The value of tailored communication in promoting medication intake behavior

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Chapter 2

Effects of eHealth interventions on medication intake behavior¹: a systematic review of the literature

Patient: “Do you have any advice? How I can monitor my health?” Nurse: “If you like, there is an application for your smartphone [...]. You can make notes, there is a map of nearby bathrooms and you can write down your symptoms” (male, 27 years old, Colitis).

¹ For the purpose of this dissertation, the term adherence is changed into medication intake behavior

Abstract

Background: Poor medication intake behavior is considered to be an important health risk. Numerous interventions to improve medication intake behavior have been developed. The use of Internet-based interventions to improve medication intake behavior has increased rapidly. Internet interventions have the potential advantage of tailoring the interventions to the needs and situation of the patient.

Objective: The main aim of this systematic review was to investigate which tailored Internet interventions are effective in improving medication intake behavior.

Methods: We undertook comprehensive literature searches in PubMed, PsycINFO, EMBASE, CINAHL, and Communication Abstracts, following the guidelines of the Cochrane Collaboration. The methodological quality of the randomized controlled trials and clinical controlled trials and methods for measuring medication intake behavior were independently reviewed by two researchers.

Results: A total of 13 studies met the inclusion criteria. All included Internet interventions clearly used moderately or highly sophisticated computer-tailored methods. Data synthesis revealed that there is evidence for the effectiveness of Internet interventions in improving medication intake behavior: 5 studies (3 high-quality (HQ) studies and 2 low-quality studies (LQ)) showed a significant effect; 6 other studies (4 HQ studies and 2 LQ studies) reported a moderate effect; and 2 studies (1 HQ study and 1 LQ study) showed no effect. Most studies used self-reported measurements to assess medication intake behavior, which is generally perceived as a low-quality measurement. We did not find a clear relationship between the quality of the studies or the level of sophistication of message tailoring and the effectiveness of the intervention. This might be explained by the great difference in study designs and the way of measuring medication intake behavior, which makes results difficult to compare. There was also large variation in the measured interval between baseline and follow-up measurements.

Conclusion: This review shows promising results on the effectiveness of Internet interventions in improving medication intake behavior. Although there is evidence according to the data synthesis, the results must be interpreted with caution due to low-quality medication intake behavior measurements. Future studies using high-quality measurements to assess medication intake behavior are recommended to establish more robust evidence for the effectiveness of eHealth interventions on medication intake behavior.
Introduction

Recent reports of the World Health Organization and the National Institute for Health and Clinical Excellence reveal that 30%–50% of patients with chronic illnesses do not take their medication as prescribed (Sabaté, 2003). Other studies also show that rates of poor medication intake behavior are very high and depend on the type of disease. The highest successful medication intake behavior rates are found for patients with human immunodeficiency virus infection, while diabetes patients have the lowest rate (DiMatteo, 2004). As such, poor medication intake behavior can be considered an important health care problem. This is especially true for patients with a chronic illness because successful medication intake behavior is a crucial factor in the effectiveness of a therapy (DiMatteo, 2004). Consequently, many patient-centered interventions are developed to improve medication intake behavior, and the impact of the Internet in the development of these interventions is increasing. It is therefore important to understand how these interventions work and to know whether they are effective in improving medication intake behavior. To our knowledge, no recent review has studied the effectiveness of patient-centered Internet interventions on patients’ medication intake behavior. Therefore, we conducted a systematic literature study in which we reviewed evidence from studies on Internet interventions that were developed to assist patients in their medication management. The purpose of our study was fourfold: first, to gain insight into the current stage of development of these interventions; second, to assess the included studies for their effectiveness on medication intake behavior; third, to investigate to what degree successful medication intake behavior is determined by the characteristics of the intervention; and fourth, to investigate whether there is a relationship between the characteristics of the study and the reported effectiveness of the interventions.

Different terms are used in the literature to describe the concept of successful medication intake behavior—for example, compliance, adherence, and persistence. They have all been used to indicate that the patient is using the medication following the prescribed regimen. These terms differ in exact meaning. In this paper, we use the term successful medication intake behavior. According to this definition, poor medication intake behavior is a wide concept that varies from missing an occasional dose to never taking the prescribed medications (Kane & Robinson, 2010). Patients have different reasons for being nonadherent. These different reasons have something in common: the patient does not execute the treatment plan and does not persist. Execution is a continuous process where the actual dosing history corresponds to the ideal doses (Urquhart & Vrijens, 2005; Wroe, 2002). To improve medication intake behavior and develop target interventions, it is important to address the specific reasons why a patient is not able or willing to execute the treatment plan. From this perspective, interventions should be personalized or tailored to address individual needs and beliefs. The definition
of tailoring describes the features that make tailored health messages different from other approaches: “It is assessment-based and as a result the message can be individual-focused” (Kreuter & Skinne, 2000). In other words, tailoring is based on gathering and assessing personal data related to health outcomes or several determinants in order to determine the most effective strategy to meet that person’s needs (Lustria et al., 2009). With these characteristics, a tailored message is able to provide personal feedback, commands greater attention, is processed more deeply, and is perceived as more likable by patients than a general message (Lustria et al., 2009; Noar, Benac, & Harris, 2007). Because of these possibilities, tailored health messages are also more likely than generic information to be read, remembered, and viewed as personally relevant (Kreuter & Skinne, 2000; Kreuter & Wray, 2003).

Computer technologies can be used to tailor health messages to the personal situation of the patient and might therefore contribute significantly to the development of tailored message strategies. The Internet is potentially a powerful medium for delivering those tailored messages. The management of a chronic disease should be personalized to an individual because the person is ultimately responsible for the success of the intervention (Wantland, Portillo, Holzemer, Slaughter, & McGhee, 2004). The technology provides an opportunity to tailor the information in several formats and modalities, which enhances the user’s experience of the material and will result in a better understanding (Lustria et al., 2009; Noar et al., 2007). Moreover, Internet interventions have the advantage that they can provide interactive and responsive programs (Wantland et al., 2004). These interventions can provide effective data and information provision and retrieval. The advantages of tailored message strategies can contribute to the incorporation of interactive and continued self-monitoring, feedback, and information exchange, which play an increasingly important role in changing patients’ behavior.

**Methods**

For this review, we used the guidelines of the Cochrane Collaboration to assess the studies on their internal validity and to summarize the existing evidence about Internet interventions to improve medication intake behavior in patients. The Cochrane Collaboration method is described in more detail in the Cochrane Handbook for Systematic Reviews of Interventions (Higgins, Green, & Collaboration, 2008).

