Feasibility and Acceptance of chatbots embedded in healthcare curricula:

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see cepeh.eu for more information

Acknowledgements

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Abstract

This document details the evaluation of each resource in terms of the feasibility and acceptance from the end-users. There was evidence of identifying the feasibility of such resources into formal training and studies exist on the acceptance of such resources, with promising results. However, all these studies defined the need for further research in the area until the use of chatbots in healthcare education became common. Furthermore, the creation process of CEPEH resources was significantly different and had improvements to current methods, due to the co-creation process, and use of low cost but effective technology.

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List of Abbreviations

CEPEH . . . . Chatbot Enhance Personalised European Healthcare curricula

RLO . . . . . . Reusable Learning Object

NLP . . . . . . Natural Language Processing

NLU . . . . . . Natural Language Understanding

A.I . . . . Artificial Intelligence

Introduction

Personalised Healthcare Education is needed to meet growing demand and quality maintenance. There is a growing evidence around chatbots, namely machine conversation systems- these programs have the potential to change the way students learn and search for information.

Chatbots can quiz existing knowledge, enable higher student engagement with a learning task, or support higher-order cognitive activities. In large-scale learning scenarios with a hight student-to-lecturer ratio, chatbots can help tackle the issue of individualized student support and facilitate personalised learning. However, limited examples of chatbots in European Healthcare Curricula have been utilised to combine both the continuum of cognitive processes presented in Bloom's taxonomy, with the idea that some repetitive tasks can be done with a chatbot- to provide greater access or to scale faculty time.

Thus, CEPEH strategic partnership has co-created open access chatbots utilising artificial intelligence, promoting innovative practices in digital era, by supporting current curricula and fostering open education.

CEPEH Erasmus+ strategic partnership aimed to co-design and implement new pedagogical approaches and, in particular, chatbots for European medical and nursing schools. CEPEH used use participatory design to engage stakeholders (students, healthcare workforce staff, lecturers, clinicians, etc.) in order to co-design effective chatbots and release them as open access resources. Through CEPEH, effective use of digital technologies and open education were be incorporated into healthcare curricula. This enabled students to increase their health and medical related skills through flexible learning.

Introduction

CEPEH expected that students adopted this new digital pedagogy and improve their skills and competences through flexible personalised learning, while the teaching staff enhanced their e-learning tool co-creation competences and make use of co-design best practices and recommendations for use. It is also expected increased cooperation between the partners. Thus, in the long term, CEPEH expects to influence the development of medical and nursing curricula with this digital innovation, foster the quality of the future healthcare workforce and further improve international competitiveness of the partners' healthcare curricula. This document details the evaluation of the resources created by the CEPEH team.

The evaluation specifically explored the feasibility and acceptance from the end-users. These end-users are learners in European healthcare higher education institutions.

There was firstly evidence for the need to identify the feasibility of chatbots and similar resources into formal education and training, with a further need to improve access to these types of learning resources. Of course, studies exist on the acceptance of chatbots, virtual patients, and many other healthcare applications, with promising results. However, through various limitations, we believed there was further research to be completed to accelerate the design, development, implementation, and evaluation processes. These have financial, stakeholder, time, and efficacy benefits. The creation process of CEPEH resources was significantly different to most in the literature, and this report highlights the approach of the CEPEH team towards enhancing personalised healthcare education can be achieved.

Background

The working practices of CEPEH are aimed at maximizing efficacy of these chatbots as learning resources, and provided a sense of shared development and ownership from all stakeholders. The process normally begins with workshops in which the project is scoped and team building occurs. The CEPEH workshops involve the widest possible team of stakeholders including tutors, students, healthcare

workers, learning technologists, health service users and carers- depending on the materials being created.

For readers who are interested in using these high quality digital resources please access them for free at CEPEH.EU

The next section will now present the evaluation of all CEPEH chatbot resources.

1

Method

Contents

1.1 Participants

This dataset had 14 males and 28 females therefore a total of 42 participants. It was a repeated measure design whereby each participant used the 4 chatbots developed by the CEPEH team. Therefore, there are 42 points of data in the condition before testing, and 126 data points after testing the chatbots- for a total of 168 row of data, 5 per participant. There were 78 questions asked in total, therefore the full dataset has over 4000 cells recorded.

Location and Profession of Participants

Figure 1.1: Location and Profession

There were 22 females and 7 males from Greece. There were 3 females and 4 males from Cyprus. There were 2 females and 2 males from Sweden, and there were 2 participants from the United Kingdom (see(1.1).

The majority 36 participants, were student, with 3 being learning technologists, 2 were lecturers, and 1 was a doctor. Although there could be a difference in these groups, the design was within- groups therefore each participants pre-usage metrics were the comparative control data, and participant differences did not affect the evaluation.

1.2 Procedure

For each resource created by the Partners, the same experimental methodology was followed. For each resource created by partners, students performed a study within an online or face to face workshop or course. Student participants joined

from Greece, Cyprus, Sweden, and the United Kingdom. A repeated measures design was used as the same group measures were taken before and after usage of the chatbots. They were recruited via staff members in the CEPEH group.

Participants were asked prior to the study if they agree to participate, providing them with a PIS form. Participants had the opportunity to discuss with the research team prior to the study and before consent is given. Then, participants used the chatbot resources independently and technical support was provided. Finally, post-intervention measures were recorded.

Some of the participants were invited to participate in Focus Group Discussions (FGD), and each FGD lasted between 15 to 25 minutes, with 5-10 participants. Participants were asked if they would like to be informed of the findings of the study.

1.3   Design

The data captured from the participants were their initials and numerical day of birth, used as anonymous identifier for pre-post analysis. Their institution was captured (Aristotle University of Thessaloniki, CYENS Centre of Excellent, Karolinska Institute, and The University of Nottingham), and Sex (Male/Female/Other).

Before any interaction with the learning resources, various perceptions of chatbot such as confidence and easy of use, usefulness, Influence from others, and current learning resources (videos, textbooks, Google, friends etc), were captured. Descriptive data was produced alongside repeated measures t-tests. Repeated measures t-tests were the appropriate test to use as this explores differences between groups, there were no covariates and we did not have several dependant variables. There was one Independent factor being Chatbot use having 2 levels (pre/post). There were 3 chatbots therefore there was option for ANOVA to determine where differences lie if statistical differences were found however this was not wholly appropriate for the data type and not necessary for pre-post comparison.

1.4   Materials and Measures

The measures used fit within a newly developed Chatbot Evaluation Framework-which takes the best measures of 5 previous frameworks. Denecke and Warren [2] derived several quality dimensions and attributes from previous chatbot literature. They formed six perspectives from their review of articles and mobile health applications.

These six perspectives were: 1) Task-oriented, 2) Artificial intelligence, 3) System quality perspective, 4) Linguistic perspective, 5) UX Perspective, 6) Healthcare quality perspective.

To capture these perspectives, we used several validated materials that can distinguish these elements of the CEPEH chatbots.

1.4.1   Chatbot Usability Questionnaire (CUQ)

The Chatbot Usability Questionnaire (CUQ) [4] is a new questionnaire specifically designed for measuring the usability of chatbots by an interdisciplinary team from the Ulster University. CUQ can be used alongside the prevalent System Usability Scale Score (SUS) [5]. Multiple metrics are more appropriate when measuring usability of chatbots [6] therefore a combination of two scores can provide an all-inclusive overview.

1.4.2   UTAUT2 (Unified Theory of Acceptance and Use of Technology)

The underpinning theory of the UTAUT2 is that there are four key constructs to the intentions of using technology based resources: 1) performance expectancy, 2) effort expectancy, 3) social influence, and 4) enabling conditions.

The TAM and the UTAUT2 have cross over in measuring technology acceptance, however the UTAUT2 has more applied probing questions. Few studies exist that use technology acceptance theories for the intention to use products that explicitly incorporate AI. A recent extension of the UTAUT2 model added five (health,

convenience comfort, sustainability, safety, security, and personal innovativeness) additional influencing factors to accommodate for AI [7]. This can be used for products in either health, household use, or mobility and can help to explain behavioural intention and use behaviour of chatbots.

1.4.3 System Usability Scale

The System Usability Scale (SUS) was used [10] and is a widely used and adopted usability questionnaire. It is popular due to its unbiased and agnostic properties, a non proprietary, and a quick scale of 10 questions. However, as there are the CUQ and parts of the UTAUT2 we have selected only 2 questions which do not cross-over with the other measures. These are improtant statements however and good indicators of usability when assessed with the other results.

1.4.4 Computer Self-Efficacy Scale Tool

The 10 question CSEST is based on the 32-item questionnaire by Murphy, Coover, and Owen (1989). It can be adapted for any technology and we have selected only a few pertinent questions. Participants are asked to think about using the CEPEH chatbots and answer that they would use the chatbots if I had never used a product like it before; If they could call someone for help if I got stuck, or if someone showed them how to do it first, and other similar usage questions.

1.4.5 Technology Acceptance Model (TAM)

The Technology Acceptance Model (TAM) [1] was specifically developed with the primary aim of identifying the determinants involved in computer acceptance in general; secondly, to examine a variety of information technology usage behaviours; and thirdly, to provide a parsimonious theoretical explanatory model. TAM suggests that attitude would be a direct predictor of the intention to use technology, which in turn would predict the actual usage of the technology. The only modification to the nine sub-scales of the questionnaire consists of applying the items to the context of chatbots. All the items, except those measuring attitudes, utilize a

seven-point Likert scale ranging from "strongly agree" to "strongly disagree" with a middle neutral point [2].

The nine sub-scales of the questionnaire:

Ease of use of chatbots Perceived usefulness of chatbots Intention of use. Attitude toward usage of chatbots. Perception of personal efficacy to use a chatbot resource. Perception of external control toward chatbots. Anxiety toward chatbot use. Intrinsic motivation to use chatbot resources. Perceived costs of chatbots.

