Reanalysis of 06-Wile

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Reference

Wile et al. (2017). Serotonin and dopamine transporter PET changes in the premotor phase of LRRK2 parkinsonism: cross-sectional studies. $Lancet\ Neurology,\ 16(5),\ 351-359.\ https://doi.org/10.1016/S1474-4422(17)30056-X$

Notes from reading methods section

- Dependant variable: PET radiotracer called 11C-DASB in cortex (first of many ANCOVAs reported)
- Independant variable: Group

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```
Healty controls (n=9)
LRRK2 w/o PD (n=9)
LRRK2 w PD (n=7)
sporadic PD (n=13)
```

- Covariate: age
- age was not estimable for LRRK2 without manifest Parkinson's disease
- Design: 1-way ANCOVA with group (4 leves) as IV and age as covariate

Reading data

Data is loaded, reshaped if necessary, and factors are specified.

Descriptives

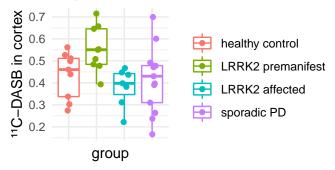
Dependant variable

Number of samples and mean (SD) in levels of the independant variables. We reproduce Table 3 and Figure 2A of the study.

```
idx = c(1, 3, 2, 4) # sorting as in publication
tab.dv = array(NA, dim=c(4,3))
tab.dv[,1] = levels(data$Group.factor)
tab.dv[,2] = summary(data$Group.factor)
tab.dv[,3] = tapply(data$Cortex, data$Group.factor,
                 function (x) sprintf(\%0.2f(\%0.2f)), mean(x), sd(x))
colnames(tab.dv) = c("group", "n", "mean (SD)")
print(tab.dv)
                                 mean (SD)
       group
## [1,] "healthy control"
                            "9" "0.43 (0.10)"
## [2,] "LRRK2 premanifest" "9" "0.55 (0.10)"
## [3,] "LRRK2 affected"
                            "7" "0.38 (0.09)"
                            "13" "0.42 (0.15)"
## [4,] "sporadic PD"
ggplot(data, aes(y=Cortex, x=Group.factor, color=Group.factor)) +
  geom_boxplot() +
  geom_point(position = position_jitter(width = 0.15, height = 0)) +
  theme_minimal() +
```

```
theme(axis.text.x = element_blank(), legend.title = element_blank()) +
xlab("group") + ylab("''C-DASB in cortex") + ggtitle("Dependent variable")
```

Dependant variable



Covariate(s)

```
idx = c(1, 3, 2, 4) # sorting as in publication
tab.cv = array(NA, dim=c(4,3))
tab.cv[,1] = levels(data$Group.factor)
tab.cv[,2] = summary(data$Group.factor)
tab.cv[,3] = tapply(data$`Age at PET`, data$Group.factor,
                  function (x) sprintf("%0.1f(%0.1f)", mean(x), sd(x)))
colnames(tab.cv) = c("group", "n", "mean (SD)")
print(tab.cv)
                                 mean (SD)
        group
## [1,] "healthy control"
                            "9"
                                 "57.4 (18.4)"
## [2,] "LRRK2 premanifest" "9"
                                 "49.8 (10.8)"
                            "7" "65.9 (14.7)"
## [3,] "LRRK2 affected"
## [4,] "sporadic PD"
                            "13" "57.8 (8.5)"
ggplot(data, aes(y=`Age at PET`, x=Group.factor, color=Group.factor)) +
  geom_boxplot() +
```

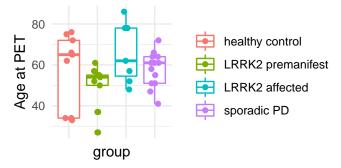
geom_point(position = position_jitter(width = 0.15, height = 0)) +

xlab("group") + ylab("Age at PET") + ggtitle("Covarate")

theme(axis.text.x = element_blank(), legend.title = element_blank()) +

Covarate

theme_minimal() +



Main analysis ANCOVA

