

# The exploration of use of an online knowledge base and effect on patient experiences and health care quality: An observational study in primary health care

Christian Gerdeskold (✉ [christian.gerdeskold@ki.se](mailto:christian.gerdeskold@ki.se))

Karolinska Institutet <https://orcid.org/0000-0002-7772-5451>

Eva Toth-Pal

Karolinska Institutet Department of Neurobiology Care Sciences and Society

Inger Wårdh

Karolinska Institutet Department of Dental Medicine

Gunnar H. Nilsson

Karolinska Institutet Department of Neurobiology Care Sciences and Society

Anna Nager

Karolinska Institutet Department of Neurobiology Care Sciences and Society

---

## Research article

**Keywords:** clinical decision support, information systems, knowledge bases, registries, primary care

**Posted Date:** July 20th, 2020

**DOI:** <https://doi.org/10.21203/rs.2.23948/v2>

**License:**  This work is licensed under a Creative Commons Attribution 4.0 International License.

[Read Full License](#)

---

**Version of Record:** A version of this preprint was published on November 16th, 2020. See the published version at <https://doi.org/10.1186/s12911-020-01313-9>.

# Abstract

**Background:** Evidence-based information available at the point of care improves patient care outcomes. Online knowledge bases can increase the application of evidence-based medicine and influence patient outcome data which may be captured in quality registries. The aim of this study was to explore the effect of use of an online knowledge base on patient experiences and health care quality.

**Methods:** The study was conducted as a retrospective, observational study of 24 primary health care centers in Sweden exploring their use of an online knowledge base. Frequency of use was compared to patient outcomes in two national quality registries. A socio-economic care need index was applied to assess whether the burden of care influenced the results from those quality registries. Non-parametric statistical methods and linear regression were used.

**Results:** Frequency of knowledge base use showed two groups: frequent and non-frequent users, with a significant use difference between the groups ( $p < .001$ ). Outcome data showed significant higher values for all seven National Primary Care Patient Survey dimensions in the frequent compared to the non-frequent knowledge base users ( $p < .001$ ), whereas 10 out of 11 parameters in the National Diabetes Register showed no differences between the groups ( $p > .05$ ). Adjusting for care need index had almost no effect on the outcomes for the groups.

**Conclusions:** Frequent users of a national online knowledge base received higher ratings on patient experiences, but figures on health care quality in diabetes showed near to no correlation.

The findings indicate that some effects may be attributed to the use of knowledge bases and requires a controlled evaluation.

## Background

The present study explored the effect of use of an online knowledge base on patient experiences and health care quality in primary health care centers in Sweden. The possible effects of knowledge base use on patient care outcomes in quality registries are scarcely reported. To the best of our knowledge, the present study is the first to examine such possible associations.

Medicine is a knowledge-intensive area with a continuous need to keep up-to-date with the latest evidence and to apply it to everyday patient care. Evidence-based medicine (EBM) connects the current best evidence with clinical practice (1, 2). In order to apply EBM in practice, knowledge is needed at the point of care (3). Online knowledge bases may provide for these needs (4). By capturing EBM health care outcomes in medical quality registries, real-world evidence can be used to improve the quality of health care (5, 6).

EBM is “the use of the best available evidence for decision-making related to the treatment of a specific patient by applying results of systematic, reproducible, unbiased research in clinical practice” (7, 8). EBM

used at the point of care has been reported to mitigate risk, effectively improve patient care outcomes, and reduce cognitive overload which can lead to medical errors (1, 2, 7, 9). A crucial step in EBM is to translate the evidence and apply the results in clinical practice. Knowledge does not necessarily change practice and mere dissemination of scientific evidence may be insufficient to change professional behavior (10).

Online knowledge bases have been shown to increase the application of EBM in clinical practice (11-13). In this study, we used Lobach's definition of a knowledge base: "Electronic (computer-based) resources comprising distilled (synthesized) or curated information that allows clinicians to select content germane to a specific patient to facilitate medical decision making" (14). The use of knowledge bases is associated with a positive impact on clinician behavior and patient outcomes, and evidence suggests that use of knowledge bases may be associated with improved knowledge and patient outcomes (12, 15, 16).