1.4.6   Qualitative Measure- Focus Group Discussions

Focus groups are a pervasive means of market research and provides credible acceptance evaluators regarding the penetration that a product or service will have on a target demographic.Focus groups are a form of qualitative research consisting of interviews or structured discussions, in which a group of people are asked about their perceptions, opinions, beliefs, and attitudes towards a product, service, concept, advertisement, idea, or packaging.

Questions are asked in an interactive group setting where participants are free to talk with other group members.During this process, the researcher either takes notes or records the vital points he or she is getting from the group. Researchers select members of the focus group carefully for effective and authoritative responses.Relevant stakeholders, then, can use the information collected through focus groups to receive insights on a specific product, issue, or topic focus [7].

A series of short focus group sessions identified the feasibility of CEPEH resources for formal curricular integration.These sessions, spanning no more than 1-1.5 hours and consisting of no more than 5-7 persons each explored all axes of curricular integration such as accessibility in the classroom, use case scenarios, technology requirements for curricular integration etc.These axes were formalized by the research team, in each evaluation site, to consider the curricular details of each institution.

1. Method

Figure 1: Flow diagram of the recruitment process

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Results

Contents

2.1 Learner Characteristics

When asked the amount of time participants have used a chatbot- in any form or subject: 23 stated they had never used a chatbot, educational or not. Further, 18/42 having used a chatbot at least once for between 0-4 hours of use in total. Two individuals had spent much longer time with usage- these were a learning technologist associated with the CEPEH team, and a mature student. In short, this sample of participants did not regularly use chatbots for their course learning, but 1/3 had used one in some form previously.

Most learners use books or online books as resources. They may use multiple sources however they were asked to note the primary source. Only 6 stated their primary sources were *Online videos/interactive materials* which includes such tools as chatbots.

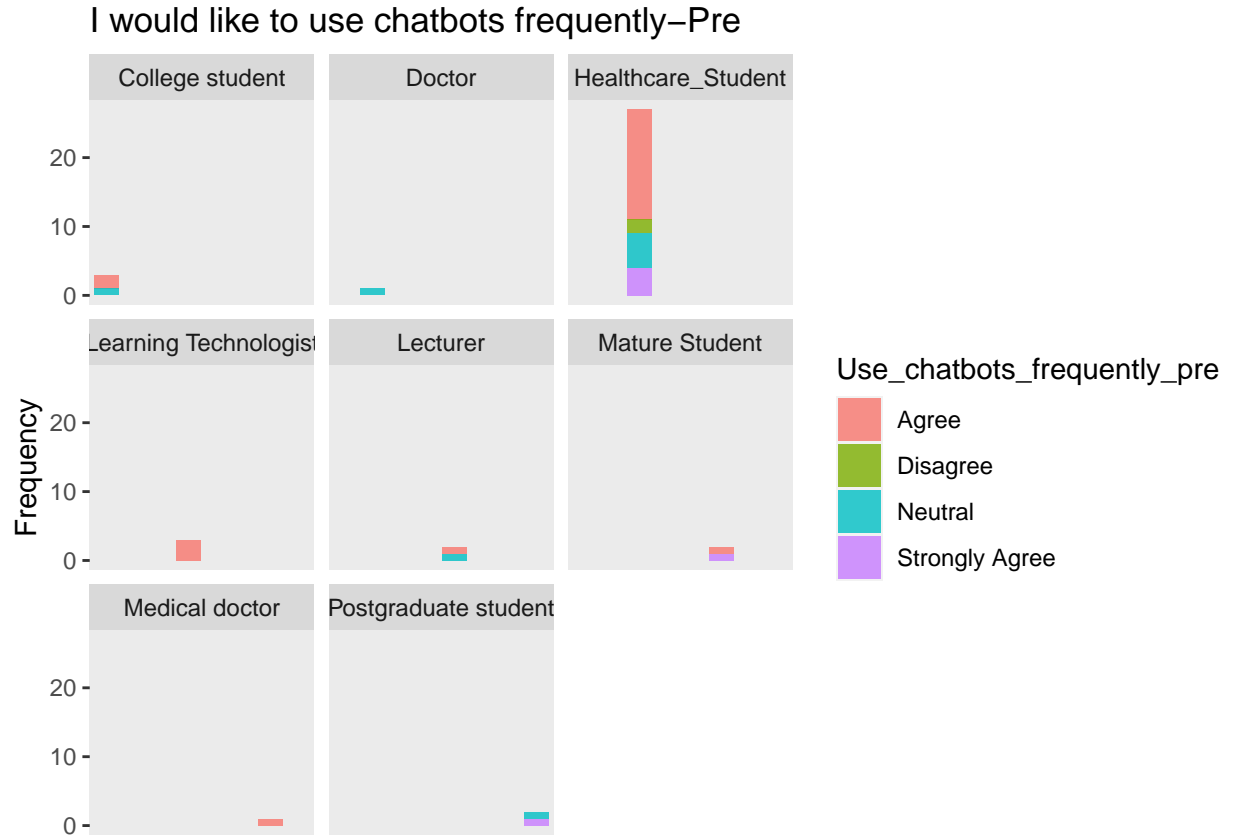## I would like to use chatbots frequently–Pre

Figure 2.1: Usage Frequency-Pre

The first boxplot (2.1) shows learners perceptions of easy of use of mobile app and other educational mobile resources

Previous_Chatbot_Usage	n
1-4 hours	15
10-19 hours	1
20+ hours	1
5-9 hours	2
Never	23

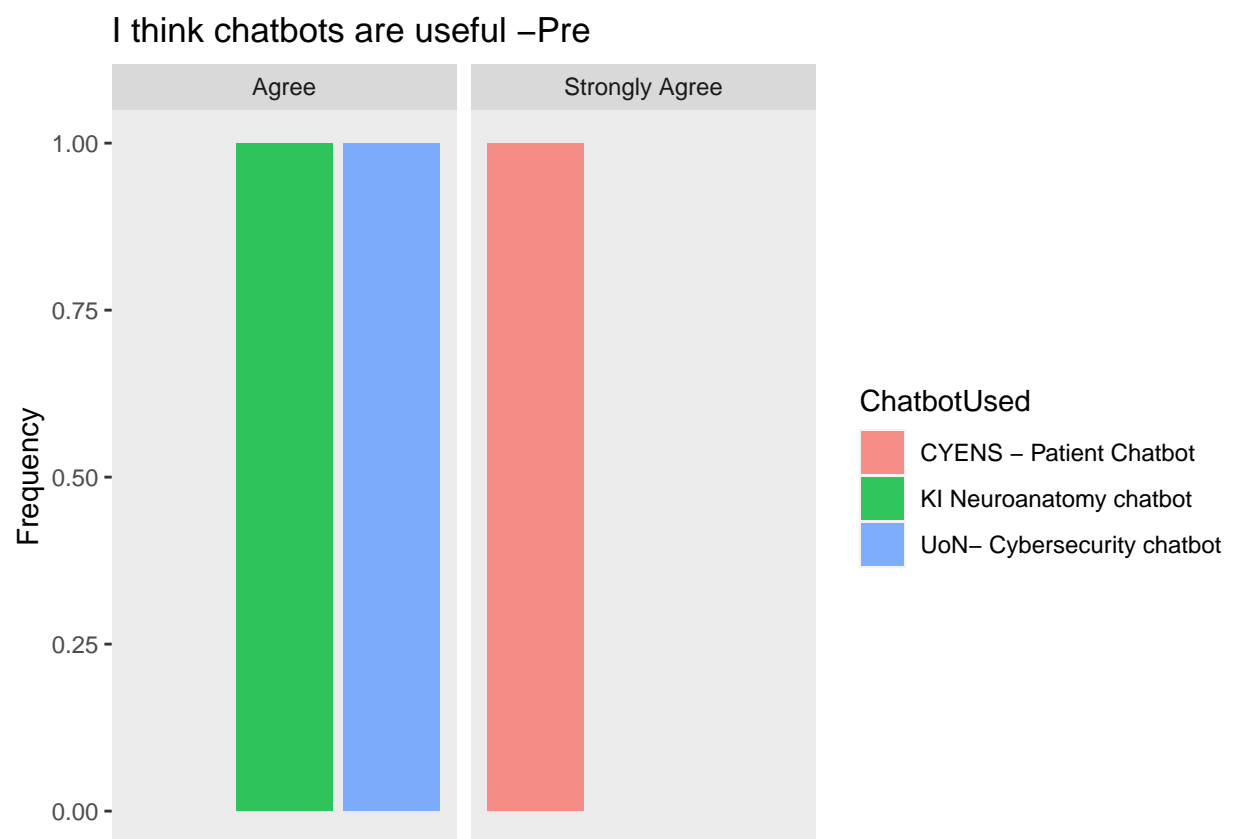I think chatbots are useful –Pre

Figure 2.2: Useful Opinion-Pre

3

Post-Intervention Results and Comparision

3.1 System Usability Scale (SUS) Scores

Note= The amount of 'agreement' is defined as the addition of 'Agree' and 'Strongly agree' responses.

The SUS score should consist of 10 items. However, some SUS questions were improved upon by 1 or more CUQ questions, specifically to this Chatbot study. The SUS results would be overshadowed by the CUQ scores, expect 2 that did not have cross-over. The two questions were:

- I would like to use the CEPEH chatbot I tested, more frequently (SUS1)(post)
- I felt confident using the CEPEH chatbot (SUS2)(post)

This meant the score of the SUS was not created, however the CUQ score better represented the Learners' perceptions of the CEPEH chatbot in terms of feasibility of reuse and acceptability in healthcare curricula.

