

# Votre rapport

## Contents

```
#####
# Analyse pré-post (6 participants)
# -----
# 1. Charge les jeux de données preinter.csv / postinter.csv.
# 2. Sépare les variables :
#     • AE* = sentiment d'auto-efficacité
#     • COP* = régulation émotionnelle
# 3. Pour chaque variable :
#     • Shapiro-Wilk normalité (pré & post) --> si valeurs non identiques
#     • t-test Welch non apparié      + t-value, p-value
#     • d de Cohen                     + magnitude (très faible + très forte)
# 5. Concatène les résultats dans deux tableaux,
#####

#
# installation automatique des librairies
#
pkgs <- c(
  "tidyverse", # contient dplyr, readr, purrr, stringr, etc.
  "effsize",   # pour la fonction cohen.d()
  "broom",     # (optionnel) tidy() pour mettre en forme les résultats de tests
  "knitr"      # pour kable(), la mise en forme des tableaux
)
to_install <- pkgs[!(pkgs %in% rownames(installed.packages()))]
if (length(to_install)) install.packages(to_install, dependencies = TRUE)

#
# Chargement des librairies
#
invisible(lapply(pkgs, library, character.only = TRUE))
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2    3.5.2      v tibble    3.2.1
## v lubridate  1.9.4      v tidyr     1.3.1
## v purrr      1.0.4
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```

#
# Lecture des fichiers
#
pre <- read_delim(
  "preinter.csv",
  delim      = ";", # le séparateur est un ; exce francophone
  show_col_types = FALSE, # on n'affiche pas le type de données par colones
  locale       = locale(encoding = "UTF-8"), # caractères classique
  col_types    = cols(
    `n°`      = col_skip(),      # on ignore la colonne ID
    .default = col_double()     # toutes les autres colonnes sont définis comme numériques
  )
)

post <- read_delim(
  "postinter.csv",
  delim      = ";",
  show_col_types = FALSE,
  locale       = locale(encoding = "UTF-8"),
  col_types    = cols(
    `n°`      = col_skip(),
    .default = col_double()
  )
)

#
# (c) Détection des familles de variables
#
vars_ae <- names(pre) %>% str_subset("^AE")
vars_cop <- names(pre) %>% str_subset("^COP")

# verification que les 2 fichiers (pre/post) contiennent les mêmes variables
vars_ae <- intersect(vars_ae, names(post))
vars_cop <- intersect(vars_cop, names(post))

cat("Variables de sentiment d'auto-efficacité (AE) identifiées :\n")

```

## Variables de sentiment d'auto-efficacité (AE) identifiées :

```
cat(paste0("- ", vars_ae), sep = "\n")
```

```

## - AE 1
## - AE 2
## - AE 3
## - AE 4
## - AE 5
## - AE 6
## - AE 7

