

Experiment 2

General preparations

Load libraries

```
library(mousetrap)
library(ggplot2)
library(dplyr)
library(tidyr)
library(afex)
library(MBESS)
library(ordinal)
```

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)
  ))
```

Custom functions

```
# Function to compute confidence interval for partial eta-squared
get_partial_etas <- function(anova_table, conf.level=.90){
  partial_etas <- sapply(row.names(anova_table),function(i){
    F <- anova_table[i,"F"]
    df1 <- anova_table[i,"num Df"]
    df2 <- anova_table[i,"den Df"]
    ci <- conf.limits.ncf(F.value=F,conf.level=conf.level,df.1=df1,df.2=df2)
    return(
      c(pes=((F*df1)/(F*df1+df2)),
        lower=ci$Lower.Limit/(ci$Lower.Limit+df1+df2+1),
        upper=ci$Upper.Limit/(ci$Upper.Limit+df1+df2+1)))
  })
  return(t(partial_etas))
}
```

Data import

```
raw_data <- read.csv("../data/exp2.csv")
raw_data$Typicality <- factor(raw_data$Condition,levels=c("Typical","Atypical"))
raw_data$group <- factor(raw_data$group,levels=c("default","slow"))
```

Correctness

Percent of correct trials per condition

```
with(raw_data, table(group, correct)/c(table(group)))
```

```
##           correct
## group           0           1
## default 0.06512043 0.93487957
## slow    0.05530776 0.94469224
```

Chi-squared test

```
chisq.test(with(raw_data, table(group, correct)), correct = FALSE)
```

```
##
## Pearson's Chi-squared test
##
## data: with(raw_data, table(group, correct))
## X-squared = 0.95372, df = 1, p-value = 0.3288
```

Generalized linear mixed model

```
contrasts(raw_data$group) <- c(0.5, -0.5)
summary(glmer(correct ~ (1 | subject_nr) + group, family = "binomial", data = raw_data))
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: correct ~ (1 | subject_nr) + group
## Data: raw_data
##
##      AIC      BIC   logLik deviance df.resid
## 1008.6   1025.8   -501.3   1002.6     2239
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.5190  0.1804  0.2213  0.2337  0.5278
##
## Random effects:
## Groups      Name                Variance Std.Dev.
## subject_nr (Intercept) 0.573      0.757
## Number of obs: 2242, groups: subject_nr, 118
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    2.9990     0.1401  21.409  <2e-16 ***
## group1         -0.1648     0.2358  -0.699    0.485
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## Correlation of Fixed Effects:  
##      (Intr)  
## group1 -0.031
```

Exclude incorrect trials

```
raw_data <- subset(raw_data, correct==1)
```

Trajectory preprocessing

```
mt_data <- mt_import_mousetrap(raw_data)  
mt_data <- mt_remap_symmetric(mt_data)  
  
# Start position descriptives (xpos)  
mean(mt_data$trajectories[,1,"xpos"])  
  
## [1] -0.2477456  
  
sd(mt_data$trajectories[,1,"xpos"])  
  
## [1] 45.62261  
  
# Start position descriptives (ypos)  
mean(mt_data$trajectories[,1,"ypos"])+1050/2  
  
## [1] 103.6232  
  
sd(mt_data$trajectories[,1,"ypos"])  
  
## [1] 32.45008  
  
mt_data <- mt_align_start(mt_data, start=c(0,0))  
mt_data <- mt_derivatives(mt_data)  
mt_data <- mt_measures(mt_data)  
mt_data <- mt_time_normalize(mt_data)
```

Manipulation check: maximum velocity and acceleration

Aggregate data per participant and condition

```
agg_vel_acc <- mt_aggregate_per_subject(mt_data, use_variables=c("vel_max", "acc_max"),  
  use2_variables="group", subject_id="subject_nr")
```

Descriptives

```
mean_vel_acc <- agg_vel_acc %>%  
  group_by(group) %>%  
  summarize(  
    N = n(),
```

```

M_vel = mean(vel_max),
SD_vel = sd(vel_max),
M_acc = mean(acc_max),
SD_acc = sd(acc_max)
) %>%
as.data.frame()

print(mean_vel_acc, digits=2)

```

```

##      group  N M_vel SD_vel M_acc SD_acc
## 1 default 59 10.0   2.2  0.54  0.119
## 2   slow 59  4.3   1.1  0.22  0.056

```

Compare maximum velocity

```

# Between groups t-test
vel_t <- t.test(vel_max~group, data=agg_vel_acc, paired=FALSE,var.equal=TRUE)
vel_t

##
## Two Sample t-test
##
## data:  vel_max by group
## t = 18.086, df = 116, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  5.117120 6.375717
## sample estimates:
## mean in group default      mean in group slow
##           10.004925           4.258506

