Voice-Based Screening For SARS-CoV-2 Exposure In Cardiovascular Clinics

Short title: Amazon Alexa and SARS-CoV-2 screening

Abhinav Sharma MD, PhD[1,2]; Emily Oulousian BSc[1,2]; Jiayi Ni MSc[2]; Renato Lopes MD, PhD[3]; Matthew Pellan Cheng MD[4]; Julie Label BSc[2]; Filipe Henriques BSc[2]; Claudia Lighter[1,2]; Nadia Giannetti MD[2]; Robert Avram MD, MSc[5,6]

1. DREAM-CV Lab, McGill University Health Centre (Montreal, Quebec, Canada)
2. Division of Cardiology, McGill University (Montreal, Quebec, Canada)
3. Duke Clinical Research Institute, Duke University (Durham, North Carolina, USA)
4. Divisions of Infectious Diseases and Medical Microbiology, McGill University Health Centre, McGill University (Montreal, Quebec, Canada)
5. Division of Cardiology, University of Ottawa (Ottawa, Ontario, Canada)
6. Montreal Heart Institute, University of Montreal (Montreal, Quebec, Canada)

Article type: Manuscript
Abstract word count: 250 words
Word count: 2,459 (including text, references, legend)
Table/figure count: Two figures and three table
References: 11

*Corresponding author

Abhinav Sharma MD, PhD
McGill University Health Centre
1001 Decarie Blvd
H4A3J1
Abhinav.sharma@mcgill.ca
Tele: (514) 934-1934 x35414
Fax: (514) 938-7396

Abstract

Aims

Artificial intelligence (A.I) driven voice-based assistants may facilitate data capture in clinical care and trials; however, the feasibility and accuracy of using such devices in a healthcare environment are unknown. We explored the feasibility of using the Amazon Alexa ('Alexa') A.I. voice-assistant to screen for risk-factors or symptoms relating to SARS-CoV-2 exposure in quaternary care cardiovascular clinics.

Methods

We enrolled participants to be screened for signs and symptoms of SARS-CoV-2 exposure by a healthcare provider and then subsequently by the Alexa. Our primary outcome was interrater reliability of Alexa to healthcare provider screening using Cohen's Kappa statistic. Participants rated the Alexa in a post-study survey (scale of 1 to 5 with 5 reflecting strongly agree). This study was approved by the McGill University Health Centre ethics board.

Results

We prospectively enrolled 215 participants. The mean age was 46 years (17.7 years standard deviation [SD]), 55% were female, and 31% were French speakers (others were English). In total, 645 screening questions were delivered by Alexa. The Alexa mis-identified one response. The simple and weighted Cohen's kappa statistic between Alexa and healthcare provider screening was 0.989 (95% CI: 0.982, 0.997) and 0.992 (955 CI 0.985, 0.999) respectively. The participants gave an overall mean rating of 4.4 (out of 5, 0.9 SD).

Conclusion

Our study demonstrates the feasibility of an A.I. driven multilingual voice-based assistant to collect data in the context of SARS-CoV-2 exposure screening. Future studies integrating such devices in cardiovascular healthcare delivery and clinical trials are warranted.

Registration: https://clinicaltrials.gov/ct2/show/NCT04508972

Introduction

Artificial intelligence (A.I.) driven voice-based technologies have significant potential to streamline and optimize delivery of healthcare and conduct of clinical trials.[1–3] Numerous commercial voice-based are now on the market and are poised to potentially play a larger role in healthcare environments.[2] Consumer data suggested that the use of voice assistant grew by 103% from 21.5 million in 2018 to 43.7 million in 2020.[4] However, the current major use case of these technologies is for search engine queries.[5] There is a lack of data surround the evaluation of such technologies to within a healthcare environment and furthermore, the reliability, accuracy, and concordance with manual or healthcare provider data collection has not been extensively explored. The SARS-CoV2 pandemic has provided an opportunity to test the feasibility of using A.I. driven voice-based devices to collect patient level information. Furthermore, such devices may facilitate optimal healthcare environment. For instance, within a healthcare setting, screening for risk factors or symptoms relating to SARS-CoV-2 exposure is often done directly by healthcare personnel, using standardized institution-specific questionnaires. Such methods are inconsistent and inefficient in screening multiple people. Amazon Alexa® ('Alexa') is a voice-based A.I. enabled personal assistant that can activate cloud based 'Skills' through verbal triggers.[6] The ability to create skill that comply with privacy regulation (for example the Health Insurance Portability and