```

```
cat("\nVariables de régulation émotionnelle (COP) identifiées :\n")
```

```
##
## Variables de régulation émotionnelle (COP) identifiées :
```

```
cat(paste0("- ", vars_cop), sep = "\n")
```

```
## - COP1
## - COP2
## - COP3
## - COP4
## - COP5
## - COP6
## - COP7
## - COP8
## - COP9
## - COP10
## - COP11
## - COP12
## - COP13
## - COP14
## - COP15
## - COP16
## - COP17
## - COP18
```

```
# création des tableaux vides
table_ae <- tibble(
  Variable = character(),
  `Pré intervention moyenne (ET)` = character(),
  `Post intervention moyenne (ET)` = character(),
  t = numeric(),
  p = numeric(),
  `d de Cohen` = numeric(),
  `Taille d'effet` = character(),
  `Shapiro pré (p)` = numeric(),
  `Shapiro post (p)` = numeric()
)

table_cop <- tibble(
  Variable = character(),
  `Pré intervention moyenne (ET)` = character(),
  `Post intervention moyenne (ET)` = character(),
  t = numeric(),
  p = numeric(),
  `d de Cohen` = numeric(),
  `Taille d'effet` = character(),
  `Shapiro pré (p)` = numeric(),
  `Shapiro post (p)` = numeric()
)

# --- Analyse AE (sentiment d'auto-efficacité) ---
for (var in vars_ae) {
  # Extraction des données pré- et post-intervention
  pre_vals <- (pre[[var]])
```

```

post_vals <- (post[[var]])

# Test de normalité Shapiro-Wilk (pré)
shapiro_pre_test <- shapiro.test(pre_vals)
p_shapiro_pre <- shapiro_pre_test$p.value
# Affichage du W et de la p-value
cat(sprintf(
  "Shapiro-Wilk pré [%s] : W = %.3f, p = %.3f\n",
  var,
  shapiro_pre_test$statistic,
  signif(p_shapiro_pre, 3)
))

# Test de normalité Shapiro-Wilk (post)
shapiro_post_test <- shapiro.test(post_vals)
p_shapiro_post <- shapiro_post_test$p.value
# Affichage du W et de la p-value
cat(sprintf(
  "Shapiro-Wilk post [%s] : W = %.3f, p = %.3f\n",
  var,
  shapiro_post_test$statistic,
  signif(p_shapiro_post, 3)
))

# Test t apparié (Welch)
t_res <- t.test(post_vals, pre_vals, paired = TRUE)

# Calcul du d de Cohen
# d de Cohen pour données appariées
d_res <- effsize::cohen.d(
  post_vals,      # vecteur post-intervention
  pre_vals,       # vecteur pré-intervention
  paired = TRUE,  # appariées
)

# Interprétation de la taille d'effet je sais pas si c'est ça que tu veux associer d à un type d'effet
magnitude <- cut(
  abs(d_res$estimate),
  breaks = c(0, 0.2, 0.5, 0.8, 1.3, Inf),
  labels = c("très faible", "faible", "moyenne", "forte", "très forte"),
  right = FALSE
)

# remplissage du tableau AE
table_ae <- table_ae %>%
  add_row(
    Variable = var,
    `Pré intervention moyenne (ET)` = sprintf("%.2f (%.2f)", mean(pre_vals), sd(pre_vals)),
    `Post intervention moyenne (ET)` = sprintf("%.2f (%.2f)", mean(post_vals), sd(post_vals)),
    t = round(t_res$statistic, 3),
    p = signif(t_res$p.value, 3),
    `d de Cohen` = round(d_res$estimate, 3),
    `Taille d'effet` = as.character(magnitude),
    `Shapiro pré (p)` = signif(p_shapiro_pre, 3),
  )

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```

    `Shapiro post (p)` = signif(p_shapiro_post, 3)
  )
}

```

```

## Shapiro-Wilk pré [AE 1] : W = 0.823, p = 0.093
## Shapiro-Wilk post [AE 1] : W = 0.822, p = 0.091
##
## Shapiro-Wilk pré [AE 2] : W = 0.832, p = 0.111
## Shapiro-Wilk post [AE 2] : W = 0.775, p = 0.035
##
## Shapiro-Wilk pré [AE 3] : W = 0.866, p = 0.212
## Shapiro-Wilk post [AE 3] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [AE 4] : W = 0.960, p = 0.820
## Shapiro-Wilk post [AE 4] : W = 0.822, p = 0.091
##
## Shapiro-Wilk pré [AE 5] : W = 0.640, p = 0.001
## Shapiro-Wilk post [AE 5] : W = 0.822, p = 0.091
##
## Shapiro-Wilk pré [AE 6] : W = 0.918, p = 0.493
## Shapiro-Wilk post [AE 6] : W = 0.781, p = 0.039
##
## Shapiro-Wilk pré [AE 7] : W = 0.891, p = 0.324
## Shapiro-Wilk post [AE 7] : W = 0.950, p = 0.737

```

```

# --- COP (régulation émotionnelle) ---
for (var in vars_cop) {
  # Extraction des données pré- et post-intervention
  pre_vals <- (pre[[var]])
  post_vals <- (post[[var]])

  # Test de normalité Shapiro-Wilk (pré)
  if (length(unique(pre_vals)) > 1) {
    sh_pre <- shapiro.test(pre_vals)
    p_shapiro_pre <- sh_pre$p.value
    cat(sprintf(
      "Shapiro-Wilk pré [%s] : W = %.3f, p = %.3f\n",
      var,
      sh_pre$statistic,
      signif(p_shapiro_pre, 3)
    ))
  } else {
    # Cas où toutes les réponses sont identiques ou trop peu nombreuses
    p_shapiro_pre <- NA_real_
    cat(sprintf(
      "Shapiro-Wilk pré [%s] : NA (valeurs identiques)\n",
      var
    ))
  }
}

# Test de normalité Shapiro-Wilk (post)
if (length(unique(post_vals)) > 1) {
  sh_post <- shapiro.test(post_vals)