**Inclusion Criteria**

We included a study when the following inclusion criteria were met: (1) the study described a patient-centered Internet intervention, (2) the study described an intervention for patients who use prescribed medication for a chronic condition, (3) at least one of the outcome measures was medication intake behavior, (4) the study was quantitative, and (5) the study was published in either the English or Dutch language.
Search Methods
We conducted a systematic literature search to identify articles containing information about the effect of Internet interventions on medication intake behavior. Comprehensive literature searches were undertaken in the databases PubMed, PsycINFO, EMBASE, CINAHL, and Communication Abstracts. The search strategies used the following keywords: (medication therapy management OR medication adherence OR patient compliance OR self-care) AND (Internet) AND (intervention study OR randomized controlled trial OR clinical controlled trial). We then continued with the snowball method by looking for references in publications, especially those of the included studies and reviews on interventions to promote medication intake behavior. The search was conducted in June 2010. Since Internet interventions is a relatively new topic, no time limits were applied. Application of the search strategy to the specified databases resulted in a total of 620 hits (Table 1). In total, we selected 13 studies from these results.

Table 1. Results of database searches.

<table>
<thead>
<tr>
<th>Source</th>
<th>Hits per strategy</th>
<th>Unique studies</th>
<th>Relevant studies</th>
</tr>
</thead>
<tbody>
<tr>
<td>PubMed</td>
<td>388</td>
<td>388</td>
<td>11</td>
</tr>
<tr>
<td>Communication Abstracts</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>PsycINFO</td>
<td>47</td>
<td>40</td>
<td>0</td>
</tr>
<tr>
<td>EMBASE</td>
<td>169</td>
<td>82</td>
<td>0</td>
</tr>
<tr>
<td>Snowball method</td>
<td>3</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>CINAHL</td>
<td>13</td>
<td>10</td>
<td>1</td>
</tr>
<tr>
<td>Total</td>
<td>620</td>
<td>520</td>
<td>13</td>
</tr>
</tbody>
</table>

Reference Manager version 11.0 (Thomson Reuters, New York, NY, USA) was used to manage the citations. Duplications were logged, leaving 520 unique results (see Figure 1). On the basis of title and abstract, two researchers (pairs of AL, MV, LvD, JvW) independently selected studies for inclusion. If the study seemed to meet the inclusion criteria or if there were doubts about the inclusion, the full text of the article was obtained. Based on the full articles two reviewers independently reviewed whether these studies fit all the inclusion criteria. Disagreements were solved by discussions between the two researchers. For a more detailed description of the excluded studies see Appendix 1.
Assessment of Methodological Quality

The methodological quality of included randomized controlled trials (RCTs) and clinical controlled trials (CCTs) was independently reviewed by two researchers (AL and JvW) using the list from the Cochrane Collaboration Back Review Group (Van Tulder, Furlan, Bombardier, & Bouter, 2003). The list consists of 11 criteria for internal validity, namely:

- 3 criteria regarding selection bias: whether (a) randomization was adequate, (b) treatment allocation was concealed, and (c) groups were similar at baseline regarding the most important indicators,
- 4 criteria for performance bias: whether (d) patients were blinded to the intervention, (e) care provider was blinded to the intervention, (g) co-interventions were avoided, and (h) compliance with the intervention was acceptable,
- 2 criteria regarding attrition bias: whether (i) the dropout rate after baseline was acceptable, and (k) the analysis included an intention-to-treat analysis, and,
- 2 criteria for detection bias: whether (f) the outcome assessor was blinded to the intervention, and (j) outcome assessments in all groups were similar.

For each included study, all criteria were scored as “yes,” “no,” or “unclear.” All unclear scores were later rated as “no.” Studies were rated as high quality (HQ) when at least 6 of the 11 criteria for internal validity were met. Otherwise, studies were considered of low
quality (LQ). Disagreements were discussed until consensus was reached. If disagreement or indistinctness persisted a third reviewer (LvD) was consulted.

In addition, two researchers (AL and LvD) independently assessed the quality of the methods for measuring medication intake behavior to a medical regimen. A standard method to assess medication intake behavior does not exist and every method has its limits (Sluijs et al., 2006; Wetzels, Nelemans, Schouten, Van Wijk, & Prins, 2006). In clinical trials, medication intake behavior can be measured based on, for example, interviews, diary, questionnaire-based self-reporting, prescription refills, pill counts, or electronic monitoring (Farmer, 1999; Wetzels et al., 2006). We categorized the measurements in high- and low-quality medication intake behavior assessments based on previous findings concerning the objectivity of these measurements (Farmer, 1999; Wetzels et al., 2006). In this review electronic monitoring and physiological/biomedical measures are defined as high-quality medication intake behavior assessment. These measurements are considered the most objective standard (Farmer, 1999). Previous research has shown that data from pill counts and electronic monitoring are strongly correlated (Velligan et al., 2007). Yet, others consider pill counts not to be accurate (Pullar, Kumar, Tindall, & Feely, 1989; Rudd et al., 1989). In addition, meta-analyses have shown that self-reported medication intake behavior is also strongly correlated with electronic monitoring (Shi, Liu, Fonseca et al., 2010; Shi, Liu, Koleva et al., 2010). Like pill counts, the accuracy of self-reported measurements is debatable. Some argue that self-reports may be an accurate measurement for measuring medication intake behavior (DiMatteo, Giordani, Lepper, & Croghan, 2002; Grymonpre, Didur, Montgomery, & Sitar, 1998), while others state that the use of self-reported measurements is not an accurate method (Farmer, 1999; McMahon et al., 2011; Sajatovic, Velligan, Weiden, Valenstein, & Ogedegbe, 2010; Wetzels et al., 2006). Taking all arguments into account, we considered self-measurements, such as questionnaires, pill counts, prescription refills, interviews, and diaries, to be most subjective for measuring medication intake behavior (Sluijs et al., 2006). We therefore considered these measurements low-quality medication intake behavior assessment. However, if two or more different low-quality medication intake behavior measurements were used in the same study, such as a combination of questionnaires and prescription refills, the method was considered high-quality medication intake behavior assessment.
**Data Extraction**

One researcher (AL) documented the following characteristics of the included studies: (1) method (type of study), (2) participants (total number of participants, sex per group, mean age per group, type of disease), (3) intervention (name of experimental intervention, name of control condition, period, number of times/minutes per week), (4) outcome measures (type of outcome measures, time of measurement), (5) results (short description), and (6) author’s conclusion.