# Cohen's d including 95 % CI
ci.smd(ncp = vel_t$statistic[[1]],n.1 = mean_vel_acc$N[1],n.2 = mean_vel_acc$N[2], conf.level = .95)

## $Lower.Conf.Limit.smd
## [1] 2.766925
##
## $smd
## [1] 3.329905
##
## $Upper.Conf.Limit.smd
## [1] 3.886533

```

Compare maximum acceleration

```

# Between groups t-test
acc_t <- t.test(acc_max~group, data=agg_vel_acc, paired=FALSE,var.equal=TRUE)
acc_t

##
## Two Sample t-test
##

```

```
## data: acc_max by group
## t = 18.667, df = 116, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.2854992 0.3532744
## sample estimates:
## mean in group default      mean in group slow
##      0.5420214              0.2226346

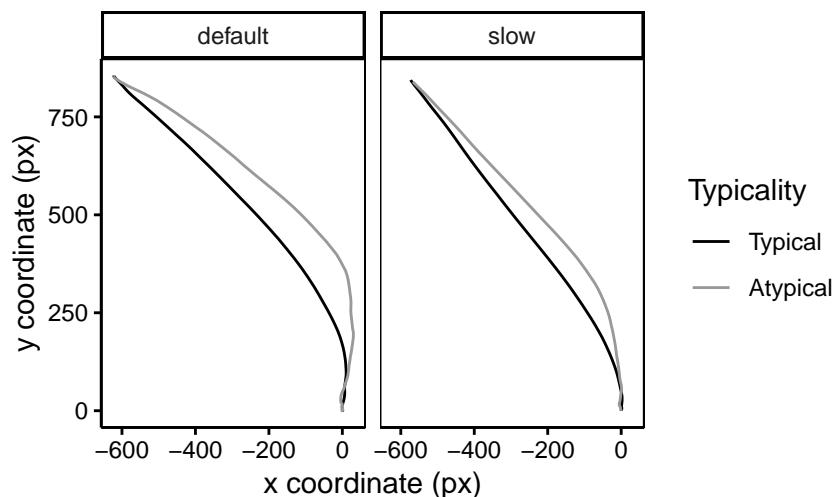
# Cohen's d including 95 % CI
ci.smd(ncp = acc_t$statistic[[1]], n.1 = mean_vel_acc$N[1], n.2 = mean_vel_acc$N[2], conf.level = .95)

## $Lower.Conf.Limit.smd
## [1] 2.86335
##
## $smd
## [1] 3.436905
##
## $Upper.Conf.Limit.smd
## [1] 4.00412
```

Aggregate trajectory curvature

Average time-normalized trajectories

```
mt_plot_aggregate(mt_data, use = "tn_trajectories", facet_col = "group",
  x = "xpos", y = "ypos", color = "Typicality", subject_id = "subject_nr") +
  xlab("x coordinate (px)") + ylab("y coordinate (px)") +
  scale_color_manual(values = c("black", "grey60"))
```



Comparison of MAD aggregated per participant

Aggregate data per participant and condition

```
agg_mad <- mt_aggregate_per_subject(mt_data, subject_id = "subject_nr",  
  use_variables = "MAD", use2_variables = c("Typicality", "group"))
```

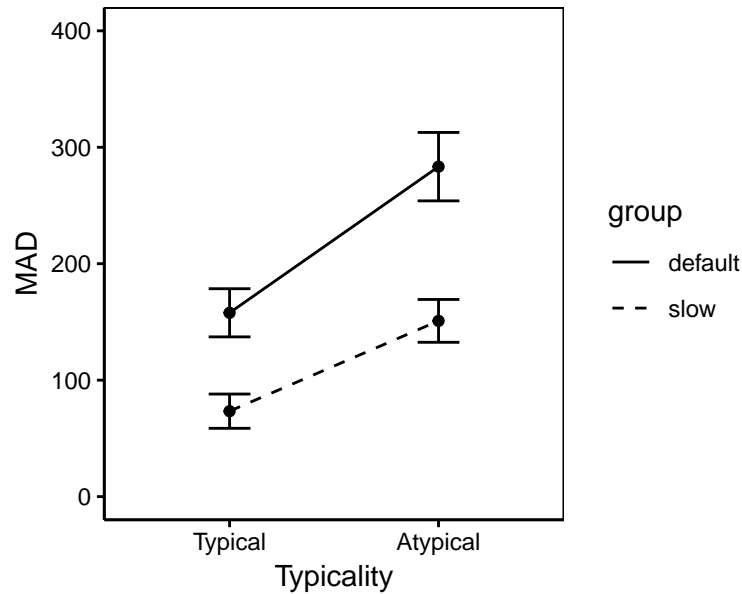
Descriptives and paired t-tests

