Antidepressant effects of esketamine in adolescents with major depressive disorder and suicidal ideation: a case series Reproducible analysis report

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1 Note

This R Markdown file will exactly reproduce our data tidying, transformation, analysis and graphing processes in the R software and produce an illustrative PDF file. Code-only scripts are also available in the Open Science Framework repository for this study, at https://osf.io/2ujqn/. The reproduction of this markdown file requires that the dataset available in the OSF repository is located in the same directory of this file, and that the user has the necessary dependencies to create a PDF using the bookdown package. Further information is available at the package website. Please refer to the repository for a detailed description of each variable contained herein. All files in the repository, including this one, are licensed under CC-BY-4.0.

To help with reproducibility, the {renv} R package was used. The renv.lock file and the package folder are also available and should be imported to the same location on your computer as this script. For further information on how to use {renv} visit the package website: https://rstudio.github.io/renv/articles/renv.html.

2 Load the required packages

```
# Required Packages
library(tidyverse)
library(haven)
library(lsr)
library(rstatix)
library(boot)
library(kableExtra)
library(patchwork)
library(RColorBrewer)
library(ggsci)
library(car)
library(purrr)
library(broom)
```

3 Load the dataset

The first step is to load the dataset into R. You can load the dataset directly from the GitHub repository.

```
## Load from GitHub repository

urlRemote <- "https://raw.githubusercontent.com/"
pathGithub <- "brenosouzamarques/LANP_Esketamine_Adolescents/master/"
fileName <- "Dataset.csv"

df <- pasteO(urlRemote, pathGithub, fileName) %>%
    read.csv()
```

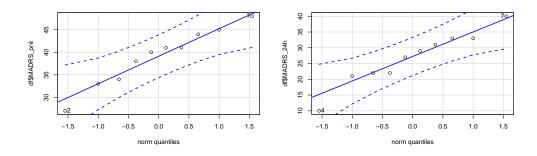
4 Visualizing distribution of dependent variables

Q-Q Plots are used to assess the symmetry of dependent variables, that is, **MADRS total score** and **MADRS item 10** at baseline and 24-h post-infusion.

4.1 MADRS total score at baseline and 24h post-treatment

```
qqplot_madrs_pre <- car::qqPlot(df$MADRS_pré)
qqplot_madrs_post <- car::qqPlot(df$MADRS_24h)
print(qqplot_madrs_pre, qqplot_madrs_post)</pre>
```

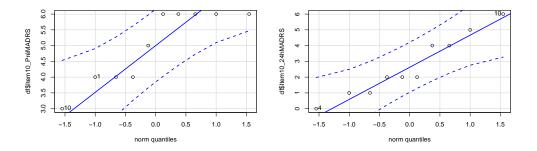
[1] 2 7



4.2 MADRS item 10 score at baseline and 24h post-treatment

```
qqplot_item10_pre <- car::qqPlot(df$Item10_PréMADRS)
qqplot_item10_post <- car::qqPlot(df$Item10_24hMADRS)
print(qqplot_item10_pre, qqplot_item10_post)</pre>
```

[1] 10 1



5 Descriptives

Total depressive symptoms and suicidal ideation scores are presented as means and standard deviations.

5.1 MADRS total scores

Table 1: MADRS total score

time	Mean	Std dev	Maximum	Minimum
baseline	39.1	6.297266	48	27
post	26.8	8.403703	40	10

5.2 MADRS item 10 scores

```
item_10_total <- c(df$Item10_PréMADRS, df$Item10_24hMADRS)
time <- rep(c("baseline", "post"), each = 10)
data_madrs_total <- tibble(time, item_10_total)
kbl(
    data_madrs_total %>%
    group_by(time) %>%
    summarise(
        Mean = mean(item_10_total),
        "Std dev" = sd(item_10_total),
        Maximum = max(item_10_total),
        Minimun = min(item_10_total)),
        caption = "MADRS item 10 score",
        booktabs = T) %>% kable_styling(latex_options = c("hold_position"))
```

Table 2: MADRS item 10 score

time	Mean	Std dev	Maximum	Minimun
baseline	5.0	1.154700	6	3
post	2.7	1.946507	6	0

6 Hypothesis tests

6.1 T-tests

We first aim to determine if the difference between pre-and post-treatment outcomes is significant. We test this hypothesis with a paired samples t-test for each outcome, as follows.