**Data Synthesis**

Due to diversity in the features of the interventions and the methods used to measure medication intake behavior, it was not possible to pool the data. Therefore, we conducted a best evidence synthesis (see Textbox 1) based on (Van Tulder et al., 2003) and adapted by a Dutch study (Steultjens et al., 2009). The best-evidence synthesis was conducted by attributing various levels of evidence to the effectiveness of the interventions. The synthesis takes into account the design, methodological quality, and outcomes of the studies. Textbox 1 shows that at least 1 HQ RCT or 2 HQ CCTs were needed to establish robust evidence for the effectiveness of an intervention.

**Sensitivity Analysis**

We conducted a sensitivity analysis to identify how sensitive the results of the best-evidence synthesizes were to changes in the way the study quality was assessed. For the sensitivity analysis, the best-evidence synthesis was repeated in two different ways, using the following principles: (1) LQ studies were excluded, (2) studies were rated as HQ if they met at least 4 of the 11 criteria of internal validity instead of 6. We then compared the results of the sensitivity analysis with the results of the best-evidence synthesis and described the sensitivity of the results (Steultjens et al., 2009; Verkaik, Van Weert, & Francke, 2005).

**Effectiveness**

Study effectiveness was categorized as significant effect on medication intake behavior, moderate effect on medication intake behavior, and no effect on medication intake behavior. We defined a study effect as moderate if the authors reported a positive effect of the intervention on medication intake behavior but there were limitations, such as the following: improvement of medication intake behavior was found only in a subgroup of the intervention group; medication intake behavior was measured indirectly (e.g., the study drew conclusions about the use of beta-agonist indicating that medication intake behavior was improved); or the significance of the results in medication intake behavior was not tested, but the authors used convincing arguments to explain the effectiveness of the intervention (see results section for explanation per study).
Textbox 1. Principles of Best Evidence Synthesis

**Evidence:**
Provided by consistent, statistically significant findings in outcome measures in at least two high quality RCTs.

**Moderate evidence:**
Provided by consistent, statistically significant findings in outcome measures in at least one high quality RCT and at least one low quality RCT or high quality CCT

**Limited evidence:**
Provided by statistically significant findings in outcome measures in at least one high quality RCT

Or
Provided by consistent, statistically significant findings in outcome measures in at least two high quality CCTs (in the absence of high quality RCTs)

**Indicative findings:**
Provided by statistically significant findings in outcome measures in at least one high quality CCT or low quality RCT (in the absence of high quality RCTs)

**No/Insufficient evidence:**
If the number of studies that have significant findings is less than 50% of the total number of studies found within the same category of methodological quality and study design

Or
In case the results of eligible studies do not meet the criteria for one of the above stated levels of evidence

Or
In case of conflicting (statistically significantly positive and statistically significantly negative) results among RCTs and CCTs

Or
In case of no eligible studies

**Intervention: Tailoring Level of Sophistication of the Website**
Tailored Internet interventions differ in how they deliver their message (Lustria et al., 2009). The difference is based on the sophistication of the way the message is tailored. We categorized the interventions in being low, moderate, or high in sophistication. Some interventions involve a form of online assessments (low sophistication), and others use online assessments, tailored feedback, and content matching (moderate sophistication). The third group of interventions provides instant feedback and a complex tailored health program with several tools and activities that would enable patients to achieve their health goals (high sophistication; see Figure 2; Lustria et al., 2009).

![Figure 2. Level of Sophistication (Lustria et al., 2009).](image-url)
Results

The main characteristics of the included studies (Artinian et al., 2003; Chan, Callahan, Sheets, Moreno, & Malone, 2003; Chan et al., 2007; Cherry, Moffatt, Rodriguez, & Dryden, 2002; DeVito Dabbs et al., 2009; Dew et al., 2004; Dilorio et al., 2009; Guendelman, Meade, Benson, Chen, & Samuels, 2002; Jan et al., 2007; Joseph et al., 2007; Ross, Moore, Earnest, Wittevrongel, & Lin, 2004; Van der Meer et al., 2009; Van der Meer et al., 2010) are presented in Table 2 and further described below (for a more detailed description of the included studies see Appendix 2).

Methodological quality: assessment of internal validity

For this review 10 RCTs were included, and 9 of them were assessed on their internal validity (Chan et al., 2003; Chan et al., 2007; DeVito Dabbs et al., 2009; Guendelman et al., 2002; Jan et al., 2007; Joseph et al., 2007; Ross et al., 2004; Van der Meer et al., 2009; Van der Meer et al., 2010; See Table 3). We included 1 RCT with no data on our primary outcome variable (i.e., medication intake behavior) for the control group (Artinian et al., 2003). This means that for this tenth study, we could not assess validity criteria. Moreover, we reviewed 2 prospective cohort designs and 1 survey. A total of 7 RCTs met 6 or more of the 11 validity criteria and therefore qualified as HQ studies.

Intervention: tailoring level of sophistication of the website

All Internet interventions reported computer-tailoring methods. Interventions were categorized as having low sophistication (online assessments), moderate sophistication (online assessments, tailored feedback, and content matching), and high sophistication (a more complex tailored health program; see Figure 2; Lustria et al., 2009).

Online assessment and feedback

Online assessment and feedback are used in interventions with single-incident computer-assisted risk or health assessments. For example, feedback is emailed to the patients or provided online (Lustria et al., 2009). In addition, these interventions are brief and usually done once at the beginning of the intervention. None of the reviewed studies used online assessment and feedback.

