

Data and Analysis Underlying the Research into The Artificial Social Agent Questionnaire (ASAQ) - Development and evaluation of a validated instrument for capturing human interaction experience with an artificial social agent

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2024-10-02

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doi: <http://>

This document provides links to files underlying the data and analysis presented in the paper:

Validating claims and replicating findings on the impact of artificial social agents (ASA), such as virtual agents, conversational agents and social robots, require a standardised measurement instrument that researchers can employ in different settings and for various agents. Such an instrument would allow researchers to evaluate their agents and establish insights beyond their specific study context. Therefore, we present the long and short versions of the ASA questionnaire (ASAQ) for evaluating human-ASA interaction on 19 constructs, such as the agent's believability, sociability, and coherence. It has been developed by an international workgroup with more than 100 ASA-researchers over multiple years who identified community-relevant constructs and associated questionnaire items and examined the questionnaire's reliability, validity, and interpretability. The result is a questionnaire that can capture more than 80% of the constructs that studies in the intelligent virtual agent community investigate, with acceptable levels of reliability, content validity, construct validity, and cross-validity. We suggest that ASA-researchers use the ASAQ short version to report their agent's psychographic information and the ASAQ long version to analyse any constructs in depth that are specifically relevant to their agent or study. Finally, this paper gives clear instructions for practical use, such as sample size estimations, and how to present results.

Requirements:

1. Pdf reader to read .pdf file
2. Text editor to read .csv file
3. R (v4.2.1)

Input: 1. data/constructASAQdesc.csv 2. data/itemASAQdesc.csv 3. data/1stReliabilityAnalysis/result_part_1.csv 4. data/1stReliabilityAnalysis/result_part_2.csv 5. data/constructDefining/expertReview.csv 6. data/predictiveValidation/scoreExpertA.csv 7. data/predictiveValidation/scoreExpertB.csv 8. data/predictiveValidation/scoreExpertC.csv 9. data/contentValidation/q431_expertGeneratedItems.csv 10. data/contentValidation/qTargets_vs_Distractors.csv 11. data/constructValidation/result_all.csv 12. data/crossValidation/result_all.csv

Libraries:

```
library(ltm)
library(psych)
library(MBESS)
library(dplyr)
library(stringi)
library(lavaan)
library(factoextra)
library(lattice)
library(boot)
library(nFactors)
library(Rcsdp)
```

```
library(GPARotation)
library(pwr)
library(knitr)
library(kableExtra)
library(distributions3)
library(rlist)
```

Dataset

!! Important note:

- For all datasets, all questionnaire-item data (with the column name) starting with “R_” have been reversed.

```
##### All Purposes Variables #####
# ASAQ's 7 points Likert-scales
scoreASAQ <- c(-3, -2, -1, 0, 1, 2, 3)

# 29 agents of the ASAQ Representative Set 2024 (14 agents in the Study Mid 2021 and 15
# in the Study 2022). Note:
# -SAM is the short name used to call Samantha (from the movie HER),
# -POPPIE is the name we used in the dataset for POPPY (from the Semaine project)
# -ROBOT BOSS is one of Double Robot agents,
# -SARAH is one of the Soul Machine agents, and
# -MARCUS is one of characters in the movie Terminator,
# More info about other ASAs see: https://osf.io/7wmjh/
# ASAs used in study Mid 2021:
name14ASAs <- c("AIBO", "AMY", "CHAPPIE", "DEEPBLUE", "DOG", "FURBY", "HAL 9000",
               "iCAT", "MARCUS", "NAO", "POPPIE", "SARAH", "SIM SENSEI", "SIRI")
# ASAs used in study 2024:
name15ASAs <- c("ALEXA", "ALICE", "C3PO", "EFFIE", "THE FISH", "FURHAT", "GEMINOID",
               "KIT", "LOLA", "PARO", "ROBOT BOSS", "SAM", "THE AMBIENT LIGHT TV",
               "THE NEGOTIATOR", "THE ZOMBIE")
name29ASAs <- c("AIBO", "ALEXA", "ALICE", "AMY", "CHAPPIE", "C3PO", "DEEPBLUE", "DOG",
               "EFFIE", "THE FISH", "FURBY", "FURHAT", "GEMINOID", "HAL 9000", "iCAT",
               "KIT", "LOLA", "MARCUS", "NAO", "PARO", "POPPIE", "ROBOT BOSS", "SAM",
               "SARAH", "SIM SENSEI", "SIRI", "THE AMBIENT LIGHT TV",
               "THE NEGOTIATOR", "THE ZOMBIE")

#probabilities used in calculating percentiles
probs<-c(.05, .10, .20, .25, .30, .40, .50, .60, .70, .75, .80, .90, .95)
probTitle<-c("q5th", "q10th", "q20th", "q25th", "q30th", "q40th", "q50th", "q60th",
             "q70th", "q75th", "q80th", "q90th", "q95th")

probsSML<-c(.25, .50, .75)
probTitleSML<-c("Small_q25th", "Medium_q50th", "Large_q75th")

# Description ASAQ's constructs/dimensions
# CID = construct/dimension's id used before study Mid 2021.
# It uses notation "C<construct-number>D<dimension-number>"
# CABR = construct/dimension's id used after study Mid 2021. They are the abbreviation
```

```

# of the construct/dimension' name
# CNAME = construct/dimension's name (see Table 1 for the complete version)
# NUM_ITEM = the number of items in the construct/dimension
# SHORT_ITEM_ID = the ids of 24 representative items that are used to establish the
# short version of ASAQ. This id used before study Mid 2021
# SHORT_ITEM_REF = the ids of the short version of ASAQ that are corresponding to
# the long version of ASAQ
constructASAQdesc <- read.csv2("data/constructASAQdesc.csv",
                              header = TRUE, sep = ";", na.strings = "NNN")

# Description ASAQ's items
# ITEM_ID = item's id used before study Mid 2021
# It used notation "C<construct-number>D<dimension-number>Q<item-number>"
# ITEM_REF = item's id used after study Mid 2021. It uses notation
# <construct-abbreviation><item-number>
itemASAQdesc <- read.csv2("data/itemASAQdesc.csv",
                          header = TRUE, sep = ";")

##### Construct Definition dataset (n=13) #####
# Expert voting on classifying 189 constructs found in IVA proceedings 2013-2018
# into 7 concepts
# No = IVA construct id [C1__ .. C189]
# Construct = construct's name
# APprop, ASoc, ARole, IntQ, HImpr, AImpr, HAttr = the number of voters classifying the
# corresponding construct into concept, respectively Agent's Properties, Agent's
# Social Trait, Agent's Role (Performance), Interaction Quality, Human's Impression
# of the Interaction, Impression of the agent's after the Interaction, and
# Human's Attributes (to Support the Interaction)
# Other = the number of voters classifying the corresponding construct other than the
# 7 concepts
# nVoters = the total number of voters for the corresponding construct
# SingleCls = the concept that holds the majority (>= 50%) voters for the corresponding
# construct
# PairCls_1 = the concept that holds the first largest number of voters but below 50%
# voters for the corresponding construct
# PairCls_2 = the concept that holds the second largest number of voters but below 50%
# voters for the corresponding construct
constructDefData = read.csv2("data/constructDefining/expertReview.csv", header = TRUE,
                             sep = ";")

##### Content Validation dataset (n=31) #####
# In the content validation study, experts (n=31) were asked by 223 questions. In each
# question, they asked whether each item presented was intended for the questioned
# construct/dimension. In a question, at most two target items (i.e., items that are
# intended for the questioned construct/dimension) and two random distractor items
# (i.e., items that should not be for the questioned construct/dimension).
# See the following dataset, which contained 431 expert-generated-questionnaire items,
# the construct/dimension id and name of each target item.
# QUESTIONID = to the question id. It's composed by C?? is the construct id, D?? is the
# dimension id, and S* is the number of questions [1..223]
# CONID = the construct id of the target. It uses the notation before study Mid 2021.
# CONNAME = the construct name of the target
# DIMID = the dimension id of the target. It uses the dimensional and notation

```

```

# before study Mid 2021.
# DIMNAME = the dimension name of the target
# ITEMID = the item id of the target. It uses the notation before study Mid 2021.
# ITEMQID = the id of the (YES or NO) questions in qualtrics of the target items;
# It's composed by <QUESTIONID>_1 or _3;
# ITEMRID = the reference id the target items C??D??S*_1_TRGT and C??D??S*_2_TRGT
# ITEMQ = the questionnaire item statement
contentValItems <-
  read.csv("data/contentValidation/q431_expertGeneratedItems.csv", header = TRUE,
    sep = ";")

# The result of the content validity study with 31 experts. It contained the list of
# pairing of one target item and one distractor item in a (qualtrics) question,
# and the answer of the experts for both items.
# PRID = anonymous id of the experts. An expert was asked to answer 233 questions
# It was allowed not to answer.
# QUESTIONID = to the question id (i.e. task number). It's composed by C?? is the
# construct id, D?? is the dimension id, and S* is the number of questions [1..223]
# TRGT_ITEMID = the item id of the target. It uses the notation before study Mid 2021.
# TRGT_ITEMQID = the id of the (YES or NO) questions in qualtrics of the target items;
# It's composed by <QUESTIONID>_1 or _3.
# TRGT_ITEMRID = the reference id of the target items C??D??S*_1_TRGT and C??D??S*_2_TRGT
# TRGT_ITEMQ = the questionnaire item statement
# DSTR_ITEMID = the item id of the distractor. It uses the notation before study Mid 2021.
# DSTR_ITEMQID = the id of the (YES or NO) questions in qualtrics of the distractor items;
# It's composed by <QUESTIONID>_2 or _4;
# DSTR_ITEMRID = the reference id of the distractor items C??D??S*_1_DSTR and
# C??D??S*_2_DSTR
# DSTR_ITEMQ = the questionnaire item statement
# TRGT_ANS = the answer of the expert for the target item
# DSTR_ANS = the answer of the expert for the distractor item
contentValData <- read.csv("data/contentValidation/qTargets_vs_Distractors.csv",
  header = TRUE, sep = ";")

##### Study Early '21 dataset (n=192) #####
# The Study Early '21 with 192 crowd-workers on Prolific
# PRID = Anonymous id of the participant
# STARTDATE = starting datetime of the data entry
# ENDDATE = ending datetime of the data entry
# RECORDDATE = datetime when the data was recorded
# GROUP = group of participants, i.e., one of four groups - 1stPOV_part_1,
# 1stPOV_part_2, 3rdPOV_part_1, and 3rdPOV_part_2
# CheckScore = the number of attention-check-correct answers given by participants
# C01D01Q?? .. C19D00Q?? = the participant's rates on the questionnaire items. The id
# of items was based the notation before Study Mid 2021
reliabilityData_part_1_raw = read.csv2("data/1stReliabilityAnalysis/result_part_1.csv",
  header = TRUE, sep = ";")
reliabilityData_part_2_raw = read.csv2("data/1stReliabilityAnalysis/result_part_2.csv",
  header = TRUE, sep = ";")

# Read obtained data based on conditions: 1st Person of View and 3rd Person of View from
# both parts
reliabilityData_part_1 <-

```

```

reliabilityData_part_1_raw[7:ncol(reliabilityData_part_1_raw)]
reliabilityData_part_2 <-
  reliabilityData_part_2_raw[7:ncol(reliabilityData_part_2_raw)]
reliabilityData_1stPOV_part_1 =
  reliabilityData_part_1_raw[reliabilityData_part_1_raw$GROUP == '1stPOV_part_1',]
reliabilityData_1stPOV_part_2 =
  reliabilityData_part_2_raw[reliabilityData_part_2_raw$GROUP == '1stPOV_part_2',]
reliabilityData_3rdPOV_part_1 =
  reliabilityData_part_1_raw[reliabilityData_part_1_raw$GROUP == '3rdPOV_part_1',]
reliabilityData_3rdPOV_part_2 =
  reliabilityData_part_2_raw[reliabilityData_part_2_raw$GROUP == '3rdPOV_part_2',]
reliabilityData_1stPOV_part_1 =
  reliabilityData_1stPOV_part_1[7:ncol(reliabilityData_1stPOV_part_1)]
reliabilityData_1stPOV_part_2 =
  reliabilityData_1stPOV_part_2[7:ncol(reliabilityData_1stPOV_part_2)]
reliabilityData_3rdPOV_part_1 =
  reliabilityData_3rdPOV_part_1[7:ncol(reliabilityData_3rdPOV_part_1)]
reliabilityData_3rdPOV_part_2 =
  reliabilityData_3rdPOV_part_2[7:ncol(reliabilityData_3rdPOV_part_2)]
reliabilityData <- cbind(reliabilityData_part_1, reliabilityData_part_2)

##### Study Mid 2021 Dataset (n=532) #####
# The Study Mid 2021 study involving 532 crowd-workers on Prolific
# PRID = Anonymous id of the participant
# STARTDATE = starting datetime of the data entry
# ENDDATE = ending datetime of the data entry
# RECORDDATE = datetime when the data was recorded
# GROUP = group of participants, i.e. 14 agents
# CheckScore = the number of attention-check-correct answers given by participants
# C01D01Q?? .. C19D00Q?? = the participant's rates on the questionnaire items. The id
# of items was based the notation before Study Mid 2021
constructValData_raw <-
  read.csv2("data/constructValidation/result_all.csv", header = TRUE, sep = ";")
constructValData <-
  cbind(constructValData_raw$GROUP, constructValData_raw[, 7:ncol(constructValData_raw)])
names(constructValData) <-
  c("ASA", colnames(constructValData_raw[, 7:ncol(constructValData_raw)]))

##### Study '22 dataset (n=534) #####
# The Cross Validation study involving 534 crowd-workers on Prolific
# PRID = Anonymous id of the participant
# STARTDATE = starting datetime of the data entry
# ENDDATE = ending datetime of the data entry
# RECORDDATE = datetime when the data was recorded
# GROUP = group of participants, i.e. 15 agents
# CheckScore = the number of attention-check-correct answers given by participants
# C01D01Q?? .. C19D00Q?? = the participant's rates on the questionnaire items. The id
# of items was based the notation before Study Mid 2021
crossValData_raw <- read.csv2("data/crossValidation/result_all.csv", header = TRUE,
  sep = ";")
crossValData <- cbind(crossValData_raw$GROUP,
  crossValData_raw[, 7:ncol(crossValData_raw)])
names(crossValData) <- c("ASA", colnames(crossValData_raw[, 7:ncol(crossValData_raw)]))

```

```
##### Predictive Validity dataset (n=3) #####
# Three experts' prediction on the scores of ASAQ constructs for 54 agents.
# One dataset file per expert
# ASA = the agent's name
# C1D1 .. C19 = the participant's rates on the questionnaire items. The id
# of constructs was based the notation before Study Mid 2021
predictiveValData_expertA = read.csv("data/predictiveValidation/scoresExpertA.csv",
                                     header = TRUE, sep = ";")
predictiveValData_expertB = read.csv("data/predictiveValidation/scoresExpertB.csv",
                                     header = TRUE, sep = ";")
predictiveValData_expertC = read.csv("data/predictiveValidation/scoresExpertC.csv",
                                     header = TRUE, sep = ";")
```

Functions

```
##### ALL PURPOSES #####

# Convert to numeric and return a subset of a dataframe starting from startCol
# to ncol(data)
convertToNumeric <- function(data, startCol){
  return(as.data.frame(sapply(data[startCol:ncol(data)], as.numeric)))
}

# Standardize data. The observed data was standardized to allow us to compare the
# ratings between different ASAs. Standardization procedure:
# 1. calculate the mean and the standard deviation per item per ASA.
# 2. for each observed value of an item, subtract it by the mean based on its
# corresponding ASA
# 3. for each observed value of an item, divide it by the standard deviation based on
# its corresponding ASA
getStandardize <- function(agents, data){
  d_M = data.frame()
  d_SD = data.frame()
  row = 1
  for (agent in agents){
    d_M[row,1] <- agent
    d_SD[row,1] <- agent
    for(j in 2:ncol(data)){
      d<-data[data[,1]==agent,j]
      if (length(d)==0) {
        print(j)
        print(paste("agent ", agent, " is empty", sep=" "))
      }
      M <- mean(d)
      std <- sd(d)
      d_M[row,j] <- M
      d_SD[row,j] <- std
    }
    row = row + 1
  }
}
```

```

d_standard = data.frame()
for(col in 2:ncol(data)){
  newCol = col - 1
  for (row in 1:nrow(data)){
    M = d_M[d_M[,1]==data[row,1], col]
    std = d_SD[d_SD[,1]==data[row,1], col]
    d_standard[row,newCol] <- (data[row, col] - M)/std
  }
}
names(d_standard) <- colnames(data[c(2:ncol(data))])
return(d_standard)
}

# Get data of individual constructs/dimensions
# INPUT
# arr: the number of columns of items of each construct/dimension in the dataset
# d: data
# inc: the order of the constructs/dimensions [1..26] (see the variable constructs)
#
# OUTPUT
# data of items of the i-th construct/dimension in inc
getDataColWise <- function(arr, d, inc){
  dat <- data.frame()
  end <- 0
  for(i in 1:inc){
    begin <- end + 1
    end <- end + arr[i]
  }
  dat <- d[begin:end]
  return(convertToNumeric(dat, 1))
}

# Get data of individual constructs/dimensions
# INPUT
# arr: the number of columns of items of each construct/dimension in the dataset
# d: data
# inc: the order of the constructs/dimensions [1..26] or [1 ..24]
#       (see the variable constructs)
#
# OUTPUT
# data of items of the i-th construct/dimension in inc
getDataRowWise <- function(arr, d, inc){
  dat <- data.frame()
  end <- 0
  for(i in 1:inc){
    begin <- end + 1
    end <- end + arr[i]
  }
  dat <- d[begin:end,]
  return(dat)
}

# Extract 90 items of the long ASAQ from data

```



```

# INPUT
# data_ori = dataset from study before Mid 2021
# itemDesc = description of ASAQ items containing the list of ASAQ items' id
# (ids used before study Mid 2021) and ref (ids used after study Mid 2021)
# startCol = the starting column number to read the dataset
# OUTPUT
# dataset of 90 items of ASAQ, with description
getLongASAQ <- function(data_ori, itemId, itemRef, startCol){
  # adding additional columns to the ASAQ items, e.g. ASAs
  startCols = colnames(data_ori)[1:startCol-1]
  longASAQCols = c(startCols, itemId)

  dat <- data.frame()
  for (col in longASAQCols){
    if (nrow(dat) == 0) dat <- data_ori[col]
    else dat <- cbind(dat, data_ori[col])
  }

  names(dat) <- c(startCols, itemRef)
  return(dat)
}

# Extract 24 items of the short ASAQ from data
# INPUT
# data_ori = dataset from study before Mid 2021
# constructDesc = description of ASAQ construct/dimension containing short_item_id
# and short_item_ref
# startCol = the starting column number to read the dataset
# OUTPUT
# dataset of 90 items of ASAQ, with description
getShortASAQ <- function(data_ori, constructDesc, startCol){
  # adding additional columns to the ASAQ items, e.g. ASAs
  startCols = colnames(data_ori)[1:startCol-1]
  shortASAQCols = c(startCols, constructDesc$SHORT_ITEM_ID)

  dat <- data.frame()
  for (col in shortASAQCols){
    if (nrow(dat) == 0) dat <- data_ori[col]
    else dat <- cbind(dat, data_ori[col])
  }

  names(dat) <- c(startCols, constructDesc$SHORT_ITEM_REF)
  return(dat)
}

# Display a dataframe for the manuscript purposes
# INPUT
# data_ori = the data to be displayed
# idConstructs, abrConstructs and nameConstructs = the list of ASAQ
# constructs/dimensions' ids, abbreviations and names
# type = [0 = displays the data,
#         1 = before displaying it is necessary to transpose the data,
#         2 = displaying data combining from both short and long versions of ASAQ]

```

```

# showDesc = displaying (1) or not (0) data description, i.e., min, max, mean, sd, median
#
# OUTPUT
# Displaying a table starts with the columns construct id, construct abbreviation, and
# construct name, and then continues with the data that is corresponding to 24 ASAQ
# constructs/dimensions.
# If showDesc == 1 then displaying the data description underneath the table
showTable <- function(data_ori, abrConstructs, nameConstructs, type,
                      showDesc, showMean, showSD, showMedian, showMin, showMax){
  dat <- data.frame(matrix(nrow = length(abrConstructs), ncol = 0))
  data <- data.frame()
  if (type == 0)
    data <- convertToNumeric(data_ori, 2)
  if (type == 1){
    headers <- data_ori[, 1]
    data_ori <- convertToNumeric(data_ori, 2)
    for(i in 1:length(headers)){
      for (j in 1:length(abrConstructs)){
        data[j, headers[i]] <- data_ori[i, j]
      }
    }
    colnames(data) <- headers
  }
  if (type == 2){
    data <- convertToNumeric(data_ori, 2)
    n = ncol(data)/2
    h1 <- paste("LONG", colnames(data)[1:n], sep = "_")
    n = n + 1
    h2 <- paste("SHORT", colnames(data)[n:ncol(data)], sep = "_")
    colnames(data) <- c(h1,h2)
  }
  data <- round(data,2)

  dat[, "CABR"] <- abrConstructs
  dat[, "CNAME"] <- nameConstructs
  dat <- cbind(dat, format(data, nsmall=2))

  print(dat)

  if (showDesc == 1){
    # Do data description
    rows = showMean + showSD + showMedian + showMin + showMax
    i = 1
    desc <- data.frame(matrix(nrow = rows, ncol = ncol(dat)))
    if (showMean == 1){
      vals = round(apply(data, 2, mean, na.rm=TRUE), 2)
      desc [i, ] <- c("", "Mean:", format(vals, nsmall=2))
      i = i + 1
    }
    if (showSD == 1){
      vals = round(apply(data, 2, sd, na.rm=TRUE), 2)
      desc [i, ] <- c("", "SD:", format(vals, nsmall = 2))
      i = i + 1
    }
  }
}

```

```

}
if (showMedian == 1){
  vals = round(apply(data,2,median,na.rm=TRUE),2)
  desc [i, ] <- c("", "Median:", format(vals, nsmall=2))
  i = i + 1
}
if (showMin == 1){
  vals = apply(data, 2, min, na.rm=TRUE)
  desc [i, ] <- c("", "Min:", format(vals, nsmall = 2))
  i = i + 1
}
if (showMax == 1){
  vals = apply(data, 2, max, na.rm=TRUE)
  desc [i, ] <- c("", "Max:", format(vals, nsmall = 2))
  i = i + 1
}

# for printing purpose
colnames(desc) <- c("", "Description", names(data))
print(desc)
}
}

# Display a dataframe of sample sizes for the manuscript purposes
# INPUT
# data_ori = the data to be displayed
# idConstructs, abrConstructs and nameConstructs = the list of ASAQ
# constructs/dimensions' ids, abbreviations and names
# type = [0 = displays the data,
#         1 = before displaying it is necessary to transpose the data,
#         2 = displaying data combining from both short and long versions of ASAQ]
#
# OUTPUT
# Displaying a table starts with the columns construct id, construct abbreviation, and
# construct name, and then continues with the data that is corresponding to 24 ASAQ
# constructs/dimensions.
showTableSample <- function(data_ori, abrConstructs, nameConstructs, type){
  data <- data.frame()
  if (type == 0)
    data <- convertToNumeric(data_ori, 2)
  if (type == 1){
    headers <- data_ori[, 1]
    data_ori <- convertToNumeric(data_ori, 2)
    for(i in 1:length(headers)){
      for (j in 1:length(abrConstructs)){
        data[j, headers[i]] <- data_ori[i, j]
      }
    }
    colnames(data) <- headers
  }
  if (type == 2){
    data <- convertToNumeric(data_ori, 2)
    n = ncol(data)/2

```

```

h1 <- paste("LONG", colnames(data)[1:n], sep = "_")
n = n + 1
h2 <- paste("SHORT", colnames(data)[n:ncol(data)], sep = "_")
colnames(data) <- c(h1,h2)
}
data <- round(data,0)

dat <- data.frame(matrix(nrow = length(abrConstructs), ncol = 0))
dat[, "CABR"] <- abrConstructs
dat[, "CNAME"] <- nameConstructs
dat <- cbind(dat, data)

print(dat)

desc <- data.frame(matrix(nrow = 1, ncol = ncol(dat)))
desc [1, ] <- c("", "Max:", round(apply(data, 2, max, na.rm=TRUE),0))
colnames(desc)<-c("", "Description", names(data))
print(desc)
}

# Display the dataframe of the ASAQ constructs/dimensions scores of the ASAQ
# representative set 2024 for the manuscript purposes
# INPUT
# data_mean = the data to be displayed (the constructs/dimensions scores of the ASAs)
# data_sd = the data to be displayed (the standard deviation of the constructs/dimensions
# scores of the ASAs)
# data_sum = the ASAQ score of the ASAs
# data_samples = the sample size of each ASA's dataset
# idConstructs, abrConstructs and nameConstructs = the list of ASAQ
# constructs/dimensions' ids, abbreviations and names
# showDesc = displaying (1) or not (0) data description, i.e., min, max, mean, sd, median
#
# OUTPUT
# Displaying a table starts with the columns construct id, construct abbreviation, and
# construct name, and then continues with the data that is corresponding to 24 ASAQ
# constructs/dimensions.
# If showDesc == 1 then displaying the data description underneath the table
showTableASA <- function(data_mean, data_sd, data_sum, data_samples,
                        abrConstructs, nameConstructs){
  d_mean <- convertToNumeric(data_mean, 2)
  d_sd <- convertToNumeric(data_sd, 2)

  headers <- c()
  for(h in data_mean[,1]){
    h_mean <- paste0(h, "_M")
    h_sd <- paste0(h, "_SD")
    headers <- c(headers, h_mean, h_sd)
  }

  data <- data.frame(matrix(nrow = length(abrConstructs), ncol = length(headers)))
  col = 1
  for(i in 1:length(data_mean[,1])){
    for (j in 1:length(abrConstructs)){

```

```

    data[j, col] <- d_mean[i, j]
  }
  col = col + 1
  for (j in 1:length(abrConstructs)){
    data[j, col] <- d_sd[i, j]
  }
  col = col + 1
}
data <- format(round(data,2), nsmall=2)
colnames(data) <- headers

dat <- data.frame(matrix(nrow = length(abrConstructs), ncol = 0))
dat[, "CABR"] <- abrConstructs
dat[, "CNAME"] <- nameConstructs
dat <- cbind(dat, format(data, nsmall=2))

print(dat)

col = 1
val_samples<-c()
val_sums <- c()
for(h in data_mean[,1]){
  val_samples[col] <- data_samples[data_samples$ASA==h,2]
  val_sums[col] <- data_sum[data_sum$ASA==h,2]
  col = col + 1
}

desc <- data.frame(matrix(nrow = 2, ncol = length(data_mean[,1])+2))
desc[1,]<-c("", "Sample Size", val_samples)
desc[2,]<-c("", "ASAQ score", val_sums)
colnames(desc)<-c("", "Description", data_mean[,1])

print(desc)
}

##### Construct Definition #####
# Calculating the total number of single-concept and pair-of-concepts constructs based
# on the expert's voting
# INPUT
# data: expertReview.csv
#
# OUTPUT
# the table of the number of constructs classified into seven anchor points
calcExpertVotingClassification <- function(data){
  dat<-data.frame()
  anchorPoints <- c("Agent's properties", "Agent's social traits", "Agent's role",
                    "Impression of the agent after the interaction",
                    "Interaction quality", "Human's impression of the interaction",
                    "Human's attributes")
  abbr= c("AProp", "ASoc", "ARole", "AImpr", "IntQ", "HImpr", "HAttr")

  # calculate the number of constructs classified in 7 concepts and Other
  for (i in 1:length(anchorPoints)){
    dat[i, "AnchorPoint"] <- anchorPoints[i]

