

NonlinearityAnalysis

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##Introduction

This notebook presents the code used to create a non-linearity analysis on the basis of keystroke data from the logger Inputlog, and accompanies the paper *Measuring non-linearity of long-term writing processes* (submitted to Reading & Writing special issue on methodology). For context, the abstract of our paper follows below.

Abstract

Non-linearity in writing provides important insight into the dynamics of writing and writing disfluencies. Currently, a range of linearity measures are available. These metrics are calculated based upon the leading edge, and are mostly used for short texts and single writing sessions. However, for longer, multi-session writing processes, the concept of the leading edge, as the singular outer boundary of the text-in-progress, is not enough to distinguish between linear production and non-linear text alterations. Therefore, in the current study, we propose a novel automatized non-linearity analysis. Within this approach, all backwards and forwards cursor and mouse operations from the point of utterance are extracted from keystroke data, and characterized both based on duration and distance. We illustrate this approach by analyzing the writing process of a complete novel based on more than 400 writing sessions totaling 276 hours of writing. The results show that this approach allows us to successfully cluster these writing sessions using the non-linearity characteristics.

Overview of the steps we have taken

We are using Inputlog-General Analysis files as our source material. You could apply the steps to the output of other keystroke loggers as well. The General Analysis is a table in which each keystroke action has its own row. It is quite similar to the raw output from other academic keystroke loggers.

Steps: 1. load all General Analysis files (one for each writing session) into R and merge them into one table, preserving the original session-numbering. (not including in this notebook) 2. aggregate/annotate the keystroke events in this table into larger segments of typing, deleting, jumping/navigating and focus (activities outside of the work doc). 3. Calculate characteristics for each jump event and each typing event 4. Create a summary table with descriptive statistics for each session 5. Running a correlation matrix, removing highly correlation variables, then performing a cluster analysis to explore similarities between sessions. This step is not included in this notebook.

Installation, use and requirements

You need the package Tidyverse to run this notebook.If you do not have it yet, you can use this code:

```
install.packages('tidyverse')
```

```
## Warning: package 'tidyverse' is in use and will not be installed
```

Load data

```
# this is a demo session. Our real data is not shared, due to privacy concerns and agreements with the  
# for example, create a new project in Rstudio, place csv-file in the same folder as this rmd-file.  
all_data <- read.csv("demo_data_nonlinearity.csv", stringsAsFactors = F, fileEncoding = 'latin1')
```

Raw data sneakpeak

This is what our raw keystroke files look like. Each line is a keystroke event.

```
# for just the first five lines  
head(all_data)
```

```
##   event_type__E_ event_id__E_ event_output__E_ event_charProduction__E_  
## 1      mouse           0      Movement           0  
## 2    keyboard           1      LSHIFT           1  
## 3    keyboard           2          T           1  
## 4    keyboard           3          h           2  
## 5    keyboard           4          i           3  
## 6    keyboard           5          s           4  
##   event_actionTime__E_ event_startClock__E_ event_endClock__E_  
## 1           516      00:00:01.953      00:00:02.469  
## 2           225      00:00:03.265      00:00:03.490  
## 3            80      00:00:03.362      00:00:03.442  
## 4           152      00:00:03.634      00:00:03.786  
## 5            96      00:00:03.746      00:00:03.842  
## 6            87      00:00:03.843      00:00:03.930  
##   event_startTime__E_ event_endTime__E_ event_pauseLocation__E_  
## 1           1953           2469           8  
## 2           3265           3490           8  
## 3           3362           3442           4  
## 4           3634           3786           1  
## 5           3746           3842           1  
## 6           3843           3930           1  
##   event_pauseLocationFull__E_ event_pauseTime__E_ event_intervalNumber__E_  
## 1           INITIAL           1953           1  
## 2           INITIAL           796           1  
## 3    BEFORE SENTENCES           97           1  
## 4    WITHIN WORDS           272           1  
## 5    WITHIN WORDS           112           1  
## 6    WITHIN WORDS           97           1  
##   event_intervalSize__E_ event_x__E_ event_y__E_ event_positionFull__E_  
## 1           1           594           229           0  
## 2           1           NA           NA           0
```

```
## 3          1          NA          NA          0
## 4          1          NA          NA          1
## 5          1          NA          NA          2
## 6          1          NA          NA          3
##   event_doclengthFull__E_ event_position__E_ event_doclength__E_
## 1          0          NA          NA
## 2          1          0          1
## 3          1          0          1
## 4          2          1          2
## 5          3          2          3
## 6          4          3          4
##   sessionID_LogCreationDate__S_ sessionID_Restricted_Logging__SL_
## 1      13-9-2021 15:54:00          NA
## 2      13-9-2021 15:54:00          NA
## 3      13-9-2021 15:54:00          NA
## 4      13-9-2021 15:54:00          NA
## 5      13-9-2021 15:54:00          NA
## 6      13-9-2021 15:54:00          NA
##   session_number
## 1          3
## 2          3
## 3          3
## 4          3
## 5          3
## 6          3
```

```
# for all of the data (fine for our sample session, may crash when you have large files)
view(all_data)
```