```

```

p_shapiro_post <- sh_post$p.value
cat(sprintf(
  "Shapiro-Wilk post [%s] : W = %.3f, p = %.3f\n\n",
  var,
  sh_post$statistic,
  signif(p_shapiro_post, 3)
))
} else {
  p_shapiro_post <- NA_real_
  cat(sprintf(
    "Shapiro-Wilk post [%s] : NA (valeurs identiques)\n\n",
    var
  ))
}
# Test t apparié
t_res <- t.test(post_vals, pre_vals, paired = TRUE)

# Calcul du d de Cohen
d_res <- effsize::cohen.d(
  post_vals,      # vecteur post-intervention
  pre_vals,       # vecteur pré-intervention
  paired = TRUE,  # appariées
)

# Interprétation de la taille d'effet
magnitude <- cut(
  abs(d_res$estimate),
  breaks = c(0, 0.2, 0.5, 0.8, 1.3, Inf),
  labels = c("très faible", "faible", "moyenne", "forte", "très forte"),
  right = FALSE
)

# Remplissage du tableau COP
table_cop <- table_cop %>%
  add_row(
    Variable = var,
    `Pré intervention moyenne (ET)` = sprintf("%.2f (%.2f)", mean(pre_vals), sd(pre_vals)),
    `Post intervention moyenne (ET)` = sprintf("%.2f (%.2f)", mean(post_vals), sd(post_vals)),
    t = round(t_res$statistic, 3),
    p = signif(t_res$p.value, 3),
    `d de Cohen` = round(d_res$estimate, 3),
    `Taille d'effet` = as.character(magnitude),
    `Shapiro pré (p)` = signif(p_shapiro_pre, 3),
    `Shapiro post (p)` = signif(p_shapiro_post, 3)
  )
}

```

```

## Shapiro-Wilk pré [COP1] : W = 0.751, p = 0.020
## Shapiro-Wilk post [COP1] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [COP2] : W = 0.921, p = 0.514
## Shapiro-Wilk post [COP2] : W = 0.866, p = 0.212
##
## Shapiro-Wilk pré [COP3] : W = 0.805, p = 0.065

```

```

## Shapiro-Wilk post [COP3] : NA (valeurs identiques)
##
## Shapiro-Wilk pré [COP4] : W = 0.640, p = 0.001
## Shapiro-Wilk post [COP4] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [COP5] : W = 0.827, p = 0.101
## Shapiro-Wilk post [COP5] : W = 0.775, p = 0.035
##
## Shapiro-Wilk pré [COP6] : W = 0.907, p = 0.415
## Shapiro-Wilk post [COP6] : W = 0.721, p = 0.010
##
## Shapiro-Wilk pré [COP7] : W = 0.831, p = 0.110
## Shapiro-Wilk post [COP7] : W = 0.773, p = 0.033
##
## Shapiro-Wilk pré [COP8] : W = 0.869, p = 0.223
## Shapiro-Wilk post [COP8] : W = 0.908, p = 0.425
##
## Shapiro-Wilk pré [COP9] : W = 0.701, p = 0.006
## Shapiro-Wilk post [COP9] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [COP10] : W = 0.640, p = 0.001
## Shapiro-Wilk post [COP10] : W = 0.683, p = 0.004
##
## Shapiro-Wilk pré [COP11] : W = 0.683, p = 0.004
## Shapiro-Wilk post [COP11] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [COP12] : W = 0.908, p = 0.421
## Shapiro-Wilk post [COP12] : W = 0.866, p = 0.212
##
## Shapiro-Wilk pré [COP13] : W = 0.822, p = 0.091
## Shapiro-Wilk post [COP13] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [COP14] : W = 0.683, p = 0.004
## Shapiro-Wilk post [COP14] : W = 0.701, p = 0.006
##
## Shapiro-Wilk pré [COP15] : W = 0.640, p = 0.001
## Shapiro-Wilk post [COP15] : W = 0.640, p = 0.001
##
## Shapiro-Wilk pré [COP16] : W = 0.822, p = 0.091
## Shapiro-Wilk post [COP16] : W = 0.683, p = 0.004
##
## Shapiro-Wilk pré [COP17] : W = 0.683, p = 0.004
## Shapiro-Wilk post [COP17] : W = 0.496, p = 0.000
##
## Shapiro-Wilk pré [COP18] : W = 0.908, p = 0.421
## Shapiro-Wilk post [COP18] : W = 0.702, p = 0.006

