effective and scalable strategies to address what the World Health Organization terms 'the obesity epidemic'. Indeed, the number of overweight and obese individuals has increased from 857 million in 1980 to 2.1 billion in 2013.<sup>1</sup> However, we have yet to see strong evidence for a successful population-wide obesity intervention. In standard lifestyle interventions, behavioural therapy is delivered to participants by trained interventionists to support modification of diet and physical activity levels in an individual or group setting; it has been found to be effective in producing weight loss when compared with pharmacological strategies.<sup>2</sup> This approach is resource-intensive, requiring significant input from both the patient and the interventionist, and can be costly to deliver, which poses a challenge for widespread translation of these approaches.<sup>3</sup> Accessible and cost-effective interventions are necessary to bring about impact on a large scale.

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Two developments that appear promising for scalable, effective interventions are the use of eHealth tools in the delivery of interventions and tailoring of intervention content.

### eHealth approaches to treating obesity

Research suggests that by delivering interventions via the Internet (eHealth), some of the cost and scalability problems of obesity interventions are mitigated.<sup>4,5</sup> eHealth interventions can be deployed through websites, emails, text messages, monitoring devices, mobile applications, computer programmes, podcasts and personal digital assistants.<sup>5</sup> eHealth systems can set individualised goals for users, monitor diet and physical activity behaviours, and respond to data with personalised guidance, among other behaviourchange strategies.

To establish the viability of these systems for largescale impact, research into the implementation and outcomes of eHealth interventions is needed. Hutchesson et al.<sup>5</sup> conducted a series of meta-analyses drawing on 88 eHealth weight loss or weight maintenance interventions. eHealth weight-loss interventions achieved modest weight-loss compared with no treatment (mean difference: -2.70 [-3.33, -2.08], p < 0.001; nine studies pooled in meta-analysis) or minimal treatment (mean difference: -1.40 [-1.98, -0.82], p < 0.001; 16 studies pooled in meta-analysis). Studies were found to be heterogeneous in both groups of pooled studies,  $I^2 = 49\%$ , p = 0.04 and  $(I^2 = 72\%)$ , p < 0.001, respectively in terms of intervention strategy, technology usage and interaction style. eHealth interventions with extra features/behavioural components (e.g. selfmonitoring, personalised feedback) or technologies (e.g. applications, texts or social networks) were more effective than standard eHealth programmes (drawing on either the Internet or podcasts only) (mean difference: 1.46 [0.80, 2.13], p < 0.001).<sup>5</sup> Arem and Irwin<sup>6</sup> also found the Internet-based weight-loss interventions included in their review to be heterogeneous. This was both in terms of the nature of the intervention (e.g. contact with participants, sample size and study duration) and their outcomes, with weight loss ranging from <1 kg to 4.9 kg. Crucially, since the interventions included in these reviews differed from each other in so many ways, it was impossible for the authors to make specific recommendations regarding which intervention components were necessary to bring about weight loss.

These effect sizes are modest in comparison with what may be achieved using traditional behavioural approaches. The modest weight losses demonstrated by the Hutchesson<sup>5</sup> review are several times less than in-person interventions. For example, an in-person

behavioural weight-loss intervention of 4–6 months of weekly group-based lifestyle counselling and intensive dietary restriction showed that participants lose on average 10.7 kg.<sup>7</sup> These may be due in part to the varying intensity of behavioural-change strategies, personalisation strategies and supportive accountability provided by a human counsellor. To ascertain the effective components in these interventions, it is necessary to draw together the interventions that implement a common strategy.

A significant challenge to the wide-scale usage of eHealth interventions is that they typically suffer a steep decline in usage over the course of the intervention and have poor completion rates.<sup>4</sup> Neve et al.<sup>8</sup> highlight that use of eHealth technological components (both frequency and duration) may prove a useful metric to enable the establishment of an 'optimal dose' required for behaviour change. The review demonstrates that the higher the number of log-ins, self-monitoring occasions, chatroom attendances and bulletin board posts, resulted in greater weight loss, and this pattern was evident in the majority of studies included. Participant attrition is a key issue in assessing the efficacy of Internet-based interventions, as low levels of completion compromise the feasibility of implementing such interventions on a larger scale.

These findings provide promising support for the efficacy of eHealth interventions for weight loss.<sup>5,6</sup> They confer an additional advantage over traditional methods, for example, in-person behavioural counselling in delivering weight-loss interventions in a cost-effective, effectual manner suited equally to the individual and at a scale needed to address the global obesity epidemic. Efforts to investigate specific strategies that are effective for weight loss, which address the issues of modest effects and low levels of engagement are needed. Delineating the precise strategies within eHealth efforts that are effective in bringing about the greatest decrease in weight requires a granulated examination and specification of the various components, modes of delivery, technological features and behavioural theory that inform these eHealth approaches, as well of usage of technological components as part of these efforts. Web-based weight-loss interventions have been reported as being less effective than face-to-face interventions,<sup>6,9</sup> however, enhanced web-based interventions (e.g. more interactive and tailored) are more efficacious than basic, more generic interventions (e.g. purely informational).<sup>9</sup> Web-based interventions may prove to be clinically effective and feasible if they can mimic some of the tools and strategies of the in-person interventions, while reaching a larger population.<sup>6</sup> Enhancing web-based interventions through tailoring the eHealth interventions to characteristics of the user or patterns of behaviour is supported by a recent systematic review that evaluated web-based interventions for weight loss and maintenance.<sup>9</sup>

## Tailoring as a strategy to enhance eHealth interventions

Tailoring is a process whereby the provision of information, advice and support is individualised to the user.<sup>10</sup> Noar et al.<sup>11</sup> explain that the process of tailoring involves an assessment of individual-level characteristics gathered by a person or self-administered (input: the basis for tailoring). This information then gets processed by either a human (human tailoring), or an expert system (computer tailoring) that uses algorithms to select content (i.e. text, images, recommendations and intervention messages) from an expert-developed database for the individual (tailoring process). Tailored material (output) is adapted using a variety of strategies to be delivered to the individual via multiple delivery modes.

Compared with non-tailored, generic materials, tailored health-messages command greater attention and are more likely to be read, elaborated upon, recalled and understood.<sup>12</sup> As such, tailored health-messages are considered to be processed more deeply, contain less redundant information, and are perceived more positively by health consumers.<sup>13</sup> Tailoring strategies can range from relatively simple, for example, employing the user's name, to the more complex, for instance adapting content to personally relevant variables.14 Work has also been undertaken to draw together concrete definitions of tailoring strategies.<sup>13,15,16</sup> Krebs et al.<sup>17</sup> add to this by defining 'dynamic tailoring' as the assessment of intervention variables prior to each feedback and 'static tailoring' where one baseline assessment provides the basis of all subsequent feedback.

### Impact of tailored approaches on health outcomes

While there is little evidence for the impact of tailored eHealth interventions for weight loss, evidence has accrued for the effectiveness of tailored web-based approaches to health interventions. For example, Lustria et al.<sup>10</sup> conducted a meta-analysis to assess the impact of tailored web-based interventions targeting physical activity, nutrition, smoking/tobacco use, drinking, medication adherence (asthma management), stress management and faecal soiling (encopresis). Forty experimental and quasi-experimental studies were analysed and web-based tailored interventions effected significantly greater improvement in health outcomes compared with controls both at post-testing, d=0.139 (95% CI=0.111, 0.166, p < 0.001, k=40) and at

follow-up, d = 0.158 (95% CI = 0.124, 0.192, p < 0.001, k = 21). While these results are encouraging, (similarly to the eHealth reviews discussed earlier<sup>5,6</sup>) the authors identified that there was great variability in how tailoring was carried out among their included articles, including differences in intervention features, formats, and levels of interactivity.

A series of meta-analyses by Krebs et al.<sup>17</sup> also assessed the effect of 88 computer-tailored interventions, using computer, print, or telephone communication channels that focused on four health behaviours: smoking cessation, physical activity, diet, and mammography screening. A significant, small effect size (Hedges's g = 0.17) was found for tailored interventions, taking the average of the four health behaviours. This is encouraging evidence for tailoring as an approach to health behaviour change in general, but more specific approaches, honing in on one specific outcome may allow for in-depth conclusions to be drawn, considering that the determinants of health behaviour change interventions differ by outcome.