The knowledge base explored in this study was Medibas, a web-based knowledge source for general practitioners (GPs) in Sweden providing access to evidence-based medical knowledge in everyday clinical life (17). Medibas' editorial staff of GPs and its network of over 200 specialist doctors gather and summarize new scientific studies, reviews, national guidelines and recommendations and incorporate these into the knowledge base.

A national quality registry contains individual-based information on diagnoses and medical outcome measures in health care. National quality registries can identify factors that may impact on patient survival (18). There are over a hundred national quality registries in Sweden (19). Quality registries have the potential to collect real-world data, i.e. data collected outside of randomized controlled trials showing the unbiased results of real-life daily clinical practice. Patient experience data can be collected as patient reported outcome measures – questionnaires to record their experience of health care services. This can provide an understanding of how health care interventions impact on patients' quality of life and allows for comparisons of health care providers' performances (20). Real-world data can also be collected in quality registries as objective outcome data, e.g. laboratory results or findings in physical examinations.

There is a lack of knowledge on whether the use of knowledge bases reflects the patient outcomes data in nationwide quality registries and whether burden of care, measured in Sweden as a care need index (CNI), influences the register outcomes in any way (21). The aim of this study was to explore the effect of use of an online knowledge base on patient experiences and health care quality.

## Methods

### Study design and setting

A retrospective, observational ecological study design was used in the present study. Data on frequency of use of the knowledge base during 2018 was collected from primary health care centers in Stockholm, Sweden.

In parallel, a cross-sectional set of standardized outcome measures were collected from two national and regional quality registries: one containing subjective data of patient experiences in health care encounters and one containing objective data of health care outcomes from diabetes care.

## **Study material**

A total of 24 privately-owned primary health care centers participated in this study. No individual physicians were selected. Eligible participating sites were primary health care centers in Stockholm that used Medibas during full year 2018, and all centers had online access to the Internet. Primary health care centers in Stockholm were chosen because they have individual IP addresses and are thereby traceable. The knowledge base's central customer server was used to obtain data on frequency use during 2018 for each primary health care center. A "session" in the knowledge base was defined as one occasion where the user was active on the web site.

## **The knowledge base**

Medibas is a Swedish medical online knowledge base with a focus on primary care (17). The primary target audience is GPs, but Medibas also targets other occupational groups in primary care such as nurses and physiotherapists. The aim of the knowledge base is to provide access to evidence-based knowledge in everyday clinical practice. The knowledge base is based on the 'Norsk Elektronisk Legehåndbok' (Norwegian Electronic Physician Handbook) which has been used by Norwegian GPs since the 1990s and was adapted to Swedish health care in 2013.

The knowledge base used for this study contains more than 4,000 articles and covers a wide range of diagnoses in healthcare. The texts are written by specialists in general medicine and are reviewed and adapted to Swedish guidelines. In addition to facts about symptoms, diagnostics, treatment and follow-up, Medibas also features illustrations and patient information which can be easily printed out. Each text also contains references with direct links to studies in PubMed, the Cochrane library and national or regional guidelines. The content is updated on a weekly basis to include new findings from national and international evidence-based sources of knowledge. Longer texts feature a summary at the beginning to provide an overview. Medibas receives financial support through subscription fees and does not contain advertising or commercial promotion. This knowledge base was chosen for the present study as it is the most comprehensive knowledge base accessible to primary health care centers on a nationwide scale in Sweden. It also contains patient education handouts, which may increase the willingness of patients to be compliant, according to an earlier study (22).