```

```

#
# (f) Affichage final
#
cat("## Tableau AE - auto-efficacité\n")

```

```

## ## Tableau AE - auto-efficacité

```

```
print(knitr::kable(table_ae, align="lccc", caption="Items AE"))
```

```
##
##
## Table: Items AE
##
## |Variable | Pré intervention moyenne (ET) | Post intervention moyenne (ET) | t | p | d de Col
## |-----| :-----| :-----| :-----| :-----| :-----|
## |AE 1 | 5.33 (1.37) | 6.33 (0.82) | 2.739 | 0.0409 | 0.747
## |AE 2 | 5.50 (1.76) | 6.17 (0.98) | 1.085 | 0.3280 | 0.434
## |AE 3 | 6.17 (0.75) | 6.33 (0.52) | 0.542 | 0.6110 | 0.254
## |AE 4 | 5.50 (1.05) | 6.33 (0.82) | 2.712 | 0.0422 | 0.857
## |AE 5 | 5.67 (1.03) | 6.33 (0.82) | 2.000 | 0.1020 | 0.700
## |AE 6 | 4.67 (2.07) | 5.33 (1.86) | 0.791 | 0.4650 | 0.338
## |AE 7 | 4.17 (2.40) | 4.33 (1.86) | 0.210 | 0.8420 | 0.076
```

```
cat("\n## Tableau COP - régulation émotionnelle\n")
```

```
##
## ## Tableau COP - régulation émotionnelle
```

```
print(knitr::kable(table_cop, align="lccc", caption="Items COP"))
```

```
##
##
## Table: Items COP
##
## |Variable | Pré intervention moyenne (ET) | Post intervention moyenne (ET) | t | p | d de C
## |-----| :-----| :-----| :-----| :-----| :-----|
## |COP1 | 3.83 (1.47) | 4.67 (0.52) | 1.185 | 0.2890 | 0.795
## |COP2 | 3.17 (1.33) | 3.67 (1.51) | 0.808 | 0.4560 | 0.351
## |COP3 | 3.83 (1.33) | 5.00 (0.00) | 2.150 | 0.0842 | 1.241
## |COP4 | 1.67 (0.52) | 1.67 (1.03) | 0.000 | 1.0000 | 0.000
## |COP5 | 2.00 (0.63) | 1.83 (0.98) | -0.307 | 0.7710 | -0.204
## |COP6 | 3.33 (1.21) | 4.00 (1.55) | 1.348 | 0.2350 | 0.467
## |COP7 | 2.00 (1.26) | 1.83 (1.17) | -0.542 | 0.6110 | -0.136
## |COP8 | 2.83 (1.72) | 2.83 (1.60) | 0.000 | 1.0000 | 0.000
## |COP9 | 4.50 (0.84) | 4.67 (0.52) | 0.542 | 0.6110 | 0.229
## |COP10 | 4.67 (0.52) | 4.50 (0.55) | -1.000 | 0.3630 | -0.312
## |COP11 | 1.50 (0.55) | 1.33 (0.52) | -0.542 | 0.6110 | -0.313
## |COP12 | 3.17 (1.17) | 3.33 (1.51) | 0.307 | 0.7710 | 0.122
## |COP13 | 3.67 (0.82) | 4.67 (0.52) | 3.873 | 0.0117 | 1.356
## |COP14 | 1.50 (0.55) | 1.50 (0.84) | 0.000 | 1.0000 | 0.000
## |COP15 | 4.67 (0.52) | 4.67 (0.52) | NaN | NaN | NaN
## |COP16 | 4.33 (0.82) | 4.50 (0.55) | 1.000 | 0.3630 | 0.188
## |COP17 | 1.50 (0.55) | 1.33 (0.82) | -0.349 | 0.7410 | -0.243
## |COP18 | 2.17 (1.17) | 1.83 (1.33) | -0.791 | 0.4650 | -0.264
```