These studies provide support for tailoring as a strategy for use in health behaviour-change interventions, but we argue that previous research on tailored health interventions has focused primarily on understanding *whether* tailoring works, rather than focusing on *how* it works, or which approaches are most useful under which circumstances.<sup>12,15</sup>

### The current study

Without meticulous description of intervention designs, the science and practice of tailoring within eHealth intervention design and implementation cannot be advanced. Previous work on tailoring has met with difficulty in differentiating the range of approaches to tailoring, differences in modality, intervention features and components.<sup>10</sup> In addition, Harrington and Noar<sup>16</sup> have called for improved reporting standards in tailored research.

The aim of the current paper was to review the evidence for tailored eHealth weight-loss interventions. We aim to describe in detail: 1. how tailoring was implemented in these studies and 2. whether these tailored approaches were effective in producing weight loss compared with generic or inactive controls. Specifically, we sought to identify what individual factors were assessed as part of the tailoring process, what tailoring strategies and eHealth tools were used, how engagement with eHealth interventions was conceptualised and whether tailoring increased this and, lastly, whether tailored approaches produced larger effect sizes than generic information and waitlist control approaches for weight loss.

### **Methods**

### Design

Systematic review of pre–post intervention trials with a control group, following Cochrane methodology and Preferred Reporting Items for Systematic Reviews and Meta-analyses (PRISMA) guidelines (see Appendix A).<sup>18</sup> The review was registered with PROSPERO (CRD42017072901).<sup>19</sup>

A systematic review of tailored eHealth interventions for weight loss was carried out, with a title and abstract screening phase (Phase 1) and a full text screening phase (Phase 2).

### Eligibility criteria

All papers that were deemed relevant at Phase 1 were included if they were quantitative peer-reviewed articles, published at any time, in the English language, with a pre-test-post-test intervention design (either randomised controlled trials (RCT) or pilot study) using a control comparison group (generic or inactive) approach. Interventions of any duration that aimed to bring about weight loss in overweight or obese (Body Mass Index (BMI) >25), free-living participants without specified clinical disease were included. Articles were required to have a stated tailored approach, which we operationalised as 'if the intervention was personalised, and based on an assessment of individual characteristics'.<sup>20</sup> The intervention was required to be behavioural in nature (e.g. aimed to change diet and physical activity), and delivered primarily via the Internet (Internet-based); in-person measurement of weight was permissible, as was briefing to randomisation group components. Articles were excluded if the intervention was not tailored at the individual level (e.g. generic in nature or tailored based on shared rather than individual characteristics, e.g. gender tailoring), not delivered online (e.g. delivered in-person), not aiming to change behaviours (e.g. pharmacological), and if the primary aim of the intervention was not specifically for weight loss; however, articles were included if there was more than one main aim, as long as weight loss was stated as one of the aims. Articles were excluded if they stated that the participants had a clinical population (e.g. type 2 Diabetes patients who may have been taking medication to control their condition), or if they were not published in the English language. Authors of studies that met the criteria but did not report the body weight of participants at baseline and follow-up (e.g. reported the BMI only) were contacted in an attempt to include them in the review.

### Sources and Searching

Searches were run in December 2016 in five databases, including: EBSCO, Science Direct, Pubmed, EMBASE and Web of Science. Combinations of the concepts 'tailoring', 'eHealth' and 'overweight' were searched as part of the search strategy. For example, EBSCO search included the following combination: TI personalis\* OR individualis\* OR tailor\* OR custom\* AND AB (ehealth or e-health or website or web or web-based or website-based or online or Internet) AND SU (overweight or obesity or obese or weight management or weightloss or weight loss or diet\* or physical activity). Searches were re-run on 15 March, 2018. An example of the search strategy is included in Supplemental material (Appendix B). Hand-searching was conducted of reference lists of included articles, systematic reviews and meta-analyses of eHealth interventions for weight loss, as well as in the Journal of Medical Internet Research.

### Data extraction

Data extraction was conducted by KR. Fifty per cent of the extracted data were checked for accuracy by author SD and 50% by author CL. Two accepted articles had associated adjunctive articles, published separately but reporting details of the development of the tailored eHealth weight-loss intervention. Data from these articles were also included in the extraction and qualitative synthesis, in tandem with the original accepted article. The extracted information included participant information, intervention information and weight-loss outcomes. This included details on study setting; study population and participant demographics and baseline characteristics; details of the tailoring intervention and control conditions; study methodology; recruitment and study completion rates; outcomes and times of measurement; suggested mechanisms of intervention action; and information for assessment of the risk of bias. Specific to the process of tailoring, we identified the individual factors that were assessed as part of the tailoring process, how the assessment was conducted (e.g. what eHealth technology/modes of delivery were used), how often it occurred (e.g. static tailoring or dynamic), how the message output was devised, what specific tailoring and behavioural strategies were implemented, how engagement with the eHealth intervention was conceptualised and measured, whether tailoring increased this and, lastly, whether tailored approaches produced greater weight loss compared with generic or waitlist approaches. Authors were contacted for any missing information.

### Data analysis

There were two components to data analysis: qualitative and quantitative synthesis. Qualitative synthesis involved examining how tailoring approaches were conducted within each intervention using information in each article's intervention description and any corresponding intervention development articles.<sup>21,22</sup> As part of this process, terms listed in Harrington and Noar<sup>16</sup> were used to identify the tailoring strategies described. These included the following terms: 'content-matching' (adapting intervention content based on theoretical determinants), 'feedback' (providing messages to individuals about their psychological or behavioural states, which may involve description of objective data back to participant ('descriptive feedback'), comparing their data to norms ('comparativenormative feedback'), their previous states ('comparative-progress feedback') and providing an interpretation or judgement of their data ('evaluative feedback') and 'personalisation' where recognisable information relating to the participant is relayed to them, including the use of their name ('identification'), telling participants that the intervention content has been created especially for them ('raising expectation of customisation') and framing messages in a context that is meaningful to them, e.g. through gender-, cultural- or ethnicity-related cues ('contextualisation').

Quantitative synthesis involved evaluating the effect of the tailored eHealth interventions on weight loss: effect sizes were calculated to estimate the difference between the intervention and control groups (both active and waitlist controls). Effect sizes for weight change at follow-up assessments were calculated based on the mean pre-post change in the treatment group minus the mean pre-post change in the control group, divided by the pooled pre-test standard deviation (SD).<sup>23</sup> In one case, the post-test SD was used as a proxy where the pre-test SD was not reported.<sup>24</sup> A pooled SD was calculated using Cohen's<sup>25</sup> formula, squaring the baseline SD of weight of the intervention group and adding the squared baseline SD for weight for the control group, and taking the square root of these. Effect sizes are therefore reported in SD units of change. Changes between the intervention and control were considered to be small (0.2 to <0.5); moderate  $(0.5 \text{ to } < 0.8) \text{ or large } (>0.8).^{25}$ 

### Quality appraisal

Quality of methodology was assessed by authors KR and SD using the Cochrane Collaboration's tool for assessing risk of bias for RCTs.<sup>26</sup> Bias was assessed as a judgement (high, low, or unclear) in the following domains of bias: selection, performance, detection,

attrition, reporting and other bias. In this review, as the outcome of significance was weight, we included an assessment of measurement bias for weight as part of the quality review. Studies using self-reported weight, or using one-off (not averaged) weight measurements were considered to have an increased risk of measurement bias.

### Results

Initially 516 articles were captured in the searches and a further 14 records were identified from hand-searching reference lists of accepted articles and relevant systematic reviews.<sup>5,6,8</sup> Following the removal of duplicates (N=218), there were 312 articles for title and abstract screening, of which 80 were subjected to full text review. Searches were re-run in March 2018, which led to an additional three articles suitable for full text review.<sup>27-29</sup> Articles were excluded due to relating to wrong outcomes (e.g. cardiovascular risk, BMI), wrong study design (e.g. review articles, face validity evaluations); being an RCT development article; wrong intervention aims (e.g. aiming to increase physical activity, diabetes prevention); wrong route of administrations (e.g. in-person), wrong population (e.g. those with diabetes mellitus) and wrong source (e.g. conference abstracts). A systematic review of tailored eHealth weight-loss interventions ultimately yielded a total of eight articles.<sup>21,22,24,30–34</sup> describing six interventions. See the PRISMA flow diagram below (Figure 1).

