Paper analyses

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General preparations

Load libraries

```
library(mousetrap)
library(ggplot2)
library(dplyr)
```

Custom ggplot2 theme

```
theme_set(theme_classic()+
  theme(
    axis.line = element_line(colour = "black"),
    axis.ticks = element_line(colour = "black"),
    axis.text = element_text(colour = "black"),
    panel.border = element_rect(colour = "black", fill=NA)
))
```

Data preparation

Read in merged raw data

```
# Read in raw data from csv.gz file
# (commented out as the raw data are now included in the mousetrap R package)
# raw_data <- read.csv("merged_data/raw_data.csv.gz",stringsAsFactor=FALSE)
# Instead, get raw data directly from data stored in mousetrap package
raw_data <- KH2017_raw</pre>
```

Filter trials

```
# Check that each subject has at least one typical and atypical trial left
table(with(raw_data,table(subject_nr,Condition))>0)
##
## TRUE
## 120
```

Import mouse-tracking data

```
mt_data <- mt_import_mousetrap(raw_data)</pre>
```

Trajectory preprocessing

```
mt_data <- mt_remap_symmetric(mt_data)
mt_data <- mt_align_start(mt_data)
mt_data <- mt_measures(mt_data)
mt_data <- mt_time_normalize(mt_data)</pre>
```

Analyses

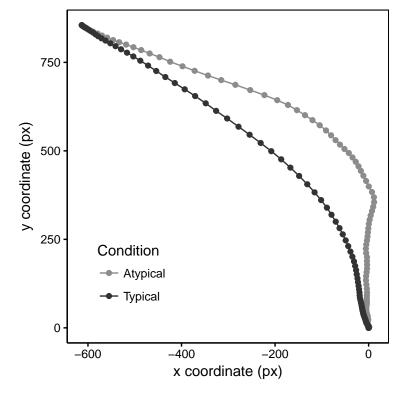
Logging resolution

```
mt_check_resolution(mt_data, desired = 10)
## $summary
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                Max.
##
     1.000 10.000 10.000
                              9.981 10.000 40.000
##
## $sd
## [1] 0.430673
##
## $frequencies
## log_diffs
               2
##
        1
                      3
                              4
                                     5
                                            6
                                                    7
                                                           8
                                                                  9
                                                                         10
       93
             133
                     85
                            128
                                    92
                                          146
                                                   63
                                                                 85 212191
##
                                                         124
##
       11
              12
                     13
                             14
                                    15
                                           16
                                                   17
                                                          20
                                                                 23
                                                                         25
##
      182
              10
                      9
                             10
                                     4
                                            3
                                                    3
                                                           1
                                                                  1
                                                                          1
       27
              29
                      32
                             33
##
                                    34
                                            36
                                                   40
##
               2
                              3
                                     2
                                             2
                                                    1
##
## $relative_frequencies
## log_diffs
               2
                      3
                              4
                                            6
                                                    7
                                                           8
                                                                         10
##
                                     5
                                                                   9
        1
## 0.0004 0.0006 0.0004 0.0006 0.0004 0.0007 0.0003 0.0006 0.0004 0.9944
                     13
##
       11
              12
                             14
                                    15
                                            16
                                                   17
                                                          20
## 0.0009 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000
##
                     32
                             33
                                    34
## 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000
##
```

```
## $frequencies_desired
## log_diffs_class
## smaller desired greater
## 949 212191 239
##
## $relative_frequencies_desired
## log_diffs_class
## smaller desired greater
## 0.0044 0.9944 0.0011
```

Plot average time-normalized tajectories

```
mt_plot_aggregate(mt_data, use = "tn_trajectories", points=TRUE,
    x = "xpos", y = "ypos", color = "Condition", subject_id = "subject_nr")+
    theme(legend.position=c(.2,.2))+
    xlab("x coordinate (px)") + ylab("y coordinate (px)")+
    scale_color_grey(start = .55, end = .20)
```



```
# ggsave("figures/Figure5.png",width=8.5, height=10,unit="cm",dpi=1200)
# ggsave("figures/Figure5.pdf",width=8.5, height=10,unit="cm")
# ggsave("figures/Figure5.eps",width=8.5, height=10,unit="cm")
```