Tailored content

With tailored content a program provides (1) tailored text messages composed in a unique way according to how patients respond to certain questions, or (2) restricted access to content sections per patient (Lustria et al., 2009). Tailored content was used by 9 of the studies we reviewed; 1 study (Jan et al., 2007), Blue Angel for Asthma Kids, conducted an Internet-based interactive asthma educational and monitoring program in which patients were able to complete an electronic diary, record symptoms and need for rescue
medication, and upload their videos when they were using their inhaler. Based on these outcomes, the program comprised both an action plan with a warning system and a written treatment plan. A similar customized educational and monitoring website for patients with asthma was conducted in 2 studies (Chan et al., 2003; Chan et al., 2007), and 1 study (Joseph et al., 2007) tested the asthma management program Puff City. The program used tailoring to alter behavior through individualized health messages based on the patients’ beliefs, attitudes, and personal barriers to change or maintain the behavior. Another form of a tailored website, System Providing Access to Records Online (SPPARO), was examined in 1 study (Ross et al., 2004). This website provided the medical record, an educational guide, and a message system for patients. Moreover, patients could contact the health provider by email. The telemedicine diabetes disease management program Health Buddy was tested in 2 studies (Cherry et al., 2002; Guendelman et al., 2002). However, the modality used in the diabetes program was different from the previously mentioned monitoring programs. Patients answered personalized questions that enabled them to monitor their disease symptoms, medication intake behavior, and disease knowledge by pressing buttons for response. The 2 studies using the Health Buddy differed in the intensity of the feedback. One study (Artinian et al., 2003) tested a medication compliance device. Data and answers to questions were recorded by the device and uploaded daily to a central server. Based on these answers health providers were able to monitor the patients, provide advice, and update the treatment regimens in the Med-eMonitor devices. One study (DeVito Dabbs et al., 2009) also tested a handheld device, Pocket Personal Assistant for Tracking Health (PATH), developed for patients after lung transplantation to record health data, review data trends, and report their condition changes to the transplant team. The device included decision-support programs to promote self-care behaviors.
<table>
<thead>
<tr>
<th>Study; method</th>
<th>Intervention*</th>
<th>Participants; sex; mean age</th>
<th>Medication intake behavior measurement; timing of measuring</th>
<th>Main conclusion</th>
</tr>
</thead>
<tbody>
<tr>
<td>(Artinian et al., 2003); RCT</td>
<td>Web-based monitoring system; tailored content; nature of expert/therapist contact</td>
<td>N = 18 (17 males; mean age 68 years); intervention group n = 9, control group n = 9</td>
<td>Pill counts; baseline and 3 months</td>
<td>Medication intake behavior rate was 94% for the monitor group as measured by the monitor system</td>
</tr>
<tr>
<td>(Jan et al., 2007); RCT</td>
<td>Blue Angel for Asthma Kids variability; tailored content; nature of expert/therapist contact</td>
<td>N = 164, intervention group n = 88 (35 males; mean age 10.9 years); control group n = 76 (28 males; mean age 9.9 years)</td>
<td>Self-reported at baseline and 12 weeks</td>
<td>The Blue Angel for Asthma Kids has the potential for improving asthma outcome compared with conventional treatment over a period of 12 weeks</td>
</tr>
<tr>
<td>(Chan et al., 2007); RCT</td>
<td>Customized educational and monitoring Web site; tailored content; nature of expert/therapist contact</td>
<td>N = 120; intervention group n = 60 (37 males; mean age 10.2 years); control group n = 60 (38 males; mean age 9.0 years)</td>
<td>Computerized prescription refill record at baseline, 26 weeks, and 52 weeks</td>
<td>No difference in medication intake behavior between groups</td>
</tr>
<tr>
<td>(Chan et al., 2003); RCT</td>
<td>Customized educational and monitoring Web site; tailored content; nature of expert/therapist contact</td>
<td>N = 10; intervention group n = 5 (1 male; mean age 6.6 years); control group n = 5 (4 males; mean age 8.7 years)</td>
<td>Self-reported asthma diary and computerized prescription refill record at 90 days and 180 days</td>
<td>After the intervention, the use of beta-agonist decreased, which is an indication of better medication intake behavior</td>
</tr>
<tr>
<td>(Joseph et al., 2007); RCT</td>
<td>Web-based asthma management program; tailored content; user control</td>
<td>N = 314 (36.6% male; mean age 15.3 years); intervention group n = 162; control group n = 52</td>
<td>Self-reported at baseline and 12 months</td>
<td>Positive changes in controller medication intake behavior were seen</td>
</tr>
<tr>
<td>(Ross et al., 2004); RCT</td>
<td>SPPARO (System Providing Access to Records Online); tailored content; nature of expert/therapist contact</td>
<td>N = 104; intervention group n = 54 (80% male; mean age 57 years); control group n = 50 (74% male; mean age 55 years)</td>
<td>Self-reported at baseline, 6 months, and 12 months</td>
<td>Providing patients access to an online medical record improved medication intake behavior</td>
</tr>
<tr>
<td>(Cherry et al., 2002); prospective design</td>
<td>Telemedicine diabetes disease management program; tailored content; nature of expert/therapist contact</td>
<td>Intervention group n = 169 (39 males; mean age 53 years); historical group (usual care)</td>
<td>Self-reported</td>
<td>Outcomes offer encouraging evidence that telemedicine technology coupled with daily remote monitoring may improve appropriate use medication</td>
</tr>
<tr>
<td>(Guendelman et al., 2002); RCT</td>
<td>Health Buddy, an interactive device connected to a home telephone; tailored content; nature of expert/therapist contact</td>
<td>N = 134; intervention group (40 males; mean age 12.2 years); control group (37 males; mean age 12.0 years)</td>
<td>Self-reported at baselines, 6 weeks, and 12 weeks</td>
<td>Patients were more likely to take their asthma medication without additional reminders</td>
</tr>
<tr>
<td>(DeVito Dabbs et al., 2009); RCT&lt;sup&gt;b&lt;/sup&gt;</td>
<td>Pocket Personal Assistant for Tracking Health (PATH); tailored content; nature of expert/therapist contact</td>
<td>N = 30; intervention group n = 15 (60% male; mean age 55 years); control group n = 15 (60% male; mean age 57 years)</td>
<td>Self-reported at baseline and 2 months</td>
<td>Patients who received the PATH were more likely to show high medication intake behavior to the medical regimen</td>
</tr>
<tr>
<td>(Van der Meer et al., 2009); RCT&lt;sup&gt;b&lt;/sup&gt;</td>
<td>Internet-based self-management program; customized health program; user control</td>
<td>N = 200; intervention group n = 101 (29% male; mean age 36 years); control group n = 99 (29% male; mean age 37 years)</td>
<td>Self-reported at baseline, 3 months, and 6 months</td>
<td>After 3 months asthma control improved</td>
</tr>
<tr>
<td>(Van der Meer et al., 2010); RCT&lt;sup&gt;b&lt;/sup&gt;</td>
<td>Internet-based self-management program; customized health program; user control</td>
<td>N = 200; intervention group n = 111 (28 males; mean age 36 years); control group n = 89 (28 males; mean age 36.6 years)</td>
<td>Self-reported at baseline, 3 months, and 1 year</td>
<td>Weekly self-monitoring leads to improved medication intake behavior in patients with partly and uncontrolled asthma at baseline and tailors asthma medication to individual patients’ needs</td>
</tr>
<tr>
<td>(Dilorio et al., 2009); Survey</td>
<td>WebEase; customized health program; user control</td>
<td>N = 35 (40% male; mean age 37.5 years)</td>
<td>Self-reported at baseline and 6 weeks</td>
<td>Participants showed some improvement in medication intake behavior following the program</td>
</tr>
<tr>
<td>(Dew et al., 2004); prospective design</td>
<td>Website including skills workshops, discussion group, ask an expert, question and answer, health tips, recourses, and references; customized health program; nature of expert/therapist contact</td>
<td>N = 64; intervention group n = 24 (18 males; mean age 45.8 years); control group n = 40 (30 males; mean age 57.5 years)</td>
<td>Self-reported at baseline and 4 months</td>
<td>The intervention appeared to be weakly associated with improved medication intake behavior</td>
</tr>
</tbody>
</table>

