

Humpback Whales and Ship Noise

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1 Data and Descriptive Statistics

```
# import data (this dataset is unfortunately not public)
openxlsx::read.xlsx("../Data/Humpback_Whales_Data.xlsx") -> dat_whale

# first a quick look at the missing values in the data
sapply(dat_whale, \(x) sum(is.na(x))) |> knitr::kable(col.names = "NAs")
```

	NAs
Individuum	0
Treatment	0
Szenario	0
ruhezeit	0
speed	5
Atem	0

```
# harmonize names
colnames(dat_whale) <- tolower(colnames(dat_whale))

# to numeric
lapply(dat_whale[, c("ruhezeit", "speed", "atem")], as.numeric) -> dat_whale[, c("ruhezeit", "speed", "atem")]

# to factor
lapply(dat_whale[, !(colnames(dat_whale) %in% c("ruhezeit", "speed", "atem"))],
       as.factor) -> dat_whale[, !(colnames(dat_whale) %in% c("ruhezeit", "speed", "atem"))]

# relevel
factor(dat_whale[, "szenario"],
       levels = c("Before", "During", "After")) -> dat_whale[, "szenario"]
factor(dat_whale[, "treatment"], c("Control", "Medium", "High")) -> dat_whale[, "treatment"]

# add log
within(dat_whale,{
  logspeed <- log(speed)
  atem[atem == 0] <- 0.001
  ruhezeit[ruhezeit == 0] <- 0.001
  logatem <- log(atem)
  logruhezeit <- log(ruhezeit)
  sqrtatem <- sqrt(atem)
  sqrtspeed <- sqrt(speed)
  sqrtruhezeit <- sqrt(ruhezeit * 100)
}) -> dat_whale