Add segmentation (step 2)

We start by labeling several groups of keyboard input. We define the category of arrow keys (as they are used to initiate non-linearity), and we distinguish between visible characters, whitespace, deletion keys and function keys (such as the Control key).

```
#####----- Add info -----#####
arrow_keys <- c("UP", "DOWN", "LEFT", "RIGHT", "END", "HOME", "PAGE_DOWN",
               "PAGE_UP")

# Check all possible keystrokes
keys <- all_data %>% filter(event_type__E_ == "keyboard")
keyst <- data.frame(event_output__E_ = unique(keys$event_output__E_),
                   stringsAsFactors = F)

keyst_add <- keyst %>%
  mutate(length = nchar(as.character(event_output__E_)),
         type = ifelse((length < 3 & event_output__E_ != "UP") |
                       grepl("OEM_", event_output__E_), "visible_char",
                       ifelse(event_output__E_ %in% c("SPACE", "RETURN", "TAB"),
                              "whitespace",
                              ifelse(grepl("UP|DOWN|LEFT|RIGHT|END|HOME", event_output__E_),
                                     "arrow_key",
                                     ifelse(event_output__E_ %in% c("BACK", "DELETE"),
```

```
      "delete_key",
      "function_key")))))))
```

Then, we add boundaries between non-linear jumps and the other types of events. In the *#*notes in the code below, you can see how we distinguish between 6 different situations in which we are placing a boundary.

```
# Add boundaries for non-linearity
data_add <- all_data %>%
  # Remove keystrokes outside doc (e.g., save as XXX)
  filter(!(event_type__E_ == "keyboard" & is.na(event_position__E_))) %>%

  left_join(keyst_add) %>%
  # Calculate per session/file separately
  group_by(session_number) %>%
  mutate(jump_start = ifelse(
    row_number() == 1 |
    #1) When a typist moves from typing a character to a mouse event
    # (click, movement, scroll, selection), or vice versa.
    (event_type__E_ == "keyboard" &
      (type %in% c("visible_char", "whitespace") |
        event_output__E_ == "CAPS LOCK" |
          (type == "function_key" &
            event_pauseLocationFull__E_ == "COMBINATION KEY"))) &
    (lag(event_type__E_) == "mouse" | event_type__E_ == "replacement")) |
    ((event_type__E_ == "mouse" | lag(event_type__E_) == "replacement") &
      lag(event_type__E_) == "keyboard" &
        (lag(type) %in% c("visible_char", "whitespace"))) |
    event_output__E_ == "CAPS LOCK") |
    #2) When a typists moves from an insertion to another event, or vice versa.
    (lag(event_type__E_) %in% c("insert") &
      event_type__E_ != lag(event_type__E_) |
      (event_type__E_ %in% c("insert") &
        lag(event_type__E_) != event_type__E_)) |
    #3) When a typist moves from typing a character to typing an arrow key,
    # or vice versa.
    (event_type__E_ == "keyboard" &
      (type %in% c("visible_char", "whitespace") |
        event_output__E_ == "CAPS LOCK" |
          (type == "function_key" &
            event_pauseLocationFull__E_ == "COMBINATION KEY"))) &
      lag(event_type__E_) == "keyboard" & lag(type) == "arrow_key") |
    (event_type__E_ == "keyboard" & type == "arrow_key"
      & (lag(event_type__E_) == "keyboard" &
        lag(type) %in% c("visible_char", "whitespace"))) |
      lag(event_type__E_) == "replacement") |
    #4) When a typists moves from a keystroke or mouse event to a