```

```

    single <- length(which(data$SingleCls == abbr[i]))
    pair <- length(which(data$PairCls_1 == abbr[i])) +
      length(which(data$PairCls_2 == abbr[i]))
    dat[i, "SingleAnchor"] <- single
    dat[i, "PairAnchors"] <- pair
  }
  return(dat)
}

##### Study Early '21 #####
# Calculate Cronbach's alpha of construct/dimension
# INPUT
# arr: the number of columns of items of each construct/dimension in the dataset
# d: data
# constructs: construct-ids
#
# OUTPUT
# table of the cronbach's alpha coefficients of all constructs/dimensions in the dataset
calcAlpha <- function(arr, data, constructs){
  dat <- data.frame()
  for(i in 1:length(constructs)){
    d <- getDataColWise(arr, data, i)
    a <- cronbach.alpha(d, standardized=TRUE)
    dat[i, "CONSTRUCTID"] <- constructs[i]
    dat[i, "std.a"] <- round(a$alpha, 2)
  }
  return(dat)
}

# Calculate the cronbach's alpha coefficient for the reliability data 1stPOV only,
# 3rdPOV only and the combination data of 1stPOV and 3rdPOV
# INPUT
# all: the results of the Study Early '21 from the combination data of 1stPOV and
# 3rd POV
# a1stPOV_part1, a1stPOV_part2, a3rdPOV_part1, a3rdPOV_part2: the results of
# the Study Early '21 of each part of dataset
#
# OUTPUT
# the aggregation of all results of the Study Early '21
getAlphaReliability <- function(all, a1stPOV_part1, a1stPOV_part2, a3rdPOV_part1,
                                a3rdPOV_part2){
  a1stPOV <- rbind(a1stPOV_part1, a1stPOV_part2)
  a3rdPOV <- rbind(a3rdPOV_part1, a3rdPOV_part2)
  diff <- abs(a1stPOV[2]-a3rdPOV[2])
  dat <- cbind(all, a1stPOV[2], a3rdPOV[2], diff)
  names(dat)<-c("CONSTRUCTID", "std.a_all", "std.a_1stPOV", "std.a_3rdPOV",
              "abs_diff_1st_3rd")
  return(dat)
}

##### Content Validation #####
# Calculate agreement coefficient of ASAQ items
# INPUT
# itemDesc: description of the long ASAQ's 90 items

```

```

# targetVsDistractor: the pairing of target items and distractors and the experts'
# answers for both corresponding pair-items
# Calculation:
# tp: true positive (if target is classified as YES)
# fn: false negative (if target is classified as NO)
# tn: true negative (if distractor is seen as a YES)
# fp: false positive (if distractor is seen as a NO)
#
# OUTPUT
# The calculation of tp, fn, tn, fp, TPRs and FPRs for each ASAQ item
calcItemAgreementCoeff <- function(itemDesc, targetVsDistractor){
  dat <- data.frame()

  for (row in 1:nrow(itemDesc)){
    dat[row, "ITEM_REF"] <- itemDesc[row, "ITEM_REF"]
    item = itemDesc[row, "ITEM_ID"]
    # Calculate TPR_s
    d_trgt <- targetVsDistractor[targetVsDistractor$TRGT_ITEMID == item,]
    TP <- length(which(d_trgt$TRGT_ANS == 1)) #if target is classified as YES
    FN <- length(which(d_trgt$TRGT_ANS == 0)) #if target is classified as NO
    dat[row, "TP"] <- TP
    dat[row, "FN"] <- FN
    APos <- TP + FN
    dat[row, "APos"] <- APos
    dat[row, "TPR_s"] <- round(((TP - (0.5 * APos)) / APos) + 0.5, 2)

    # Calculate FPR_s
    d_dstr <- targetVsDistractor[targetVsDistractor$DSTR_ITEMID == item,]
    FP <- length(which(d_dstr$DSTR_ANS == 1)) # if distractor is seen as a YES
    TN <- length(which(d_dstr$DSTR_ANS == 0)) # if distractor is seen as a NO
    dat[row, "FP"] <- FP
    dat[row, "TN"] <- TN
    ANeg <- FP + TN
    dat[row, "ANeg"] <- ANeg
    dat[row, "FPR_s"] <- round(((FP - (0.5 * ANeg)) / ANeg) + 0.5, 2)
  }

  return(dat)
}

# Calculate agreement coefficient of ASAQ construct/dimensions
# INPUT
# constructDesc: description of the long ASAQ's construct/dimension
# itemAgreementCoeffs: TPRs and FPRs values of all ASAQ items
#
# OUTPUT
# TPRs and FPRs values of all ASAQ constructs/dimensions
calcConstructAgreementCoeff <- function(constructDesc, itemAgreementCoeffs){
  dat <- data.frame()
  coeffs <- convertToNumeric(itemAgreementCoeffs, 2)

  for (row in 1:nrow(constructDesc)){
    d <- getDataRowWise(constructDesc$NUM_ITEMS, coeffs, row)
  }
}

```

```

dat[row, "CABR"] <- constructDesc[row, "CABR"]
dat[row, "M_TPRs"] <- round(mean(d[, "TPR_s"]), 2)
dat[row, "SD_TPRs"] <- round(sd(d[, "TPR_s"]), 2)
dat[row, "M_FPRs"] <- round(mean(d[, "FPR_s"]), 2)
dat[row, "SD_FPRs"] <- round(sd(d[, "FPR_s"]), 2)
}

return(dat)
}

##### Study Mid 2021 (n=532) #####
##### Study 2022 (n=534) #####
# Run CFA based on the list of factorial models
# INPUT
# modelNames = name used to distinguish the model
# factorialModels = the list of factorial models
# longASAQ_data_std = standardized long ASAQ dataset
#
# OUTPUT
# CFI coefficient of cfa of the factorial models
getCFIcoefficient <- function(modelNames, factorialModels, longASAQ_data_std){
  dat <- data.frame()

  for (i in 1:length(modelNames)){
    dat[i, "Model"] <- modelNames[i]
    set.seed(1234)
    dat[i, "CFI"] <-
      format(round(fitMeasures(cfa(factorialModels[i], data=longASAQ_data_std),
                                "cfi"), 2), nsmall = 2)
  }

  return(dat)
}

# Calculate predicted latent scores of long ASAQ constructs/dimensions
# INPUT
# individualConstructModels = the list of individual construct models
# longASAQ_data_std = standardized long ASAQ dataset
#
# OUTPUT:
# Predicted latent scores of long ASAQ constructs/dimensions
getPredictedData_longASAQ <- function(individualConstructModels, longASAQ_data_std){

  dat <- NULL
  for (i in 1:length(individualConstructModels)){
    set.seed(1234)
    fit <- cfa(individualConstructModels[i], data=longASAQ_data_std)
    pred <- data.frame(lavPredict(fit, type="lv"))
    if (is.null(dat)) dat <- pred
    else dat <- cbind(dat, pred)
  }

  return(dat)
}

```



```

# Maximum Likelihood Factor Analysis
# INPUT
# predictedData_longASAQ: predicted latent scores of long ASAQ constructs/dimensions
# nfactors: the number of intended factors
#
# OUTPUT
# Display the results of EFA
getEFA <- function(predictedData_longASAQ, nfactors){
  set.seed(1234)
  data_cor <- cor(predictedData_longASAQ)
  factors_data <- fa(data_cor, nfactors, rotate = "promax", SMC=FALSE, fm="ml",
                     max.iter=100)
  return(factors_data)
}

# Run CFA factorial models of the long ASAQ and returning the factor loading scores
# INPUT
# modelNames = name used to distinguish the model
# factorialModels = predefined factorial models based on the EFA results
# longASAQ_data_std = standardized long ASAQ dataset
#
# OUTPUT
# Factor loadings of items based on the factorial models
printStdFactorLoadingModels_longASAQ <- function(modelNames, factorialModels,
                                                  longASAQ_data_std, printing){

  dat <- list()
  for (i in 1:length(factorialModels)){
    set.seed(1234)
    fit <- cfa(factorialModels[i], data=longASAQ_data_std)
    loadings <- round(data.frame(inspect(fit, what="std")$lambda),2)
    ccov <- round(data.frame(lavInspect(fit, what="std")$psi),2)
    dat <- list.append(dat, loadings)

    if (printing == 1){
      cat("\n\n", modelNames[i] , "(CFI = ",
          format(round(fitMeasures(fit, "cfi"), 2), nsmall = 2),
          ")\n")
      for (j in 1:ncol(loadings)){
        d <- loadings[j]
        d <- d %>% dplyr::filter(d[,1]>0.0)
        print(format(d, nsmall=2))
        cat("\n")
      }
      cat("Covariance Matrix \n")
      print(ccov)
      cat("\n")
    }
  }
  return(dat)
}

# Return only factor loadings scores of all models
# INPUT

```

```

# loading_models = factor loadings of items resulted from
# getStdFactorLoadingModels_longASAQ()
#
# OUTPUT
# Factor loadings of all items
getStdFactorLoadings <- function(loading_models){
  dat <- NULL
  for (i in 1:length(loading_models)){
    loading_model <- loading_models[[i]]
    for (j in 1:ncol(loading_model)){
      d <- loading_model[j]
      d <- d[d[,1]>0.0,]
      if (is.null(dat)) dat <- d
      else dat <- c(dat, d)
    }
  }
  return(dat)
}

# Run CFA factorial models of the long ASAQ and returning the covariance matrix
# INPUT
# modelNames = name used to distinguish the model
# factorialModels = predefined factorial models based on the EFA results
# longASAQ_data_std = standardized long ASAQ dataset
#
# OUTPUT
# Covariance matrix based on the factorial models
getStdCovarianceMatrixModels_longASAQ <- function(modelNames, factorialModels,
                                                  longASAQ_data_std, printing){
  dat <- list()
  for (i in 1:length(factorialModels)){
    set.seed(1234)
    fit <- cfa(factorialModels[i], data=longASAQ_data_std)
    ccov <- round(data.frame(lavInspect(fit, what="std")$psi),2)
    dat <- list.append(dat, c)

    if (printing == 1){
      cat("\n", modelNames[i] , "\n")
      print(format(ccov, nsmall=2))
      cat("\n")
    }
  }
  return(dat)
}

```

```

##### Predictive Validation #####
# Calculate the constructs' score of each ASA by calculating the median of the the score
# of each construct/dimension per ASA given by the experts. The calculation results in
# one score for each construct/dimension per ASA
# INPUT:
# d_expertA, d_expertB, d_expertC = data containing the scores of all ASAQ
# constructs/dimensions given by respectively expertA, expertB and expertC with one of
# the columns is "ASA" containing the list of ASAs followed columns of the number of
# ASAQ constructs/dimensions

```

```

# constructs: the list of ASAQ constructs/dimensions
#
# OUTPUT:
# the constructs'/dimensions' score of each ASA
getPredictiveValScores_medianBase<- function(d_expertA, d_expertB, d_expertC,
                                              constructs){

  dat <- data.frame()
  A <- convertToNumeric(d_expertA, 2)
  B <- convertToNumeric(d_expertB, 2)
  C <- convertToNumeric(d_expertC, 2)

  id26constructs <- colnames(A)
  for (row in 1:nrow(A)){
    col = 1
    for (construct in constructs){
      dat[row, "ASA"] <- d_expertA[row, "ASA"]
      if (id26constructs[col] == "C3D1" | id26constructs[col] == "C11D1" |
          id26constructs[col] == "C3D2" | id26constructs[col] == "C11D2") {
        # ignore
        col = col + 1
      }
      else
        dat[row, construct] <- median(c(A[row, col], B[row, col], C[row, col]))
        col = col + 1
      }
    }
  }
  return(dat)
}

# Calculate construct/dimension scores for ASAQ Representative Set 2024 based on median
# INPUT
# agents: a list of agents used in the Study Mid 2021 and Study 2022
# data: the constructs'/dimensions' score of each ASA calculating based on the median of
# the scores
# constructs: the list of 24 ASAQ constructs/dimensions
#
# OUTPUT
# the constructs'/dimensions' score of each ASA
get29ASAPredictiveValData_medianBase<- function(agents, data, constructs){
  dat <- data.frame()

  row <- 1
  for(agent in agents){
    d <- data[data$ASA == agent,]
    dat[row, "ASA"] <- agent
    for (construct in constructs){
      if (construct == "PF" | construct == "UAL")
        dat[row, construct] <- ""
      else
        dat[row, construct] <- d[,construct]
    }
    row <- row + 1
  }
}

```

```

return(dat)
}

# Calculate the correlation between the scores of ASAQ constructs/dimensions of the ASAQ
# Representative Set 2024 of the experts' (predictive analysis) and the ASAQ long and
# short version
# INPUT
# data = dataset with one column "ASA" containing the list of ASAs followed with columns
# containing scores of all ASAQ's constructs/dimensions, which are calculated based on
# constructs = the id of 24 constructs/dimensions
#
# OUTPUT
# dataframe containing the ASAQ-scores of the ASAs based on the mean of
# the ASAQ's constructs/dimensions
calcCorrPredictiveAndASAQConstructsScores <- function(pred_constructScores,
                                                    long_constructScores,
                                                    short_constructScores,
                                                    constructs){

  dat <- data.frame()

  pred_constructScores <- convertToNumeric(pred_constructScores, 2)
  long_constructScores <- convertToNumeric(long_constructScores, 2)
  short_constructScores <- convertToNumeric(short_constructScores, 2)
  tasks <- c("PRED_LONG", "PRED_SHORT")

  for (i in 1:length(tasks)){
    dat[i, "COR"] <- tasks[i]
    for(construct in constructs){
      if (construct == "PF" | construct == "UAL"){
        dat[i, construct] <- ""
      }
      else if (i == 1)
        dat[i, construct] <- round(cor(cbind(pred_constructScores[, construct],
                                              long_constructScores[, construct]),
                                      method = "spearman",
                                      use = "complete.obs")[1, 2], 2)

      else if (i == 2)
        dat[i, construct] <- round(cor(cbind(pred_constructScores[, construct],
                                              short_constructScores[, construct]),
                                      method = "spearman",
                                      use = "complete.obs")[1, 2], 2)

    }
  }

  return(dat)
}

```

```

##### ASAQ CONSTRUCT SCORES based on COMBINED dataset (n=1066) #####
##### ASAQ Representative Set 2024 #####
# Get the score of all ASAQ constructs/dimensions of the short version of ASAQ.
# The score of each construct/dimension is taken from a representative item of the
# construct/dimension. This results in one score for each construct/dimension for each
# individual participant
# INPUT

```

```

# selectedItems: representative items of all ASAQ constructs/dimensions
# data: data of the long version of ASAQ,
# with one of the columns is "ASA" containing the list of ASAs followed columns of 90
# questionnaire items
# constructs: the list of ASAQ constructs/dimensions
#
# OUTPUT
# the constructs'/dimensions' scores of the short version of ASAQ
getShortASAQConstructsScore <- function(selectedItems, data, constructs){
  dat <- data[, names(data) %in% c("ASA", selectedItems)]
  names(dat) <- c("ASA", constructs)
  return(dat)
}

# Calculate the score of all ASAQ constructs/dimensions of the long version of ASAQ by
# calculating the mean of items in each construct/dimension. The calculation results in
# one score for each construct/dimension for each individual participant
# INPUT
# arr: the number of columns of items of each construct/dimension in the dataset
# data: data of the long version of ASAQ,
# with one of the columns is "ASA" containing the list of ASAs followed columns of 90
# questionnaire items
# constructs: the list of ASAQ constructs/dimensions
#
# OUTPUT
# the constructs'/dimensions' scores of the long version of ASAQ
getLongASAQConstructsScore <- function(arr, data, constructs){
  d <- data[, names(data)!="ASA"]
  d <- convertToNumeric(d, 1)

  dat <- data.frame()
  end <- 0
  for(i in 1:length(arr)){
    begin <- end + 1
    end <- end + arr[i]
    if (begin == end)
      dat[1:nrow(data), i] <- d[begin]
    else
      dat[1:nrow(data), i] <- rowMeans(d[begin:end])
  }

  dat <- cbind(data$ASA, dat)
  names(dat) <- c("ASA", constructs)
  return(dat)
}

# Calculate the constructs' score of each ASA by calculating the mean of each
# construct/dimension per ASA. The calculation results in one score for each
# construct/dimension per ASA
# INPUT:
# agents: the list of ASAs in the study
# data: data containing the scores of all ASAQ constructs/dimensions given by the
# participants (n=the number of participants) with one of the columns is "ASA"

```

```

# containing the list of ASAs followed columns of the number of ASAQ
# constructs/dimensions
# constructs: the list of ASAQ constructs/dimensions
#
# OUTPUT:
# the constructs'/dimensions' score of each ASA
calcASQscores <- function(agents, data){
  d_scoreASA <- data.frame()

  row <- 1
  for(agent in agents){
    d <- data[data$ASA == agent,]
    d_scoreASA[row, 1:ncol(data)] <-
      c(agent, colMeans(convertToNumeric(d,2)))
    row <- row + 1
  }

  names(d_scoreASA) <- colnames(data)
  return(d_scoreASA)
}

# Calculate the standard deviation of the ASAs' score
# based on the constructs/dimensions' score
# INPUT:
# agents: the list of ASAs in the study
# data: data containing the scores of all ASAQ constructs/dimensions given by the
# participants (n=the number of participants) with one of the columns is "ASA"
# containing the list of ASAs followed columns of the number of ASAQ
# constructs/dimensions
# constructs: the list of ASAQ constructs/dimensions
#
# OUTPUT:
# the standard deviation of each ASA's score
calcASQscores_StdDev <- function(agents, data){
  d_scoreASA <- data.frame()

  row <- 1
  for(agent in agents){
    d <- data[data$ASA == agent,]
    d_scoreASA[row, 1:ncol(data)] <-
      c(agent, sapply(convertToNumeric(d,2), sd))
    row <- row + 1
  }

  names(d_scoreASA) <- colnames(data)
  return(d_scoreASA)
}

# Calculate the total ASAQ-Scores of the ASAQ Representative Set 2024 based on the ASAQ
# long version
# INPUT:
# longASQscores = the scores of all ASAQ constructs/dimensions of the ASAQ Representative
# Set 2024. It has one column "ASA" containing the list of the ASAQ

```

```

# Representative Set 2024 followed by the scores of 24 ASAQ constructs/dimensions.
# The input data is calculated by calculating
# the mean the scores given by the participants (n=1066)
#
# OUTPUT:
# the total ASAQ-Scores of the ASAQ Representative Set 2024
calcTotalLongASAQscore <- function(longASAQscores){
  dat = data.frame(matrix(nrow = nrow(longASAQscores), ncol = 0))
  dat$ASA <- longASAQscores$ASA
  dat$ASAQ_SCORE <- round(rowSums(convertToNumeric(longASAQscores,2)),0)
  return(dat)
}

# Get the number of sample per ASA
# INPUT:
# data = raw dataset containing groups of participants based on ASA
# agents = the list of ASAs in the study
#
# OUTPUT:
# the number of sample of ASAs
getSampleASA <- function(agents, data){
  dat = data.frame()
  row = 1
  for (agent in agents){
    d <- data[data$ASA == agent,]
    dat[row, "ASA"] <- agent
    dat[row, "N"] <- nrow(d)
    row <- row + 1
  }
  return(dat)
}

##### Concurrent Validity Short Version #####
# Calculate the correlation and delta Score between the long and short version of ASAQ
# INPUT
# data = dataset (constructValData, crossValData and combinedData) with one column "ASA"
# containing the list of ASAs followed with columns containing scores of all ASAQ's
# constructs/dimensions, which are calculated based on the mean of the scores of items
# of the constructs/dimensions
# constructs = the id of 24 constructs/dimensions
#
# OUTPUT
# dataframe containing the correlation and delta Mean between the long and short ASAQ
# versions
getCorDiffLongShortASAQ <- function(longASAQ_constructValData,
                                     shortASAQ_constructValData,
                                     longASAQ_crossValData,
                                     shortASAQ_crossValData,
                                     longASAQ_combinedData,
                                     shortASAQ_combinedData,
                                     constructs){

  dat <- data.frame()

  # constructs/dimensions scores

```

```

longASAQ_constructValData <-
  convertToNumeric(longASAQ_constructValData, 2)
shortASAQ_constructValData <-
  convertToNumeric(shortASAQ_constructValData, 2)
longASAQ_crossValData <-
  convertToNumeric(longASAQ_crossValData, 2)
shortASAQ_crossValData <-
  convertToNumeric(shortASAQ_crossValData, 2)
longASAQ_combinedData <-
  convertToNumeric(longASAQ_combinedData, 2)
shortASAQ_combinedData <-
  convertToNumeric(shortASAQ_combinedData, 2)

for(i in 1:length(constructs)){
  dat[i, "CID"] <- constructs[i]
  dat[i, "COR_StudyM21"] <-
    round(cor(cbind(longASAQ_constructValData[, i],
                    shortASAQ_constructValData[, i]),
              method = "spearman",
              use = "complete.obs")[1, 2], 2)
  dat[i, "DM_StudyM21"] <-
    round(abs(mean(longASAQ_constructValData[, i]) -
              mean(shortASAQ_constructValData[, i])), 2)
  dat[i, "COR_Study22"] <-
    round(cor(cbind(longASAQ_crossValData[, i],
                    shortASAQ_crossValData[, i]), method = "spearman",
              use = "complete.obs")[1, 2], 2)
  dat[i, "DM_Study22"] <-
    round(abs(mean(longASAQ_crossValData[, i]) -
              mean(shortASAQ_crossValData[, i])), 2)
  dat[i, "COR_COMB"] <-
    round(cor(cbind(longASAQ_combinedData[, i],
                    shortASAQ_combinedData[, i]), method = "spearman",
              use = "complete.obs")[1, 2], 2)
  dat[i, "DM_COMB"] <-
    round(abs(mean(longASAQ_combinedData[, i]) -
              mean(shortASAQ_combinedData[, i])), 2)
}

return(dat)
}

```

```

##### Interpretability #####
# Calculate the frequency of 7-points of the participants' answer scale per
# construct/dimension of the short and long version of ASAQ
# INPUT
# arr = a list of the number of items for each individual ASAQ's construct/dimension
# data = with one column "ASA" containing the list of ASAs followed with columns
# containing participants' answers for all ASAQ's items/constructs/dimensions
# shortASAQdata = shortASAQConstructScore_combinedData, longASAQdata = combinedData
# constructs = the id of 24 constructs/dimensions
#
# OUTPUT
# The frequency of 7-points of the participants' answer scale per construct/dimension

```



```

# of the short and long version of ASAQ
calcFreqShortASAQConstructsScore <- function(shortASAQdata, scores, constructs){
  dat <- data.frame()
  title <- c("Min3", "Min2", "Min1", "Zero", "Plus1", "Plus2", "Plus3")
  shortASAQdata <- convertToNumeric(shortASAQdata, 2)

  for(i in 1:length(constructs)){
    dat[i,"CID"] <- constructs[i]
    for(j in 1:length(scores)){
      dat[i, title[j]] <-
        length(which(round(shortASAQdata[,i]) == scores[j]))
    }
  }
  return(dat)
}

calcFreqLongASAQConstructsScore <- function(arr, longASAQdata, scores, constructs){
  dat <- data.frame()
  title <- c("Min3", "Min2", "Min1", "Zero", "Plus1", "Plus2", "Plus3")
  longASAQdata <- convertToNumeric(longASAQdata, 2)

  end <- 0
  for(i in 1:length(arr)){
    begin <- end + 1
    end <- end + arr[i]
    d<- longASAQdata[begin:end]
    dat[i, "CID"] <- constructs[i]
    for(j in 1:length(scores)){
      dat[i, title[j]] <-
        length(which(d == scores[j]))
    }
  }
  return(dat)
}

# Calculate the relative frequency of 7-points of the participants' answer scale per
# construct/dimension of the short and long version of ASAQ
# INPUT
# arr = a list of the number of items for each individual ASAQ's construct/dimension
# data = with one column "ASA" containing the list of ASAs followed with columns
# containing participants' answers for all ASAQ's items/constructs/dimensions
# shortASAQdata = shortASAQConstructScore_combinedData, longASAQdata = combinedData
# constructs = the id of 24 constructs/dimensions
#
# OUTPUT
# The relative frequency of 7-points of the participants' answer scale per
# construct/dimension of the short and long version of ASAQ
calcRelFreqShortASAQConstructsScore <- function(shortASAQdata, scores, constructs){
  dat <- data.frame()
  title <- c("Min3", "Min2", "Min1", "Zero", "Plus1", "Plus2", "Plus3")
  shortASAQdata <- convertToNumeric(shortASAQdata, 2)

  for(i in 1:length(constructs)){
    dat[i,"CID"] <- constructs[i]

```

```

    for(j in 1:length(scores)){
      dat[i, title[j]] <-
        round(length(which(round(shortASAQdata[,i]) == scores[j])) /
              nrow(shortASAQdata),2)
    }
  }
  return(dat)
}

calcRelFreqLongASAQConstructsScore <- function(arr, longASAQdata, scores, constructs){
  dat <- data.frame()
  title <- c("Min3", "Min2", "Min1", "Zero", "Plus1", "Plus2", "Plus3")
  longASAQdata <- convertToNumeric(longASAQdata, 2)

  end <- 0
  for(i in 1:length(arr)){
    begin <- end + 1
    end <- end + arr[i]
    d<- longASAQdata[begin:end]
    dat[i, "CID"] <- constructs[i]
    for(j in 1:length(scores)){
      dat[i, title[j]] <-
        round(length(which(d == scores[j])) / (nrow(longASAQdata)*ncol(d)),2)
    }
  }
  return(dat)
}

# Drawing the frequency of total scores of the participants' answer scale for the short
# and long version of ASAQ
# INPUT
# data = (the sum of) participants' rate of all constructs/dimensions of the short
# and long version of ASAQ. The data contains a columns of the list of the ASAQ
# constructs/dimensions following (the sum of) the rates
# title = the title of the plot
#
# OUTPUT
# the bar chart of participants' total score
drawBarPlotFreqTotalASAQscore <- function(data, title, colStart){

  data <- convertToNumeric(data, colStart)
  xx <- barplot(table(rowSums(data)), col = "steelblue",
               ylim =c(0, max(table(rowSums(data))))))
  title(main = title,
        xlab = 'Sum of Item Scores', ylab = 'Frequency')
}

# Drawing the histogram of the ASAQ scores of the short and long version of ASAQ
# INPUT
# data = the ASAQ scores of all constructs/dimensions of the short and long version of
# ASAQ. The dataset has a column containing the list of ASAs followed by columns of all
# ASAQ constructs/dimensions
# title = the title of the plot
#

```

```

# OUTPUT
# the histogram of the ASAQ scores
drawHistogramASAScore <- function(data, title, startCol){

  data <- as.matrix(convertToNumeric(data, startCol))
  h<-hist(data,
    main=title,
    xlab="ASAQ Scores",
    xlim=c(min(data)-0.5,max(data)+0.5),
    col="steelblue",
    breaks = seq(min(data)-0.5,max(data)+0.5,1),
    freq = FALSE
  )
}

# Calculate the sum of item scores of long ASAQ constructs/dimensions
# INPUT:
# arr = the number of items of each construct/dimension
# longASAScore = 90 items of long ASAQ
# constructs = the id of 24 ASAQ constructs/dimensions
#
# OUTPUT:
# the sum of item scores of ASAQ constructs/dimensions
calcSumLongConstructASAScore <- function(arr, longASAScore, constructs){
  dat <- data.frame(matrix(nrow = nrow(longASAScore), ncol = 0))
  longASAScore <- convertToNumeric(longASAScore, 2)

  end <- 0
  for(i in 1:length(arr)){
    begin <- end + 1
    end <- end + arr[i]
    d<- longASAScore[begin:end]
    dat[, constructs[i]] <- rowSums(d)
  }

  return(dat)
}

# Calculate the relative frequency of lowest and highest extreme of each long ASAQ
# constructs/dimensions
# INPUT:
# arr = the number of items of each construct/dimension
# sumASAScores = the sum of item scores of long ASAQ constructs/dimensions
# constructs = the id of 24 ASAQ constructs/dimensions
#
# OUTPUT:
# The relative frequency of lowest and highest extreme of each long ASAQ
# constructs/dimensions
calcRelFreqExtrmLongASAScoreConstructScores <- function(arr, sumASAScores, constructs){
  dat <- data.frame()

  for(i in 1:length(constructs)){
    dat[i, "CID"] <- constructs[i]