Keep_Using_Chatbots	Confident	Count
Agree	Agree	44
Agree	Disagree	5

Keep_Using_Chatbots	Confident	Count
Agree	Neutral	11
Agree	Strongly Agree	6
Disagree	Agree	6
Disagree	Disagree	5
Disagree	Neutral	4
Neutral	Agree	10
Neutral	Disagree	1
Neutral	Neutral	6
Not Applicable	Not Applicable	3
Strongly Agree	Agree	10
Strongly Agree	Not Applicable	1
Strongly Agree	Strongly Agree	12
Strongly Disagree	Agree	1
Strongly Disagree	Strongly Agree	1

3.2 Technology Acceptance Model

The TAM had 3 sections (Ease of Use, Perceived Usefulness, and Intention of Use). Ease of Use results showed significant increases in Users' usage with each Chatbot. Perceived Usefulness: There were not significant findings for the Perceived usefulness. The justification for this may be due to being early versions of applications with limited functionality and functions which can be difficult for user to experience the intended further range of features and learning exercises.

Intention of Use: For users' intentions to use within their course, the result of the Mann-Whitney U test was not significant, U = , z = , p = . in their intentions before use (m=xx, mode=xx) compared to after (m=xx, mode=x), however there was improvement therefore the chatbots may have more benefit than expected by students.

3.3 Chatbot Usabilty Questionanire (CUQ)

3.3.1 CUQ Calcuation tool

The CUQ was developed by researchers at Ulster University,Link and as the calculation can be complex a dedicated calculation tool has been created. Please

3. Post-Intervention Results and Comparision

Chatbot Usability Questionnaire Results

	Results				Mean Question Scores	
					Q	**Score**
Chatbot	CEPEH chatbots				1	3.5 ± 0.9
					2	3.0 ± 0.9
Participants	160				3	3.9 ± 0.7
					4	2.1 ± 0.9
CUQ Score	65.2±11.4				5	3.9 ± 0.8
					6	2.3 ± 0.9
Lowest Score	28.1	**Participant**	24		7	3.8 ± 0.8
					8	2.5 ± 0.9
Highest Score	96.9	**Participant**	29		9	3.4 ± 0.9
					10	2.8 ± 0.9
Median Score	65.6	**Participant**	12		11	3.8 ± 0.7
					12	2.5 ± 0.9
					13	3.4 ± 0.8
					14	2.6 ± 0.8
					15	4.0 ± 0.7
					16	2.1 ± 0.7

This is the results page. Mean CUQ score, lowest, highest and median scores are above. Mean scores per question are on the right. **Note: It is normal for Median Score participant to be listed as N/A if you have an even number of participants!**
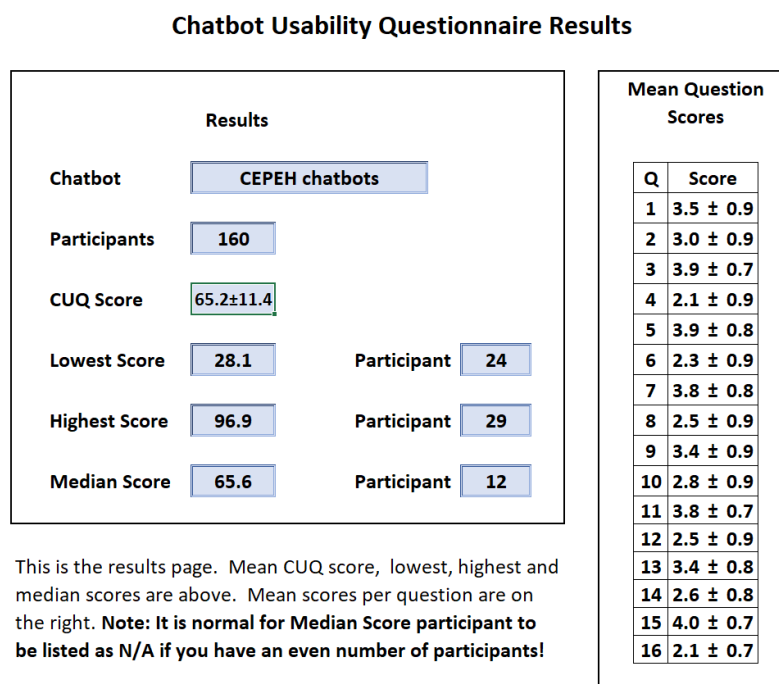
Figure 3.1: A marvel-lous meme
(#fig:cuq image)

download the CEPEH CUQ calculation tool which has all of the data entered, so you can see the CEPEH CUQ scoring.

click here to download CUQ calc tool

click here to download CUQ score image *mobile download disabled

The score for all 3 chatbots grouped was 65.2/100, This scoring system was designed to be comparable to SUS and may be freely used alongside it, or in combination with other usability metrics. There has been evidence of correlation of 76% between the CUQ and SUS therefore we expect the SUS scored to be between 48.75 and 81%. We believe the CUQ has more validity towards measuring the concepts of interest on this study.

Read the CUQ development paper, see page 3 for correlation

Figure shows the CUQ scores as a box plot to highlight the range of Usability of the resources. Further exploration is required to understand which elements are causing this spread.

My Knowledge of the Topic Improved after Use
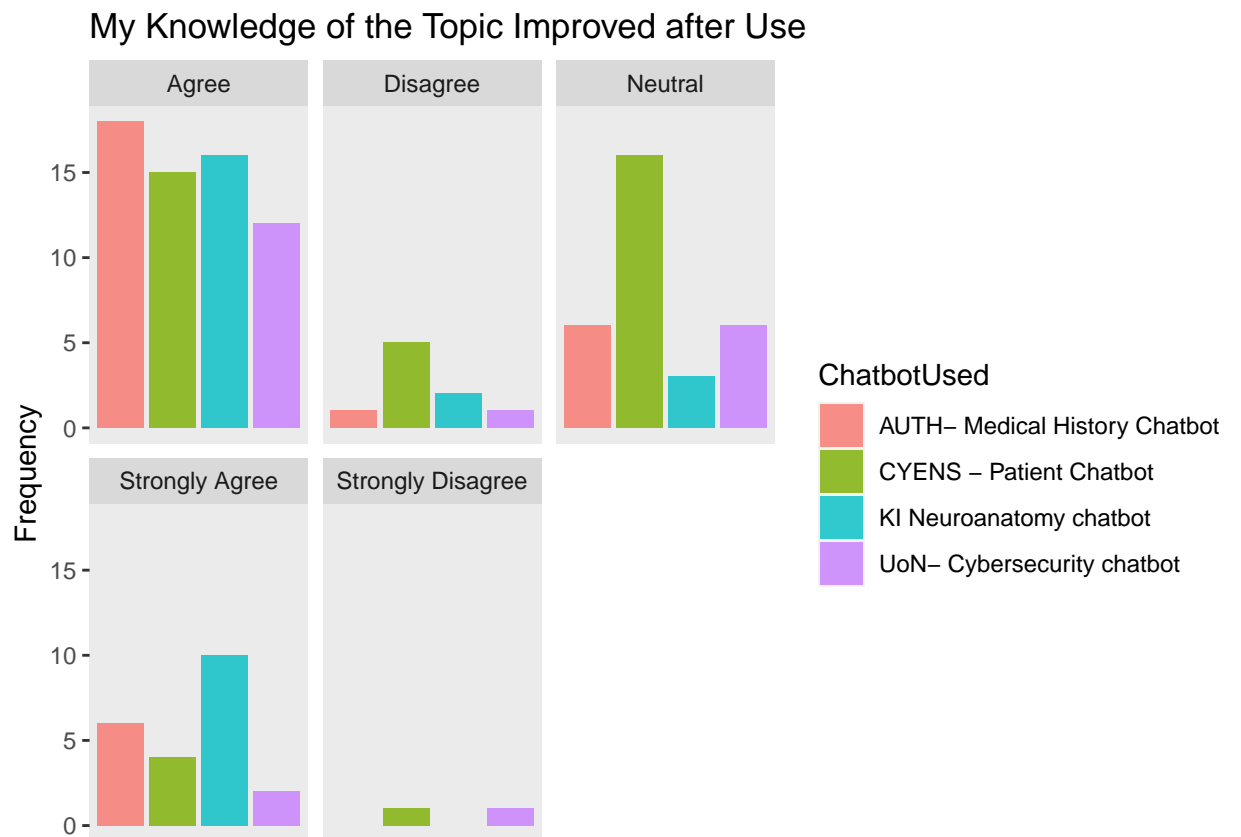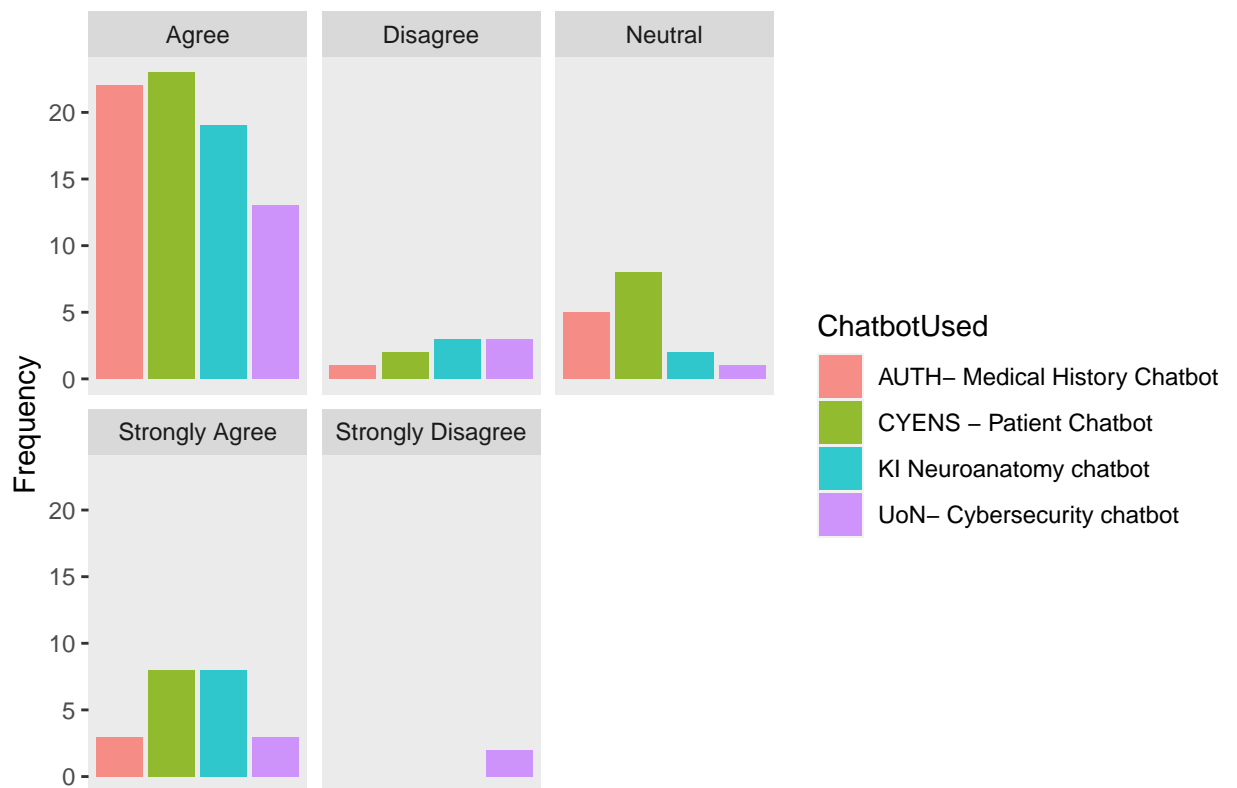

