

# Basic Research Methods: eye tracking tutorial

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# Hands-on exercise

## Task:

- ▶ a passive-viewing task
- ▶ Three conditions: positive, negative and neutral pictures from IAPS (International Affective Picture System)
- ▶ 3000 ms presentation, 500 ms ITI
- ▶ 30 x each condition (no repeating items)
- ▶ time: ~5 min

## Sample:

- ▶ you!

# Procedure

1. Positioning
2. Calibration (+ re-calibration)
3. Data collection
4. Data pre-processing
5. Data visualisation
6. Data interpretation

## Hands-on time!

Now we collect data from volunteers. Meanwhile, form groups of 5-10 people and prepare the experimental design and predictions by considering the following points:

### 1. **Experimental design:**

- ▶ conditions (3: pos, neg, neu)
- ▶ no of repetitions of stimuli
- ▶ sample size + statistical power
- ▶ type of statistical analysis

### 2. **Research questions / hypotheses**

- ▶ backward reasoning: what could be the research question asked in this study?
- ▶ hypotheses and predictions: based on what you already know about eye tracking, pupillometry, and processing of emotional and neutral images in humans, propose your hypotheses and predictions for our results.

## Raw data

```
# show the first 4 rows  
head(rd,4)
```

```
##                timestamp                time  
## 1                MSG 2020-01-07 10:17:43.26 1061923  
## 2 2020-01-07 10:17:43.023                1061923567  
## 3 2020-01-07 10:17:43.040                1061923584  
## 4 2020-01-07 10:17:43.056                1061923600  
##  
##                state      rawx      rawy      avgx      avgy  
## 1 start_recording      NA      NA      NA      NA  
## 2                7 1063.299 719.7043 1036.940 699.1066 26  
## 3                7 1040.149 710.0520 1037.394 698.9050 25  
## 4                7 1051.085 682.4178 1038.015 699.0016 25  
##  
##      Lrawy      Lavgx      Lavgy      Lpsize      Lpupilx      Lpupily  
## 1      NA      NA      NA      NA      NA      NA  
## 2 757.5413 1050.300 710.9240 24.6516 0.3618 0.5723 108  
## 3 739.2700 1050.397 712.3792 24.8139 0.3617 0.5719 102  
## 4 715.2422 1050.301 712.5284 24.4666 0.3614 0.5715 105
```

## Structured raw data

```
# show selected columns  
data <- rd[,c(1,3:5,13,20)]  
# ...and the first 5 rows  
head(data,5)
```

```
##           timestamp           fix      rawx      r  
## 1 2020-01-07 10:17:43.259 start_recording      NA  
## 2 2020-01-07 10:17:43.023           True 1063.299 719.7  
## 3 2020-01-07 10:17:43.039           True 1040.149 710.0  
## 4 2020-01-07 10:17:43.055           True 1051.085 682.4  
## 5 2020-01-07 10:17:43.072           True 1130.159 803.0  
##      Rpsize  
## 1      NA  
## 2 27.8472  
## 3 26.6744  
## 4 27.4394  
## 5 31.6633
```

# Pre-processing

Steps of pre-processing. . .

## Pre-processed data

##	code	t_ms	sample	cond	trial_no	segmentName	size_f
## 1	1111	4.361	24.52612	NEG	1	baseline	-0.0616
## 2	1111	4.362	24.52464	NEG	1	baseline	-0.0633
## 3	1111	4.363	24.52319	NEG	1	baseline	-0.0646
## 4	1111	4.364	24.52176	NEG	1	baseline	-0.0660
## 5	1111	4.365	24.52036	NEG	1	baseline	-0.0674
## 6	1111	4.366	24.51898	NEG	1	baseline	-0.0688
## 7	1111	4.367	24.51763	NEG	1	baseline	-0.0701
## 8	1111	4.368	24.51632	NEG	1	baseline	-0.0714
## 9	1111	4.369	24.51503	NEG	1	baseline	-0.0727
## 10	1111	4.370	24.51377	NEG	1	baseline	-0.0740
## 11	1111	4.371	24.51255	NEG	1	baseline	-0.0752
## 12	1111	4.372	24.51137	NEG	1	baseline	-0.0764
## 13	1111	4.373	24.51022	NEG	1	baseline	-0.0775
## 14	1111	4.374	24.50911	NEG	1	baseline	-0.0786
## 15	1111	4.375	24.50803	NEG	1	baseline	-0.0797
## 16	1111	4.376	24.50700	NEG	1	baseline	-0.0808
## 17	1111	4.377	24.50600	NEG	1	baseline	-0.0818



## Basic descriptives

Sample size:

```
## [1] 1
```

Mean, max, and min (relative to baseline):

```
## [1] 0.53
```

```
## [1] 4.35
```

```
## [1] -3.15
```

*Question:* What do negative values mean?

## Aggregate data

*Question:* What is data aggregation?

For plotting, we need the time course, but for statistics we have to aggregate the data over conditions and a time window. Here we will use 1 - 3 secs.

*Question:* Why won't we use the data  $< 1$  sec?

*# For plotting:*

```
data_P_av_trials <- aggregate(list(data$size_fdb),by=list(c(
colnames(data_P_av_trials) <- c("cond","t","trial","size")
```

```
data_P_av <- aggregate(list(data$size_fdb),by=list(data$con
colnames(data_P_av) <- c("cond","t","size")
```

*# For stats:*

```
data2 <- data[data$t_fdb >= 1 & data$t_fdb <= 3,]
data_av_trials <- aggregate(list(data2$size_fdb),by=list(da
colnames(data_av_trials) <- c("cond","trial","size")
```

## Plotting the data - visual inspection

Let's plot the averaged data over participants for each condition (the grey bands are the 95% confidence intervals):

```
## Warning: package 'ggplot2' was built under R version 3.5
```

```
## `geom_smooth()` using method = 'gam' and formula 'y ~ s
```