## **The Swedish National Primary Care Patient Survey**

The Swedish National Primary Care Patient Survey (NPCPS) is a recurrent national survey of patient experiences (23). Since 2009, all Swedish health care regions (n=21) have participated and the survey is coordinated by the Swedish Association of Local Authorities and Regions. The survey is carried out every two years and includes both primary and specialized care. The most recent survey in Stockholm was

carried out in 2018 (n=57,384) and the response rate was 35.4%. Thus, 20,313 patients responded to the questionnaire. A random sample of patients who had visited primary health care centers received an invitation to respond to a web or postal questionnaire. Confidentiality was ensured and it was not possible to read an individual's responses when the results were compiled. The questionnaire consisted of seven dimensions on a five to seven graded Likert scale: overall impression, emotional support, participation and involvement, respect and treatment, continuity and coordination, information and knowledge, and accessibility.

### **The Swedish National Diabetes Register**

The Swedish National Diabetes Register (NDR), founded in 1996, has long been a cornerstone of diabetes care in Sweden, providing clinicians with evidence-based information and supporting the improvement of health care quality (24). Both hospitals and primary health care centers input diabetes patient data online, and in return benefit from opportunities to monitor risk factors, receive help in identifying needs for treatment improvements, and minimize the potential consequences of diabetes. The NDR currently contains data on 425,000 patients and has a coverage of 97% of all Swedish diabetic patients. Data is either entered manually and reported online or directly by transmission from the patient's electronic medical records (22). The Swedish Society for Diabetology is the owner of the registry and receives financial aid from the Swedish Association of Local Authorities and Regions. There are over 50 variables for each patient in the NDR, e.g. blood pressure, HbA1c and blood lipids.

### **National quality registries**

Data was extracted from the NPCPS and the NDR on a single occasion in April 2019 for the time period of a full 12 months from January to December 2018. Eligible data sets had no missing data. The number of sessions in the knowledge base per listed patient during 2018 were dichotomized into two groups of primary health care centers: frequent and non-frequent users of the knowledge base. We did a binary split into these two groups in order to analyze how outcomes in the NPCPS and NDR were related to frequency of use of the knowledge base.

### **Care Need Index**

CNI is a socio-economic needs index which describes the expected risk of developing ill health based on socio-economic factors on an individual level (21, 25).

The seven variables and their relative weights are:

CNI variable	Relative weight (0-9)
1. Aged over 65 years and single	6.15
2. Born abroad (Eastern Europe, Asia, Africa or South America)	5.72
3. Unemployed (or in employment measure), 16-64 years	5.13
4. Single parent with children who are 17 years or younger	4.19
5. Persons, aged one year or older, who recently moved into the healthcare center's catchment area	4.19
6. Low educational status, 25-64 years	3.97
7. Aged under five years	3.23

The CNI is an estimate to measure the workload of Swedish GPs. Compensation for socio-economic weight is paid per listed patient. Based on the socio-economic weight in CNI for each listed person, an index is calculated for the entire primary health care center's patient list, which then forms the basis for the compensation. CNI is not dependent on the number of visits to the primary health care center. About 60% of the Swedish population (n=10 million) do not sort into a CNI variable while 7-8% of the population Re actual for two or more variables. In order to explore whether the outcomes data from the NPCPS and the NDR were influenced by socio-economic weights, we selected CNI as a measure to reveal a potential relationship.

## Statistical Analysis

Descriptive statistics (mean and standard deviation) were used to describe the knowledge base user groups and register data (26). The Mann-Whitney test was used to calculate differences between knowledge base user groups (27). We tested dimensions of the NPCPS against CNI using linear regression (28). In order to test the seven dimensions from the NPCPS against CNI for the dichotomized user groups, the CNI data, originally showing an uneven distribution (non-parametric), was transformed to a normal distribution (parametric) using the Johnson Transformation Method (29). Using linear regression, we tested the difference between the NPCPS dimensions, NDR parameters and CNI for the dichotomized user groups. A p-value of <.05 was considered significant for all statistical analyzes. SPSS statistical software version 26 was used to analyze the data (30).

## Results

Frequency of use, as measured by the number of sessions in the knowledge base, and the division of the 24 primary health care centers into two groups: frequent users and non-frequent users are displayed in Figure 1. The number of listed patients was on average 8,903 in the frequent user group and 11,911 in the non-frequent user group. The listed patients were equally distributed (p=.320) between the two groups (Table 1). CNI was 1.95 in the frequent user group and 2.55 in the non-frequent user group (p=.052). The frequency of sessions per listed patient was significantly higher in the frequent user group than in the

non-frequent user group ( $p < .001$ ), as was the response rate for the NPCPS, 39.91% vs. 33.53% ( $p = .013$ ) (Table 1).

