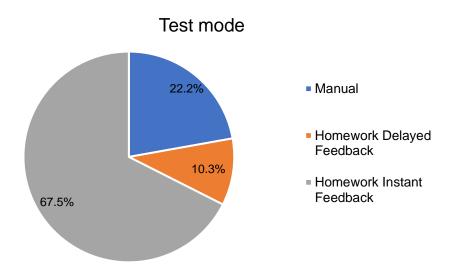
### Analysis based on report-v4.csv file from Paul's email of 13.09.2022

I excluded the records coming from Paul's testing account (school name: Fradgley High, School ID: 5) leaving 212,258 observations with 18 variables. On such a cleaned file, I tried to find regular users of the platform based on Julia's division.

### Characteristics of the entire participants' sample:

N = 2,555

- <u>Number of questions:</u> During the entire period of usage of the GCSE Prepper, an average student attempted to responded to about **67.75** (SD = 107.32) unique questions. When I included all the questions students tried to answer, including the repetition of questions in subsequent sessions, they practiced about **83.07** (SD = 225.98) questions.
- <u>Testing Mode:</u>
   The percentages of questions practiced in each mode:



• <u>Time Spent with the GSCE Prepper:</u> An average period of usage of the GCSE Prepper was about **12 weeks** – the time between the first log in and the last log in. However, when I considered weeks of active usage of the platform – weeks at which a given participant tried to answer at least one question – that was on average **4 active weeks** per student. In total, students spent approximately **119 minutes** (about 2 hours) on trying to answer questions (including repetitions), and **246 minutes** (about 4 hours) to read feedback.

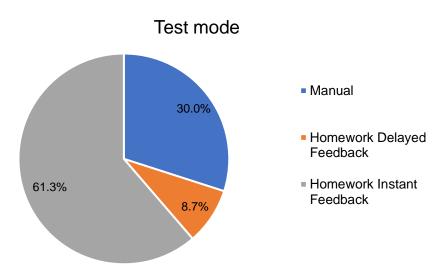
### Group A – students who are actively and regularly engaged with the platform

N = 566

### **CONDITIONS**

Necessary conditions:

- 1. 8 weeks period of usage of the GCSE Prepper (week of last log in minus week of first log in)
- 2. Practiced at least 50 unique questions (not included reptations)
  Additional conditions (it is enough to meet one of them):
- 3. During the period of usage of the platform, they had at least 50% active weeks (active week they did some activity with the program that week)
- 4. Practiced with at least 100 unique questions
- Number of questions: An average student in Group A attempted to respond to about 196.50 (SD = 167.35) unique questions, and 256.40 (SD = 434.25) questions with repetitions.
- <u>Testing Mode:</u>
- The percentages of questions practiced in each mode:



<u>Time Spent with the GSCE Prepper:</u> The average period of usage of the GCSE
Prepper was about 25 weeks, including about 10 active weeks. In total, in Group A,
students spent about 5 hours 40 minutes trying to answer questions (including
repetitions), and about 9 hours 20 minutes to read feedback.

### Group B – students who accessed the platform once (and did not use it thereafter)

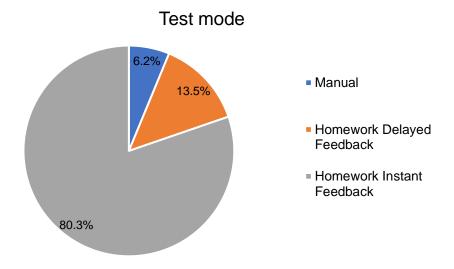
N = 681

### **CONDITION**

Students who used the GSCE Prepper for only one day.

- Number of questions: In Group B, students attempted to respond about 9.41 (SD = 9.47) unique questions, and 9.78 (SD = 10.29) questions including repetitions.
- Testing Mode:

The percentages of questions practiced in each mode:

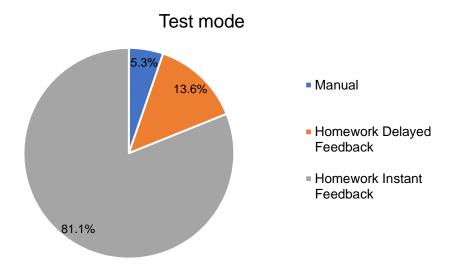


• <u>Time Spent with the GSCE Prepper:</u> In Group B, students spent about **15 min** trying to answer questions (including repetitions), and about **2 minutes** to read feedback.

### Group C – students who use the platform sporadically

N = 1,308

- Number of questions: In Group C, on average, students attempted to respond about **42.42** (*SD* = 27.55) unique questions, and **46.22** (*SD* = 32.34) questions including repetitions (for comparison, Group A: 196.50 and 256.40, respectively).
- Testing Mode:
- The percentages of questions practiced in each mode:



• <u>Time Spending with the GSCE Prepper:</u> The average period of usage of the GCSE Prepper was about **12 weeks**, including about **3 active weeks** (Group A: 25 weeks, 10 active weeks). In Group C, participants spent in total about **1 hours 17 minutes** on trying to answer questions (including repetitions), and in total about **3 hours 56 minutes** to read feedback. (Group A: 5 hours 40 answering questions, 9 hours 20 minutes for feedback).

Group D – students who never accessed the platform (despite being invited to). *No data so far.* 

### Additional Information

All	Min.	Max.	Median	Mean	Standard Dev.
number of unique questions	1	1,633	34	67.75	107.32
number of questions including repetitions in subsequent sessions	1	8,295	38	83.07	225.98
total time spent on responding answers (sec.)	4	456,356	3,173	7,153.00	15,622.62
total time spent on reading feedback (sec.)	0	6,182,584	333	14,748.00	224,903.40
total number of days (from firts to last log in)	1	729	49	77.47	87.36
total number of weeks (from first to last log in)	1	105	8	11.88	12.47
active weeks	1	44	3	4.20	4.03
Group A					
number of unique questions	51	1633	150	196.50	167.35
number of questions including repetitions in subsequent sessions	51	8295	169	256.40	434.25
total time spent on responding answers (sec.)	1,705	456,356	15,073	20,463.00	25,056.56
total time spent on reading feedback (sec.)	0	4,457,340	1,608	33,682.00	297,587.20
total number of days (from firts to last log in)	47	729	154	171.10	90.23
total number of weeks (from first to last log in)	8	105	23	25.24	12.91
active weeks	2	44	9	9.80	4.56
Group B					
number of unique questions	1	86	7	9.41	9.47
number of questions including repetitions in subsequent sessions	1	90	7	9.78	10.29
total time spent on responding answers (sec.)	4	8,396	629	903.90	936.64
total time spent on reading feedback (sec.)	0	2,039	56	110.70	182.92
Group C	Min.	Max.	Median	Mean	Standard Dev.

number of unique questions	2	276	37	42.42	27.55
number of questions including repetitions in subsequent sessions	2	278	40	46.22	32.34
total time spent on responding answers (sec.)	40	346,328	3,374	4,647.00	10,085.48
total time spent on reading feedback (sec.)	0	6,182,584	355	14,176.00	245,542.60
total number of days (from firts to last log in)	2	655	56	76.77	67.43
total number of weeks (from first to last log in)	1	94	9	11.77	9.63
active weeks	1	13	3	3.44	1.81

### **Analysis of Questions Repeated Three Times**

Graph from Cepeda et al. (2008).

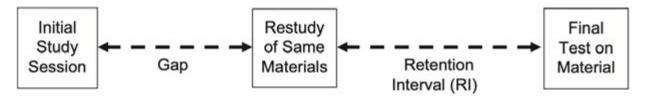


Fig. 1. Structure of a typical study of spacing effects on learning. Study episodes are separated by a varying gap, and the final study episode and test are separated by a fixed retention interval.

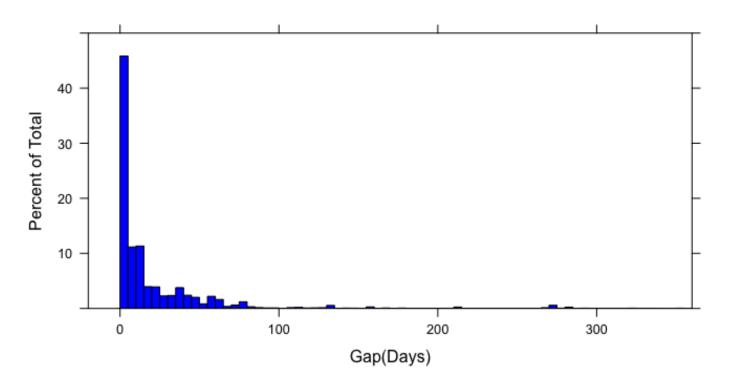
### ALL QUESTIONS INCLUDED

The total number of questions in each session and a test: 7,175

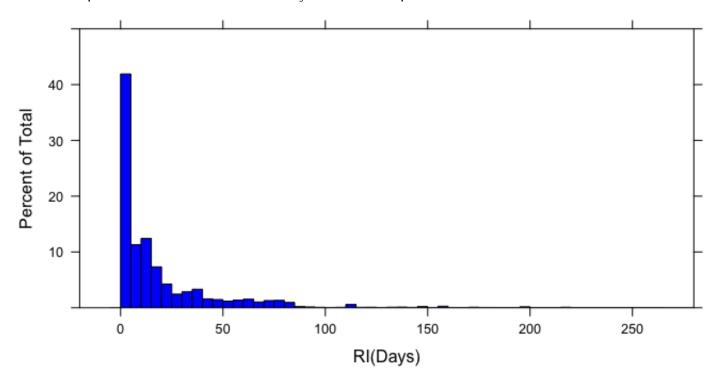
The number of unique questions: 1,833s

	Median	Mean	SD	1st Qu.	3rd Qu.	Min.	Max.
Initial Session Accuracy	100.00	64.93	41.76	20.00	100.00	0.00	100.00
Restudy Session Accuracy	100.00	70.06	40.27	50.00	100.00	0.00	100.00
Final Test Accuracy	100.00	72.16	39.81	50.00	100.00	0.00	100.00
Gap (days)	7.00	21.12	39.92	1.00	24.00	0.00	352.00
RI (days)	8.0	18.7	27.73	1.0	22.0	0.0	227.0

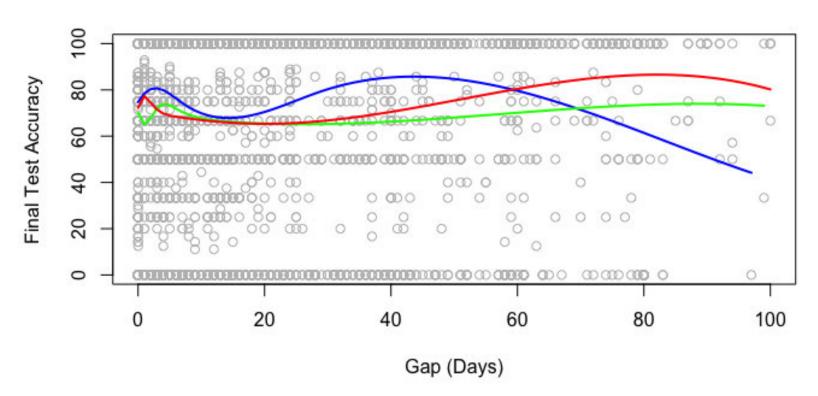
28% of questions were restudied the same day