<sup>a</sup> Sophistication of tailoring classification based on Figure 2.

<sup>b</sup> Randomized controlled trial.
Table 3.  
Results of methodological quality

<table>
<thead>
<tr>
<th>Study</th>
<th>Validity criteria(^a) met</th>
<th>Study quality(^b)</th>
<th>Quality measurement medication intake behavior</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Randomized clinical trials</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(Artinian et al., 2003)</td>
<td>Not applicable(^c)</td>
<td>Low</td>
<td>Low</td>
</tr>
<tr>
<td>(Jan et al., 2007)</td>
<td>a, b, c, d, i, j</td>
<td>High</td>
<td>Low</td>
</tr>
<tr>
<td>(Chan et al., 2007)</td>
<td>a, b, f, i, j</td>
<td>Low</td>
<td>Low</td>
</tr>
<tr>
<td>(Chan et al., 2003)</td>
<td>a, b, c, f, h, i, j</td>
<td>High</td>
<td>High</td>
</tr>
<tr>
<td>(Joseph et al., 2007)</td>
<td>a, b, c, d, e, h, i, j</td>
<td>High</td>
<td>Low</td>
</tr>
<tr>
<td>(Ross et al., 2004)</td>
<td>a, b, c, d, e, h, i, j, k</td>
<td>High</td>
<td>Low</td>
</tr>
<tr>
<td>(Guendelman et al., 2002)</td>
<td>a, b, d, i, j</td>
<td>Low</td>
<td>Low</td>
</tr>
<tr>
<td>(DeVito Dabbs et al., 2009)</td>
<td>a, b, c, e, i, j</td>
<td>High</td>
<td>Low</td>
</tr>
<tr>
<td>(Van der Meer et al., 2009)</td>
<td>a, b, c, i, j, k</td>
<td>High</td>
<td>Low</td>
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<tr>
<td>(Van der Meer et al., 2010)</td>
<td>a, c, h, i, j, k</td>
<td>High</td>
<td>Low</td>
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<tr>
<td><strong>Prospective design/clinical trial or cohort design</strong></td>
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</tr>
<tr>
<td>(Cherry et al., 2002)</td>
<td>Low</td>
<td>Low</td>
<td></td>
</tr>
<tr>
<td>(Dew et al., 2004)</td>
<td>Low</td>
<td>Low</td>
<td></td>
</tr>
<tr>
<td><strong>Survey</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(Dilorio et al., 2009)</td>
<td>Low</td>
<td>Low</td>
<td></td>
</tr>
</tbody>
</table>

\(^a\) a: randomization adequate; b: treatment allocation concealed; c: groups similar at baseline regarding most important indicators; d: patients blinded to intervention; e: care provider blinded to intervention; f: outcome assessor blinded to intervention; g: co-interventions avoided; h: compliance with intervention acceptable; i: dropout rate after baseline acceptable; j: outcome assessed similarly in all groups; k: intention-to-treat analysis included.

\(^b\) That is, 6 of 11 validity criteria were met.

\(^c\) Compliance was acceptable in the first interval (<90 days).

\(^d\) Compliance was acceptable in the first interval (6 months).

\(^e\) No data on medication medication intake behavior for the control group and therefore judged as low quality.

\(^f\) Compliance was acceptable in the first interval (3 months).

**Customized health programs**

Interventions that provide not only tailored content but also individualized instructions for meeting certain health goals, self-management goals, or goal-setting activities are so-called customized health programs (Lustria et al., 2009), used by 4 of the included studies. Of these, 2 studies (Van der Meer et al., 2009; Van der Meer et al., 2010) tested the effects of an Internet-based self-management program for asthma patients. This website allowed monitoring through the website, text messages, use of an Internet-based treatment plan, online education, and the possibility to communicate with the health provider. The intervention WebEase (Dilorio et al., 2009) consisted of three modules that were designed to assess an individual’s status related to self-management practices and create a plan for change or to maintain the behavior. The modules in WebEase required the patient to answer questions related to these topics. Feedback was provided based on these responses. Patients entered data into MyLog, which is a screen for recording data about medication-taking behavior, stress, etc. In addition, the intervention included a knowledge component and a discussion board. This means that each patient was directed...
to another path (DiLorio et al., 2009). Another study (Dew et al., 2004) tested a customized health program where patients chose which components of the website they wanted to use. The website included a home page, post-transplant skills workshops, discussion groups, “ask an expert,” question-and-answer possibility, healthy-living tips, resources, and reference library. The way the patients used the website was based more on voluntary participation than in the study that used MyLog (DiLorio et al., 2009). Table 4 and Table 5 show for each study which method for delivering the tailored message was used (see column 2).

