Smart Stress Care: Usability, Feasibility and Preliminary Efficacy of Fully Automated Stress Management Application for Employees

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Objective We developed a fully automated smartphone-based stress management application and explored its usability, potential feasibility, and preliminary efficacy for stress management in Korean employees.

Methods Healthy employees working in large public enterprises were enrolled. Participants used our automated stress management application for four weeks. With the application, they monitored their stress level and life style factors. Personalized stress management techniques, including psychoeducation and cognitive behavioral technique, were also provided based on their stress level and lifestyle pattern. In 2014, additional relaxation techniques were incorporated. Participants' mental health status and lifestyle pattern were self-assessed at baseline and at 4 weeks after using the application.

Results A total of 68 subjects were recruited. The application generally received high satisfaction rating. After the intervention, perceived stress level was significantly decreased, both in 2013 and 2014 (BEPSI-K score pre. vs. post. 14.27 vs. 11.00, F=12.49, p=0.001 in 2013; 12.05 vs. 10.00, F=17.18, p<0.001). In 2014, depression symptom severity was also significantly decreased (CES-D score pre- vs. Post-, 17.66 vs. 11.95, F=9.76, p=0.004). The effects were more significant in females and in those <35 years.

Conclusion Our fully automated stress management application is acceptable and usable, showing preliminary efficacy for reducing employees' stress levels.

Key Words Smartphone-based, Application, Stress management intervention, Employees.

INTRODUCTION

Stress of employees has been known as an important risk factor for major mental health problems. It is associated not only with the development of various mental health problems, but also with unhealthy lifestyle behaviors, eventually leading to negative consequences, including absenteeism, loss of productivity, and disability.1,2 For these reasons, early stress management has attracted much attention as the primary focus of the prevention of mental health problems.3

Although there various stress management interventions, most of them have been underutilized, resulting in many affected employees left untreated.4 This may partly be due to limited accessibility,3,5 barriers in time or treatment cost, or even fear of potential stigmatization attached to mental health problems. These factors work as major barriers in implementing stress management interventions in Korea.4 According to a study conducted by the Korean society of occupational stress with approximately 7,000 workers from 254 businesses across the country, 67.73% of the subjects were exposed to potential occupational stress and 22% of the subjects were suffering from severe stress. However, psychiatric services were severely underutilized, likely due to fear of stigmatization and lack of knowledge about mental health services.7

Technological advances in the internet and smartphones may allow affected employees to have easier access to stress management interventions. In particular, a smartphone-based intervention may maximize its effects on reducing mental stress by overcoming barriers like time or place—providing flexibility to meet the individual's needs—as well as stigmatization.
by providing anonymity upon request. Considering cultural obstacles, i.e., social stigma, related to using mental health services and high penetration rate of smartphones (94% of all population) in Korea, a smartphone-based stress management intervention can be a promising tool.

Hence, we developed a fully automated stress management application that is accessible 24 hours a day and can be utilized at the individual's own pace and convenience. In this study, we aimed to evaluate its usability, potential feasibility, and preliminary efficacy in stress management of employees.

METHODS

Development process of the fully automated smartphone application

We developed a smartphone application entitled “Stress Guide” for employees and a web-based system for mental health professionals in collaboration with KT, a Korean wireless communication company. KT assisted in the technical aspect of application development. The application was developed for iOS. Prior to development, we randomly distributed a questionnaire to employees in three public enterprises to explore what types of interventions were primarily desired by employees. The application was developed based on their feedback. The application was initially composed of four parts; 1) mental health and lifestyle assessment, 2) daily lifestyle management, and 3) forum for sharing their sources of trouble and getting psychological support among participants, including a frequently asked questions section, and 4) a tool for searching nearby mental health professionals.

After 4 weeks of intervention, we collected participants’ feedback on their satisfaction using a 5-point Likert scale and asked open-ended question about additional features they wanted to see in the next iteration. After finishing the study in 2013, most participants wanted additional practical features that can help relaxation. Following their feedback, we also included 5) a relaxation training part in 2014.