# first split into different intensities
dat_whale_intens <- split(dat_whale, dat_whale[, "treatment"])

# build formulas
formulae <- paste(c("ruhezeit", "speed", "atem"), "~", "szenario")

# max and min for plot y-axis
sapply(c(min, max), \(x){
```

```

sapply(dat_whale[, c("ruhezeit", "speed", "atem")], \(y) x(y, na.rm = TRUE))

}) -> ylims

# over szenarios
invis.Map(\(y, nom, lims){

  # safe for presentation
  # pdf(paste0("../Presentation/", nom, ".pdf"))

  # align
  par(mfrow = c(3, 1), mar = c(2, 4, 4, 2) + 0.1)

  # over treatment
  invis.Map(\(x, nom){

    # boxplots
    boxplot(as.formula(y), data = x,
            col = c("cornflowerblue", "deepskyblue4", "darkblue"),
            ylim = c(lims[1], lims[2]))

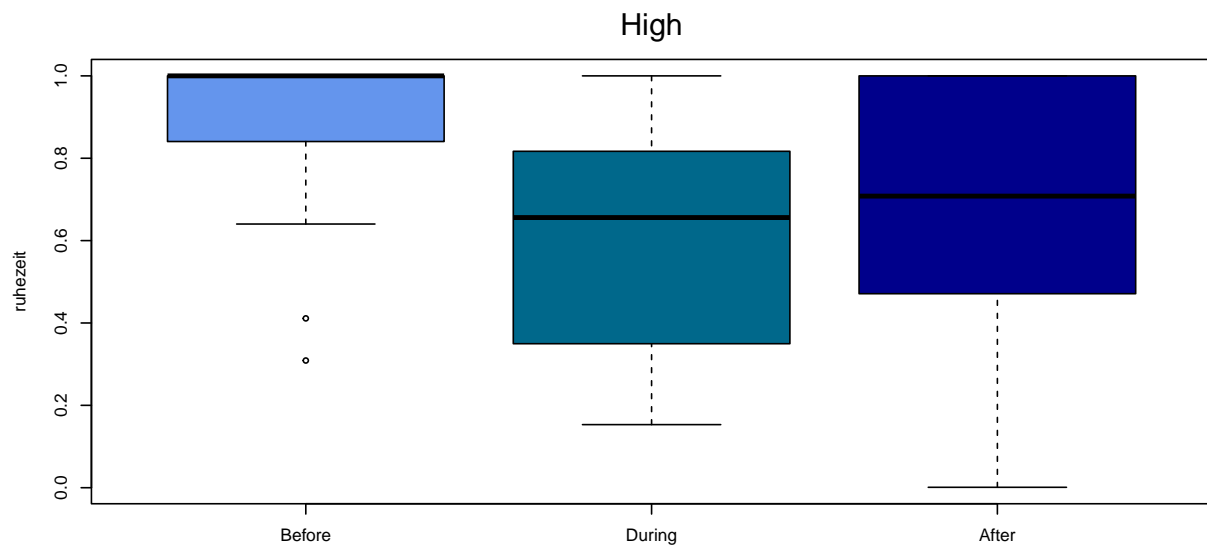
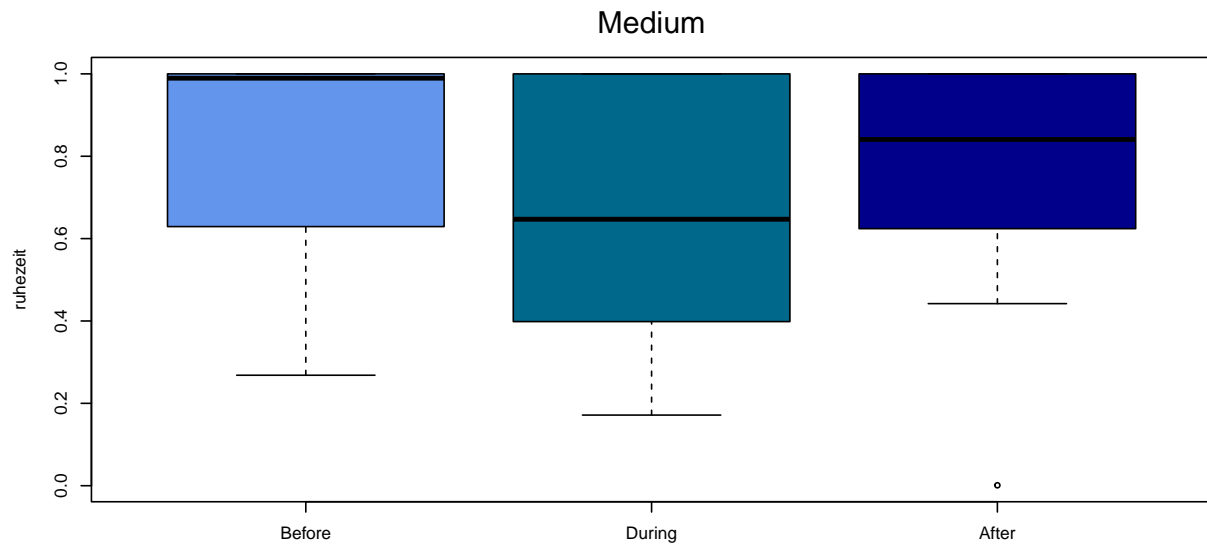
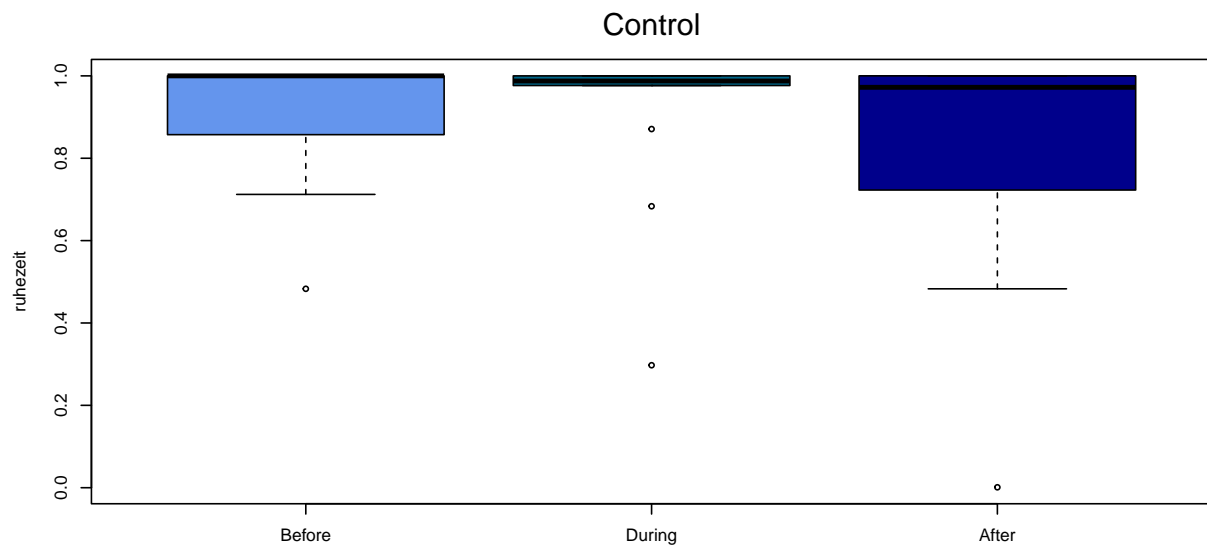
    # add label
    mtext(nom, side = 3, line = 1, cex = 1.2)

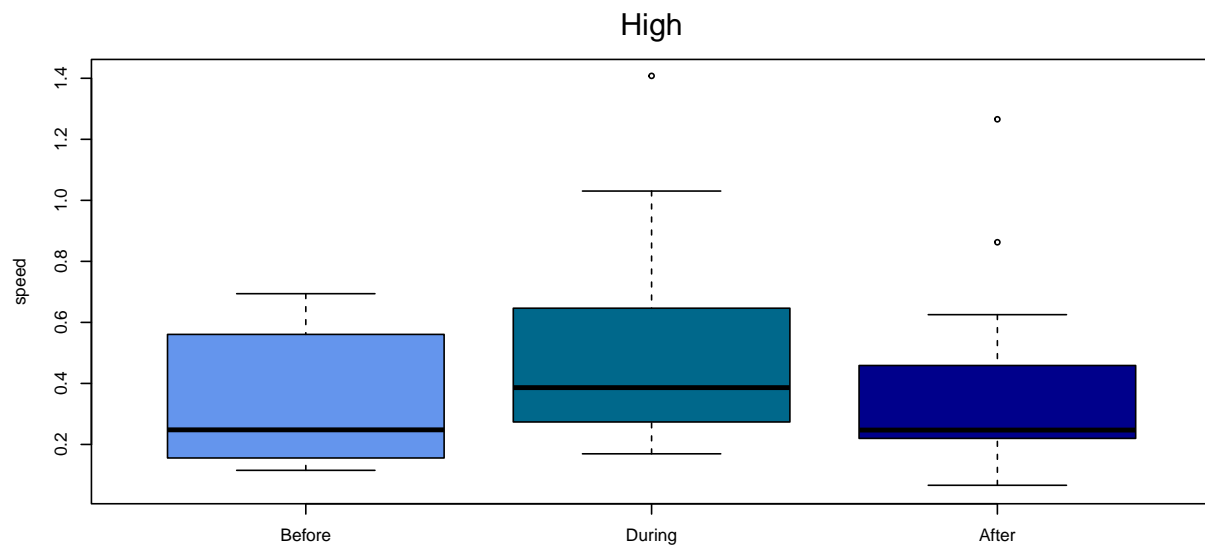
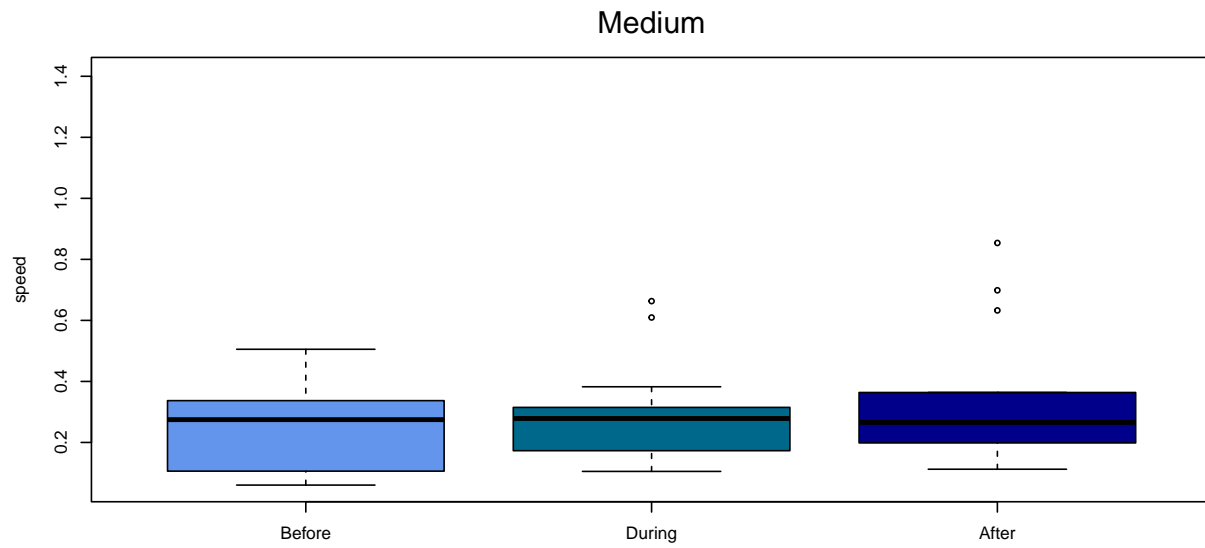
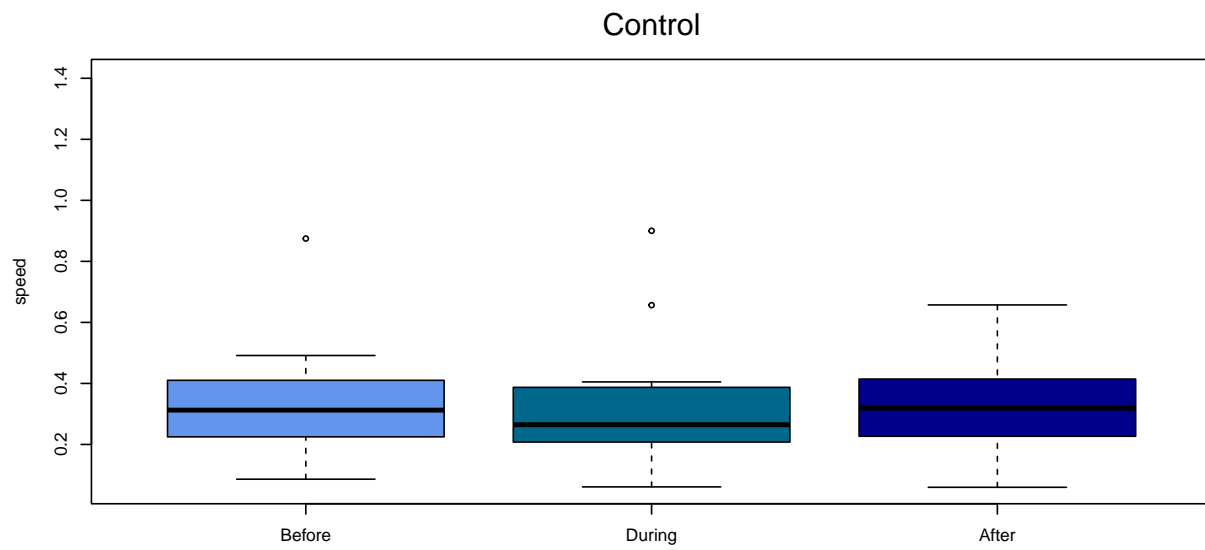
  }, dat_whale_intens, names(dat_whale_intens))

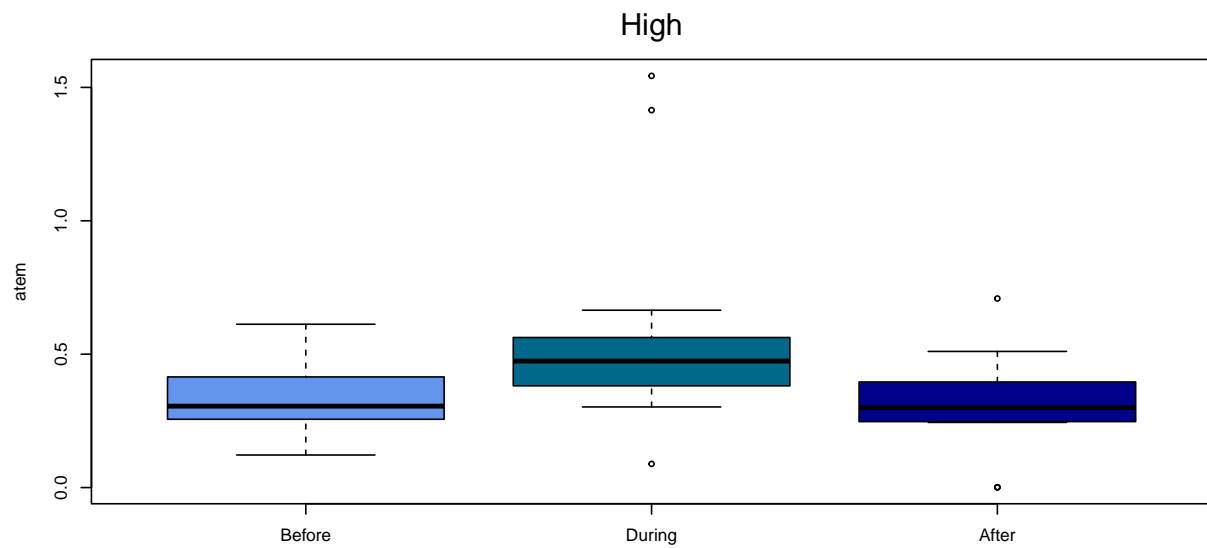
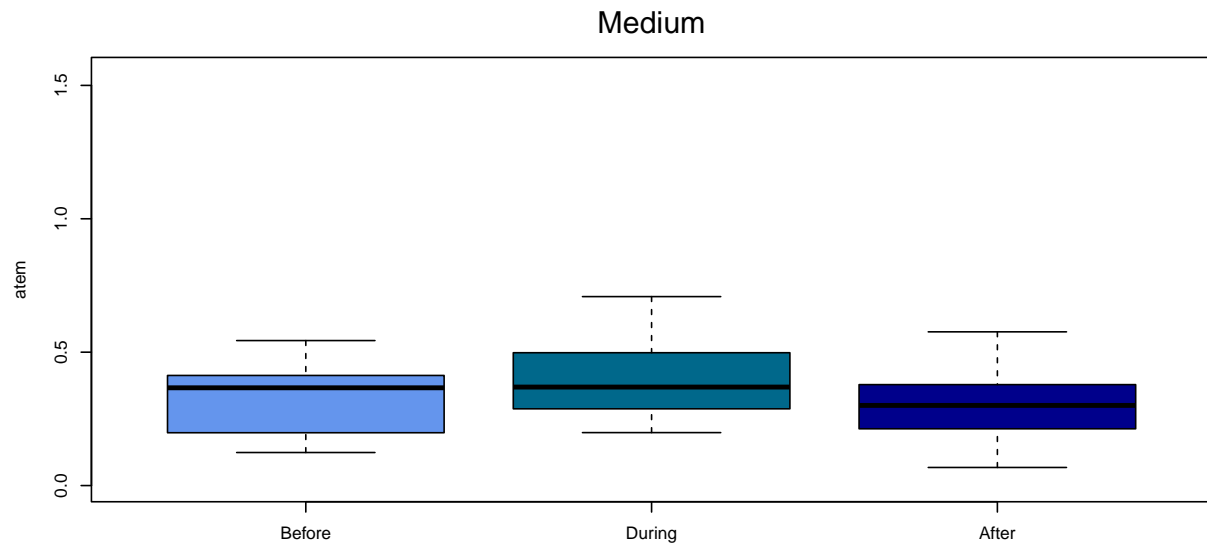
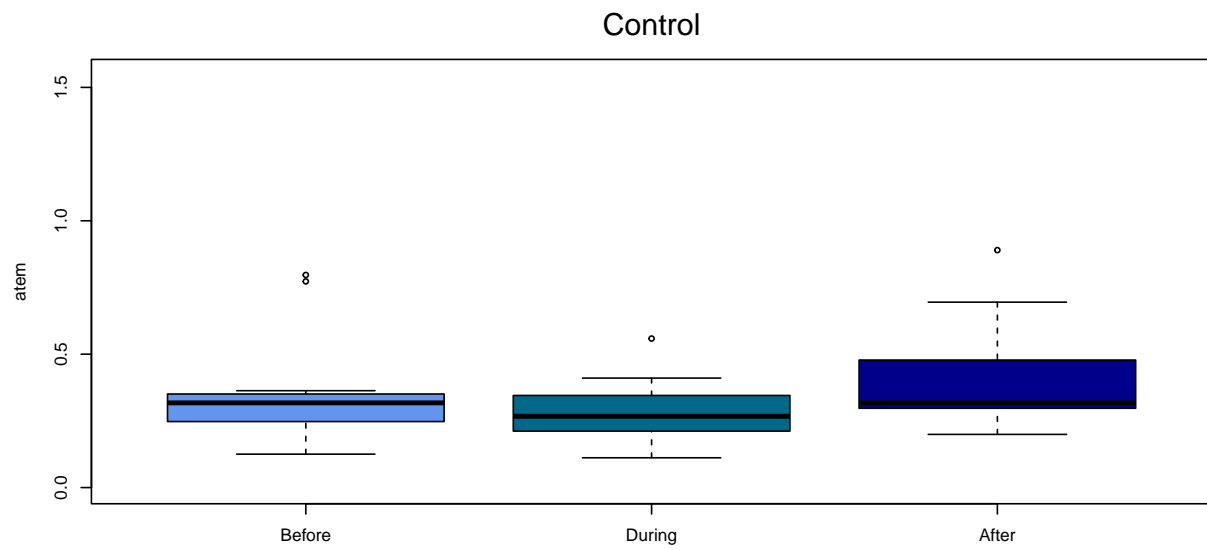
  # close graph. device
  # dev.off()

}, formulae, c("resting", "speed", "respatory"), ylims |> t() |> as.data.frame())