```
mad_table <- agg_mad %>%  
  group_by(group) %>%  
  select(MAD, group, Typicality) %>%  
  summarize(  
    N = length(MAD[Typicality=="Typical"]),  
    M_t = mean(MAD[Typicality=="Typical"]),  
    SD_t = sd(MAD[Typicality=="Typical"]),  
    M_a = mean(MAD[Typicality=="Atypical"]),  
    SD_a = sd(MAD[Typicality=="Atypical"]),  
    t = t.test(MAD[Typicality=="Atypical"], MAD[Typicality=="Typical"], paired=TRUE)$statistic,  
    p = t.test(MAD[Typicality=="Atypical"], MAD[Typicality=="Typical"], paired=TRUE)$p.value,  
    d = (M_a - M_t) / sd(MAD[Typicality=="Atypical"] - MAD[Typicality=="Typical"])  
  )  
  
mad_table %>%  
  as.data.frame() %>%  
  print(digits=3)
```

```
##      group  N   M_t SD_t M_a SD_a    t      p      d  
## 1 default 59 157.8 159 283  226 5.49 9.32e-07 0.715  
## 2   slow 59  73.4 113 151  141 4.52 3.05e-05 0.589
```

Figure

```
ggplot(agg_mad, aes(x=Typicality, y=MAD, linetype=group, group=group)) +  
  geom_line(stat="summary", fun.y="mean") +  
  geom_point(stat="summary", fun.y="mean") +  
  geom_errorbar(stat="summary", fun.data="mean_se", width=.2, linetype=1) +  
  scale_linetype_manual(values=c(1, 2)) +  
  coord_cartesian(ylim=c(0, 400))
```



ANOVA

```
anova_mad <- aov_ez(data=agg_mad, dv = "MAD", between = "group", within = "Typicality",
  id = "subject_nr")
```

```
## Contrasts set to contr.sum for the following variables: group
```

```
nice(anova_mad, es = c("pes", "ges"))
```

```
## Anova Table (Type 3 tests)
```

```
##
```

```
## Response: MAD
```

##	Effect	df	MSE	F	ges	pes	p.value
## 1	group	1, 116	42401.22	16.37 ***	.10	.12	<.0001
## 2	Typicality	1, 116	12045.29	50.49 ***	.09	.30	<.0001
## 3	group:Typicality	1, 116	12045.29	2.82 +	.005	.02	.10

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '+' 0.1 ' ' 1
```

```
# 90 % confidence interval for partial eta-squared
```

```
round(get_partial_etas(anova_mad$anova_table, conf.level=.90), 2)
```

##		pes	lower	upper
## group		0.12	0.04	0.22
## Typicality		0.30	0.19	0.40
## group:Typicality		0.02	NA	0.09

Distribution of trajectory shapes

Bimodality coefficient

```
# Standardize MAD per participant
mt_data <- mt_standardize(mt_data, use_variables = "MAD", within = "subject_nr")

# Calculate bimodality coefficient
mt_check_bimodality(mt_data, use_variables = "z_MAD",
  grouping_variables = c("group", "Typicality"), methods = "BC")

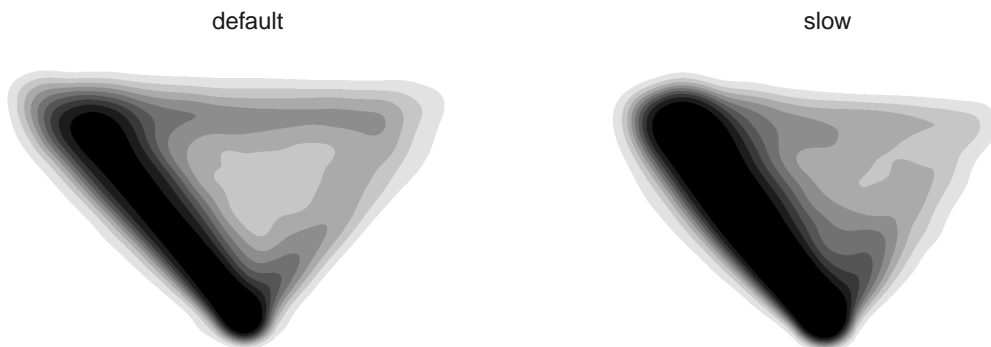
## $BC
##      group Typicality      z_MAD
## 1 default    Typical 0.5582705
## 2 default   Atypical 0.5762162
## 3  slow     Typical 0.5727287
## 4  slow     Atypical 0.5926053
```

