Eye_tracking

May 3, 2017

1 Eye tracking - Quantitative analysis

This part focuses on quantitative analysis on eye tracking data.

1.1 Import and global variables

```
In [1]: import codecs
                                               # to read '.txt' files files
       import csv
                                               # to read and write '.csv' files
                                               # to plot graphs
       import matplotlib.pyplot as plt
       import numpy as np
       import os
                                               # to deal with os calls
       import pandas as pd
       import seaborn as sns
       import collections
       from scipy.signal import argrelextrema, argrelmax
       from pylab import savefig
In [2]: participant = 0
                                  # current participant ID
       recording_name = 1
                                # name of the recording
                                # recording duration
       recording_duration = 2
       time_column = 3
                                # time indication
                                # x-position of the gaze point
       gaze_x_column = 4
       gaze_y_column = 5
                                # y-position of the gaze point
       pupil_diam_left = 6
                                # diameter of left pupil over time
       pupil_diam_right = 7
                              # diameter of right pupil over time
       mt_column = 8
                                # movement type column
                                # movement duration column
       md column = 9
       mi_column = 10
                                # movement index column
       event_column = 11  # Event type
```

1.2 Helper functions

```
csvread.next()
    for row in csvread:
        data.append(row)
    # Filtering the data where Eye Tracking is not working
    return [line for line in data if not line[mt_column] == "EyesNotFound"]
# Returns True if string can be converted as float
def is_number(str):
   try:
        float(str)
        return True
    except:
        return False
# Returns list of users contained in data
def get_users(data):
    users = list(set([l[participant] for l in data[1:]]))
    users.sort()
    return users
```

1.3 Analysis of means and standard deviations of fixations

1.3.1 Main functions

```
In [4]: # Defining functions for titles in quantitative analysis
        def fixation_key(user):
            return "all " + str(float(user)) + ' - Fixation time (ms)'
        def saccade_key(user):
            return "all " + str(float(user)) + ' - Saccade time (ms)'
        def gaze_x_key(user):
            return "all " + str(float(user)) + ' - Gaze X (pixels)'
        def gaze_y_key(user):
            return "all " + str(float(user)) + ' - Gaze Y (pixels)'
        keys = {'fixation': fixation_key, 'saccade': saccade_key, 'gaze_x': gaze_x_key, 'gaze_y'
In [5]: def quantitative(raw_data, user_id, type):
            raw_data = raw_data[1:]
            def analyse_records(data, user_id):
                # Fixation and saccade times
                pd1 = pd.DataFrame(list(set([float(line[md_column]) for line in data if line[mt_
                                   columns=[str(user_id) + ' - Fixation time (ms)']).describe(in)
                pd2 = pd.DataFrame(list(set([float(line[md_column]) for line in data if line[mt_
```

```
columns=[str(user_id) + ' - Saccade time (ms)']).describe(ind
                # X and Y coordinates of gaze points
                pd3 = pd.DataFrame([float(line[gaze_x_column]) for line in data if is_number(line
                                   columns=[str(user_id) + ' - Gaze X (pixels)']).describe(inclu
                pd4 = pd.DataFrame([float(line[gaze_y_column]) for line in data if is_number(line
                                   columns=[str(user_id) + ' - Gaze Y (pixels)']).describe(incluser)
                return pd.concat([pd1, pd2, pd3, pd4], axis=1)
            if type == "all":
                data = [line for line in raw_data if float(line[participant]) == user_id]
                return analyse_records(data, "all " + str(user_id))
            if type == "each":
                records = list(set([line[recording_name] for line in raw_data if float(line[part
                records.sort()
                dfs = pd.DataFrame()
                for rec in records:
                    data = [line for line in raw_data if line[recording_name] == rec]
                    dfs = pd.concat([dfs, analyse_records(data, str(user_id) + " - " + str(rec))
                return dfs
In [10]: def describe_quantitative(d, users, save=True):
             plt.clf()
             # Showing: fixation, saccade, gaze X, gaze Y on average
             f, axarr = plt.subplots(4, sharex=False)
             users_int = [int(user) for user in users]
             axarr[0].set_xlim([min(users_int)-1, max(users_int)+1])
             axarr[0].set_title("Average fixation time")
             axarr[0].set_ylabel("time (ms)")
             axarr[1].set_title("Average saccade time")
             axarr[1].set_ylabel("time (ms)")
             axarr[2].set_title("Average Gaze X position")
             axarr[2].set_ylim([0, 1920])
             axarr[2].set_ylabel("position (pixel)")
             axarr[3].set_title("Average Gaze Y position (not reversed)")
             axarr[3].set_ylim([0, 1080])
             axarr[3].set_ylabel("position (pixel)")
             def plot_describe(data_frame, pos):
                 data_array = data_frame.as_matrix()
                 for index in range(0, len(data_array[0])):
                     axarr[pos].errorbar(int(users[index]), data_array[1][index], data_array[2][
                                         color='red', lw=3)
                     axarr[pos].errorbar(int(users[index]), data_array[1][index],
                                   [[data_array[1][index] - data_array[3][index]], [data_array[7]
```

```
fmt='.k', color='gray', lw=1)
def plot_one_value(value, pos):
    data_frame = pd.DataFrame()
    for user in users:
        result = pd.concat([quantitative(d, float(user), "all"), quantitative(d, fl
        # Saving the results into CSV
        if save is True:
            # base = os.path.dirname(os.path.abspath(__file__))
            # location = base + "\\..\\results\\" + str(user) + "_results.csv"
            location = str(user) + "_results.csv"
            result.to_csv(location, sep=';', encoding='utf-16')
        key = keys[value](user)
        data_frame = pd.concat([data_frame, pd.DataFrame(result, columns=[key])], a
    plot_describe(data_frame, pos)
    return data_frame
plot_one_value('fixation', 0)
plot_one_value('saccade', 1)
plot_one_value('gaze_x', 2)
plot_one_value('gaze_y', 3)
plt.tight_layout()
plt.xlabel('User ID')
f.set_size_inches(20, 18, forward=True)
# Save pictures
savefig('foo.png', bbox_inches='tight')
savefig('foo.pdf', bbox_inches='tight')
# Show the curves
plt.show()
```