Accountability Act [HIPAA] in the U.S) may enable devices to be integrated into clinical environments. There is limited evidence in the evaluation of voice-based technologies in healthcare settings or in the context of clinical trials.[1,2,6–8] By testing the utility and accuracy of using the Alexa in screening for symptoms or risk factors relating to SARS-CoV2, we will obtain greater evidence into the ability to integrate such devices into clinical care and future clinical trials. We prospectively evaluated the ability of Alexa to screen for risk-factors or symptoms relating to SARS-CoV-2 exposure (VOICE-COVID-19 Study; NCT04508972).[9]

Methods

The device utilized was the Amazon Echo Show 8. With Amazon Web Services, an Alexa Skill was developed using the Amazon Alexa Developer Kit.[7] Questions relating to risk factors for SARS-CoV-2 exposure were based on Health Canada recommendations **(Table 1)**.[10] The content of the skill was initially developed in English and then translated into French. As Alexa can only have one default language we set this in French but entered skill questions in both English and French. This resulted in a skill where the participant could select either French or English. The resulting English voice was spoken with a slight French accent.

Data capture using the Alexa

An Alexa skill uses a vocal interaction model and application logic to determine the participant's request (Figure 1). The voice algorithm is based on Natural Language Understanding (NLU) and Automatic Speech Recognition (ASR).[11] The ASR involves the recognition and translation of speech into text. However, as vocal intonations and inflections vary widely, the device needs NLU to first rearrange spoken data into a machine-readable format. The skill built for this study is a pre-determined sequence of screening questions driven by a vocal request: answering yes/no, numbers

for age, etc. The questionnaire is activated through an utterance (a specific phrase): "Alexa, ___"
The patient's words are streamed to the Alexa service in the cloud, ASR and NLU will respond through voice recognition and structure the information into a request, and the screening process will begin. Each response is stored as a slot value and will act as a request for the next question.

Recruitment and screening process

The participants were recruited from October 1ˢᵗ – December 31ˢᵗ 2020. Potential participants were recruited by verbally discussing with the individual upon entry into the cardiovascular clinic. No compensation was provided to the participant. Among participants who verbally consented, the screening questions were initially asked by the healthcare professional. Subsequently, the participant would then move towards the Amazon Echo device and initiate the screening process with the device. The first question asked by the Alexa was whether the participant wanted to proceed in French or English. Subsequent questions were asked about the purpose of visit to the clinic (**Table 1**). Then the Alexa then proceeded to initiate the screening questions using the language the participant selected. The screening questions asked by the healthcare provider and the Alexa were the same: Any international travel in past 14 days out of Canada; in close contact with a person with Covid-19 without personal protective equipment; or presence of any symptoms (in the prior 14 days) including cough, fever, runny nose, sore throat, diarrhea (**Table 1**). All responses regarding the screening questions were binary (yes/no) and if participants provided additional details this would not be recorded. Furthermore, if the Alexa device could not understand the participants response, the Alexa would indicate that the response was not understood and would ask the participant to re-respond. The participant was only able to respond to Alexa once the question was completely delivered else Alexa would restart the question. People

entering the cardiovascular clinic at the McGill University Health Centre were first manually screened by healthcare personnel. The entry location had a small enclave that enable a more private discussing with the healthcare personnel but still had ambient noise from the surrounding areas. If participants responded 'yes' to the screening questions either with the healthcare provider or the Alexa, following completion of the post-survey questions, the healthcare provider would then move the participant to an isolation area and conduct a more detailed history.

Post screening survey

After completion of the screening process, the Alexa Skill was rated by participants on a five-point scale. The scale was based on a previously validated scale of user preference and app engagement.[12] Participants were subsequently asked about any privacy concerns related to data storage or use of the Alexa application by asking the following question with a binary yes/no response: "Do you have any privacy concerns regarding the use of the Amazon Alexa to screen for symptoms".

Data flow, consent, and study ethics approval

The high-level data flow and study architecture for the VOICE-COVID-19 study is described in **Figure 1**. When the researchers are defining the specific questions and triggered, it gets uploaded onto the AWS cloud platform through the AWS S3 bucket. This will integrate into the AWS Lambda server which will transfer information into the Aws DynamoDB. With DynamoDB researchers can create database tables that can store and retrieve any amount of data and serve any level of request data flow.