27-28% of questions were tested the same day as their last repetition



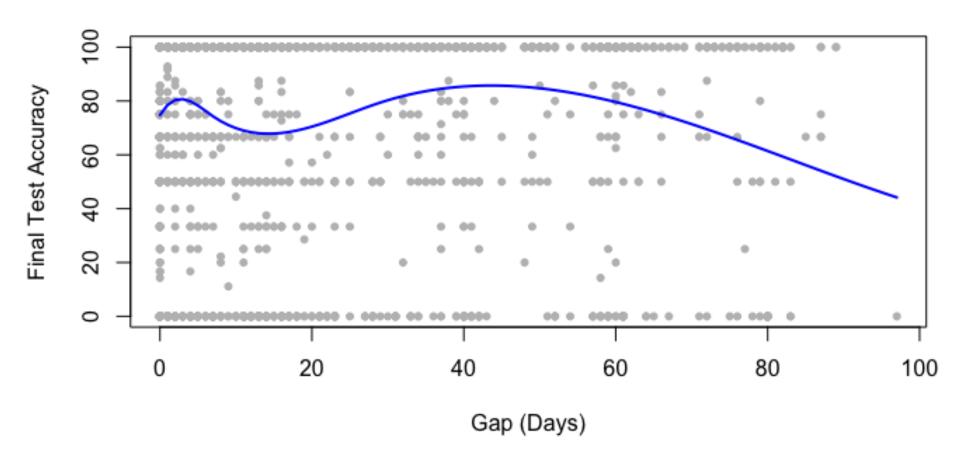
### **General Plot**



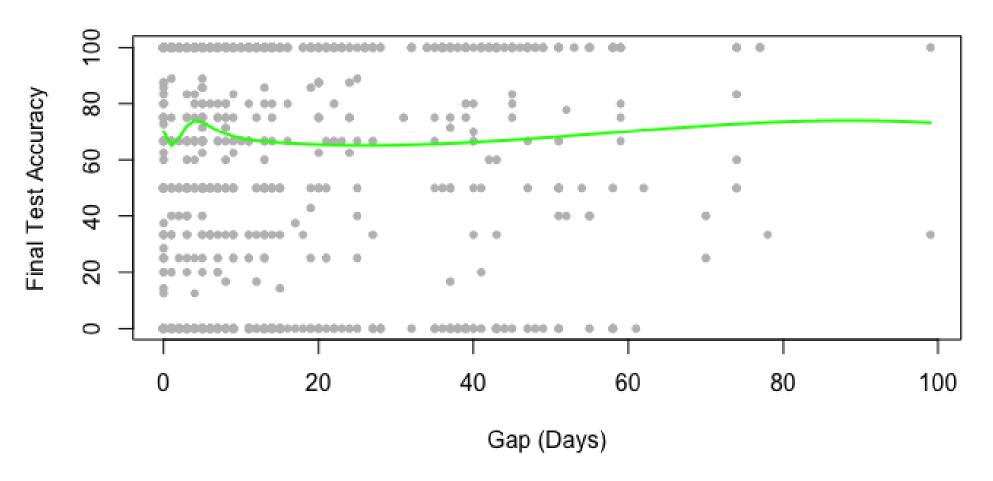
### RI:

- 0-7 days
- 8-20 days
- 21-100 days -

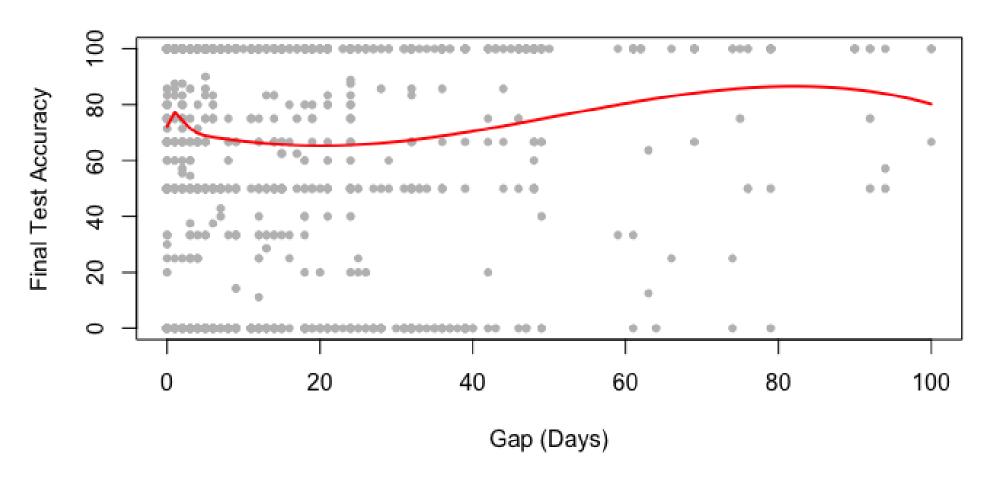
RI: 0-7 days



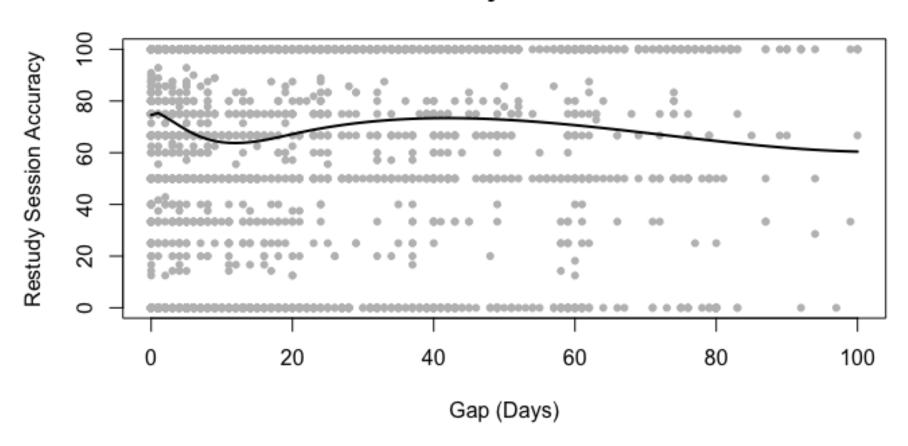
RI: 8-20 days



RI: 21-100 days



# **Restudy Session**



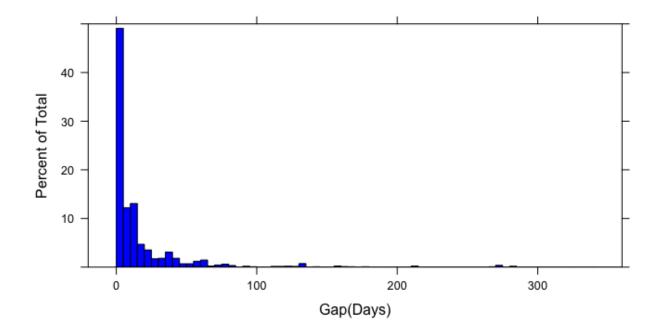
# CONTROLLED ACCURACY ON THE INITIAL STUDY SESSION (deleted questions with more than 50% accuracy on the initial study session)

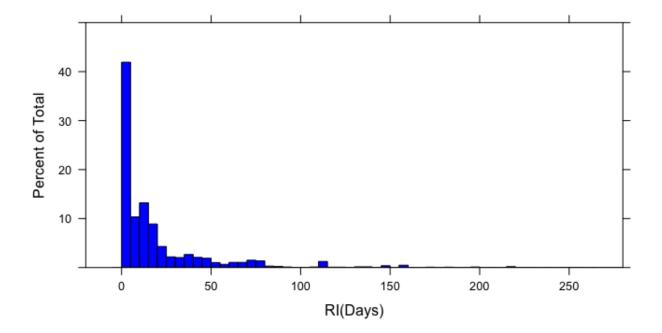
The total number of questions in each session and a test: 2,746

The number of unique questions: 1,079

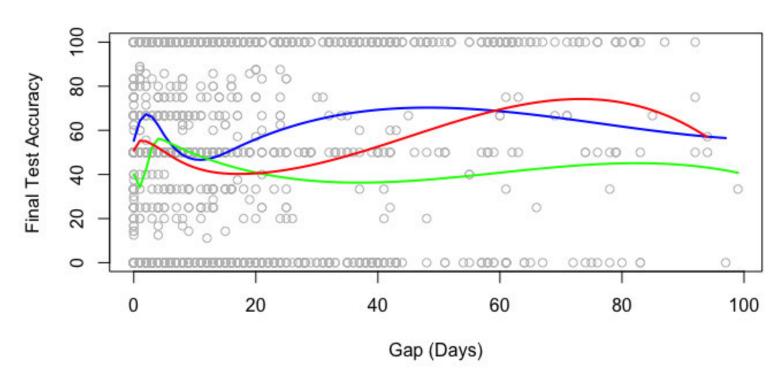
	Median	Mean	SD	1st Qu.	3rd Qu.	Min.	Max.
Initial Session Accuracy	0.00	15.96	21.81	0.00	40.00	0.00	50.00
Restudy Session Accuracy	50.00	47.85	43.27	0.00	100.00	0.00	100.00
Final Test Accuracy	50.00	51.96	44.13	0.00	100.00	0.00	100.00
Gap (days)	6.00	17.16	34.54	1.00	16.00	0.00	325.00
RI (days)	9.00	19.14	29.85	1.00	21.00	0.0	227.0

32% of questions were restudied the same day





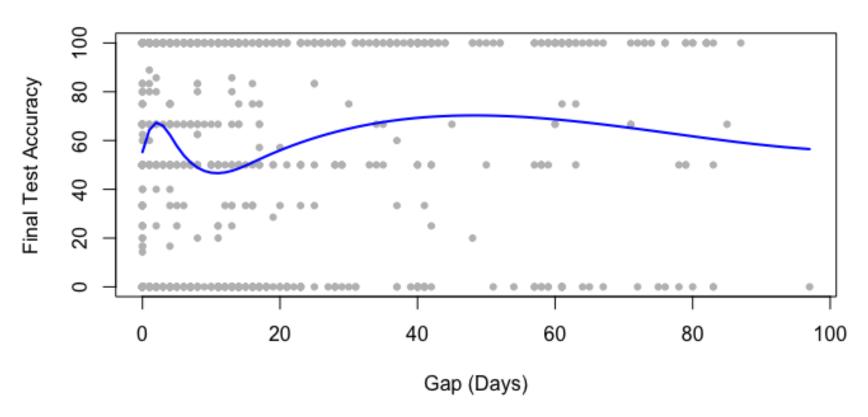
### **General Plot**



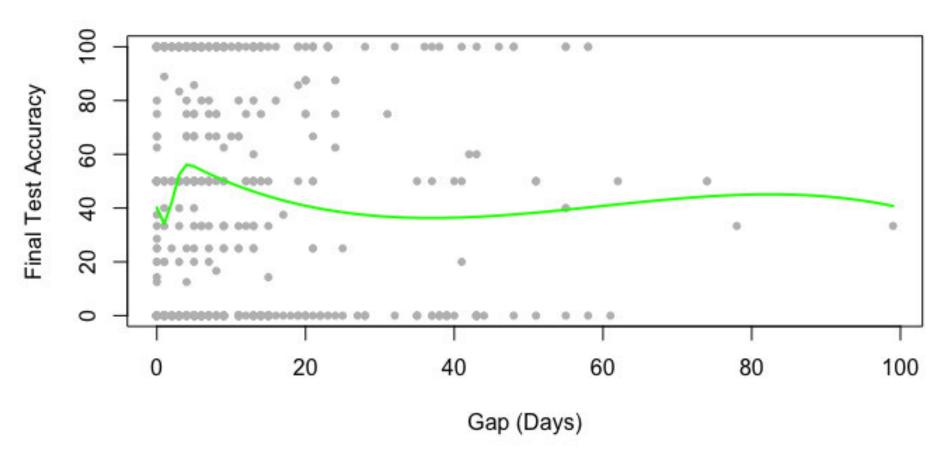
RI:

- 0-7 days
- 8-20 days
- 21-100 days -

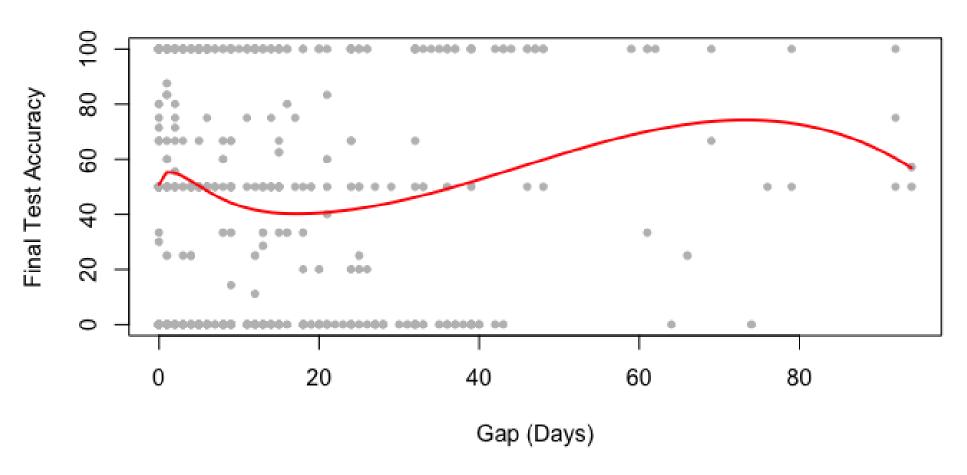
RI: 0-7 days



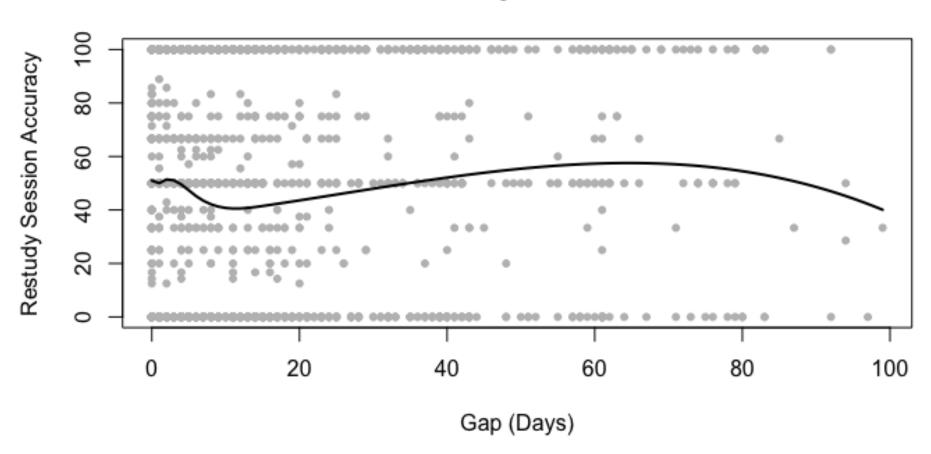
RI: 8-20 days



RI: 21-100 days



## **Restudy Session**



# CONTROLLED ACCURACY ON THE INITIAL STUDY SESSION (deleted questions with more than 50% accuracy on the initial study session) AND DELETED 0-SEC. QUESTIONS

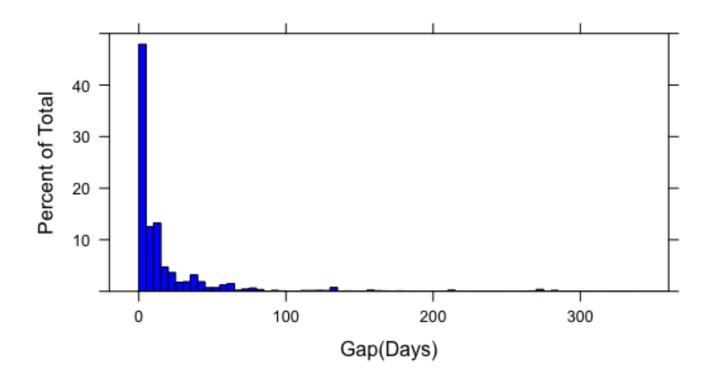
Subjects: 277

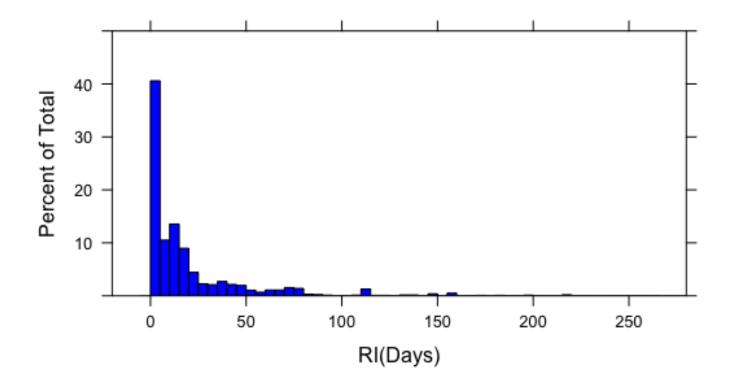
The total number of questions in each session and a test: 2,665

The number of unique questions: 1,040

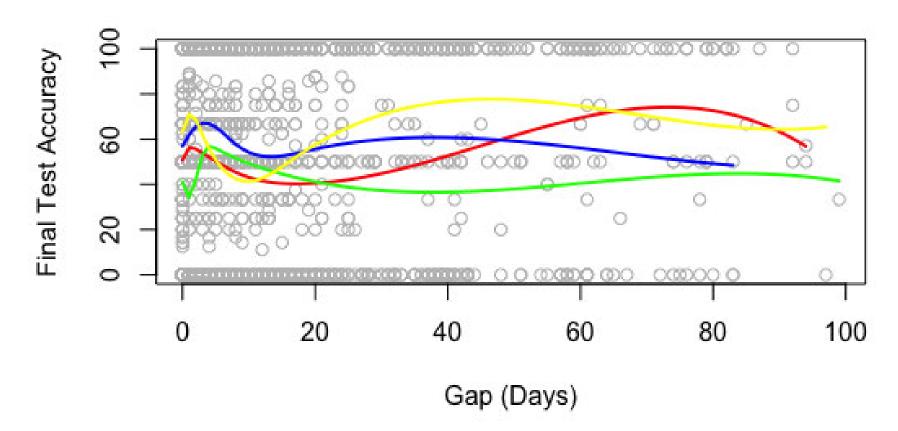
	Median	Mean	SD	1st Qu.	3rd Qu.	Min.	Max.
Initial Session Accuracy	0.00	16.40	21.94	0.00	50.00	0.00	50.00
Restudy Session Accuracy	50.00	48.15	43.13	0.00	100.00	0.00	100.00
Final Test Accuracy	50.00	53.16	43.91	0.00	100.00	0.00	100.00
Gap (days)	6.00	17.62	34.96	1.00	17.00	0.00	325.00
RI (days)	10.00	19.61	30.15	1.00	22.00	0.00	227.00

29% of questions were restudied the same day





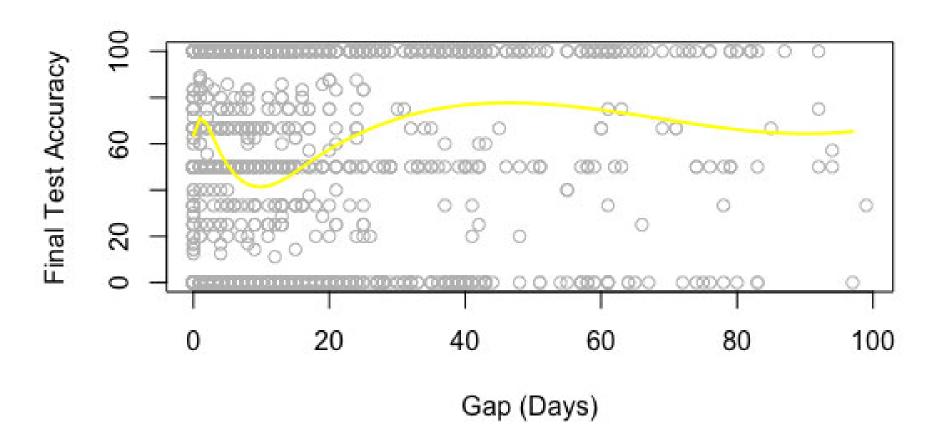
## **General Plot**



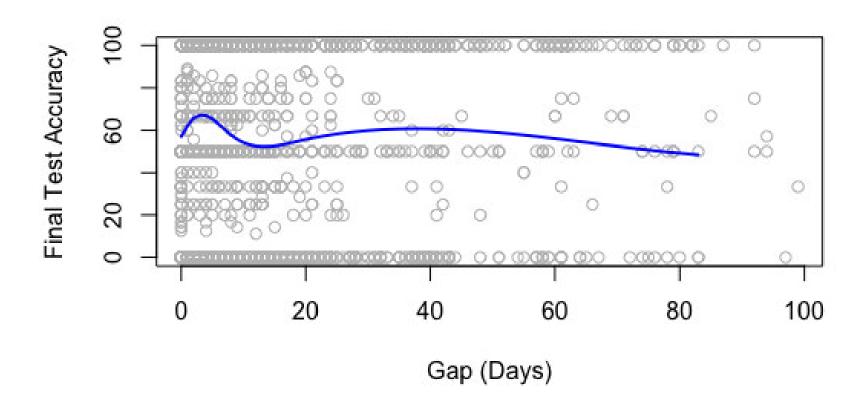
### RI:

- 1 day
- 2-7 days
- 8-20 days
- 21-100 days -

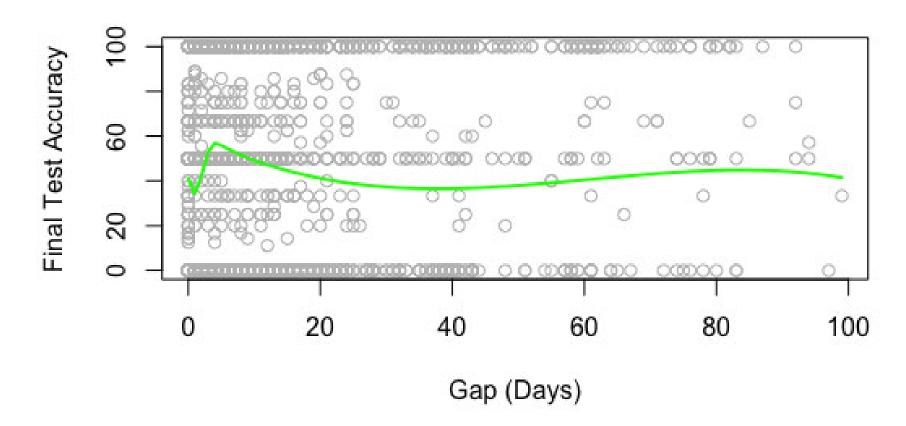
RI: 1 day



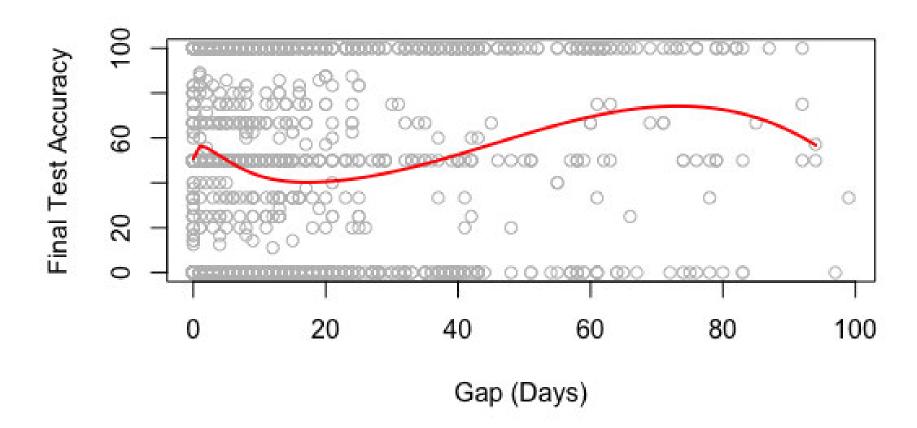
**RI: 2-7 days** 



RI: 8-20 days



RI: 21-100 days



### **ALL QUESTIONS INCLUDED**

### Intact dataset:

```
> summary(lmer(accuracy_session_3 ~ GAP*RI + (GAP*RI|Student_ID), data = report))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy_session_3 ~ GAP * RI + (GAP * RI | Student_ID)
  Data: report
REML criterion at convergence: 73551.9
Scaled residuals:
   Min
            1Q Median
                           3 Q
                                  Max
-2.9247 -0.4699 0.3402 0.6154 2.6855
Random effects:
Groups
                      Variance Std.Dev. Corr
           Name
Student ID (Intercept) 1214.1327 34.8444
           GAP
                      268.0455 16.3721 -0.01
           RI
                     267.7851 16.3641 0.00 0.00
           GAP:RI
                       0.4514 0.6719 0.00 -0.69 -0.72
Residual
                      1094.6718 33.0858
Number of obs: 7175, groups: Student_ID, 345
Fixed effects:
            Estimate Std. Error t value
(Intercept) 66.359014 3.198576 20.746
GAP
    -0.001084 1.120366 -0.001
RI
   -0.307924 1.145013 -0.269
GAP:RI
          0.007343 0.045918 0.160
Correlation of Fixed Effects:
      (Intr) GAP
                  RI
GAP -0.114
      -0.123 -0.028
GAP:RI 0.159 -0.672 -0.721
fit warnings:
Some predictor variables are on very different scales: consider rescaling
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

Logarithmic transformation:

```
> summary(lmer(accuracy session 3 ~ GAPlog*RIlog + (GAPlog*RIlog|Student ID), data = report))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy_session_3 ~ GAPlog * RIlog + (GAPlog * RIlog | Student_ID)
  Data: report
REML criterion at convergence: 71689.7
Scaled residuals:
            10 Median
   Min
                          30
                                 Max
-2.7029 -0.5306 0.3609 0.6240 2.5810
Random effects:
Groups
                     Variance Std.Dev. Corr
          Name
Student_ID (Intercept) 896.633 29.944
                   22.267 4.719 -0.61
          GAPlog
                 29.003 5.385 -0.37 0.89
          RIlog
          GAPlog:RIlog 1.184 1.088 -0.03 -0.59 -0.89
                       1171.865 34.233
Residual
Number of obs: 7175, groups: Student ID, 345
Fixed effects:
            Estimate Std. Error t value
(Intercept) 67.8965 2.7905 24.331
GAPlog
         0.4839 0.7687 0.630
RIlog
      -1.6577 0.8993 -1.843
GAPlog:RIlog -0.2825 0.2766 -1.021
Correlation of Fixed Effects:
          (Intr) GAPlog RIlog
          -0.748
GAPlog
          -0.668 0.737
RIlog
GAPlog:RIlg 0.429 -0.691 -0.819
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

### Square-root transformation:

```
> summary(lmer(accuracy_session_3 ~ GAPsqrt*RIsqrt + (GAPsqrt*RIsqrt|Student_ID), data = report))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAPsqrt * RIsqrt + (GAPsqrt * RIsqrt | Student ID)
  Data: report
REML criterion at convergence: 71693.2
Scaled residuals:
   Min
            10 Median
                           30
                                  Max
-2.7235 -0.5291 0.3469 0.6313 2.4973
Random effects:
Groups
                        Variance Std.Dev. Corr
           Name
Student_ID (Intercept) 8.559e+02 29.2557
                      6.049e+00 2.4594 -0.67
           GAPsqrt
           RIsgrt
                         6.892e+00 2.6252 -0.34 0.92
           GAPsqrt:RIsqrt 3.739e-02 0.1934 -0.19 -0.57 -0.85
Residual
                         1.172e+03 34.2371
Number of obs: 7175, groups: Student_ID, 345
Fixed effects:
              Estimate Std. Error t value
(Intercept) 67.79615 2.46276 27.529
GAPsart
            0.11869 0.36165 0.328
RIsart
             -1.05019 0.39667 -2.647
GAPsqrt:RIsqrt -0.05118
                        0.06106 -0.838
Correlation of Fixed Effects:
           (Intr) GAPsqr RIsqrt
          -0.726
GAPsqrt
          -0.621 0.768
RIsgrt
GAPsqrt:RIs 0.307 -0.639 -0.684
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

```
Intact dataset:
```

```
> summary(lmer(accuracy session 3 ~ GAP*RI + GAPx2 + (GAP*RI|Student ID), data = report))
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy_session_3 ~ GAP * RI + GAPx2 + (GAP * RI | Student_ID)
   Data: report
REML criterion at convergence: 73579.7
Scaled residuals:
   Min
            10 Median
                            30
                                   Max
-2.9315 -0.4695 0.3419 0.6217 2.6873
Random effects:
 Groups
           Name
                       Variance Std.Dev. Corr
 Student ID (Intercept) 1226.1777 35.0168
           GAP
                       274.6257 16.5718 -0.02
           RI
                       277.5433 16.6596 -0.05 0.02
           GAP:RI
                         0.6268 0.7917 0.04 -0.66 -0.76
 Residual
                       1094.8265 33.0882
Number of obs: 7175, groups: Student_ID, 345
Fixed effects:
             Estimate Std. Error t value
(Intercept) 65.8979553 3.2099339 20.529
GAP
            0.0264930 1.1353488 0.023
          -0.1200095 1.1588361 -0.104
RI
GAP×2
         0.0001297 0.0001490 0.871
GAP:RI
            0.0010173 0.0541911 0.019
Correlation of Fixed Effects:
      (Intr) GAP
                    RI
                           GAPx2
     -0.120
GAP
      -0.139 0.006
RI
GAPx2 0.042 -0.012 -0.001
GAP:RI 0.170 -0.650 -0.763 -0.007
fit warnings:
Some predictor variables are on very different scales: consider rescaling
optimizer (nloptwrap) convergence code: 0 (OK)
unable to evaluate scaled gradient
Model failed to converge: degenerate Hessian with 4 negative eigenvalues
```

### Logarithmic transformation:

```
> summary(lmer(accuracy session 3 ~ GAPlog*RIlog + GAPlogx2 +(GAPlog*RIlog|Student_ID), data = repor
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAPlog * RIlog + GAPlogx2 + (GAPlog * RIlog |
                                                                             Student ID)
  Data: report
REML criterion at convergence: 71685.8
Scaled residuals:
   Min
            1Q Median
                           3Q
                                  Max
-2.7024 -0.5357 0.3650 0.6238 2.5977
Random effects:
Groups
                       Variance Std.Dev. Corr
           Name
Student_ID (Intercept) 894.261 29.904
           GAPlog
                        21.583 4.646 -0.60
           RIlog
                         28.423 5.331 -0.39 0.90
           GAPlog:RIlog 1.122 1.059 -0.05 -0.55 -0.86
Residual
                       1170.671 34.215
Number of obs: 7175, groups: Student_ID, 345
Fixed effects:
            Estimate Std. Error t value
(Intercept) 67.1812
                        2.8036 23.962
             2.6750
GAPlog
                        1.2553 2.131
RIlog
            -1.8434
                        0.9027 -2.042
GAPlogx2
            -0.5436
                        0.2430 -2.238
GAPlog:RIlog -0.2735
                        0.2776 -0.985
Correlation of Fixed Effects:
           (Intr) GAPlog RIlog GAPlg2
GAPlog
           -0.543
RIlog
           -0.646 0.353
           0.122 -0.794 0.111
GAPlogx2
GAPlog:RIlg 0.409 -0.387 -0.809 -0.030
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

### Square-root transformation:

```
> summary(lmer(accuracy session 3 ~ GAPsqrt*RIsqrt + GAPsqrtx2 + (GAPsqrt*RIsqrt|Student ID). data = r
eport))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy_session_3 ~ GAPsqrt * RIsqrt + GAPsqrtx2 + (GAPsqrt *
   RIsgrt | Student ID)
  Data: report
REML criterion at convergence: 71696.6
Scaled residuals:
   Min
            10 Median
                           30
-2.7226 -0.5351 0.3488 0.6352 2.5091
Random effects:
Groups
                         Variance Std.Dev. Corr
           Name
Student_ID (Intercept)
                        881.1180 29.6836
           GAPsqrt
                            6.4455 2.5388 -0.68
                            6.5991 2.5689 -0.35 0.92
           RIsart
           GAPsqrt:RIsqrt 0.0319 0.1786 -0.17 -0.60 -0.86
Residual
                         1171.8389 34.2321
Number of obs: 7175, groups: Student_ID, 345
Fixed effects:
              Estimate Std. Error t value
              67.24751
                        2.49706 26.931
(Intercept)
GAPsqrt
               0.51976
                        0.46246 1.124
RIsart
              -1.07200
                        0.38851 -2.759
GAPsqrtx2
              -0.04249
                       0.03011 -1.412
GAPsqrt:RIsqrt -0.04360 0.05998 -0.727
Correlation of Fixed Effects:
           (Intr) GAPsqr RIsqrt GAPsq2
GAPsqrt
           -0.646
RIsgrt
           -0.601 0.540
GAPsqrtx2 0.112 -0.614 0.136
GAPsqrt:RIs 0.288 -0.424 -0.681 -0.137
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

# CONTROLLED ACCURACY ON THE INITIAL STUDY SESSION (deleted questions with more than 50% accuracy on the initial study session) AND DELETED 0-SEC, OUESTIONS

### **Intact dataset:**