1.3.2 Results - application on data

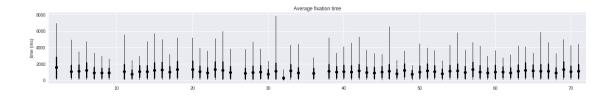
This section aims at reading and performing the analysis of the data captured by the glasses.

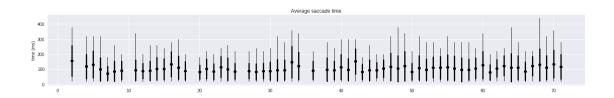
Data are exported from the glasses as '.tsv' file and converted using Excel as 'unicode .txt' file in order to enable a cross-platform encoding. The data of all the users (containing all together 2,155 045,rows) are exported in 3 different files due to the size limitation of Excel which cannot handle more than 1,048,576 rows.

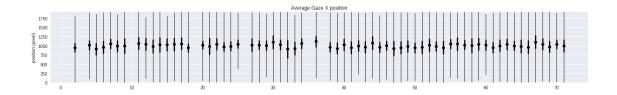
After loading the data, some IDs are deleted manually from the list of users to consider since their related participant didn't fulfill all the experiment requirements.

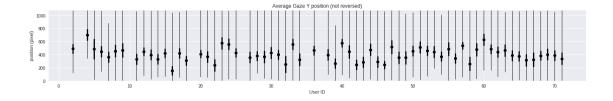
```
In [7]: # Reading the eye tracking data from 3 different files
```

```
if False:
            tmp = load("/media/sf_EyeTracking/data/short_1_35_unicode.txt")
            tmp.extend(load("/media/sf_EyeTracking/data/short_36_51_unicode.txt"))
            tmp.extend(load("/media/sf_EyeTracking/data/short_52_71_unicode.txt"))
        print "Data contains " + str(len(tmp)) + " rows."
Data contains 2155045 rows.
In [8]: # Getting the list of users
        users_str = get_users(tmp)
        users_float = [int(user) for user in users_str]
        users_float.sort()
        print "Users IDs from the data: " + str(users_float)
        # Some user ID are known as being missing.
        print "Missing IDs in the data: " + str([x for x in range(1,71) if x not in users_float]
        # Some other users are not considered.
        not_users = [35, 37]
        print "IDs not considered: " + str(not_users) +'\n'
        users_float = [user for user in users_float if user not in not_users]
        users_str = [str(user) for user in users_float]
        # List of users for the experiment
        print str(len(users_float)) + " participants to analyze: " + str(users_float)
Users IDs from the data: [2, 4, 5, 6, 7, 8, 9, 11, 12, 13, 14, 15, 16, 17, 18, 20, 21, 22, 23, 2
Missing IDs in the data: [1, 3, 10, 19, 26]
IDs not considered: [35, 37]
64 participants to analyze: [2, 4, 5, 6, 7, 8, 9, 11, 12, 13, 14, 15, 16, 17, 18, 20, 21, 22, 23
In [11]: # Perform the analysis on the list of users ID selected.
         describe_quantitative(tmp, users_str, save=False) # If True, csv files with data of
<matplotlib.figure.Figure at 0x7f91dc0d2750>
```