```

```

dat[i, "ITEM_NUM"] <- arr[i]
dat[i, "fRelLowEXTRM"] <-
  round(length(which(sumASAScores[,i] == (arr[i] * -3))) /
    nrow(sumASAScores), 2)
dat[i, "fRelHighEXTRM"] <-
  round(length(which(sumASAScores[,i] == (arr[i] * 3))) /
    nrow(sumASAScores), 2)
dat[i, "SumfRelEXTRM"] <-
  round(length(which(sumASAScores[,i] == (arr[i] * -3))) /
    nrow(sumASAScores), 2) +
  round(length(which(sumASAScores[,i] == (arr[i] * 3))) /
    nrow(sumASAScores), 2)
}

return(dat)
}

# Calculate the percentile scores of ASAS-Scores of the ASAS Representative Set 2024 based
# on the short and long
# versions of ASAS
# INPUT:
# data = the scores of all ASAS constructs/dimensions of the ASAS Representative Set 2024.
# It has one column "ASA" containing the list of the ASAS Representative Set 2024
# followed by the scores of 24 ASAS constructs/dimensions. The input data is calculated
# by calculating the mean of the scores given by the participants (n=1066)
# probs = the percentile
# constructs = the id of 24 ASAS constructs/dimensions
# header = the header of the output dataframe
# OUTPUT:
# the percentile scores of ASAS-Scores of the ASAS Representative Set 2024
calcPercentileASASscore <- function(data, probs, constructs, header){
  dat = data.frame()
  data <- convertToNumeric(data, 2)

  row <- 1
  for (i in 1:length(constructs)){
    dat[row, "CID"] <- constructs[i]
    for (j in 1:length(probs)){
      dat[row, header[j]] <- round(quantile(data[, i], probs[j]), 2)
    }
    row = row + 1
  }
  return(dat)
}

# Calculate the percentile of the total ASAS-Scores of the ASAS Representative Set 2024
# based on the short and long versions of ASAS
# INPUT:
# shortASASscore, longASASscore = the scores of all ASAS constructs/dimensions of
# ASAS Representative Set 2024. It has one column "ASA" containing the list of the ASAS
# Representative Set 2024 followed by the scores of 24 ASAS constructs/dimensions.
# The input data is calculated by calculating the mean of the scores given by the
# participants (n=1066)

```

```

# probs = the percentile
# constructs = the id of 24 ASAQ constructs/dimensions
# header = the header of the output dataframe
#
# OUTPUT:
# the percentile total ASAQ-Scores of the ASAQ Representative Set 2024
calcPercentileTotalASAQscore <- function(longASAQscore, shortASAQscore, probs,
                                         header){

  dat = data.frame()
  dat[1, "ASAQ"] <- "Long"
  longASAQscore <- rowSums(convertToNumeric(longASAQscore, 2))
  for (j in 1:length(probs)){
    dat[1, header[j]] <- round(quantile(longASAQscore, probs[j]),2)
  }
  dat[2, "ASAQ"] <- "Short"
  shortASAQscore <- rowSums(convertToNumeric(shortASAQscore, 2))
  for (j in 1:length(probs)){
    dat[2, header[j]] <- round(quantile(shortASAQscore, probs[j]),2)
  }
  return(dat)
}

# Calculate the absolute difference of ASAQ-scores of all combinations of
# the ASAQ Representative Set 2024 for each construct and dimension based on the short
# and long versions of the ASAQ
# INPUT:
# agents = the names of the ASAQ Representative Set 2024
# data = the scores of all ASAQ constructs/dimensions of the ASAQ Representative Set 2024.
# It has one column "ASA" containing the list of the ASAQ Representative Set 2024
# followed by the scores of 24 ASAQ constructs/ dimensions. The input data is
# calculated using the mean of the scores given by the participants (n=1066)
# constructs = the id of 24 ASAQ constructs/dimensions
#
# OUTPUT:
# the difference of ASAQ-scores of all combinations of the ASAQ Representative Set 2024
# for each construct and dimension based on the short and long version of the ASAQ
calcDiffTwoASAQscore <- function(agents, data, constructs){
  dat <- data.frame()

  row <- 1
  for (i in 1:length(agents)){
    agent1 = agents[i]
    for (j in 1:length(agents)){
      agent2 = agents[j]
      if (agent1 != agent2){
        dat[row, "ASA"] <- paste(agent1, agent2, sep="_")
        for(construct in constructs){
          score1 <- as.numeric(data[data$ASA == agent1, construct])
          score2 <- as.numeric(data[data$ASA == agent2, construct])
          dat[row, construct] <- round(abs(score1 - score2), 2)
        }
        row <- row + 1
      }
    }
  }
}

```

```

    }
  }

  return(dat)
}

##### Sample Size #####
# Calculate the effect size of ASAQ based on the scores of all ASAQ constructs/
# dimensions of the ASAQ Representative Set 2024 based on the short and long version of
# the ASAQ. The calculation of the effect size is based on Cohen's d (Wikipedia
# "Effect Size" https://en.wikipedia.org/wiki/Effect\_size#Cohen's\_d)
#
# INPUT:
# agents = the names of the ASAQ Representative Set 2024
# data = the scores of all ASAQ constructs/dimensions of the ASAQ Representative Set 2024.
# It has one column "ASA" containing followed by the scores of 24 ASAQ
# constructs/dimensions from given by participants. In particular, the long version
# data is by calculating the mean of the scores of the construct/dimension's items
# given by the participants (n=1066)
# constructs = the id of 24 ASAQ constructs/dimensions
#
# OUTPUT:
# the effect size for each construct and dimension based on the short and long version
# of the ASAQ
calcEffectSizeASQConstructScore <- function(agents, constructScoreData, constructs){
  dat <- data.frame()

  row <- 1
  for (i in 1:length(agents)){
    agent1 = agents[i]
    for (j in 1:length(agents)){
      agent2 = agents[j]
      if (agent1 != agent2){
        dat[row, "ASA"] <- paste(agent1, agent2, sep="_")
        d_agent1 <- as.data.frame(sapply(
          constructScoreData[constructScoreData$ASA == agent1,
                               2:ncol(constructScoreData)], as.numeric))
        d_agent2 <- as.data.frame(sapply(
          constructScoreData[constructScoreData$ASA == agent2,
                               2:ncol(constructScoreData)], as.numeric))

        n1 <- nrow(d_agent1)
        n2 <- nrow(d_agent2)
        M1 <- apply(d_agent1, 2, mean)
        M2 <- apply(d_agent2, 2, mean)
        V1 <- apply(d_agent1, 2, var)
        V2 <- apply(d_agent2, 2, var)
        for(i in 1:length(constructs)){
          diff <- abs(M1[i] - M2[i])
          #s = pooled standard deviation of two independent sample
          s <- sqrt(((n1-1)*V1[i])+((n2-1)*V2[i]))/(n1+n2-2))
          dat[row, constructs[i]]<- (diff / s)
        }
        row <- row + 1
      }
    }
  }
}

```

```

    }
  }

  return(dat)
}

# Calculate minimum sample size for comparing two or more agents based on the power
# analysis
# INPUT:
# percentileEffectSizeASAPConstructScore = the percentile scores of all ASAP
# constructs/dimensions of the ASAP Representative Set 2024. It has one column "CID"
# containing the list of 24 ASAP constructs/dimensions followed by the percentile
# scores of the ASAP constructs/dimensions
# constructs = the id of 24 ASAP constructs/dimensions
# header = the header of the output dataframe (the string version of the percentile -
# probs)
#
# OUTPUT:
# the minimum sample size per ASAP construct/dimension
calcSamplePowerAnalysisASAPConstructScore <-
  function(percentileEffectSizeASAPConstructScore, constructs, headers){
    dat <- data.frame()

    percentileEffectSizeASAPConstructScore <-
      convertToNumeric(percentileEffectSizeASAPConstructScore, 2)

    for (i in 1:length(constructs)){
      dat[i,"CID"] <- constructs[i]

      for (header in headers){
        effectSize <- percentileEffectSizeASAPConstructScore[i, header]
        pwr <- pwr.t.test(d = effectSize, power = 0.80, sig.level = 0.05,
                          type = "two.sample", alternative = "two.sided")
        dat[i, header] <- round(pwr$n,0)
      }
    }

    return(dat)
  }

# Calculate maximum sample size for comparing two ASAAs based on the power analysis
# INPUT:
# sampleSizeLongASAP = minimum sample size for the long ASAP
# sampleSizeShortASAP = minimum sample size for the short ASAP
#
# OUTPUT:
# the maximum sample size per percentile in interest for both long and short versions of
# ASAP
getMaxSamplePowerAnalysis <- function(sampleSizeLongASAP, sampleSizeShortASAP){
  dat <- data.frame(matrix(nrow = 2, ncol = 4))
  sampleSizeLong <- convertToNumeric(sampleSizeLongASAP, 2)
  sampleSizeShort <- convertToNumeric(sampleSizeShortASAP, 2)

```

```

dat[1,] <- c("Long ASAQ", apply(convertToNumeric(sampleSizeLongASAQ,2), 2, max))
dat[2,] <- c("Short ASAQ", apply(convertToNumeric(sampleSizeShortASAQ,2), 2, max))
names(dat) <- c("Version", colnames(sampleSizeLong))

return(dat)
}

# Calculate minimum sample size for evaluating an ASA based on the confident interval
# INPUT:
# agents = the list of the ASAQ Representative Set 2024
# constructScoreData = the ASAQ scores of 24 ASAQ constructs/dimensions given by the
# participants (shape: 1066 x 24)
# ASAQscoreData = the ASAQ scores of all ASAQ constructs/dimensions of the ASAQ
# Representative Set 2024 (shape: 29 x 24). Both constructScoreData and ASAQscoreData
# have one column "ASA" then followed by the column of the scores of 24 ASAQ
# constructs/dimensions
# probs = the percentile distance of interest, e.g., 5th, 10th, 20th or 25th
# constructs = the id of 24 ASAQ constructs/dimensions
# Note: Calculate the std_err using mean of difference in percentile
#
# OUTPUT:
# the minimum sample size per ASAQ construct/dimension
calcSampleConfIntervASAQConstructScore <-
  function(agents, constructScoreData, ASAQscoreData, probs, constructs){

    ##calculate variance of agents
    d_N <- c() #the number of samples with the same agent

    #the variance of constructs/dimension per agents
    d_vars <- data.frame(matrix(nrow = length(agents), ncol = length(constructs)))
    for (i in 1:length(agents)){
      d_agent <- as.data.frame(apply(constructScoreData[
        constructScoreData$ASA == agents[i], 2:ncol(constructScoreData)],
        as.numeric))
      d_N[i] <- nrow(d_agent)
      d_vars[i,] <- apply(d_agent, 2 , var)
    }

    ##calculate s_pooled
    d_s_pooled <- c() #the s_pooled of constructs/dimension
    for (j in 1:length(constructs)){
      V_pooled <- 0
      N <- 0
      for(k in 1:length(agents)){
        V_pooled <- V_pooled + ((d_N[k] - 1) * d_vars[k, j])
      }
      d_s_pooled[j]<- sqrt(V_pooled / (nrow(constructScoreData) - length(agents)))
    }

    ##calculate percentile (probs) of the ASAQ scores the ASAQ constructs/dimensions
    d_perc = data.frame()
    ASAQscoreData <- convertToNumeric(ASAQscoreData, 2)
    for (p in 1:length(constructs)){

```



```

    for (q in 1:length(probs)){
      d_perc[p, q] <- quantile(ASAQscoreData[, p], probs[q])
    }
  }

  ##calculate std_err
  d_stdErr = c()
  diffPerc = c()      #the difference of percentile scores
  nDiff = ncol(d_perc) - 1 #the number of difference in the interest
  for (r in 1:length(constructs)){
    for(n in 1:(nDiff)){
      perc1 <- d_perc[r, n]
      perc2 <- d_perc[r, n + 1]
      diffPerc[n] <- abs(perc2 - perc1)
    }
    d_stdErr[r]<-mean(diffPerc)
  }

  #calculate minimal sample size based on the confident interval
  Z <- Normal(0,1)
  d_samples = data.frame()
  confs = c(0.90, 0.95, 0.975, 0.99)      #choices confidence interval
  headers <- c("90CI", "95CI", "97.5CI", "99CI") #table headers
  for (m in 1:length(constructs)){
    d_samples[m,"CID"] <- constructs[m]
    for (l in 1:length(confs)){
      zScore <- quantile(Z, (1-confs[l]) / 2)
      sampleSize = ((zScore * d_s_pooled[m]) /d_stdErr[m])^2
      d_samples[m, headers[l]] <- round(sampleSize, 0)
    }
  }
  return(d_samples)
}

# Calculate maximum sample size for evaluating an ASA based on the confident interval
# INPUT:
# long5thData, short5thData, long10thData, short10thData, long20thData, short20thData,
# long25thData, short25thData = minimum sample size for evaluating an ASA based on the
# confident interval calculated on respectively on the 5th, 10th, 20th and 25th
# percentiles
#
# OUTPUT:
# the maximum sample size per percentile in interest for both long and short versions of
# ASAQ
getMaxSampleSizeConfidenceIntervalASAQConstructScore <-
function(long5thData, short5thData, long10thData, short10thData,
         long20thData, short20thData, long25thData, short25thData){

  long5thData <- convertToNumeric(long5thData,2)
  short5thData <- convertToNumeric(short5thData,2)
  long10thData <- convertToNumeric(long10thData,2)
  short10thData <- convertToNumeric(short10thData,2)
  long20thData <- convertToNumeric(long20thData,2)

```

```

short20thData <- convertToNumeric(short20thData,2)
long25thData <- convertToNumeric(long25thData,2)
short25thData <- convertToNumeric(short25thData,2)

#table headers
headers <- c("PERC", "LONG_90CI", "LONG_95CI", "LONG_97.5CI", "LONG_99CI",
             "SHORT_90CI", "SHORT_95CI", "SHORT_97.5CI", "SHORT_99CI")

percs <- c("5%", "10%", "20%", "25%")
dat <- data.frame(matrix(nrow = 4, ncol = length(headers)))
dat[1, ] <- c(percs[1], apply(long5thData, 2, max), apply(short5thData, 2, max))
dat[2, ] <- c(percs[2], apply(long10thData, 2, max), apply(short10thData, 2, max))
dat[3, ] <- c(percs[3], apply(long20thData, 2, max), apply(short20thData, 2, max))
dat[4, ] <- c(percs[4], apply(long25thData, 2, max), apply(short25thData, 2, max))
colnames(dat) <- headers

return(dat)
}

```

1 Introduction

- Previous work: Publications, documents and result data

2 The ASA Questionnaire

3 Creation of the Questionnaire

3.1 Constructs and Dimensions

Table 2: The number of constructs classified into the single anchor points and anchor pairs.

```
calcExpertVotingClassification(constructDefData)
```

##	AnchorPoint	SingleAnchor	PairAnchors
## 1	Agent's properties	4	24
## 2	Agent's social traits	27	30
## 3	Agent's role	9	23
## 4	Impression of the agent after the interaction	1	2
## 5	Interaction quality	12	32
## 6	Human's impression of the interaction	23	49
## 7	Human's attributes	2	6

```

unclassified = length(which(constructDefData$SingleCls == "None"))
other = length(which(constructDefData$SingleCls == "Other"))

```

... This left us with 177 (93.70%) for detailed grouping, (189 starting - 1 unclassified - 11 out of scope = 177).

3.2 Questionnaire Items

4 Characteristics of the ASAQ

4.1 Reliability of the ASAQ

```
# Study Early '21: Reliability Analysis (n=192)
longASAQ_reliabilityData <-
  getLongASAQ(reliabilityData, itemASAQdesc$ITEM_ID, itemASAQdesc$ITEM_REF, 1)
longASAQ_reliabilityData_1stPOV_part_1 <-
  getLongASAQ(reliabilityData_1stPOV_part_1, itemASAQdesc$ITEM_ID[1:44],
    itemASAQdesc$ITEM_REF[1:44], 1)
longASAQ_reliabilityData_1stPOV_part_1 <-
  getLongASAQ(reliabilityData_1stPOV_part_1, itemASAQdesc$ITEM_ID[1:44],
    itemASAQdesc$ITEM_REF[1:44], 1)
longASAQ_reliabilityData_1stPOV_part_2 <-
  getLongASAQ(reliabilityData_1stPOV_part_2, itemASAQdesc$ITEM_ID[45:90],
    itemASAQdesc$ITEM_REF[45:90], 1)
longASAQ_reliabilityData_3rdPOV_part_1 <-
  getLongASAQ(reliabilityData_3rdPOV_part_1, itemASAQdesc$ITEM_ID[1:44],
    itemASAQdesc$ITEM_REF[1:44], 1)
longASAQ_reliabilityData_3rdPOV_part_2 <-
  getLongASAQ(reliabilityData_3rdPOV_part_2, itemASAQdesc$ITEM_ID[45:90],
    itemASAQdesc$ITEM_REF[45:90], 1)
alpha_reliabilityData <- getAlphaReliability(
  calcAlpha(constructASAQdesc$NUM_ITEMS, longASAQ_reliabilityData,
    constructASAQdesc$CABR),
  calcAlpha(constructASAQdesc$NUM_ITEMS[1:12], longASAQ_reliabilityData_1stPOV_part_1,
    constructASAQdesc$CABR[1:12]),
  calcAlpha(constructASAQdesc$NUM_ITEMS[13:24], longASAQ_reliabilityData_1stPOV_part_2,
    constructASAQdesc$CABR[13:24]),
  calcAlpha(constructASAQdesc$NUM_ITEMS[1:12], longASAQ_reliabilityData_3rdPOV_part_1,
    constructASAQdesc$CABR[1:12]),
  calcAlpha(constructASAQdesc$NUM_ITEMS[13:24], longASAQ_reliabilityData_3rdPOV_part_2,
    constructASAQdesc$CABR[13:24]))

# Study Mid 2021: Construct Validity (n=532)
longASAQ_constructValData <-
  getLongASAQ(constructValData, itemASAQdesc$ITEM_ID, itemASAQdesc$ITEM_REF, 2)
longASAQ_constructValData_std <-
  getStandardize(name14ASAs, longASAQ_constructValData)
alpha_constructValData <-
  calcAlpha(constructASAQdesc$NUM_ITEMS, longASAQ_constructValData_std,
    constructASAQdesc$CABR)

# Study 2022: Cross Validity (n=534)
longASAQ_crossValData <-
  getLongASAQ(crossValData, itemASAQdesc$ITEM_ID, itemASAQdesc$ITEM_REF, 2)
```

```

longASAQ_crossValData_std <-
  getStandardize(name15ASAs, longASAQ_crossValData)
alpha_crossValData <-
  calcAlpha(constructASAQdesc$NUM_ITEMS, longASAQ_crossValData_std,
            constructASAQdesc$CABR)

# Combination of study Mid 2021 and study 2022
combinedData_std = rbind(longASAQ_constructValData_std, longASAQ_crossValData_std)
alpha_combinedData <-
  calcAlpha(constructASAQdesc$NUM_ITEMS, combinedData_std, constructASAQdesc$CABR)

alphas <-
  cbind(alpha_reliabilityData[,1:4], alpha_constructValData$std.a,
        alpha_crossValData$std.a, alpha_combinedData$std.a)
names(alphas) <-
  c("CID", "STUDYE21_all", "STUDYE21_1stPOV", "STUDYE21_3rdPOV", "STUDYM21",
    "STUDY22", "COMBINED_DATA")

```

... We combine the data of these last two to efficiently reach a large n, see Table 3. The combined studies have values ranging from 0.60 to 0.86, with an average of 0.72, which is classified as a respectable reliability ... The differences in reliability between the first-person and third-person perspectives are not substantial, see Table 3 (column Study Early '21). The mean absolute difference between the two perspectives seems small ($M = 0.12$, $SD = 0.12$) and ...

Table 3: Reliability analysis results (Cronbach's alpha) of three studies.

```

showTable(alphas, constructASAQdesc$CABR, constructASAQdesc$CNAME,
          0, 0, 0, 0, 0, 0, 0)

```

##	CABR	CNAME	STUDYE21_all	STUDYE21_1stPOV
## 1	HLA	Human-Like Appearance	0.86	0.88
## 2	HLB	Human-Like Behavior	0.82	0.83
## 3	NA	Natural Appearance	0.70	0.66
## 4	NB	Natural Behavior	0.56	0.58
## 5	AAS	Agent's Appearance Suit.	0.77	0.74
## 6	AU	Agent's Usability	0.80	0.77
## 7	PF	Performance	0.53	0.72
## 8	AL	Agent's Likeability	0.84	0.84
## 9	AS	Agent's Sociability	0.56	0.66
## 10	APP	Agent's Personality Pr.	0.55	0.38
## 11	UAA	User Acceptance of the A.	0.65	0.74
## 12	AE	Agent's Enjoyability	0.76	0.79
## 13	UE	User's Engagement	0.65	0.88
## 14	UT	User's Trust	0.62	0.83
## 15	UAL	User Agent Alliance	0.72	0.66
## 16	AA	Agent's Attentiveness	0.79	0.58
## 17	AC	Agent's Coherence	0.75	0.74
## 18	AI	Agent's Intentionality	0.70	0.77
## 19	AT	Attitude	0.74	0.72
## 20	SP	Social Presence	0.71	0.84
## 21	IIS	Interaction Impact on Self.	0.79	0.66
## 22	AEI	Agent's Emotional Int. Pr.	0.87	0.38
## 23	UEP	User's Emotion Presence	0.81	0.74
## 24	UAI	User Agent Interplay	0.70	0.79
##	STUDYE21_3rdPOV	STUDYM21	STUDY22	COMBINED_DATA

## 1	0.84	0.83	0.86	0.84
## 2	0.80	0.84	0.85	0.84
## 3	0.73	0.73	0.73	0.73
## 4	0.55	0.60	0.60	0.60
## 5	0.80	0.73	0.72	0.72
## 6	0.83	0.72	0.81	0.77
## 7	0.30	0.68	0.64	0.66
## 8	0.84	0.80	0.80	0.80
## 9	0.47	0.71	0.66	0.69
## 10	0.64	0.60	0.61	0.60
## 11	0.51	0.65	0.67	0.66
## 12	0.74	0.70	0.77	0.74
## 13	0.84	0.70	0.65	0.68
## 14	0.80	0.66	0.67	0.67
## 15	0.73	0.77	0.78	0.77
## 16	0.55	0.72	0.70	0.71
## 17	0.80	0.67	0.69	0.68
## 18	0.83	0.68	0.73	0.70
## 19	0.30	0.78	0.81	0.79
## 20	0.84	0.71	0.62	0.67
## 21	0.47	0.72	0.73	0.73
## 22	0.64	0.86	0.85	0.86
## 23	0.51	0.64	0.69	0.66
## 24	0.74	0.67	0.61	0.64

4.2 Content Validity of the ASAQ

```
items_contentValData <- calcItemAgreementCoeff(itemASAQdesc, contentValData)
construct_contentValData <-
  calcConstructAgreementCoeff(constructASAQdesc, items_contentValData)
```

Table 4 shows the average TPR_s and FPR_s of the ASAQ constructs. The TPR_s ranges from 0.81 to 1.00 with an average of 0.95 (SD = 0.04), while FPR_s ranges from 0.00 to 0.25 with an average of 0.12, (SD = 0.06). ...

Table 4: The average TPR_s and FPR_s for each ASAQ construct or dimension.

```
showTable(construct_contentValData, constructASAQdesc$CABR, constructASAQdesc$CNAME,
          0, 0, 0, 0, 0, 0, 0)
```

##	CABR	CNAME	M_TPRs	SD_TPRs	M_FPRs	SD_FPRs
## 1	HLA	Human-Like Appearance	0.96	0.05	0.06	0.08
## 2	HLB	Human-Like Behavior	0.98	0.04	0.21	0.07
## 3	NA	Natural Appearance	0.92	0.05	0.08	0.08
## 4	NB	Natural Behavior	0.97	0.05	0.22	0.03
## 5	AAS	Agent's Appearance Suit.	0.93	0.12	0.07	0.06
## 6	AU	Agent's Usability	0.94	0.05	0.03	0.06
## 7	PF	Performance	0.91	0.09	0.13	0.12
## 8	AL	Agent's Likeability	0.81	0.05	0.11	0.07
## 9	AS	Agent's Sociability	0.95	0.05	0.13	0.12
## 10	APP	Agent's Personality Pr.	0.97	0.05	0.14	0.10
## 11	UAA	User Acceptance of the A.	0.96	0.07	0.08	0.08
## 12	AE	Agent's Enjoyability	0.96	0.05	0.17	0.05
## 13	UE	User's Engagement	0.89	0.11	0.12	0.06

## 14	UT	User's Trust	0.97	0.06	0.04	0.07
## 15	UAL	User Agent Alliance	0.97	0.05	0.19	0.07
## 16	AA	Agent's Attentiveness	1.00	0.00	0.20	0.05
## 17	AC	Agent's Coherence	0.96	0.05	0.14	0.10
## 18	AI	Agent's Intentionality	0.95	0.06	0.10	0.07
## 19	AT	Attitude	1.00	0.00	0.15	0.09
## 20	SP	Social Presence	0.93	0.06	0.25	0.04
## 21	IIS	Interaction Impact on Self.	0.95	0.06	0.00	0.00
## 22	AEI	Agent's Emotional Int. Pr.	1.00	0.00	0.10	0.07
## 23	UEP	User's Emotion Presence	0.98	0.04	0.11	0.04
## 24	UAI	User Agent Interplay	0.95	0.05	0.04	0.07

4.3 Construct Validity of the ASAQ

We took two sequential analyses: 1. Convergent validity analysis. 2. Discriminant validity analysis.

Both analyses are reported in:

- Siska Fitrianie, Merijn Bruijnes, Fengxiang Li, Amal Abdulrahman, and Willem-Paul Brinkman. 2022. The Artificial-Social-Agent Questionnaire: Establishing the long and short questionnaire versions. In ACM International Conference on Intelligent Virtual Agents (IVA'22), September, 2022, Faro, Portugal. ACM, New York, NY, USA. DOI: <https://doi.org/10.1145/3514197.3549612>
- Related scripts and dataset can be found at:
 - Siska Fitrianie, Merijn Bruijnes, Fengxiang Li, Amal Abdulrahman, and Willem-Paul Brinkman. 2022. Data and analysis underlying the research into the Artificial-Social-Agent Questionnaire: Establishing the long and short questionnaire versions. (2022). <https://doi.org/10.4121/19758436> 4TU.ResearchData.

4.4 Cross-Validity of the ASAQ

Using the obtained data (n=534), we again took two sequential analyses:

1. Convergent validity analysis.

```
individualFactorialModel_longASAQ <-
c('C_HLA =~ HLA1 + HLA2 + HLA3 + HLA4',
  'C_HLB =~ HLB1 + HLB2 + HLB3 + HLB4 + HLB5',
  'C_NA =~ NA1 + NA2 + NA3 + NA4 + NA5',
  'C_NB =~ NB1 + NB2 + NB3',
  'C_AAS =~ AAS1 + AAS2 + AAS3',
  'C_AU =~ AU1 + AU2 + AU3',
  'C_PF =~ PF1 + PF2 + PF3',
  'C_AL =~ AL1 + AL2 + R_AL3 + AL4 + AL5',
  'C_AS =~ AS1 + AS2 + AS3',
  'C_APP =~ APP1 + R_APP2 + APP3',
  'C_UAA =~ UAA1 + UAA2 + R_UAA3',
  'C_AE =~ R_AE1 + AE2 + AE3 + R_AE4',
  'C_UE =~ UE1 + UE2 + UE3',
  'C_UT =~ UT1 + UT2 + UT3',
  'C_UAL =~ UAL1 + UAL2 + UAL3 + UAL4 + UAL5 + UAL6',
  'C_AA =~ AA1 + AA2 + AA3',
  'C_AC =~ R_AC1 + R_AC2 + R_AC3 + R_AC4',
  'C_AI =~ AI1 + AI2 + R_AI3 + AI4',
  'C_AT =~ AT1 + AT2 + R_AT3',
```

```

'C_SP =~ SP1 + SP2 + SP3',
'C_IIS =~ IIS1 + IIS2 + IIS3 + IIS4',
,
C_AEI =~ AEI1 + AEI2 + R_AEI3 + AEI4 + R_AEI5
C_UEP =~ UEP1 + UEP2 + UEP3 + UEP4
,
'C_UAI =~ UAI1 + UAI2 + UAI3 + UAI4'
)

convergentValmodels <-
c("Human-Like Appearance", "Human-Like Behavior", "Natural Appearance",
  "Natural Behavior", "Agent's Appearance Suit.", "Agent's Usability",
  "Performance", "Agent's Likeability", "Agent's Sociability",
  "Agent's Personality Pr.", "User Acceptance of the A.", "Agent's Enjoyability",
  "User's Engagement", "User's Trust", "User Agent Alliance", "Agent's Attentiveness",
  "Agent's Coherence", "Agent's Intentionality", "Attitude", "Social Presence",
  "Interaction Impact on Self.", "User's and Agent's Emotional State",
  "User Agent Interplay")

# CFI scores of convergent validation of the long ASAQ
cfi_convergentVal_crossValData <-
  getCFIcoefficient(convergentValmodels, individualFactorialModel_longASAQ,
    longASAQ_crossValData_std)

```

... The 24 CFAs (i.e. 16 constructs and 8 dimensions) as part of the convergence analysis, showed good model fit with, on average, a CFI score of 0.99 (SD = 0.02, range [0.94 .. 1.00]). ...