CYENS chatbot had around 10 more participants stating that they were neutral

on gaining knowledge of the topic

I Trust CEPEH Chatbots to Provide me with my Course Information

CEPEH Chatbot Personailty was Realistic and Engaging
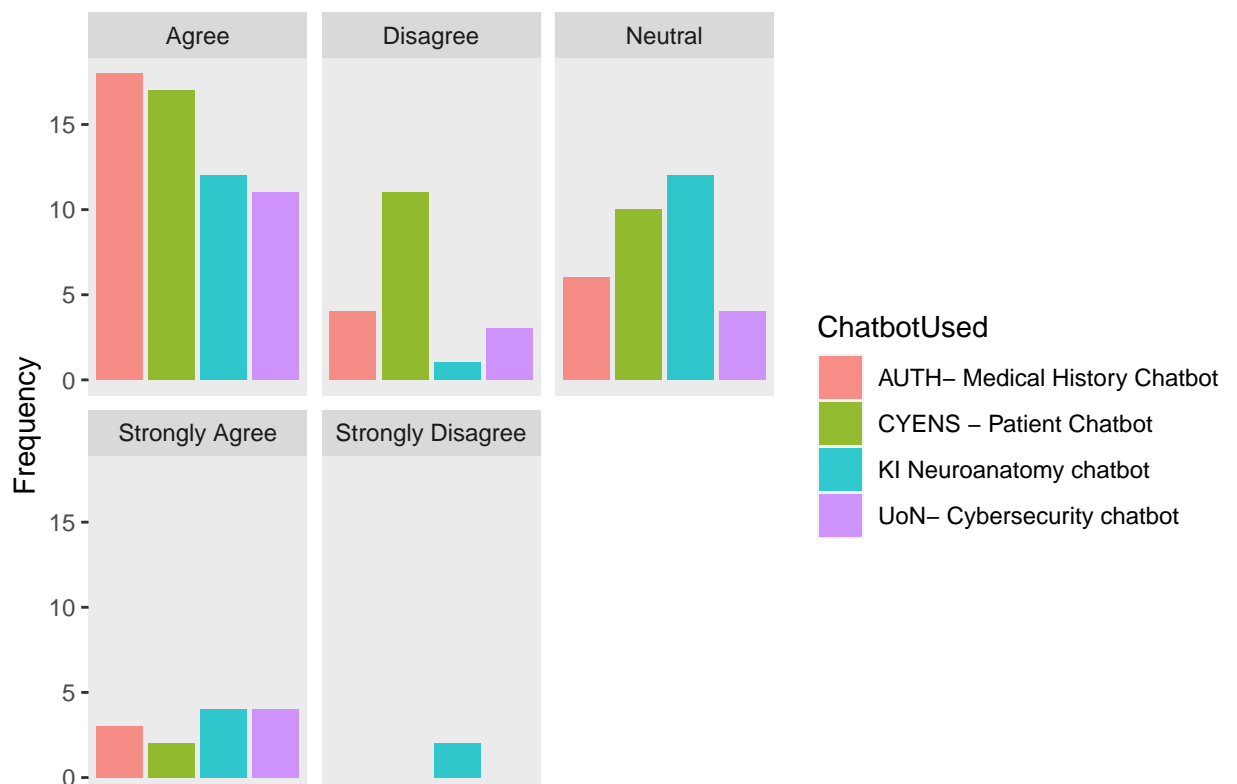
3. Post-Intervention Results and Comparision