```
> summary(lmer(accuracy session 3 ~ GAP*RI + (GAP*RI|Student ID), data = report))
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAP * RI + (GAP * RI | Student ID)
  Data: report
REML criterion at convergence: 28634.7
Scaled residuals:
    Min
             10 Median
                               3Q
                                      Max
-2.38948 -0.68397 0.00413 0.83093 2.56955
Random effects:
Groups
                 Variance Std.Dev. Corr
           Name
Student ID (Intercept) 1231.0207 35.0859
           GAP
                      236.0023 15.3624 -0.07
           RI
                      288.4559 16.9840 -0.09 0.00
           GAP:RI
                      0.7067 0.8407 0.10 -0.56 -0.82
Residual
                      1248.9539 35.3405
Number of obs: 2665, groups: Student_ID, 277
Fixed effects:
           Estimate Std. Error t value
(Intercept) 53.79296 4.03532 13.331
GAP
         0.46939 1.21759 0.386
          -0.14557 1.32518 -0.110
RI
GAP:RI
       -0.01366 0.06596 -0.207
Correlation of Fixed Effects:
      (Intr) GAP
                   RI
GAP
    -0.181
      -0.176 0.004
GAP:RI 0.227 -0.571 -0.822
```

```
Logarithmic transformation:
```

```
> summary(lmer(accuracy session 3 ~ GAPlog*RIlog + (GAPlog*RIlog|Student ID), data = repor
t))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAPlog * RIlog + (GAPlog * RIlog | Student ID)
  Data: report
REML criterion at convergence: 27274.3
Scaled residuals:
    Min
             10 Median
                             30
                                     Max
-2.26765 -0.83137 0.04067 0.82925 2.17881
Random effects:
Groups
                 Variance Std.Dev. Corr
          Name
Student ID (Intercept) 840.464 28.991
          GAPlog
                    18.011 4.244 -0.51
          RIlog 51.877 7.203 -0.61 0.95
          GAPlog:RIlog 1.973 1.405 -0.08 -0.70 -0.73
                     1455.332 38.149
Residual
Number of obs: 2665, groups: Student ID, 277
Fixed effects:
           Estimate Std. Error t value
(Intercept) 52.8630 3.7895 13.950
GAPlog 1.5552 1.1753 1.323
      -1.6038 1.3713 -1.170
RIlog
GAPlog:RIlog -0.7367 0.4667 -1.579
Correlation of Fixed Effects:
          (Intr) GAPlog RIlog
GAPlog
          -0.705
RIlog
        -0.768 0.681
GAPlog:RIlg 0.476 -0.774 -0.770
```

### Square-root transformation:

```
> summary(lmer(accuracy_session_3 ~ GAPsqrt*RIsqrt + (GAPsqrt*RIsqrt|Student_ID), data = report))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAPsqrt * RIsqrt + (GAPsqrt * RIsqrt | Student ID)
   Data: report
REML criterion at convergence: 27281.5
Scaled residuals:
    Min
              10 Median
                               3Q
                                      Max
-2.29629 -0.82675 0.03268 0.82920 2.22578
Random effects:
Groups
           Name
                      Variance Std.Dev. Corr
 Student ID (Intercept) 7.669e+02 27.6932
           GAPsqrt
                      3.459e+00 1.8597 -0.60
           RIsart
                        1.091e+01 3.3035 -0.58 0.98
           GAPsgrt:RIsgrt 6.645e-02 0.2578 -0.28 -0.53 -0.61
 Residual
                         1.458e+03 38.1791
Number of obs: 2665, groups: Student_ID, 277
Fixed effects:
              Estimate Std. Error t value
(Intercept) 52.6882 3.1678 16.632
GAPsqrt
           0.5710 0.5183 1.102
RIsart
              -1.0195 0.6107 -1.670
GAPsqrt:RIsqrt -0.1577 0.1111 -1.419
Correlation of Fixed Effects:
           (Intr) GAPsqr RIsqrt
GAPsqrt -0.654
RIsart
        -0.723 0.638
GAPsqrt:RIs 0.361 -0.728 -0.652
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

#### **Intact dataset:**

```
> summary(lmer(accuracy_session_3 ~ GAP*RI + GAPx2 + (GAP*RI|Student_ID), data = report))
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAP * RI + GAPx2 + (GAP * RI | Student ID)
  Data: report
REML criterion at convergence: 28690.5
Scaled residuals:
    Min
              10 Median
                               3Q
                                       Max
-2.39018 -0.68324 0.00463 0.82209 2.57298
Random effects:
Groups
           Name
                      Variance Std.Dev. Corr
Student_ID (Intercept) 1219.561 34.9222
           GAP
                       349.818 18.7034 -0.07
           RI
                       250.655 15.8321 0.00 0.01
           GAP:RI
                        0.744 0.8625 0.03 -0.65 -0.76
Residual
                      1247.633 35.3219
Number of obs: 2665, groups: Student ID, 277
Fixed effects:
             Estimate Std. Error t value
(Intercept) 53.8134008 4.0510841 13.284
GAP
            0.5170211 1.4510928 0.356
RI
           -0.1875454 1.2565781 -0.149
GAPx2
         0.0002195 0.0003230 0.680
GAP:RI
          -0.0145829 0.0673886 -0.216
Correlation of Fixed Effects:
      (Intr) GAP
                   RI
                          GAPx2
GAP
     -0.166
      -0.141 0.003
RI
GAPx2 0.071 -0.021 -0.007
GAP:RI 0.192 -0.639 -0.768 -0.007
```

#### Logarithmic transformation:

```
> summary(lmer(accuracy_session_3 ~ GAPlog*RIlog + GAPlogx2 +(GAPlog*RIlog|Student_ID), data = report))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy_session_3 ~ GAPlog * RIlog + GAPlogx2 + (GAPlog * RIlog |
                                                                            Student_ID)
  Data: report
REML criterion at convergence: 27271.8
Scaled residuals:
    Min
              10 Median
                               30
                                      Max
-2.23107 -0.84025 0.02583 0.83012 2.20002
Random effects:
Groups
           Name
                       Variance Std.Dev. Corr
Student_ID (Intercept) 845.973 29.086
           GAPlog
                         20.188 4.493 -0.47
           RIlog
                         53.165 7.291 -0.63 0.94
           GAPlog:RIlog 2.111 1.453 -0.05 -0.71 -0.72
Residual
                       1451.676 38.101
Number of obs: 2665, groups: Student ID, 277
Fixed effects:
            Estimate Std. Error t value
(Intercept) 51.8660
                        3.8721 13.395
GAPlog
             4.3912
                        2.1843 2.010
RIlog
            -1.9310
                      1.3919 -1.387
            -0.7135
GAPlogx2
                     0.4451 -1.603
GAPlog:RIlog -0.6720
                      0.4743 -1.417
Correlation of Fixed Effects:
           (Intr) GAPlog RIlog GAPlg2
GAPlog
           -0.531
RIlog
           -0.727 O.252
GAPlogx2
          0.185 -0.838 0.138
GAPlog:RIlg 0.456 -0.369 -0.770 -0.059
```

#### Square-root transformation:

```
> summary(lmer(accuracy_session_3 ~ GAPsqrt*RIsqrt + GAPsqrtx2 +(GAPsqrt*RIsqrt|Student_ID), data = report))
boundary (singular) fit: see help('isSingular')
Linear mixed model fit by REML ['lmerMod']
Formula: accuracy session 3 ~ GAPsqrt * RIsqrt + GAPsqrtx2 + (GAPsqrt *
                                                                     RIsgrt | Student ID)
  Data: report
REML criterion at convergence: 27285
Scaled residuals:
    Min
             10 Median
                                      Max
                              30
-2.28847 -0.83077 0.03255 0.83334 2.22551
Random effects:
Groups
           Name
                       Variance Std.Dev. Corr
Student ID (Intercept) 7.829e+02 27.9798
           GAPsqrt
                       3.668e+00 1.9151 -0.61
                       1.131e+01 3.3633 -0.59 0.98
           RIsgrt
           GAPsqrt:RIsqrt 6.558e-02 0.2561 -0.23 -0.57 -0.64
Residual
                        1.457e+03 38.1690
Number of obs: 2665, groups: Student ID, 277
Fixed effects:
             Estimate Std. Error t value
                       3.25331 16.094
(Intercept) 52.35890
            0.86461 0.77826 1.111
GAPsqrt
           -1.05637 0.61986 -1.704
RIsgrt
GAPsqrtx2
           -0.03162 0.05892 -0.537
GAPsqrt:RIsqrt -0.14841 0.11247 -1.320
Correlation of Fixed Effects:
           (Intr) GAPsqr RIsqrt GAPsq2
GAPsqrt -0.578
RIsgrt -0.677 0.316
GAPsqrtx2 0.198 -0.743 0.148
GAPsqrt:RIs 0.322 -0.358 -0.662 -0.168
```

#### The most promising model on all questions:

```
> summary(lmer(accuracy session 3 new ~ GAPlog*RIlog + GAPlogx2 + RIlogx2+ (1|Student ID), data = report))
Linear mixed model fit by REML. t-tests use Satterthwaite's method ['lmerModLmerTest']
Formula: accuracy session 3 new ~ GAPlog * RIlog + GAPlogx2 + RIlogx2 +
                                                                          (1 | Student ID)
  Data: report
REML criterion at convergence: 5736.4
Scaled residuals:
   Min
            10 Median
                           30
                                  Max
-2.7069 -0.5345 0.3953 0.6179 2.7716
Random effects:
 Groups
           Name
                       Variance Std.Dev.
 Student ID (Intercept) 0.06339 0.2518
 Residual
                       0.11991 0.3463
Number of obs: 7175, groups: Student ID, 345
Fixed effects:
              Estimate Std. Error
                                         df t value Pr(>|t|)
(Intercept) 6.619e-01 2.066e-02 7.274e+02 32.039 < 2e-16 ***
GAPlog
            2.302e-02 1.061e-02 7.061e+03 2.168 0.03016 *
RIlog
            1.080e-02 1.218e-02 6.946e+03 0.887 0.37517
GAPlogx2
            -5.798e-03 2.247e-03 7.107e+03 -2.581 0.00988 **
RIlogx2
           -6.163e-03 2.643e-03 7.010e+03 -2.332 0.01971 *
GAPlog:RIlog -2.669e-03 1.979e-03 7.158e+03 -1.348 0.17761
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Correlation of Fixed Effects:
           (Intr) GAPlog RIlog GAPlg2 RIlgx2
GAPlog
           -0.349
RIlog
          -0.367 -0.022
GAPlogx2
         0.167 -0.886 0.117
RIlogx2
            0.200 0.067 -0.899 -0.031
GAPlog:RIlg 0.298 -0.232 -0.313 -0.128 -0.020
```