2. Discriminant validity analysis.

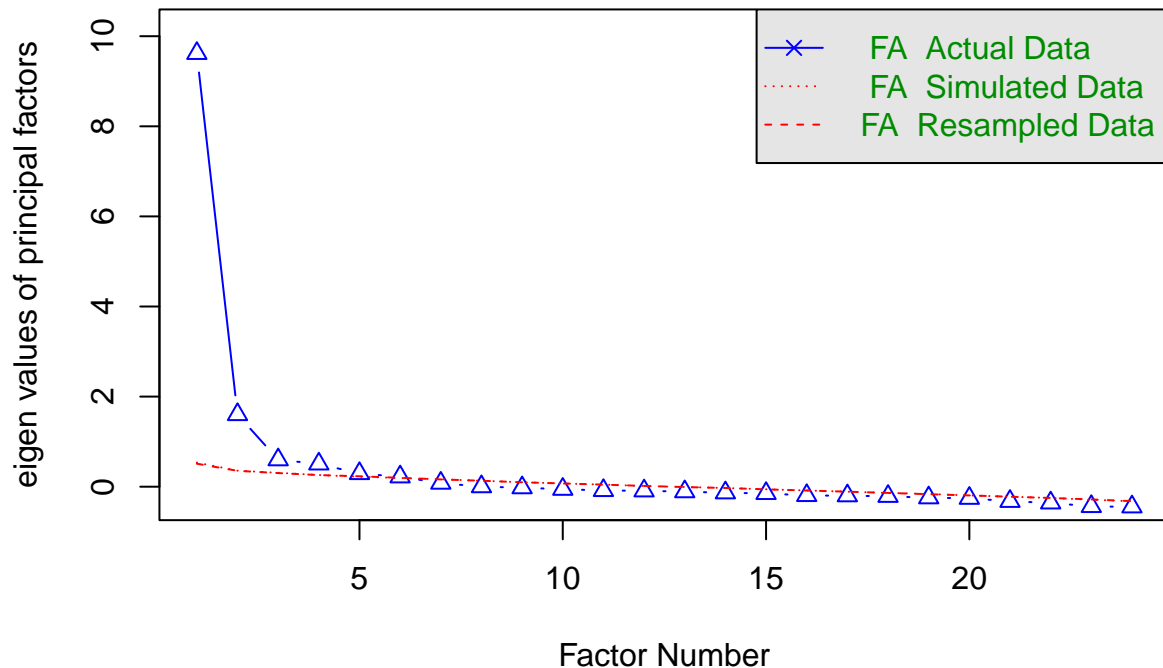
```

# Retrieve predicted scores of long ASAQ constructs/dimensions based on
# their individual CFA result
predicted_crossValData <-
  getPredictedData_longASAQ(individualFactorialModel_longASAQ, longASAQ_crossValData_std)

# Determine Number of Factors to Extract
parallel_crossValData <-
  fa.parallel(predicted_crossValData, fm='ml', fa='fa')

```

Parallel Analysis Scree Plots



Parallel analysis suggests that the number of factors = 6 and the number of components = NA

Note: The suggestion of parallel analysis may vary between 5 to 6 factors. In this study we took 6 factorial models (the largest) because we try to avoid the co-linearity issue between constructs when we run the factor analysis, while at the same time we still allow the association between these constructs.

```
# Maximum Likelihood Factor Analysis
efa_crossValData <- getEFA(predicted_crossValData, 6)

# Factorial models based on efa_crossValData
factorialModels_crossValData <- c(
  '
  C_HLB =~ HLB1 + HLB2 + HLB3 + HLB4 + HLB5
  C_NB =~ NB1 + NB2 + NB3
  C_AS =~ AS1 + AS2 + AS3
  C_APP =~ APP1 + R_APP2 + APP3
  C_SP =~ SP1 + SP2 + SP3
  C_AEI =~ AEI1 + AEI2 + R_AEI3 + AEI4 + R_AEI5
  C_UEP =~ UEP1 + UEP2 + UEP3 + UEP4
  C_UAI =~ UAI1 + UAI2 + UAI3 + UAI4
  ',
  '
  C_UAA =~ UAA1 + UAA2 + R_UAA3
  C_AE =~ R_AE1 + AE2 + AE3 + R_AE4
  C_AT =~ AT1 + AT2 + R_AT3
  C_IIS =~ IIS1 + IIS2 + IIS3 + IIS4
  ',
  '
  C_UE =~ UE1 + UE2 + UE3
  C_AA =~ AA1 + AA2 + AA3
  '
)
```



```

C_UEP =~ UEP1 + UEP2 + UEP3 + UEP4
C_UAI =~ UAI1 + UAI2 + UAI3 + UAI4
',
'

C_AU =~ AU1 + AU2 + AU3
C_PF =~ PF1 + PF2 + PF3
C_UT =~ UT1 + UT2 + UT3
C_UAL =~ UAL1 + UAL2 + UAL3 + UAL4 + UAL5 + UAL6
C_AC =~ R_AC1 + R_AC2 + R_AC3 + R_AC4
C_AI =~ AI1 + AI2 + R_AI3 + AI4
',
'

C_HLA =~ HLA1 + HLA2 + HLA3 + HLA4
C_NA =~ NA1 + NA2 + NA3 + NA4 + NA5
C_AAS =~ AAS1 + AAS2 + AAS3
',
'

C_AAS =~ AAS1 + AAS2 + AAS3
C_AL =~ AL1 + AL2 + R_AL3 + AL4 + AL5
C_AE =~ R_AE1 + AE2 + AE3 + R_AE4
C_AC =~ R_AC1 + R_AC2 + R_AC3 + R_AC4
'
)
discriminantValModels_crossValData <- c(
  "FACTORIAL MODEL 1", "FACTORIAL MODEL 2", "FACTORIAL MODEL 3", "FACTORIAL MODEL 4",
  "FACTORIAL MODEL 5", "FACTORIAL MODEL 6"
)

```

Figure 4: Confirmatory factor analysis diagrams. Links between constructs that are of marginal and moderate concern “rho” > .80 ... are shown.

```

loadings_discriminantVal_crossValData <-
  printStdFactorLoadingModels_longASAP(discriminantValModels_crossValData,
                                         factorialModels_crossValData,
                                         longASAP_crossValData_std, 1)

##
##
## FACTORIAL MODEL 1 (CFI = 0.93 )
## C_HLB
## HLB1 0.68
## HLB2 0.62
## HLB3 0.76
## HLB4 0.79
## HLB5 0.77
##
## C_NB
## NB1 0.44
## NB2 0.70
## NB3 0.64
##
## C_AS
## AS1 0.68
## AS2 0.58
## AS3 0.62

```

```

##
##      C_APP
## APP1    0.62
## R_APP2  0.54
## APP3    0.60
##
##      C_SP
## SP1 0.72
## SP2 0.68
## SP3 0.42
##
##      C_AEI
## AEI1    0.72
## AEI2    0.76
## R_AEI3  0.73
## AEI4    0.75
## R_AEI5  0.67
##
##      C_UEP
## UEP1    0.59
## UEP2    0.71
## UEP3    0.52
## UEP4    0.57
##
##      C_UAI
## UAI1    0.53
## UAI2    0.44
## UAI3    0.54
## UAI4    0.63
##
## Covariance Matrix
##      C_HLB C_NB C_AS C_APP C_SP C_AEI C_UEP C_UAI
## C_HLB  1.00 0.90 0.74 0.71 0.69 0.58 0.51 0.54
## C_NB   0.90 1.00 0.85 0.82 0.76 0.75 0.57 0.63
## C_AS   0.74 0.85 1.00 0.78 0.92 0.63 0.66 0.74
## C_APP  0.71 0.82 0.78 1.00 0.83 0.75 0.68 0.63
## C_SP   0.69 0.76 0.92 0.83 1.00 0.66 0.54 0.61
## C_AEI  0.58 0.75 0.63 0.75 0.66 1.00 0.52 0.56
## C_UEP  0.51 0.57 0.66 0.68 0.54 0.52 1.00 0.76
## C_UAI  0.54 0.63 0.74 0.63 0.61 0.56 0.76 1.00
##
##
## FACTORIAL MODEL 2 (CFI = 0.93 )
##      C_UAA
## UAA1    0.79
## UAA2    0.71
## R_UAA3  0.45
##
##      C_AE
## R_AE1 0.59
## AE2   0.60
## AE3   0.78
## R_AE4 0.68

```

```

##
##      C_AT
## AT1  0.83
## AT2  0.75
## R_AT3 0.72
##
##      C_IIS
## IIS1  0.77
## IIS2  0.73
## IIS3  0.46
## IIS4  0.56
##
## Covariance Matrix
##      C_UAA C_AE C_AT C_IIS
## C_UAA  1.00 0.87 0.88 0.88
## C_AE   0.87 1.00 0.92 0.83
## C_AT   0.88 0.92 1.00 0.82
## C_IIS  0.88 0.83 0.82 1.00
##
##
##
## FACTORIAL MODEL 3 (CFI = 0.93 )
##      C_UE
## UE1 0.64
## UE2 0.62
## UE3 0.59
##
##      C_AA
## AA1 0.66
## AA2 0.62
## AA3 0.70
##
##      C_UEP
## UEP1 0.62
## UEP2 0.69
## UEP3 0.53
## UEP4 0.56
##
##      C_UAI
## UAI1 0.52
## UAI2 0.43
## UAI3 0.57
## UAI4 0.62
##
## Covariance Matrix
##      C_UE C_AA C_UEP C_UAI
## C_UE  1.00 0.71 0.55 0.49
## C_AA  0.71 1.00 0.49 0.65
## C_UEP 0.55 0.49 1.00 0.75
## C_UAI 0.49 0.65 0.75 1.00
##
##
##
## FACTORIAL MODEL 4 (CFI = 0.96 )

```

```

##      C_AU
## AU1 0.80
## AU2 0.77
## AU3 0.72
##
##      C_PF
## PF1 0.66
## PF2 0.52
## PF3 0.64
##
##      C_UT
## UT1 0.56
## UT2 0.59
## UT3 0.76
##
##      C_UAL
## UAL1 0.56
## UAL2 0.62
## UAL3 0.51
## UAL4 0.69
## UAL5 0.60
## UAL6 0.67
##
##      C_AC
## R_AC1 0.70
## R_AC2 0.62
## R_AC3 0.62
## R_AC4 0.49
##
##      C_AI
## AI1 0.53
## AI2 0.79
## R_AI3 0.70
## AI4 0.52
##
## Covariance Matrix
##      C_AU C_PF C_UT C_UAL C_AC C_AI
## C_AU 1.00 0.69 0.61 0.57 0.52 0.44
## C_PF 0.69 1.00 0.96 0.89 0.77 0.72
## C_UT 0.61 0.96 1.00 0.89 0.66 0.70
## C_UAL 0.57 0.89 0.89 1.00 0.50 0.71
## C_AC 0.52 0.77 0.66 0.50 1.00 0.54
## C_AI 0.44 0.72 0.70 0.71 0.54 1.00
##
##
## FACTORIAL MODEL 5 (CFI = 0.97 )
##      C_HLA
## HLA1 0.79
## HLA2 0.86
## HLA3 0.76
## HLA4 0.72
##
##      C_NA

```

```

## NA1 0.54
## NA2 0.69
## NA3 0.53
## NA4 0.72
## NA5 0.50
##
##      C_AAS
## AAS1  0.66
## AAS2  0.60
## AAS3  0.78
##
## Covariance Matrix
##      C_HLA C_NA C_AAS
## C_HLA  1.00 0.72 0.39
## C_NA   0.72 1.00 0.61
## C_AAS  0.39 0.61 1.00
##
##
##
## FACTORIAL MODEL 6 (CFI = 0.94 )
##      C_AAS
## AAS1  0.70
## AAS2  0.57
## AAS3  0.77
##
##      C_AL
## AL1   0.59
## AL2   0.86
## R_AL3 0.82
## AL4   0.52
## AL5   0.61
##
##      C_AE
## R_AE1 0.71
## AE2   0.68
## AE3   0.60
## R_AE4 0.70
##
##      C_AC
## R_AC1 0.72
## R_AC2 0.65
## R_AC3 0.57
## R_AC4 0.47
##
## Covariance Matrix
##      C_AAS C_AL C_AE C_AC
## C_AAS  1.00 0.70 0.62 0.54
## C_AL   0.70 1.00 0.93 0.56
## C_AE   0.62 0.93 1.00 0.63
## C_AC   0.54 0.56 0.63 1.00

```

CFI scores of discriminant validation of the long ASAQ

```

cfi_discriminantVal_crossValData <-
  getCfIcoefficient(discriminantValModels_crossValData, factorialModels_crossValData,

```

```
longASAQ_crossValData_std)
```

... The factor loadings of the items ranged from 0.42 to 0.86 ($M = 0.64$, $SD = 0.10$) in the six CFA, while the CFI scores ranged from 0.93 to 0.97 ($M = 0.94$, $SD = 0.02$), see Figure 4. All CFIs are above the .90 threshold considered an indication for a good fit to the data ...

4.5 Predictive Validity of the ASAQ

```
# The constructs'/dimensions' score of each ASA
predictiveValScores_medianBase <-
  getPredictiveValScores_medianBase(predictiveValData_expertA, predictiveValData_expertB,
                                     predictiveValData_expertC,
                                     constructASAQdesc$CABR)

predictiveVal29ASA_medianBase <-
  get29ASAPredictiveValData_medianBase(name29ASAs,
                                       predictiveValScores_medianBase,
                                       constructASAQdesc$CABR)

# Combined dataset of Study Mid 2021 and Study 2022
combinedData <- rbind(longASAQ_constructValData, longASAQ_crossValData)

# Calculate the score of all ASAQ constructs/dimensions of the long version of ASAQ by
# calculating the mean of items in each construct/dimension. The calculation results in
# one score for each construct/dimension for each individual participant
longASAQConstructsScore_combinedData =
  getLongASAQConstructsScore(constructASAQdesc$NUM_ITEMS,
                             combinedData,
                             constructASAQdesc$CABR)

# Calculate the constructs' score of each ASA by calculating the mean of each
# construct/dimension per ASA The calculation results in one score for each
# construct/dimension per ASA
longASAQscores_combinedData <-
  calcASAQscores(name29ASAs, longASAQConstructsScore_combinedData)

# Get the score of all ASAQ constructs/dimensions of the short version of ASAQ.
# The score of each construct/dimension is taken from a representative item of the
# construct/dimension.
# for PF: only for APF (since the representative for the short version is PF1)
# for UAL: only for TUAL (since the representative for the short version is UAL1)
shortASAQConstructsScore_combinedData =
  getShortASAQConstructsScore(constructASAQdesc$SHORT_ITEM_REF,
                              combinedData,
                              constructASAQdesc$CABR)

# Calculate the constructs' score of each ASA by calculating the mean of
# each construct/dimension per ASA. The calculation results in one score for each
# construct/dimension per ASA
shortASAQscores_combinedData <-
  calcASAQscores(name29ASAs, shortASAQConstructsScore_combinedData)
```

```

# The correlation between the scores of ASAQ constructs/dimensions of the ASAQ
# Representative Set 2024 of the experts' predictive analysis (i.e. the median of ASAQ
# constructs/dimensions' scores from 3 experts) and the ASAQ long and short versions
# (i.e., calculated using the mean of items of each ASAQ construct/dimension)
corrPredictiveAndASAQConstructsScores <-
  calcCorrPredictiveAndASAQConstructsScores(predictiveVal29ASA_medianBase,
                                             longASAQscores_combinedData,
                                             shortASAQscores_combinedData,
                                             constructASAQdesc$CABR)

```

... Thus excluding the constructs Performance and User Agent Alliance, the Spearman correlations for the long ASAQ version range from 0.10 to 0.92 with a median of 0.50. According to Hinkle (2003) we can classify this as a moderate correlation ...

Table 5: The correlation (rho) between the median of expert predictions and the mean ASAQ scores of 29 agents.

```

showTable(corrPredictiveAndASAQConstructsScores, constructASAQdesc$CABR,
          constructASAQdesc$CNAME, 1, 0, 0, 0, 0, 0, 0)

```

##	CABR	CNAME	PRED_LONG	PRED_SHORT
## 1	HLA	Human-Like Appearance	0.92	0.91
## 2	HLB	Human-Like Behavior	0.58	0.57
## 3	NA	Natural Appearance	0.74	0.72
## 4	NB	Natural Behavior	0.50	0.48
## 5	AAS	Agent's Appearance Suit.	0.58	0.52
## 6	AU	Agent's Usability	0.35	0.45
## 7	PF	Performance	NA	NA
## 8	AL	Agent's Likeability	0.64	0.66
## 9	AS	Agent's Sociability	0.50	0.54
## 10	APP	Agent's Personality Pr.	0.61	0.38
## 11	UAA	User Acceptance of the A.	0.41	0.32
## 12	AE	Agent's Enjoyability	0.67	0.46
## 13	UE	User's Engagement	0.41	0.43
## 14	UT	User's Trust	0.32	0.34
## 15	UAL	User Agent Alliance	NA	NA
## 16	AA	Agent's Attentiveness	0.39	0.54
## 17	AC	Agent's Coherence	0.10	0.09
## 18	AI	Agent's Intentionality	0.53	0.34
## 19	AT	Attitude	0.45	0.49
## 20	SP	Social Presence	0.71	0.74
## 21	IIS	Interaction Impact on Self.	0.25	0.18
## 22	AEI	Agent's Emotional Int. Pr.	0.41	0.46
## 23	UEP	User's Emotion Presence	0.51	0.46
## 24	UAI	User Agent Interplay	0.47	0.50

4.6 Concurrent Validity of the Long and Short ASAQ Versions

```

# Calculate the score of all ASAQ constructs/dimensions of the long version of ASAQ by
# calculating the mean of items in each construct/dimension. The calculation results in
# one score for each construct/dimension for each individual participant
longASAQConstructsScore_constructValData =
  getLongASAQConstructsScore(constructASAQdesc$NUM_ITEMS,

```

```

                                longASAQ_constructValData,
                                constructASAQdesc$CABR)
longASAQConstructsScore_crossValData =
  getLongASAQConstructsScore(constructASAQdesc$NUM_ITEMS,
                              longASAQ_crossValData,
                              constructASAQdesc$CABR)

# Get the score of all ASAQ constructs/dimensions of the short version of ASAQ.
# The score of each construct/dimension is taken from a representative item of the
# construct/dimension. This results in one score for each construct/dimension for each
# individual participant
shortASAQConstructsScore_constructValData =
  getShortASAQConstructsScore(constructASAQdesc$SHORT_ITEM_REF,
                              longASAQ_constructValData,
                              constructASAQdesc$CABR)
shortASAQConstructsScore_crossValData =
  getShortASAQConstructsScore(constructASAQdesc$SHORT_ITEM_REF,
                              longASAQ_crossValData,
                              constructASAQdesc$CABR)

# Calculate the constructs' score of each ASA by calculating the mean of each
# construct/dimension per ASA. The calculation results in one score for each
# construct/dimension per ASA
longASAQscores_constructValData <-
  calcASAQscores(name29ASAs, longASAQConstructsScore_constructValData)
longASAQscores_crossValData <-
  calcASAQscores(name29ASAs, longASAQConstructsScore_crossValData)

shortASAQscores_constructValData <-
  calcASAQscores(name29ASAs, shortASAQConstructsScore_constructValData)
shortASAQscores_crossValData <-
  calcASAQscores(name29ASAs, shortASAQConstructsScore_crossValData)

# The correlation and SMD between the long (based on the MEAN of the construct/dimension
# scores) and short version of ASAQ
corDiffLongShortASAQ <-
  getCorDiffLongShortASAQ(longASAQConstructsScore_constructValData,
                          shortASAQConstructsScore_constructValData,
                          longASAQConstructsScore_crossValData,
                          shortASAQConstructsScore_crossValData,
                          longASAQConstructsScore_combinedData,
                          shortASAQConstructsScore_combinedData,
                          constructASAQdesc$CABR)

```

... Combining these datasets shows an average correlation of 0.81 over all constructs (range [0.70 .. 0.92]). This can be interpreted as high concurrent validity ... Likewise, the absolute mean differences between the means of the constructs and dimensions of the long version and the representative item of the short version shows an acceptable variation, with a mean of the absolute mean differences of 0.21, (range [0.00 .. 0.58], SD = 0.17).

Table 6: Correlation (rho) and absolute standardised mean difference (Delta M) between the long and short versions of ASAQ.


```
showTable(corDiffLongShortASAQ, constructASAQdesc$CABR, constructASAQdesc$CNAME,
          0, 0, 0, 0, 0, 0, 0)
```

##	CABR	CNAME	COR_StudyM21	DM_StudyM21	COR_Study22
## 1	HLA	Human-Like Appearance	0.92	0.03	0.93
## 2	HLB	Human-Like Behavior	0.87	0.15	0.82
## 3	NA	Natural Appearance	0.83	0.09	0.82
## 4	NB	Natural Behavior	0.85	0.61	0.85
## 5	AAS	Agent's Appearance Suit.	0.85	0.08	0.85
## 6	AU	Agent's Usability	0.83	0.11	0.85
## 7	PF	Performance	0.78	0.13	0.73
## 8	AL	Agent's Likeability	0.87	0.10	0.86
## 9	AS	Agent's Sociability	0.86	0.53	0.84
## 10	APP	Agent's Personality Pr.	0.74	0.39	0.79
## 11	UAA	User Acceptance of the A.	0.80	0.05	0.82
## 12	AE	Agent's Enjoyability	0.80	0.48	0.77
## 13	UE	User's Engagement	0.83	0.07	0.72
## 14	UT	User's Trust	0.82	0.09	0.80
## 15	UAL	User Agent Alliance	0.69	0.43	0.71
## 16	AA	Agent's Attentiveness	0.79	0.19	0.76
## 17	AC	Agent's Coherence	0.73	0.01	0.79
## 18	AI	Agent's Intentionality	0.75	0.25	0.76
## 19	AT	Attitude	0.88	0.00	0.90
## 20	SP	Social Presence	0.85	0.07	0.81
## 21	IIS	Interaction Impact on Self.	0.77	0.24	0.77
## 22	AEI	Agent's Emotional Int. Pr.	0.86	0.38	0.83
## 23	UEP	User's Emotion Presence	0.74	0.07	0.77
## 24	UAI	User Agent Interplay	0.78	0.36	0.73
##	DM_Study22	COR_COMB	DM_COMB		
## 1	0.03	0.92	0.03		
## 2	0.21	0.85	0.18		
## 3	0.02	0.82	0.05		
## 4	0.53	0.85	0.57		
## 5	0.00	0.85	0.04		
## 6	0.20	0.84	0.16		
## 7	0.35	0.75	0.24		
## 8	0.19	0.87	0.15		
## 9	0.64	0.85	0.58		
## 10	0.32	0.77	0.35		
## 11	0.12	0.81	0.09		
## 12	0.10	0.78	0.29		
## 13	0.05	0.77	0.01		
## 14	0.16	0.81	0.13		
## 15	0.56	0.70	0.50		
## 16	0.18	0.78	0.18		
## 17	0.04	0.76	0.03		
## 18	0.35	0.76	0.30		
## 19	0.00	0.89	0.00		
## 20	0.05	0.83	0.06		
## 21	0.26	0.77	0.25		
## 22	0.24	0.85	0.31		
## 23	0.10	0.76	0.08		
## 24	0.37	0.76	0.36		

4.7 Interpretability of the ASAQ

```
# Calculate the mean, standard deviation, and relative frequency of how often the 1066
# (532 + 534) participants use one of the seven points on the answer scale for each
# of the representative items of the long and short versions of the ASAQ
freqLongASAQConstructsScore_combinedData <-
  calcFreqLongASAQConstructsScore(constructASAQdesc$NUM_ITEMS, combinedData, scoreASAQ,
                                   constructASAQdesc$CABR)
freqShortASAQConstructsScore_combinedData <-
  calcFreqShortASAQConstructsScore(shortASAQConstructsScore_combinedData,
                                   scoreASAQ,
                                   constructASAQdesc$CABR)
relFreqLongASAQConstructsScore_combinedData <-
  calcRelFreqLongASAQConstructsScore(constructASAQdesc$NUM_ITEMS, combinedData, scoreASAQ,
                                      constructASAQdesc$CABR)
relFreqShortASAQConstructsScore_combinedData <-
  calcRelFreqShortASAQConstructsScore(shortASAQConstructsScore_combinedData,
                                       scoreASAQ,
                                       constructASAQdesc$CABR)

# Calculate the sum of item scores of long ASAQ constructs/dimensions
sumLongConstructASAQscore_combinedData <-
  calcSumLongConstructASAQscore (constructASAQdesc$NUM_ITEMS,
                                 combinedData,
                                 constructASAQdesc$CABR)

# Calculate relative frequency of extreme scores of
# long ASAQ constructs/dimensions
relFreqExtrmLongASAQConstructScores <-
  calcRelFreqExtrmLongASAQConstructScores(constructASAQdesc$NUM_ITEMS,
                                           sumLongConstructASAQscore_combinedData,
                                           constructASAQdesc$CABR)
```

... The percentage ranges from 2% to 28% with an average of 8.50%, which is lower than 15%, a cut-off point suggested to indicate a floor or ceiling effect ...

... Consequently, this is reflected in an average of (mean of 10.00% (score -3) + 19.21% (score 3) =) 29.21% extreme scores for the short ASAQ. Consequently, the short version is less capable of capturing extreme experiences because of potential floor and ceiling effects.

However, to investigate the impact of potential floor and ceiling effects on the total ASAQ performance, we computed the range of sum of the item scores for both the long and short versions. Where, in theory, the total score of the long version can range from 0 to 0, the scores of our 1066 participants range from -225 to 255 ($M = 50.50$, $SD = 76.76$, ruling out floor or ceiling effects. Similarly, for the short version of the ASAQ, the sum of item scores can range from -72 to 72, while the score of our participants ranges from -67 to 69 ($M = 15.75$, $SD = 22.15$). Furthermore, the bell-shaped curves suggest that there are no holes in the distribution of the sum of item scores for both the long and short versions ...