A summary of the included studies is presented in Table 1. There were 14 intervention arms, 2 interventions had three arms, and 4 interventions had two. There were N = 4356 participants (baseline), with 2243 assigned to the intervention groups, and 2113 assigned to the control groups. Participants were recruited from the general population (82.5%), university students (10.5%) and university staff (7%). The mean retention rate to final follow-up was 39.04%. The average age of participants was 37.11, SD = 7.54. Their mean BMI at baseline was 30.06, SD = 4.05. They were mostly female, 82.18% and Caucasian, 56.67%. Studies were published between 2006 and 2016. There were five RCTs<sup>30-34</sup> and one randomised pilot study<sup>24</sup> with the majority (N = 5) conducted in the  $USA^{24,30-33}$  and one in Europe.<sup>34</sup> The intervention durations ranged from 5 weeks to 24 months, with the average length being 24.5 weeks. There was an average of 2.33 follow-ups where participants' body weight data were collected. Four of the six studies measured weight objectively<sup>24,30,33,34</sup> with two relying on selfreport methods.<sup>31,32</sup> Four out of these six studies compensated participants for taking part.24,30,33,34

There were two articles considered to be of low risk of bias, three of moderate quality, and one of high risk.

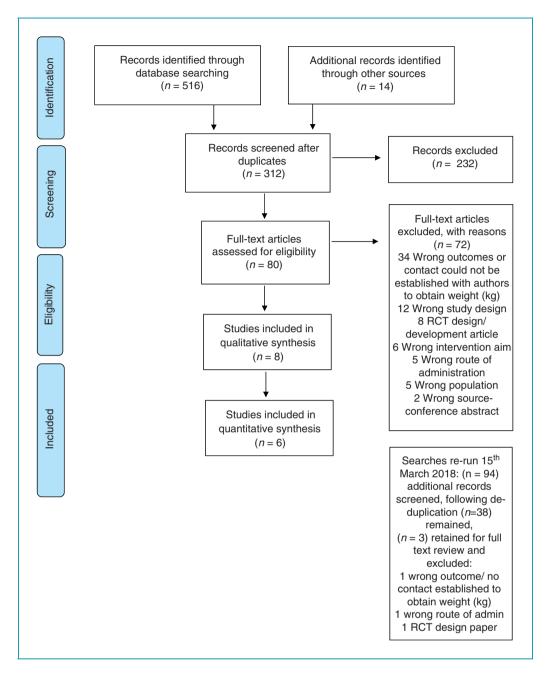


Figure 1. PRISMA flow diagram.

Random sequence generation, allocation concealment and blinding of participants and personnel were judged to be adequate strategies in three articles, blinding of outcomes assessment was described in two articles. Five articles were deemed to be low risk in regards to reporting incomplete outcome data. Two articles had published protocols and thus were considered to have a low risk of bias for selective outcome reporting. Two articles relied upon participants' self-reported weight for their analysis, and this was considered to have a high risk of bias in relation to the measurement and reporting of weight. One was considered to be low risk owing to the inclusion of two objective measurements of weight and taking the average. A summary of the risk of bias results is included in Figure 2.

### How was tailoring implemented?

There were six interventions included in the review, referring to 14 intervention arms, four articles

Study design	Randomised control trial (RCT)	RCT	RCT	, , , , , , , , , , , , , , , , , , ,
Outcomes (primary and secondary, where stated in article)	Dietary intake frequencies; weight loss; opinions of intervention (intervention group only)	Weight loss (body weight change): secondary out- comes: dietary intake, exercise	Anthropometric: (Body Mass Index, skin folds and waist circumference) and energy balance-related behav- iours (physical activity: intake of fat, snacks and sweetened drinks)	
Comparison	Waitlist control; completed three assessments at same time as intervention group evaluating dietary intake, weight	Baseline: in-person group session for randomisation; education (principles of behavioural weight-loss, diet, exercise and behav- iour change); orientation to website (differed by group), instruction on meal replacement usage (440 kcals - coupons given) with unstructured hird meal, increase exercise to 30 mins walking per day, self-monitor diet and exercise; access to website, which provided some- where to log weight, and see graphs of weight, weight usekly weight usekly weight loss tips via email, recipes, and a weight-loss e-buddy network system for comparison and peer support for weight loss via email	Log-in name and a password received by email for access to separate control website. Three web-based educational modules with general information on weight-gain prevention. The first module aimed to	
Intervention (duration; number of follow-ups; retention to final weight measurement)	5-week duration; two follow-ups (6 weeks; 8 weeks); 85% retention Register on study website; fill in online Personal Nutrition Planner questionnaire; receive personalised dietary guidelines; participants self-selected a long- term goal (e.g. lose weight, reduce risk of chronic disease), and short-term action for next 7 days rel- evant to this. Opt out/in for weekly newsletter with personalised information relating to their goal	<ul> <li>6-month duration; 2 follow-ups (3 and 6 months); 80.7% retention Two active intervention groups: computer-automated feedback (CT), or human email counselling (HC) Both HC and CT groups: access to separate website with electronic diary and within group message board, extra weekly email prompting logging of behaviour and educational content (as per the diabetes prevention programme)</li> <li>HC only: weekly feedback via email from health coun- sellor - unstructured and based on clinical judge- ment (HCs had behavioural weight-loss experience and degrees in nutrition, psychology, health educa- tion, exercise physiology) including answering ques- tions, behavioural feedback on erroise goals, weight loss feedback on calorie goals, consumed and burned compared with individualised diet and exercise goals, behavioural strategies for improving adherence to self-monitoring diet and exercise, overcoming barriers, and motivation or praise depending on logging frequency</li> </ul>	2-month duration; 2 follow-ups (at 1 and 6 months); 57.8% Log-in name and a password received by email, for access to intervention website. Participants were asked to visit the websites at least three or four times during a 2-month period. They received email reminders to visit the intervention every 2 weeks. At 1 month and 6 months after the intervention period,	
Population (number, participant type, recruitment strategy) and BMI (M, SD)	307 female university staff, recruited by email adver- tisement; BMI: M = 27.72; SD = 6.97	192 participants recruited from local newspaper; BMI: M = 32.7; SD = 3.5	539 participants, recruited by local newspapers, flyers delivered door-to-door/in waiting rooms of $GP^{*}s'$ among the employees of four large companies; BMI: M = 28.04; SD = 1.94	
Country	USA	ASU	Netherlands	
Article aim	To evaluate the impact of the Personal Nutrition Planner on dietary intake (dairy, fruit, vegetable intake) and weight loss	To determine the short-term efficacy of a self-disc pro- fintermet weight-loss pro- gramme (compared with the same programme sup- plemented with behaviou- ral counselling from either a computer-automated tailored system or from a human counsellor)	To evaluate the efficacy of the computer-tailored inter- vention in weight-related anthropometric measures and energy balance-relat- ed behaviours (physical activity: dietary intake)	
First author and Year	Mouttapa et al. (2011)	Tate et al. (2006)	Van Genugten et al. (2012) Design article: Van Genugten et al. (2010)	

Table 1. Characteristics of eligible studies.