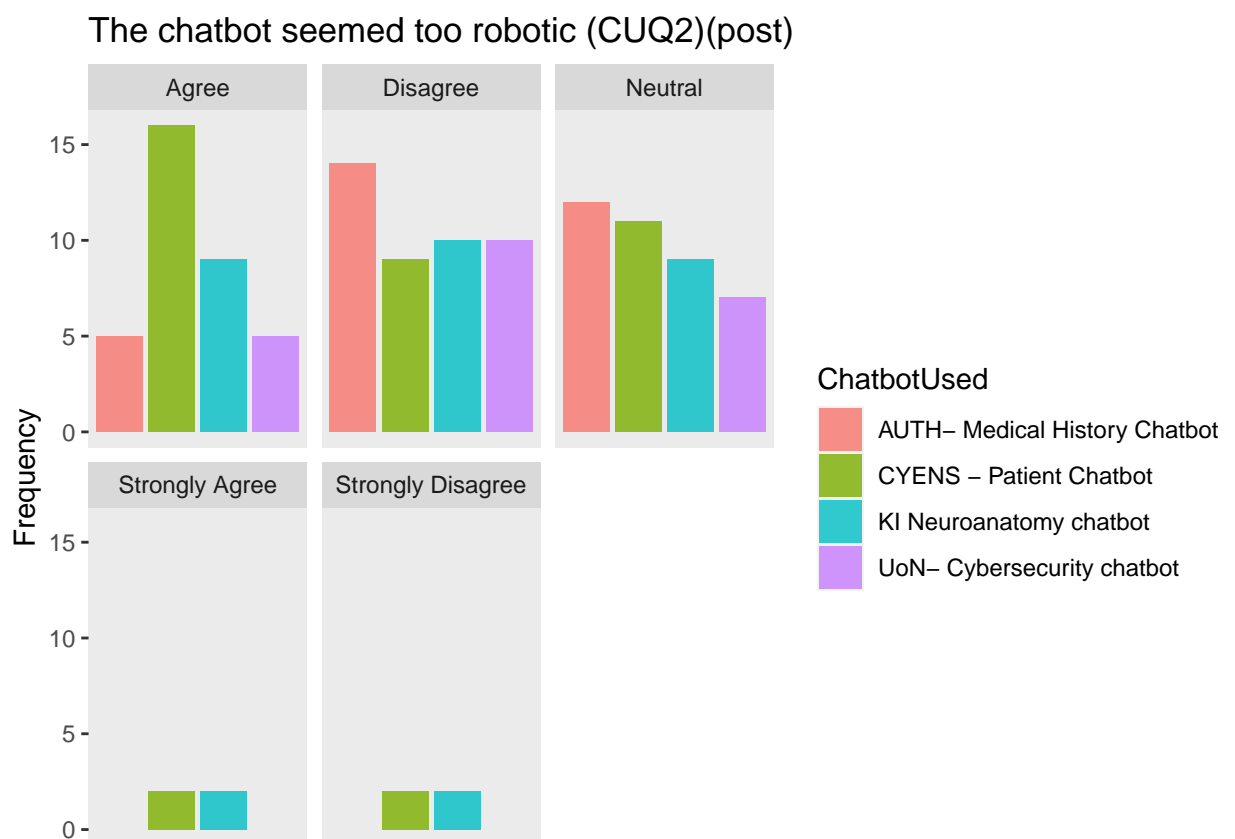There was mixed results for the chatbot used being realistic and engaging. This question has two descriptive terms however based on the other results we understand that the chatbot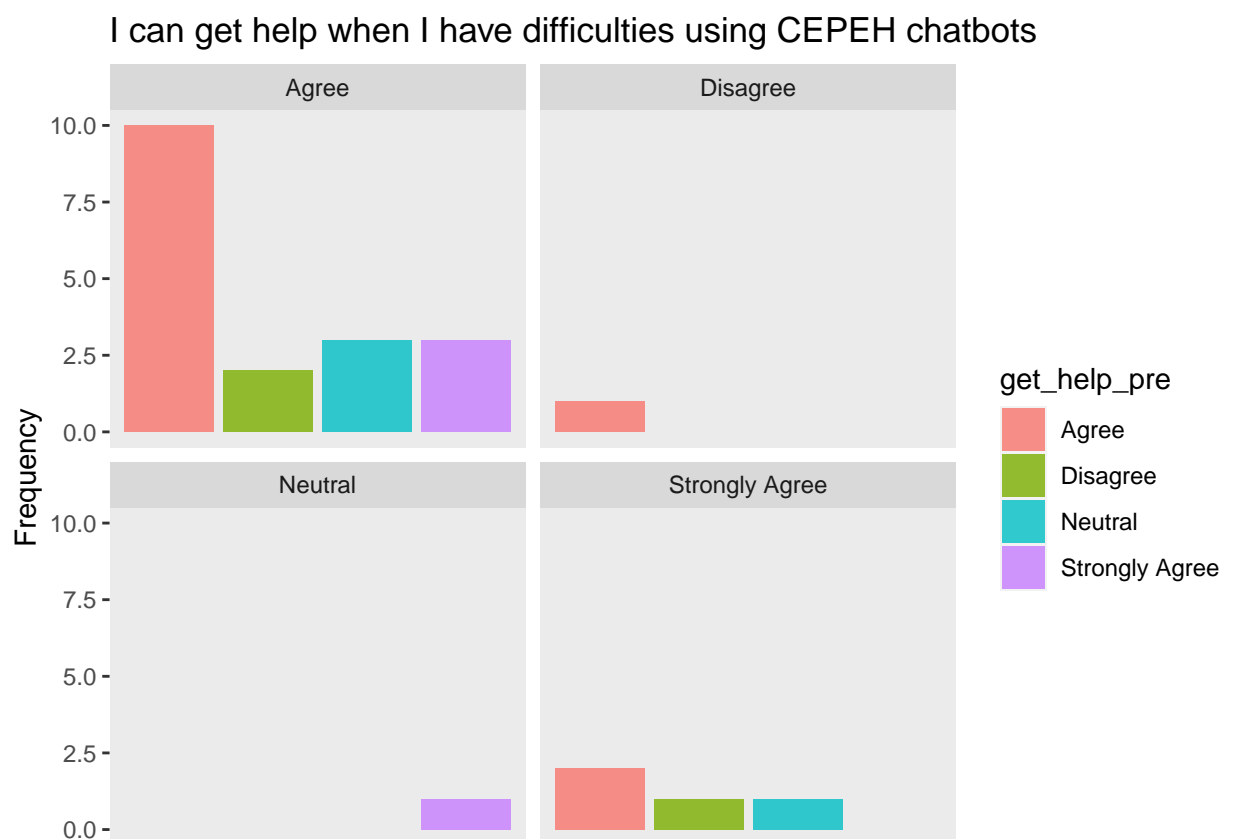s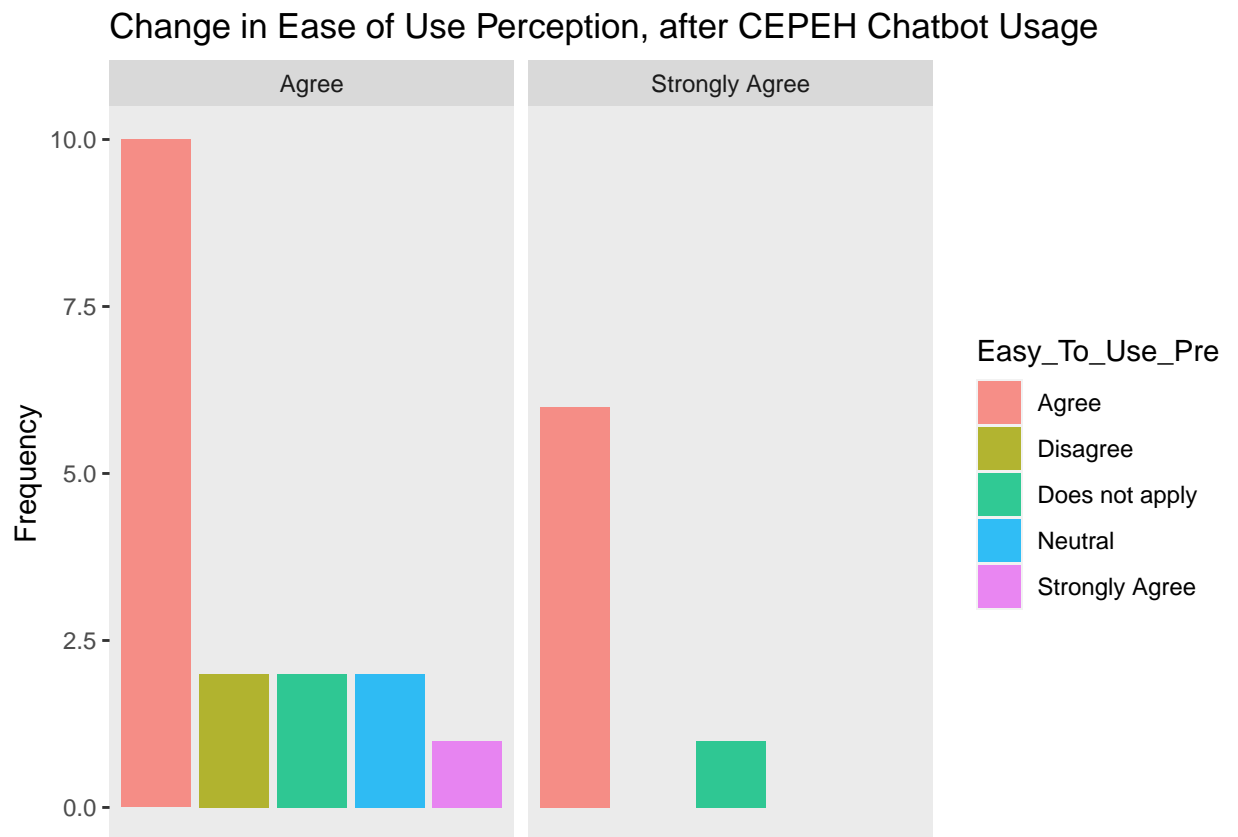' NLP logic, or ability to respond required improvement to be more 'smooth' in replying. The primary limitation was found in the 'robotic' interactions(See Figure 10). This was investigated further in the 'Text Mining' and 'Sentiment Analysis' sections.

Change in Ease of Use Perception, after CEPEH Chatbot Usage

I can get help when I have difficulties using CEPEH chatbots

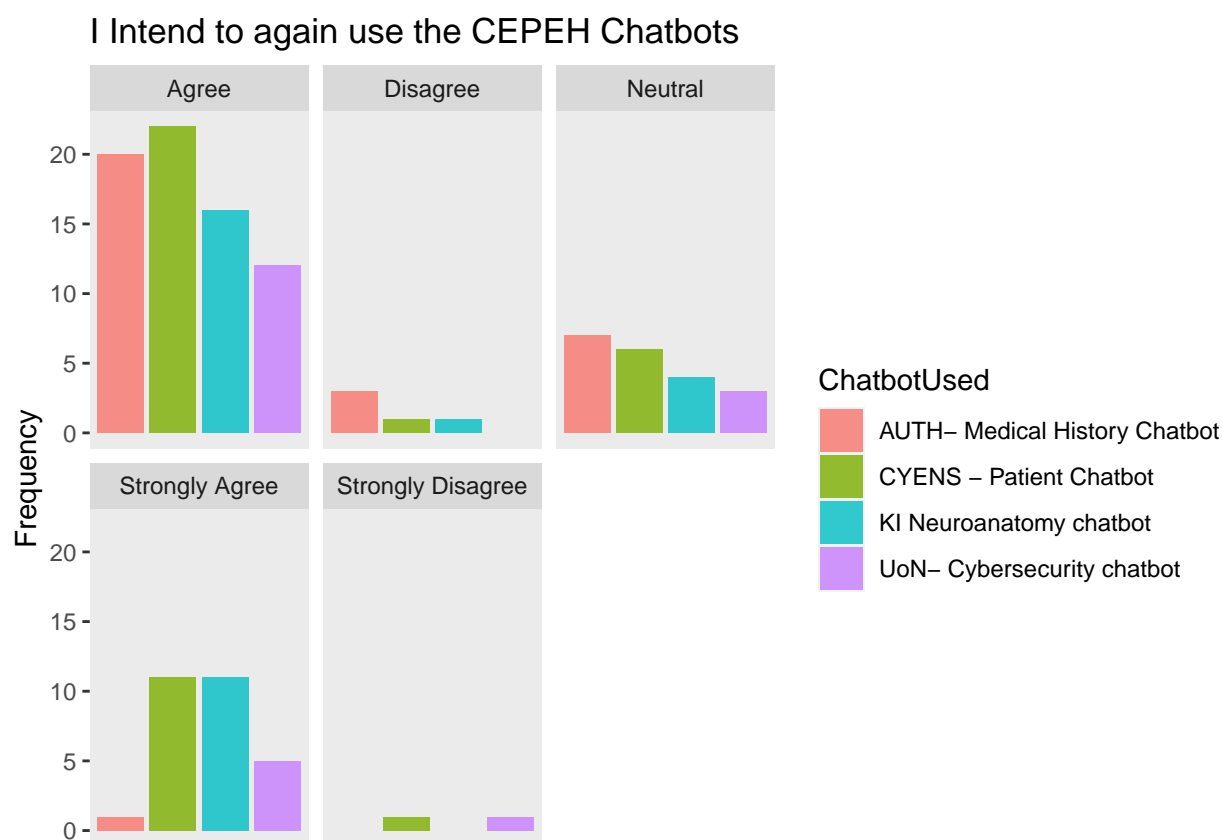## I Intend to again use the CEPEH Chatbots

Figure 3.2: Intend to Reuse-Post

Those who disagreed or were neutral in the pre usage measure, improved their understanding that help was available with the CEPEH chatbots. After usage, 40 participants agreed they could get help if they had difficulty using the resources.


3.4 Inferential Statistics

3.4.1 Repeated Measures T-test results

After using the CEPEH chatbots, majority of participants stated they would reuse the chatbots. However, there was 6 counts of *disagree* or *strongly disagree* for all 4 chatbots. Further, there were 17 counts of neutral in reuse, which was approximately 4 participants per chatbot (see (3.2).