Comparison of measures based on time-normalized vs. raw trajectories

```
mt_data <- mt_measures(mt_data,use="tn_trajectories",save_as="tn_measures")
diag(cor(mt_data$measures[,-1],mt_data$tn_measures[,-1]))</pre>
```

```
ypos_max
                                                           ypos_min
##
          xpos_max
                          {\tt xpos\_min}
                                                          0.9985444
##
         0.9999950
                         0.9999948
                                          0.9999992
                                          MD above
##
               MAD
                         MAD time
                                                     MD above time
         0.9998875
                         0.9990594
                                          0.9999922
                                                          0.9978766
##
##
          MD below
                     MD_below_time
                                                 ΑD
                                                                 AUC
##
         0.9998770
                                          0.9999912
                                                          0.9999688
                         0.9718700
##
        xpos_flips
                        ypos_flips xpos_reversals ypos_reversals
##
         0.9977138
                         0.9897775
                                          1.0000000
                                                           1.0000000
##
                RT initiation_time
                                          idle_time
##
         1.0000000
                         0.9997353
                                          0.9929374
```

Comparison of MAD aggregated per participant

```
agg_mad <- mt_aggregate_per_subject(mt_data, subject_id = "subject_nr",</pre>
  use_variables = "MAD", use2_variables = "Condition")
agg_mad %>%
  group_by(Condition) %>%
  select(MAD,Condition) %>%
  summarize_all(.funs = c("mean","sd"))
## # A tibble: 2 x 3
##
     Condition
                   mean
                               sd
##
        <fctr>
                  <dbl>
                            <dbl>
## 1 Atypical 343.7954 218.6368
      Typical 172.2093 110.8397
t.test(MAD~Condition, data=agg_mad, paired=TRUE)
##
##
   Paired t-test
##
## data: MAD by Condition
## t = 6.731, df = 59, p-value = 7.706e-09
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 120.5772 222.5949
## sample estimates:
## mean of the differences
                  171.5861
# Cohen's dz
mad_diff <- with(agg_mad,MAD[Condition=="Atypical"]-MAD[Condition=="Typical"])</pre>
mean(mad_diff)/sd(mad_diff)
## [1] 0.8689738
```

Bimdality analyses

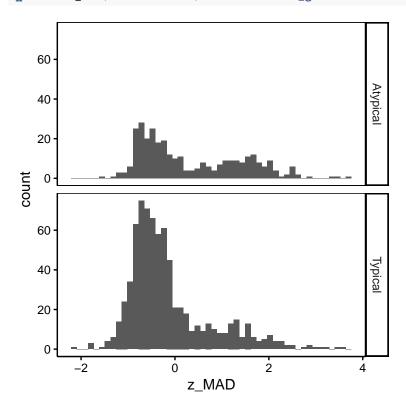
```
# Standardize MAD per participant
mt_data <- mt_standardize(mt_data,
   use_variables = "MAD", within = "subject_nr",prefix = "z_")</pre>
```

```
# Calculate bimodality coefficient
mt_check_bimodality(mt_data,use_variables = "z_MAD",
    grouping_variables = "Condition", methods = "BC")

## $BC
## Condition    z_MAD
## 1 Atypical 0.5934582
## 2 Typical 0.6078283

# Merge trial level data (needed for distribution qplot with facets)
results <- merge(mt_data$data,mt_data$measures,by="mt_id")

# Distribution
qplot(x=z_MAD,data=results,bins=50)+facet_grid(Condition~.)</pre>
```



Repeat analyses without large MAD values

```
# Exclude trials with abs(z_MAD)>1.5
table(abs(mt_data$measures$z_MAD)<=1.5)/nrow(mt_data$measures)

##
## FALSE TRUE
## 0.1165414 0.8834586

mt_data_sub <- mt_subset(mt_data,abs(z_MAD)<=1.5,check="measures")

# Check that each subject has at least on typical and atypical trial left
table(with(mt_data_sub$data,table(subject_nr,Condition))>0)
```

```
##
## TRUE
## 120
# Comparison of MAD aggregated per participant
agg_mad_sub <- mt_aggregate_per_subject(mt_data_sub, subject_id = "subject_nr",
  use_variables = "MAD", use2_variables = "Condition")
agg_mad_sub %>%
 group_by(Condition) %>%
  select(MAD,Condition) %>%
summarize_all(.funs = c("mean", "sd"))
## # A tibble: 2 x 3
##
    Condition
                 mean
                              sd
##
                  <dbl>
        <fctr>
                           <dbl>
## 1 Atypical 200.7050 209.2332
       Typical 121.7434 117.4904
t.test(MAD~Condition, data=agg_mad_sub, paired=TRUE)
##
## Paired t-test
##
## data: MAD by Condition
## t = 4.1599, df = 59, p-value = 0.0001047
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
   40.97963 116.94341
## sample estimates:
## mean of the differences
##
                 78.96152
```