```





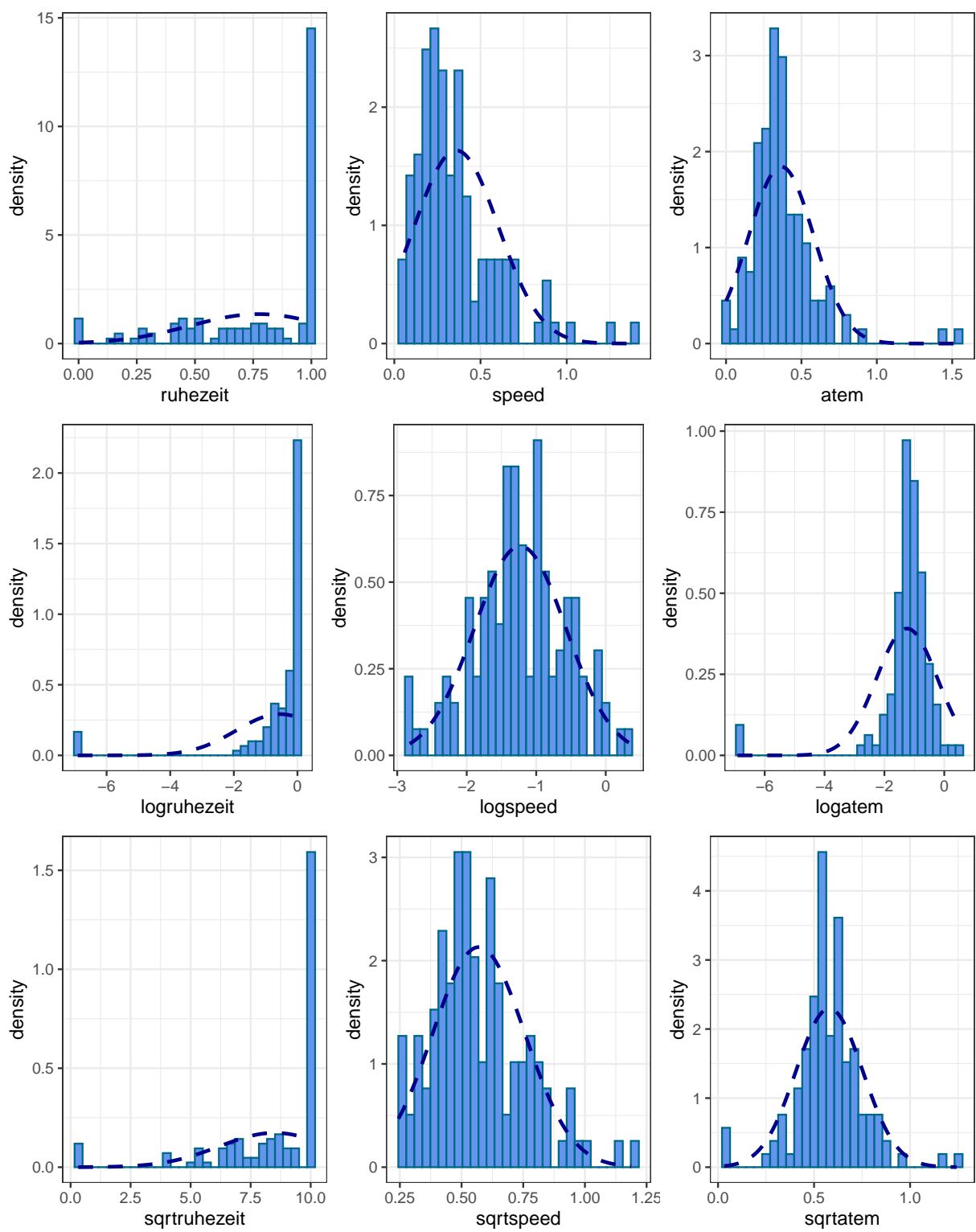


2 Distribution of Covariates

```
# vars
nom <- c("ruhezeit", "speed", "atem")
nom <- c(nom, paste0("log", nom), paste0("sqrt", nom))

# loop to generate plots
lapply(nom, \(x) Dens_norm_plot(y = x)) -> plots

# display (remove useless bin width messages)
print((plots[[1]] + plots[[2]] + plots[[3]]) /
      (plots[[4]] + plots[[5]] + plots[[6]]) /
      (plots[[7]] + plots[[8]] + plots[[9]])) |>
  suppressWarnings() |>
  suppressMessages()
```



3 Model

```
# formulas
formulae_lmm <- c("speed ~ I(treatment) * I(szenario)",
                  "atem ~ I(treatment) * I(szenario)",
                  "logspeed ~ I(treatment) * I(szenario)",
                  "sqrtatem ~ I(treatment) * I(szenario)")

formulae_glmm <- "ruhezeit ~ I(treatment) * I(szenario)"

# comb
formulae_cmb <- list(formulae_lmm, formulae_glmm)

# fit models
Map(\(type, bool){

  lapply(type, \(x){

    if(bool){ # LMMs

      # fit LMM
      nlme::lme(as.formula(x), random = ~ 1 | individuum,
                data = dat_whale, na.action = na.omit,
                method = "REML") -> fit

      # summary
      list(fit,
            summary(fit))

    } else { # GLMMs

      # fit glmm PQL
      MASS::glmmPQL(as.formula(x), random = ~ 1 | individuum,
                    family = binomial(link = "logit"),
                    data = dat_whale) -> fit

      # summary
      list(fit,
            summary(fit))

    }

  }) |> setNames(type)

}, formulae_cmb, c(TRUE, FALSE)) |> setNames(c("LMM", "GLMM")) -> models

## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
## iteration 1
## iteration 2
## iteration 3
## iteration 4
```

```

# fits and summaries
LMM_fits <- lapply(models[[1]], "[", 1)
LMM_summaries <- lapply(models[[1]], "[", 2)
glmmPQL_fits <- lapply(models[[2]], "[", 1)
glmmPQL_summaries <- lapply(models[[2]], "[", 2)

# rebind fits for plotting
fits <- c(LMM_fits, glmmPQL_fits)
summaries <- c(LMM_summaries, glmmPQL_summaries)