Recruitment

Employees were recruited through our advertisement posted at Seoul National University Bundang Hospital (SNUBH) health promotion center (HPC), or by phone and/or e-mail correspondence to employees of three large public enterprises. SNUBH HPC provides standardized annual medical examinations, including lifestyle risk factor evaluations and laboratory tests, as well as national cancer screening services for employees. Any employees from one of these three enterprises or any employed individual who visited HPC for a health checkup were eligible to participate in this study. Another eligibility criterion for participation was smartphone literacy. After explaining the purpose and procedures, those who agreed to participate were included. Informed consent was received from all participants. This study was approved by the Institutional Review Board of SNUBH (B-1206/158-112).

The contents of the app (Figure 1)

Mental health and lifestyle assessment part

At baseline, participants evaluated their mental health status, occupational stress level, and lifestyle factors, including smoking, alcohol drinking, exercise, eating habits, and sleep. Once they filled out all the measures, test results were presented with a web diagram and detailed explanation about their mental health status. When clicking on each section of the diagram, results were automatically provided with personalized recommendations from mental health professionals based on participants’ test results (Figure 2). Recommendations were mainly composed of psychoeducation and cognitive-behavioral techniques to modify their maladaptive thoughts and behaviors.

Daily mood with life style monitoring

Participants were also able to voluntarily monitor their daily lifestyle pattern, including includes alcohol consumption, smoking, coffee intake, sleep, exercise, and dietary patterns.
with simple mood evaluation on a daily, weekly, and monthly basis. When they answer 10 brief questions regarding lifestyle management on the app, they are provided with a short, customized comment with recommendations to improve daily habits, which is automatically generated based on the score from their daily condition. Participants can set push alarms to be notified at certain times of the day for self-evaluation.

Forum and frequently asked question part
In the forum section, participants were able to share their problems with other users anonymously. Additionally, participants had free access to the frequently asked questions (FAQs) section, which addressed topics relating to common stressors of employees, including marital problem, interpersonal relationship, career development, coworker relationship, child rearing, in-law issues, emotional problems, and general self-management. This section was answered by employee assistance program professionals with more than three years of experience in stress management programs.

A tool for searching nearby mental health professionals
In case participants wanted an offline mental health service, our application provided a search tool to help find nearby mental health professionals. The search results provided a list of professionals with their addresses, phone numbers, and direct links to their websites.

Relaxation training part
Following the participants’ feedback in 2013, we additionally included the relaxation training part in 2014. All 38 participants who took part in this study during 2014 were able to use this feature. Abdominal breathing, progressive muscular relaxation, and meditation were included. A voice guide was provided to each relaxation method. Participants can choose the gender of voice guide and background music. Logs were left on the calendar for participants to monitor their application activity.

Implementation
After obtaining informed consents, an iPhone with the preinstalled stress guide application was provided to participants who did not already possess an iPhone during the 4-week study period. For those who already had an iPhone, instruction on installing the application was provided. A baseline evaluation was self-assessed using measures included in the mental health and lifestyle assessment part of the application. Participants were able to use the fully automated application during the study period without limit at no cost. Prior to participation, participants were instructed to use the app at least twice or three times a week. No direct human contact was made to encourage or monitor their usage. Instead, a brief text message—one during the study period—was sent to encourage usage of the application. After four weeks, a final self-assessment was done, with the same measures used for the baseline evaluation, and the rented iPhones were collected at this time. All self-assessed data were automatically sent and stored in the internet system for analysis and were available for download as excel files.