This study was approved by the ethics board at the McGill University Health Centre. Verbal consent (written consent waiver was granted) was obtained from all adults participants 18 years and older (or from their legal representatives) prior to screening by the healthcare provider.

Statistical analysis

Categorical variables were presented as counts (percentages) and continuous variables were presented as median, 25th and 75th percentiles. The primary endpoint of the current study was the interrater reliability of Alexa versus manual screening using simple and weighted Cohen's Kappa statistic. To facilitated pragmatic data collection, data collection on the specific comorbidities of patients were not collected. Data were analyzed using SAS version 9.4 software (SAS, Cary, North Carolina). Statistical significance was based on a p value of ≤0.05.

Coding availability

Coding for the Amazon Alexa skill will be made available by requests to the corresponding author.

Results

Demographics

In total, 215 participants were screened (mean age of 46.1 ± 17.7 years; 118 [55%] females; **Table 2**). The range of age was 16 to 84 years. There were no participants that were offered the opportunity to participate who refused. There were 66 (31%) French speakers and the remainder were English. There were 114 (53%) patients or family members and the rest were hospital workers. In total, 645 screening questions were delivered by Alexa.

Screening results

There were 13 responses missing due to participant withdrawal (n=2) and loss of wireless internet connection (due to our hospital network) after starting. In 15 questions (out of 645; 2%) the study coordinator had to clarify how the respond to the Alexa due to the participant interrupting the Alexa delivered questions. In 8 attempts (out of 645; 1.2%) , the Alexa did not understand the response and asked the participant to repeat the response. There were 14 positive responses to screening questions. Alexa mis-identified one English response - one screening question was responded to as 'yes' while the Alexa capture this as 'no'. The overall simple Cohen's kappa statistic reflecting the correlation between Alexa and the healthcare personnel screening was 0.989 (95% CI: 0.982, 0.997). The weighted Cohen's Kappa correlation was 0.992 (955 CI 0.985, 0.999)(Figure 2). In the oldest tertile of age (range 57-84 years), the concordance was perfect with Cohen's kappa correlation of 1.0. There was perfect concordance in French with a Cohen's Kappa 1.00 (95% 1.00-1.00); for English the concordance was 0.96 (95% CI 0.886-1.000).

Post-screening survey

The median time for completion of the Alexa screening questions was 40 seconds (interquartile range 33 to 44 seconds). The mean (with standard deviation) participant rating (out of five) was as follows: 4.7 (0.7) for appropriately asking the questions; 4.7 (0.7) for ease of language comprehension; 4.5 (1.0) for ease of device use; and 4.4 (0.9) for overall rating (**Table 3**). No participant expressed any privacy concern.

Discussion

There has been limited evaluation of the use of voice-based devices in healthcare settings. In the present study, we demonstrated that the Amazon Alexa demonstrated near perfect concordance with screening questions compared to healthcare personnel. Furthermore, it appeared that there was no decrement in the ability to be screened by Alexa among older participants. The

ability to use commercial voice-based devices to facilitate SARS-CoV-2 screening may represent a strategy to reduce burden on healthcare personnel and minimize their exposure to potentially infected individuals.[2] Furthermore, these results suggest that future studies evaluating the utility of integrating the such devices into routine clinical care and clinical trials in non-SARS-CoV-2 related settings are warranted.

There are several aspects to the functionality of voice-based technologies. The hands-free SARS-CoV-2 screening is built on the Alexa Skills Kit's interactive voice-driven interface.[13] The ability to activate the skill without requiring additional step, and just through voice, makes the use of such technologies ideal for populations where data capture via electronic devices is challenging, such as amongst older participants.

As healthcare delivery and future clinical trials become more pragmatic, there will be increasing need to leverage digital technologies to facilitate rapid data acquisition. Our result demonstrate the potential for such devices to collect medical grade data with high degree of accuracy and reliability. Such technologies may have the potential to reduce the burden of healthcare delivery and trial conduct by enabling accurate remote data collection. In addition, leveraging voice-based A.I. to supplement such routine screening can possibly enable more effective workforce utilization. Our results provides important feasibility information to identify that data can be accurately collected through the Alexa. Extending the use of such voice-based technologies in clinical and non-clinical environments (such as schools, workplaces, or home settings) to evaluate clinical workflow and burden warrants further evaluation.