    # delete/backspace keypress, or vice versa.
    (event_type__E_ %in% c("mouse", "keyboard", "insert") &
      (is.na(type) | type != "delete_key") &
      lag(event_type__E_) == "keyboard" & lag(type) == "delete_key") |
    (event_type__E_ == "keyboard" & type == "delete_key"
      & lag(event_type__E_) %in% c("mouse", "keyboard", "insert") &
      !lag(event_output__E_) %in% c("DELETE", "BACK")) |
```

```

#5) When a typist moves from one mode of deletion to another (e.g., from
#   delete key to backspace key press).
  (type == "delete_key" & lag(type) == "delete_key" &
   (event_output__E_) != lag(event_output__E_)) |
#6) When a typist moves from the main text to a different source
#   (e.g., online dictionary)
  (lag(event_type__E_) %in% c("focus") &
   event_type__E_ != lag(event_type__E_)) |
  (event_type__E_ %in% c("focus") &
   event_type__E_ != lead(event_type__E_) , 1, 0),

#Part B
# Set to zero if selection is directly followed by insert/delete.
# (A series of deletions count as one event.)
#delete - replacement
jump_start = ifelse((event_type__E_ == "replacement" &
  lag(event_output__E_) == "DELETE" &
  event_startClock__E_ == lag(event_startClock__E_) &
  event_endClock__E_ == lag(event_endClock__E_))
| (lag(event_type__E_) == "replacement" &
  event_output__E_ == "DELETE" &
  lead(event_type__E_) == "replacement" &
  lag(event_output__E_, 2) == "DELETE" ) |
  (event_type__E_ == "replacement" &
  lag(event_type__E_) == "replacement") |
  is.na(jump_start), 0, jump_start),

# Create count number of linear event
jump_number = ifelse(jump_start == 1,
  cumsum(jump_start == 1 |
    row_number() == 1), NA),
  prev_loc = lag(event_pauseLocationFull__E_),
  next_loc = lead(event_pauseLocationFull__E_),
  endTime_session = max(event_endTime__E_)
) %>%
fill(jump_number)

```

```
## Joining, by = "event_output__E_"
```

```
## Replacing 0 with NA for further processing
```

```
data_add %>%
mutate(event_charProduction__E_ = ifelse(event_charProduction__E_ == 0, NA, event_charProduction__E_))
```

```
## # A tibble: 728 x 30
## # Groups:   session_number [2]
##   event_type__E_ event_id__E_ event_output__E_ event_charProduction__E_
##   <chr>          <int> <chr>          <int>
## 1 mouse          0 Movement      NA
## 2 keyboard       1 LSHIFT        1
## 3 keyboard       2 T            1
## 4 keyboard       3 h            2
## 5 keyboard       4 i            3
## 6 keyboard       5 s            4
## 7 keyboard       6 SPACE        5
```

```
## 8 keyboard          7 i          6
## 9 keyboard          8 s          7
## 10 keyboard         9 SPACE       8
## # ... with 718 more rows, and 26 more variables: event_actionTime__E_ <int>,
## #   event_startClock__E_ <chr>, event_endClock__E_ <chr>,
## #   event_startTime__E_ <int>, event_endTime__E_ <int>,
## #   event_pauseLocation__E_ <int>, event_pauseLocationFull__E_ <chr>,
## #   event_pauseTime__E_ <int>, event_intervalNumber__E_ <int>,
## #   event_intervalSize__E_ <int>, event_x__E_ <int>, event_y__E_ <int>,
## #   event_positionFull__E_ <int>, event_doclengthFull__E_ <int>,
## #   event_position__E_ <int>, event_doclength__E_ <int>,
## #   sessionID_LogCreationDate__S_ <chr>,
## #   sessionID_Restricted_Logging__SL_ <lgl>, session_number <int>,
## #   length <int>, type <chr>, jump_start <dbl>, jump_number <int>,
## #   prev_loc <chr>, next_loc <chr>, endTime_session <int>
```