After 4 weeks’ intervention, we collected participants’ feedback on satisfaction using a 5-point Likert scale and asked open-ended question regarding additional feature that users might want for the next iteration. Following the participants’ feedback in 2013, we also included the relaxation training part in 2014.
Assessment measures
Changes in depressive symptoms, perceived stress levels, and lifestyles were monitored at baseline evaluation and the final evaluation. Among them, changes in individual's perceived stress level and depressive symptoms were used as primary outcome measures. Changes in occupational stress level, resilience, and lifestyle were also additionally evaluated.

Primary symptom outcomes
Changes in individual's perceived stress level and depressive mood were used as the primary outcome measures. Individual's perceived stress level was evaluated using the Korean version of the brief encounter psychosocial instrument (BEPSI-K). The brief encounter psychosocial instrument (BEPSI) evaluates the subjectively perceived stress levels using five questions with a 5-point Likert scale and is a widely used measure of perceived stress with high reliability and validity. Likewise, the Korean version, BEPSI-K, also showed high reliability and validity.

Changes in depressive mood were evaluated using the center for epidemiologic studies depression scale (CES-D). The CES-D is a 20-item questionnaire with a 5-point Likert scale used for screening depression in the general population. The Korean translated version of CES-D also showed high reliability and validity.

Additional measures
The social readjustment rating scale (SRRS) was applied to measure the current stress factors that can work as a confounding factor in evaluating psychological distress. The SRRS assesses the experiences of 43 major stressful events during the past year. Each event has a different weight depending on its severity, and the total score generates the intensity of recent stressors, which can work as a confounding factor in evaluating psychological distress.

Type A personality—generally known to be vulnerable to stress—was evaluated using Framingham type A behavior pattern (FTA). The FTA contains a 20-item questionnaire with an 11-point Likert scale.

The Korean occupational stress scale (KOSS) brief version was additionally used to evaluate stress levels related to occupational settings. KOSS uses 24 questions with a 4-point Likert scale. The KOSS brief version is composed of seven subscales used to evaluate the main causes of occupational stress, including job demand, insufficient job control, interpersonal conflict, job insecurity, occupational system, lack of reward, and organizational climate.

Individual's resilience was evaluated using the Korean resilience quotient-53 (KRQ-53). The KRQ-53 has 53 questions with a 5-point Likert scale, which evaluates various components of resilience, including controllability, positivity, and sociability.

In evaluating exercise and diet patterns, the international physical activity questionnaire (IPAQ) and the mini dietary assessment (MDA) were used. IPAQ has 6 questions regarding the duration and intensity of exercise in the past one week. A higher score indicates high activeness. The IPAQ scores were classified into three categories: 1) the ‘active’ category, which was characterized with intense exercise at least three times a week with a total IPAQ score of ≥1500 or exercise every day with a total IPAQ score of ≥3000; 2) the ‘minimally active’ defined by 20 minutes of intense exercise at least three times a week or 30 minutes of light exercise every day, or a total IPAQ score of ≥600; and 3) the ‘non-active’ category, which includes other cases not previously defined. MDA evaluates whether the individual maintains a regular and balanced diet with 6 questions; a higher score means healthier diet.

Smoking and alcohol consumption patterns were explored using the Fagerström test for nicotine dependence (FTND) and the Korean version of the alcohol use disorders identification test (AUDIT), respectively. FTND contains six items regarding nicotine craving. The sum of the individual item scores is associated with biological nicotine dependence. AUDIT is a screening questionnaire that uses 10 sub-items to identify individuals with hazardous and harmful patterns of alcohol consumption. In both questionnaires, higher scores indicate greater dependence.

Usability assessment
At the final visit, we collected participants’ feedback on satisfaction using a 5-point Likert scale and asked open-ended questions about additional features that they want included in the next iteration. Their satisfaction about each content, speed, and design of the application was measured with a 5-point Likert scale. We also asked their willingness to pay, and how much, for continual usage of the application.