Our study also highlights the practical elements faced with such devices (**Table 4**). Voice-based technologies generally need continuous internet access. As we experienced, loss of wireless internet access may impair device functionality. Additional testing into functioning in 'offline'

modes will be needed to ensure continuous function in clinical settings. Furthermore, while privacy concerns are significant issues, especially around devices and technologies related to identifying those at risk of SARS-CoV2 exposure,[14] in our study, no participant expressed any privacy concern. Such privacy concerns will need to be explored with larger deployment of such voice-based technologies.

Expanding on the results of our study, there is great need in cardiovascular clinical studies and care delivery for accurate remote patient monitoring.[15,16] Currently, most remote monitoring strategies focus on either telephone or video-based follow-up along with additional wearable devices or patches.[17,18] Voice-based assistants may provide one potential tool to facilitate remote patient follow-up, especially in settings where directed synchronous interaction with a patient (through a phone call or via video-platform) is limited. While our study does not directly demonstrate the use of Alexa in these use cases, the high degree of concordance of data collected by Alexa to standardized questions, compared to a healthcare provider, suggests that further exploration of such technologies are warranted within cardiovascular studies and care delivery.

Study Limitations

There are several limitations. Our results cannot be extrapolated to other voice-based devices. In the one case where the Alexa mistook the response, an English response was recognized as French and miscoded Future studies will have to evaluate the accuracy of data capture in other languages. Challenges and solutions in using voice-based device are shown in Table 4. Our study was a non-randomized study but provides significant evidence to consider randomized studies of voice-based assistants in similar clinical environments. A larger sample size with more diverse population would enable greater generalizability of our findings. The time required to complete the screening

by the healthcare provider was not completed. Further studies comparing duration of interaction with Alexa and healthcare providers are needed. Our entry location still had ambient noise from the surrounding areas which may have impacted on the ability of the Alexa to correctly identify responses; however, these results highlight real-life environments in which such devices can be used. Currently, the participant responds to the questions delivered by the Alexa. Future studies enabling participants to ask questions to the Alexa regarding COVID-19 represents a novel opportunity to explore the ability of such devices to provide medical information.
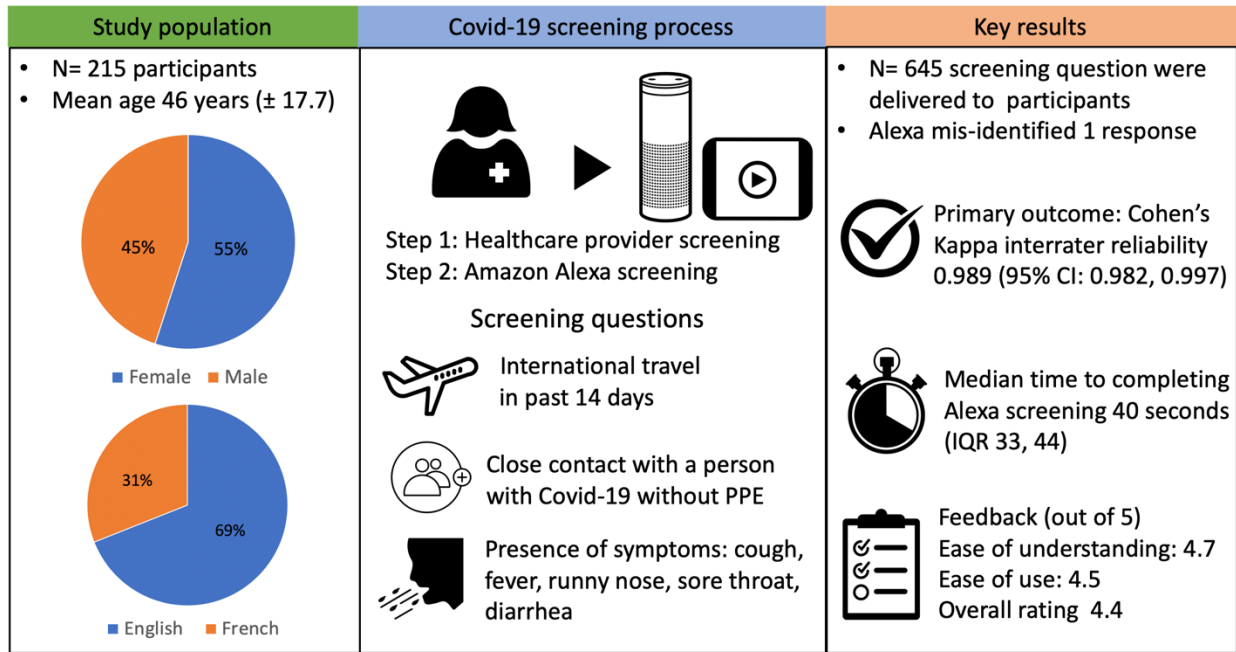
Conclusion