Table 7: The relative frequency (RF) of the lowest, highest and sum of extreme scores for each ASAQ construct and dimension, based on the long ASAQ version.

```
showTable(relFreqExtrmLongASAQConstructScores, constructASAQdesc$CABR,
          constructASAQdesc$CNAME, 0, 0, 0, 0, 0, 0, 0)
```

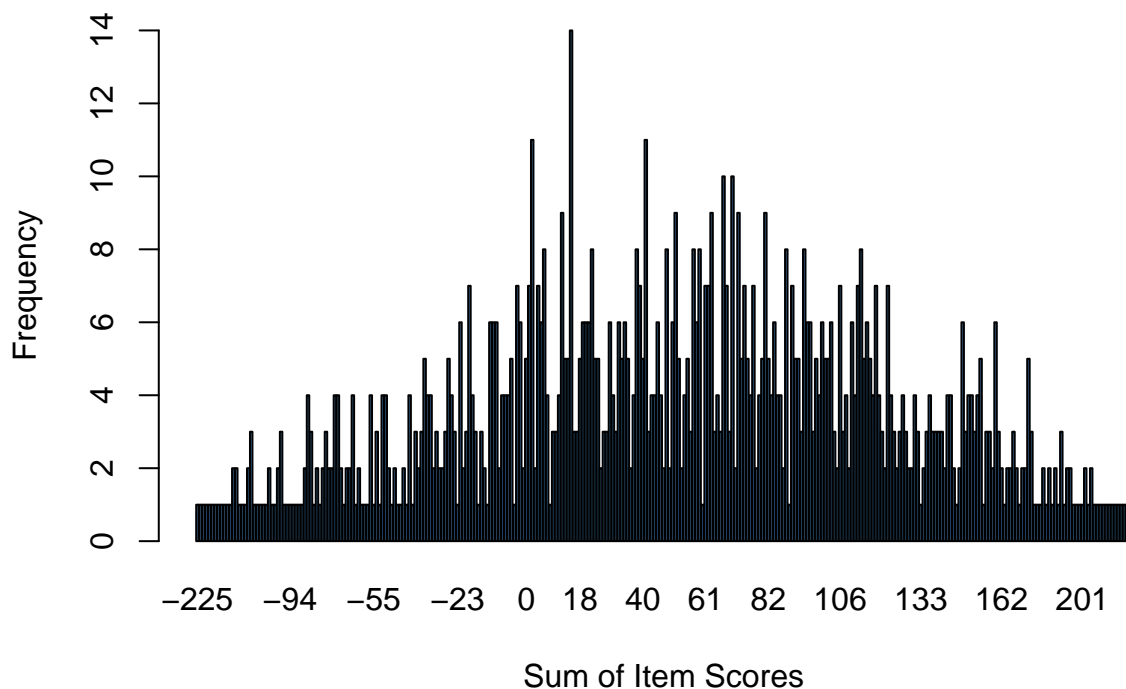
```
##      CABR      CNAME ITEM_NUM fRelLowEXTRM fRelHighEXTRM
```

## 1	HLA	Human-Like Appearance	4.00	0.23	0.05
## 2	HLB	Human-Like Behavior	5.00	0.03	0.02
## 3	NA	Natural Appearance	5.00	0.02	0.01
## 4	NB	Natural Behavior	3.00	0.06	0.02
## 5	AAS	Agent's Appearance Suit.	3.00	0.00	0.12
## 6	AU	Agent's Usability	3.00	0.01	0.11
## 7	PF	Performance	3.00	0.00	0.06
## 8	AL	Agent's Likeability	5.00	0.00	0.05
## 9	AS	Agent's Sociability	3.00	0.02	0.03
## 10	APP	Agent's Personality Pr.	3.00	0.03	0.02
## 11	UAA	User Acceptance of the A.	3.00	0.01	0.10
## 12	AE	Agent's Enjoyability	4.00	0.00	0.09
## 13	UE	User's Engagement	3.00	0.00	0.17
## 14	UT	User's Trust	3.00	0.01	0.02
## 15	UAL	User Agent Alliance	6.00	0.01	0.01
## 16	AA	Agent's Attentiveness	3.00	0.00	0.16
## 17	AC	Agent's Coherence	4.00	0.00	0.11
## 18	AI	Agent's Intentionality	4.00	0.01	0.03
## 19	AT	Attitude	3.00	0.01	0.15
## 20	SP	Social Presence	3.00	0.05	0.01
## 21	IIS	Interaction Impact on Self.	4.00	0.01	0.02
## 22	AEI	Agent's Emotional Int. Pr.	5.00	0.12	0.00
## 23	UEP	User's Emotion Presence	4.00	0.00	0.03
## 24	UAI	User Agent Interplay	4.00	0.00	0.02
##	SumfRelEXTRM				
## 1		0.28			
## 2		0.05			
## 3		0.03			
## 4		0.08			
## 5		0.12			
## 6		0.12			
## 7		0.06			
## 8		0.05			
## 9		0.05			
## 10		0.05			
## 11		0.11			
## 12		0.09			
## 13		0.17			
## 14		0.03			
## 15		0.02			
## 16		0.16			
## 17		0.11			
## 18		0.04			
## 19		0.16			
## 20		0.06			
## 21		0.03			
## 22		0.12			
## 23		0.03			
## 24		0.02			

Figure 5: Range sum of scores of each item of the long and short versions of ASAQ.

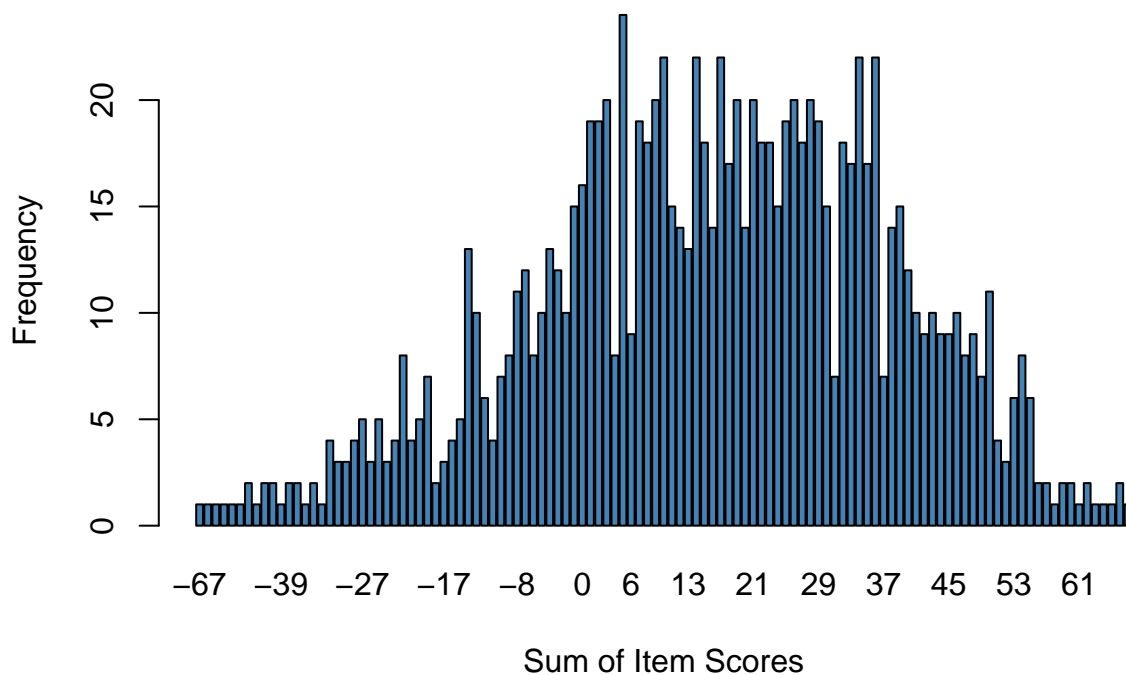
```
# Drawing the histogram of the ASAQ constructs' scores
drawBarPlotFreqTotalASQscore(sumLongConstructASQscore_combinedData,
                              "Long ASAQ (n=1066)", 1)
```

Long ASAQ (n=1066)



```
drawBarPlotFreqTotalASAQscore(shortASAQConstructsScore_combinedData,
                                "Short ASAQ (n=1066)", 2)
```

Short ASAQ (n=1066)



```
# Calculate percentile score of the long and short ASAQ
percentileLongASAQScore_combinedData <-
```

```

    calcPercentileASAQscore(longASAQscores_combinedData, probs,
                           constructASAQdesc$CABR, probTitle)
percentileShortASAQScore_combinedData <-
    calcPercentileASAQscore(shortASAQscores_combinedData, probs,
                           constructASAQdesc$CABR, probTitle)

percentileTotalASAQscore <-
    calcPercentileTotalASAQscore(longASAQscores_combinedData, shortASAQscores_combinedData,
                                probs, probTitle)

```

Table 9: The percentile of ASAQ scores of the ASAQ Representative Set 2024 (n =29).

```

percentileTotalASAQscore

##   ASAQ  q5th q10th q20th q25th q30th q40th q50th q60th q70th q75th q80th q90th
## 1 Long -0.72  2.68  9.36 10.37 12.54 13.58 14.32 16.41 17.82 19.28 20.45 24.91
## 2 Short -0.05  6.21  9.81 11.57 12.79 14.07 14.95 17.88 19.67 20.85 21.81 25.86
##   q95th
## 1 27.69
## 2 30.20

```

```

# Calculate the percentile difference score observed between the mean scores
# of all combinations of the ASAQ Representative Set 2024 for each construct
# and dimension based on the short and long version of the ASAQ
percentileDiffLongTwoASAQscore_combinedData <-
    calcPercentileASAQscore(calcDiffTwoASAQscore(name29ASAs,
                                                  longASAQscores_combinedData,
                                                  constructASAQdesc$CABR),
                           probs, constructASAQdesc$CABR, probTitle
    )
percentileDiffShortTwoASAQscore_combinedData <-
    calcPercentileASAQscore(calcDiffTwoASAQscore(name29ASAs,
                                                  shortASAQscores_combinedData,
                                                  constructASAQdesc$CABR),
                           probs, constructASAQdesc$CABR, probTitle)

```

5 Using the ASAQ

5.1 Choosing the Long or Short Version

5.2 Sample Size

```

# Power analysis:
effectSizeLongASAQConstructScore_combinedData <-
    calcEffectSizeASAQConstructScore(name29ASAs,
                                     longASAQConstructsScore_combinedData,
                                     constructASAQdesc$CABR)
effectSizeShortASAQConstructScore_combinedData <-
    calcEffectSizeASAQConstructScore(name29ASAs,
                                     shortASAQConstructsScore_combinedData,
                                     constructASAQdesc$CABR)

```

```

# Calculate the percentile scores of the effect size
percentileEffectSizeLongASAQConstructScore_combinedData <-
  calcPercentileASAQscore(effectSizeLongASAQConstructScore_combinedData,
    probs, constructASAQdesc$CABR, probTitle)
percentileEffectSizeShortASAQConstructScore_combinedData <-
  calcPercentileASAQscore(effectSizeShortASAQConstructScore_combinedData,
    probs, constructASAQdesc$CABR, probTitle)

# Calculate minimum sample size for comparing two or more agents:
samplePowerAnalysisLongASAQConstructScore_combinedData <-
  calcSamplePowerAnalysisASAQConstructScore(
    calcPercentileASAQscore(effectSizeLongASAQConstructScore_combinedData,
      probsSML, constructASAQdesc$CABR, probTitleSML),
    constructASAQdesc$CABR, probTitleSML)
samplePowerAnalysisShortASAQConstructScore_combinedData <-
  calcSamplePowerAnalysisASAQConstructScore(
    calcPercentileASAQscore(effectSizeShortASAQConstructScore_combinedData,
      probsSML, constructASAQdesc$CABR, probTitleSML),
    constructASAQdesc$CABR, probTitleSML)

# Get max sample size for studies comparing two ASAs based on the power analysis
maxSamplePowerAnalysisASAQConstructScore_combinedData <-
  getMaxSamplePowerAnalysis(samplePowerAnalysisLongASAQConstructScore_combinedData,
    samplePowerAnalysisShortASAQConstructScore_combinedData)

```

Table 10: Most conservative sample size estimations for studies comparing two ASAs with .80 power and .05 alpha level.

```
maxSamplePowerAnalysisASAQConstructScore_combinedData
```

```

##      Version Small_q25th Medium_q50th Large_q75th
## 1 Long ASAQ      485      116      41
## 2 Short ASAQ     614      154      53

# Calculate minimum sample size for evaluating an ASA:
probs5th <- c(.05, .10, .15, .20, .25, .30, .35, .40, .45, .50, .55, .60, .65, .70,
  .75, .80, .85, .90, .95)
sampleConfIntervLongASAQConstructScore_combinedData_p5th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    longASAQConstructsScore_combinedData,
    longASAQscores_combinedData,
    probs5th,
    constructASAQdesc$CABR)
sampleConfIntervShortASAQConstructScore_combinedData_p5th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    shortASAQConstructsScore_combinedData,
    shortASAQscores_combinedData,
    probs5th,
    constructASAQdesc$CABR)

probs10th <- c(.10, .20, .30, .40, .50, .60, .70, .80, .90)
sampleConfIntervLongASAQConstructScore_combinedData_p10th <-
  calcSampleConfIntervASAQConstructScore(

```

```

    name29ASAs,
    longASAQConstructsScore_combinedData,
    longASAQscores_combinedData,
    probs10th,
    constructASAQdesc$CABR)
sampleConfIntervShortASAQConstructScore_combinedData_p10th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    shortASAQConstructsScore_combinedData,
    shortASAQscores_combinedData,
    probs10th,
    constructASAQdesc$CABR)

probs20th <- c(.20, .40, .60, .80)
sampleConfIntervLongASAQConstructScore_combinedData_p20th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    longASAQConstructsScore_combinedData,
    longASAQscores_combinedData,
    probs20th,
    constructASAQdesc$CABR)
sampleConfIntervShortASAQConstructScore_combinedData_p20th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    shortASAQConstructsScore_combinedData,
    shortASAQscores_combinedData,
    probs20th,
    constructASAQdesc$CABR)

probs25th <- c(.25, .50, .75)
sampleConfIntervLongASAQConstructScore_combinedData_p25th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    longASAQConstructsScore_combinedData,
    longASAQscores_combinedData,
    probs25th,
    constructASAQdesc$CABR)
sampleConfIntervShortASAQConstructScore_combinedData_p25th <-
  calcSampleConfIntervASAQConstructScore(
    name29ASAs,
    shortASAQConstructsScore_combinedData,
    shortASAQscores_combinedData,
    probs25th,
    constructASAQdesc$CABR)

# Calculate maximum sample size for evaluating an ASA based on the confident interval
maxSampleSizeConfidenceIntervalASAQConstructScore_combinedData <-
  getMaxSampleSizeConfidenceIntervalASAQConstructScore(
    sampleConfIntervLongASAQConstructScore_combinedData_p5th,
    sampleConfIntervShortASAQConstructScore_combinedData_p5th,
    sampleConfIntervLongASAQConstructScore_combinedData_p10th,
    sampleConfIntervShortASAQConstructScore_combinedData_p10th,
    sampleConfIntervLongASAQConstructScore_combinedData_p20th,

```

```
sampleConfIntervShortASAQConstructScore_combinedData_p20th,
sampleConfIntervLongASAQConstructScore_combinedData_p25th,
sampleConfIntervShortASAQConstructScore_combinedData_p25th)
```

Table 11: Sample size for studies investigating a single ASA (and comparing it with the ASAQ representative set 2024) that are interested in all ASAQ constructs, for the various error margins (percentiles), specific confidence intervals (CI), and for both ASAQ versions. In bold are the suggested sample sizes considered appropriate in most studies. ...

```
maxSampleSizeConfidenceIntervalASAQConstructScore_combinedData
```

##	PERC	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI	SHORT_90CI	SHORT_95CI
## 1	5%	689	979	1280	1691	871	1237
## 2	10%	201	286	374	493	350	497
## 3	20%	70	100	130	172	76	108
## 4	25%	46	66	86	114	68	97
##	SHORT_97.5CI	SHORT_99CI					
## 1	1618	2137					
## 2	650	859					
## 3	141	186					
## 4	126	167					

5.3 Reporting and Visualising Your Results

6 Discussion and Conclusion

```
# registered participants active in the past 90 days (retrieved on 17-07-2024)
activeTotal = 216627
man = 65603 # Man including Trans Male/Trans Man
woman = 94475 # Woman including Trans Female/Trans Woman
nonBinary = 3340 #Non-binary (would like to give more detailed)
notSay = 1138 #Rather not say
genderUnknown = activeTotal - man - woman - nonBinary - notSay
age18to25 = 59734
age26to35 = 79101
age36to45 = 40597
age46to55 = 22222
age56to65 = 10780
ageOlder65 = 4063
ageUnknown = activeTotal-age18to25-age26to35-age36to45-age46to55- age56to65-ageOlder65
```

... Although we refrained from collecting demographic data such as gender and age, the gender distribution of participants registered on this platform was 30.3% (Men), 43.6% (Women), 1.5% (non-Binary), and 24.6% (not reported), and the age distribution was 27.6% (18-25 years), 36.5% (26-35 years), 18.7% (36-45 years), 10.3% (46-55 years), 5.0% (56-65 years), 1.9% (older than 65) and 0.1% (not reported). ...

6.1 Future Work

Acknowledgement

Appendix A The Artificial-Social-Agent Questionnaire

Appendix B The Short Version of The Artificial-Social-Agent Questionnaire

Appendix C Artificial Social Agents used in Studies

Table C.1: The stimuli used in studies: 26 ASAs, a dog, a fish and a zombie.

```
calcTotalLongASAScore(longASAScores_combinedData)
```

##	ASA	ASAScore
## 1	AIBO	19
## 2	ALEXA	10
## 3	ALICE	12
## 4	AMY	9
## 5	CHAPPIE	18
## 6	C3PO	18
## 7	DEEPBLUE	7
## 8	DOG	29
## 9	EFFIE	-1
## 10	THE FISH	32
## 11	FURBY	13
## 12	FURHAT	14
## 13	GEMINOID	18
## 14	HAL 9000	14
## 15	iCAT	-2
## 16	KITT	16
## 17	LOLA	16
## 18	MARCUS	25
## 19	NAO	23
## 20	PARO	19
## 21	POPPIE	14
## 22	ROBOT BOSS	0
## 23	SAM	25
## 24	SARAH	22
## 25	SIM SENSEI	17
## 26	SIRI	13
## 27	THE AMBIENT LIGHT TV	3
## 28	THE NEGOTIATOR	14
## 29	THE ZOMBIE	10

Appendix D Statistical Analysis based on the Combined Studies Mid-2021 and 2022

Table D.1: The relative frequency of how often a score was used (on items, n=1066) based on the long version of the ASAQ.

```
showTable(relFreqLongASAQConstructsScore_combinedData, constructASAQdesc$CABR,
          constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)
```

##	CABR	CNAME	Min3	Min2	Min1	Zero	Plus1	Plus2	Plus3
## 1	HLA	Human-Like Appearance	0.37	0.09	0.08	0.07	0.15	0.12	0.12
## 2	HLB	Human-Like Behavior	0.16	0.11	0.12	0.11	0.24	0.16	0.11
## 3	NA	Natural Appearance	0.22	0.10	0.11	0.13	0.19	0.13	0.11
## 4	NB	Natural Behavior	0.26	0.10	0.10	0.12	0.18	0.12	0.12
## 5	AAS	Agent's Appearance Suit.	0.03	0.03	0.04	0.14	0.23	0.25	0.27
## 6	AU	Agent's Usability	0.03	0.03	0.05	0.16	0.23	0.26	0.24
## 7	PF	Performance	0.02	0.03	0.06	0.19	0.22	0.26	0.22
## 8	AL	Agent's Likeability	0.11	0.06	0.07	0.16	0.17	0.18	0.25
## 9	AS	Agent's Sociability	0.11	0.10	0.11	0.17	0.22	0.17	0.12
## 10	APP	Agent's Personality Pr.	0.15	0.09	0.12	0.15	0.19	0.16	0.13
## 11	UAA	User Acceptance of the A.	0.04	0.04	0.07	0.20	0.18	0.21	0.26
## 12	AE	Agent's Enjoyability	0.05	0.05	0.08	0.13	0.19	0.22	0.29
## 13	UE	User's Engagement	0.02	0.02	0.04	0.08	0.20	0.29	0.36
## 14	UT	User's Trust	0.08	0.06	0.07	0.31	0.19	0.17	0.12
## 15	UAL	User Agent Alliance	0.08	0.07	0.09	0.21	0.25	0.18	0.12
## 16	AA	Agent's Attentiveness	0.03	0.02	0.05	0.08	0.21	0.25	0.36
## 17	AC	Agent's Coherence	0.02	0.03	0.06	0.14	0.16	0.24	0.34
## 18	AI	Agent's Intentionality	0.09	0.06	0.08	0.16	0.21	0.20	0.19
## 19	AT	Attitude	0.04	0.05	0.08	0.15	0.19	0.21	0.28
## 20	SP	Social Presence	0.18	0.12	0.14	0.18	0.19	0.12	0.07
## 21	IIS	Interaction Impact on Self.	0.06	0.06	0.08	0.30	0.20	0.18	0.12
## 22	AEI	Agent's Emotional Int. Pr.	0.31	0.12	0.11	0.14	0.15	0.10	0.07
## 23	UEP	User's Emotion Presence	0.08	0.07	0.08	0.16	0.24	0.20	0.16
## 24	UAI	User Agent Interplay	0.07	0.06	0.08	0.18	0.25	0.20	0.16
##	Description	Min3	Min2	Min1	Zero	Plus1	Plus2	Plus3	
## 1	Mean:	0.11	0.07	0.08	0.16	0.20	0.19	0.19	
## 2	SD:	0.10	0.03	0.03	0.06	0.03	0.05	0.09	
## 3	Median:	0.08	0.06	0.08	0.16	0.20	0.19	0.16	
## 4	Min:	0.02	0.02	0.04	0.07	0.15	0.10	0.07	
## 5	Max:	0.37	0.12	0.14	0.31	0.25	0.29	0.36	

Table D.2: The relative frequency of how often a score was used (n=1066) based on the short version of the ASAQ.

```
showTable(relFreqShortASAQConstructsScore_combinedData, constructASAQdesc$CABR,
          constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)
```

##	CABR	CNAME	Min3	Min2	Min1	Zero	Plus1	Plus2	Plus3
## 1	HLA	Human-Like Appearance	0.39	0.08	0.08	0.06	0.14	0.11	0.13
## 2	HLB	Human-Like Behavior	0.14	0.10	0.12	0.09	0.26	0.18	0.12
## 3	NA	Natural Appearance	0.21	0.12	0.14	0.11	0.18	0.12	0.12
## 4	NB	Natural Behavior	0.15	0.09	0.11	0.10	0.23	0.17	0.15
## 5	AAS	Agent's Appearance Suit.	0.04	0.03	0.04	0.13	0.22	0.25	0.29
## 6	AU	Agent's Usability	0.02	0.03	0.04	0.16	0.21	0.27	0.27
## 7	PF	Performance	0.02	0.02	0.04	0.15	0.20	0.30	0.27

## 8	AL	Agent's Likeability	0.09	0.05	0.06	0.19	0.16	0.19	0.26
## 9	AS	Agent's Sociability	0.17	0.15	0.13	0.20	0.18	0.11	0.07
## 10	APP	Agent's Personality Pr.	0.09	0.06	0.11	0.17	0.25	0.19	0.13
## 11	UAA	User Acceptance of the A.	0.04	0.03	0.04	0.23	0.19	0.21	0.26
## 12	AE	Agent's Enjoyability	0.06	0.07	0.11	0.16	0.17	0.19	0.25
## 13	UE	User's Engagement	0.01	0.02	0.03	0.08	0.21	0.32	0.33
## 14	UT	User's Trust	0.06	0.07	0.10	0.22	0.24	0.19	0.13
## 15	UAL	User Agent Alliance	0.12	0.10	0.09	0.30	0.20	0.12	0.07
## 16	AA	Agent's Attentiveness	0.03	0.03	0.04	0.10	0.26	0.26	0.28
## 17	AC	Agent's Coherence	0.02	0.02	0.06	0.14	0.16	0.26	0.34
## 18	AI	Agent's Intentionality	0.07	0.05	0.08	0.15	0.16	0.22	0.27
## 19	AT	Attitude	0.03	0.05	0.08	0.14	0.23	0.20	0.27
## 20	SP	Social Presence	0.17	0.10	0.13	0.21	0.19	0.13	0.07
## 21	IIS	Interaction Impact on Self.	0.05	0.05	0.06	0.25	0.24	0.20	0.15
## 22	AEI	Agent's Emotional Int. Pr.	0.25	0.12	0.12	0.13	0.16	0.12	0.10
## 23	UEP	User's Emotion Presence	0.08	0.07	0.08	0.15	0.23	0.20	0.18
## 24	UAI	User Agent Interplay	0.09	0.08	0.09	0.23	0.23	0.17	0.10
##	Description	Min3	Min2	Min1	Zero	Plus1	Plus2	Plus3	
## 1	Mean:	0.10	0.07	0.08	0.16	0.20	0.20	0.19	
## 2	SD:	0.09	0.04	0.03	0.06	0.04	0.06	0.09	
## 3	Median:	0.08	0.06	0.08	0.15	0.21	0.19	0.16	
## 4	Min:	0.01	0.02	0.03	0.06	0.14	0.11	0.07	
## 5	Max:	0.39	0.15	0.14	0.30	0.26	0.32	0.34	

Table D.3 - Table D.6: The ASAQ scores of the ASAQ Representative Set 2024 (n=29) based on the long version of ASAQ (n=1066).

```
showTableASA(longASAQscores_combinedData,
              calcASAQscores_StdDev(name29ASAs, longASAQConstructsScore_combinedData),
              calcTotalLongASAQscore(longASAQscores_combinedData),
              getSampleASA(name29ASAs, combinedData),
              constructASAQdesc$CABR, constructASAQdesc$CNAME)
```

##	CABR	CNAME	AIBO_M	AIBO_SD	ALEXA_M	ALEXA_SD	ALICE_M
## 1	HLA	Human-Like Appearance	-2.54	0.78	-2.47	0.90	0.96
## 2	HLB	Human-Like Behavior	-1.59	1.11	-0.34	1.35	0.21
## 3	NA	Natural Appearance	0.04	1.47	-1.30	1.02	0.11
## 4	NB	Natural Behavior	-0.30	1.30	-1.06	1.23	-1.09
## 5	AAS	Agent's Appearance Suit.	2.06	0.81	1.13	1.36	1.44
## 6	AU	Agent's Usability	1.78	0.77	2.02	0.97	1.62
## 7	PF	Performance	1.56	0.95	1.78	0.77	1.05
## 8	AL	Agent's Likeability	1.48	1.23	0.77	1.19	0.77
## 9	AS	Agent's Sociability	0.93	1.36	-0.03	1.24	0.01
## 10	APP	Agent's Personality Pr.	0.32	1.14	-0.53	1.40	-0.64
## 11	UAA	User Acceptance of the A.	1.82	0.98	1.94	1.00	0.98
## 12	AE	Agent's Enjoyability	1.96	0.97	1.30	1.13	0.95
## 13	UE	User's Engagement	2.19	0.97	1.54	1.06	1.93
## 14	UT	User's Trust	-0.09	1.12	1.01	1.12	0.73
## 15	UAL	User Agent Alliance	0.36	1.17	0.60	0.99	0.48
## 16	AA	Agent's Attentiveness	1.85	0.96	1.80	0.98	2.13
## 17	AC	Agent's Coherence	1.76	0.89	1.89	0.89	1.65
## 18	AI	Agent's Intentionality	0.13	1.42	0.61	1.06	0.44
## 19	AT	Attitude	2.37	0.67	1.71	0.86	1.08
## 20	SP	Social Presence	0.03	1.45	-0.77	1.41	-0.98
## 21	IIS	Interaction Impact on Self.	1.06	1.16	0.74	0.92	0.16

## 22	AEI	Agent's Emotional Int. Pr.	-0.69	1.58	-1.71	1.45	-1.46	
## 23	UEP	User's Emotion Presence	1.51	0.90	-0.38	1.15	-0.33	
## 24	UAI	User Agent Interplay	1.30	1.22	0.13	1.09	0.25	
##	ALICE_SD	AMY_M	AMY_SD	CHAPPIE_M	CHAPPIE_SD	C3PO_M	C3PO_SD	DEEPBLUE_M
## 1	1.58	0.88	1.38	-1.09	1.54	-0.31	1.72	-1.96
## 2	1.34	0.09	1.51	0.31	1.09	0.42	1.29	-0.55
## 3	1.18	-0.25	1.18	-0.92	1.16	-0.64	1.18	-1.10
## 4	1.14	-0.85	1.40	0.16	1.36	0.04	1.28	-1.14
## 5	1.08	1.21	1.27	0.95	1.28	1.61	0.84	1.09
## 6	1.25	1.28	0.83	0.75	1.09	1.07	1.31	0.80
## 7	1.38	0.87	1.08	1.13	0.85	1.15	1.28	1.92
## 8	1.06	0.08	1.28	1.27	0.98	1.44	1.26	0.33
## 9	1.42	-0.28	1.25	0.63	1.07	0.42	1.20	-1.23
## 10	1.57	-0.44	1.22	0.77	1.42	1.31	1.38	-1.31
## 11	1.09	0.63	0.98	1.39	1.13	0.48	1.51	1.34
## 12	1.12	0.31	1.09	1.82	0.88	1.10	1.38	1.18
## 13	0.92	1.31	0.93	2.04	0.77	1.43	0.94	1.80
## 14	1.20	0.61	0.93	0.07	1.14	1.20	1.12	1.18
## 15	1.06	0.10	1.04	0.65	0.89	0.70	1.06	0.63
## 16	1.04	1.66	1.11	1.15	1.19	1.73	0.91	1.32
## 17	0.94	1.50	1.01	1.05	0.95	1.33	1.07	2.11
## 18	1.35	0.33	1.44	0.18	1.39	1.41	0.97	1.54
## 19	1.33	0.36	1.05	1.86	1.04	0.03	1.72	1.47
## 20	1.32	-0.46	1.41	0.25	1.21	-0.27	1.21	-1.21
## 21	0.97	-0.10	0.81	0.88	0.98	0.48	1.15	0.83
## 22	1.36	-0.92	1.40	0.26	1.37	0.86	1.20	-2.17
## 23	1.27	0.30	1.03	1.18	1.20	0.13	0.96	-0.40
## 24	0.91	0.69	0.95	1.02	1.12	0.61	0.89	0.17
##	DEEPBLUE_SD	DOG_M	DOG_SD	EFFIE_M	EFFIE_SD	THE FISH_M	THE FISH_SD	FURBY_M
## 1	1.56	-1.91	1.28	0.62	1.68	-1.87	1.24	-2.14
## 2	1.61	-0.62	1.47	-0.19	1.65	-0.16	1.61	-0.14
## 3	1.20	1.70	0.82	-0.35	1.51	1.85	1.06	-0.85
## 4	1.55	1.89	1.00	-1.41	1.37	2.21	0.84	-0.61
## 5	1.27	1.85	0.96	0.67	1.49	2.14	0.86	1.80
## 6	1.25	0.42	1.07	0.85	1.44	0.81	1.26	1.92
## 7	0.97	1.28	1.50	0.58	1.33	1.69	0.76	1.16
## 8	1.07	2.31	0.79	-0.32	1.51	2.10	0.74	0.91
## 9	1.62	1.48	1.04	-0.53	1.52	1.22	0.82	0.73
## 10	1.43	1.47	1.04	-0.60	1.42	1.39	0.86	0.74
## 11	1.06	1.99	0.89	0.03	1.35	1.58	0.93	1.02
## 12	0.98	2.29	0.76	0.06	1.46	2.32	0.75	1.44
## 13	1.01	2.33	0.60	0.81	1.49	1.78	0.93	1.50
## 14	1.23	0.21	1.12	0.11	1.12	0.05	1.10	-0.46
## 15	1.25	1.03	1.17	-0.17	1.33	1.14	0.83	-0.06
## 16	1.49	1.91	1.18	0.53	1.48	1.98	0.94	1.14
## 17	0.88	1.07	1.10	0.68	1.43	1.74	1.16	1.41
## 18	1.31	0.21	1.05	-0.11	1.38	1.42	0.93	-0.49
## 19	1.06	2.50	0.82	0.31	1.26	2.32	0.82	1.56
## 20	1.58	0.48	1.05	-0.97	1.48	0.54	1.22	-0.12
## 21	1.07	1.41	1.04	-0.15	1.26	1.22	1.05	0.63
## 22	1.27	1.10	1.27	-1.21	1.69	1.15	0.95	-0.15
## 23	1.35	1.51	0.91	-0.19	1.43	1.55	1.02	1.03
## 24	1.07	1.58	1.08	-0.18	1.41	1.68	0.71	1.03
##	FURBY_SD	FURHAT_M	FURHAT_SD	GEMINOID_M	GEMINOID_SD	HAL 9000_M	HAL 9000_SD	