3. Post-Intervention Results and Comparision

For CYENS, even though the knowledge of the topic was not perceived to improve by some participants, this box plot shows how 34/42 stated they would reuse the chatbot developed by CYENS.
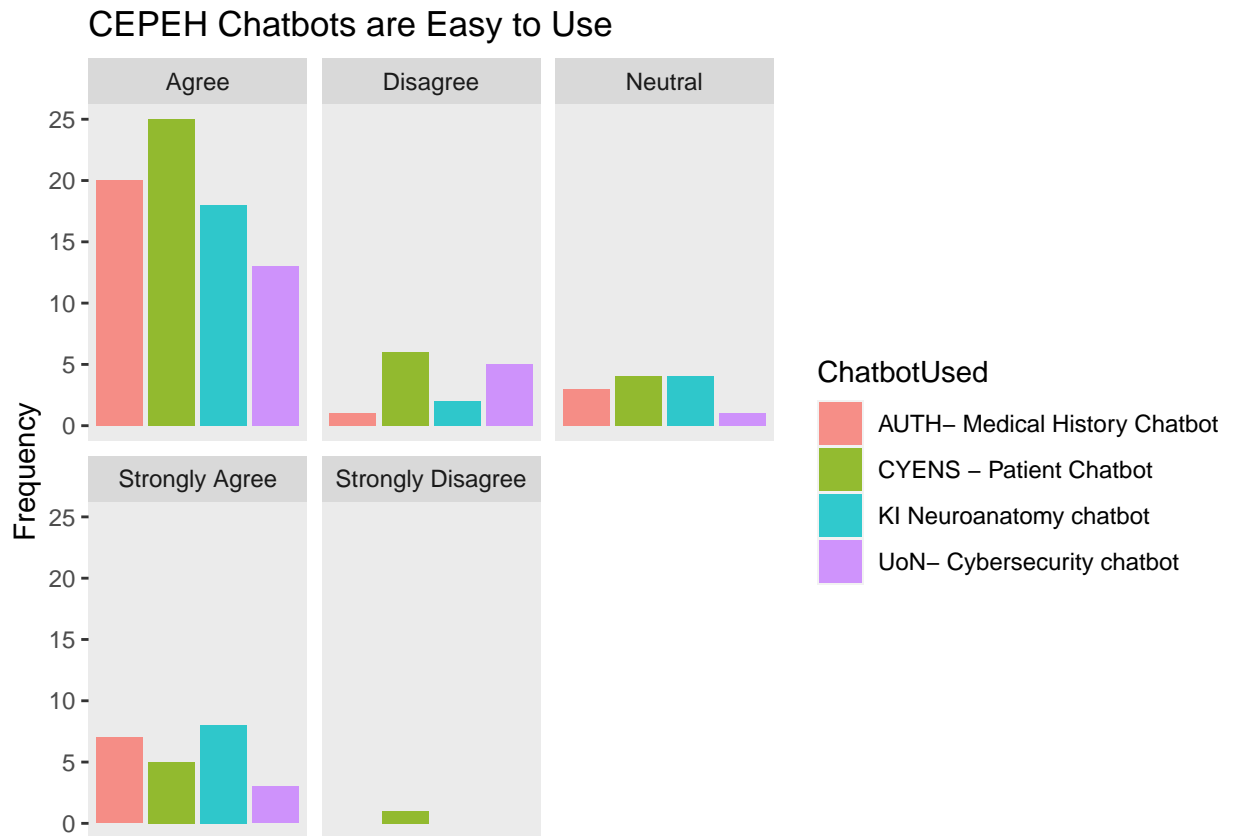
There was only 1 'Strongly Disagree' response. The agreement options counted for the majority of the data.

3.4.2   Other Findings

Other questions

I intend to continue using chatbots in the future (BI1)

The chatbot provided the information I needed with minimal commands

My knowledge of the topic improved after i had used the Chatbot

My confidence in understanding the topic improved after I had used the Chatbot

The chatbot provided me with the type of response i expected from asking

a tutor/lecturer

3. Post-Intervention Results and Comparision

The information provided was reliable

The chatbot has a high level of trustworthiness

The duration of conversations to find my answer was too long

The videos/images provided were useful to my questions

The chatbot exceeded my expectation of how it could help me

The chatbot exceeded my expectation of how it could engage with me

I think this learning method could help me to acquire knowledge

I would use this tool again as it has some value to me

I think i will actively use this learning method

I believe I had some choice about learning during chatbot use

I would trust the chatbot to provide me with information for my course

One piece of knowledge i learned from the chatbot was..

Repeated Measures t-test, aka paired t-test (before and after measurements)

This t-test compares confident using mobile chatbots before and after CEPEH chatbot usage.

Appendix

3.5 CEPEH Qualatative Feedback

The focus group discussions provided a lot of feedback for how the participants experienced their interactions with the chatbots, and how the CEPEH team can improve them, improve the design and development processes, and improve uptake and sharing.

One method of analysing this data is with use of text mining and data manipulation, creating word clouds, sentiment analysis, and using a model which can distinguish the unique themes in text, and highlights for us what text is used to create these themes.

Therefore, we have created a model to allow efficient and intelligent analysis of this open/free focus group data.

3.6 Tokenising

Firstly, we tokenised the words from the FGDs. A Token is "a meaningful unit of text, most often a word, that we are interested in using for further analysis". For each word we give it a property that we can call upon later.

The data manipulation for this included removing punctuation, converting to lower-case, and setting word type to word (and not such types as "characters", "ngrams", "sentences", "lines" etc)

3.6.1 Stop words

The model then removed words with meaningless function. These are called stop words. Words like "the", "of" and "to" are the most frequent words found, tech-

nically, but are of little interest to us.

We also created a custom list of stop words for CEPEH. We know participants may mention other objects, and the list was as followed: found; chatbot; chatbots; presentation.

The data was ready for analysis by the model. We ordered it to find the most frequent words.

```
## # A tibble: 384 x 3
## # Groups:   doc_id [1]
##    doc_id word            n
##    <fct>  <chr>       <int>
##  1 1      information    11
##  2 1      helpful        8
##  3 1      understand     8
##  4 1      idea           7
##  5 1      ideas          7
##  6 1      workshop       7
##  7 1      beginning      6
##  8 1      dont           6
##  9 1      medical        6
## 10 1      technology     6
## # ... with 374 more rows
```