Table 4.
Effectiveness of short-term interventions (<6 months)

<table>
<thead>
<tr>
<th>Study</th>
<th>Study quality</th>
<th>Sophistication of tailoring</th>
<th>Quality measurement medication intake behavior</th>
<th>Short-term effectiveness (&lt;6 months)*</th>
</tr>
</thead>
<tbody>
<tr>
<td>(DeVito Dabbs et al., 2009)</td>
<td>High</td>
<td>Moderate</td>
<td>Low</td>
<td>++</td>
</tr>
<tr>
<td>(Jan et al., 2007)</td>
<td>High</td>
<td>Moderate</td>
<td>Low</td>
<td>++</td>
</tr>
<tr>
<td>(Dew et al., 2004)</td>
<td>Low</td>
<td>High</td>
<td>Low</td>
<td>–</td>
</tr>
<tr>
<td>(DiLorio et al., 2009)</td>
<td>Low</td>
<td>High</td>
<td>Low</td>
<td>++</td>
</tr>
<tr>
<td>(Artinian et al., 2003)</td>
<td>Low</td>
<td>Moderate</td>
<td>Low</td>
<td>+</td>
</tr>
<tr>
<td>(Guendelman et al., 2002)</td>
<td>Low</td>
<td>Moderate</td>
<td>Low</td>
<td>++</td>
</tr>
</tbody>
</table>

* ++ = significant effect on medication intake behavior; + = moderate effect on medication intake behavior; – = no effect on medication intake behavior.

Role of health providers
Interventions also differ in the type and extent of health provider involvement. User control allows individuals to take a major role in managing their own care, whereas in expert control an expert or therapist takes a more directive role (Lustria et al., 2009). Only 4 of the interventions were based on user control and 9 interventions used contact with the health provider. A web-based asthma management program was developed in 1 study (Joseph et al., 2007). The program used tailoring to alter behavior through individualized health messages based on the user’s beliefs, attitudes, and personal barriers to change. The health provider did not interfere. Three studies were also user based with treatment algorithms to give feedback (DiLorio et al., 2009; Van der Meer et al., 2009; Van der Meer et al., 2010). In contrast, in the Blue Angel for Asthma Kids, a customized educational and monitoring website site providing secure email contact between patients and their therapist, the therapist had a more directive role (Jan et al., 2007). Like the Blue Angel for Asthma Kids, SPPARO included a messaging system that made it possible to exchange secure messages with the health provider (Ross et al., 2004). The intervention manager who reviewed the data, sent emails about the peak flow, inhaler technique, and symptoms, and forwarded them the website. Patients (the virtual group as well as the
office-based group) had access to their case manager 24 hours a day, 7 days a week. One study (Dew et al., 2004) conducted a website including skills workshops, discussion group, ask an expert, question and answer, health tips, recourses, and references. In this case, the role of health providers was to provide the possibility for a patient to ask an expert. Several interventions used a special device to exchange information between patients and health care providers. The Med-eMonitor recorded data and answers to questions that could be uploaded by the health provider. Based on these outcomes, the health provider provided advice and updated the treatment regimens (Artinian et al., 2003). The PATH (DeVito Dabbs et al., 2009) recorded data and provided a tailored decision-support program and email contact with the health provider. The Health Buddy device in 2 studies (Cherry et al., 2002; Guendelman et al., 2002) is like the Med-eMonitor and PATH based on the feedback and participation of the health provider, but there is a difference in the intensity of participation of the health provider. In 1 study (Cherry et al., 2002) the health provider contacted the patient only when necessary, while in the other 2 studies (DeVito Dabbs et al., 2009; Guendelman et al., 2002) the patient received feedback instantly after sending a question. Our analysis to examine the extent to which medication intake behavior is determined by different tailoring levels revealed no clear relationship between the intervention’s level of sophisticated tailoring and the extent to which the intervention was effective (Table 4).

Summary of effects on successful medication intake behavior

We found 5 studies with a significant effect on medication intake behavior (3 HQ studies and 2 LQ studies; DeVito Dabbs et al., 2009; Dilorio et al., 2009; Guendelman et al., 2002; Jan et al., 2007; Joseph et al., 2007). The first study (Jan et al., 2007) concluded that the intervention had a positive and significant effect on use of the inhaled corticosteroid and that this effect significantly differed from the baseline. In addition, the second study (Joseph et al., 2007) found positive changes in controller medication intake behavior. The third study (Dilorio et al., 2009) tested the WebEase intervention and found a significant
effect on successful medication intake behavior. The fourth study (Guendelman et al., 2002) found that patients were more likely to take their asthma medication when they used the Health Buddy. In the fifth study (DeVito Dabbs et al., 2009), patients who received the PATH intervention were more likely to adhere to their medical regimen. A moderate effect on successful medication intake behavior was reported in 6 studies (4 HQ studies and 2 LQ studies). The SPPARO proved to be feasible and improved general medication intake behavior. Improved medication intake behavior showed a similar trend but these results did not reach significance (Ross et al., 2004). The second study (Chan et al., 2003) concluded that, after the intervention, the use of beta-agonist decreased, which is an indication of improved medication intake behavior. The third study (Cherry et al., 2002) reported that medication intake behavior improved from 65% at pre-test to 94% at post-test, but the difference was not statistically tested. The medication intake behavior rate in the fourth study (Artinian et al., 2003) was 94% for the monitor group as measured by the monitor system. However, because there was no pre-test and the data of the control group were not available, it is unknown whether the results were significant as compared with the pre-test or the control group. The intervention of the last 2 studies (Van der Meer et al., 2009; Van der Meer et al., 2010) improved medication intake behavior in patients with partly and uncontrolled asthma at baseline. The authors concluded that the intervention was most effective in improving medication intake behavior for patients with partly or uncontrolled asthma at baseline. No significant results on patients’ successful medication intake behavior were found in 2 studies (1 HQ and 1 LQ) (Chan et al., 2007; Dew et al., 2004).

Assessment of medication intake behavior measurements

Regarding the measurement of medication intake behavior, the 13 studies we reviewed showed a large variability of methods: 12 studies used a low-quality measurement to assess medication intake behavior and 1 used a combination of these methods (i.e., a high-quality measurement to assess medication intake behavior; Table 3).

Low quality of medication intake behavior measurement

In 10 studies, self-reported scales were used to obtain the medication intake behavior rate. Although 5 studies (Guendelman et al., 2002; Jan et al., 2007; Joseph et al., 2007; Van der Meer et al., 2009; Van der Meer et al., 2010) used self-reports to measure medication intake behavior, they did not describe what kind of instrument they used. One study (Dew et al., 2004) used self-reported data by asking questions regarding medication intake behavior during the initial interview. Reports of therapist and patients were compared. In addition, 1 study (Cherry et al., 2002) used a self-developed medication intake behavior survey on the Health Buddy appliance. The other 3 studies chose existing, valid, self-reported medication intake behavior scales. One study (Ross et al., 2004) used a
combination of the Morisky scale and the General Adherence Scale from the Medical Outcomes Study, and one study (DeVito Dabbs et al., 2009) used the Health Habits Assessment, a self-reported scale to measure medication intake behavior. One study (Dilorio et al., 2009) used the self-report USCF Adherence Questionnaire and the Antiretroviral General Adherence Scale. Finally, two studies used measurements such as counting pills (Artinian et al., 2003), and one study (Chan et al., 2007) used a computerized prescription refill record (after evaluation of the pilot study in which completing the diary turned out to be time consuming and inconvenient (see Chan et al., 2003).