1.4 Analysing gaze points over X and Y coordinates

1.4.1 X coordinates: working over value distribution

```
In [14]: # Selecting the values for one user and one recording.
         x_pos = [line[gaze_x_column] for line in tmp
                      if line[participant] == '42'
                      and line[recording_name] == 'Recording172']
         x_pos = [int(line) for line in x_pos if line != '']
         x_{pos} = np.asarray(x_{pos})
         print "List of x values: " + str(x_pos)
List of x values: [ 979 981 982 ..., 1276 1278 1280]
In [15]: # Plotting the distribution based on SeaBorn
         x_pos = pd.Series(x_pos, name="Gaze X distribution")
         x_dist = sns.distplot(x_pos)
         plt.show()
     0.0035
     0.0030
     0.0025
     0.0020
     0.0015
     0.0010
     0.0005
     0.0000
                250
                         500
                                   750
                                           1000
                                                    1250
                                                             1500
                                                                      1750
                                                                                2000
```

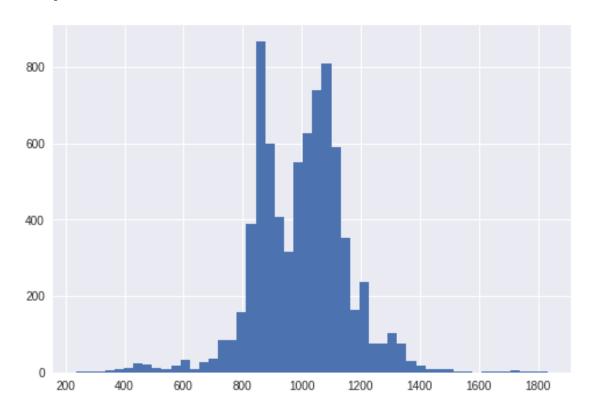
This represents where people look at when driving. They mostly look at their lane, but also check the opposite one, which is more on the left.

Gaze X distribution

It is then interesting to have a look at the distance between the two peaks (at first counted here, in number of bins - arbitrary value, stable over measurement). The ratio between the first and second extrema is then computed.

Some curves might create other extremum that are not related to the driving lanes, this can then be corrected manually by choosing the right extremum.

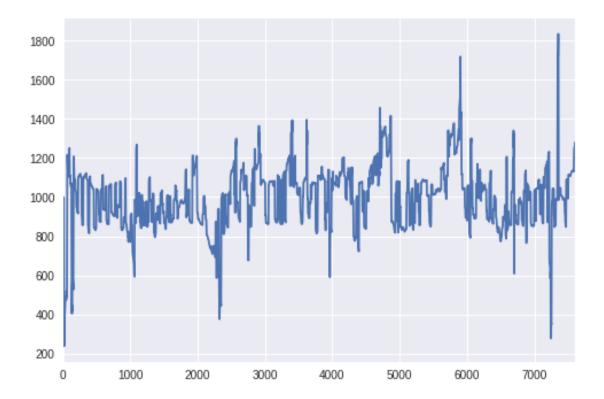
```
In [16]: # Reconstructing the values based on NumPy
     a = x_pos.hist(bins=50)
     plt.show()
```



```
bin_len.append(bins[i + 1] - bins[i])
             return bin_len
         bin_len = bin_length(bins)
         bin_length = bin_len[0]
In [18]: n_{values} = n[argrelmax(n)[0]]
         print "Local optimums values: " + str(n_values)
         # Return maximum
         fst_max = np.partition(n_values.flatten(), -1)[-1]
         fst_index = np.argmax(n)
         print "First extremum: " + str(fst_index) + ", " + str(fst_max)
         # Return second value
         scd_max = np.partition(n_values.flatten(), -2)[-2]
         scd_index = np.nonzero(n==scd_max)[0][0]
         print "Second extremum: " + str(scd_index) + ", " + str(scd_max)
         diff_max = fst_max - scd_max
         diff_index = fst_index - scd_index
         print "Difference between optimums positions: " + str(diff_index) + " bins"
         print "Difference between optimums position: " + str(diff_index * bin_length) + " pxls"
         print "Difference between optimums value: " + str(diff_max) + " pxls"
         print "Ratio between optimums: " + str(fst_max/scd_max)
Local optimums values: [
                            3.
                                 25.
                                      34. 866. 808. 236. 104.
                                                                       10.
                                                                              6.]
First extremum: 19, 866.0
Second extremum: 26, 808.0
Difference between optimums positions: -7 bins
Difference between optimums position: -223.44 pxls
Difference between optimums value: 58.0 pxls
Ratio between optimums: 1.07178217822
   Results on participant 42: - recording 169: 5, 456.0 (tr2) - recording 170: 5, 296.0 (tr3) - recording
171: 7, 573.0 (tr1) - recording 172: -7, 58.0 1.071 (tr2)
   Results on participant 43: - recording 173: 5, 419.0 1.570 (tr3) - recording 175: -2, 149.0, 1.198
```