## **Patient experiences**

All seven dimensions of the NPCPS - overall impression, emotional support, participation and involvement, respect and treatment, continuity and coordination, information and knowledge, and accessibility - were rated significantly higher among patients listed primary health care centers in the frequent user group vs. those in the non-frequent user group (Table 2).

## **Health care quality**

The NDR parameters encompass different criteria, from medical history and physical findings such as blood pressure measurements and foot examinations to laboratory test

results of blood and urine. There was no significant difference in the NDR parameters between the user groups except for the parameter "HbA1C >70", which was significant ( $p = .045$ ) and therefore separated the user groups (Table 3).

## **Adjusting for CNI**

In order to adjust for the possible effect of CNI, an additional multiple regression analysis was performed (not shown). The addition of CNI of the primary health care centers in a multiple regression analysis did not affect the statistical outcomes from the NPCPS. For outcomes from the NDR, the addition of CNI of the primary health care centers changed the parameter "HbA1C >70" from significant to non-significant ( $p > .05$ ) and the parameter "Physical inactive" from non-significant to significant ( $p < .05$ ), i. e significantly higher percentage physical inactive patients reported to NDR among primary health care centers with non-frequent use of the knowledge base. Adding mean age of primary health care centers' patients reported to NDR did not affect the results from the multiple regression analysis.

## **Discussion**

Primary health care centers using the knowledge base grouped into two categories: frequent and non-frequent user groups. The number of sessions in the knowledge base per listed patient, i.e. frequency of use of Medibas, showed a significant difference between these two groups.

The results showed that primary health care centers using the knowledge base frequently scored higher in the NPCPS. This may suggest that physicians who use the knowledge base frequently are able to fulfil the needs of patients more effectively. The NPCPS contains subjective ratings by patients and reflects their views on the standard of care. The results from the NDR showed no differences between frequent and non-frequent users of the knowledge base. This could indicate that diabetic variables such as average blood glucose level, blood lipids and blood pressure are more static and the features influencing these values are multifactorial, genotype- and phenotype-wise. Physicians' use of electronic knowledge

bases might have a diminutive influence on these physical parameters. Other researchers have found that some behavioral changes can be more easily moderated than physical parameters (31). Based on the findings of the present study, there may be a relationship between patient outcome measurements in the NPCPS and frequency of use of the knowledge base. On the other hand, objective parameters such as those in the NDR may not be impacted by the use of an online knowledge base. It cannot be ruled out that there is no effect, but results of this study neither confirm nor refute this.

CNI describes the expected risk of developing ill health based on socio-economic factors, and could therefore possibly affect how burden of care influences patient outcomes between frequent and non-frequent users of the knowledge base. It could be argued that patients with low expected risk of developing ill health due to socio-economic factors tend to cluster in primary health care centers where physicians are highly committed to satisfying individual patients' needs and demands. We therefore added CNI to test this hypothesis. We found no effect (apart from one minor parameter in the NDR) when adding CNI to the two groups. This may reflect that burden of care is already embodied in the NPCPS as well as in the results from the NDR. The addition of CNI, whose inherent values may already be reflected in the register data, seems to have small effect and therefore not susceptible to influence from the knowledge provided by the knowledge base.