Our results demonstrated the feasibility of a multilingual voice-based assistant to accurately capture medical grade information and facilitate accurate SARS-CoV-2 exposure screening. Such results suggest the feasibility of integrating voice-based technologies in healthcare delivery and future clinical trials. Given the strong concordance of data collected by Alexa and the healthcare provider, evaluating the use of such technologies to facilitate cardiovascular studies and delivery of clinical care is warranted.

Figure 1: Architecture diagram for VOICE-COVID study data flow
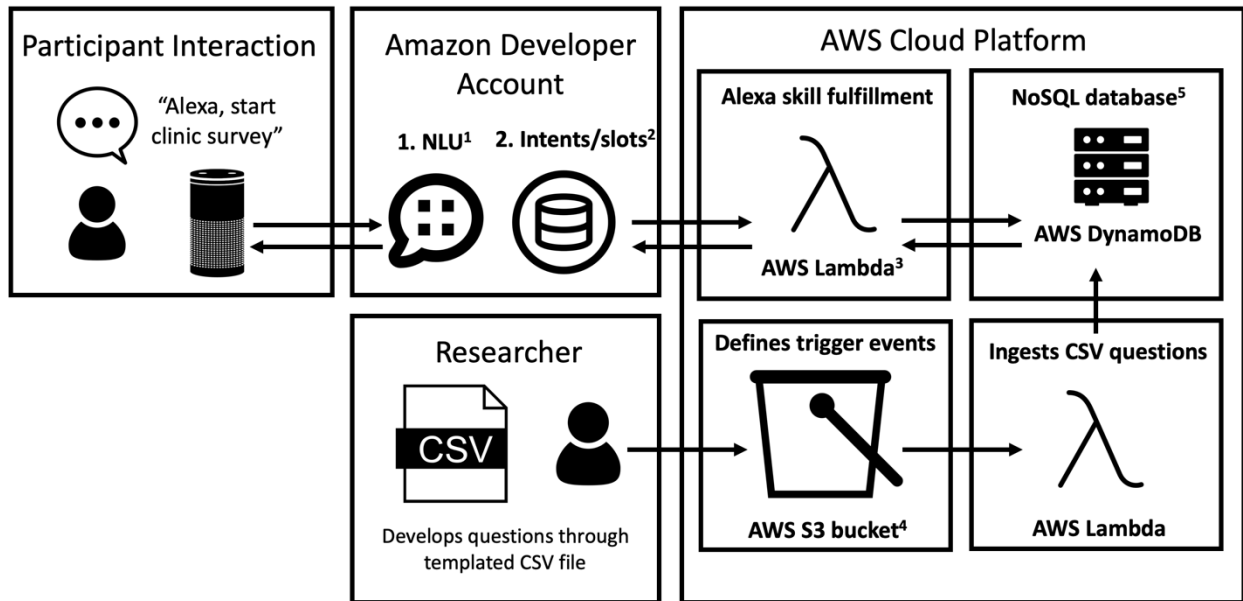
AWS amazon webservices; CSV comma separated value; NLU natural language understanding. 1. With natural language understanding (NLU), the data transferred from the Alexa device can deduce through artificial intelligence what a speaker is uttering. 2. The intent and slots will define the utterance into the coded values that have been defined by the researcher. 3. When the utterance data is transferred to the AWS cloud platform, it gets integrated into AWS Lambda which is a serverless compute service that enables the running of the question algorithms.