3.7 Plotting word frequencies - bar graphs

With this information a Bar graph of top words from the participants in the FGD can be rendered.
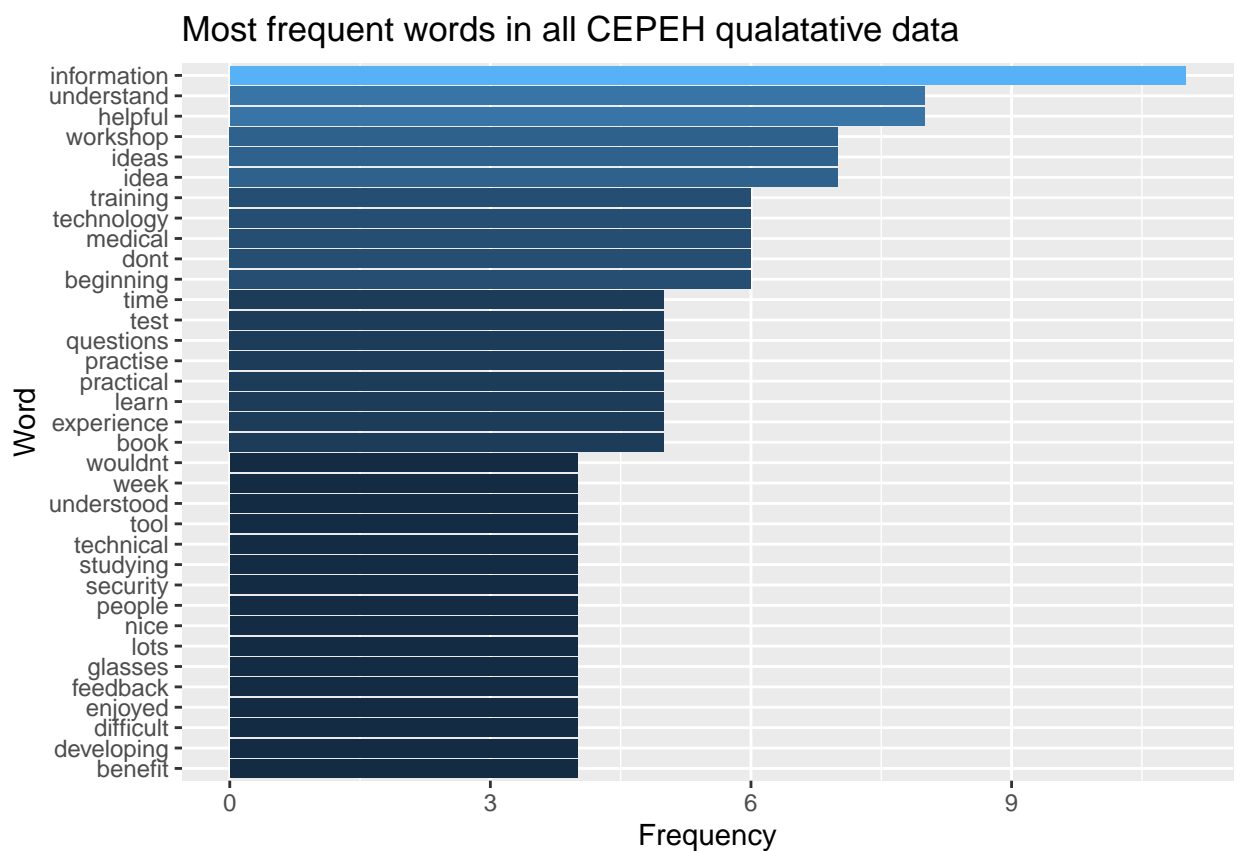
3. Post-Intervention Results and Comparision
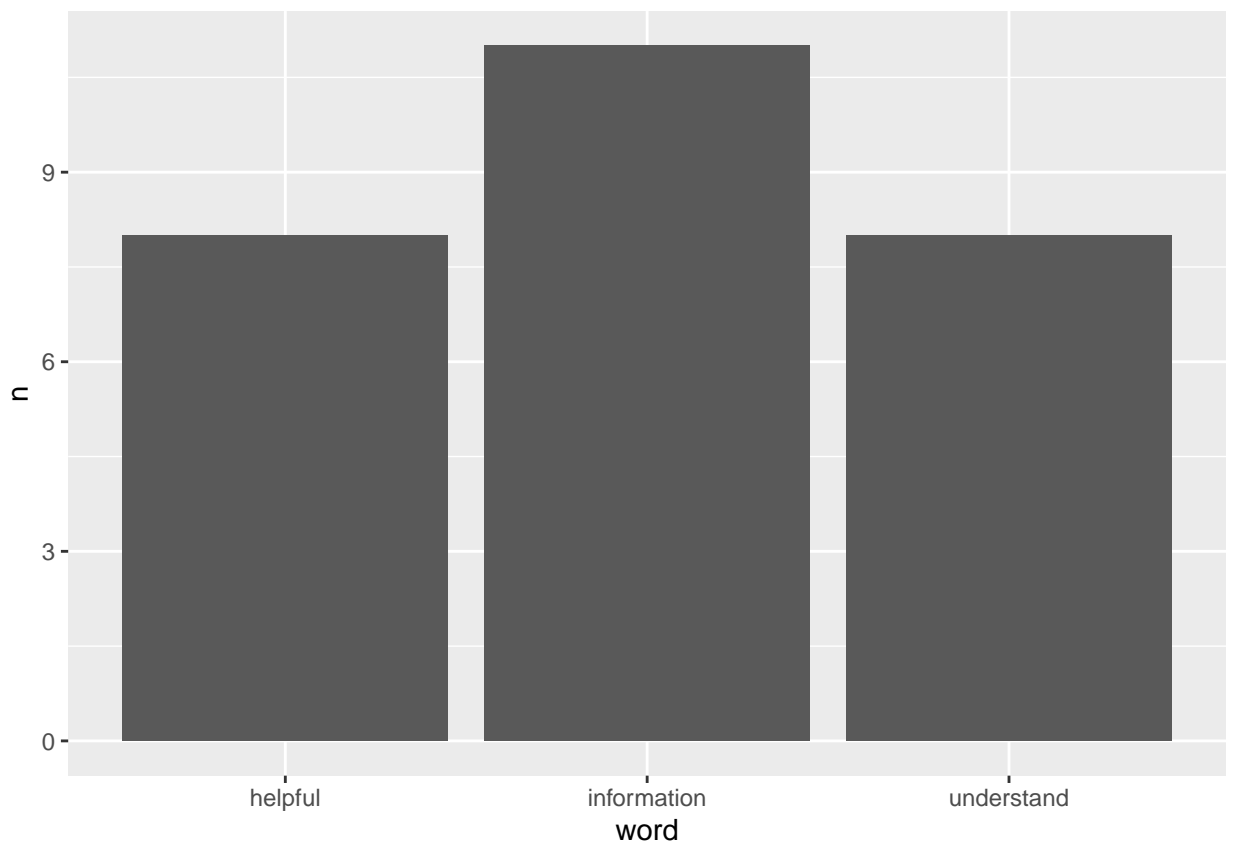
and after some modifications, a graph of the top 35 words is produced, with

better aesthetics.The most frequent words present in focus group discussions after

using the 4 chatbots, are in the Figure below.

Most frequent words in all CEPEH qualatative data

Although the frequency is not high for each word, we are able to get a general picture of the sentiments, intensities, and concerns which would be immediately occurring when plotted.

3.7.1 Normalised frequency

A better way to understand this data is to normalise the frequency of occurrences in accordance with the source text. The raw text had 2827 words in total. Therefore we can mutate the ratios to reflect this.

Plotting normalised frequency

Now we can plot, for example, the 20 most frequent words when normalised by the source text.

Most frequent words in CEPEH FGDs
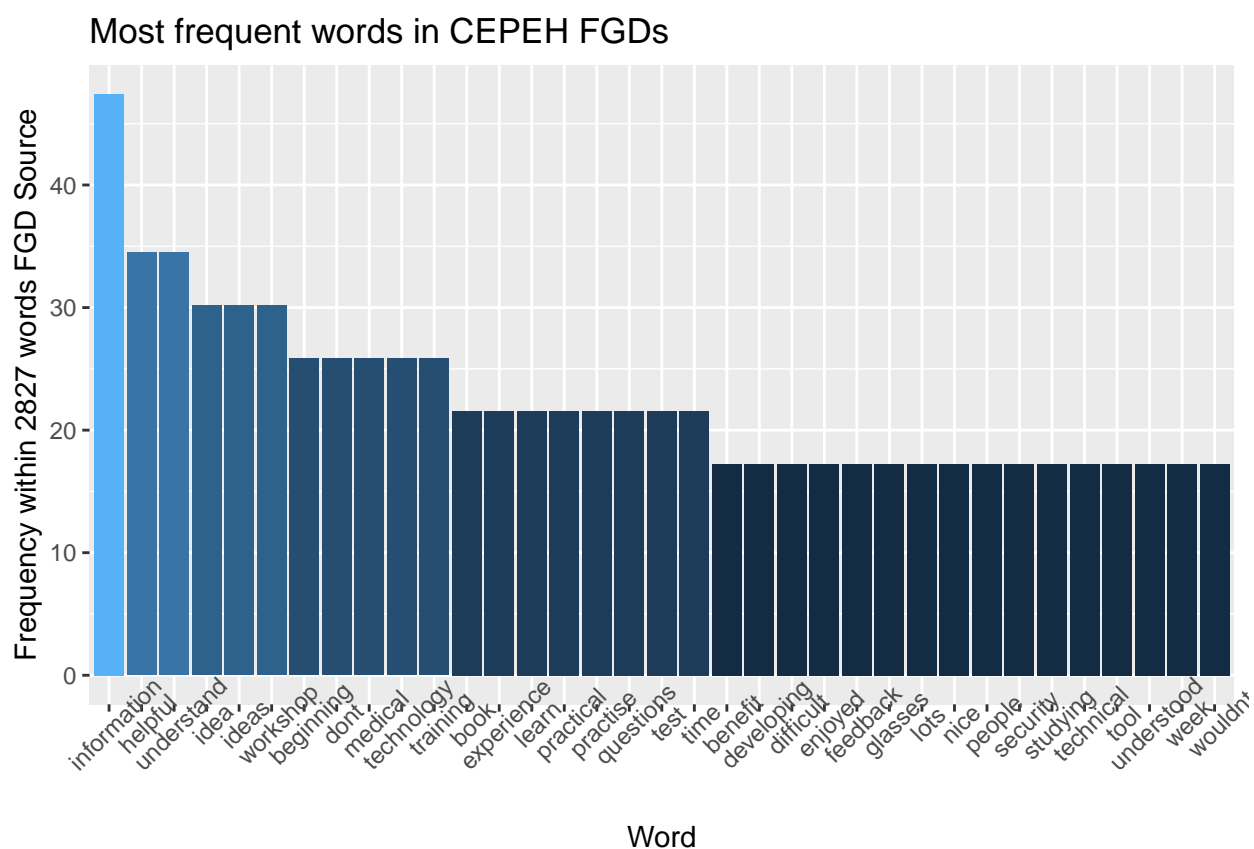
In summary, this understanding of frequent words can help to understand common concurrences and extrapolate to a larger audience. If scope and impact of CEPEH chatbots increased we can understand the type of themes and trends may occur, based on such FGD analysis.

3.8 Word clouds

To visualise the most frequent words in another format, below is a word cloud which presents the word size to indicate the frequency- words that occur more often being displayed in a larger font size. This has a normalised data frequency in accordance to the FGD source document analysed.
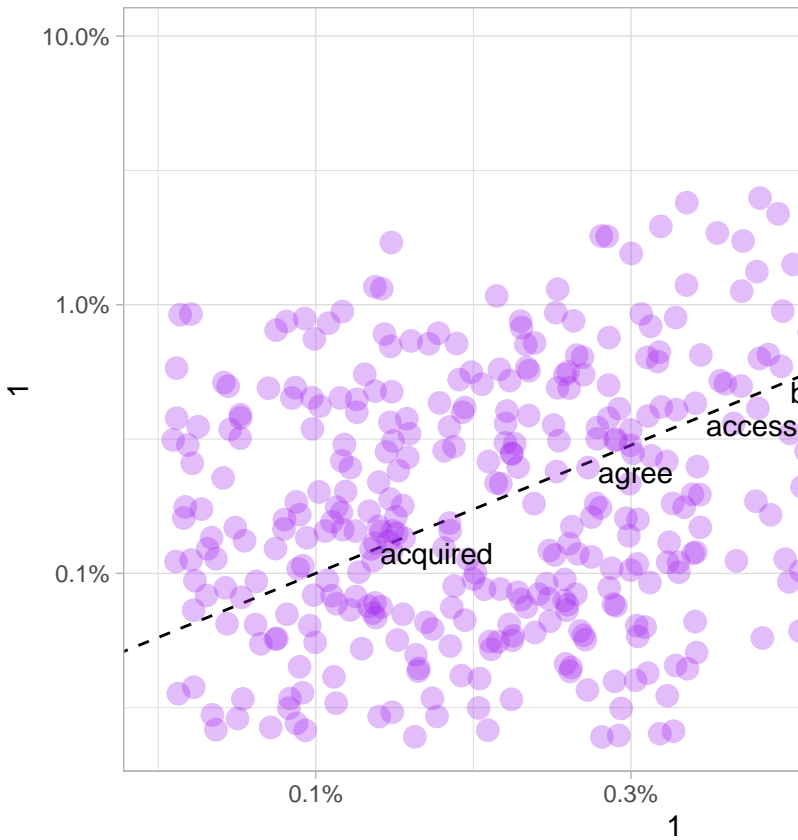
We understand the context has been reduced for each word. However, in general there can be categorised positive/negative words from the word cloud: Positive words are- benefit, practical, nice, helpful, learn, ideas, and enjoyed Negative words are- difficult, test (who likes a test?), don't, and 'lot' may be negative if there is a 'lot' of information.