Table 3: T-tests

Dependent Variable	estimate	statistic	p.value
MADRS total score	12.3	1.221010	0.0022350
MADRS item 10	2.3		0.0118109

6.2 Effect sizes

Effect sizes allow further interpretation of how significant are these changes from baseline to 24 hours post-ketamine treatment.

Table 4: Effect sizes

Outcome	Effect Size (Cohen's D)
MADRS total score	1.3349078
MADRS item 10	0.9948934

7 Bootstrapped 95% confidence intervals

7.1 Create a bootstrap function

This functions creates a main element, the data_dens data frame. It stores the bootstrapped samples of means as they are calculated using the bias-corrected and accelerated (Bca) method. These are later used to plot density graphs that compare bootstrapped distributions of pre-and post-treatment symptoms.

```
b <- df[[f2]]</pre>
boot_a <- boot(a,</pre>
                function(x, j) mean(x[j], na.rm = T), #Boot baseline variable
boot_b <- boot(b,</pre>
                function(x, j) mean(x[j], na.rm = T), #Boot 7 days variable
                R = 10000)
boot_ci_a <- boot.ci(boot.out = boot_a, #Calculate baseline CI</pre>
                       conf = 0.95,
                       type = "bca")
boot_ci_b <- boot.ci(boot.out = boot_b, #Calculate post-treatment CI</pre>
                       conf = 0.95,
                       type = "bca")
print(boot_ci_a) # Return baseline bootstrapped CI
print(boot_ci_b) # Return post-treatment bootstrapped CI
## Code to extract bootstrapped sample means
densitiy_a <- boot_a$t</pre>
densitiy_b <- boot_b$t</pre>
dens <- c(densitiy_a, densitiy_b)</pre>
time <- rep(c("Baseline", "Post-Treatment"), each = 10000)</pre>
data_dens <<- data.frame(dens, time)</pre>
data_dens$time <- factor(data_dens$time)</pre>
```

7.2 Run bootstraps and create ggplot graph objects

7.2.1 Bootstrap - MADRS total score

```
### Bootstrap CI - MADRS total score
boot_funs(df, "MADRS_pré", "MADRS_24h")
## BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
## Based on 10000 bootstrap replicates
##
## boot.ci(boot.out = boot_a, conf = 0.95, type = "bca")
## Intervals :
               BCa
## Level
        (34.8, 42.3)
## 95%
## Calculations and Intervals on Original Scale
## BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
## Based on 10000 bootstrap replicates
##
## CALL :
## boot.ci(boot.out = boot_b, conf = 0.95, type = "bca")
## Intervals :
## Level
               BCa
## 95% (21.2, 31.1)
## Calculations and Intervals on Original Scale
plot_madrs <- ggplot(data_dens)+</pre>
            geom_density(aes(x = dens, fill = time),
                         color = "white",
                         alpha = 0.9) +
            labs(x = "Total MADRS scores")+
            scale_fill_brewer(name = "Time of assessment:",
                              palette = "Dark2")+
  theme_classic()+
  theme(text = element_text(family = "serif"),
       axis.text.x = element_text(size = 14),
       axis.title.x = element_text(size = 14),
       axis.text.y = element_text(size = 14),
       axis.title.y = element_text(size = 14),
       legend.text = element_text(size = 13),
        legend.title = element_text(size = 13)) ##ggplot2 graph
```