In part B (in the code block above), we add a specific rule to aggregate deletions of multiple characters that are done by selecting each character separately - this is something which occurred quite a lot for one particular writer, it may not be necessary for other materials. Also, we add an identification number for each segment (note that although the code calls everything ‘jump’ here it is actually all segments in between the boundaries we set before.)

preview of segmented data

```
head(data_add)
```

```
## # A tibble: 6 x 30
## # Groups:   session_number [1]
##   event_type__E_ event_id__E_ event_output__E_ event_charProdu~ event_actionTim~
##   <chr>          <int> <chr>          <int>          <int>
## 1 mouse          0 Movement          0          516
## 2 keyboard        1 LSHIFT            1          225
## 3 keyboard        2 T              1           80
## 4 keyboard        3 h              2          152
## 5 keyboard        4 i              3           96
## 6 keyboard        5 s              4           87
## # ... with 25 more variables: event_startClock__E_ <chr>,
## #   event_endClock__E_ <chr>, event_startTime__E_ <int>,
## #   event_endTime__E_ <int>, event_pauseLocation__E_ <int>,
## #   event_pauseLocationFull__E_ <chr>, event_pauseTime__E_ <int>,
## #   event_intervalNumber__E_ <int>, event_intervalSize__E_ <int>,
## #   event_x__E_ <int>, event_y__E_ <int>, event_positionFull__E_ <int>,
## #   event_doclengthFull__E_ <int>, event_position__E_ <int>,
## #   event_doclength__E_ <int>, sessionID_LogCreationDate__S_ <chr>,
## #   sessionID_Restricted_Logging__SL_ <lgl>, session_number <int>,
## #   length <int>, type <chr>, jump_start <dbl>, jump_number <int>,
## #   prev_loc <chr>, next_loc <chr>, endTime_session <int>
```

```
# Or for the full table
# view(data_add)
```

Now we will label each segment - adding a column with their type (delete, focus, jumps, typing et cetera). This enables us to later calculate certain values for only the jumps or only the typing chunks, for example.

Add characteristics for each jump (step 3)

For a description of the characteristics, please see our paper.

Add characteristics for each typing segment (step 3B)

Session-level summary table (step 4)

```
# Summary statistics for each session (all, time and word count)
sum_session <- data_add %>%
  group_by(session_number)%>%
  summarize(
    total_time = last(event_endTime__E_) - first(event_startTime__E_),
    total_time_seconds = total_time/1000,
    total_time_minutes = total_time_seconds/60,
    total_charproduced = last(event_charProduction__E_) - first(event_charProduction__E_)
  ) %>%
  mutate(
    char_produced2 = total_charproduced - lag(total_charproduced))

#added total jump time per session
sum_session2 <- jump_filt %>%
  group_by(session_number)%>%
  summarize(
    Jump_time = sum(jump_duration, na.rm = T)
  ) %>% left_join(sum_session)
```

```
## Joining, by = "session_number"
```

```
#view(sum_session2)
```

```
## Added TYPING chunks (size & duration & position relative to leading edge, also total amount of chars
sum_session3 <- sum_typing %>%
  group_by(session_number)%>%
  summarise(
    MeanTypingChars = mean(typing_size_chars),
    sdTypingChars = sd(typing_size_chars),
    totalTypingChars = sum(typing_size_chars, na.rm = T),
    MeanDurationTyping = mean(typing_duration),
    sdDurationTyping = sd(typing_duration),
    MeanTypingPositionRel = mean(start_position_rel),
    sdTypingPositionRel = sd(start_position_rel),
    MeanTypingPositionEdge = mean(start_position_edge),
    sdTypingPositionEdge = sd(start_position_edge),
    totalTypingChars = sum(typing_size_chars)
  ) %>% left_join(sum_session2)
```

```
## Joining, by = "session_number"
```