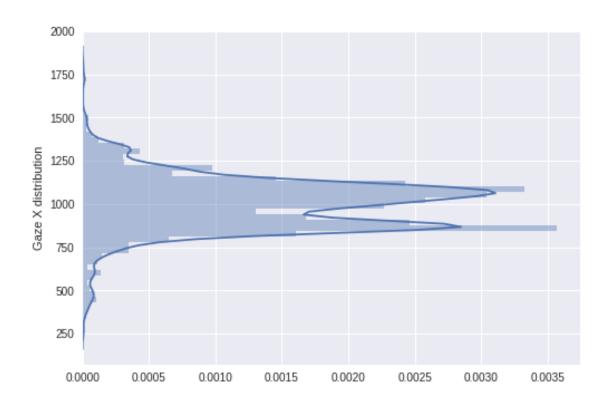
1.4.2 X coordinates: working over time distribution

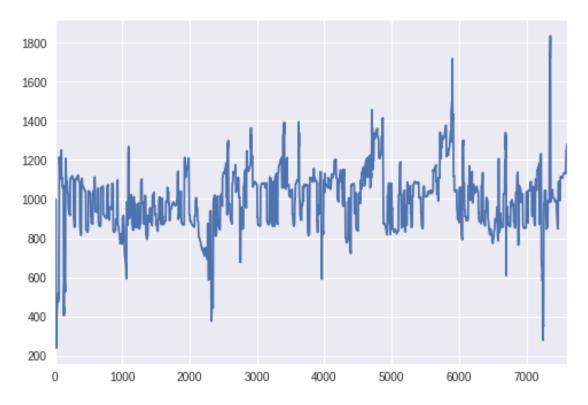
(tr1) - recording 176: -4, 262.0, 1.287 (tr2) - recording 177: 5, 262.0, 1.420 (tr3)



First one might represent looking in the side mirror when starting. Other ones might also represent crossing or checking on the sides.

1.4.3 Y coordinates: working over value distribution

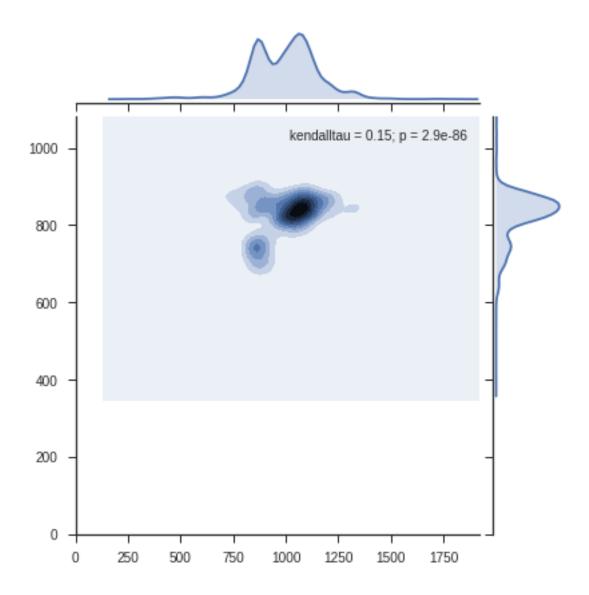




Peaks show the number of time the driver is checking the upper mirror. Down shows the checking of commands (biggest ones) or of the speed indication (less big ones).

