

Untitled

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```
library(data.table)
library(dplyr)

## -----

## data.table + dplyr code now lives in dtplyr.
## Please library(dtplyr)!

## -----

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:data.table':
##
##   between, first, last

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

library(ggplot2)

# Load data

post.attitude <- fread(input = "./REVISED_FINAL_POST_ATTITUDE_rename.csv",
  header = TRUE, stringsAsFactors = TRUE)
Pre_post <- fread(input = "./Pre_post.csv",
  header = TRUE, stringsAsFactors = TRUE)

### linear Regression on A0.3, A0.4, A0.57,A0.58,Experience
#### not enough variaion in 0.5, 0.60

# #check levels
# sapply(post.attitude %>%
#   select(A0.3,A0.4,A0.56,A0.57,A0.58,
#     A0.67, A0.68, A0.69, A0.70, A0.71, A0.72,
#     Experience),function(x)levels(x))

# merge pre_post and post.attitude
pre_post.attitude <- merge(Pre_post,post.attitude,by='BPL.BLD.ID', all.x = TRUE)

# select E group
pre_post.attitude_E <- pre_post.attitude %>%
  filter(LogCheck == "Y")
```

```

# ONE - NO TECH, NO AGE OR EXPERIENCE
all_pvals <- data.frame(varname=as.character(), col=as.character(), p_val=as.numeric())

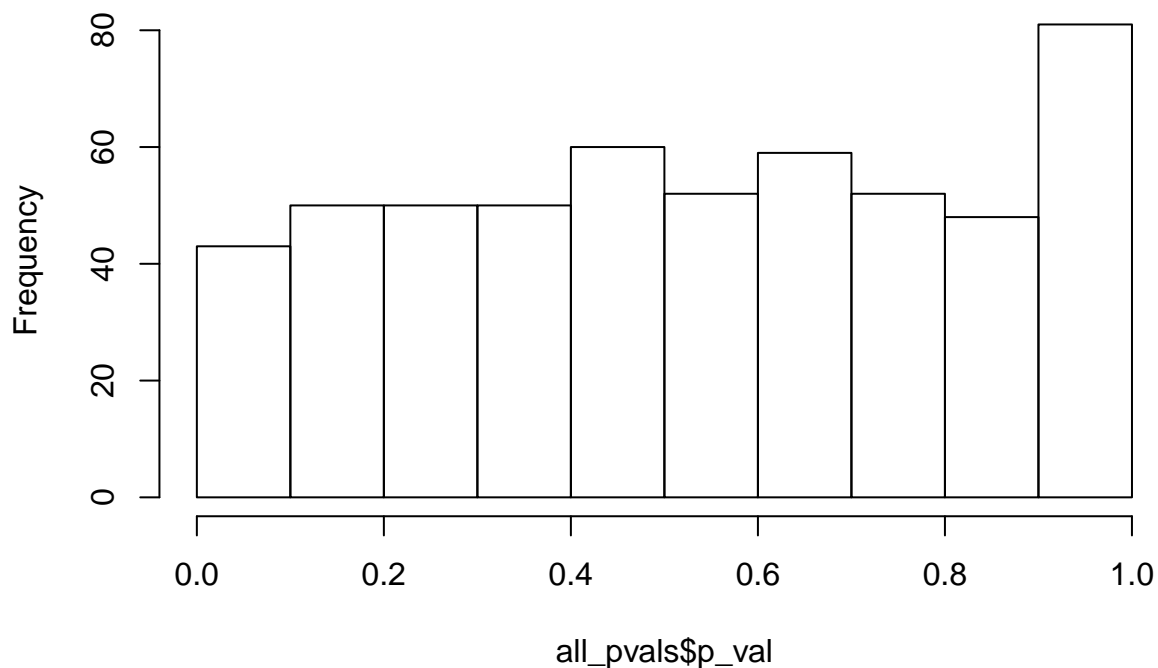
for (col in names(pre_post.attitude_E)[grep("X", names(pre_post.attitude_E))]) {
  # cat(col, "\n")
  smy <- summary(lm(paste(col, "~ A0.3 + A0.4 + A0.56 + A0.57 + A0.58 + A0.59 + Age + Experience + A0.6"),
                    data = pre_post.attitude_E))

  all_pvals <- rbind(all_pvals,
                    data.frame(varname = names(smy$coefficients[,4]), col = col, p_val = smy$coefficients[,4]))
}

hist(all_pvals$p_val)

```

Histogram of all_pvals\$p_val



```

all_pvals %>% filter(p_val < .05)