Next, we are adding information on where in the text a jump starts - for example: within a word, or after a sentence -. We then calculate the % of jumps that fall into each location-category for each session.

```
pivottableStartLoc <- count(jump_filt, start_location)

# Pivot_wider helps to glue the two tables together at the appropriate junctions
pivotwide <- pivot_wider(pivottableStartLoc, names_from = start_location, values_from = n)
pivotfancy <- pivotwide %>%
  rename(
    StartPos_AfterWords = `AFTER WORDS`,
    # StartPos_BeforeParagraphs = `BEFORE PARAGRAPHS`,
    # StartPos_BeforeWords = `BEFORE WORDS`,
    # StartPos_Change = CHANGE,
    StartPos_Deletion = REVISION,
    # StartPos_WithinWords = `WITHIN WORDS`,
    StartPos_BeforeSentences = `BEFORE SENTENCES`,
    StartPos_AfterSentences = `AFTER SENTENCES`)
    # StartPos_unknown = UNKNOWN)

## a number of categories do not occur in our sample sessions. You can remove the hashtag to include

# Adding count of instances within each category

pivotfancy2 <- pivotfancy %>%
  mutate(N = sum(StartPos_AfterWords,
    # StartPos_BeforeParagraphs, StartPos_BeforeWords, StartPos_Change,
    # StartPos_WithinWords, StartPos_unknown,
    StartPos_Deletion, StartPos_BeforeSentences,
    StartPos_AfterSentences, na.rm = T))

# And changing raw counts to percentages ##
pivotfancy3 <- pivotfancy2 %>%
  mutate( StartPos_AfterWords_perc = 100/(N/StartPos_AfterWords),
    # StartPos_BeforeParagraphs_perc = 100/(N/StartPos_BeforeParagraphs),
    # StartPos_BeforeWords_perc = 100/(N/StartPos_BeforeWords),
    # StartPos_Change_perc = 100/(N/StartPos_Change),
    StartPos_Deletion_perc = 100/(N/StartPos_Deletion),
    # StartPos_WithinWords_perc = 100/(N/StartPos_WithinWords),
    StartPos_BeforeSentences_perc = 100/(N/StartPos_BeforeSentences),
    StartPos_AfterSentences_perc = 100/(N/StartPos_AfterSentences))

## Removing unnecessary columns ##
pivotfancy31 <- pivotfancy3 %>%
  select(session_number, StartPos_AfterWords_perc,
    # StartPos_BeforeParagraphs_perc,
    # StartPos_BeforeWords_perc,
    # StartPos_Change_perc,
    StartPos_Deletion_perc,
```



```

      # StartPos_WithinWords_perc,
      StartPos_BeforeSentences_perc,
      StartPos_AfterSentences_perc)

## Replacing missing values by zero (0% instead of NA)
pivotfancy4 <- mutate_at(pivotfancy31, vars(StartPos_AfterWords_perc:StartPos_AfterSentences_perc), ~rep(
  0, n()))

sum_session4 <- sum_session3 %>%
  group_by(session_number)%>%
  mutate( Perc_time_jumps = ifelse(sum(Jump_time > 0) == 0, 0,
                                   100/( total_time / Jump_time))) %>%

  left_join(pivotfancy4)

```

```
## Joining, by = "session_number"
```

And now on to the most important bit - adding descriptive writing session statistics from the jump characteristics.

```

descriptives <- jump_filt %>%
  group_by(session_number)%>%
  summarize(
    # Time-based
    MeanDurationJumps = mean(jump_duration, na.rm = TRUE),
    SD_DurationJumps = sd(jump_duration, na.rm = TRUE),

    # Position change
    MeanJumpsize_chars = mean(jump_size_chars, na.rm = T),
    SDJumpsize_chars = sd(jump_size_chars, na.rm = T),
    MeanJumpsize_rel = mean(jump_size_rel, na.rm = T),
    SDJumpsize_rel = sd(jump_size_rel, na.rm = T),

    # All distances transformed to positive values
    MeanJumpsize_chars_Plus = mean(jump_size_charsPlus, na.rm = T),
    sdJumpsize_chars_Plus = sd(jump_size_charsPlus, na.rm = T),

    # Log transform of MeanJumpsize
    logmeanJumpsize_chars_Plus = log(MeanJumpsize_chars_Plus),

    # Total jumpsize
    TotalJumpsize =sum(jump_size_charsPlus),
    logTotalJumpsize = log(TotalJumpsize),

    # Size of backwards jumps
    MeanJumpSize_BackW = mean(jump_size_chars[jump_size_chars < 0],
                              na.rm = TRUE),
    SDJumpSize_BackW = sd(jump_size_chars[jump_size_chars < 0],
                          na.rm = TRUE),

    # Size of forwards jumps
    MeanJumpSize_Forw = mean(jump_size_chars[jump_size_chars > 0],
                              na.rm = TRUE),