3.8.1 The vocabulary of Texts

Here is a graph that has plotted the words in places depending on the word frequencies. Additionally, colour hotspots shows how different the frequencies are - darker items are more similar in terms of their frequencies, lighter-coloured ones more fre-

quent in one text compared to the other.

3.9 Sentiment analysis

What is the sentiment of all participants? What is types of emotional words are being used? The preparation of these words has some use in understanding the frequencies, but their emotional valence are not compared. The table above has the word *'helpful'* which has a positive connotation, however there are 386 words, with many having several occurrences.

```
##   max(total_score) min(total_score)
## 1               38               38
```

negative	positive	total_score
24	62	38

4

Discussion

Contents

##Summary of Findings

The Chatbots were beneficial.

Learners have lots of other choices such as YouTube, but there is a certain need for personalised information gathering , this can save time and prevent learning incorrect information.

This was one reason why they were rated positive as they are able to streamline data finding for learners in a format that is understandable and easy to them.

4.1 Quantatative Results

4.2 Qualatative Results

4.3 Limitations

4.4 Conclusions

The usual way to include citations in an *R Markdown* document is to put references in a plain text file with the extension **.bib**, in **BibTex** format.[1] Then reference the path to this file in **index.Rmd**'s YAML header with `bibliography: example.bib`.

Most reference managers can create a .bib file with you references automatically. However, the **by far** best reference manager to use with *R Markdown* is Zotero with the Better BibTex plug-in, because the `citr` plugin for RStudio (see below) can read references directly from your Zotero library!

Here is an example of an entry in a **.bib** file:

```
@article{Shea2014,
  author =        {Shea, Nicholas and Boldt, Annika},
  journal =       {Trends in Cognitive Sciences},
  pages =         {186--193},
  title =         {{Supra-personal cognitive control}},
  volume =        {18},
  year =          {2014},
  doi =           {10.1016/j.tics.2014.01.006},
}
```

In this entry highlighted section, 'Shea2014' is the **citation identifier**. To default way to cite an entry in your text is with this syntax: `[@citation-identifier]`.

So I might cite some things (Lottridge et al., 2012; Mill, 1965 [1843]; Shea et al., 2014).

---

[1]The bibliography can be in other formats as well, including EndNote (**.enl**) and RIS (**.ris**), see rmarkdown.rstudio.com/authoring_bibliographies_and_citations.

4.4.1 Appearance of citations and references section (pandoc)

By default, `oxforddown` lets Pandoc handle how citations are inserted in your text and the references section. You can change the appearance of citations and references by specifying a CSL (Citation Style Language) file in the `csl` metadata field of **index.Rmd**. By default, `oxforddown` by the Americal Psychological Association (7th Edition), which is an author-year format.

With this style, a number of variations on the citation syntax are useful to know:

- Put author names outside the parenthesis

  - This: `@Shea2014 says blah.`
  - Becomes: Shea et al. (2014) says blah.

- Include only the citation-year (in parenthesis)

  - This: `Shea et al. says blah [-@Shea2014]`
  - Becomes: Shea et al. says blah (2014)

- Add text and page or chapter references to the citation

  - This: `[see @Shea2014, pp. 33-35; also @Wu2016, ch. 1]`
  - Becomes: Blah blah (see Shea et al., 2014, pp. 33–35; also Wu, 2016, ch. 1).

If you want a numerical citation style instead, try `csl: bibliography/transactions-on-comput` or just have a browse through the Zotero Style Repository and look for one you like. For convenience, you can set the line spacing and the space between the bibliographic entries in the reference section directly from the YAML header in **index.Rmd**.

If you prefer to use `biblatex` or `natbib` to handle references, see this chapter.

4.4.2 Insert references easily with RStudio's Visual Editor

For an easy way to insert citations, use RStudio's Visual Editor. Make sure you have the latest version of RStudio – the visual editor was originally really buggy, especially in relation to references, but as per v2022.02.0, it's great!

4.5 Cross-referencing

We can make cross-references to **sections** within our document, as well as to **figures** (images and plots) and **tables**.

The general cross-referencing syntax is `\@ref(label)`

4.5.1 Section references

Headers are automatically assigned a reference label, which is the text in lower caps separated by dashes. For example, `# My header` is automatically given the label `my-header`. So `# My header` can be referenced with `\@ref(my-section)`

Remember what we wrote in section **??**?

We can also use **hyperlink syntax** and add # before the label, though this is only guaranteed to work properly in HTML output:

- So if we write `Remember what we wrote up in [the previous section](#citations)`?
- It becomes Remember what we wrote up in the previous section?

Creating custom labels

It is a very good idea to create **custom labels** for our sections. This is because the automatically assigned labels will change when we change the titles of the sections - to avoid this, we can create the labels ourselves and leave them untouched if we change the section titles.

We create custom labels by adding `{#label}` after a header, e.g. `# My section {#my-label}`. See our chapter title for an example. That was section **??**.

4.5.2   Figure (image and plot) references

- To refer to figures (i.e. images and plots) use the syntax `\@ref(fig:label)`
- **GOTCHA**: Figures and tables must have captions if you wish to cross-reference them.

Those results can be interpreted that the learning objectives of the training event was chosen appropriately for the diverse audience including clinicians, academics, researchers, and learning technologists/IT specialist resulting to a successful training event that enable participants to take the acquired knowledge back to their organisations in order to co-design and implement. As it was expected and can be depicted from self-confidence statements that some participants being very confident before the event, not all the objectives expected to be reached by everyone, since the training was targeting both technical and non-technical participants. However, on both average and individual matched responses participants self-statements showed that they improved their knowledge and understanding in using co-creation approaches to develop digital education resources and in designing and developing chatbots as educational resources.

4.6   Reach, Impact, and Qualatative analysis

```
# Appendix {-}
```

```
<!--chapter:end:04-Discussion.Rmd-->
```

```
output:
  #bookdown::html_document2: default
  #bookdown::word_document2: default
  bookdown::pdf_document2:
    template: templates/template.tex
```

4. Discussion

```
documentclass: book
#bibliography: [bibliography/references.bib, bibliography/additional-references.bib
---
```

(Additional Analyses) Training Events {#cites-and-refs}

```
\chaptermark{Citations and cross-refs}
\minitoc <!-- this will include a mini table of contents-->
```

CEPEH Training Event C1

The CEPEH training event C1 held at the premises of University of Nottingham aiming
15 participants were from RISE, AUTH, UoN.

Project managers of partners signposted the person involved, and relevant announcem

Participants were asked to highlight what they liked for each day and how each day

Day 1
The participants comment that they liked the design method for educational resource

Day 2

Participants enjoyed the presentation from the invited speaker from another facult

Day3

Participants liked the hands-on activities of the day also enjoyed the creativity o

CEPEH Training Event 2

Pre-Training Event survey May 9th-13th 2022 Thessaloniki, Greece

Twenty-six participants attended the Training Event, along with approximately 10 st

Table 1 suggested attendees had minimal intention to share their own ideas due to l

There were 77% of participants stated they were novices in experience with chatbots

One day had several events regarding cybersecurity in healthcare. When asked before

There were 90% (23) of students who heard about the event through a lecturer or a p

There were six individuals who stated neutral or disagree when asked if having issu

As this was face-to-face, participants were asked about sufficient Covid-19 precaut
In summary, most participants were undergraduate students with novice experience, h

<!--chapter:end:05-extensions.Rmd-->

\startappendices

<!-- If you feel it necessary to include an appendix, it goes here. The first appen

The First Appendix

This first appendix includes an R chunk that was hidden in the document (using `ech

4. Discussion

In 02-rmd-basics-code.Rmd

And here's another one from the same chapter, i.e. Chapter \\ref{code}:

The Second Appendix, for Fun

<!--chapter:end:front-and-back-matter/98-appendices.Rmd-->

<!-- This .Rmd file serves only to add the References heading, irrespective of outp

References {-}

<!-- the below latex command is needed after unnumbered chapter headings -- without

```{=tex}
\markboth{References}{}
```

Goethe, J. W. von. (1829). *Wilhelm Meisters Wanderjahre oder die Entsagen-den.* Cotta.

Lottridge, D., Marschner, E., Wang, E., Romanovsky, M., & Nass, C. (2012). Browser design impacts multitasking. *Proceedings of the Human Factors and Ergonomics Society 56th Annual Meeting.* https://doi.org/10.1177/10 71181312561289

Mill, J. S. (1965 [1843]). *A system of logic, ratiocinative and inductive: Being a connected view of the principles of evidence and the methods of scientific investigation.* Longmans.

Shea, N., Boldt, A., Bang, D., Yeung, N., Heyes, C., & Frith, C. D. (2014). Supra-personal cognitive control and metacognition. *Trends in Cognitive Sciences,*

18(4), 186–193. https://doi.org/10.1016/j.tics.2014.01.006

Wu, T. (2016). *The Attention Merchants: The Epic Scramble to Get Inside Our Heads.* Knopf Publishing Group.