# CONTROLLED ACCURACY ON THE INITIAL STUDY SESSION (deleted questions with more than 50% accuracy on the initial study session) AND DELETED 0-SEC. QUESTIONS

```
> summary(lmer(accuracy session 3 new ~ GAPlog*RIlog + GAPlogx2 + RIlogx2+ (1|Student ID), data = report))
Linear mixed model fit by REML. t-tests use Satterthwaite's method [
lmerModLmerTest1
Formula: accuracy session 3 new ~ GAPlog * RIlog + GAPlogx2 + RIlogx2 +
   (1 | Student ID)
  Data: report
REML criterion at convergence: 2809.9
Scaled residuals:
    Min
              10 Median
                               30
                                       Max
-2.25484 -0.82954 0.04115 0.85902 2.67513
Random effects:
Groups
           Name
                      Variance Std.Dev.
Student ID (Intercept) 0.04926 0.2219
Residual
                      0.14940 0.3865
Number of obs: 2665, groups: Student ID, 277
Fixed effects:
              Estimate Std. Error
                                        df t value Pr(>|t|)
(Intercept) 5.267e-01 2.983e-02 8.681e+02 17.655 <2e-16 ***
GAPlog
           3.478e-02 1.991e-02 2.438e+03 1.747
                                                     0.0808 .
RIlog
          -1.493e-02 2.228e-02 2.304e+03 -0.670
                                                     0.5029
GAPlogx2 -5.602e-03 4.259e-03 2.464e+03 -1.315
                                                     0.1885
RIlogx2
         -9.972e-04 4.829e-03 2.283e+03 -0.206 0.8364
GAPlog:RIlog -6.656e-03 3.748e-03 2.595e+03 -1.776 0.0759 .
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Correlation of Fixed Effects:
           (Intr) GAPlog RIlog GAPlg2 RIlgx2
GAPlog
          -0.439
RIlog
         -0.450 -0.068
GAPlogx2 0.207 -0.881 0.160
RIlogx2
           0.238 0.115 -0.902 -0.084
GAPlog:RIlg 0.383 -0.259 -0.318 -0.105 0.001
```

### **Exploration of the additional data:**

### N = 240

	GCSE results
the number of questions	0.43
the number of questions with repetitions	0.44
time spent on answering questions	0.44
time spent on answering questions - adjusted	0.50
time spent on reading feedback	0.22
the number of "active" weeks	0.45
the period of using GCSE prepper	0.43
accuracy of answered questions	0.32

*p* < .001

#### > rcorr(correlations) exam accuracy period active weeks feedback time time ad rep questions questions 1.00 0.32 0.43 0.45 0.22 0.44 0.50 0.44 0.43 exam 0.32 1.00 0.05 -0.03 0.10 0.16 0.17 0.03 0.01 accuracy period 0.43 0.05 1.00 0.75 0.45 0.56 0.61 0.66 0.67 active\_weeks 0.45 -0.03 0.75 1.00 0.41 0.68 0.78 0.83 0.84 0.66 0.57 feedback time 0.22 0.10 0.45 0.41 1.00 0.76 0.57 0.44 0.16 0.56 0.68 0.76 1.00 0.94 0.77 0.77 time time ad 0.50 0.17 0.61 0.78 0.66 0.94 1.00 0.90 0.90 rep questions 0.44 0.03 0.66 0.83 0.57 0.77 0.90 1.00 1.00 questions 0.43 0.01 0.67 0.84 0.57 0.77 0.90 1.00 1.00 n= 240

P									
	exam	accuracy	period	active_weeks	feedback_time	time	time_ad	rep_questions	questions
exam		0.0000	0.0000	0.0000	0.0006	0.0000	0.0000	0.0000	0.0000
accuracy	0.0000		0.4776	0.6383	0.1350	0.0106	0.0094	0.6992	0.8248
period	0.0000	0.4776		0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
active_weeks	0.0000	0.6383	0.0000		0.0000	0.0000	0.0000	0.0000	0.0000
feedback_time	0.0006	0.1350	0.0000	0.0000		0.0000	0.0000	0.0000	0.0000
time	0.0000	0.0106	0.0000	0.0000	0.0000		0.0000	0.0000	0.0000
time_ad	0.0000	0.0094	0.0000	0.0000	0.0000	0.0000		0.0000	0.0000
rep_questions	0.0000	0.6992	0.0000	0.0000	0.0000	0.0000	0.0000		0.0000
questions	0.0000	0.8248	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	

```
N = 240
```

**GCSE results:** M = 5.16, SD = 1.80

time spent on answering questions – adjusted (min) M = 45.01, SD = 46.62

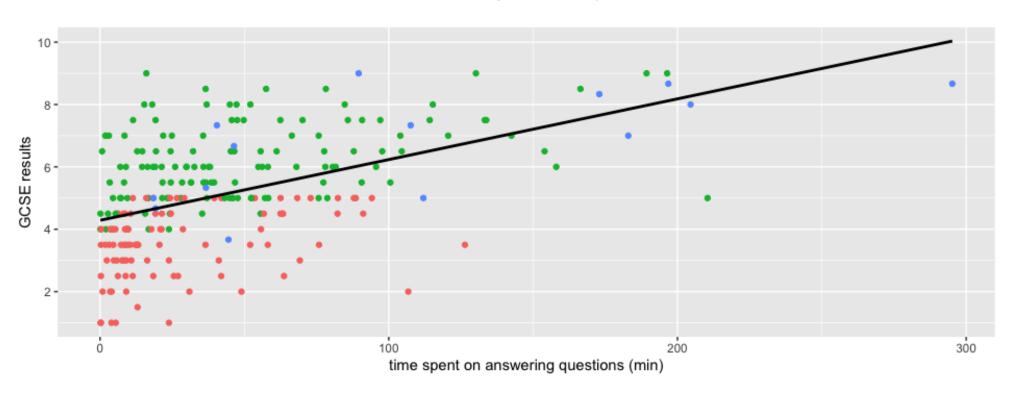
```
> exp1.lm = lm(exam ~ time ad min, data = GCSE predictions)
> summary(exp1.lm)
Call:
lm(formula = exam ~ time ad min, data = GCSE predictions)
Residuals:
    Min
            1Q Median
                           3 Q
                                  Max
-4.3654 -0.9514 -0.0321 0.9964 4.4010
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 4.286834 0.140029 30.614
                                        <2e-16 ***
time ad min 0.019471 0.002163
                                9.001
                                        <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.559 on 238 degrees of freedom
                             Adjusted R-squared: 0.2508
Multiple R-squared: 0.254,
F-statistic: 81.02 on 1 and 238 DF, p-value: < 2.2e-16
```

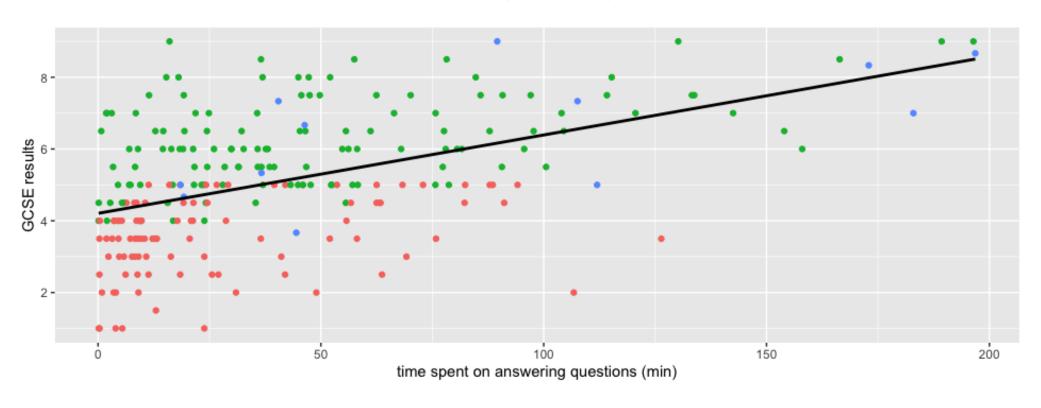
#### **Foundation**

```
> exp2.lm = lm(GCSE_Foundation$GCSE ~ GCSE_Foundation$sum_response_time_min_adjusted, data = GCSE_Foundation)
 > summary(exp2.lm)
Call:
 lm(formula = GCSE_Foundation$GCSE ~ GCSE_Foundation$sum_response_time_min_adjusted,
    data = GCSE_Foundation)
 Residuals:
    Min
             10 Median
                                    Max
                             3Q
 -2.5807 -0.7134 0.1973 0.7408 1.6581
Coefficients:
                                               Estimate Std. Error t value Pr(>|t|)
                                               3.194525 0.152195 20.990 < 2e-16 ***
 (Intercept)
GCSE_Foundation$sum_response_time_min_adjusted 0.012985 0.003763
                                                                    3.451 0.000842 ***
 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
 Residual standard error: 1.057 on 93 degrees of freedom
Multiple R-squared: 0.1135, Adjusted R-squared: 0.104
F-statistic: 11.91 on 1 and 93 DF, p-value: 0.0008422
>
```

#### Higher

```
> exp3.lm = lm(GCSE Higher$GCSE ~ GCSE Higher$sum response time min adjusted, data = GCSE Higher
> summary(exp3.lm)
Call:
lm(formula = GCSE Higher$GCSE ~ GCSE Higher$sum response time min adjusted,
    data = GCSE Higher)
Residuals:
    Min
            1Q Median
                            3 Q
                                   Max
-3.1235 -0.9069 -0.0812 0.8239 3.2633
Coefficients:
                                         Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                         5.539883 0.153466
                                                               36.10 < 2e-16 ***
GCSE Higher$sum response time min adjusted 0.012279 0.002308 5.32 4.44e-07 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.161 on 129 degrees of freedom
Multiple R-squared: 0.1799, Adjusted R-squared: 0.1736
F-statistic: 28.31 on 1 and 129 DF, p-value: 4.44e-07
```





**GCSE** results: M = 5.16, SD = 1.80

time spent on answering questions – adjusted (min), average M=0.95, SD=0.37

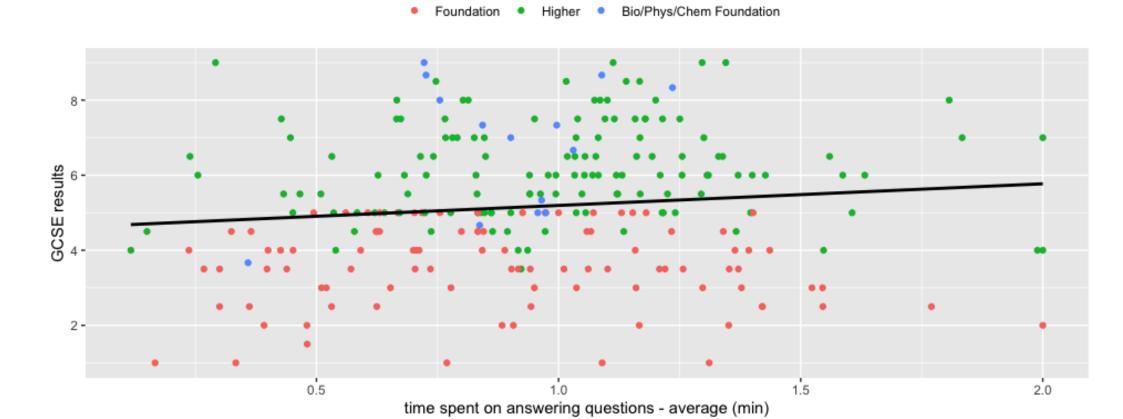
```
> exp4.lm = lm(exam ~ time ad average min, data = GCSE predictions)
> summary(exp4.lm)
Call:
lm(formula = exam ~ time ad average min, data = GCSE predictions)
Residuals:
   Min
            1Q Median
                                   Max
                            3 Q
-4.3737 -1.1573 -0.0816 1.2168 4.2148
Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
                     4.6170
                                0.3214 14.365 <2e-16 ***
(Intercept)
time ad average min 0.5772
                                0.3168 1.822
                                                0.0698 .
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.793 on 238 degrees of freedom
Multiple R-squared: 0.01375, Adjusted R-squared: 0.009607
F-statistic: 3.318 on 1 and 238 DF, p-value: 0.06976
>
```