```

```

SDJumpSize_Forw = sd(jump_size_chars[jump_size_chars > 0],
                      na.rm = TRUE),

# Percentile of jumps that is backwards
Countrows = n(),
PercentageBackwardsJumps = ifelse( sum(jump_size_chars < 0) == 0,0,
                                   100/(Countrows / sum(jump_size_chars < 0))),

# Slope = jump size in chars / duration
MeanJumpSlope = mean(jump_slope),
SDJumpSlope = sd(jump_slope),

# Start position relative to the leading edge (in percentile )
MeanStartPos_rel = mean(start_position_rel),
sdStartPos_rel = sd(start_position_rel),

# Start position in characters from leading edge
MeanStartPos_edge = mean(start_position_edge),
sdStartPos_edge = sd(start_position_edge),

# End position relative to the leading edge (in percentile)
MeanEndPos_rel = mean(end_position_rel),
sdEndPos_rel = sd(end_position_rel),

# Content of jumps
Mean_n_events = mean(n_events),
sd_n_events = sd(n_events),
Mean_n_scroll_movements = mean(n_scroll_movements),
sd_n_scroll_movements = sd(n_scroll_movements)

```

Sneakpeak at the output from the previous code block

```

# Viewing the entire table
view(descriptives)

## Overview of the added variables
colnames(descriptives)

## [1] "session_number"           "MeanDurationJumps"
## [3] "SD_DurationJumps"         "MeanJumpsize_chars"
## [5] "SDJumpsize_chars"         "MeanJumpsize_rel"
## [7] "SDJumpsize_rel"           "MeanJumpsize_chars_Plus"
## [9] "sdJumpsize_chars_Plus"    "logmeanJumpsize_chars_Plus"
## [11] "TotalJumpsize"            "logTotalJumpsize"
## [13] "MeanJumpSize_BackW"       "SDJumpSize_BackW"
## [15] "MeanJumpSize_Forw"        "SDJumpSize_Forw"
## [17] "Countrows"                "PercentageBackwardsJumps"
## [19] "MeanJumpSlope"            "SDJumpSlope"
## [21] "MeanStartPos_rel"         "sdStartPos_rel"

```

```
## [23] "MeanStartPos_edge"      "sdStartPos_edge"
## [25] "MeanEndPos_rel"        "sdEndPos_rel"
## [27] "Mean_n_events"         "sd_n_events"
## [29] "Mean_n_scroll_movements" "sd_n_scroll_movements"
```

Merging two tables and adding a few other relative variables using information from one of the generic tables.

```
# Merging two tables
descriptivesBig <- sum_session4 %>%
  group_by(session_number) %>%
  left_join(descriptives)
```

```
## Joining, by = "session_number"
```

```
# Relative jump count added
descriptivesBigger <- descriptivesBig %>%
  group_by(session_number) %>%
  mutate(RelCountJumps = totalTypingChars/Countrows)
```

```
# Detour added, it's a ratio (total jumpsize /characters typed)
descriptivesFinal <- descriptivesBigger %>%
  group_by(session_number) %>%
  mutate(Detour = TotalJumpsize/char_produced2, Author = "GB")
```