Previous research has focused on interventions to either increase the use of, or find new ways of using, electronic knowledge sources, whereas this study specifically examined the effect of the use of a knowledge base on patient outcome measures in two nationwide registries. Earlier studies have found that use of register data may play a vital role in patient care (12, 15, 32). Furthermore, an excess of research has been devoted to evaluating electronic knowledge sources by employing self-reported use, which is prone to biases (33-35). We are not aware of any prior studies examining outcomes of the use of an online knowledge base by relating frequency of use to objective data from quality registries. Recent studies have given valuable clarification on factors influencing knowledge seeking such as lack of time, resource use and accessibility (4, 36-38). The present study adds to these findings by exploring the frequency of use of an online knowledge base and investigating its potential impact on unbiased outcome measures, e.g. objective outcome data from quality registries. The study brings new knowledge of how to evaluate the use of a clinical knowledge base and its possible impact on health care quality. The technical novelty in this study lies in its front line approach to evaluating the effects of clinical knowledge applied to patient care. To the best of our knowledge there are few, if any, studies which take this new approach to evaluation. Previous studies in this field have mostly been in the form of self-assessments, and thereby subject to recall bias, whereas our study looks at objective data of knowledge base use combined with results from national quality registries (33-35).

The NPCPS was chosen for the study as it represents a high-quality nationwide patient reported outcomes programme. The NDR was chosen because it is the quality register for diabetes care in Sweden and contains nationwide diabetes data of high validity, reliability and granularity. The collected information is objective as it is transferred automatically from the electronic patient records.

This study suffers from several limitations including major confounding issues. The study design only looks at associations and not evidence for causation. Hence the conclusions of how a knowledge base can improve health care quality has to be interpreted with great care. The purpose of this “first of its kind” study was to establish a starting point for a non-self-reported way of looking at collected user data. In the future there needs to be a focus on individual data, both from the user perspective as well as from the patient’s side. Measuring effects of health care by studying outcomes data, e.g. in quality registries, has inherent limitations and should primarily be used for hypotheses generation. Other limitations in this study are the small sample size of the study population, the response rates of patient surveys, the low frequency of response in the NPCPS and the selection of privately-owned primary health care centers. The NPCPS represents an important qualitative source of information on patient preferences, but it may be difficult to extrapolate our findings to other contexts. The generalizability of our findings into other areas, such as rural geographical areas and publicly-run primary health care centers, may be limited. Further, the use of “sessions” in the knowledge base may not adequately reflect actual use of the knowledge base. Meanwhile, strengths of this study include the use of objective outcome measures (e.g. quality register data) and no self-reported results. Another strength is that a total geographical group of the knowledge base’s users was investigated. Finally the frequent and non-frequent user groups of the knowledge base have significant differences. We believe that these differences are characteristics of the two groups; the frequent users tend to have more satisfied patients who consequently score higher on satisfaction of given care. Vice versa does the non-frequent user group seem to have less satisfied patients.

Future research in this field is needed in the form of results from an unbiased selection of patients’ and caregivers’ experiences of knowledge base use in the form of a randomized controlled trial. It should aim to find new methods to support causation between the use of a knowledge base and impact on health care quality.

## **Conclusions**

Frequent users of a national online knowledge base received higher ratings on patient experiences, but figures on health care quality in diabetes showed near to no correlation. The findings indicate that some effects may be attributed to the use of knowledge bases and requires a controlled evaluation.

## **Declarations**

### **Ethical approval**

This study was subject of an ethical application and was deemed exempt from full review by the Swedish Ethical Review Authority. The reviewing body gave an advisory statement declaring that no ethical conflicts were found in the present study (Registration number: 2019-04035).

### **Consent for publication**

Not applicable.

### **Availability of data and materials**

The data sets in the present study can be made available from the corresponding author on request.

### **Competing interests**

Christian Gerdesköld has received research funding support from Medibas. Anna Nager is employed part-time by Medibas. All other authors declare no conflict of interest.

### **Funding**

This study was funded by Medibas.

### **Authors' contributions**

Christian Gerdesköld (C.G.) and Anna Nager (A.N.) developed the study concept. All authors contributed to the study design. Testing and data collection were performed by A.N. and C.G.

All authors drafted, analyzed and discussed the results, contributed to the final manuscript and approved the final version of the manuscript for submission.

### **Acknowledgements**

We wish to acknowledge the contributions of Sven-Erik Johansson for his kind assistance in the statistical set-up and analysis of data.