Study design			(continued)
Stud		RCT	(co
Outcomes (primary and secondary, where stated in article)		Weight (at 24 months) sec- ondary: weight at 6,12,18 months; waist circumfer- ence (cm), arm circumfer- ence (cm), systolic blood pressure (mm Hg), dia- stolic blood pressure (mm Hg), heart rate (beats per min), and the level of	
Comparison	increase the motivation for weight management/ pre- vention. The second module provided informa- tion about possible behav- iour changes. The thind module provided general information about a healthy diet and safe physical activity	Active control: access given to an alternative website to intervention participants (asked to visit once per week) and were sent quarterly newsletters via email with generic health information (no specific behavioural strategies)	
Intervention (duration; number of follow-ups; retention to final weight measurement)	<ul> <li>participants were asked by email to fill out the online questionnaire again.</li> <li>The entire intervention could be finished in 90 minutes; The intervention consisted of four modules, each to be visited 1 week after the previous one and followed the string a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal for one relevant change in Dl or PA and making a goal-setting) and make a plan for what to change (guided goal-setting) and make a plan for where, when, and how to make that change (implementation plan).</li> <li>The second and third modules were focused around evaluation of progress towards behaviour change, and provided feedback on past week performance. If necessary, it supported adaptation of action and coping plans (when attempts to change behaviour coping plans (when attempts to change behaviour were unsuccesful). The fourth module aimed at provided feedback on past week performance. If necessary, it supported adaptation of body weight without use of the programme.</li> <li>A tool to monitor and evaluate (changes in) body weight without use of the programme.</li> <li>A tool to monitor and evaluate (changes in) body weight were unsuccesful). The fourth module aimed at promoting sustained as a short guideline with sequences of actions for long-term WGP, reflecting on the self-regulatory skills that had been practised in the rend, the participants signed a personalised contractives, and was accessible through the luteret.</li> </ul>	24-month duration; four follow-ups (at 6, 12, 18, 24 months); 84% retention. Access to range of six remote modes of delivery apps, social media networks, website, text/ emails and phone contact with health coaches. Intervention content was tailored to participants' physical location (e.g. text to determine location and advice depending on answer) and social environment (e.g. whether using apps, Facebook, study website). Tailoring	
Population (number, partici- pant type, recruitment strate- gy) and BMI (M, SD)		404 college students, recruited by print (e.g. newspapers, flyers, posters, magnets) and digital (e.g. emails, electronic bulletins, web- sites, and Facebook) advertisements; in-person student orientation events; BM: M= 28.5; SD = 2.75	
Country		USA	
Article aim		To assess the efficacy of a 2- year social and mobile intervention designed to reduce weight by improv- ing weight-related behav- iours among college students	
First author and Year		Godino et al. (2016); Design article: Patrick et al. (2014)	

Table 1. Continued

	Study design	int of iav- ake, sting tit, wway life, cial iet, iet,	pilot RCT goal fifica- t for -	
	Outcomes (primary and secondary, where stated in article)	engagement (ie, amount of use) of the intervention components, physical activity, sedentary behav- iours, total dietary intake, eating behaviours relating to weight management, sugar-sweetened beverage consumption, eating away from home, quality of life, depression, self-esteem, bddy image, psychosocial constructs relating to physical activity and diet, social support and social network composition (Facebook data)	Weight loss: secondary: phys- ical activity behaviour: goal setting and planning: physical activity self-effica- cy: weight self-efficacy: adapted social support for diet and exercise: engage- ment/compliance: con- sumer satisfaction	
	Comparison		Two control groups (active control and wait list con- trol): Facebook group (active con- trol): educational content (handouts/ podcasts); access to polls and healthy activity or eating event invitations (e.g. on-campus farmer's and cycling event) messages via group post or messages via group post or messaging); goal setting for calorie intake based on weight, increase PA;wait list control	
	Intervention (duration; number of follow-ups; retention to final weight measurement)	information' gathered at baseline via Facebook and apps captured participant information (preferred mode of delivery; physical activity and diet goals; preferred frequency and time of tracking, feedback and participation in goal review; health coaches issued feedback on performance and progress towards goals)	2-month duration; 2 follow-ups (at 4 and 8 weeks); 96% retention. (Facebook) + daily text messages with prompt to (a) monitor their behaviours; feedback on performance of self-monitoring (e.g. whether or not they monitored each behaviour, diet only, PAonly, both, or neither);(b) report their behaviours (e.g. text back their daily calories, PA, weight) Participant returns a text response of these, and receives immediate feedback by text (e.g. actnowledgement of submithing self-monitoring data); or (c) address high-risk habits identified at baseline (e.g. late night snacking, liquid calories). Weekly tailored feedback summary reports based on self-monitoring data); or (c) address high-risk habits identified at baseline (e.g. late night anacking, liquid calories).	
	Population (number, partici- pant type, recruitment strate- gy) and BMI (M, SD)		52 college students, recruited in-person and online through websites; BMI: M = 31.36; SD = 5.3	
	Country		ASU	
pər	Article aim		To examine the feasibility, acceptability, and initial efficacy of a technology- based 8-week weight-loss intervention among col- lege students	
Table 1. Continued	First author and Year		Napolitano et al. (2013)	

Study design	ζŢ
Outcomes (primary and secondary, where stated in article)	Weight-loss percentage; sec- ondary: process measures (whether the user read the information completely, found the information helpful, easy to under- stand, and personally rel- evant, and whether they would recommend the programme to others)
Comparison	Access to freely accessible website with weight loss/ health information-only and a determination of whether participant was overweight. The pro- gramme included an over- view and sections related to the importance of weight and weight man- agement; definitions of a healthy weight; determi- nations of whether the participant is overweight; preparation for weight management; facts about weight-loss diets and pro- grammes; and weight management strategies. Through a menu, the user had the option of selecting any section for viewing and reviewing. They could also view other health topics on the site of their choosing, such as diabetes or asthma. Using this pro- gramme, participants were able to create their own educational experience. As in the TES condition, par- ticipants were allowed to return to the site through- out the course of the study
Intervention (duration; number of follow-ups; retention to final weight measurement)	6-week duration; 2 follow-ups (at 3 and 6 months); 20% retention. Register on study website and fill in baseline question-naire (same for both conditions); within 24 hrs, access to respective website (tailored or information-only). The tailored condition employed a tailored expert system (TES) that used 'Balance', a 6-week self-help weight management programme devised on an algorithm using baseline assessment data and links between data elements. The Balance programme creates individually tailored weight management plans, which focus on a healthy dist, better understanding of the relationship between food consumption and energy expenditure, calorie and fat consumption and energy expenditure, calorie and fat consumption, attributions for previous weight management efforts, body image, and social support. Authors provided an example of the tailoring processes: participants who reported greater ability to change diet than physical activity received more diteary advice; specifically cited barriers and lack of efficacy were addressed with messages tailored more distary advice; specifically cited barriers and here based materials consisted of an initial guide follow-up tailored action plans. Allow-up materials were designed to reinforce distary and physical activity improvements, adress specific barriers and physical activity improvements. Adress specific barriers and physical activity improvements, adress specific barriers and physical activity improvements, adress pacific barriers and provide support and self-monitoring resources. Participants were allowed to return to any of the materials throughout the course of the study of the materials throughout the course of the study of the materials throughout the course of the study of the materials throughout the course of the study of the materials throughout the course of the study of the materials throughout the course of the study of the materials throughout the course of the study of the materials throughout the course of the study of the materials
Population (number, partici- pant type, recruitment strate- gy) and BMI (M, SD)	2.862 Kaiser Permanente members, recruited by let- ters, newsletters and flyers; BMI: M = 32.05; SD = 3.85
Country	USA
Article aim	To assess the efficacy of a web-based tailored behavioural weight man- agement web-based information-only weight management materials
First author and Year	Rothert et al. (2006)

Note: CT = computer tailoring; HT = human tailoring; PA = physical activity; M = mean; SD = standard deviation; SMS = short message service (text); NR = not reported.

# Table 1. Continued

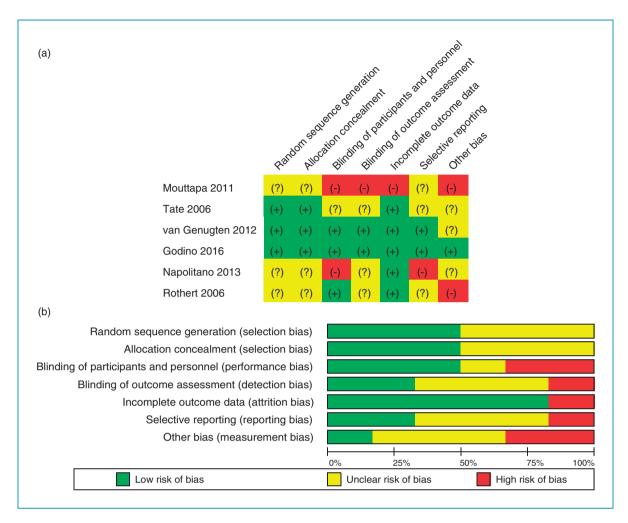


Figure 2. Risk of bias graph for quantitative articles (n = 6). (a) Summary of included articles using Cochrane's Risk of Bias tool. (b) Summary of risk of bias across studies.

compared computer tailoring to generic information control,<sup>24,32–34</sup> or to waitlist control;<sup>24,31</sup> one compared human tailoring (devised by human health counsellors) to generic information control;<sup>33</sup> one used a combination of computer tailoring and human tailoring<sup>30</sup> compared with a generic information control.