inspecting the output table

```
view(descriptivesFinal)
```

```
# and to obtain a first impression of the dataset
summary(descriptivesFinal)
```

```
## session_number MeanTypingChars sdTypingChars totalTypingChars
## Min. :3.00 Min. :11.00 Min. :10.61 Min. :154.0
## 1st Qu.:3.25 1st Qu.:11.01 1st Qu.:10.84 1st Qu.:162.5
## Median :3.50 Median :11.03 Median :11.07 Median :171.0
## Mean :3.50 Mean :11.03 Mean :11.07 Mean :171.0
## 3rd Qu.:3.75 3rd Qu.:11.04 3rd Qu.:11.30 3rd Qu.:179.5
## Max. :4.00 Max. :11.06 Max. :11.54 Max. :188.0
##
## MeanDurationTyping sdDurationTyping MeanTypingPositionRel sdTypingPositionRel
## Min. :2066 Min. :2109 Min. :0.7573 Min. :0.3145
## 1st Qu.:2193 1st Qu.:2281 1st Qu.:0.7573 1st Qu.:0.3145
## Median :2321 Median :2453 Median :0.7573 Median :0.3145
## Mean :2321 Mean :2453 Mean :0.7573 Mean :0.3145
## 3rd Qu.:2448 3rd Qu.:2626 3rd Qu.:0.7573 3rd Qu.:0.3145
## Max. :2575 Max. :2798 Max. :0.7573 Max. :0.3145
## NA's :1 NA's :1
## MeanTypingPositionEdge sdTypingPositionEdge Jump_time total_time
## Min. :27.64 Min. :34.91 Min. :3186 Min. : 93281
```

```

## 1st Qu.:29.32      1st Qu.:37.16      1st Qu.:3545      1st Qu.: 98289
## Median :31.00      Median :39.41      Median :3904      Median :103297
## Mean   :31.00      Mean   :39.41      Mean   :3904      Mean   :103297
## 3rd Qu.:32.68      3rd Qu.:41.66      3rd Qu.:4262      3rd Qu.:108305
## Max.   :34.35      Max.   :43.91      Max.   :4621      Max.   :113313
##
## total_time_seconds total_time_minutes total_charproduced char_produced2
## Min.   : 93.28      Min.   :1.555      Min.   :199.0      Min.   :41
## 1st Qu.: 98.29      1st Qu.:1.638      1st Qu.:209.2      1st Qu.:41
## Median :103.30      Median :1.722      Median :219.5      Median :41
## Mean   :103.30      Mean   :1.722      Mean   :219.5      Mean   :41
## 3rd Qu.:108.31      3rd Qu.:1.805      3rd Qu.:229.8      3rd Qu.:41
## Max.   :113.31      Max.   :1.889      Max.   :240.0      Max.   :41
##                                     NA's :1
## Perc_time_jumps StartPos_AfterWords_perc StartPos_Deletion_perc
## Min.   :2.812      Min.   : 0.0      Min.   : 0.0
## 1st Qu.:3.347      1st Qu.:12.5      1st Qu.:12.5
## Median :3.883      Median :25.0      Median :25.0
## Mean   :3.883      Mean   :25.0      Mean   :25.0
## 3rd Qu.:4.418      3rd Qu.:37.5      3rd Qu.:37.5
## Max.   :4.954      Max.   :50.0      Max.   :50.0
##
## StartPos_BeforeSentences_perc StartPos_AfterSentences_perc MeanDurationJumps
## Min.   : 0.0      Min.   : 0.0      Min.   :1593
## 1st Qu.:12.5      1st Qu.:12.5      1st Qu.:1772
## Median :25.0      Median :25.0      Median :1952
## Mean   :25.0      Mean   :25.0      Mean   :1952
## 3rd Qu.:37.5      3rd Qu.:37.5      3rd Qu.:2131
## Max.   :50.0      Max.   :50.0      Max.   :2310
##
## SD_DurationJumps MeanJumpsize_chars SDJumpsize_chars MeanJumpsize_rel
## Min.   :1570      Min.   : -36.0      Min.   : 9.899      Min.   :0.05893
## 1st Qu.:1670      1st Qu.: -28.5      1st Qu.:22.627      1st Qu.:0.14527
## Median :1769      Median : -21.0      Median :35.355      Median :0.23161
## Mean   :1769      Mean   : -21.0      Mean   :35.355      Mean   :0.23161
## 3rd Qu.:1868      3rd Qu.: -13.5      3rd Qu.:48.083      3rd Qu.:0.31794
## Max.   :1967      Max.   : -6.0      Max.   :60.811      Max.   :0.40428
##
## SDJumpsize_rel      MeanJumpsize_chars_Plus sdJumpsize_chars_Plus
## Min.   :0.06987      Min.   : 7      Min.   : 8.485
## 1st Qu.:0.17803      1st Qu.:16      1st Qu.:19.092
## Median :0.28619      Median :25      Median :29.698
## Mean   :0.28619      Mean   :25      Mean   :29.698
## 3rd Qu.:0.39435      3rd Qu.:34      3rd Qu.:40.305
## Max.   :0.50252      Max.   :43      Max.   :50.912
##
## logmeanJumpsize_chars_Plus TotalJumpsize logTotalJumpsize MeanJumpSize_BackW
## Min.   :1.946      Min.   :14      Min.   :2.639      Min.   : -79.0
## 1st Qu.:2.400      1st Qu.:32      1st Qu.:3.093      1st Qu.: -62.5
## Median :2.854      Median :50      Median :3.547      Median : -46.0
## Mean   :2.854      Mean   :50      Mean   :3.547      Mean   : -46.0
## 3rd Qu.:3.307      3rd Qu.:68      3rd Qu.:4.001      3rd Qu.: -29.5
## Max.   :3.761      Max.   :86      Max.   :4.454      Max.   : -13.0
##