```
# Orthogonal contrasts
contrasts(data$Group.factor) = contr.helmert(4)
fit.ancova = aov(Cortex ~ `Age at PET` + Group.factor, data = data)
# result = summary(fit) # Type I
result = Anova(fit.ancova, type=3) # Type III
print(result)
## Anova Table (Type III tests)
##
## Response: Cortex
                Sum Sq Df F value
                                     Pr(>F)
## (Intercept) 0.36361 1 25.1466 1.767e-05 ***
## `Age at PET` 0.00073 1 0.0502
                                     0.8241
## Group.factor 0.12125 3 2.7951
                                     0.0555 .
## Residuals
               0.47717 33
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Comparing ANCOVA in original study with reanalysis

Independant variable

```
tab.IV = rbind(stats.orig.IV, stats.rep.IV)
rownames(tab.IV) = c("original Study", "reanalysis")
print(t(tab.IV))
          original Study reanalysis
                          "2.80"
## Fvalue NA
## df1
                          " 3"
          NA
## df2
          NA
                          "33"
## pvalue "n.s."
                         "0.056"
## MD
          NA
                         NA
## lowerCI NA
                          NA
## upperCI NA
                         NA
```

Covariate

```
tab.CV = rbind(stats.orig.CV, stats.rep.CV)
rownames(tab.CV) = c("original Study", "reanalysis")
print(t(tab.CV))
           original Study reanalysis
                          "0.05"
## Fvalue NA
                          " 1"
## df1
          NA
                          "33"
## df2
           NA
## pvalue NA
                          "0.82"
## MD
          NA
                          NA
```

```
## lowerCI NA NA NA NA
```

Assumptions

1. Homogeneity of variance

- ANOVA/ANCOVA is fairly robust in terms of the error rate when sample sizes are equal.
- When groups with larger sample sizes have larger variances than the groups with smaller sample sizes, the resulting F-ratio tends to be conservative. That is, it's more likely to produce a non-significant result when a genuine difference does exist in the population.
- Conversely, when the groups with larger sample sizes have smaller variances than the groups with smaller samples sizes, the resulting F-ratio tends to be liberal and can inflate the false positive rate.

```
tapply(data$Cortex, data$Group.factor, sd)

## healthy control LRRK2 premanifest LRRK2 affected sporadic PD
## 0.10480960 0.10221440 0.08699411 0.14748558

leveneTest(Cortex ~ Group.factor, data = data)

## Levene's Test for Homogeneity of Variance (center = median)
## Df F value Pr(>F)
## group 3 0.821 0.4914
## 34
```

2. Independence between covariate and IV

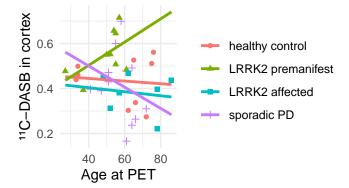
When the covariate and the experimental effect (independent variable) are not independent the treatment effect is obscured, spurious treatment effects can arise and the interpretation of the ANCOVA is seriously compromised.

We test whether our groups differ on the CV. If the groups do not significantly differ then is appropriate to use the covariate.

3. Homogeneity of regression slopes

```
fit.hrs = aov(Cortex ~ `Age at PET`*Group.factor, data = data)
Anova(fit.hrs, type=3)

## Anova Table (Type III tests)
##
## Response: Cortex
## Sum Sq Df F value Pr(>F)
## (Intercept) 0.31357 1 21.8078 5.908e-05 ***
## `Age at PET` 0.00032 1 0.0226 0.8816
```



Additional analyses: MANCOVA

The paper performed four ANCOVAs for cortex, striatum, brainstem, and hypothalamus but did not correct for multiple testing. Therefore, we also perform a MANCOVA.

```
fit = manova(cbind(Cortex, Striatum, Brainstem, Hypothalamus) ~ Group.factor + `Age at PET`,
             data=data)
summary(fit)
                Df Pillai approx F num Df den Df
                                                      Pr(>F)
                 3 0.95385
                             3.7293
                                         12
                                                96 0.0001241 ***
## Group.factor
## 'Age at PET'
                 1 0.36886
                             4.3832
                                          4
                                                30 0.0065618 **
## Residuals
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Notes

- The first reported ANCOVA which was n.s. was reproduced
- p-value was not reported
- Authors also report age unadjusted analyses and report these as ANCOVAs instead of ANOVAs (p. 357)
 which was confusing
- Assumptions were met, except homogeneity of regression slopes showing a different relationship of outcome and covariate in the LRRK2 premanifest group
- Many analyses, they corrected for multiplicity

Data was analyzed according to recommendations by Field, Miles, & Field (2012).