**High quality of medication intake behavior measurement**
One study used a combination of methods. This study (Chan et al., 2003) used a diary in combination with a computerized prescription refill record. Table 3 shows the results of the assessment of the internal validity and the quality of medication intake behavior measurement.

**Relation between quality of medication intake behavior measurement and effectiveness**
Our investigation of the relationship between the quality of the medication intake behavior measurement and the effectiveness of the interventions revealed no clear relationship (there was only 1 study using a high-quality method to assess medication intake behavior), although self-reported medication intake behavior measurements seemed to result more often in significant effects than did pill counts and pharmacist medication intake behavior measurements (Table 4 and Table 5). Of the 10 studies using self-reports (low-quality medication intake behavior measurement), 5 reported a significant effect of the intervention on medication intake behavior (DeVito Dabbs et al., 2009; Dilorio et al., 2009; Guendelman et al., 2002; Jan et al., 2007; Joseph et al., 2007), 4 a moderate effect (Cherry et al., 2002; Ross et al., 2004; Van der Meer et al., 2009; Van der Meer et al., 2010), and 1 no effect (Dew et al., 2004). From the 2 studies in which pharmacist data or pill counting was used, 1 reported a moderate effect (Artinian et al., 2003) and 1 no effect (Chan et al., 2007). The 1 study that used a combination of methods to measure (Chan et al., 2003) found a moderate effect on medication intake behavior.

**Relation between interval of medication intake behavior measurement and effectiveness**
There was no clear relationship between the timing of the medication intake behavior measurements and the effectiveness of the intervention. The intervals between baseline and follow-up measurements differed between projects. Short-term medication intake behavior (i.e., within 6 months) was measured in 6 studies. The first study (Dilorio et al., 2009) showed that WebEase improved medication intake behavior 6 weeks after baseline. Patients who used PATH were more likely to show better medication intake behavior than the control group after an interval of 8 weeks after baseline (DeVito Dabbs et al., 2009).
The third study (Jan et al., 2007) found a significant effect on successful medication intake behavior after 12 weeks, and the fourth study (Guendelman et al., 2002) reported an improvement in medication intake behavior after 12 weeks. The fifth study (Artinian et al., 2003) found a successful medication intake behavior rate of 94% in the experimental group after 12 weeks. Because of the lack of medication intake behavior data for the control group and the lack of a pretest, the effects on medication intake behavior could not be established. The sixth study (Dew et al., 2004) examined the proportion of nonadherent patients in both an intervention and a control group after 16 weeks. That study’s authors found uniformly small and nonsignificant differences between the control and intervention groups. However, they found an important difference within the intervention group. Subgroup differences appeared when the intensity of using parts of the intervention was related to the effectiveness. For example, patients who used the Managing Medical Regimen Workshop more often or intensely appeared to improve their medication intake behavior than those using the intervention less often or intensely (Dew et al., 2004).

Long-term medication intake behavior was measured in 7 studies—that is, medication intake behavior with an interval of 6 months or longer, mostly of 1 year or more. Two studies (Chan et al., 2003; Van der Meer et al., 2009) reported a moderate effect in their pilot on medication intake behavior after 6 months. In 2 studies (Joseph et al., 2007; Ross et al., 2004), they found a moderate (Ross et al., 2004) and significant (Joseph et al., 2007) improvement in medication intake behavior after 1 year. Two studies (Cherry et al., 2002; Van der Meer et al., 2010) found a moderate effect on medication intake behavior after 1 year. This means that all of the included studies using an interval of 6 months or longer showed an effect (significant or moderate) on long-term successful medication intake behavior. One study (Chan et al., 2007) did not find an effect on successful medication intake behavior after 1 year.

Table 4 and Table 5 give an overview of the methodological quality of the studies, the level of sophistication of each intervention, the quality of measurement of medication intake behavior, and an overview of the short-term and long-term effects.

Data synthesis
Using the principles of the best-evidence synthesis (see Textbox 1), taking into account the design, methodological quality, and outcomes of the studies, the following conclusions can be drawn. In total, 7 studies were considered HQ. We found 3 HQ studies (DeVito Dabbs et al., 2009; Jan et al., 2007; Joseph et al., 2007) and 2 LQ studies (Dilorio et al., 2009; Guendelman et al., 2002) that had a significant effect on improved medication intake behavior and that met 6 of 11 criteria. This means that there is evidence that tailored Internet interventions are successful in improving medication intake behavior.
**Sensitivity analysis**

The sensitivity analysis showed the same results as the best-evidence synthesis. The results remained the same when the analysis was repeated with the 6 LQ studies excluded (i.e., taking only the 7 HQ studies into account). Moreover, when studies were rated to be HQ if 4 instead of 6 criteria of interval validity were met, results stayed the same.

**Discussion**

**Principal results**

First, our objective was to gain insight into the current state of the use of Internet interventions to improve medication intake behavior. Results of this review indicate that this is still a new field. This is visible in the differences in interventions with respect to crucial aspects such as the level of sophisticated tailoring and the role of health care providers. Despite the differences, it is remarkable that none of the interventions used a low level of tailoring and the majority (nine of thirteen) provided the opportunity to contact a health provider.

Second, the studies were assessed on their effectiveness on medication intake behavior. There is evidence that Internet interventions can improve medication intake behavior. This evidence comes from three HQ studies and two LQ studies, finding significant results on medication intake behavior.

Third, we wanted to investigate to what degree medication intake behavior is determined by the characteristics of the intervention. All interventions discussed in this review used tailored methods and used a moderately or highly sophisticated tailored intervention. These types of health programs, especially customized health programs, are more complex, generally long-term, allowing the patients to access the programs several times (Lustria et al., 2009), and are considered appropriate for difficult-to-influence behaviors. We did not find a clear relationship between how sophisticated the tailoring of the intervention was and the extent to which the intervention appeared to be effective, possibly due to the various methods that were used.