```

4 Residual Diagnostics

```

# align
par(mfrow = c(5, 2), mar = c(2, 4, 4, 2) + 0.1)

# residual plots
invisible.Map(\(x, nom){

  invisible.Map(\(y, col){

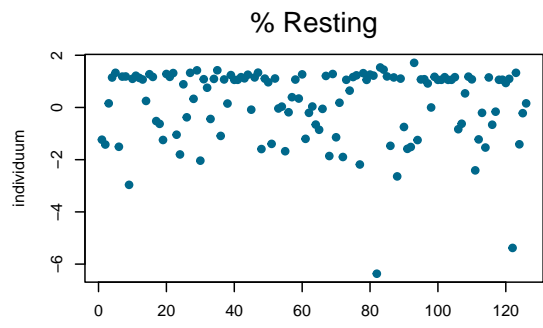
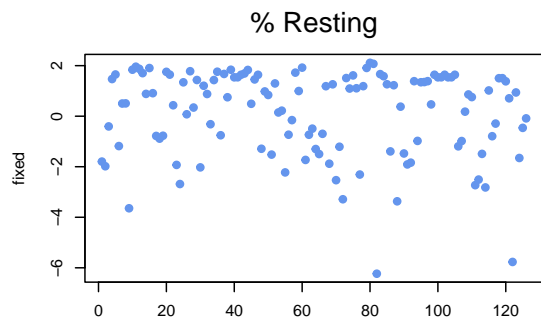
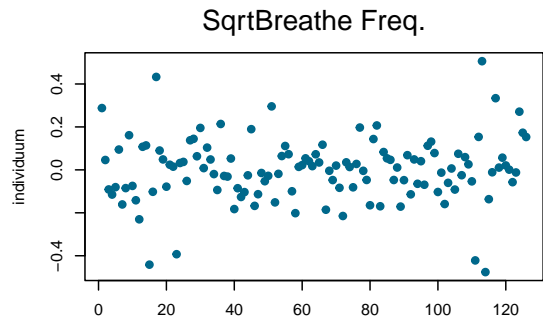
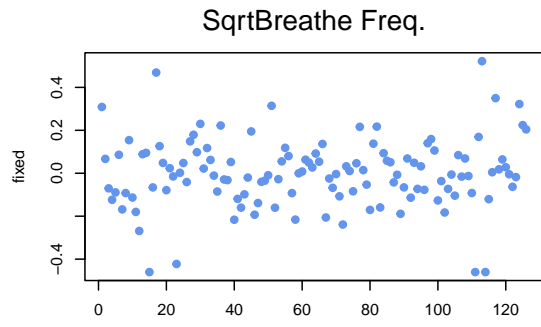
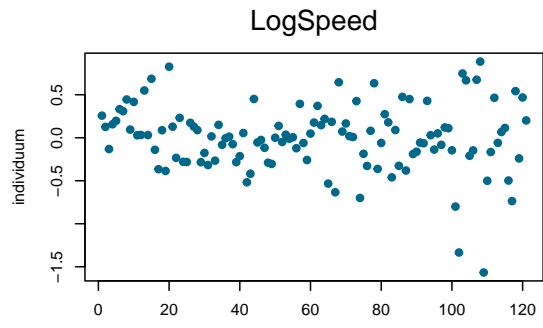
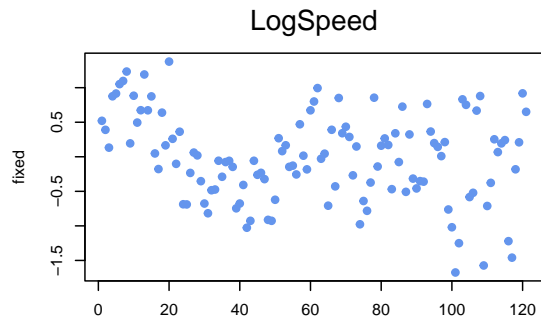
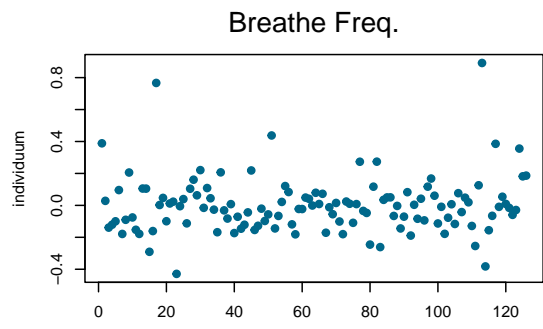
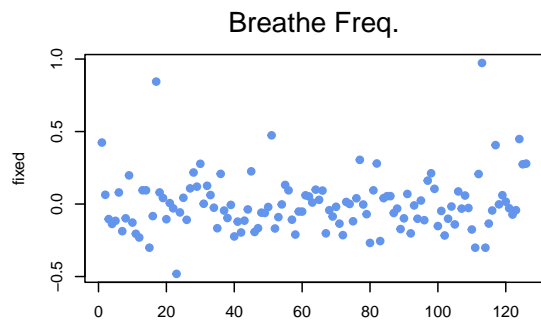
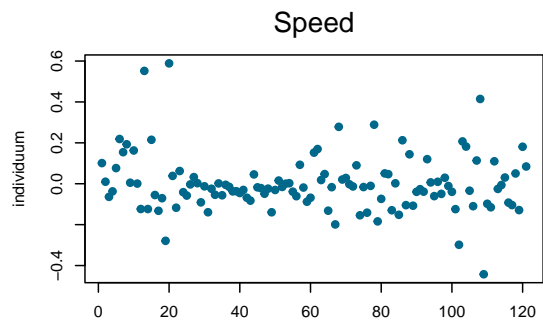
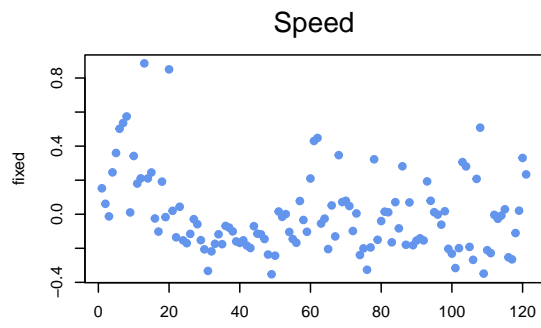
    # plots
    plot(x[["residuals"]][, y], type = "p", ylab = y, xlab = "", pch = 19,
         col = col)

    # label
    mtext(nom, side = 3, line = 1, cex = 1.2)

  }, c("fixed", "individuum"), c("cornflowerblue", "deepskyblue4"))

}, fits, c("Speed", "Breathe Freq.", "LogSpeed", "SqrtBreathe Freq.", "% Resting"))