Figure 2: Summary of VOICE-COVID design and results [Graphical abstract]

Study population	Covid-19 screening process	Key results
• N= 215 participants • Mean age 46 years (± 17.7) **Female** 55% / **Male** 45% **English** 69% / **French** 31%	Step 1: Healthcare provider screening Step 2: Amazon Alexa screening **Screening questions** International travel in past 14 days Close contact with a person with Covid-19 without PPE Presence of symptoms: cough, fever, runny nose, sore throat, diarrhea	• N= 645 screening question were delivered to participants • Alexa mis-identified 1 response Primary outcome: Cohen's Kappa interrater reliability 0.989 (95% CI: 0.982, 0.997) Median time to completing Alexa screening 40 seconds (IQR 33, 44) Feedback (out of 5) Ease of understanding: 4.7 Ease of use: 4.5 Overall rating 4.4

CI confidence interval; IQR Interquartile range; PPE personal protective equipment

Table 1: Specific questions asked by the Amazon Alexa.

Baseline questions asked by the Amazon Alexa: 1. Which language do you prefer to continue in? You can select French or English. Dans quelle langue préférez-vous continuer? Vous pouvez sélectionner le français ou l'anglais. 2. Do you work at this hospital? 3. What is the purpose of your visit? You can indicate work, appointment, visit, delivery, pickup or other.
Screening questions for exposure to SARS-CoV2 1. Do you work at this hospital? 2. What is the purpose of your visit? You can say appointment, visitor, delivery, pick up or other* 3. Have you traveled outside of Canada in the last fourteen days? 4. Have you been in close contact with a confirmed case of COVID-19 without protective personal equipment? 5. Do you currently have any of the following symptoms: cough, fever, runny nose, sore throat, and diarrhea?

*If the participant works in the hospital, this question is skipped.

Table 2: Baseline characteristics of the participants

Demographics	N (%)
Number of subjects	215
Preferred language:	
English	149 (69.3%)
French	66 (30.7%)
Mean Age (SD) [Range], years	46.1 (17.7) [16, 84]
Male	97 (45.12%)
Work at this hospital	101 (47)
Purpose of visit:	
Work	101 (47.0%)
Appointment	76 (35.4%)
Visit	27 (13.6%)
Delivery	1 (0.5%)
Pick up	1 (0.4%)
Other	9 (4.2%)
Travelled outside of Canada in the last 14 days	5 (2.3%)
Have been in close contact with a confirmed case of COVID-19 without protective personal equipment	4 (1.9%)
Currently have any of the following symptoms: cough, fever, runny nose, sore throat, diarrhea	6 (2.8%)

SD standard deviation

Table 3: Post-screening survey

Questions	Rating					Mean (SD)
	1	**2**	**3**	**4**	**5**	
Did you feel the "check-in" appropriately asked the screening questions?	4 (1.86)	0 (0.00)	7 (3.26)	27 (12.56)	177 (82.33)	4.7 (0.70)
Did you feel the "check-in" minimized your chance of contracting COVID-19?	26 (12.09)	15 (6.98)	24 (11.16)	34 (15.81)	116 (53.95)	3.9 (1.42)
Was the Alexa language easy to understand?	0 (0.00)	5 (2.33)	12 (5.58)	30 (13.95)	168 (78.14)	4.7 (0.69)
Was the "check-in" easy to use	5 (2.33)	9 (4.19)	19 (8.84)	33 (15.35)	149 (69.30)	4.5 (0.98)
How would you rate this "check-in" overall?	4 (1.86)	2 (0.93)	21 (9.77)	67 (31.16)	121 (56.28)	4.5 (0.85)

Note, score of 5 reflects 'strongly agree' while the score of '1' represents 'strongly disagree'. SD standard deviation

Table 4: Challenges and solutions in the evaluation of voice-based devices

Challenges	Solutions
Accuracy in data collection has not been assessed for devices	Academic – industry collaborations to validate accuracy of devices for data collection
Privacy concerns regarding collection of data	Limit data collection and adhere to local and national data privacy laws
Data capture of voice-based devices in noisy environments	Field testing of device in real-world settings is critical before deployment
Multiple language skill development may remain challenging	Ensure multi-language support is available for the device and enabling validation/training of device algorithms for multiple languages
Need for continuous wireless internet access	Enable testing of possible 'offline' modes of device.