Last, we wanted to investigate whether there is a relationship between the characteristics of a study and the reported effectiveness of the interventions. We found that there was variation not only in the level of tailoring, but also in the measurement of medication intake behavior, the timing of measuring medication intake behavior, and the intensity of the intervention. The included studies used self-reporting measurements (i.e., interviews, diary, self-reporting via questionnaires) or pill counts, or prescription refills, or a combination. No study used electronic monitoring, which is perceived as a high-quality method for assessing medication intake behavior (Farmer, 1999). Of the thirteen studies we reviewed, seven measured long-term medication intake behavior, using an interval of six months or longer. There is no clear evidence that the duration of the intervention is related to the effectiveness of the intervention. Nevertheless, of the seven studies
measuring long-term medication intake behavior, one HQ study showed positive effects and four HQ studies and one LQ study showed moderate effects on successful medication intake behavior. This indicates that long-term interventions are promising. However, more research in this field is needed.

There is evidence that Internet interventions can be effective in improving medication intake behavior. The evidence comes from three HQ studies. However, the results should be interpreted with caution. Self-reported scales were used in 10 studies, which is considered a low-quality medication intake behavior measurement: five reported a significant effect of the intervention on successful medication intake behavior, four a moderate effect, and one no effect. Self-measurements can contribute to overestimating of the effects of interventions (Nieuwkerk & Oort, 2005). This could be explained by the possibility that patients may forget that they missed a dose. Biases that appear most prominent in estimating medication intake behavior by the patient from structured questionnaires are social desirability and social approval (Sluijs et al., 2006; Urquhart & Vrijens, 2005). In other words, studies relying on self-reporting may have a tendency to err on the optimistic side when it comes to medication intake behavior, certainly compared with more objective pill-counting studies. However, it must be noted that anonymous self-report questionnaires are found to be significantly correlated with electronic monitoring (Hugen et al., 2002) and virologic response (Nieuwkerk & Oort, 2005), considered more objective methods. Research also shows that using specific strategies, such as ensuring patients that their responses will be kept confidential (Nieuwkerk & Oort, 2005) or stratifying patients according to their socially desirable response (Nieuwkerk, De Boer-Van der Kolk, Prins, Locadia, & Sprangers, 2010), improves the prediction of medication intake behavior by self-reports. This indicates that self-reports are not useless, but future research should examine more strategies to reinforce accurate reporting by patients (Nieuwkerk & Oort, 2005). On the other hand, we included RCTs and, consequently, self-reported medication intake behavior can be expected to be overestimated in both treatment arms. Thus, the intervention effect (i.e., difference between intervention and control group) was not necessarily overestimated. Additionally, a distinction can be made between valid self-report measurements and measurements that are not. If self-reported measurements are used, for instance because this is a cost-effective method, using validated measurements is recommended.

Electronic monitoring or observation is considered to be more accurate. It electronically records the time and date of the actual dosing events. Because every single method has its limitations, the best approach is to use multiple assessment techniques concurrently, as a way to improve the accuracy of medication intake behavior assessment (Wetzels et al., 2006). One study in our review (Chan et al., 2003) used self-reported diary and prescription refills to measure medication intake behavior and made a distinction in interpreting the results. They used a self-reported diary to assess how the patients used
their inhaler, based on the idea that when patients are not using their inhaler according to the health providers’ advice, they can be considered nonadherent. In addition, they used prescription refills to measure how many refills the patient was obtaining. In line with this method of measuring medication intake behavior, the optimal approach could be suggested to be a combination of self-reports and more objective measurements (Drotar, 2000). In addition, every single measurement needs a different interpretive approach because it has different relationships to clinical outcomes (Sluijs et al., 2006).

Of the studies we reviewed, six measured short-term medication intake behavior that varied from six weeks to four months: four of them were found highly effective and one moderately effective regarding medication intake behavior. The question is whether medication intake behavior improved in the long term, because the period was too short to measure persistence. According to an international expert forum on medication intake behavior (Sluijs et al., 2006), it is not easy to identify adherent and nonadherent patients beforehand. There is a large body of evidence dominated by reports identifying factors that are predictive or associated with nonadherence. Medication intake behavior could be seen as a dynamic behavior that is determined or influenced by unrelated factors that fluctuate and change over time (Reynolds, 2004). As an adherent patient can become nonadherent over time, the importance of time (i.e., persistence) has been emphasized (Urquhart & Vrijens, 2005). The quality of execution of the treatment plan can influence persistence. Factors such as perceptions of treatment outcomes, beneficial effects, and adverse effects can influence the quality of execution over time. Therefore, conducting interventions that address long-term medication intake behavior and overcome reasons why a patient is not able or not willing to adhere are recommended.

**Methodological limitations**

The search method was top-down in that we relied on existing databases and search terms. This approach has the possibility of missing important articles due to miscoding of search terms. A bottom-up strategy is more time consuming but has the advantage of being more comprehensive.

**Clinical implications**

Monitoring medication intake behavior to optimize effects and minimize nonadherence could be time consuming. Computers, however, are very good at collecting data concerning the monitoring of medication intake behavior. Internet interventions can be tailored, collect data, and monitor medication intake behavior. In addition, based on this systematic literature review, there is evidence that tailored Internet interventions can be an effective method to improve medication intake behavior. This means that Web-based interventions can be effective at increasing medication intake behavior among chronically ill patients. Health providers, who want to enhance patients’ medication intake behavior,
are encouraged to use tailored websites or reminder systems. They could use these interventions in addition to their everyday work.

Implications for research
Because we did not find a clear relationship between the effectiveness and the degree of tailoring, we recommend that future studies should be conducted with variation in the level of sophistication of tailoring to further test which characteristics of the tailored messages have the most positive effects on successful medication intake behavior. Moreover, website compliance is often not completely reported. While the frequency in which the patient used the website is often reported, studies do not describe how exactly patients used the website. Therefore, it is difficult to compare the results of different interventions, because the way patients use the website can have implications for the effectiveness of the website.

Conclusion
With more than 40 million people using the Internet for a variety of purposes, health communication programs in the future are more likely to be delivered online (Kreuter & Wray, 2003). These types of interventions especially have the potential to address difficult-to-change behaviors such as successful medication intake behavior. This review shows promising results on the effectiveness of tailored Internet interventions to enhance successful medication intake behavior of chronically ill patients. There is evidence that these interventions can enhance successful medication intake behavior. But it remains a relatively new field, and studies using more objective measurements to assess medication intake behavior are recommended.