```



```

# qqplots
par(mfrow = c(5, 2), mar = c(3.8, 4, 4, 2) + 0.1)

# residual plots
invisible.Map(\(x, nom){

  invisible.Map(\(y, col){

    # plots
    qqnorm(x[["residuals"]][, y], pch = 19, col = col, main = "")

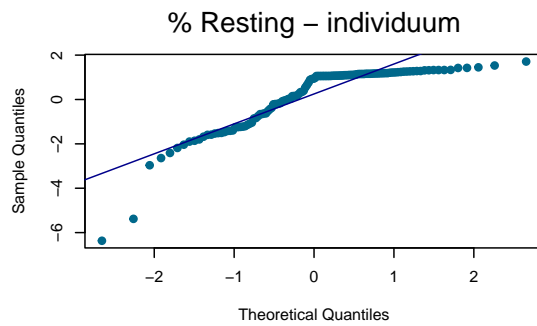
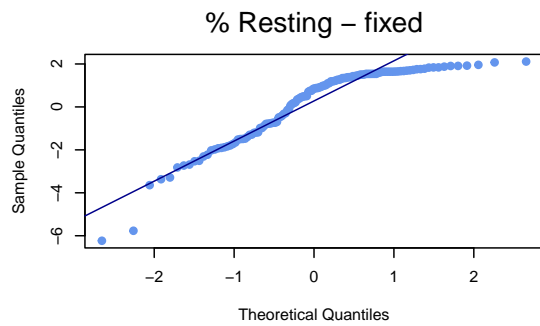
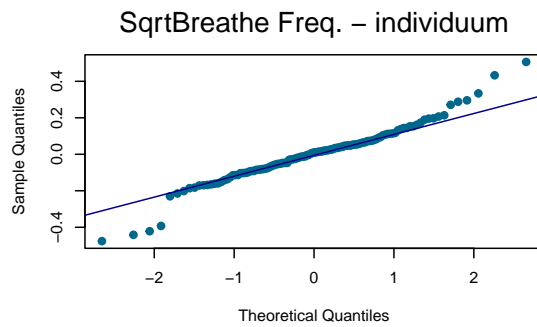
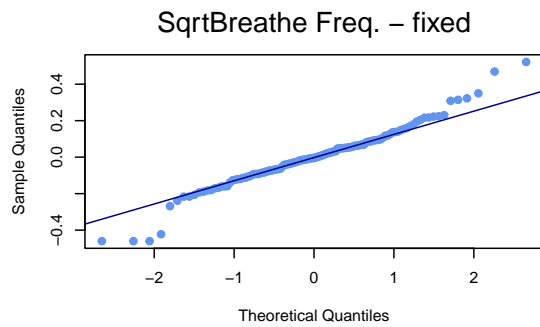
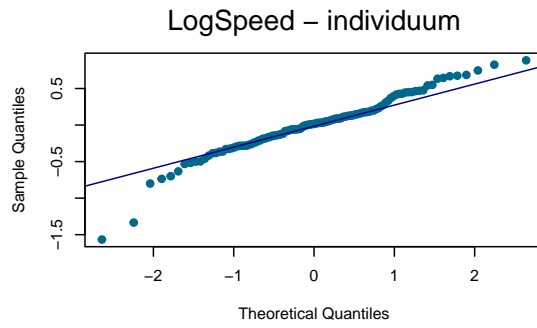
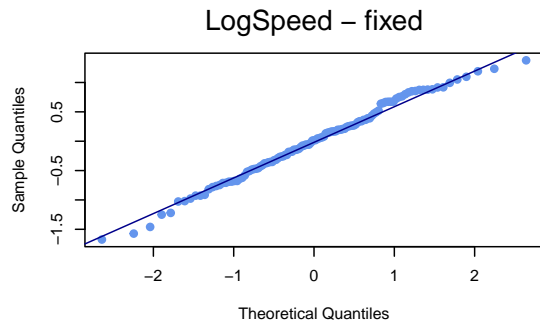
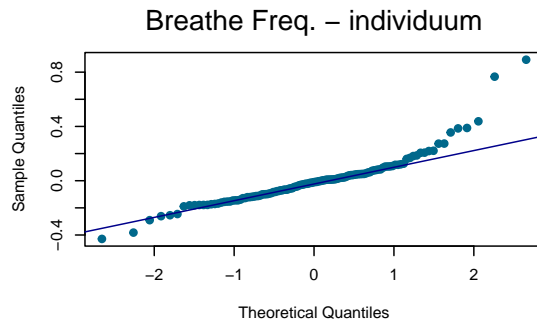
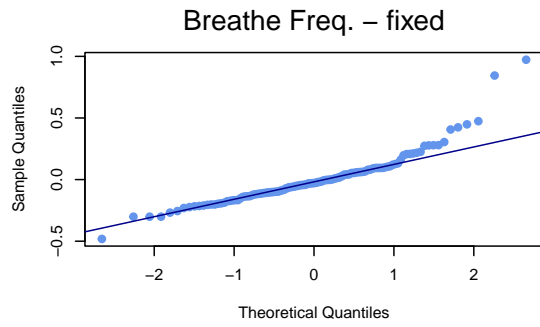
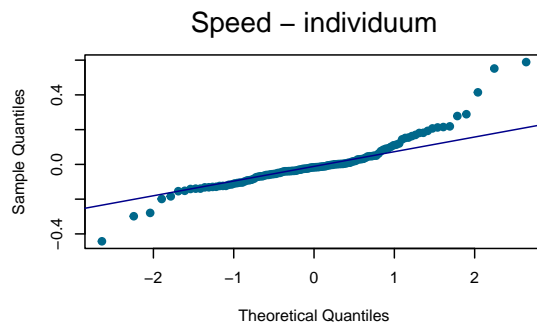
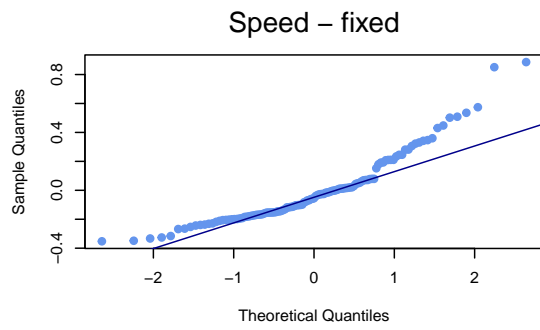
    # line
    qqline(x[["residuals"]][, y], col = "darkblue")

    # label
    mtext(paste(nom, "-", y), side = 3, line = 1, cex = 1.2)

  }, c("fixed", "individuum"), c("cornflowerblue", "deepskyblue4"))

}, fits, c("Speed", "Breathe Freq.", "LogSpeed", "SqrtBreathe Freq.", "% Resting"))