Computer tailoring, when feedback is delivered by a system, was implemented through a tailored website devised by Tailorbuilder<sup>34</sup> and a tailored expert system.<sup>32</sup>

## What was tailoring based on and how was it gathered? (Tailoring input)

Theoretical models of behaviour change are integral to tailoring and so any difference in theoretical components that provides the basis for the tailoring process may underlie any similarities or differences in intervention outcomes. Four out of six studies mentioned one or more theories that informed the development of the intervention: Social Cognitive Theory;<sup>31</sup> Cognitive Behavioural Theory;<sup>33</sup> Self-Regulation Theory, the Theory of Planned Behaviour, the Precaution Adoption Process Model;<sup>34</sup> and Behaviour Change Theory comprising Social Cognitive Theory, Control Theory, Operant Conditioning, Ecological Theory, and Social Network Theory.<sup>30</sup> Tailoring was based on anthropometric data and personal characteristics including age, height, weight, gender,<sup>31,32</sup> family history and prior weight-loss experience;<sup>32</sup> behaviours including dietary factors, e.g. calories, level of physical activity,<sup>24,30–32,34</sup> theoretical concepts such as future weight goals,<sup>31,32,34</sup> physical activity and diet goals,<sup>30</sup> advantages and disadvantages of weight management, confidence and willingness to change behaviours,<sup>34</sup> physical location,<sup>30</sup> whether self-monitoring occurred;<sup>24</sup> high-risk habits,<sup>24</sup> health-related habits, barriers to physical activity; as well as a range of psychosocial characteristics around stress management, attitudes to overweight people, comfort eating, and sources of motivation to lose weight, among others<sup>32</sup> (see Table 2 for further details).

Tailoring input, which provided the basis for the process of tailoring was gathered using online questionnaires at baseline,<sup>31,32</sup> web-based diaries,<sup>33</sup> via purposebuilt tailored website,<sup>33</sup> study-designed mobile applications,<sup>30</sup> SMS messaging,<sup>29</sup> Facebook, and emails from a remote health coach.<sup>30</sup>

Self-tailoring or 'customisation'<sup>15</sup> reflected personal preferences, and was based on desired features, e.g. whether they wanted a newsletter,<sup>31</sup> their desired intervention mode of delivery,<sup>31</sup> and how often users wanted feedback, tracking and goal reviews.<sup>30</sup>

### How was tailoring conducted? Computer tailoring

Dynamic tailoring (based on dietary or physical activity behaviours or weight, self-monitoring status or location) was conducted through daily and/or weekly input<sup>24,30,33</sup> or less regular input and based on theory, e.g. recommended every two weeks<sup>34</sup> or a selfdetermined pattern of tracking and feedback.<sup>30</sup> Static tailoring involved a process of distributing an online questionnaire at baseline.<sup>31,32</sup>

Two tailoring approaches were explicitly named and described: a tailored website<sup>34</sup> and a tailored expert system.<sup>32</sup>

One intervention delivered computer-tailored feedback and human tailoring via health coaches who also delivered feedback,<sup>30</sup> and whether or not participants received the intervention depended on the usage of these components. In the other instance, daily feedback appeared to be automated as messages were programmed and feedback was immediate, however it was not explicitly stated who or what devised the weekly personalised messages; an automated system appeared to have gathered daily behavioural data and this was provided to participants through a secure link delivered by Facebook message.<sup>24</sup> The computertailored feedback in Tate and colleagues' study<sup>33</sup> was based on the participants' responses and matched with a response from a bank of messages previously devised by experts; a series of decision rules ensured that automatically generated feedback was provided on the webpage.

As an example of an in-depth level of computer tailoring, Rothert et al.<sup>32</sup> described the action of the tailored expert system in their intervention:

participants reporting a family history of a particular disease received information regarding the connection of obesity to this class of diseases; participants who reported greater ability to change diet than physical activity received more dietary advice; specifically cited barriers and lack of efficacy were addressed with messages tailored to those issues; psychosocial stress was compared with reported coping abilities and accompanied with tailored stress management advice; and participants who reported that overweight individuals lacked willpower were given messages attempting to change this perception to a more controllable, external attribution (p. 268).

### Human tailoring

Tate et al.<sup>33</sup> drew on human tailoring by human counsellors who had behavioural weight-loss experience and degrees in nutrition, psychology, health education, and exercise physiology. Feedback was based on their clinical judgement in interpreting the participants' selfreported weekly weight, daily caloric intake, use of meal replacements and exercise, in their electronic, web-based diaries on the study website. The content of feedback for approaches involved comparativeprogress feedback for weight, comparative-progress feedback on calorie goals consumed and burned compared with individualised diet and exercise goals, behavioural strategies for improving adherence to self-monitoring diet and exercise, overcoming barriers, and motivation or praise depending on logging frequency. This detail was not provided in the other study.<sup>21,30</sup> Based on the description provided we determined that feedback was based on app usage, behaviours and progress to goals.

### How tailored interventions were delivered

A range of modes were used in the delivery of tailored interventions. Websites were used in all studies.<sup>24,30–34</sup> Two used SMS messages as well as Facebook.<sup>24,30</sup> Two studies used humans to tailor information: one used health counsellors who contacted participants via email<sup>33</sup> and one used health coaches who could instant message, telephone or video-call participants.<sup>30</sup>

### Type of Tailored Strategies and Output

Tailoring strategies were identified from the descriptions in the articles. Content-matching,  $^{31,32,34}$  evaluative feedback,  $^{24,30,31,33,34}$  descriptive feedback,  $^{24,34}$  and comparative-progress feedback,  $^{24,30,33,34}$  identification,  $^{34}$  contextualisation,  $^{30}$  as well as self-tailoring (customisation) $^{30,31,34}$  and personalisation  $^{34}$  were implemented by a range of eHealth systems. Location-tailoring was also used.  $^{30}$ 

Output typically consisted of feedback on weight loss, physical activity and diet (e.g. calories burned and consumed) compared with individualised diet and exercise goals.<sup>24,30,31,33,34</sup> Visual graphs of progress,

Dehational principles         Moder: how often         Taloning based on         Taloning based based date based based date based based date based bas	G
Dynamic tailoring (weekly) via Web-based diaries, including weekly Hur participants logging to a weight, daily caloric intake, use of web-based diary meal replacements and exercise Either computer-tailoring (HT), depending on intervention condition	MyPyramid guidelines used to recommend Questionnaire for opinions of the amounts for all food groups and how intervention (not given nutrients in each food group benefit health to control) ( <i>Content-matching</i> ): mode: website. Output 2: personalised newsletter based on their selection of long-term and short-term goals ( <i>content-matching</i> , self-tailoring/customisation), mode: (if opt in) emailed newsletter
	Human-tailoring group: weekly feedback - unstructured and based on clinical judge- ment (counsellors had behavioural weight- ilos sperience and degrees in nurrition, psychology health education, exercise ilos sperience and degrees in nurrition, psychology health education, exercise in psychology including answering questions, physiology) including answering questions, behavioural feedback on progress towards goals and weight loss (comparative-progress feedback; evaluative feedback based on behaviours from past week (SD= 9.2), p<0.001 on behaviours from past week (SD= 9.2), p<0.001 on behaviours from past week (SD= 9.2), p<0.001 on behaviours from past week (SD= 9.2), p<0.001 
Self-Regulation Theory:Dynamic computer-tailoringModule 1: Input 1:assessments basedModule 1:0Theory of Planned(asked to visit websiteon personal details and futurehistory aBehaviour: Precautionevery 2weeks)on personal details and futurehistory aBased on main steps for self-Tailored website made usingpersonal advantages and disad-lored feeBased on main steps for self-Tailored website made usingvantages of weight-gain prevention:weight instructioRegulation for weight con-trol: self-monitoringnored feeinstructiotrol: self-monitoringmodule softwareInput 3: assessment of confidenceinstructiotrol: self-monitoringprevention:input 4: assessment ofOutput 2: feeweight, diet and PA), goalsetting, action planning,evaluation.fee daotkevaluation.add willingness for weight-gainpros andweight, diet and PA), goalgoal-settingadvantageevaluation.add willingness for weight-feainpros andevaluation.planning,tool and preparation tool.feedback	Module 1:Output 1: graphs depicting weightNot explicitly defined.history and predicted weight (comparative- progress feedback: evaluative feedback); tai- lored feedback on participants' reasons for weight management and alternate views, instruction to consider both and reweighNot explicitly defined.Module 1:Output 2: feedbackTailored vs control: visited Module 1: 93.3% vs 81.5%; pros and cons.Not explicitly defined.Dutput 2: feedback dortModule 2: 74.1% vs 66.7%; pros and cons.Module 2: 74.1% vs 66.7%; module 2: 75.2% (no generic advantages (descriptive Module 4)