## 1	1.22	-0.76	1.77	2.09	1.21	-2.08	1.15			
## 2	1.71	-0.29	1.53	0.14	1.26	0.23	1.49			
## 3	1.28	-0.82	1.40	0.75	1.07	-1.20	1.16			
## 4	1.41	-1.09	1.19	-0.74	1.37	-0.36	1.35			
## 5	1.14	0.79	1.68	1.90	0.87	1.39	0.99			
## 6	0.98	1.71	1.00	1.27	1.13	1.39	1.00			
## 7	1.07	1.50	0.79	1.37	0.98	1.18	1.29			
## 8	1.59	0.69	1.26	0.87	1.40	0.36	1.45			
## 9	1.43	0.15	1.21	0.59	1.20	0.05	1.39			
## 10	1.43	-0.54	1.04	-0.13	1.54	0.29	1.34			
## 11	1.34	1.05	1.48	1.10	1.13	1.36	1.11			
## 12	1.37	1.23	1.04	1.29	1.31	0.86	1.26			
## 13	1.11	2.01	0.81	1.72	0.94	2.12	1.02			
## 14	1.37	1.04	1.00	0.30	1.37	0.41	1.40			
## 15	1.32	0.83	0.95	0.59	0.92	0.84	1.08			
## 16	1.30	2.13	0.86	1.75	0.90	2.10	0.79			
## 17	1.27	2.04	0.82	1.25	0.98	1.61	1.21			
## 18	1.32	0.76	1.14	0.52	1.53	1.80	0.98			
## 19	1.28	1.46	1.40	1.28	1.04	0.41	1.71			
## 20	1.75	-0.27	1.38	-0.02	1.20	-0.26	1.17			
## 21	1.33	0.29	0.95	0.54	1.02	0.75	1.15			
## 22	1.76	-1.48	1.15	-1.18	1.39	-1.06	1.33			
## 23	1.24	0.81	1.24	0.06	1.22	0.80	1.21			
## 24	1.10	0.55	1.07	0.55	1.07	0.68	1.00			
##	iCAT_M	iCAT_SD	KITT_M	KITT_SD	LOLA_M	LOLA_SD	MARCUS_M	MARCUS_SD	NAO_M	NAO_SD
## 1	-2.21	1.23	-2.01	1.31	1.14	1.55	1.69	1.13	-0.93	1.55
## 2	-1.28	1.31	0.52	1.38	0.58	1.52	1.70	0.97	0.74	1.25
## 3	-1.50	1.07	-0.94	1.31	0.62	1.21	0.67	1.21	-0.57	1.33
## 4	-1.88	0.97	-0.22	1.37	-0.57	1.25	1.21	1.13	-0.36	1.51
## 5	0.49	1.49	1.31	0.98	1.45	1.24	1.41	1.02	1.74	0.94
## 6	1.19	1.25	1.60	1.18	1.75	1.18	0.65	1.25	1.00	1.15
## 7	0.94	1.04	1.11	1.17	1.22	1.10	1.09	0.87	1.31	1.01
## 8	-0.57	1.28	1.21	1.22	0.59	1.55	0.41	1.09	1.74	1.33
## 9	-0.35	1.56	0.44	1.42	0.45	1.32	0.75	1.07	0.97	1.13
## 10	-0.52	1.45	0.89	1.39	-0.20	1.40	1.35	1.10	1.06	1.26
## 11	0.66	1.27	0.92	1.44	0.46	1.25	0.76	1.09	1.57	1.23
## 12	0.47	1.18	1.34	0.99	0.64	1.47	0.71	1.19	1.91	1.06
## 13	1.67	1.02	1.44	1.08	1.28	1.12	2.08	0.89	2.11	0.73
## 14	-0.03	1.01	0.99	1.36	1.08	1.06	0.31	0.97	0.51	1.17
## 15	-0.26	1.08	0.98	1.04	0.68	1.11	1.07	0.84	0.94	1.14
## 16	1.38	1.23	1.72	1.08	2.22	0.82	1.96	0.94	1.59	1.16
## 17	1.68	0.90	1.76	1.06	1.95	0.90	1.04	1.08	1.61	0.86
## 18	0.07	1.22	1.50	0.98	0.68	1.40	1.63	0.96	0.82	1.05
## 19	1.10	1.43	0.21	1.48	0.33	1.12	-0.01	1.24	1.90	0.99
## 20	-1.12	1.35	-0.41	1.37	0.05	1.19	0.75	1.50	0.40	1.46
## 21	-0.08	1.03	0.80	1.27	-0.06	1.22	0.24	0.84	1.12	1.34
## 22	-1.84	1.27	-1.07	1.67	-1.42	1.32	0.78	1.36	0.07	1.80
## 23	0.26	1.22	0.72	1.12	0.33	1.14	1.24	0.93	0.99	1.05
## 24	0.06	1.24	1.06	1.06	0.54	0.97	1.40	0.88	1.14	1.27
##	PARO_M	PARO_SD	POPPIE_M	POPPIE_SD	ROBOT	BOSS_M	ROBOT	BOSS_SD	SAM_M	SAM_SD
## 1	-2.41	1.02	0.64	1.42		-1.18		1.58	-1.31	1.69
## 2	-1.09	1.25	0.51	1.50		-0.06		1.42	1.58	1.18
## 3	0.38	1.35	-0.31	1.35		-1.24		1.12	-0.41	1.43
## 4	-0.48	1.16	-0.38	1.37		-0.68		1.46	0.42	1.22

## 5	1.99	0.89	0.70	1.21	0.24	1.66	0.35	1.32
## 6	1.77	0.93	0.84	1.53	1.25	1.43	1.99	0.92
## 7	1.62	1.00	0.94	1.19	0.41	1.28	1.54	0.88
## 8	1.57	1.22	0.22	1.24	-0.71	1.45	1.38	1.20
## 9	1.25	1.00	0.75	1.24	-0.25	1.59	1.51	1.20
## 10	0.04	1.29	-0.04	1.22	0.05	1.38	0.41	1.27
## 11	2.05	0.95	1.35	1.23	0.01	1.15	1.44	1.22
## 12	2.03	0.93	1.03	1.26	-0.29	1.48	1.86	0.94
## 13	1.84	0.93	1.99	1.04	0.67	1.68	2.25	0.65
## 14	-0.04	0.96	0.32	1.15	0.28	1.17	0.79	1.02
## 15	0.56	0.99	0.22	1.13	0.15	1.37	0.77	1.01
## 16	1.40	1.11	1.50	1.24	0.84	1.24	2.16	0.90
## 17	1.46	1.02	1.22	1.34	0.85	1.00	1.87	0.98
## 18	-0.66	1.14	0.44	1.24	0.51	1.40	1.13	1.20
## 19	2.41	0.93	1.68	1.36	-0.63	1.52	1.41	1.35
## 20	0.23	0.91	0.04	1.45	-0.59	1.37	0.10	1.21
## 21	1.45	1.09	0.21	1.18	-0.68	1.47	0.90	1.20
## 22	-0.87	1.40	-0.73	1.60	-0.26	1.54	-0.06	1.60
## 23	1.88	0.92	0.49	1.19	0.49	1.44	1.72	0.70
## 24	0.91	1.00	0.68	1.34	0.68	1.14	1.22	1.23
##	SARAH_M	SARAH_SD	SIM	SENSEI_M	SIM	SENSEI_SD	SIRI_M	SIRI_SD
## 1	1.66	1.14		1.28		1.42	-1.76	1.23
## 2	0.71	1.21		0.92		1.40	-0.28	1.42
## 3	0.87	1.21		0.61		1.29	-0.70	1.07
## 4	-0.32	1.35		-0.36		1.10	-0.81	1.08
## 5	1.74	0.92		1.55		1.29	0.80	1.23
## 6	1.71	1.10		1.25		1.01	2.21	0.73
## 7	1.72	0.88		1.04		1.21	2.07	0.90
## 8	0.77	1.18		0.61		1.27	0.77	1.08
## 9	0.04	1.37		0.13		1.43	-0.14	1.59
## 10	-0.17	1.34		0.01		1.17	-0.64	1.32
## 11	1.51	0.96		0.74		1.20	2.13	0.78
## 12	1.11	1.08		0.86		1.06	1.53	1.04
## 13	1.67	0.70		1.04		1.35	1.56	0.99
## 14	1.30	0.86		0.48		1.02	1.17	1.04
## 15	0.79	0.82		0.32		1.00	0.57	1.18
## 16	2.04	0.91		1.76		1.14	1.79	0.95
## 17	1.92	0.84		1.58		1.00	2.08	0.80
## 18	1.36	0.89		0.85		1.15	0.79	1.05
## 19	1.73	0.85		0.92		1.07	2.03	0.80
## 20	0.06	1.32		-0.06		1.50	-0.99	1.40
## 21	0.78	0.85		0.36		1.03	0.93	1.03
## 22	-1.36	1.47		-0.65		1.24	-1.91	1.11
## 23	0.15	1.03		0.36		1.01	-0.62	1.34
## 24	0.40	1.08		0.95		1.15	0.05	1.05
##	THE AMBIENT	LIGHT	TV_M	THE AMBIENT	LIGHT	TV_SD	THE NEGOTIATOR_M	
## 1			-2.19			1.07		0.99
## 2			-1.37			1.17		0.65
## 3			-0.84			1.18		0.39
## 4			-0.96			1.39		-0.35
## 5			1.81			1.05		1.53
## 6			1.29			1.05		0.89
## 7			0.95			1.17		0.53
## 8			1.06			1.17		0.04

## 9	-0.95	1.14	-0.24									
## 10	-0.80	1.44	0.08									
## 11	1.33	1.26	0.52									
## 12	1.63	1.16	0.63									
## 13	1.56	1.32	2.25									
## 14	-0.30	1.38	0.40									
## 15	-0.20	1.46	0.35									
## 16	0.25	1.60	1.40									
## 17	1.56	1.12	1.48									
## 18	-0.06	1.51	1.49									
## 19	1.67	1.25	0.11									
## 20	-1.43	1.25	-0.32									
## 21	0.93	1.25	0.09									
## 22	-1.50	1.41	-1.11									
## 23	0.44	1.42	0.98									
## 24	-0.51	1.55	0.80									
##	THE NEGOTIATOR_SD	THE ZOMBIE_M	THE ZOMBIE_SD									
## 1	1.35	1.79	1.25									
## 2	1.26	0.52	1.12									
## 3	1.22	0.41	1.11									
## 4	1.40	0.75	1.23									
## 5	0.92	1.01	1.27									
## 6	1.31	-0.62	1.21									
## 7	1.11	0.18	0.92									
## 8	1.19	-0.60	1.37									
## 9	1.35	-0.64	1.31									
## 10	1.23	1.04	1.18									
## 11	1.06	-0.26	1.50									
## 12	1.31	-0.35	1.17									
## 13	0.75	2.18	0.86									
## 14	1.13	-0.41	1.18									
## 15	1.05	-0.14	1.22									
## 16	1.22	1.87	0.91									
## 17	1.22	0.20	1.04									
## 18	0.99	1.16	1.29									
## 19	1.35	-0.54	1.52									
## 20	1.25	-0.18	1.20									
## 21	0.88	-0.68	1.34									
## 22	1.33	0.35	1.41									
## 23	1.24	1.76	0.90									
## 24	0.79	0.86	1.00									
##	Description	AIBO	ALEXA	ALICE	AMY	CHAPPIE	C3PO	DEEPBLUE	DOG	EFFIE	THE FISH	
## 1	Sample Size	39	37	35	39	38	36	39	39	34	37	
## 2	ASAQ score	19	10	12	9	18	18	7	29	-1	32	
##	FURBY	FURHAT	GEMINOID	HAL	9000	iCAT	KITT	LOLA	MARCUS	NAO	PARO	POPPIE
## 1	39	34	35	37	36	36	37	36	36	37	38	
## 2	13	14	18	14	-2	16	16	25	23	19	14	
##	ROBOT	BOSS	SAM	SARAH	SIM	SENSEI	SIRI	THE AMBIENT	LIGHT	TV	THE NEGOTIATOR	
## 1	34	36	39	38	39	35	36					
## 2	0	25	22	17	13	3	14					
##	THE ZOMBIE											
## 1	35											
## 2	10											

Table D.7 - Table D.10: The ASAQ scores of the ASAQ Representative Set 2024 (n=29) based on the short version of ASAQ (n=1066).

```
showTableASA(shortASAScores_combinedData,
              calcASAScores_StdDev(name29ASAs, shortASAQConstructsScore_combinedData),
              calcTotalLongASAScore(shortASAScores_combinedData),
              getSampleASA(name29ASAs, combinedData),
              constructASAQdesc$CABR, constructASAQdesc$CNAME)
```

##	CABR	CNAME	AIBO_M	AIBO_SD	ALEXA_M	ALEXA_SD	ALICE_M	
## 1	HLA	Human-Like Appearance	-2.59	0.88	-2.70	0.81	1.34	
## 2	HLB	Human-Like Behavior	-1.77	1.42	0.03	1.80	0.77	
## 3	NA	Natural Appearance	-0.56	2.21	-1.03	1.98	-0.20	
## 4	NB	Natural Behavior	0.69	1.88	-0.89	1.87	-0.40	
## 5	AAS	Agent's Appearance Suit.	2.00	1.19	1.49	1.66	1.51	
## 6	AU	Agent's Usability	1.95	0.86	2.22	0.98	1.60	
## 7	PF	Performance	1.90	1.07	2.30	0.88	0.66	
## 8	AL	Agent's Likeability	1.44	1.65	1.30	1.63	0.71	
## 9	AS	Agent's Sociability	0.54	1.82	-1.00	1.67	-0.71	
## 10	APP	Agent's Personality Pr.	0.82	1.65	0.03	1.88	-0.37	
## 11	UAA	User Acceptance of the A.	1.74	1.09	2.05	1.25	0.77	
## 12	AE	Agent's Enjoyability	1.41	1.45	0.54	1.76	0.03	
## 13	UE	User's Engagement	2.31	1.26	1.54	1.24	1.51	
## 14	UT	User's Trust	0.21	1.59	1.30	1.24	0.60	
## 15	UAL	User Agent Alliance	-0.13	1.59	-0.08	1.64	-0.37	
## 16	AA	Agent's Attentiveness	1.69	1.08	1.62	1.26	1.97	
## 17	AC	Agent's Coherence	1.92	1.18	2.22	1.00	1.54	
## 18	AI	Agent's Intentionality	0.59	2.09	1.14	1.90	1.11	
## 19	AT	Attitude	2.33	0.77	1.51	1.35	0.97	
## 20	SP	Social Presence	-0.13	2.00	-0.81	1.85	-1.14	
## 21	IIS	Interaction Impact on Self.	1.28	1.45	1.22	1.40	0.57	
## 22	AEI	Agent's Emotional Int. Pr.	0.05	2.26	-1.49	1.85	-0.86	
## 23	UEP	User's Emotion Presence	1.82	1.10	-0.27	1.64	-0.34	
## 24	UAI	User Agent Interplay	1.33	1.56	-0.65	1.62	-0.86	
##	ALICE_SD	AMY_M	AMY_SD	CHAPPIE_M	CHAPPIE_SD	C3PO_M	C3PO_SD	DEEPBLUE_M
## 1	1.68	1.13	1.67	-1.21	1.76	-0.19	2.10	-2.10
## 2	1.42	0.44	1.82	0.42	1.69	0.64	1.84	-0.90
## 3	1.62	-0.46	1.82	-1.13	1.60	-0.97	1.78	-1.15
## 4	2.05	-0.18	1.92	0.89	1.59	0.33	1.62	-0.87
## 5	1.67	1.13	1.59	0.87	1.70	1.19	1.60	1.18
## 6	1.26	1.51	1.10	0.71	1.37	1.28	1.56	1.28
## 7	1.85	0.90	1.48	1.29	1.04	1.53	1.34	2.51
## 8	1.62	-0.05	1.97	1.47	1.50	1.92	1.42	0.85
## 9	1.93	-1.05	1.54	0.13	1.47	-0.31	1.69	-1.54
## 10	1.90	0.03	1.56	0.68	1.54	1.39	1.71	-0.56
## 11	1.55	0.36	1.20	1.71	1.27	1.19	1.56	1.59
## 12	1.74	-0.26	1.79	1.47	1.47	1.42	1.70	0.33
## 13	1.22	1.10	1.23	2.13	0.91	0.78	1.64	1.69
## 14	1.82	0.33	1.36	-0.08	1.51	1.17	1.83	1.56
## 15	1.61	-0.23	1.51	-0.03	1.53	1.00	1.53	0.46
## 16	1.29	1.23	1.55	1.11	1.43	2.22	0.93	1.49
## 17	1.44	1.31	1.56	1.45	1.33	1.56	1.42	2.00
## 18	1.79	0.82	1.68	0.53	1.93	1.78	1.53	1.72
## 19	1.32	0.44	1.25	2.00	1.29	0.00	1.93	1.38
## 20	1.48	-0.18	1.86	0.42	1.59	-0.11	1.75	-1.38

## 21	1.31	0.51	1.39	1.21	1.34	0.94	1.49	1.10
## 22	1.96	-0.36	2.05	0.53	1.77	1.14	1.59	-2.41
## 23	1.53	0.03	1.46	1.55	1.52	-0.53	1.83	-0.03
## 24	1.80	0.38	1.33	0.79	1.32	0.31	1.58	-1.00
##	DEEPBLUE_SD	DOG_M	DOG_SD	EFFIE_M	EFFIE_SD	THE FISH_M	THE FISH_SD	FURBY_M
## 1	1.60	-2.13	1.36	0.85	2.03	-2.19	1.56	-2.69
## 2	2.05	-0.51	2.00	0.18	1.82	-0.41	2.03	0.00
## 3	1.83	2.10	1.39	-0.56	2.05	2.16	1.01	-1.10
## 4	2.30	2.28	0.92	-1.00	1.71	2.57	0.87	0.21
## 5	1.54	2.28	1.23	0.44	1.91	2.14	1.23	1.87
## 6	1.41	0.74	1.53	1.03	1.45	1.24	1.40	2.18
## 7	0.79	1.23	1.80	0.71	1.51	1.62	1.11	1.51
## 8	1.37	2.62	0.75	-0.38	2.15	2.51	0.84	0.85
## 9	1.73	1.18	1.30	-1.09	1.75	0.43	1.32	0.00
## 10	1.96	1.21	1.44	-0.38	1.84	1.43	1.07	1.33
## 11	1.09	2.08	1.06	-0.15	1.64	1.62	1.42	1.00
## 12	1.66	2.03	1.39	-0.15	2.16	2.27	1.12	0.82
## 13	1.28	2.36	0.87	0.79	1.74	2.08	1.04	1.74
## 14	1.65	1.05	1.49	0.24	1.52	0.57	1.64	-0.59
## 15	1.79	0.82	1.70	-0.56	1.46	0.41	1.54	-0.54
## 16	1.67	1.87	1.28	0.29	1.92	1.68	1.25	0.90
## 17	1.19	1.33	1.81	0.62	1.97	1.65	1.44	1.13
## 18	1.90	-0.54	1.82	0.09	2.09	1.57	1.26	-0.62
## 19	1.46	2.49	0.97	0.47	1.50	2.19	1.31	1.54
## 20	1.90	1.03	1.61	-1.09	1.86	0.81	1.41	0.00
## 21	1.59	1.31	1.42	-0.03	1.82	0.76	1.42	0.90
## 22	1.14	1.51	1.54	-0.94	1.98	1.43	1.63	0.79
## 23	2.08	2.26	0.97	-0.15	1.88	1.70	1.08	1.10
## 24	1.86	1.26	1.58	-0.79	1.57	1.38	1.01	1.05
##	FURBY_SD	FURHAT_M	FURHAT_SD	GEMINOID_M	GEMINOID_SD	HAL 9000_M	HAL 9000_SD	
## 1	0.80	-0.65	2.01	2.20	1.39	-2.14	1.58	
## 2	2.20	-0.35	1.67	0.77	1.52	0.54	1.97	
## 3	1.92	-0.97	1.88	0.71	1.89	-1.62	1.69	
## 4	2.00	-0.82	1.80	-0.09	1.69	0.00	2.00	
## 5	1.26	0.74	1.91	1.94	0.97	1.70	1.10	
## 6	1.21	1.74	1.05	1.43	1.29	1.65	1.16	
## 7	1.23	1.97	0.94	1.54	1.17	1.43	1.39	
## 8	2.06	0.97	1.51	0.43	2.08	0.41	1.79	
## 9	2.09	-0.47	1.46	-0.40	1.75	-0.38	1.80	
## 10	1.74	-0.15	1.64	0.14	1.78	0.73	1.74	
## 11	1.69	0.74	1.78	1.03	1.27	1.92	1.23	
## 12	1.99	1.06	1.43	0.89	1.76	0.54	1.92	
## 13	1.31	1.97	1.36	1.57	1.12	1.97	1.28	
## 14	1.85	0.82	1.38	0.40	1.79	0.76	1.80	
## 15	1.85	0.35	1.70	0.00	1.24	0.51	1.98	
## 16	1.73	1.82	1.03	1.66	1.14	2.24	1.06	
## 17	1.73	2.09	1.06	1.57	1.31	1.27	1.50	
## 18	2.21	1.35	1.76	0.97	1.71	2.30	1.08	
## 19	1.50	1.41	1.60	1.26	1.24	0.49	1.77	
## 20	2.27	-0.26	1.54	0.20	1.45	-0.14	1.51	
## 21	1.65	0.56	1.56	0.69	1.23	0.95	1.51	
## 22	2.09	-1.26	1.69	-1.29	1.62	-0.92	1.79	
## 23	1.73	1.00	1.81	0.23	1.65	0.65	1.64	
## 24	1.82	-0.03	1.77	0.17	1.38	0.14	1.46	

##	iCAT_M	iCAT_SD	KITT_M	KITT_SD	LOLA_M	LOLA_SD	MARCUS_M	MARCUS_SD	NAO_M	NAO_SD
## 1	-2.19	1.51	-2.28	1.58	1.11	1.76	1.67	1.45	-1.17	1.75
## 2	-1.22	1.81	1.06	1.51	0.65	1.64	1.97	1.13	0.97	1.59
## 3	-1.81	1.62	-0.92	2.13	0.49	1.84	0.61	2.11	-0.89	1.89
## 4	-1.64	1.73	0.53	1.86	0.14	1.93	1.67	1.24	0.50	2.05
## 5	0.64	1.66	1.28	1.50	1.62	1.40	0.97	1.52	2.00	1.15
## 6	1.25	1.52	1.83	1.30	1.92	1.14	0.25	1.68	1.06	1.15
## 7	0.75	1.42	1.64	1.29	1.54	1.46	1.42	1.08	0.69	1.60
## 8	-0.75	1.75	1.72	1.45	0.41	2.15	0.58	1.57	2.00	1.60
## 9	-1.22	1.91	-0.25	2.01	-0.19	1.49	0.08	1.68	0.50	1.48
## 10	0.28	1.99	1.42	1.46	0.30	1.76	1.56	1.32	1.31	1.33
## 11	0.50	1.48	1.50	1.87	0.35	1.69	0.72	1.49	1.56	1.56
## 12	-0.50	1.73	1.69	1.12	0.27	1.81	1.31	1.88	1.75	1.61
## 13	1.39	1.34	1.61	1.36	1.32	1.27	2.06	1.04	1.78	1.12
## 14	-0.11	1.62	1.31	1.56	0.84	1.59	0.42	1.18	0.17	1.48
## 15	-0.86	1.71	0.83	1.48	-0.22	1.69	1.14	1.33	0.67	1.53
## 16	0.92	1.75	1.75	1.30	1.95	1.05	1.78	1.10	1.56	1.27
## 17	1.31	1.58	1.69	1.39	1.76	1.32	1.17	1.48	2.00	1.15
## 18	-0.08	1.87	1.69	1.49	1.11	1.93	1.83	1.30	1.14	1.69
## 19	1.11	1.53	0.67	1.79	0.24	1.59	-0.06	1.47	2.03	1.11
## 20	-1.14	1.90	-0.19	1.69	-0.08	1.86	0.53	1.65	0.50	1.90
## 21	-0.06	1.35	1.03	1.75	0.49	1.66	0.33	1.51	1.39	1.42
## 22	-1.83	1.54	-1.28	1.80	-1.22	1.58	0.69	1.88	0.47	2.25
## 23	0.44	1.46	0.78	1.84	0.54	1.71	1.39	1.38	1.00	1.47
## 24	-0.22	1.73	0.56	1.78	-0.38	1.53	1.22	1.42	1.19	1.62
##	PARO_M	PARO_SD	POPPIE_M	POPPIE_SD	ROBOT	BOSS_M	ROBOT	BOSS_SD	SAM_M	SAM_SD
## 1	-2.62	1.01	0.95	1.54		-0.97		1.99	-1.75	1.86
## 2	-1.59	1.74	0.87	1.80		0.12		1.89	1.78	1.40
## 3	0.54	1.79	-0.18	1.72		-1.24		1.69	-0.42	2.41
## 4	0.54	1.61	0.45	1.81		-0.47		1.81	1.11	1.94
## 5	1.95	1.56	0.87	1.55		0.24		1.79	0.42	1.89
## 6	2.05	1.03	0.68	1.86		1.06		1.76	2.22	0.96
## 7	1.76	1.42	0.92	1.65		0.88		1.79	1.97	1.16
## 8	1.62	1.53	0.13	1.77		-0.65		2.12	1.72	1.34
## 9	0.84	1.57	0.42	1.65		-0.62		1.94	1.06	1.67
## 10	0.73	1.71	0.16	1.42		0.21		1.79	1.03	1.72
## 11	2.03	1.19	1.53	1.45		0.47		1.71	1.89	1.28
## 12	1.30	1.70	-0.13	2.20		-0.24		1.88	1.75	1.36
## 13	2.27	0.90	1.84	1.48		1.44		1.85	2.36	0.83
## 14	0.68	1.65	0.18	1.57		0.32		1.82	0.86	1.48
## 15	-0.11	1.74	-0.26	1.50		-0.41		1.83	-0.06	1.62
## 16	0.95	1.39	1.00	1.61		0.59		1.60	1.75	1.11
## 17	1.73	1.17	1.18	1.75		0.91		1.68	2.25	0.97
## 18	-0.38	1.38	0.61	1.76		0.79		1.82	1.53	1.58
## 19	2.49	1.02	1.63	1.57		-0.59		1.78	1.47	1.44
## 20	0.65	1.38	0.05	1.68		-0.50		1.81	0.58	1.65
## 21	1.65	1.48	0.37	1.76		-0.32		1.87	1.17	1.58
## 22	-0.30	1.98	-0.21	1.85		-0.06		2.01	0.42	2.06
## 23	2.05	0.97	0.61	1.62		0.62		1.81	1.69	1.37
## 24	0.62	1.57	0.37	1.73		0.62		1.72	1.25	1.20
##	SARAH_M	SARAH_SD	SIM	SENSEI_M	SIM	SENSEI_SD	SIRI_M	SIRI_SD		
## 1	1.87	1.26		1.42		1.57	-1.72	1.61		
## 2	0.87	1.40		1.21		1.61	-0.03	1.77		
## 3	1.00	1.56		0.71		1.74	-0.28	1.65		