Smoothed heatmaps

```
heatmap_smoothed <- mt_heatmap_ggplot(mt_data,
  xres = 1000,
  smooth_radius = 20,
  n_shades = 10,
  mean_image = 0.2,
  colors=c("white", "black"),
  facet_col="group")

## spatializing trajectories
## calculate image
## smooth image
## enhance image by 4.8
## spatializing trajectories
## calculate image
## smooth image
## enhance image by 9.1

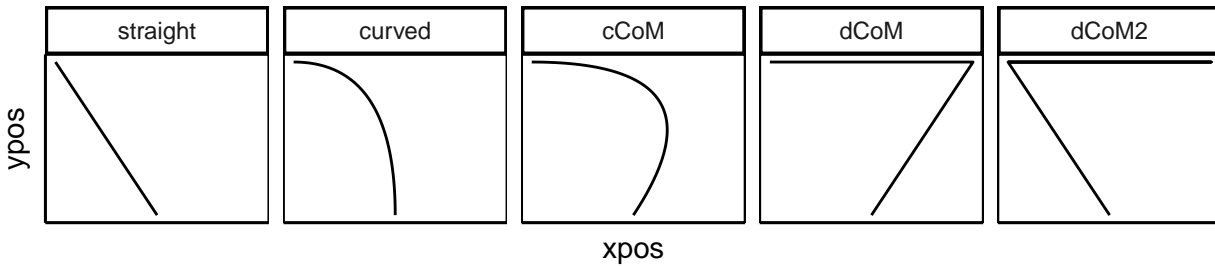
heatmap_smoothed+
  theme(strip.background = element_rect(colour = NA))
```



Prototype classification

Plot prototypes

```
mt_plot(mt_prototypes, facet_col="mt_id", only_ggplot = TRUE)+  
  geom_path()+  
  facet_grid(cols = vars(factor(mt_id, levels=rownames(mt_prototypes))))+  
  theme(axis.text=ggplot2::element_blank(), axis.ticks=ggplot2::element_blank())
```



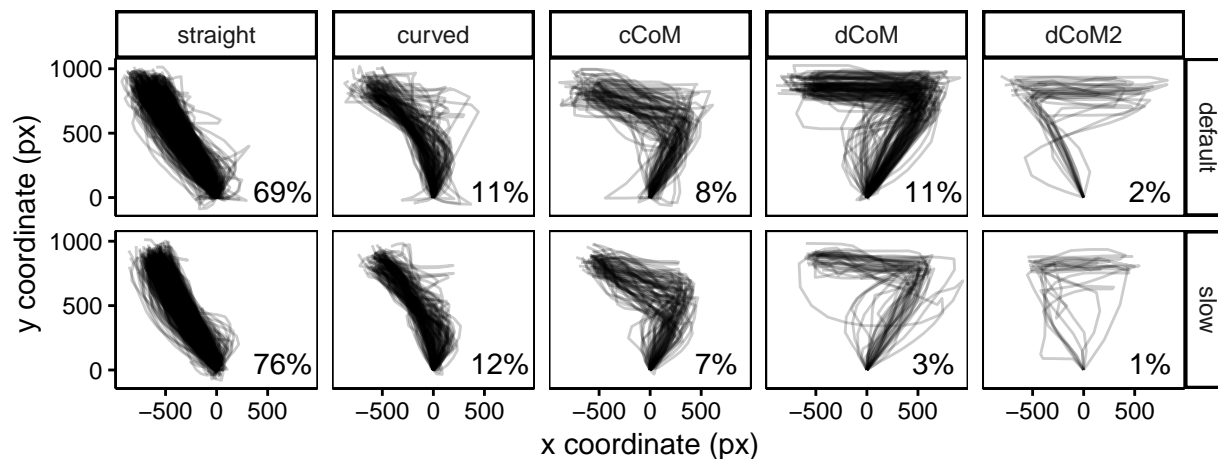
Map trajectories onto prototypes

```
mt_data <- mt_spatialize(mt_data)  
mt_data <- mt_map(mt_data, prototypes = mt_prototypes,  
  save_as = "measures", grouping_variables = "group")  
mt_data$data$prototype_label <- mt_data$measures$prototype_label
```