##           varname col      p_val
## 1      A0.57Yes X1.4 0.017476011
## 2      A0.57Yes X2.1 0.030605236
## 3    (Intercept) X3.1 0.041709267
## 4      Age44-54 X3.1 0.004290497
## 5      Age55+ X3.1 0.011591706
## 6 Experience15-20Y X3.1 0.006519213
## 7 Experience20+Y X3.1 0.002421040
## 8 Experience5-10Y X3.1 0.003974522

```

```
## 9      A0.56Yes X3.3 0.035350768
## 10     A0.57Yes X3.3 0.001050115
## 11     A0.59Yes X3.3 0.027619137
## 12     A0.56Yes X3.6 0.040277584
## 13     A0.56Yes X4.5 0.029526885
## 14     A0.58Yes X4.5 0.048757364
## 15     Age44-54 X4.5 0.009357408
## 16     Age55+ X4.5 0.035390824
## 17 Experience15-20Y X4.5 0.008122051
## 18 Experience20+Y X4.5 0.018161658
## 19     A0.69Yes X4.8 0.023024443
## 20     A0.57Yes X6.2 0.038793549
## 21 Experience5-10Y X6.3 0.039922465
## 22 A0.3Very involved X7.1 0.035656837
## 23     A0.56Yes X7.1 0.049678794
## 24     Age55+ X7.1 0.013745740
## 25 Experience15-20Y X7.1 0.029065456
## 26 Experience20+Y X7.1 0.013113864
```

```
p.adjust(all_pvals$p_val, method = "BH")
```

```
## [1] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [8] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [15] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [22] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [29] 1.0000000 1.0000000 1.0000000 1.0000000      NaN      NaN      NaN
## [36]      NaN      NaN      NaN      NaN      NaN      NaN      NaN
## [43]      NaN      NaN      NaN      NaN      NaN      NaN 1.0000000
## [50] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 0.8248419 1.0000000
## [57] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [64] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 0.9471479
## [71] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [78] 1.0000000 1.0000000      NaN      NaN      NaN      NaN      NaN
## [85]      NaN      NaN      NaN      NaN      NaN      NaN      NaN
## [92]      NaN      NaN      NaN      NaN 1.0000000 1.0000000 1.0000000
## [99] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [106] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [113] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [120] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [127] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [134] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [141] 1.0000000 1.0000000 0.9471479 1.0000000 0.9961561 1.0000000 1.0000000
## [148] 1.0000000 1.0000000 0.9731922 1.0000000 0.5845802 0.7491428 0.7105942
## [155] 0.5845802 0.5845802 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [162] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [169] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [176] 1.0000000 1.0000000 1.0000000 1.0000000 0.9471479 0.5723125 1.0000000
## [183] 0.9471479 1.0000000 1.0000000 0.9731922 1.0000000 1.0000000 1.0000000
## [190] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [197] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [204] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [211] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [218] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [225] 1.0000000 1.0000000 1.0000000 0.9471479 1.0000000 1.0000000 1.0000000
## [232] 1.0000000 1.0000000 1.0000000 0.9731922 1.0000000 1.0000000 1.0000000
```

[illegible]

```
p.adjust(all_pvals$p_val[all_pvals$p_val < 0.05], method = "BH")
```

```
## [1]      NA      NA      NA      NA      NA      NA
## [7]      NA      NA      NA      NA      NA      NA
## [13]     NA      NA      NA      NA 0.03935026 0.04518504
## [19]     NA      NA      NA      NA      NA      NA
## [25]     NA      NA      NA      NA      NA      NA
## [31]     NA      NA      NA      NA 0.04518504 0.02788823
## [37] 0.03573892 0.03389991 0.02788823 0.02788823 0.04518504 0.02730298
## [43] 0.04518504 0.04518504 0.04518504 0.04967879 0.03475609 0.04518504
## [49] 0.03475609 0.03935026 0.04518504      NA      NA      NA
## [55]      NA      NA      NA      NA      NA      NA
## [61]      NA      NA      NA 0.04518504 0.04518504 0.04518504
## [67] 0.04967879 0.03573892 0.04518504 0.03573892      NA      NA
## [73]      NA      NA      NA      NA      NA      NA
## [79]      NA      NA      NA      NA      NA      NA
## [85]      NA      NA      NA      NA      NA      NA
## [91]      NA      NA      NA      NA      NA
```