## 4	0.36	1.75	0.50	1.80	-0.46	1.85
## 5	1.85	1.01	1.66	1.49	0.90	1.68
## 6	1.72	1.30	1.45	1.33	2.26	0.88
## 7	1.87	1.10	1.26	1.37	2.28	0.89
## 8	0.59	1.73	0.53	1.57	1.38	1.29
## 9	-0.46	1.76	-0.26	1.73	-0.87	2.10
## 10	0.18	1.37	0.32	1.56	0.33	1.59
## 11	1.31	1.15	0.58	1.33	2.36	0.96
## 12	0.41	1.85	0.66	1.76	0.97	1.74
## 13	1.59	0.91	1.11	1.27	1.38	1.33
## 14	1.49	1.00	0.37	1.75	1.49	1.05
## 15	0.41	1.35	-0.79	1.23	0.03	1.72
## 16	1.79	1.15	1.61	1.31	1.36	1.58
## 17	2.08	1.09	1.61	1.35	2.10	1.14
## 18	1.69	1.28	1.24	1.75	1.87	1.40
## 19	1.62	1.04	0.84	1.42	2.05	0.97
## 20	0.03	1.71	0.05	1.80	-0.95	1.78
## 21	1.08	1.18	0.71	1.52	1.28	1.30
## 22	-0.90	1.82	0.16	1.60	-1.67	1.56
## 23	0.05	1.62	-0.45	1.74	-0.64	1.84
## 24	-0.10	1.19	0.79	1.61	-1.03	1.56
##	THE AMBIENT	LIGHT TV_M	THE AMBIENT	LIGHT TV_SD	THE NEGOTIATOR_M	
## 1		-2.20		1.57	0.81	
## 2		-1.43		1.72	0.72	
## 3		-0.60		1.82	0.33	
## 4		-0.86		2.03	0.25	
## 5		2.03		0.89	1.72	
## 6		2.00		1.19	1.03	
## 7		2.06		1.30	1.17	
## 8		2.00		1.28	0.03	
## 9		-1.69		1.51	-0.64	
## 10		-0.49		2.21	0.33	
## 11		1.57		1.48	0.56	
## 12		1.86		1.44	0.67	
## 13		1.74		1.54	2.22	
## 14		0.66		1.81	0.25	
## 15		-0.77		1.85	-0.28	
## 16		0.17		1.98	1.50	
## 17		1.69		1.53	1.14	
## 18		0.69		1.79	1.42	
## 19		1.46		1.40	0.03	
## 20		-1.69		1.66	-0.72	
## 21		1.34		1.55	0.39	
## 22		-1.26		1.74	-0.67	
## 23		0.66		1.85	1.17	
## 24		-0.66		1.88	0.81	
##	THE NEGOTIATOR_SD	THE ZOMBIE_M	THE ZOMBIE_SD			
## 1		1.79	2.00	1.35		
## 2		1.54	1.37	1.35		
## 3		1.77	0.29	1.58		
## 4		1.79	1.74	1.31		
## 5		1.26	0.60	1.72		
## 6		1.50	-0.34	1.53		
## 7		1.44	0.63	1.21		

## 8		1.61	-0.54	2.06							
## 9		1.57	-1.11	1.83							
## 10		1.47	0.86	1.77							
## 11		1.08	-0.20	1.80							
## 12		1.84	1.00	1.39							
## 13		0.93	2.20	0.90							
## 14		1.40	-0.37	1.68							
## 15		1.65	-0.80	1.71							
## 16		1.38	1.29	1.23							
## 17		1.78	-0.09	1.56							
## 18		1.42	1.17	1.52							
## 19		1.30	-0.40	1.70							
## 20		1.65	-0.23	1.55							
## 21		1.46	-0.51	1.90							
## 22		1.96	0.23	2.06							
## 23		1.83	2.31	0.93							
## 24		1.51	1.31	1.18							
##	Description	AIBO	ALEXA	ALICE	AMY	CHAPPIE	C3PO	DEEPBLUE	DOG	EFFIE	THE FISH
## 1	Sample Size	39	37	35	39	38	36	39	39	34	37
## 2	ASAQ score	21	12	10	9	19	20	7	32	-2	32
##	FURBY FURHAT	GEMINOID	HAL	9000	iCAT	KITT	LOLA	MARCUS	NAO	PARO	POPPIE
## 1	39 34	35	37	36	36	37	36	36	37	38	
## 2	13 14	18	15	-5	20	15	25	24	21	14	
##	ROBOT BOSS	SAM	SARAH	SIM	SENSEI	SIRI	THE AMBIENT	LIGHT	TV	THE NEGOTIATOR	
## 1	34 36	39	38	39					35	36	
## 2	2 28	22	17	14					8	14	
##	THE ZOMBIE										
## 1	35										
## 2	12										

Table D.11: The percentile scores of the ASAQ constructs/dimensions based on the long version of the ASAQ representative set 2024 (n=1066).

```
showTable(percentileLongASAQScore_combinedData, constructASAQdesc$CABR,
           constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)
```

##	CABR	CNAME	q5th	q10th	q20th	q25th	q30th	q40th	q50th
## 1	HLA	Human-Like Appearance	-2.45	-2.25	-2.11	-2.01	-1.94	-1.67	-1.09
## 2	HLB	Human-Like Behavior	-1.34	-1.13	-0.42	-0.29	-0.24	-0.12	0.14
## 3	NA	Natural Appearance	-1.28	-1.21	-0.93	-0.85	-0.83	-0.63	-0.35
## 4	NB	Natural Behavior	-1.30	-1.10	-1.00	-0.85	-0.78	-0.60	-0.38
## 5	AAS	Agent's Appearance Suit.	0.41	0.63	0.80	0.95	1.04	1.23	1.41
## 6	AU	Agent's Usability	0.51	0.73	0.83	0.85	0.93	1.21	1.27
## 7	PF	Performance	0.46	0.57	0.94	0.95	1.04	1.12	1.16
## 8	AL	Agent's Likeability	-0.59	-0.37	0.16	0.33	0.38	0.63	0.77
## 9	AS	Agent's Sociability	-0.83	-0.55	-0.26	-0.24	-0.09	0.04	0.15
## 10	APP	Agent's Personality Pr.	-0.74	-0.64	-0.53	-0.52	-0.35	-0.12	0.04
## 11	UAA	User Acceptance of the A.	0.02	0.37	0.59	0.66	0.75	0.99	1.10
## 12	AE	Agent's Enjoyability	-0.15	0.26	0.64	0.71	0.86	1.04	1.18
## 13	UE	User's Engagement	0.91	1.23	1.44	1.50	1.55	1.67	1.78
## 14	UT	User's Trust	-0.37	-0.13	0.02	0.07	0.15	0.30	0.40
## 15	UAL	User Agent Alliance	-0.18	-0.15	0.13	0.22	0.33	0.50	0.59
## 16	AA	Agent's Attentiveness	0.65	1.08	1.35	1.40	1.44	1.67	1.75
## 17	AC	Agent's Coherence	0.75	1.00	1.16	1.25	1.36	1.48	1.58
## 18	AI	Agent's Intentionality	-0.34	-0.07	0.16	0.21	0.38	0.52	0.68

```

## 19  AT                      Attitude -0.33  0.02  0.27  0.33  0.38  1.08  1.41
## 20  SP                      Social Presence -1.18 -1.02 -0.85 -0.59 -0.44 -0.27 -0.18
## 21  IIS Interaction Impact on Self. -0.47 -0.11  0.03  0.16  0.22  0.38  0.63
## 22  AEI Agent's Emotional Int. Pr. -1.88 -1.74 -1.46 -1.42 -1.30 -1.10 -0.92
## 23  UEP      User's Emotion Presence -0.39 -0.34  0.10  0.15  0.28  0.37  0.49
## 24  UAI      User Agent Interplay -0.09  0.06  0.22  0.40  0.54  0.62  0.68
##    q60th q70th q75th q80th q90th q95th
## 1  -0.40  0.78  0.96  1.05  1.67  1.75
## 2   0.29  0.51  0.52  0.61  0.78  1.31
## 3  -0.02  0.39  0.41  0.61  0.78  1.37
## 4  -0.36 -0.31 -0.22  0.09  0.84  1.62
## 5   1.51  1.69  1.74  1.81  1.92  2.03
## 6   1.37  1.67  1.71  1.76  1.94  2.01
## 7   1.27  1.45  1.54  1.58  1.73  1.87
## 8   0.85  1.15  1.27  1.41  1.61  1.96
## 9   0.45  0.69  0.75  0.83  1.22  1.39
## 10  0.25  0.61  0.77  0.95  1.32  1.37
## 11  1.35  1.42  1.51  1.58  1.95  2.02
## 12  1.30  1.49  1.63  1.84  1.97  2.19
## 13  1.91  2.02  2.08  2.11  2.20  2.25
## 14  0.50  0.77  0.99  1.02  1.17  1.19
## 15  0.65  0.74  0.79  0.83  0.99  1.05
## 16  1.80  1.90  1.96  2.01  2.13  2.15
## 17  1.64  1.75  1.76  1.87  1.97  2.06
## 18  0.81  1.15  1.36  1.41  1.51  1.59
## 19  1.55  1.70  1.73  1.88  2.33  2.40
## 20 -0.03  0.04  0.06  0.15  0.41  0.52
## 21  0.78  0.86  0.90  0.93  1.14  1.33
## 22 -0.70 -0.20 -0.06  0.14  0.79  1.00
## 23  0.81  1.01  1.18  1.35  1.58  1.74
## 24  0.85  0.99  1.03  1.09  1.32  1.51
##    Description  q5th q10th q20th q25th q30th q40th q50th q60th q70th q75th
## 1      Mean: -0.42 -0.20  0.05  0.14  0.24  0.43  0.60  0.77  1.01  1.10
## 2      SD:  0.85  0.89  0.89  0.87  0.87  0.87  0.80  0.73  0.65  0.64
## 3     Median: -0.36 -0.12  0.15  0.22  0.36  0.51  0.66  0.81  1.00  1.10
## 4      Min: -2.45 -2.25 -2.11 -2.01 -1.94 -1.67 -1.09 -0.70 -0.31 -0.22
## 5      Max:  0.91  1.23  1.44  1.50  1.55  1.67  1.78  1.91  2.02  2.08
##    q80th q90th q95th
## 1   1.20  1.47  1.65
## 2   0.61  0.53  0.46
## 3   1.22  1.54  1.68
## 4   0.09  0.41  0.52
## 5   2.11  2.33  2.40

```

Table D.12: The percentile scores of the ASAQ constructs/dimensions based on the short version of the ASAQ representative set 2024 (n=1066).

```

showTable(percentileShortASAQScore_combinedData, constructASAQdesc$CABR,
           constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)

```

```

##    CABR                      CNAME  q5th q10th q20th q25th q30th q40th q50th
## 1   HLA      Human-Like Appearance -2.66 -2.60 -2.20 -2.19 -2.13 -1.74 -1.17
## 2   HLB      Human-Like Behavior  -1.53 -1.26 -0.45 -0.35 -0.02  0.13  0.44
## 3    NA      Natural Appearance  -1.47 -1.17 -1.06 -0.97 -0.95 -0.59 -0.46
## 4    NB      Natural Behavior   -0.96 -0.88 -0.61 -0.46 -0.31  0.03  0.25

```

## 5	AAS	Agent's Appearance Suit.	0.43	0.57	0.82	0.87	0.93	1.18	1.49		
## 6	AU	Agent's Usability	0.42	0.71	1.03	1.06	1.13	1.28	1.45		
## 7	PF	Performance	0.67	0.70	0.89	0.92	1.19	1.31	1.51		
## 8	AL	Agent's Likeability	-0.61	-0.41	0.09	0.41	0.41	0.58	0.85		
## 9	AS	Agent's Sociability	-1.41	-1.14	-1.02	-0.87	-0.68	-0.47	-0.38		
## 10	APP	Agent's Personality Pr.	-0.44	-0.37	0.03	0.14	0.17	0.28	0.33		
## 11	UAA	User Acceptance of the A.	0.05	0.36	0.53	0.58	0.73	1.01	1.31		
## 12	AE	Agent's Enjoyability	-0.25	-0.16	0.17	0.33	0.46	0.66	0.89		
## 13	UE	User's Engagement	0.92	1.10	1.39	1.44	1.52	1.59	1.74		
## 14	UT	User's Trust	-0.27	-0.09	0.20	0.24	0.28	0.37	0.57		
## 15	UAL	User Agent Alliance	-0.80	-0.78	-0.46	-0.37	-0.27	-0.20	-0.08		
## 16	AA	Agent's Attentiveness	0.41	0.84	0.98	1.11	1.25	1.49	1.61		
## 17	AC	Agent's Coherence	0.74	1.08	1.18	1.27	1.31	1.47	1.57		
## 18	AI	Agent's Intentionality	-0.47	-0.14	0.56	0.61	0.73	1.00	1.14		
## 19	AT	Attitude	-0.26	-0.01	0.36	0.47	0.56	1.00	1.38		
## 20	SP	Social Presence	-1.29	-1.14	-0.87	-0.72	-0.41	-0.19	-0.13		
## 21	IIS	Interaction Impact on Self.	-0.22	-0.03	0.38	0.49	0.53	0.69	0.90		
## 22	AEI	Agent's Emotional Int. Pr.	-1.77	-1.52	-1.27	-1.26	-1.11	-0.89	-0.36		
## 23	UEP	User's Emotion Presence	-0.50	-0.36	-0.07	0.03	0.12	0.55	0.65		
## 24	UAI	User Agent Interplay	-0.94	-0.81	-0.49	-0.22	-0.07	0.20	0.38		
##	q60th	q70th	q75th	q80th	q90th	q95th					
## 1	-0.28	0.91	1.11	1.21	1.71	1.95					
## 2	0.65	0.77	0.87	0.91	1.24	1.62					
## 3	-0.22	0.31	0.49	0.57	0.77	1.66					
## 4	0.43	0.52	0.54	0.77	1.68	2.07					
## 5	1.65	1.80	1.87	1.94	2.01	2.09					
## 6	1.64	1.79	1.92	1.97	2.19	2.22					
## 7	1.54	1.71	1.87	1.93	2.10	2.29					
## 8	1.23	1.46	1.62	1.72	2.00	2.31					
## 9	-0.25	0.05	0.13	0.43	0.60	0.97					
## 10	0.72	0.84	1.03	1.25	1.39	1.43					
## 11	1.55	1.61	1.71	1.80	2.03	2.07					
## 12	1.05	1.37	1.42	1.56	1.77	1.96					
## 13	1.83	2.02	2.08	2.16	2.28	2.34					
## 14	0.67	0.83	0.86	1.10	1.34	1.49					
## 15	-0.01	0.38	0.41	0.48	0.82	0.93					
## 16	1.67	1.75	1.78	1.81	1.95	2.12					
## 17	1.68	1.75	1.92	2.00	2.09	2.17					
## 18	1.22	1.48	1.57	1.69	1.79	1.86					
## 19	1.47	1.58	1.63	2.01	2.22	2.43					
## 20	-0.02	0.05	0.20	0.45	0.60	0.75					
## 21	1.01	1.14	1.21	1.24	1.31	1.37					
## 22	-0.09	0.20	0.42	0.49	0.86	1.32					
## 23	0.96	1.14	1.39	1.61	1.87	2.18					
## 24	0.62	0.80	1.05	1.21	1.27	1.33					
##	Description	q5th	q10th	q20th	q25th	q30th	q40th	q50th	q60th	q70th	q75th
## 1	Mean:	-0.51	-0.31	0.00	0.11	0.22	0.45	0.66	0.86	1.09	1.21
## 2	SD:	0.89	0.92	0.89	0.89	0.87	0.84	0.80	0.71	0.62	0.61
## 3	Median:	-0.45	-0.26	0.13	0.29	0.34	0.56	0.75	0.98	1.14	1.30
## 4	Min:	-2.66	-2.60	-2.20	-2.19	-2.13	-1.74	-1.17	-0.28	0.05	0.13
## 5	Max:	0.92	1.10	1.39	1.44	1.52	1.59	1.74	1.83	2.02	2.08
##	q80th	q90th	q95th								
## 1	1.35	1.58	1.79								
## 2	0.58	0.54	0.49								

```
## 3  1.41  1.74  1.96
## 4  0.43  0.60  0.75
## 5  2.16  2.28  2.43
```

Table D.13: The percentile scores of the difference of the ASAQ construct/dimension scores (based on the representative set 2024 of the long version of the ASAQ, n=1066).

```
showTable(percentileDiffLongTwoASAQscore_combinedData, constructASAQdesc$CABR,
           constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)
```

##	CABR	CNAME	q5th	q10th	q20th	q25th	q30th	q40th	q50th
## 1	HLA	Human-Like Appearance	0.12	0.21	0.41	0.58	0.73	1.09	1.56
## 2	HLB	Human-Like Behavior	0.08	0.14	0.28	0.34	0.42	0.57	0.77
## 3	NA	Natural Appearance	0.09	0.16	0.30	0.38	0.47	0.66	0.90
## 4	NB	Natural Behavior	0.05	0.12	0.26	0.32	0.38	0.52	0.72
## 5	AAS	Agent's Appearance Suit.	0.07	0.12	0.22	0.28	0.31	0.42	0.53
## 6	AU	Agent's Usability	0.04	0.09	0.21	0.27	0.34	0.43	0.51
## 7	PF	Performance	0.05	0.08	0.16	0.19	0.23	0.35	0.44
## 8	AL	Agent's Likeability	0.09	0.16	0.30	0.37	0.44	0.61	0.74
## 9	AS	Agent's Sociability	0.08	0.16	0.29	0.33	0.41	0.57	0.71
## 10	APP	Agent's Personality Pr.	0.06	0.12	0.28	0.36	0.46	0.60	0.75
## 11	UAA	User Acceptance of the A.	0.05	0.12	0.25	0.31	0.39	0.50	0.62
## 12	AE	Agent's Enjoyability	0.08	0.15	0.28	0.34	0.41	0.55	0.69
## 13	UE	User's Engagement	0.06	0.08	0.15	0.20	0.24	0.32	0.42
## 14	UT	User's Trust	0.05	0.10	0.20	0.25	0.30	0.41	0.53
## 15	UAL	User Agent Alliance	0.04	0.08	0.15	0.20	0.23	0.31	0.41
## 16	AA	Agent's Attentiveness	0.04	0.08	0.16	0.19	0.24	0.32	0.40
## 17	AC	Agent's Coherence	0.04	0.08	0.16	0.19	0.22	0.31	0.40
## 18	AI	Agent's Intentionality	0.07	0.13	0.26	0.32	0.37	0.55	0.68
## 19	AT	Attitude	0.08	0.16	0.32	0.41	0.54	0.74	0.94
## 20	SP	Social Presence	0.06	0.12	0.21	0.26	0.31	0.43	0.53
## 21	IIS	Interaction Impact on Self.	0.05	0.10	0.19	0.27	0.32	0.44	0.57
## 22	AEI	Agent's Emotional Int. Pr.	0.09	0.19	0.35	0.42	0.50	0.71	0.91
## 23	UEP	User's Emotion Presence	0.07	0.17	0.28	0.36	0.46	0.58	0.73
## 24	UAI	User Agent Interplay	0.07	0.11	0.20	0.26	0.31	0.41	0.51
##	q60th	q70th	q75th	q80th	q90th	q95th			
## 1	2.21	2.83	3.01	3.17	3.67	3.97			
## 2	0.94	1.14	1.28	1.46	1.87	2.11			
## 3	1.09	1.31	1.46	1.60	2.02	2.51			
## 4	0.84	1.13	1.42	1.57	2.30	2.76			
## 5	0.66	0.80	0.91	1.01	1.25	1.43			
## 6	0.66	0.84	0.92	0.97	1.33	1.62			
## 7	0.56	0.66	0.74	0.79	1.02	1.21			
## 8	0.93	1.17	1.32	1.40	1.77	2.08			
## 9	0.87	1.06	1.17	1.28	1.62	1.86			
## 10	0.95	1.23	1.34	1.43	1.81	1.99			
## 11	0.78	0.92	1.02	1.15	1.47	1.65			
## 12	0.85	1.05	1.17	1.28	1.62	1.92			
## 13	0.52	0.63	0.69	0.76	1.00	1.26			
## 14	0.69	0.80	0.88	0.97	1.19	1.39			
## 15	0.49	0.62	0.71	0.77	0.96	1.11			
## 16	0.50	0.65	0.74	0.84	1.20	1.50			
## 17	0.49	0.61	0.69	0.80	1.04	1.28			
## 18	0.84	1.00	1.10	1.23	1.48	1.74			
## 19	1.16	1.42	1.55	1.69	2.13	2.35			

```

## 20 0.70 0.86 0.96 1.03 1.31 1.52
## 21 0.67 0.84 0.91 0.99 1.31 1.54
## 22 1.14 1.44 1.59 1.80 2.24 2.56
## 23 0.89 1.10 1.21 1.36 1.62 1.91
## 24 0.62 0.76 0.86 0.97 1.19 1.44
##      Description q5th q10th q20th q25th q30th q40th q50th q60th q70th q75th q80th
## 1      Mean: 0.07 0.13 0.24 0.31 0.38 0.52 0.67 0.84 1.04 1.15 1.26
## 2      SD: 0.02 0.04 0.07 0.09 0.12 0.18 0.25 0.35 0.46 0.48 0.51
## 3      Median: 0.06 0.12 0.26 0.32 0.38 0.51 0.65 0.81 0.96 1.06 1.19
## 4      Min: 0.04 0.08 0.15 0.19 0.22 0.31 0.40 0.49 0.61 0.69 0.76
## 5      Max: 0.12 0.21 0.41 0.58 0.73 1.09 1.56 2.21 2.83 3.01 3.17
##      q90th q95th
## 1      1.60 1.86
## 2      0.59 0.64
## 3      1.48 1.69
## 4      0.96 1.11
## 5      3.67 3.97

```

Table D.14: The percentile scores of the difference of the ASAQ construct/dimension scores (based on the representative set 2024 of the short version of the ASAQ, n=1066).

```

showTable(percentileDiffShortTwoASAQscore_combinedData, constructASAQdesc$CABR,
           constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)

```

```

##      CABR      CNAME q5th q10th q20th q25th q30th q40th q50th
## 1      HLA      Human-Like Appearance 0.08 0.22 0.48 0.55 0.78 1.13 1.75
## 2      HLB      Human-Like Behavior 0.10 0.18 0.34 0.43 0.52 0.71 0.92
## 3      NA       Natural Appearance 0.08 0.15 0.33 0.42 0.53 0.71 0.91
## 4      NB       Natural Behavior 0.08 0.17 0.36 0.42 0.54 0.72 0.96
## 5      AAS      Agent's Appearance Suit. 0.06 0.13 0.23 0.29 0.33 0.46 0.62
## 6      AU       Agent's Usability 0.04 0.14 0.23 0.29 0.35 0.46 0.58
## 7      PF       Performance 0.05 0.10 0.22 0.26 0.33 0.44 0.56
## 8      AL       Agent's Likeability 0.10 0.18 0.38 0.46 0.56 0.78 0.97
## 9      AS       Agent's Sociability 0.08 0.15 0.32 0.38 0.45 0.61 0.75
## 10     APP      Agent's Personality Pr. 0.05 0.11 0.21 0.31 0.40 0.53 0.66
## 11     UAA      User Acceptance of the A. 0.05 0.14 0.26 0.33 0.41 0.54 0.73
## 12     AE       Agent's Enjoyability 0.10 0.16 0.33 0.40 0.46 0.63 0.79
## 13     UE       User's Engagement 0.05 0.09 0.17 0.22 0.25 0.35 0.46
## 14     UT       User's Trust 0.05 0.10 0.21 0.26 0.32 0.44 0.54
## 15     UAL      User Agent Alliance 0.05 0.11 0.21 0.28 0.33 0.46 0.57
## 16     AA       Agent's Attentiveness 0.05 0.09 0.17 0.21 0.26 0.37 0.48
## 17     AC       Agent's Coherence 0.04 0.09 0.17 0.22 0.27 0.38 0.44
## 18     AI       Agent's Intentionality 0.06 0.14 0.28 0.34 0.43 0.56 0.68
## 19     AT       Attitude 0.08 0.16 0.37 0.44 0.53 0.70 0.92
## 20     SP       Social Presence 0.05 0.11 0.23 0.30 0.38 0.56 0.70
## 21     IIS      Interaction Impact on Self. 0.05 0.11 0.19 0.24 0.30 0.39 0.53
## 22     AEI      Agent's Emotional Int. Pr. 0.08 0.23 0.38 0.52 0.60 0.83 1.05
## 23     UEP      User's Emotion Presence 0.10 0.18 0.38 0.45 0.55 0.69 0.92
## 24     UAI      User Agent Interplay 0.06 0.16 0.28 0.40 0.46 0.63 0.81
##      q60th q70th q75th q80th q90th q95th
## 1      2.51 3.08 3.31 3.53 4.00 4.30
## 2      1.18 1.40 1.62 1.80 2.30 2.63
## 3      1.20 1.46 1.60 1.72 2.35 3.01
## 4      1.20 1.40 1.55 1.78 2.30 2.74
## 5      0.76 0.96 1.05 1.13 1.40 1.56

```



```

## 6 0.73 0.92 0.98 1.12 1.47 1.77
## 7 0.68 0.82 0.89 0.99 1.25 1.40
## 8 1.16 1.39 1.52 1.69 2.12 2.48
## 9 0.90 1.13 1.24 1.41 1.69 2.06
## 10 0.79 1.04 1.11 1.21 1.41 1.75
## 11 0.89 1.08 1.19 1.31 1.63 1.81
## 12 0.97 1.15 1.28 1.44 1.78 2.00
## 13 0.58 0.67 0.75 0.81 0.99 1.26
## 14 0.68 0.84 0.93 1.03 1.26 1.43
## 15 0.69 0.83 0.94 1.07 1.28 1.54
## 16 0.62 0.75 0.83 0.94 1.31 1.50
## 17 0.55 0.69 0.78 0.87 1.12 1.48
## 18 0.90 1.13 1.21 1.36 1.78 2.16
## 19 1.08 1.37 1.51 1.62 2.04 2.38
## 20 0.86 1.01 1.14 1.25 1.61 1.87
## 21 0.64 0.78 0.88 0.96 1.27 1.54
## 22 1.32 1.54 1.72 1.88 2.37 2.73
## 23 1.10 1.31 1.47 1.64 2.04 2.34
## 24 1.00 1.21 1.36 1.46 1.89 2.11
## Description q5th q10th q20th q25th q30th q40th q50th q60th q70th q75th q80th
## 1 Mean: 0.07 0.14 0.28 0.35 0.43 0.59 0.76 0.96 1.17 1.29 1.42
## 2 SD: 0.02 0.04 0.08 0.10 0.13 0.18 0.28 0.40 0.48 0.52 0.55
## 3 Median: 0.06 0.14 0.27 0.34 0.42 0.56 0.72 0.90 1.10 1.20 1.33
## 4 Min: 0.04 0.09 0.17 0.21 0.25 0.35 0.44 0.55 0.67 0.75 0.81
## 5 Max: 0.10 0.23 0.48 0.55 0.78 1.13 1.75 2.51 3.08 3.31 3.53
## q90th q95th
## 1 1.78 2.08
## 2 0.63 0.68
## 3 1.66 1.94
## 4 0.99 1.26
## 5 4.00 4.30