Statistical analysis
Descriptive data for the sample were presented as the mean with standard deviation (SD) for continuous variables and counts with proportions for categorical variables. Symptom outcome measures were compared between the baseline and final evaluation using a generalized linear mixed effects model analysis, which is one of methods for analyzing repeated measures. It can build models for continuous or categorical outcomes while accounting for correlations between the outcomes from the same subject. To adjust the effects of recent stressful events an individual may have experienced, SRRS, recruitment year, gender, and age (<35 years or not (median)) were applied as covariates. These were chosen by Akaike in-
formation criterion (AIC)\textsuperscript{21} and Bayesian information criterion (BIC)\textsuperscript{22} as measures of model (variable) selection. Moreover, in order to determine the effects of gender and age, the same analyses were conducted after dividing the groups by gender and age. Thirty-five years was the median age, which was also associated with the duration of employment. Twenty-seven (79.4\%) subjects aged \( \geq 35 \) years were employed for more than 10 years, while only 7 (20.6\%) subjects aged <35 years were employed for more than 10 years. Thus, we created subgroups depending on age; a group younger than 35 years and a group older than 35 years. To control for family-wise error rates, Bonferroni correction was applied for multiple testing.

Statistical analysis was performed by using R version 3.2.1 (http://www.r-project.org), and the level of significance was set at \( p<0.05 \).

**RESULTS**

A total of 68 participants took part in this study (Table 1). All participants attended the final visit; no one dropped out. Based on self-reported questionnaire, the frequency of application usage was reported as 2.7 times per week and 23.9 minutes on average per use.

**Outcome measures** (Table 2)

After four weeks of intervention, the BEPSI-K score was significantly reduced in both 2013 and 2014 (\( F=12.49, p=0.001 \) in 2013; \( F=17.18, p<0.001 \) in 2014). The CES-D score (\( F=9.756, p=0.004 \)) was additionally decreased in 2014. No significant differences were observed between the participants recruited in 2013 and those in 2014 with respect to changed patterns of measures.

We conducted a power analysis for the primary outcome

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**Table 1. Demographic data of participants**

| Measurements | Participants in 2013 (N=30) | Participants in 2014 (N=38) |
|--------------|----------------------------|-----------------------------|
|              | N (%) or mean±SD           | N (%) or mean±SD            |
| Gender       |                            |                             |
| Male         | 19 (63.33)                 | 9 (23.68)                   |
| Female       | 11 (36.67)                 | 29 (76.32)                  |
| Age          | 34.80±7.96                 | 36.45±10.64                 |
| Duration of employment (year) | 34.80±7.96 | 36.45±10.64 |
| <3           | 12 (40.00)                 | 13 (34.21)                  |
| 3–5          | 2 (6.67)                   | 7 (18.42)                   |
| 5–10         | 6 (20.00)                  | 4 (10.53)                   |
| \( \geq 10 \) | 10 (33.33)                 | 14 (36.84)                  |

SD: standard deviation
measures. The power of BEPSI-K in 2013 and 2014 was 94.23% and 98.54% at \( \alpha = 0.05 \), respectively. Additionally, the CES-D showed a power of 88.77% at \( \alpha = 0.05 \) after correcting for multiple testing; only the BEPSI-K in 2014 showed satisfactory power at \( \alpha = 0.05/9 \).

We conducted additional analyses to explore the effect of age and gender (Table 3). Considering that age was strongly associated with duration of employment, we stratified participants by age (<35 vs. ≥35 years).

When participants were stratified by gender, males showed a significant decrease, only in the BEPSI-K score, while females showed a significant decrease in the BEPSI-K, FTA, CES-D, and AUDIT scores.

When participants were stratified by age, those <35 years showed a significant decrease in the scores of BEPSI-K, CES-D, and IPAQ, and an increase in the MDA score. Those ≥35 years showed a significant decrease in the BEPSI-K and KOSS scores.