#### **Foundation**

```
> exp5.lm = lm(GCSE Foundation$GCSE ~ GCSE Foundation$mean response time adjusted min, data = GCSE Foundation)
> summary(exp5.lm)
Call:
lm(formula = GCSE_Foundation$GCSE ~ GCSE_Foundation$mean_response_time_adjusted_min,
    data = GCSE_Foundation)
Residuals:
     Min
              1Q Median 3Q
                                        Max
-2.68435 -0.61347 -0.00657 0.92530 1.52411
Coefficients:
                                               Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                                3.7125
                                                           0.2882 12.881 <2e-16 ***
GCSE Foundation$mean response time adjusted min -0.1688
                                                           0.2987 -0.565 0.573
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.12 on 93 degrees of freedom
Multiple R-squared: 0.003421, Adjusted R-squared: -0.007295
F-statistic: 0.3192 on 1 and 93 DF, p-value: 0.5734
a discourse of the bed where
```

### Higher

```
> exp6.lm = lm(GCSE Higher$GCSE ~ GCSE Higher$mean response time adjusted min, data = GCSE Higher)
> summary(exp6.lm)
Call:
lm(formula = GCSE Higher$GCSE ~ GCSE Higher$mean response time adjusted min,
    data = GCSE_Higher)
Residuals:
    Min
             1Q Median
                            3 Q
                                   Max
-2.6313 -1.0483 -0.1685 0.9012 3.0481
Coefficients:
                                           Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                             5.8690
                                                        0.3320 17.679 <2e-16 ***
GCSE Higher$mean response time adjusted min 0.2843
                                                       0.3134 0.907
                                                                         0.366
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.278 on 129 degrees of freedom
Multiple R-squared: 0.006339, Adjusted R-squared: -0.001364
F-statistic: 0.823 on 1 and 129 DF, p-value: 0.366
. .
```



**GCSE results:** M = 5.16, SD = 1.80

accuracy of answered questions: M = 0.45, SD = 0.20

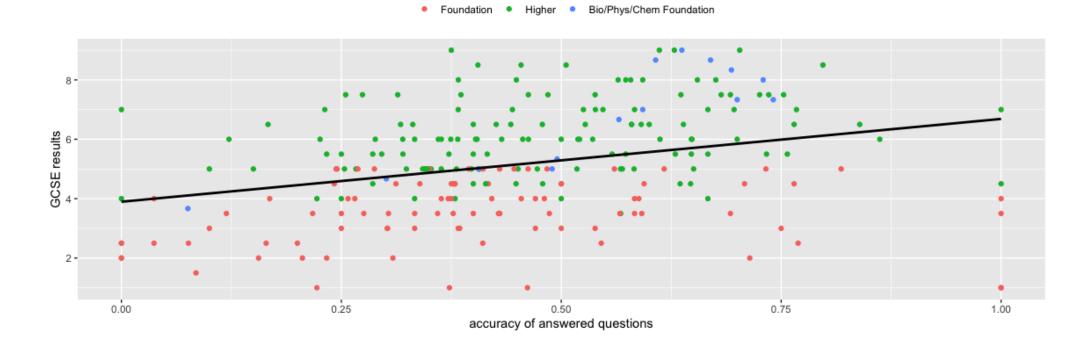
```
> ratings1.lm = lm(exam ~ accuracy, data = GCSE predictions)
> summary(ratings1.lm)
Call:
lm(formula = exam ~ accuracy, data = GCSE predictions)
Residuals:
            10 Median
   Min
                           3 Q
                                  Max
-5.6849 -1.1693 -0.0171 1.1405 4.0581
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
                      0.2696 14.452 < 2e-16 ***
            3.8961
(Intercept)
             2.7888 0.5412 5.153 5.38e-07 ***
accuracy
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 1.712 on 238 degrees of freedom
Multiple R-squared: 0.1004, Adjusted R-squared: 0.09659
F-statistic: 26.55 on 1 and 238 DF, p-value: 5.385e-07
```

#### **Foundation**

```
> exp7.lm = lm(GCSE_Foundation$GCSE ~ GCSE_Foundation$memory_accuracy, data = GCSE_Foundation)
> summary(exp7.lm)
Call:
lm(formula = GCSE_Foundation$GCSE ~ GCSE_Foundation$memory_accuracy,
    data = GCSE_Foundation)
Residuals:
    Min
              1Q Median
                                3Q
                                       Max
-2.95137 -0.63836 0.02032 0.94042 1.54080
Coefficients:
                               Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                           0.2370 13.928 <2e-16 ***
                                 3.3003
GCSE_Foundation$memory_accuracy 0.6511 0.5143 1.266 0.209
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.113 on 93 degrees of freedom
Multiple R-squared: 0.01694, Adjusted R-squared: 0.006367
F-statistic: 1.602 on 1 and 93 DF, p-value: 0.2087
> #osobno dla higher
```

#### Higher

```
> exp8.lm = lm(GCSE Higher$GCSE ~ GCSE Higher$memory accuracy, data = GCSE Higher)
> summary(exp8.lm)
Call:
lm(formula = GCSE_Higher$GCSE ~ GCSE_Higher$memory_accuracy.
    data = GCSE_Higher)
Residuals:
     Min
               1Q Median
                                        Max
                                3 Q
-2.80959 -0.91597 -0.09183 0.87070 3.03875
Coefficients:
                           Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                        0.3032 17.42 < 2e-16 ***
                             5.2829
GCSE_Higher$memory_accuracy 1.8089
                                                 3.07 0.00261 **
                                        0.5892
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.237 on 129 degrees of freedom
Multiple R-squared: 0.0681, Adjusted R-squared: 0.06088
F-statistic: 9.427 on 1 and 129 DF, p-value: 0.002608
>
```



**GCSE results:** M = 5.16, SD = 1.80

The number of questions: M = 48.03, SD = 45.07

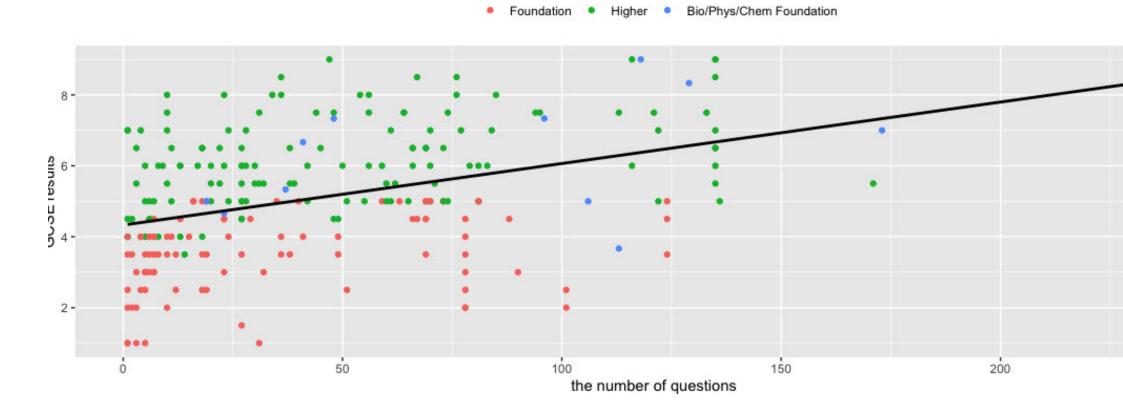
```
> ratings2.lm = lm(exam ~ questions, data = GCSE_predictions)
> summary(ratings2.lm)
Call:
lm(formula = exam ~ questions, data = GCSE_predictions)
Residuals:
            1Q Median 3Q
    Min
                                  Max
-4.0815 -0.9950 -0.0619 1.1885 3.8547
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 4.330408  0.153591  28.194  < 2e-16 ***
questions 0.017338 0.002334 7.428 1.96e-12 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.626 on 238 degrees of freedom
Multiple R-squared: 0.1882, Adjusted R-squared: 0.1848
F-statistic: 55.18 on 1 and 238 DF. p-value: 1.959e-12
```

#### **Foundation**

```
> exp9.lm = lm(GCSE_Foundation$GCSE - GCSE_Foundation$numeber_questions, data = GCSE_Foundation)
> summary(exp9.lm)
Call:
lm(formula = GCSE_Foundation$GCSE ~ GCSE_Foundation$numeber_questions,
   data = GCSE_Foundation)
Residuals:
   Min
            10 Median
                                   Max
                            3Q
-2.5219 -0.7381 0.2171 0.7370 1.6275
Coefficients:
                                Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                3.213169 0.158523 20.27 <2e-16 ***
GCSE Foundation$numeber questions 0.009958 0.003255 3.06 0.0029 **
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 1.07 on 93 degrees of freedom
Multiple R-squared: 0.09145, Adjusted R-squared: 0.08168
F-statistic: 9.361 on 1 and 93 DF, p-value: 0.002896
```

#### Higher

```
> exp10.lm = lm(GCSE Higher$GCSE ~ GCSE Higher$numeber questions, data = GCSE Higher)
> summary(exp10.lm)
Call:
lm(formula = GCSE Higher$GCSE ~ GCSE Higher$numeber questions,
    data = GCSE Higher)
Residuals:
    Min
              10 Median
                                30
                                       Max
-2.23157 -0.89568 -0.05819 0.81349 2.88348
Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
                             5.568265   0.166573   33.428   < 2e-16 ***
(Intercept)
GCSE Higher$numeber questions 0.011665 0.002595 4.495 1.53e-05 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.192 on 129 degrees of freedom
Multiple R-squared: 0.1354, Adjusted R-squared: 0.1287
F-statistic: 20.21 on 1 and 129 DF, p-value: 1.53e-05
```



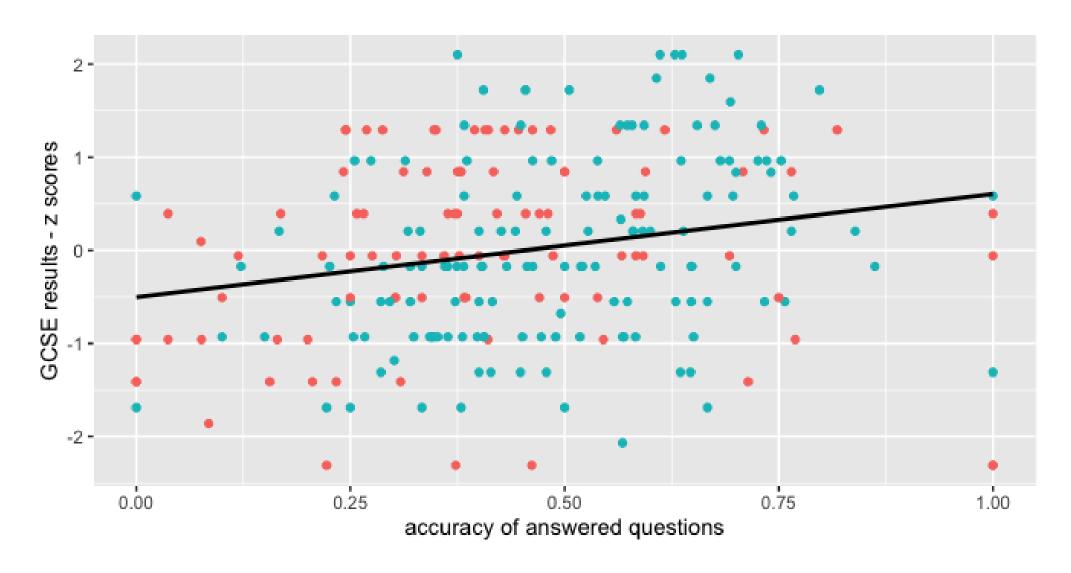
#### Interaction

```
> ratings.lm = lm(exam ~ accuracy * questions, data = GCSE predictions)
> summary(ratings.lm)
Call:
lm(formula = exam ~ accuracy * questions, data = GCSE predictions)
Residuals:
    Min
            1Q Median
                           3Q
                                  Max
-4.2324 -0.9623 -0.0738 1.0292 4.1166
Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
(Intercept)
                 3.825147 0.319187 11.984 < 2e-16 ***
accuracy
                 1.370481 0.592650 2.312 0.021613 *
questions
                  -0.003546   0.005871   -0.604   0.546436
accuracy:questions 0.040339 0.010654 3.786 0.000194 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.489 on 236 degrees of freedom
Multiple R-squared: 0.3257, Adjusted R-squared: 0.3171
F-statistic: 37.99 on 3 and 236 DF, p-value: < 2.2e-16
>
```