## **List Of Abbreviations**

CNI	Care Need Index
EBM	Evidence-based medicine
GP	General practitioner
NDR	The Swedish National Diabetes Register
NPCPS	The Swedish National Primary Care Patient Survey

## **References**

1. Fossum M, Herbst K, Harper L, Castagnetti M, Beckers G, Bagli D, et al. Evidence-based medicine 1: Background. J Pediatr Urol. 2019;15(1):78-9.
2. Gillam S, Siriwardena AN. Evidence-based healthcare and quality improvement. Qual Prim Care. 2014;22(3):125-32.

3. Van de Velde S, Heselmans A, Delvaux N, Brandt L, Marco-Ruiz L, Spitaels D, et al. A systematic review of trials evaluating success factors of interventions with computerised clinical decision support. *Implement Sci.* 2018;13(1):114.
4. Cook DA, Sorensen KJ, Wilkinson JM, Berger RA. Barriers and Decisions When Answering Clinical Questions at the Point of Care: A Grounded Theory Study. *JAMA Intern Med.* 2013;173(21):1962-9.
5. Djulbegovic B, Guyatt GH. Progress in evidence-based medicine: a quarter century on. *Lancet.* 2017;390(10092):415-23.
6. Allen-Duck A, Robinson JC, Stewart MW. Healthcare Quality: A Concept Analysis. *Nurs Forum.* 2017;52(4):377-86.
7. Kang H. How to understand and conduct evidence-based medicine. *Korean J Anesthesiol.* 2016;69(5):435-45.
8. Tebala GD. The Emperor's New Clothes: a Critical Appraisal of Evidence-based Medicine. *Int J Med Sci.* 2018;15(12):1397-405.
9. Mondoux S, Shojania KG. Evidence-based medicine: A cornerstone for clinical care but not for quality improvement. *J Eval Clin Pract.* 2019;25(3):363-8.
10. Le JV. Implementation of evidence-based knowledge in general practice. *Dan Med J.* 2017;64(12).
11. Akehurst J, Sattar Z, Gordon I, Ling J. Implementing online evidence-based care pathways: A mixed-methods study across primary and secondary care. *BMJ open.* 2018;8(12):e022991.
12. Maggio LA, Aakre CA, Del Fiol G, Shellum J, Cook DA. Impact of Clinicians' Use of Electronic Knowledge Resources on Clinical and Learning Outcomes: Systematic Review and Meta-Analysis. *J Med Internet Res.* 2019;21(7):e13315.
13. Porat T, Delaney B, Kostopoulou O. The impact of a diagnostic decision support system on the consultation: perceptions of GPs and patients. *BMC Med Inform Decis Mak.* 2017;17(1):79.
14. Lobach D, Sanders GD, Bright TJ, Wong A, Dhurjati R, Bristow E, et al. Enabling health care decisionmaking through clinical decision support and knowledge management. *Evid Rep Technol Assess.* 2012(203):1-784.
15. Aakre CA, Pencille LJ, Sorensen KJ, Shellum JL, Del Fiol G, Maggio LA, et al. Electronic Knowledge Resources and Point-of-Care Learning: A Scoping Review. *AcadMed.* 2018;93(11S Association of American Medical Colleges Learn Serve Lead: Proceedings of the 57th Annual Research in Medical Education Sessions):S60-s7.
16. Grimshaw JM, Eccles MP, Lavis JN, Hill SJ, Squires JE. Knowledge translation of research findings. *Implement Sci.* 2012;7:50.
17. Medibas. Medibas [Internet]. Stockholm 2019 [Available from: <https://medibas.se/intro/>]
18. Hoque DME, Kumari V, Hoque M, Ruseckaite R, Romero L, Evans SM. Impact of clinical registries on quality of patient care and clinical outcomes: A systematic review. *PLoS One.* 2017;12(9):e0183667.
19. Emilsson L, Lindahl B, Köster M, Lambe M, Ludvigsson JF. Review of 103 Swedish Healthcare Quality Registries. *J Intern Med.* 2015;277(1):94-136.