Classified trajectories per group

Relative frequencies

```
prototype_percentages <- mt_data$data %>%  
  group_by(group, prototype_label) %>%  
  summarise(n=n()) %>%  
  mutate(Percent=paste(round(100*n/sum(n)), "%", sep=""))  
  
mt_plot(mt_data, use = "sp_trajectories",  
  x = "xpos", y = "ypos", facet_col = "prototype_label", facet_row="group", alpha=.2)+  
  xlab("x coordinate (px)") + ylab("y coordinate (px)") +  
  geom_text(data=prototype_percentages, aes(label=Percent), x=650, y=50)+  
  scale_y_continuous(breaks=c(0, 500, 1000))+  
  coord_cartesian(xlim=c(-900, 900))
```



Chi-squared test

```
chisq.test(with(mt_data$data, table(group, prototype_label)))
```

```
##
## Pearson's Chi-squared test
##
## data: with(mt_data$data, table(group, prototype_label))
## X-squared = 49.656, df = 4, p-value = 4.261e-10
```

Classified trajectories per group X typicality condition

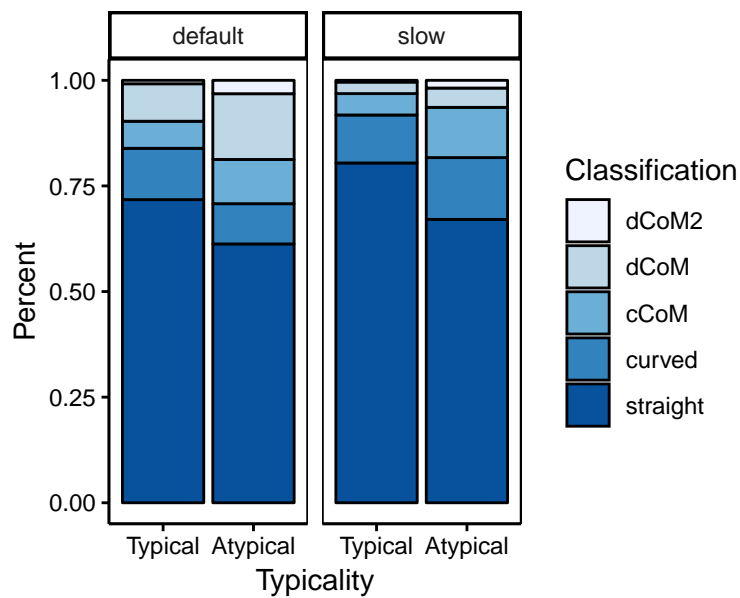
Relative frequencies

```
rel_freq_agg <- mt_data$data %>%
  group_by(group, Typicality, prototype_label) %>%
  summarise(n=n()) %>%
  mutate(Percent=n/sum(n))

spread(rel_freq_agg[, -4], "prototype_label", "Percent", fill = 0) %>%
  as.data.frame() %>%
  print(digits=2)
```

```
##      group Typicality straight curved  cCoM dCoM dCoM2
## 1 default   Typical    0.72  0.121 0.064 0.089 0.0082
## 2 default   Atypical    0.61  0.095 0.105 0.156 0.0317
## 3  slow     Typical    0.80  0.114 0.051 0.027 0.0041
## 4  slow     Atypical    0.67  0.146 0.119 0.046 0.0183
```

```
ggplot(rel_freq_agg, aes(x=Typicality, y=Percent, fill=forcats::fct_rev(prototype_label))) +
  geom_bar(stat="identity", color="black") +
  scale_fill_brewer(type="seq", name="Classification") +
  facet_grid(.~group)
```



Ordinal mixed regression

```
contrasts(mt_data$data$Typicality) <- c(-0.5,0.5)
contrasts(mt_data$data$group) <- c(0.5,-0.5)
summary(clmm(prototype_label~Typicality*group+(1|subject_nr),data=mt_data$data))
```

```
## Cumulative Link Mixed Model fitted with the Laplace approximation
##
## formula: prototype_label ~ Typicality * group + (1 | subject_nr)
## data:    mt_data$data
##
## link threshold nobis logLik    AIC      niter    max.grad cond.H
## logit flexible  2107 -1818.10 3652.20 584(2340) 1.43e-03 6.8e+01
##
## Random effects:
## Groups      Name          Variance Std.Dev.
## subject_nr (Intercept) 1.021      1.01
## Number of groups:  subject_nr 118
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## Typicality1      0.77413   0.10859   7.129 1.01e-12 ***
## group1           0.61284   0.21926   2.795 0.00519 **
## Typicality1:group1 -0.05205   0.21571  -0.241 0.80932
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Threshold coefficients:
##              Estimate Std. Error z value
## straight|curved    1.0353    0.1121   9.237
## curved|cCoM        1.8798    0.1193  15.752
## cCoM|dCoM          2.6979    0.1320  20.439
## dCoM|dCoM2         4.8252    0.2297  21.004
```