Rays R&D

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```
data <- read_csv("/Users/kenji/Documents/GitHub/Rays R&D/rays_questionnaire_dataset_2022.csv")
## Rows: 200 Columns: 3
## Delimiter: ","
## dbl (3): predictor_A, predictor_B, target_y
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
sum(is.na(data$predictor_A))
## [1] 9
summary(lm(target y ~ predictor A, data = data))
##
## Call:
## lm(formula = target_y ~ predictor_A, data = data)
## Residuals:
              1Q Median
                             3Q
      Min
                                   Max
## -50.890 -5.699 0.140 4.616 62.657
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
                          1.049 -10.253
## (Intercept) -10.752
                                         <2e-16 ***
                          1.006 1.695
                                         0.0918 .
## predictor_A 1.705
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 13.99 on 176 degrees of freedom
    (22 observations deleted due to missingness)
## Multiple R-squared: 0.01607,
                                 Adjusted R-squared: 0.01048
## F-statistic: 2.874 on 1 and 176 DF, p-value: 0.09179
summary(lm(target_y ~ predictor_A + predictor_B, data = data))
```

```
##
## Call:
## lm(formula = target_y ~ predictor_A + predictor_B, data = data)
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                         Max
## -47.597 -5.644
                    0.008
                              4.395 60.238
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -10.7136
                             1.1347 -9.442
                                               <2e-16 ***
                 1.7047
                             1.0466
                                       1.629
                                                0.105
## predictor_A
                                                0.305
## predictor_B
                 0.3679
                             0.3578
                                       1.028
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 14.11 on 163 degrees of freedom
     (34 observations deleted due to missingness)
## Multiple R-squared: 0.02188,
                                    Adjusted R-squared: 0.009883
## F-statistic: 1.823 on 2 and 163 DF, p-value: 0.1647
y = -10.7136 + 1.7047 \times A + 0.3679 \times B
result_1 <- -10.7136 + 1.7047 * 1.2 + 0.3679 * -2.1
result_1
## [1] -9.44055
To get Pr(>|t|) we need to get the t-value and p-value for each variable. To get the p-value we do:
## p-value = p-value = 2 * pt(abs(t value), residual df, lower.tail = FALSE)
P_1 \leftarrow 2 * pt(abs(0.377), 995, lower.tail = FALSE)
P_2 \leftarrow 2 * pt(abs(5.888), 995, lower.tail = FALSE)
P_3 \leftarrow 2 * pt(abs(9.923), 995, lower.tail = FALSE)
P_4 \leftarrow 2 * pt(abs(3.345), 995, lower.tail = FALSE)
P_1
## [1] 0.706254
P_2
## [1] 5.340641e-09
P 3
## [1] 3.409977e-22
```

```
P_4
## [1] 0.0008535592
\#Winston\_Walker \leftarrow P(H2/E) = [P(E/H2) \ x \ P(H2)] \ / \ [P(E/H1) \ x \ P(H1) \ + \ P(E/H2) \ x \ P(H2)]
#P(H1) = 0.5
#P(H2) = 0.5
\# P(E/H1) = 0.29
\# P(E/H2) = 0.35
Winston_Walker <- (0.35 * 0.5)/(0.29*0.5 + 0.35*0.5)*100
Winston_Walker
## [1] 54.6875
#Confidence Interval
S.E_mean <- 10/sqrt(10)
S.E_{mean}
## [1] 3.162278
alpha = 0.05
degrees.freedom = 10 - 1
t.score = qt(p = alpha/2, df = degrees.freedom, lower.tail = F)
print(t.score)
## [1] 2.262157
margin.error <- t.score * S.E_mean</pre>
margin.error
## [1] 7.153569
sample.mean <- 95</pre>
lower.bound <- sample.mean - margin.error</pre>
upper.bound <- sample.mean + margin.error</pre>
print(c(lower.bound,upper.bound))
```

[1] 87.84643 102.15357