20. Black N. Patient reported outcome measures could help transform healthcare. *BMJ*. 2013;346:f167.
21. Sundquist K, Malmström M, Johansson S, Sundquist J. Care Need Index, a useful tool for the distribution of primary health care resources. *J Epidemiol Community Health*. 2003;57(5):347-52.
22. Adolfsson ET, Rosenblad A. Reporting systems, reporting rates and completeness of data reported from primary healthcare to a Swedish quality register – the National Diabetes Register. *Int J Med Inform*. 2011;80(9):663-8.
23. NPS. National Primary Care Patient Survey 2018. Available from: <https://patientenkat.se/sv/>.
24. Diabetesregistret N. Årsrapport 2017. Nationella Diabetesregistret (NDR) – Registercentrum Västra Götaland, Göteborg, Sweden. 2018.
25. Malmström M. Care need index, social deprivation and health: epidemiological studies in Swedish health care. Malmö 1999.
26. Nick TG. Descriptive Statistics. In: Ambrosius WT, editor. *Topics in Biostatistics*. Totowa, NJ: Humana Press; 2007. p. 33-52.
27. Nachar N. The Mann-Whitney U: A test for assessing whether two independent samples come from the same distribution. *Tutorials in quantitative Methods for Psychology*. 2008;4(1):13-20.
28. Olive DJ. *Linear Regression*. Cham: Springer International Publishing; 2017.
29. Olive DJ, Springer International Publishing AG. *Linear Regression* 2018.
30. Wagner WE. *Using IBM SPSS statistics for research methods and social science statistics*. Los Angeles: SAGE; 2019.
31. Marteau T, Hollands G, Kelly M. *Emerging Behavioral and Social Science Perspectives on Population Health*. 2015.
32. Aakre CA, Maggio LA, Fiol GD, Cook DA. Barriers and facilitators to clinical information seeking: a systematic review. *J Am Med Inform Assoc*. 2019;26(10):1129-40.
33. Koppel R. Is healthcare information technology based on evidence? *Yearb Med Inform*. 2013;8:7-12.
34. Scott PJ, Brown AW, Adedeji T, Wyatt JC, Georgiou A, Eisenstein EL, et al. A review of measurement practice in studies of clinical decision support systems 1998–2017. *J Am Med Inform Assoc*. 2019;26(10):1120-8.
35. Sousa VEC, Dunn Lopez K. Towards Usable E-Health. A Systematic Review of Usability Questionnaires. *Appl Clin Inform*. 2017;8(2):470-90.
36. Cook DA, Enders F, Linderbaum JA, Zwart D, Lloyd FJ. Speed and accuracy of a point of care web-based knowledge resource for clinicians: a controlled crossover trial. *Interact J Med Res*. 2014;3(1):e7.
37. Cook DA, Sorensen KJ, Hersh W, Berger RA, Wilkinson JM. Features of effective medical knowledge resources to support point of care learning: a focus group study. *PLoS One*. 2013;8(11):e80318.
38. Cook DA, Thompson WG. Comfort and experience with online learning: trends over nine years and associations with knowledge. *BMC Med Educ*. 2014;14:128.

## Tables

**Table 1.**

Characteristics of the two user groups of the knowledge base, presented as mean values, standard deviations. (NPCPS: The Swedish National Primary Care Patient Survey)

<b>Variables</b>	<b>Frequent users (n=8) (SD)</b>	<b>Non-frequent users (n=16) (SD)</b>	<b>P- value<sup>1</sup></b>
Listed patients at the primary health care centers	8,903 (2,769)	11,911 (6,067)	.320
Care Need Index	1.95 (0.40)	2.55 (0.86)	.052
Sessions per 1,000 listed patient	169 (95.39)	24 (21.48)	.000*
Response rate as percent of NPCPS	39.91 (4.21)	33.53 (1.55)	.013*
Mean age in National Diabetes Register	68.62 (0.68)	67.20 (0.66)	.548

<sup>1</sup>Mann-Whitney test, exact significance 2-tailed.