```



5 Coefficients and t-Tests

```
invis.lapply(names(summaries), \(x){
  # mods
  mod <- summaries[[x]][["tTable"]]

  # print table
  knitr::kable(mod)
}) |> setNames(names(summaries))
```

```
## $`speed ~ I(treatment) * I(szenario)`
```

```
##
```

```
##
```

	Value	Std.Error	DF	t-value	p-value
(Intercept)	0.3392015	0.0675357	73	5.0225532	0.0000035
I(treatment)Medium	-0.0796659	0.0950416	39	-0.8382218	0.4070142
I(treatment)High	0.0128286	0.0922713	39	0.1390310	0.8901411
I(szenario)During	-0.0132775	0.0659405	73	-0.2013561	0.8409802
I(szenario)After	0.0365231	0.0677611	73	0.5389981	0.5915282
I(treatment)Medium:I(szenario)During	0.0571340	0.0932540	73	0.6126709	0.5419979
I(treatment)High:I(szenario)During	0.1831087	0.0900919	73	2.0324658	0.0457448
I(treatment)Medium:I(szenario)After	0.0556807	0.0941578	73	0.5913548	0.5561095
I(treatment)High:I(szenario)After	0.0262800	0.0935384	73	0.2809546	0.7795400

```
##
```

```
## $`atem ~ I(treatment) * I(szenario)`
```

```
##
```

```
##
```

	Value	Std.Error	DF	t-value	p-value
(Intercept)	0.3487213	0.0568680	78	6.1321165	0.0000000
I(treatment)Medium	-0.0228728	0.0789743	39	-0.2896229	0.7736389
I(treatment)High	-0.0161951	0.0776965	39	-0.2084406	0.8359698
I(szenario)During	-0.0644547	0.0757239	78	-0.8511814	0.3972742
I(szenario)After	0.0673726	0.0757239	78	0.8897145	0.3763551
I(treatment)Medium:I(szenario)During	0.1421530	0.1051600	78	1.3517785	0.1803532
I(treatment)High:I(szenario)During	0.3022035	0.1034585	78	2.9210116	0.0045609
I(treatment)Medium:I(szenario)After	-0.0945670	0.1051600	78	-0.8992675	0.3712780
I(treatment)High:I(szenario)After	-0.0978333	0.1034585	78	-0.9456288	0.3472582

```
##
```

```
## $`logspeed ~ I(treatment) * I(szenario)`
```

```
##
```

```
##
```

	Value	Std.Error	DF	t-value	p-value
(Intercept)	-1.2308608	0.1828628	73	-6.7310605	0.0000000
I(treatment)Medium	-0.3267841	0.2573970	39	-1.2695723	0.2117607
I(treatment)High	-0.0179156	0.2498382	39	-0.0717087	0.9432002
I(szenario)During	-0.1078457	0.1807837	73	-0.5965456	0.5526563
I(szenario)After	0.0853729	0.1857559	73	0.4595971	0.6471713
I(treatment)Medium:I(szenario)During	0.3393236	0.2556667	73	1.3272106	0.1885751
I(treatment)High:I(szenario)During	0.5076296	0.2469975	73	2.0552010	0.0434380
I(treatment)Medium:I(szenario)After	0.2628398	0.2581038	73	1.0183495	0.3118751

```
## |I(treatment)High:I(szenario)After      | 0.0214941| 0.2564104| 73| 0.0838269| 0.9334235|
##
## $`sqrtatem ~ I(treatment) * I(szenario)`
##
##
## |                                     |      Value| Std.Error| DF|      t-value|      p-value|
## |:-----:|-----:|-----:|--:|-----:|-----:|
## |(Intercept)                        | 0.5703680| 0.0458895| 78| 12.4291561| 0.0000000|
## |I(treatment)Medium                 | -0.0123334| 0.0637282| 39| -0.1935318| 0.8475469|
## |I(treatment)High                   | -0.0050724| 0.0626970| 39| -0.0809038| 0.9359322|
## |I(szenario)During                   | -0.0472330| 0.0622771| 78| -0.7584316| 0.4504783|
## |I(szenario)After                   | 0.0589025| 0.0622771| 78| 0.9458125| 0.3471650|
## |I(treatment)Medium:I(szenario)During | 0.1147202| 0.0864861| 78| 1.3264577| 0.1885589|
## |I(treatment)High:I(szenario)During  | 0.2023393| 0.0850868| 78| 2.3780343| 0.0198531|
## |I(treatment)Medium:I(szenario)After | -0.0878070| 0.0864861| 78| -1.0152724| 0.3131150|
## |I(treatment)High:I(szenario)After   | -0.1322060| 0.0850868| 78| -1.5537787| 0.1242856|
##
## $`ruhezeit ~ I(treatment) * I(szenario)`
##
##
## |                                     |      Value| Std.Error| DF|      t-value|      p-value|
## |:-----:|-----:|-----:|--:|-----:|-----:|
## |(Intercept)                        | 2.3808540| 0.6231463| 78| 3.8206985| 0.0002658|
## |I(treatment)Medium                 | -0.7865013| 0.7862894| 39| -1.0002696| 0.3233462|
## |I(treatment)High                   | -0.3502397| 0.8093376| 39| -0.4327486| 0.6675812|
## |I(szenario)During                   | -0.0267061| 0.8095529| 78| -0.0329887| 0.9737679|
## |I(szenario)After                   | -1.0226216| 0.7058604| 78| -1.4487589| 0.1514133|
## |I(treatment)Medium:I(szenario)During | -0.7751521| 0.9789670| 78| -0.7918062| 0.4308751|
## |I(treatment)High:I(szenario)During  | -1.5056474| 0.9912816| 78| -1.5188896| 0.1328334|
## |I(treatment)Medium:I(szenario)After | 0.7143004| 0.9101265| 78| 0.7848364| 0.4349268|
## |I(treatment)High:I(szenario)After   | -0.3409966| 0.9105983| 78| -0.3744753| 0.7090669|
```