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Theory stated and How tailoring conducted: behavioural principles when; how often
Dynamic tailoring (daily/ weekly). Combination of computertailoring and human tailoring. Unclear specifically when and how due to flexible nature of intervention delivery: choice of six modes including Facebook, three study-designed mobile applications, text messag- ing, emails, a website with blog posts, and technolo- gy-mediated communica- tion with a health coach

Study author, year and title	Theory stated and behavioural principles	How tailoring conducted; when; how often	Tailoring based on	Tailored output: type and mode of delivery	Engagement
	healthy changes in physi- cal activity and diet; blog posts that were education- al; Patrick (2014) adds: social support and accountability (friends, participants, health coach via social network); formation of health veight-related behaviour; location based support and prompts; problem solving	(up to 10 brief (5-15 min) interactions)		Mode: text. apps, Facebook, website, instant messenger, telephone call, or video call	months. Control: NR
Napolitano et al., 2013 Using Facebook and text messaging to deliver a weight-loss pro- gramme to col- lege students	None stated. Goal setting (weight-loss goal set with study staff member); self-monitoring (given advice on how, given tools including scales, book and pedome- ter, measuring utensils), and social suport (via at content sent weekly via text messages); education- al content sent weekly via handout/ podcast, video demonstrations, one topic per week, topics of; self- monitoring and navigating campus; planning and nutrition; internal vs external hunger and eating triggers, physical activity; stress and distorted think- ing; social support; special occasions, dining out and holidays; relapse prevention	Dynamic tailoring (daily/ weekly) Unclear how devised - appears to be automated/computer-tai- lored feedback (via daily text and weekly summary report) ("messages were programmed at random intervals' p. 26)	Input 1: whether they monitored (general and behaviour specific, e.g. diat and PA) on that day. Input 2: daily PA, calories, and weight: Input 3: baseline reported high- risk habits	Output 1: feedback on performance of self- monitoring (e.g. whether or not they moni- tored aach behaviour, diet only, PA only, both, or neithen) ( <i>descriptive feedback</i> ); Output 2: immediate feedback by text (e.g. acknowledgement of submitting self-moni- toring data); ( <i>descriptive feedback</i> ); Output 3: advice on habits (e.g. late night snacking, liquid calories); Output 4: weekly tailored feedback summary reports (based on col- lated Input 2), which are compiled into personalised reports that summarised progress (included text and visual feedback on progress (included text and visual feedback on progress towards reaching one's behaviou- ral goals and progress in the skills training for the week ( <i>evaluative feedback</i> ). Mode: SMS message, external link to secure page accessed via Facebook	Level of engagement was exam- ined for the Facebook groups (intervention group and active control) by quantifying the number of times participants 'liked' a study-related post, posted a comment, and RSVP'd to an event. Active control: 4.17 participants 'liked' study-related posts (M = 1.25 likes each); 7/17 commented no study-related content at least once (M = 3.05 posts per commenter); 15/17 responded to event invite (average RSVPs, M = 6.54). Intervention group: 4/18 'liked' study-related posts (M = 1 like each); 14/18 posted on study- related content at least once (M = 1.3 posts per commenter); 13/18 responded to event invit- tations (M = 8.56 RSVPs of those who responded (); 79.7% of all general monitoring texts receiving a response
Rothert et al., 2006 Web-based weight man- agement programmes in an integrated health	None stated. Social support (optional par- ticipant-nominated 'buddy' who received email	Static computer-tailoring algorithm (tailored expert system): Balance a self- help weight management	Input: demographic information; per- sonal and family health history; former weight-loss experiences (including former use of specific	Output 1: an individually tailored weight man- agement plan in line with behavioural strategies, e.g.participants who reported greater ability to change diet than physical	Not explicitly defined. Process measures (% positive): tailored vs control Read information completely 82%

stuay autnor, year and title	behavioural principles	now tailoring conducted, when; how often	Tailoring based on	Tailored output: type and mode of delivery	Engagement
care setting: Arandomised, con- trolled trial	prompts to provide sup- port): educational content (initial guide); action plans (weeks 1, 3 and 6); positive reinforcement of dietary and physical activity improvements, address specific barriers, and pro- vide support and self- monitoring resources	programme developed by HealthMedia, Inc.) via baseline questionnaire	weight-loss treatments and out- comes from weight-loss attempts); general self-care activities (includ- ing tobacco, physical examinations, flossing, seat belt use, and stress management); physical activity, and berriters to being physically active perceived difficulty in or ontrolling diet and physical activity; worry regarding body image; barriers to weight management; psychosocial stress and coping; general dietary preferences (e.g. consumption of alcoholic beverages, desserts, fast food, high-fat dairy products, fried when stressed; weight-loss goals and motivation to lose weight; source of motivation (e.g. personal choice, pressure from others); a typological assessment of eating behaviour (e.g. whether the subject east in response to certain emo- tions, restricts food intake, then eats because of hunger, etc.); attitudes regarding overweight individuals (e.g. that they lack willpower, reducing risk of disease, hawing clothes fit better, reassuring others, getting people to stop naging them to lose weight, etc.); and perceived social suport	activity received more dietary advice: spe- cifically cited barriers and lack of efficacy were addressed with messages tailored to those issues ( <i>content-matching</i> ). Output 2: tailored action plans delivered at 1, 3 and 6 weeks into the programme to rein- force initial plan ( <i>content-matching</i> ). Mode: website	vs 67%; materials helpful 75% vs 57%; information easy to understand 93% vs 82%; materials were personally rel- evant 78% vs 61%; would rec- ommend programme to others 75% vs 59%, all ps =0.0001

Note: CT = computer tailoring; HT = human tailoring; PA = physical activity; M = mean; SD = standard deviation; SMS = short message service (text); NR = not reported.

Table 2. Continued

behavioural feedback, differential behaviour-change strategies, personalised certificates and personal reward plans were also delivered to participants.<sup>34</sup> Output also included a personalised nutrition plan and newsletter around their goals.<sup>31</sup> Feedback on behaviours such as daily self-monitoring also featured,<sup>24</sup> as did a weight management plan with behavioural strategies (e.g. action plans) matched to the individual.<sup>32</sup>

### Did tailoring work to increase engagement?

Four studies did not explicitly define engagement within the context of their intervention. As such, there was no consistent method of conceptualising and measuring engagement with eHealth interventions. Some studies reported usage metrics such as number of log-ins or use of eHealth features e.g. web-based behavioural diary,<sup>33</sup> registration to a website,<sup>31</sup> while others used process evaluations using subjective participant reports of amount of provided material read.<sup>32</sup> One study drew on both online module visits and subjective process measures.<sup>34</sup>

Engagement was explicitly defined in Godino et al.<sup>27</sup> and also in Napolitano et al.<sup>24</sup> Godino et al.<sup>30</sup> defined it as the sum of a participant's recorded interactions on the study Facebook page (e.g. a post, comment, or like) and mobile apps (e.g. entry of the number of steps taken per day), text messages sent and replied to, and communication with the health coach between each study measurement. Napolitano et al. defined engagement as the number of times participants 'liked' a study-related post, posted a comment, and RSVP'd to an event.<sup>24</sup> Tate and colleagues<sup>30</sup> examined log-in frequency and number and use of online diary submissions, counting the number of weeks where behavioural diaries were submitted.