Author's contribution:

Abhinav Sharma MD, PhD: study design, analysis, manuscript writing
Emily Oulousian BSc: study conduct, critical revision for manuscript
Jiayi Ni MSc: study design, analysis
Renato Lopes MD, PhD: critical revision for manuscript
Matthew Pellan Cheng MD: manuscript writing, critical revision for manuscript
Julie Label BSc: study design

Filipe Henriques BSc: study design and analysis
Claudia Lighter: study conduct, critical revision for manuscript
Nadia Giannetti MD: study conduct, critical revision for manuscript
Robert Avram MD, MSc: study conduct, critical revision for manuscript

References:

1. Sharma A, Harrington RA, McClellan MB, et al. Using Digital Health Technology to Better Generate Evidence and Deliver Evidence-Based Care. *J Am Coll Cardiol*. 2018;71(23):2680-2690. doi:10.1016/j.jacc.2018.03.523
2. Sezgin E, Huang Y, Ramtekkar U, Lin S. Readiness for voice assistants to support healthcare delivery during a health crisis and pandemic. *npj Digit Med*. 2020;3(1):122. doi:10.1038/s41746-020-00332-0
3. Marquis-Gravel G, Roe MT, Turakhia MP, et al. Technology-Enabled Clinical Trials. *Circulation*. 2019;140(17):1426-1436. doi:10.1161/CIRCULATIONAHA.119.040798
4. Voice Assistant Use Through Smart Hearables Consumer Report - Voicebot.ai. https://voicebot.ai/2020/11/19/voice-assistant-use-through-smart-hearables-earbuds-has-doubled-since-2018-apple-airpods-extend-market-share-lead-new-report/. Accessed April 27, 2021.
5. McCaffrey M, Hayes P, Hobbs M, Wagner J. Prepare for the voice revolution. *PwC Consum Intell Ser*. 2018:1-12. https://www.pwc.com/us/en/advisory-services/publications/consumer-intelligence-series/pwc-voice-assistants.pdf.
6. Alexa Healthcare Skills. https://developer.amazon.com/en-US/blogs/alexa/alexa-skills-kit/2019/04/introducing-new-alexa-healthcare-skills. Accessed September 21, 2020.
7. Alexa Skills Kit Official Site. https://developer.amazon.com/en-US/alexa/alexa-skills-kit/frequently-asked-questions. Accessed September 21, 2020.
8. Whitelaw S, Mamas MA, Topol E, Van Spall HGC. Applications of digital technology in COVID-19 pandemic planning and response. *Lancet Digit Heal*. 2020;2(8):e435-e440. doi:10.1016/S2589-7500(20)30142-4
9. The VOICE-COVID-19 - ClinicalTrials.gov. https://clinicaltrials.gov/ct2/show/NCT04508972. Accessed September 21, 2020.
10. CANADA COVID-19 Self-Assessment Tool. https://ca.thrive.health/covid19/en. Accessed May 3, 2020.
11. What Is Automatic Speech Recognition? - Alexa Skills Kit Official Site. https://developer.amazon.com/en-US/alexa/alexa-skills-kit/asr. Accessed March 8, 2021.
12. Levine DM, Co Z, Newmark LP, et al. Design and testing of a mobile health application rating tool. *npj Digit Med*. 2020;3(1):1-7. doi:10.1038/s41746-020-0268-9
13. About Voice Interaction Models | Alexa Skills Kit. https://developer.amazon.com/en-US/docs/alexa/ask-overviews/voice-interaction-models.html. Accessed March 8, 2021.
14. Bengio Y, Janda R, Yu YW, et al. The need for privacy with public digital contact tracing during the COVID-19 pandemic. *Lancet Digit Heal*. 2020;0(0). doi:10.1016/S2589-7500(20)30133-3
15. Sana F, Isselbacher EM, Singh JP, Heist EK, Pathik B, Armoundas AA. Wearable Devices for Ambulatory Cardiac Monitoring: JACC State-of-the-Art Review. *J Am Coll Cardiol*. 2020;75(13):1582-1592. doi:10.1016/j.jacc.2020.01.046
16. Bhatia A, Maddox TM. Remote Patient Monitoring in Heart Failure: Factors for Clinical

Efficacy. *Int J Hear Fail*. 2021;3(1):31. doi:10.36628/ijhf.2020.0023

17.   Bayoumy K, Gaber M, Elshafeey A, et al. Smart wearable devices in cardiovascular care: where we are and how to move forward. *Nat Rev Cardiol*. March 2021. doi:10.1038/s41569-021-00522-7

18.   Yuan N, Pevnick JM, Botting PG, et al. Patient Use and Clinical Practice Patterns of Remote Cardiology Clinic Visits in the Era of COVID-19. *JAMA Netw Open*. 2021;4(4):e214157. doi:10.1001/jamanetworkopen.2021.4157