We conducted additional analyses to explore the effect of application usage duration (Table 4). The median duration of usage per week was 40 minutes, so we divided users into two groups based on their usage duration using the median duration as cut-off point (average use frequency per week * average spent time per use). When participants were stratified by application usage duration, participants who used the app more than 40 minutes per week showed a significant improvement in the scores of BEPSI-K, CES-D, KOSS, KRQ-53, and AUDIT, while those who used the app less than 40 minutes showed a significant improvement only in the BEPSI-K and MDA scores.

### Usability of the application

Participants generally reported high satisfaction with re-

### Table 3. Generalized linear mixed effects model analyses by gender and age subgroup

| Measures                                               | Males (N=28) |          |          |          |          |          |          |
|--------------------------------------------------------|--------------|----------|----------|----------|----------|----------|----------|
|                                                        | F            | p        | F        | p        | F        | p        | F        |
| Brief encounter psychosocial instrument (BEPSI-K)      | 18.185       | <0.001   | 11.189   | 0.002    |          |          |          |
| Center for Epidemiologic Studies Depression Scale (CES-D) | 3.835        | 0.058    |          |          | 6.835    | 0.014    |          |
| Korean occupational stress scale (KOSS)                | 0.004        | 0.949    |          |          | 7.082    | 0.012    |          |
| Framingham A type behavior pattern (FTA)               | 0.099        | 0.755    |          |          | 3.341    | 0.078    |          |
| Korean resilience quotient -53 (KRQ-53)                | 0.582        | 0.451    |          |          | 5.714    | 0.023    |          |
| International physical activity questionnaire (IPAQ)    | 0.785        | 0.379    |          |          | 2.832    | 0.098    |          |
| Mini dietary assessment (MDA)                          | 5.349        | 0.027    |          |          | 0.876    | 0.357    |          |
| Alcohol Use Disorders Identification Test (AUDIT)      | 0.118        | 0.733    |          |          | 4.255    | 0.048    |          |
| Fagerström Test for Nicotine Dependence (FTND)         | 1.086        | 0.305    |          |          | 3.008    | 0.093    |          |

The SRRS score and recruited year were applied as covariates. SRRS: social readjustment rating scale

### Table 4. Generalized linear mixed effects model analyses by application usage duration per week subgroup

| Measures                                               | The average application usage duration per week <40 minutes (N=36) | The average application usage duration per week ≥40 minutes (N=32) |
|--------------------------------------------------------|--------------------------------------------------------------------|-------------------------------------------------------------------|
|                                                        | F                    | p        | F                    | p        |
| Brief encounter psychosocial instrument (BEPSI-K)      | 18.185               | <0.001   | 11.189               | 0.002    |
| Center for Epidemiologic Studies Depression Scale (CES-D) | 3.835                | 0.058    | 6.835                | 0.014    |
| Korean occupational stress scale (KOSS)                | 0.004                | 0.949    | 7.082                | 0.012    |
| Framingham A type behavior pattern (FTA)               | 0.099                | 0.755    | 3.341                | 0.078    |
| Korean resilience quotient -53 (KRQ-53)                | 0.582                | 0.451    | 5.714                | 0.023    |
| International physical activity questionnaire (IPAQ)    | 0.785                | 0.379    | 2.832                | 0.098    |
| Mini dietary assessment (MDA)                          | 5.349                | 0.027    | 0.876                | 0.357    |
| Alcohol Use Disorders Identification Test (AUDIT)      | 0.118                | 0.733    | 4.255                | 0.048    |
| Fagerström Test for Nicotine Dependence (FTND)         | 1.086                | 0.305    | 3.008                | 0.093    |

The SRRS score and recruited year were applied as covariates. SRRS: social readjustment rating scale
spect to the overall usability of the application (4.0 points on a 5-point Likert scale). They were mostly content with appreciation meditation, ease of use, and speed of the app (4.26 points); 41 participants (60.3%) replied that they were willing to pay money to use the application.