7.2.2 Bootstrap - MADRS item 10

```
### Bootstrap CI - MADRS item 10
boot_funs(df, "Item10_PréMADRS", "Item10_24hMADRS")
## BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
## Based on 10000 bootstrap replicates
##
## CALL :
## boot.ci(boot.out = boot_a, conf = 0.95, type = "bca")
## Intervals :
## Level
               BCa
        (4.2, 5.5)
## 95%
## Calculations and Intervals on Original Scale
## BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
## Based on 10000 bootstrap replicates
## CALL :
## boot.ci(boot.out = boot_b, conf = 0.95, type = "bca")
## Intervals :
               BCa
## Level
## 95%
        (1.6, 3.8)
## Calculations and Intervals on Original Scale
plot_madrs10 <- ggplot(data_dens)+</pre>
            geom_density(aes(x = dens, fill = factor(time)),
                         color = "white",
                         alpha = 0.9) +
            labs(x = "Item 10 MADRS scores",
                 y = element_blank())+
            scale_fill_brewer(name = "Time of assessment:",
                              palette = "Dark2")+
  theme_classic()+
  theme(text = element_text(family = "serif"),
       axis.text.x = element_text(size = 14),
        axis.title.x = element_text(size = 14),
        axis.text.y = element_text(size = 14),
        axis.title.y = element_text(size = 14),
        legend.text = element text(size = 13),
        legend.title = element_text(size = 13)) ##ggplot2 graph
```

8 Graphics

This first part creates an annotated plot using the ggplot2 density plots for the bootstrapped sample means and the patchwork package.

This next part uses a long formatted dataset to compare pre-and post-infusion results for each patient.

```
## Create dataset for total MADRS scores
madrs total <- c(df$MADRS pré, df$MADRS 24h)
time <- rep(c("baseline", "post"), each = 10)</pre>
id \leftarrow rep(1:10, 2)
id <- factor(id)</pre>
madrs_data_plot <- tibble(id, time, madrs_total)</pre>
## plot total MADRS scores
madrs_plot <- madrs_data_plot %>% ggplot()+
            geom_boxplot(aes(x = time, y = madrs_total),
                          width = .3,
                          alpha = .3,
                          color = "grey")+
            scale_y_continuous(breaks = seq(from = 10,
                                               to = 50,
                                              by = 5)) +
            scale_x_discrete(labels = c("Baseline", "24 hours post-infusion"))+
            geom_point(aes(x = time, y = madrs_total,
                            color = id,
                            group = id),
                        size = 1.5) +
            geom_line(aes(x = time, y = madrs_total,
                           color = id,
                           group = id,
                           alpha = .4),
                       size = 1.5) +
            scale_color_npg()+
            theme_classic()+
```

```
labs(x = "Time of assessment",
                  y = "Total MADRS scores")+
            theme(axis.title.x = element_text(size = 14),
                  axis.text.x = element_text(size = 13),
                  axis.title.y = element_text(size = 15),
                  axis.text.y = element_text(size = 13),
                  legend.position = "none")
## Create dataset for MADRS Item 10
si_pre <- df$Item10_PréMADRS</pre>
si_post <- df$Item10_24hMADRS</pre>
madrs_item10 <- c(si_pre, si_post)</pre>
time <- rep(c("baseline", "post"), each = 10)</pre>
id \leftarrow rep(1:10, 2)
id <- factor(id)</pre>
madrs_item_plot <- tibble(id, time, madrs_item10)</pre>
item10_plot <- madrs_item_plot %>% ggplot()+
            geom_boxplot(aes(x = time, y = madrs_item10),
                          width = .3,
                          alpha = .3,
                          color = "grey")+
            scale_y_continuous(breaks = c(0, 1, 2, 3, 4, 5, 6)) +
            scale_x_discrete(labels = c("Baseline", "24 hours post-infusion"))+
            geom_point(aes(x = time, y = madrs_item10,
                            color = id,
                            group = id),
                        size = 1.5,
                        position = position_dodge(0.08))+
            geom_line(aes(x = time, y = madrs_item10,
                           color = id,
                           group = id,
                           alpha = .4),
                       size = 1.5,
                       position = position_dodge(0.08))+
            scale_color_npg()+
            theme_classic()+
            labs(x = "Time of assessment",
                 y = "MADRS item 10 scores")+
            theme(axis.title.x = element_text(size = 14),
                  axis.text.x = element_text(size = 13),
                  axis.title.y = element_text(size = 15),
                  axis.text.y = element_text(size = 13),
                  legend.position = "none",
                   text = element_text(family = "serif"))
```

8.1 Join and save plots