```

Table D.15: The percentile scores of the Cohen's effect sizes (d) (based on the difference of the ASAQ construct/dimension scores of the long version of the ASAQ representative set 2024, n=1066).

```

showTable(percentileEffectSizeLongASAQConstructScore_combinedData,
           constructASAQdesc$CABR, constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)

```

```

## CABR CNAME q5th q10th q20th q25th q30th q40th q50th
## 1 HLA Human-Like Appearance 0.09 0.16 0.33 0.43 0.55 0.76 1.06
## 2 HLB Human-Like Behavior 0.06 0.10 0.20 0.24 0.29 0.41 0.54
## 3 NA Natural Appearance 0.08 0.14 0.24 0.31 0.39 0.54 0.72
## 4 NB Natural Behavior 0.04 0.09 0.20 0.25 0.29 0.40 0.54
## 5 AAS Agent's Appearance Suit. 0.06 0.10 0.19 0.24 0.28 0.37 0.47
## 6 AU Agent's Usability 0.03 0.08 0.18 0.25 0.29 0.37 0.46
## 7 PF Performance 0.05 0.07 0.15 0.19 0.22 0.32 0.42
## 8 AL Agent's Likeability 0.07 0.13 0.24 0.31 0.37 0.49 0.61
## 9 AS Agent's Sociability 0.06 0.12 0.22 0.26 0.30 0.44 0.56
## 10 APP Agent's Personality Pr. 0.05 0.10 0.22 0.28 0.34 0.45 0.57
## 11 UAA User Acceptance of the A. 0.05 0.11 0.21 0.26 0.33 0.43 0.55
## 12 AE Agent's Enjoyability 0.07 0.13 0.24 0.31 0.36 0.48 0.62
## 13 UE User's Engagement 0.06 0.09 0.17 0.20 0.24 0.33 0.43
## 14 UT User's Trust 0.05 0.09 0.17 0.22 0.27 0.36 0.48
## 15 UAL User Agent Alliance 0.04 0.08 0.14 0.18 0.23 0.30 0.37
## 16 AA Agent's Attentiveness 0.05 0.07 0.15 0.19 0.21 0.31 0.38

```

## 17	AC	Agent's Coherence	0.04	0.08	0.15	0.19	0.22	0.30	0.39			
## 18	AI	Agent's Intentionality	0.06	0.11	0.22	0.26	0.31	0.45	0.58			
## 19	AT	Attitude	0.06	0.14	0.27	0.36	0.45	0.61	0.81			
## 20	SP	Social Presence	0.04	0.09	0.15	0.21	0.24	0.32	0.42			
## 21	IIS	Interaction Impact on Self.	0.05	0.09	0.20	0.25	0.29	0.40	0.52			
## 22	AEI	Agent's Emotional Int. Pr.	0.06	0.13	0.25	0.30	0.35	0.49	0.63			
## 23	UEP	User's Emotion Presence	0.06	0.14	0.26	0.31	0.40	0.53	0.64			
## 24	UAI	User Agent Interplay	0.06	0.11	0.19	0.24	0.29	0.38	0.47			
##	q60th	q70th	q75th	q80th	q90th	q95th						
## 1	1.48	2.04	2.17	2.36	2.99	3.32						
## 2	0.69	0.84	0.98	1.13	1.46	1.68						
## 3	0.92	1.11	1.24	1.36	1.77	2.28						
## 4	0.66	0.92	1.10	1.31	2.03	2.44						
## 5	0.57	0.70	0.76	0.85	1.02	1.21						
## 6	0.60	0.74	0.81	0.91	1.19	1.52						
## 7	0.52	0.62	0.68	0.78	1.01	1.20						
## 8	0.77	0.94	1.06	1.13	1.54	1.73						
## 9	0.67	0.81	0.89	1.02	1.27	1.45						
## 10	0.72	0.92	1.02	1.15	1.38	1.62						
## 11	0.68	0.79	0.88	1.02	1.25	1.50						
## 12	0.74	0.95	1.05	1.13	1.44	1.70						
## 13	0.52	0.64	0.68	0.78	0.94	1.07						
## 14	0.58	0.70	0.78	0.87	1.07	1.21						
## 15	0.46	0.57	0.63	0.69	0.85	0.98						
## 16	0.45	0.57	0.65	0.72	1.00	1.20						
## 17	0.48	0.60	0.64	0.74	1.03	1.23						
## 18	0.70	0.85	0.91	1.02	1.26	1.49						
## 19	0.96	1.14	1.30	1.43	1.85	2.09						
## 20	0.52	0.64	0.71	0.76	0.98	1.16						
## 21	0.65	0.77	0.84	0.92	1.16	1.32						
## 22	0.79	0.97	1.12	1.30	1.71	1.98						
## 23	0.77	0.95	1.10	1.19	1.56	1.77						
## 24	0.58	0.71	0.80	0.87	1.06	1.30						
##	Description	q5th	q10th	q20th	q25th	q30th	q40th	q50th	q60th	q70th	q75th	q80th
## 1	Mean:	0.06	0.11	0.21	0.26	0.31	0.43	0.55	0.69	0.85	0.95	1.06
## 2	SD:	0.01	0.02	0.05	0.06	0.08	0.11	0.15	0.22	0.30	0.32	0.35
## 3	Median:	0.06	0.10	0.20	0.25	0.29	0.41	0.54	0.66	0.80	0.88	1.02
## 4	Min:	0.03	0.07	0.14	0.18	0.21	0.30	0.37	0.45	0.57	0.63	0.69
## 5	Max:	0.09	0.16	0.33	0.43	0.55	0.76	1.06	1.48	2.04	2.17	2.36
##	q90th	q95th										
## 1	1.37	1.60										
## 2	0.47	0.53										
## 3	1.25	1.50										
## 4	0.85	0.98										
## 5	2.99	3.32										

Table D.16: The percentile scores of the Cohen's effect sizes (d) (based on the difference of the ASAQ construct/dimension scores of the short version of the ASAQ representative set 2024, n=1066).

```
showTable(percentileEffectSizeShortASAQConstructScore_combinedData,
           constructASAQdesc$CABR, constructASAQdesc$CNAME, 0, 1, 1, 1, 1, 1, 1)
```

##	CABR	CNAME	q5th	q10th	q20th	q25th	q30th	q40th	q50th
## 1	HLA	Human-Like Appearance	0.05	0.14	0.32	0.40	0.50	0.69	1.06
## 2	HLB	Human-Like Behavior	0.06	0.11	0.20	0.25	0.29	0.42	0.53

## 3	NA	Natural Appearance	0.05	0.07	0.17	0.23	0.29	0.39	0.52			
## 4	NB	Natural Behavior	0.04	0.09	0.20	0.23	0.30	0.40	0.52			
## 5	AAS	Agent's Appearance Suit.	0.05	0.09	0.16	0.20	0.24	0.32	0.40			
## 6	AU	Agent's Usability	0.04	0.11	0.18	0.22	0.27	0.35	0.46			
## 7	PF	Performance	0.03	0.08	0.17	0.20	0.26	0.34	0.42			
## 8	AL	Agent's Likeability	0.07	0.11	0.22	0.29	0.35	0.48	0.59			
## 9	AS	Agent's Sociability	0.05	0.08	0.18	0.22	0.26	0.36	0.44			
## 10	APP	Agent's Personality Pr.	0.03	0.07	0.12	0.19	0.24	0.32	0.38			
## 11	UAA	User Acceptance of the A.	0.04	0.10	0.18	0.25	0.29	0.39	0.51			
## 12	AE	Agent's Enjoyability	0.06	0.09	0.19	0.24	0.27	0.36	0.48			
## 13	UE	User's Engagement	0.05	0.08	0.15	0.18	0.21	0.28	0.36			
## 14	UT	User's Trust	0.03	0.07	0.14	0.17	0.20	0.28	0.34			
## 15	UAL	User Agent Alliance	0.03	0.06	0.13	0.17	0.21	0.29	0.35			
## 16	AA	Agent's Attentiveness	0.04	0.07	0.13	0.16	0.19	0.27	0.36			
## 17	AC	Agent's Coherence	0.03	0.07	0.13	0.16	0.20	0.26	0.32			
## 18	AI	Agent's Intentionality	0.04	0.09	0.17	0.21	0.24	0.32	0.40			
## 19	AT	Attitude	0.06	0.12	0.27	0.32	0.39	0.49	0.65			
## 20	SP	Social Presence	0.03	0.06	0.13	0.17	0.23	0.32	0.40			
## 21	IIS	Interaction Impact on Self.	0.03	0.07	0.13	0.16	0.19	0.26	0.35			
## 22	AEI	Agent's Emotional Int. Pr.	0.05	0.12	0.22	0.26	0.32	0.45	0.57			
## 23	UEP	User's Emotion Presence	0.06	0.11	0.23	0.27	0.33	0.44	0.57			
## 24	UAI	User Agent Interplay	0.04	0.10	0.18	0.25	0.30	0.41	0.51			
##	q60th	q70th	q75th	q80th	q90th	q95th						
## 1	1.46	1.89	2.06	2.28	2.88	3.12						
## 2	0.69	0.85	0.94	1.08	1.39	1.64						
## 3	0.66	0.83	0.89	0.98	1.41	1.83						
## 4	0.67	0.80	0.92	1.07	1.50	1.80						
## 5	0.52	0.63	0.70	0.76	0.94	1.02						
## 6	0.56	0.68	0.75	0.82	1.05	1.28						
## 7	0.53	0.63	0.69	0.75	0.96	1.14						
## 8	0.72	0.85	0.93	1.05	1.38	1.60						
## 9	0.53	0.67	0.74	0.83	1.03	1.25						
## 10	0.45	0.62	0.68	0.74	0.92	1.02						
## 11	0.62	0.77	0.86	0.96	1.16	1.38						
## 12	0.57	0.70	0.76	0.86	1.06	1.25						
## 13	0.45	0.56	0.61	0.66	0.86	1.01						
## 14	0.44	0.55	0.59	0.66	0.86	0.99						
## 15	0.44	0.52	0.58	0.67	0.82	0.96						
## 16	0.46	0.54	0.60	0.66	0.88	0.97						
## 17	0.40	0.50	0.55	0.60	0.77	0.98						
## 18	0.54	0.65	0.72	0.80	1.06	1.28						
## 19	0.79	0.95	1.05	1.19	1.47	1.71						
## 20	0.50	0.60	0.65	0.73	0.94	1.10						
## 21	0.44	0.53	0.57	0.62	0.82	0.96						
## 22	0.69	0.86	0.94	1.04	1.41	1.68						
## 23	0.68	0.85	0.93	1.07	1.40	1.64						
## 24	0.65	0.79	0.85	0.97	1.20	1.36						
##	Description	q5th	q10th	q20th	q25th	q30th	q40th	q50th	q60th	q70th	q75th	q80th
## 1	Mean:	0.04	0.09	0.18	0.22	0.27	0.37	0.48	0.60	0.74	0.81	0.91
## 2	SD:	0.01	0.02	0.05	0.06	0.07	0.10	0.15	0.21	0.28	0.30	0.34
## 3	Median:	0.04	0.09	0.17	0.22	0.26	0.36	0.45	0.55	0.68	0.74	0.82
## 4	Min:	0.03	0.06	0.12	0.16	0.19	0.26	0.32	0.40	0.50	0.55	0.60
## 5	Max:	0.07	0.14	0.32	0.40	0.50	0.69	1.06	1.46	1.89	2.06	2.28
##	q90th	q95th										

```
## 1  1.17  1.37
## 2  0.43  0.48
## 3  1.06  1.27
## 4  0.77  0.96
## 5  2.88  3.12
```

Tabel D.17: Sample sizes for each ASAQ construct/dimension of the long and short ASAQ versions based on Cohen's effect sizes (d) from Table D.15 and D.16 with .80 power and .05 alpha level (two-tailed t-test).

```
samplePowerAnalysisASAQConstructScore_combinedData <-
  cbind(samplePowerAnalysisLongASAQConstructScore_combinedData,
        samplePowerAnalysisShortASAQConstructScore_combinedData[, 2:ncol(
          samplePowerAnalysisShortASAQConstructScore_combinedData)])
names(samplePowerAnalysisASAQConstructScore_combinedData) <-
  c("CID", "LONG_Small_q25th", "LONG_Medium_q50th", "LONG_Large_q75th",
    "SHORT_Small_q25th", "SHORT_Medium_q50th", "SHORT_Large_q75th")
showTableSample(samplePowerAnalysisASAQConstructScore_combinedData,
  constructASAQdesc$CABR, constructASAQdesc$CNAME, 0)
```

##	CABR	CNAME	LONG_Small_q25th	LONG_Medium_q50th
## 1	HLA	Human-Like Appearance	86	15
## 2	HLB	Human-Like Behavior	273	55
## 3	NA	Natural Appearance	164	31
## 4	NB	Natural Behavior	252	55
## 5	AAS	Agent's Appearance Suit.	273	72
## 6	AU	Agent's Usability	252	75
## 7	PF	Performance	436	90
## 8	AL	Agent's Likeability	164	43
## 9	AS	Agent's Sociability	233	51
## 10	APP	Agent's Personality Pr.	201	49
## 11	UAA	User Acceptance of the A.	233	53
## 12	AE	Agent's Enjoyability	164	42
## 13	UE	User's Engagement	393	86
## 14	UT	User's Trust	325	69
## 15	UAL	User Agent Alliance	485	116
## 16	AA	Agent's Attentiveness	436	110
## 17	AC	Agent's Coherence	436	104
## 18	AI	Agent's Intentionality	233	48
## 19	AT	Attitude	122	25
## 20	SP	Social Presence	357	90
## 21	IIS	Interaction Impact on Self.	252	59
## 22	AEI	Agent's Emotional Int. Pr.	175	41
## 23	UEP	User's Emotion Presence	164	39
## 24	UAI	User Agent Interplay	273	72
##	LONG_Large_q75th	SHORT_Small_q25th	SHORT_Medium_q50th	SHORT_Large_q75th
## 1	5	99	15	5
## 2	17	252	57	19
## 3	11	298	59	21
## 4	14	298	59	20
## 5	28	393	99	33
## 6	25	325	75	29
## 7	35	393	90	34
## 8	15	188	46	19
## 9	21	325	82	30
## 10	16	436	110	35

## 11	21	252	61	22
## 12	15	273	69	28
## 13	35	485	122	43
## 14	27	544	137	46
## 15	41	544	129	48
## 16	38	614	122	45
## 17	39	614	154	53
## 18	20	357	99	31
## 19	10	154	38	15
## 20	32	544	99	38
## 21	23	614	129	49
## 22	14	233	49	19
## 23	14	216	49	19
## 24	26	252	61	23
##	Description LONG_Small_q25th LONG_Medium_q50th LONG_Large_q75th			
## 1	Max:	485	116	41
##	SHORT_Small_q25th SHORT_Medium_q50th SHORT_Large_q75th			
## 1	614	154	53	

Table D.18: Sample sizes for each ASAQ construct/dimension for the 5th percentile error margin.

```
sampleConfIntervASAQConstructScore_combinedData_p5th <-
  cbind(sampleConfIntervLongASAQConstructScore_combinedData_p5th,
        sampleConfIntervShortASAQConstructScore_combinedData_p5th[, 2:ncol(
          sampleConfIntervShortASAQConstructScore_combinedData_p5th)])
names(sampleConfIntervASAQConstructScore_combinedData_p5th) <-
  c("CID", "LONG_90CI", "LONG_95CI", "LONG_97.5CI", "LONG_99CI",
    "SHORT_90CI", "SHORT_95CI", "SHORT_97.5CI", "SHORT_99CI")
showTableSample(sampleConfIntervASAQConstructScore_combinedData_p5th,
  constructASAQdesc$CABR, constructASAQdesc$CNAME, 0)
```

##	CABR	CNAME	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI
## 1	HLA	Human-Like Appearance	92	130	171	225
## 2	HLB	Human-Like Behavior	235	334	436	576
## 3	NA	Natural Appearance	186	264	346	456
## 4	NB	Natural Behavior	168	239	312	412
## 5	AAS	Agent's Appearance Suit.	454	645	843	1113
## 6	AU	Agent's Usability	506	718	939	1240
## 7	PF	Performance	511	725	948	1252
## 8	AL	Agent's Likeability	209	296	387	511
## 9	AS	Agent's Sociability	305	433	567	748
## 10	APP	Agent's Personality Pr.	337	478	625	826
## 11	UAA	User Acceptance of the A.	297	422	551	728
## 12	AE	Agent's Enjoyability	211	299	391	517
## 13	UE	User's Engagement	495	703	919	1213
## 14	UT	User's Trust	466	661	865	1142
## 15	UAL	User Agent Alliance	689	979	1280	1691
## 16	AA	Agent's Attentiveness	482	684	894	1181
## 17	AC	Agent's Coherence	543	771	1009	1332
## 18	AI	Agent's Intentionality	345	490	641	846
## 19	AT	Attitude	173	245	320	423
## 20	SP	Social Presence	551	783	1023	1352
## 21	IIS	Interaction Impact on Self.	333	472	618	816
## 22	AEI	Agent's Emotional Int. Pr.	212	300	393	519
## 23	UEP	User's Emotion Presence	250	355	464	613

## 24	UAI	User Agent Interplay	414	588	769	1015	
##	SHORT_90CI	SHORT_95CI	SHORT_97.5CI	SHORT_99CI			
## 1	102	145	190	251			
## 2	261	370	484	640			
## 3	293	416	544	719			
## 4	303	430	563	743			
## 5	704	999	1307	1726			
## 6	476	676	884	1167			
## 7	591	839	1097	1449			
## 8	284	403	528	697			
## 9	452	642	840	1110			
## 10	696	988	1292	1706			
## 11	432	614	803	1060			
## 12	518	736	963	1271			
## 13	677	961	1257	1660			
## 14	701	996	1302	1720			
## 15	767	1089	1424	1881			
## 16	575	817	1068	1411			
## 17	871	1237	1618	2137			
## 18	468	665	869	1148			
## 19	243	344	450	595			
## 20	634	901	1178	1556			
## 21	804	1142	1494	1972			
## 22	309	439	574	758			
## 23	311	441	577	762			
## 24	416	591	772	1020			
##	Description	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI	SHORT_90CI	SHORT_95CI
## 1	Max:	689	979	1280	1691	871	1237
##	SHORT_97.5CI	SHORT_99CI					
## 1		1618	2137				

Table D.19: Sample sizes for each ASAQ construct/dimension for the 10th percentile error margin.

```

sampleConfIntervASAQConstructScore_combinedData_p10th <-
  cbind(sampleConfIntervLongASAQConstructScore_combinedData_p10th,
        sampleConfIntervShortASAQConstructScore_combinedData_p10th[, 2:ncol(
          sampleConfIntervShortASAQConstructScore_combinedData_p10th)])
names(sampleConfIntervASAQConstructScore_combinedData_p10th) <-
  c("CID", "LONG_90CI", "LONG_95CI", "LONG_97.5CI", "LONG_99CI",
    "SHORT_90CI", "SHORT_95CI", "SHORT_97.5CI", "SHORT_99CI")
showTableSample(sampleConfIntervASAQConstructScore_combinedData_p10th,
  constructASAQdesc$CABR, constructASAQdesc$CNAME, 0)

```

##	CABR	CNAME	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI
## 1	HLA	Human-Like Appearance	21	30	39	51
## 2	HLB	Human-Like Behavior	90	127	167	220
## 3	NA	Natural Appearance	65	93	121	160
## 4	NB	Natural Behavior	75	107	139	184
## 5	AAS	Agent's Appearance Suit.	142	202	264	349
## 6	AU	Agent's Usability	153	217	283	374
## 7	PF	Performance	148	210	275	363
## 8	AL	Agent's Likeability	68	97	127	167
## 9	AS	Agent's Sociability	94	133	174	230
## 10	APP	Agent's Personality Pr.	77	109	143	189
## 11	UAA	User Acceptance of the A.	95	135	177	234

## 12	AE	Agent's Enjoyability	77	110	144	190	
## 13	UE	User's Engagement	188	267	350	462	
## 14	UT	User's Trust	132	188	246	325	
## 15	UAL	User Agent Alliance	161	228	298	394	
## 16	AA	Agent's Attentiveness	192	273	357	471	
## 17	AC	Agent's Coherence	201	286	374	493	
## 18	AI	Agent's Intentionality	102	144	189	249	
## 19	AT	Attitude	47	67	88	116	
## 20	SP	Social Presence	152	216	282	373	
## 21	IIS	Interaction Impact on Self.	137	195	255	337	
## 22	AEI	Agent's Emotional Int. Pr.	54	77	101	133	
## 23	UEP	User's Emotion Presence	61	87	114	150	
## 24	UAI	User Agent Interplay	131	186	243	321	
##	SHORT_90CI	SHORT_95CI	SHORT_97.5CI	SHORT_99CI			
## 1	23	33	43	57			
## 2	81	115	150	199			
## 3	150	213	279	369			
## 4	84	119	155	205			
## 5	187	265	347	458			
## 6	138	196	257	339			
## 7	156	222	290	384			
## 8	82	116	152	200			
## 9	168	239	313	413			
## 10	154	218	286	377			
## 11	124	176	230	303			
## 12	133	189	247	326			
## 13	196	279	364	481			
## 14	209	297	389	513			
## 15	177	252	329	435			
## 16	267	379	496	655			
## 17	350	497	650	859			
## 18	135	191	250	330			
## 19	70	99	129	171			
## 20	172	244	319	422			
## 21	220	312	408	539			
## 22	102	145	189	250			
## 23	88	125	163	216			
## 24	98	139	182	241			
##	Description	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI	SHORT_90CI	SHORT_95CI
## 1	Max:	201	286	374	493	350	497
##	SHORT_97.5CI	SHORT_99CI					
## 1	650	859					

Table D.20: Sample sizes for each ASAQ construct/dimension for the 20th percentile error margin.

```

sampleConfIntervASAQConstructScore_combinedData_p20th <-
  cbind(sampleConfIntervLongASAQConstructScore_combinedData_p20th,
        sampleConfIntervShortASAQConstructScore_combinedData_p20th[, 2:ncol(
          sampleConfIntervShortASAQConstructScore_combinedData_p20th)])
names(sampleConfIntervASAQConstructScore_combinedData_p20th) <-
  c("CID", "LONG_90CI", "LONG_95CI", "LONG_97.5CI", "LONG_99CI",
    "SHORT_90CI", "SHORT_95CI", "SHORT_97.5CI", "SHORT_99CI")
showTableSample(sampleConfIntervASAQConstructScore_combinedData_p20th,
  constructASAQdesc$CABR, constructASAQdesc$CNAME, 0)

```

##	CABR	CNAME	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI	
## 1	HLA	Human-Like Appearance	5	6	8	11	
## 2	HLB	Human-Like Behavior	43	61	80	105	
## 3	NA	Natural Appearance	15	22	28	37	
## 4	NB	Natural Behavior	34	48	62	83	
## 5	AAS	Agent's Appearance Suit.	33	47	61	81	
## 6	AU	Agent's Usability	36	52	68	89	
## 7	PF	Performance	69	98	128	169	
## 8	AL	Agent's Likeability	24	34	45	59	
## 9	AS	Agent's Sociability	35	50	66	87	
## 10	APP	Agent's Personality Pr.	19	27	35	46	
## 11	UAA	User Acceptance of the A.	34	48	63	83	
## 12	AE	Agent's Enjoyability	22	31	41	54	
## 13	UE	User's Engagement	54	77	101	133	
## 14	UT	User's Trust	31	44	58	77	
## 15	UAL	User Agent Alliance	60	85	111	146	
## 16	AA	Agent's Attentiveness	70	100	130	172	
## 17	AC	Agent's Coherence	52	73	96	126	
## 18	AI	Agent's Intentionality	23	33	43	56	
## 19	AT	Attitude	14	20	26	34	
## 20	SP	Social Presence	44	62	82	108	
## 21	IIS	Interaction Impact on Self.	37	53	69	91	
## 22	AEI	Agent's Emotional Int. Pr.	19	27	35	46	
## 23	UEP	User's Emotion Presence	21	29	38	50	
## 24	UAI	User Agent Interplay	38	55	71	94	
##	SHORT_90CI	SHORT_95CI	SHORT_97.5CI	SHORT_99CI			
## 1	5	7	10	13			
## 2	39	55	72	95			
## 3	30	43	56	74			
## 4	40	57	74	98			
## 5	43	60	79	104			
## 6	48	68	90	118			
## 7	40	57	75	98			
## 8	25	36	47	62			
## 9	34	48	63	84			
## 10	46	65	84	112			
## 11	30	43	56	74			
## 12	36	52	68	89			
## 13	64	91	118	156			
## 14	74	105	137	181			
## 15	71	101	133	175			
## 16	68	97	127	167			
## 17	74	105	137	181			
## 18	55	79	103	136			
## 19	18	25	33	44			
## 20	42	60	78	103			
## 21	76	108	141	186			
## 22	26	37	49	64			
## 23	22	31	40	53			
## 24	21	29	39	51			
##	Description	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI	SHORT_90CI	SHORT_95CI
## 1	Max:	70	100	130	172	76	108
##	SHORT_97.5CI	SHORT_99CI					
## 1	141	186					

Table D.21: Sample sizes for each ASAQ construct/dimension for the 25th percentile error margin.

```
sampleConfIntervASAQConstructScore_combinedData_p25th <-
  cbind(sampleConfIntervLongASAQConstructScore_combinedData_p25th,
        sampleConfIntervShortASAQConstructScore_combinedData_p25th[, 2:ncol(
          sampleConfIntervShortASAQConstructScore_combinedData_p25th)])
names(sampleConfIntervASAQConstructScore_combinedData_p25th) <-
  c("CID", "LONG_90CI", "LONG_95CI", "LONG_97.5CI", "LONG_99CI",
    "SHORT_90CI", "SHORT_95CI", "SHORT_97.5CI", "SHORT_99CI")
showTableSample(sampleConfIntervASAQConstructScore_combinedData_p25th,
  constructASAQdesc$CABR, constructASAQdesc$CNAME, 0)
```

##	CABR	CNAME	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI
## 1	HLA	Human-Like Appearance	2	3	4	6
## 2	HLB	Human-Like Behavior	31	44	57	75
## 3	NA	Natural Appearance	10	14	19	25
## 4	NB	Natural Behavior	44	63	82	108
## 5	AAS	Agent's Appearance Suit.	23	33	43	57
## 6	AU	Agent's Usability	19	27	35	47
## 7	PF	Performance	37	52	68	90
## 8	AL	Agent's Likeability	19	27	35	46
## 9	AS	Agent's Sociability	19	27	35	46
## 10	APP	Agent's Personality Pr.	11	16	21	27
## 11	UAA	User Acceptance of the A.	20	29	37	49
## 12	AE	Agent's Enjoyability	17	24	31	41
## 13	UE	User's Engagement	33	47	61	81
## 14	UT	User's Trust	17	23	31	41
## 15	UAL	User Agent Alliance	40	57	75	99
## 16	AA	Agent's Attentiveness	41	59	77	101
## 17	AC	Agent's Coherence	45	64	84	111
## 18	AI	Agent's Intentionality	12	17	22	29
## 19	AT	Attitude	8	12	15	20
## 20	SP	Social Presence	46	66	86	114
## 21	IIS	Interaction Impact on Self.	25	35	46	61
## 22	AEI	Agent's Emotional Int. Pr.	12	17	22	29
## 23	UEP	User's Emotion Presence	13	19	24	32
## 24	UAI	User Agent Interplay	33	47	61	81
##	SHORT_90CI	SHORT_95CI	SHORT_97.5CI	SHORT_99CI		
## 1	2	4	5	6		
## 2	21	30	40	52		
## 3	17	24	31	41		
## 4	34	48	63	83		
## 5	24	34	44	59		
## 6	25	36	47	62		
## 7	21	30	39	52		
## 8	20	29	37	49		
## 9	31	45	58	77		
## 10	38	54	71	94		
## 11	17	24	31	42		
## 12	27	38	49	65		
## 13	41	59	77	101		
## 14	68	97	126	167		
## 15	46	66	86	114		
## 16	46	65	85	113		
## 17	52	74	97	128		

## 18	34	48	63	83		
## 19	16	23	30	39		
## 20	38	54	71	93		
## 21	48	68	89	117		
## 22	13	18	24	32		
## 23	15	21	27	36		
## 24	16	23	30	40		
##	Description	LONG_90CI	LONG_95CI	LONG_97.5CI	LONG_99CI	SHORT_90CI SHORT_95CI
## 1	Max:	46	66	86	114	68 97
##	SHORT_97.5CI	SHORT_99CI				
## 1	126	167				