**Table 2.**

Results from the NPCPS for the frequent and non-frequent user groups, mean values and (standard deviations)

Dimensions in the NPCPS	Frequent user group (n=8) (%) (SD)	Non-frequent user group (n=16) (%) (SD)	P-value <sup>1</sup>
Overall impression	89.05 (3.67)	77.12 (8.30)	.001*
Emotional support	84.25 (4.44)	72.11 (9.24)	.001*
Participation and involvement	87.14 (2.77)	77.52 (5.76)	<.001*
Respect and treatment	91.20 (2.31)	82.14 (7.29)	.001*
Continuity and coordination	85.90 (4.49)	72.14 (9.20)	.001*
Information and knowledge	84.25 (3.82)	72.84 (6.76)	<.001*
Accessibility	88.45 (2.94)	80.58 (5.97)	.001*

<sup>1</sup>Mann-Whitney test, exact significance 2-tailed.

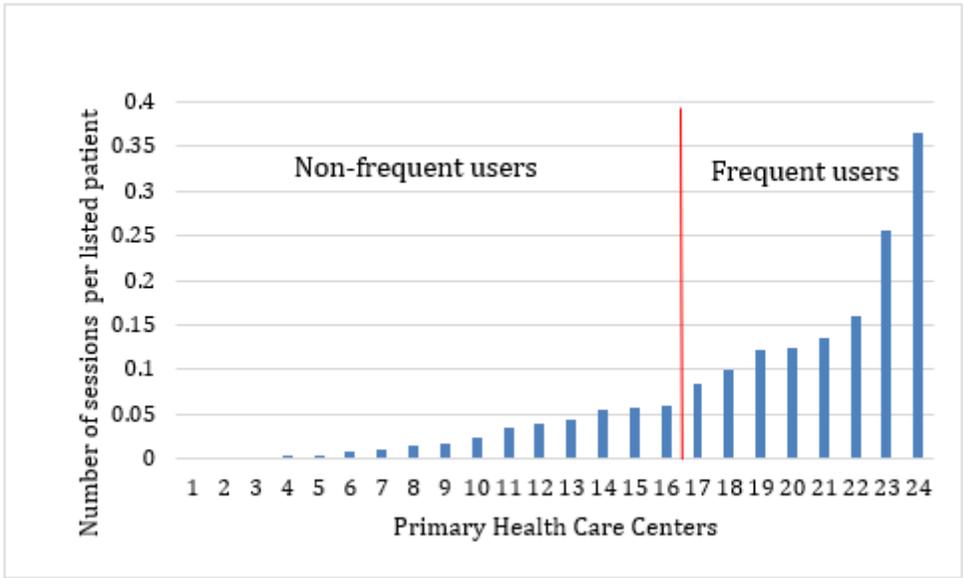
### Table 3.

Results for the NDR (The Swedish National Diabetes Register) parameters used in the study for frequent and non-frequent user groups, mean values and (standard deviations)

NDR parameters	Frequent users (n=8) (%) (SD)	Non-frequent users (n=16) (%) (SD)	P-value <sup>1</sup>
HbA1c <52 mmol/mol	61.05 (7.64)	58.44 (4.76)	.365
HbA1c >70 mmol/mol	7.01 (1.83)	9.05 (2.27)	.045*
Blood pressure ≤130/80 mm Hg	39.86 (6.71)	42.79 (5.17)	.097
Blood pressure <140/85 mm Hg	57.21 (9.13)	56.07 (5.18)	.912
LDL <2.5 mm/L	51.91 (7.03)	51.18 (8.36)	.765
Lipid lowering drug	44.06 (12.35)	52.26 (9.38)	.115
Albuminuria	24.44 (5.46)	23.01 (4.97)	.717
Foot exam	76.83 (8.01)	81.55 (12.58)	.119
Retinal examination	77.40 (13.55)	75.78 (26.06)	.265
Smoker	12.63 (4.27)	15.30 (3.50)	.184
Physically inactive	14.78 (8.43)	22.57 (8.90)	.065

<sup>1</sup>Mann-Whitney test, exact significance 2-tailed.

## Figures



**Figure 1**

The distribution of sessions in the knowledge base per listed patient for participating primary health care centers.