Two articles conducted a process evaluation, capturing participants' subjective views of the intervention through self-report questionnaires.<sup>32,34</sup> Rothert et al.<sup>32</sup> reported that the tailored group responded more favourably than the information-only group for each item, including reading the intervention information completely; rating the materials as more helpful, more personally relevant, and easier to understand; and reporting that they would recommend programme to others, all ps = 0.0001. In research by van Genugten et al.<sup>34</sup> the tailored intervention group also perceived the material to be more personally relevant, and found content to be more novel than those in the generic information group, ps < 0.01, but contrastingly reported that they read less of the information, p < 0.001. They perceived the length of the information as less appropriate than those in the generic group, p = 0.01. There was no difference among groups in

their ratings of usefulness of the information, attractiveness of the design, appreciation of the tool, whether they would recommend it to others, and overall rating.

### Did tailoring support weight loss compared with control conditions?

A summary of the intervention groups compared with the controls is provided in Table 3.

Four out of six articles found positive effects for tailoring on weight loss when compared with generic or waitlist controls.<sup>24,30,32,33</sup> Two articles found no difference between the intervention and control (waitlist<sup>31</sup> and generic information<sup>34</sup>). Follow-up time points ranged widely from 1 month to 24 months.

The effect sizes denoting the mean difference in weight between intervention and the control groups ranged from -0.02 to -0.86 but in general are very small to small.<sup>25</sup> In the only article that compared tailoring types, there was no difference between human tailoring and computer tailoring at 3-month follow-up; both resulted in statistically significant weight loss compared with the control group. However, at 6 months, human tailoring was still effective in producing weight loss compared with the control, but this was no longer the case for computer tailoring.<sup>33</sup> Napolitano et al.<sup>24</sup> found the largest effect between the tailored vs waitlist control conditions, d = -0.86, followed by the difference between the tailored and generic information, d = -0.71.

Godino et al.<sup>30</sup> found a significant difference between groups at 6 month and 12 month follow-ups, with the tailored intervention resulting in 1.33 kg more weight loss than the general information control at both 6 and 12 months (p < 0.05), but there was no significant difference at 18 or 24 months, where ultimately the effect size at 24 months was -0.06. Rothert et al.<sup>32</sup> found a significant difference between groups at 3 and 6 months, (mean weight loss for the tailored groups = 2.6 and 2.8 kg, respectively, ps < 0.001) favouring the intervention group, which used a tailored expert system.

### Discussion

### Summary of key findings

A systematic review of tailored eHealth weight-loss interventions was conducted on six interventions, outlined in eight published articles, to describe how tailoring was implemented, and whether it was effective in producing weight-loss. Tailored interventions were found to be more effective in supporting weight loss than generic or waitlist controls in four of the six articles. Effect sizes were very small to moderate, with evidence of fluctuations in effect sizes and

	Baseline bod	Baseline body weight (kg)			Follow-up	Follow-up body weight (change in kg)	ght (change in kg)			
	Int		Control		(months unless	Int		Control		
	W	SD	Σ	SD	otherwise stated)	W	SD	W	SD	Effect size (d)
Mouttapa et al., 2011	71.7	17.43	74.91	21.42	6 weeks	↓0.59	2.12	↓0.29	5.02	-0.02
Tate et al., 2006	CT = 89; HT = 89	CT = 13.2; HT = 13	88.3	13.9	u m	3 months: $CT = \downarrow 5.3$ ; $HT = \downarrow 6.1$ 6 months: $CT = \downarrow 4.9$ ; $HT = \downarrow 7.3$	3 months: CT = 4.2; HT = 3.9 months: CT = 5.9; HT = 6.2	3 months: J2.8 6 months: J2.6	3 months: 3.5; 6 months: 5.7	3 months: CT vs C = -0.19**; HT vs C = -0.26**; (no dif CT vs HT) 6 months: CT vs C = -0.17; HT vs C =35*** (HT>CT)
van Genugten et al., 2012	83.39	11.49	83.34	10.61	1 (NR); 6	↓0.3 <b>4</b>	12.36	<b>0.6</b>	10.81	0.02
Godino et al., 2016	76.5	12.7^	76.5	13.2^	6, 12, 18, 24	6 months = $1.1$ ; 12 months = $1.1$ ; 18 months = $1.3$ ; 24 months = $0.3$ ;	R	6 months = $\uparrow$ 0.3; 12 months = $\uparrow$ 0.3; 18 months = $\uparrow$ 0.4; 24 months = $\uparrow$ 1.1	N	6 months: -0.11* 12 months: -0.1** 18 months: -0.05 24 months: -0.06
Napolitano et al., 2013	л Л	N	N N	N N	1; 2	1 month = $\downarrow$ 1.7 2 months = $\downarrow$ 2.4	1 month = 1.6; 2 months = 2.5	1 month: $AC = \downarrow 0.46;$ $C = \uparrow 0.28;$ 2 month: $AC = \downarrow 0.63;$ $C = \downarrow 0.24$	1 month: AC = 1.4; C = 1.7 2 months: AC = 2.4; C = 2.6	^^1 month: T vs AC = -0.5*; T vs C = -0.79***; 2 months: T vs AC = -0.71*; T vs C = -0.86*
Rothert et al., 2006	92.2	14.4	92.5	14.3	3; 6	3 months = $\downarrow$ 2.6; 6 months = $\downarrow$ 2.8	NR	3 months = $\downarrow$ 1.4; 6 months = $\downarrow$ 1.1	NR	3 months = -0.1***; 6 months = -0.12***

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differences of effect between tailored and non-tailored interventions, and between tailoring types, over time.

We note the diversity of approaches, input and technology used in tailored interventions; tailoring included a range of approaches including computer tailoring (using purpose-built tailored websites e.g. using TailorBuilder<sup>34</sup> and tailored expert systems<sup>32</sup>), human tailoring (by remote trained health counsellors<sup>33</sup>) and a combination of the two techniques.<sup>30</sup> Tailoring was based on a range of inputs, including anthropometric data (e.g. weight) and personal characteristics (e.g. age, prior weight-loss experience), behaviours (e.g. dietary intake, physical activity), goals (e.g. weight goal), theoretical determinants (e.g. confidence/willingness to change behaviours) psychosocial factors (e.g. stress management, social support) and location. A range of tailoring strategies was implemented, including evaluative- and comparative-progress feedback and contentmatching, as outlined by Harrington and Noar.<sup>16</sup>

Tailoring input was gathered using a variety of technological modes and included static processes of data collection from online questionnaires at baseline.<sup>31,32</sup> Dynamic processes also occurred where rolling input was gathered from web-based diaries,<sup>33</sup> via a website,<sup>34</sup> tailored study-designed purpose-built applications<sup>30</sup> and messaging.<sup>24</sup> mobile SMS Conceptualisation of engagement with tailored interventions was not uniformly defined. Our findings suggest that tailored materials were rated by participants as more personally relevant, more helpful, easier to understand, more likely to be recommended to others, and more novel than those in the generic information group.

The first goal of this review was to describe how tailoring was implemented in eHealth interventions. We have described a range of methods, inputs and technology used as part of tailored interventions. Previous reviews describing eHealth interventions have also found heterogeneity in approaches, so this was no surprise given the ubiquity of the Internet and the range of delivery options it affords.<sup>5,6</sup> With this advance in delivery methods comes new opportunities for tailoring health information, and we have found that tailoring is being increasingly used in innovative ways. However, if we are to build the science of tailoring it is imperative that the strategies reported in existing tailoring specification efforts<sup>15,16</sup> are used. In addition, the input or basis for tailoring should be specified as well as who/what devises tailoring output.