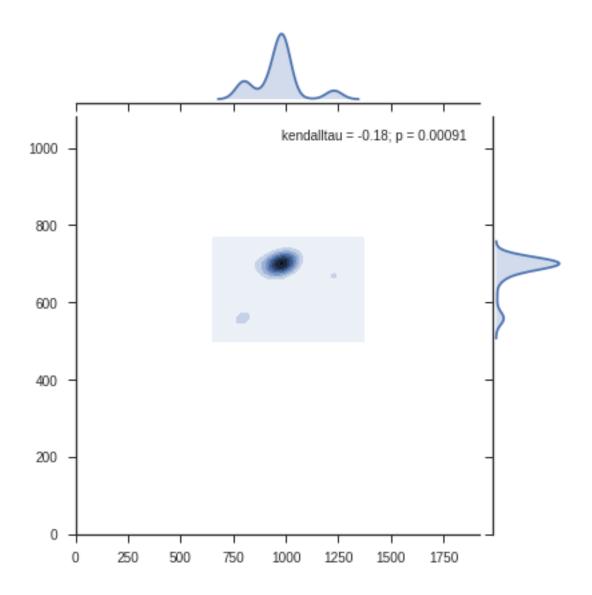
1.5 Working with heatmap around events

```
In [28]: import matplotlib.cm as cm
         import matplotlib.mlab as ml
         from scipy.stats import kendalltau
In [29]: def get_time(usr, rec):
             coord = [int(line[time_column]) for line in tmp
                      if line[participant] == usr and line[recording_name] == rec]
             return min(coord), max(coord)
In [30]: def ms2mins(millis):
             millis = int(millis)
             seconds=(millis/1000)%60
             seconds = int(seconds)
             minutes=(millis/(1000*60))%60
             minutes = int(minutes)
             return ("%dm%ds" % (minutes, seconds))
         def mins2ms(minutes, seconds, millis=0):
             return minutes * 60 * 1000 + seconds * 1000 + millis
In [31]: print ms2mins(mins2ms(2,32,0))
2m32s
In [32]: def heatmap(usr, rec, start=-1, end=-1):
             if start == -1 and end == -1:
                 start, end = get_time(usr, rec)
             # Selecting the values for one user and one recording.
             coord = [[line[gaze_x_column], line[gaze_y_column]] for line in tmp
                      if line[participant] == usr and line[recording_name] == rec
                      and int(line[time_column]) < end</pre>
                      and int(line[time_column]) > start]
             # Converting to integers
             def toInt(1):
                 return [int(i) for i in l if is_number(i)]
             coord = [toInt(1) for 1 in coord if len(toInt(1)) == 2]
             # Extracting the data
```

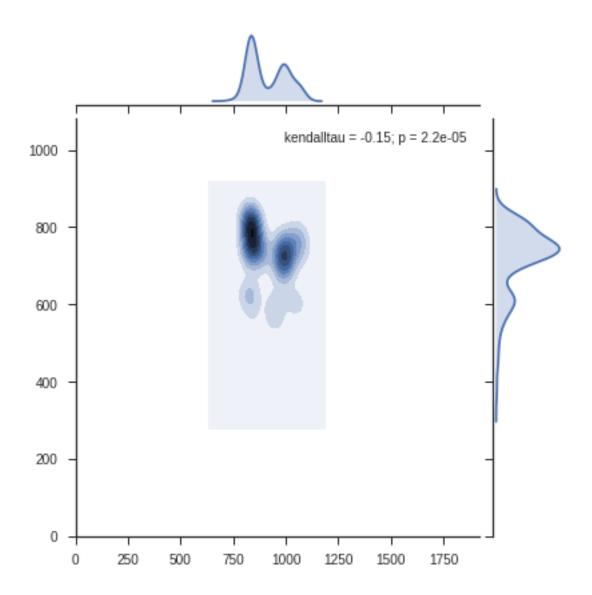


Plot from 0 m 30 s to 0 m 33 s

<matplotlib.figure.Figure at 0x7f91d6159d90>



<matplotlib.figure.Figure at 0x7f91d5d69190>



1.6 Checking for missing events in recordings

```
count = 0
             for 1 in data:
                 if l[event_column] == log_event or l[event_column] == mis_event:
                      count += 1
                 if l[event_column] == del_event:
                      count -= 1
             return count
         # Print a summary of the number of logged event per recording
         summary = []
         for r in recordings:
                 summary.append([r,events_nb([line for line in tmp if line[recording_name] == r]
In [99]: # Creating a dictionary
         D = \{1[0]: 1[1] \text{ for } 1 \text{ in summary}\}
         D = collections.OrderedDict(sorted(D.items()))
         # Printing the results
         plt.figure(figsize=(60,10))
         plt.bar(range(len(D)), D.values(), align='center')
         plt.xticks(range(len(D)), D.keys(), rotation='vertical')
         plt.tight_layout()
         savefig('nb_events.pdf', bbox_inches='tight')
         plt.show()
```