```

```

## SDJumpSize_BackW MeanJumpSize_Forw SDJumpSize_Forw Countrows
## Min. : NA Min. :1.0 Min. : NA Min. :2
## 1st Qu.: NA 1st Qu.:2.5 1st Qu.: NA 1st Qu.:2
## Median : NA Median :4.0 Median : NA Median :2
## Mean :NaN Mean :4.0 Mean :NaN Mean :2
## 3rd Qu.: NA 3rd Qu.:5.5 3rd Qu.: NA 3rd Qu.:2
## Max. : NA Max. :7.0 Max. : NA Max. :2
## NA's :2 NA's :2
## PercentageBackwardsJumps MeanJumpSlope SDJumpSlope
## Min. :50 Min. : -0.008630 Min. :0.006581
## 1st Qu.:50 1st Qu.: -0.006398 1st Qu.:0.010049
## Median :50 Median : -0.004166 Median :0.013517
## Mean :50 Mean : -0.004166 Mean :0.013517
## 3rd Qu.:50 3rd Qu.: -0.001935 3rd Qu.:0.016986
## Max. :50 Max. : 0.000297 Max. :0.020454
##
## MeanStartPos_rel sdStartPos_rel MeanStartPos_edge sdStartPos_edge
## Min. :0.5863 Min. :0.3294 Min. :34.50 Min. :47.38
## 1st Qu.:0.6291 1st Qu.:0.3904 1st Qu.:36.75 1st Qu.:50.56
## Median :0.6719 Median :0.4514 Median :39.00 Median :53.74
## Mean :0.6719 Mean :0.4514 Mean :39.00 Mean :53.74
## 3rd Qu.:0.7146 3rd Qu.:0.5123 3rd Qu.:41.25 3rd Qu.:56.92
## Max. :0.7574 Max. :0.5733 Max. :43.50 Max. :60.10
##
## MeanEndPos_rel sdEndPos_rel Mean_n_events sd_n_events
## Min. :0.4010 Min. :0.2438 Min. :3.500 Min. :1.414
## 1st Qu.:0.4359 1st Qu.:0.3067 1st Qu.:4.375 1st Qu.:1.591
## Median :0.4708 Median :0.3695 Median :5.250 Median :1.768
## Mean :0.4708 Mean :0.3695 Mean :5.250 Mean :1.768
## 3rd Qu.:0.5057 3rd Qu.:0.4323 3rd Qu.:6.125 3rd Qu.:1.945
## Max. :0.5406 Max. :0.4952 Max. :7.000 Max. :2.121
##
## Mean_n_scroll_movements sd_n_scroll_movements RelCountJumps Detour
## Min. :0 Min. :0 Min. :77.00 Min. :0.3415
## 1st Qu.:0 1st Qu.:0 1st Qu.:81.25 1st Qu.:0.3415
## Median :0 Median :0 Median :85.50 Median :0.3415
## Mean :0 Mean :0 Mean :85.50 Mean :0.3415
## 3rd Qu.:0 3rd Qu.:0 3rd Qu.:89.75 3rd Qu.:0.3415
## Max. :0 Max. :0 Max. :94.00 Max. :0.3415
## NA's :1
## Author
## Length:2
## Class :character
## Mode :character
##
##
##
##

```

Final remarks

We hope to have shown the ingredients of our analysis and the steps we took to demarcate keystroke logging files into non-linear jumps, texts bursts, focus and deletion events, followed by the application of descriptive

statistics at the writing session-level.