In this vein, our review has attempted to describe the range of tailoring approaches delivered by eHealth means. In Kreuter's<sup>35</sup> much cited original definition, tailoring is defined as the provision of individualised information, advice and support based on the participant's known characteristics, behaviours or scores on

relevant theoretical constructs. To aid the conceptualisation of tailoring in this context, we devised a model of tailoring depth, which we refer to as deep- and surface levels of tailoring (see Figure 3). This model is by no means exhaustive but we feel it is a step in the right direction in terms of differentiating between types of tailoring input to provide the basis for tailoring. It incorporates concepts discussed by Resnicow et al. in developing culturally sensitive public health interventions.<sup>36</sup> These concepts include surface structures, in other words relating to observable or behavioural characteristics of a population, and deep structures, relating to psychosocial, cultural and historical factors of a target population. This model indicates the distinctions made when the process of tailoring is based on behaviours or observable factors and when it is based on theoretical constructs, and allows for a combination of these. This builds upon Morrison's assertion that tailoring strategies can range from relatively simple e.g. employing the user's name, to the more complex, e.g. adapting content to personally relevant variables.<sup>14</sup> Deep-level tailoring refers to tailoring message content based on theoretical determinants of a specific health outcome, and the content that is delivered aims to modify these through strategies, for example like content-matching and evaluative feedback. In our review, input for deep-level tailoring was captured using a questionnaire at baseline (static tailoring).<sup>32</sup> For example, participants who reported a greater ability to change their diet at baseline compared with those who reported greater ability to change their physical activity received more dietary advice.<sup>32</sup> Deep-level tailoring was also conducted in a dynamic manner, gathering theory-based input via modules, using more than one assessment point, in order to ensure the intervention remained suitable to the needs of participants.<sup>34</sup> This is referred to as 're-tailoring' where new information from participants is obtained and feedback given at follow-up time points.<sup>37</sup> It was recently found that participants disengage from eHealth interventions because they believe they are no longer relevant to them and their needs.<sup>38</sup> As such, a more continuous process of adapting and delivering content to the individual may be a promising route.

Surface-level tailoring is a more concrete but simpler process, when tailoring is based on participants' (ongoing) self-reported behavioural data, e.g. amount of physical exercise or dietary intake, and evaluative or comparative-progress feedback is delivered on this basis with the aim of keeping participants on-track with their behavioural goals. We suggest that this type of tailoring is no less important, but is conducted on the basis of behaviours or observable factors, typically relying on more frequent input (e.g. daily or weekly). This process aligns with self-regulation

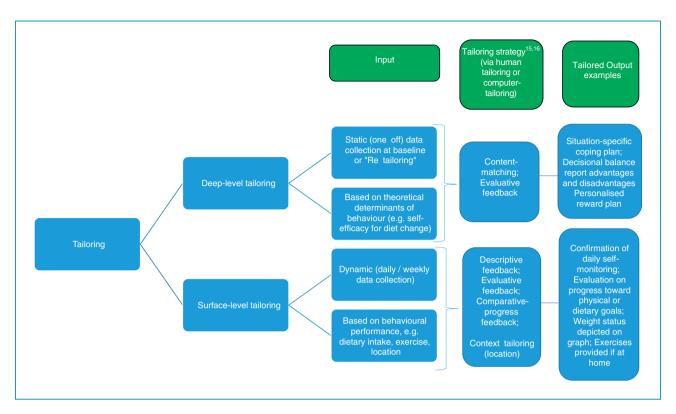


Figure 3. Model of tailoring depth.

theory in that it enables the participant to learn to selfmonitor and regulate their behaviour in line with their goals.<sup>39</sup> There may be a fundamental difference in the mechanisms at play behind this tailoring approach compared with deep-level tailoring, but this will need to be tested empirically in future research.

Tailoring may initiate processes of behaviour change, and it is possible that these processes in turn influence the efficacy of, and engagement with tailored interventions. Dynamic processes of tailoring, including movement between surface and deep level, or by tailoring type (e.g. human or computer), must be examined in relation to the efficacy of the intervention. For example, in the only study to examine human vs computer tailoring, both of these groups were effective at 3month follow-up, however at 6 months, the humantailoring group had significantly greater weight loss than the control.<sup>33</sup> This indicates the possibility that over time, the effects of tailoring will vary, and this variation may be related to an interaction on intervention-type by intervention-time. For example, at some intervention phases it may be more effective for a computer-tailored vs a human-tailored intervention to be implemented. This is linked to the suggestion that interactivity should be considered in developing effective eHealth interventions.<sup>9</sup> This also highlights the concept of engagement, an under-considered factor in the research design or assessment of factors influencing intervention efficacy. We note that engagement was not explicitly defined in the majority of included studies (N=4), and among those that did, it was conceptualised in a variety of ways mostly in relation to interaction with study-related content but also through process evaluations of the interventions.

This is in line with how engagement has been conceptualised in eHealth interventions – in terms of both participant experience and behaviours.<sup>40</sup> Future efforts to examine tailoring in eHealth interventions should measure engagement *both* in terms of participants' objective usage of intervention components (e.g. behaviouralthrough log-in metrics, usage data) *and* subjective experience (e.g. through process evaluations, participant interviews or surveys). This will enable an assessment of the extent to which tailoring (and indeed other intervention features and components) works the way it is theorised to (e.g. via mechanisms of increasing attention, increasing self-referential thinking<sup>15</sup>) and will provide a platform for future efforts to develop enhanced tailoring approaches addressing these mechanisms.

The second aim of the review was to establish whether tailored eHealth interventions result in weight loss compared with a generic control. Effect sizes were modest, comparable to other reviews in health behaviour change.<sup>10,17</sup> However, it has been shown that modest weight loss (3-5%) of pre-test body weight) is enough to have a beneficial health

impact, e.g. reducing blood pressure, HbA1c, triglycerides and LDL cholesterol<sup>41</sup> and so, even modest effects are desirable.

Four out of the six studies found beneficial effects for the tailored arm of eHealth weight-loss interventions. The two studies that found no effect were less intensively focused on weight loss, as both had a further aim in addition to weight loss, with Mouttapa et al. aiming to additionally provide a dietary intervention<sup>31</sup> and van Genugten et al. to also prevent weightgain.<sup>34</sup> While they met the criteria for the study, these findings suggest that the intensity of the intervention plays a key role, for example, in the study by van Genugten et al., even though the intervention was rigorously developed using behaviour change theory, the intervention could be finished within 90 minutes.<sup>34</sup> This may suggest that though tailoring was not found to be successful in that case, it may have been that the behavioural 'dose' of the intervention was not high enough. The other studies suggest that the effect of tailored eHealth interventions for weight loss is effective up until 1 year with no evidence at follow-up after this point.

### Limitations

The number of completed studies using tailored strategies and the Internet to support weight loss in our results belies the number of studies available for review. Despite significant effort, we were unable to secure data on weight for four studies, as weight was not reported in the published record or provided on request, and our results must be considered in this context.

Our review featured participants predominantly from the general population (82.5% of participants) and this suggests that effective tailored approaches can be delivered via the Internet in a scalable manner for weight loss. However, we did not include interventions addressing weight-loss maintenance or weightgain prevention as part of this review, which are also important foci for population-wide obesity strategies. We note that these are important areas deserving of research attention, however interventions that address these aims differ in intensity to those addressing weight loss and so we advise a delineation of behaviour type to explicate if and how tailoring is effective. The use of tailored eHealth tools in weight-gain prevention and weight-loss maintenance should be examined in future work as we attempt to support both those at risk of becoming overweight and those who need to maintain weight loss in a cost-effective, scalable and effective manner.

### Conclusions

eHealth interventions for weight loss are promising in terms of being scalable and effective,<sup>7</sup> but it is necessary to find strategies that help eHealth interventions channel the effectiveness of traditional in-person behavioural counselling approaches.<sup>2,7,42</sup> Tailoring is one approach that shows promise. We have found tailoring to have a small but beneficial effect on weight loss in four out of six studies in this review, yet tailoring approaches varied considerably even for one clearly defined health outcome (weight loss) and mode of delivery (Internet). To ameliorate this going forward and to build on previous research, we propose a model of tailoring depth, suggesting that there may be conceptual differences in tailoring approaches. Concepts of deep- vs surface-level tailoring could be tested experimentally to assess the mechanisms or effects of tailoring on the basis of input type, be it theoretical determinants, behavioural input or other tailoring input type. In order to develop the science of tailoring within the domain of eHealth interventions for behaviour change, it is vital to improve research designs (e.g. use tailoring vs generic control group) and methodologies (e.g. use the tailoring strategies and guidelines recommended in this field)<sup>16</sup> and in the wider domain of health behaviour change (e.g. by reporting behaviour-change techniques).<sup>43</sup>

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**Contributorship:** KR and SD conceived the study and devised and piloted the search strategy in consultation with CL. KR and SD conducted title and abstract screening and full text review. KR conducted data extraction and CL and SD checked this for accuracy. KR and SD completed the quality review. KR wrote the first draft of the manuscript. All authors reviewed and edited the manuscript and approved the final version of the manuscript.

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