**DISCUSSION**

Our study suggests that our application may be an acceptable strategy to manage stress for employees in Korea. After a four-week intervention, perceived stress level and depressive symptoms were significantly decreased. No one dropped out before the end of the study, and they used the application 2.7 times a week on average. Participants were generally satisfied with their experiences.

It is notable that the CES-D score in our study was higher than the general population. Previous study with Korean employees also reported high CES-D scores (mean=13.02, SD=7.75 for females; mean=12.25, SD=8.06 for males). Korean employees generally suffer from high stress level and job satisfaction due primarily to long working hours and strict occupational climate. It is also possible that employees who experienced psychological symptoms were more willing to participate in this study.

Smartphone-based stress management intervention is a promising tool for managing mental health problems with high accessibility and availability. Intervention is made available to users anytime and anywhere, at their own leisure, with anonymity. Recently published studies showed that smartphone-based stress management intervention was effective in reducing mental health-related symptoms; however, evidence for its usability and acceptability are still scarce.

Smartphone-based interventions are beginning to emerge as a method of treatment for various mental health problems. However, it is important to carefully consider the characteristics of the equipment used when planning for content. Unlike face-to-face intervention or web-based intervention using desk-top computers, smartphone users can access interventions on their devices in any location for a relatively short duration. Thus, content provided should accommodate this usage pattern. In our application, we tried to accommodate this characteristic by offering various types of interventions on-demand with time constraint in mind—each type requires only a few minutes to complete. Considering that traditional psychiatric interventions for clinical conditions (i.e., depression or anxiety disorder) require a longer time commitment with high intensity, smartphone-based intervention may not be a suitable replacement for the traditional intervention for clinical conditions. Smartphone-based stress management intervention targets relatively milder conditions with more resilient populations.

In this study, we did not provide direct human support. It is unclear whether this lack of direct human contact—unguided intervention—is effective in stress management. Several studies reported unguided electronic (i.e., Desktop computer-based) mental health interventions were less effective than guided interventions, but they targeted clinical populations, who generally require more intensive treatment than employees with stress. Christenson et al. reported that depression severity significantly affected adherence to electronic intervention programs, suggesting that less severe mental health conditions may not require direct human support. Moreover, unlike desktop computers, smartphones can utilize push notifications, i.e., alarms, to remind users, as incorporated in our study.

A meta-analysis on the effects of stress management intervention showed that cognitive behavioral techniques generated the largest effects. However, it is unclear which method was most effective in the smartphone setting. In our study, we applied various techniques, including self-monitoring, psychoeducation, cognitive behavioral technique, and relaxation components, together and let users choose their favorite methods. Hence, it is unclear which technique was actually most effective for managing stress.

It should be noted that certain groups may respond better to smartphone-based intervention. This should be considered in the planning phases. In our study, we found that younger age (<35 years of age) and female groups showed a greater significant decrease in depressive symptoms. Males and females generally showed different coping strategies and attitudes toward health problems. Previous studies showed that females put higher effort to overcome their mental health problems. This may contribute to females showing greater compliance with the self-guided intervention. In our study, females not only showed greater improvement in depressive symptoms and perceived stress levels, but they also showed improvement in type A personality patterns and alcohol use problems. This suggests that smartphone-based interventions can effectively prevent mental health problems in female employees.

The younger age group may be considered as target patient group for the smartphone-based intervention. However, in our study, those ≥35 years showed significant improvement in KOS scores while those <35 years did not. Previous studies also showed mixed results about the effect of age on adherence to electronic interventions. Further study is needed to better determine the role of age on the effects of smartphone-based intervention.

Smartphone-based application for mental health has become dramatically common recently. Following this trend, evidence-based recommendation for application develop-
Smartphone-Based Stress Management Intervention

The relationship between nicotine dependence scores and biomarkers of effectiveness on reducing depressive symptoms and stress levels. Additional studies are needed to further investigate feasibility, acceptability, and long-term effects of our application on preventing and managing mental health problems.

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