#### **Model 1: GCSE – z scores**

**Predictor: accuracy of answered questions** 

```
> model1 = lm(report$GCSE_z ~ report$memory_accuracy, data = report)
> summary(model1)
Call:
lm(formula = report$GCSE z ~ report$memory accuracy, data = report)
Residuals:
    Min
             10 Median
                             3 Q
                                    Max
-2.91431 -0.71188 -0.02681 0.70957 2.18642
Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
            (Intercept)
report$memory_accuracy 1.1088 0.3078 3.602 0.000384 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.9738 on 238 degrees of freedom
Multiple R-squared: 0.0517, Adjusted R-squared: 0.04771
F-statistic: 12.98 on 1 and 238 DF, p-value: 0.0003842
```



#### **Moderator: Exam Level (1 – Foundation, 2 – Higher)**

```
Call:
 lm(formula = GCSE z ~ Level..1.Foundation..2.Higher. * memory_accuracy,
    data = report)
 Residuals:
     Min
              1Q Median
 -2.64932 -0.68998 -0.01088 0.71245 2.30189
 Coefficients:
                                         Estimate Std. Error t value Pr(>|t|)
 (Intercept)
                                           0.4108
                                                    0.4679 0.878 0.3809
 Level..1.Foundation..2.Higher.
                                          -0.6377
                                                    0.3087 -2.066
                                                                  0.0399 *
 memory_accuracy
                                          -0.6293
                                                    0.9912 -0.635 0.5261
 Level..1.Foundation..2.Higher.:memory_accuracy 1.1963 0.6275 1.906 0.0578 .
 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
 Residual standard error: 0.9692 on 236 degrees of freedom
 Multiple R-squared: 0.06857, Adjusted R-squared: 0.05673
 F-statistic: 5.792 on 3 and 236 DF, p-value: 0.0007767
SIMPLE SLOPES ANALYSIS
Slope of memory accuracy when Level..1. Foundation..2. Higher. = 1.00 (1):
           S.E. t val.
  Est.
                      1.28 0.20
  0.57
         0.44
Slope of memory_accuracy when Level..1. Foundation..2. Higher. = 2.00 (2):
  Est.
           S.E.
                  t val.
```

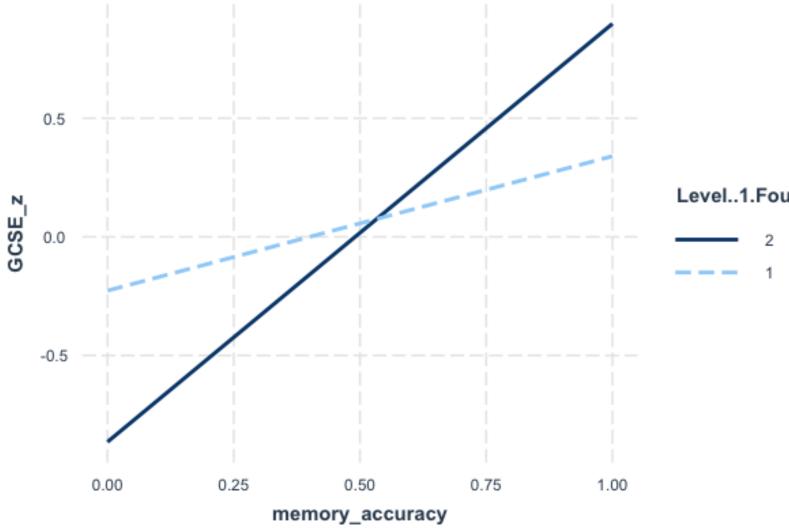
.

1.76

0.44

3.97

0.00

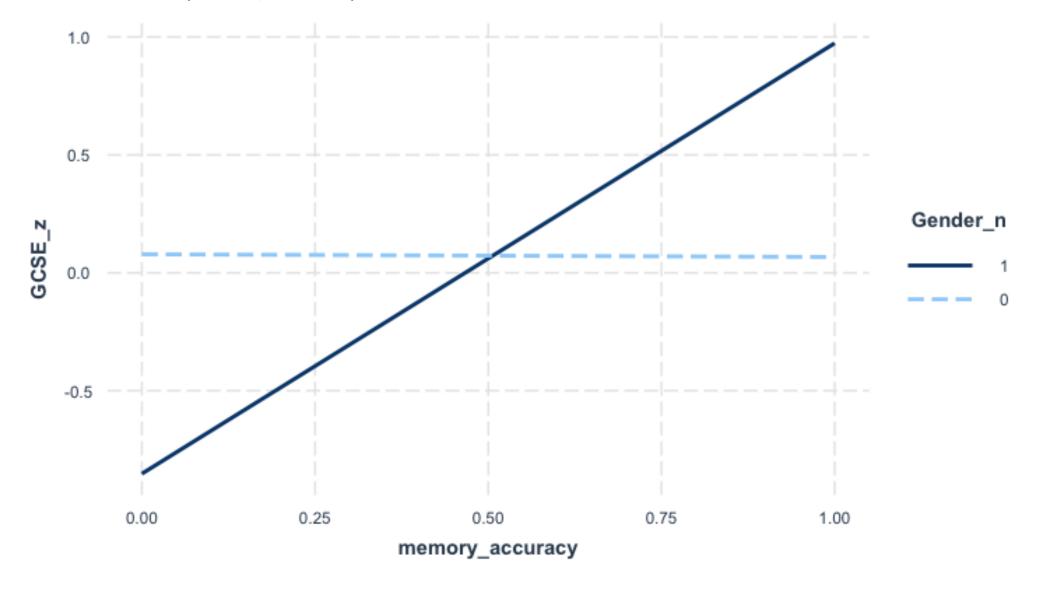


Level..1.Foundation..2.Higher.

#### **Moderator: Gender (0 – Male, 1 - Female)**

```
Call:
lm(formula = GCSE_z ~ Gender_n * memory_accuracy, data = report)
Residuals:
    Min
             1Q Median
                            3 Q
-2.38525 -0.63856 -0.01219 0.76899 2.26595
Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
                               0.25032 0.314 0.75405
(Intercept)
                    0.07852
Gender_n
                    -0.93072 0.31448 -2.960 0.00339 **
memory_accuracy -0.01160 0.48240 -0.024 0.98084
Gender_n:memory_accuracy 1.83742 0.62109 2.958 0.00341 **
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.9595 on 236 degrees of freedom
Multiple R-squared: 0.08715, Adjusted R-squared: 0.07555
F-statistic: 7.511 on 3 and 236 DF, p-value: 8.033e-05
SIMPLE SLOPES ANALYSIS
Slope of memory accuracy when Gender n = 0.00 (0):
   Est. S.E. t val.
  -0.01 0.48 -0.02 0.98
Slope of memory_accuracy when Gender_n = 1.00 (1):
  Est. S.E. t val.
  1.83 0.39 4.67 0.00
```

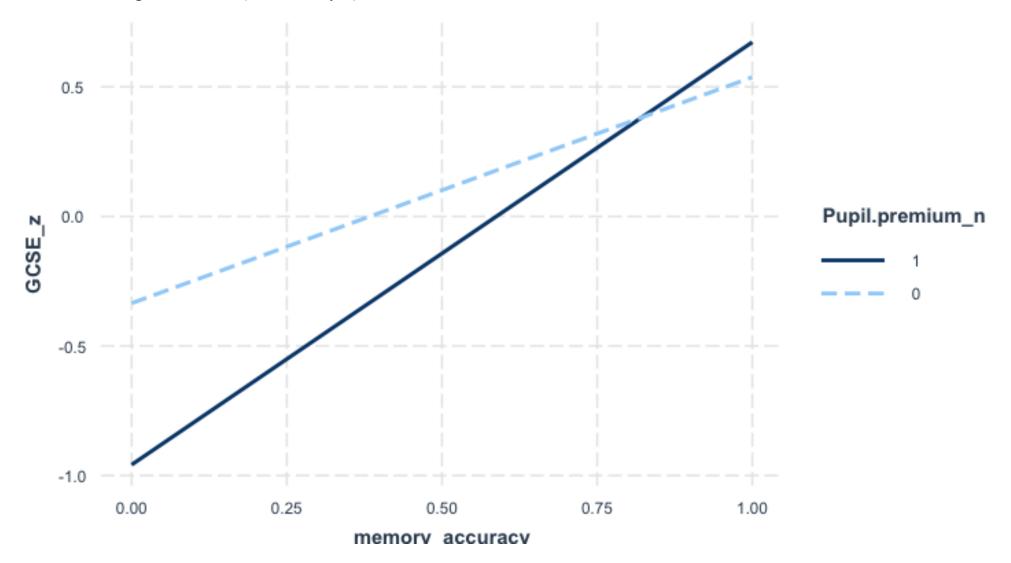
### **Moderator: Gender (0 – Male, 1 - Female)**



#### Moderator: Pupil Premium (0 - no, 1 - yes)

```
Call:
 lm(formula = GCSE_z ~ Pupil.premium_n * memory_accuracy, data = report)
 Residuals:
    Min
            1Q Median
                         3Q
                               Max
 -2.8466 -0.6603 0.0061 0.6970 2.1063
 Coefficients:
                            Estimate Std. Error t value Pr(>|t|)
 (Intercept)
                            -0.3350
                                       0.1812 -1.849 0.0658 .
 Pupil.premium_n
                            -0.6233
                                    0.3370 -1.849 0.0657 .
                             memory_accuracy
 Pupil.premium_n:memory_accuracy 0.7584 0.6676 1.136 0.2571
 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
 Residual standard error: 0.9683 on 236 degrees of freedom
 Multiple R-squared: 0.07033, Adjusted R-squared: 0.05852
 F-statistic: 5.951 on 3 and 236 DF, p-value: 0.0006286
SIMPLE SLOPES ANALYSIS
Slope of memory accuracy when Pupil.premium n = 0.00 (0):
  Est.
          S.E. t val.
         0.37
                2.38 0.02
  0.87
Slope of memory accuracy when Pupil.premium n = 1.00 (1):
  Est. S.E. t val.
         0.56
  1.63
                   2.92 0.00
```

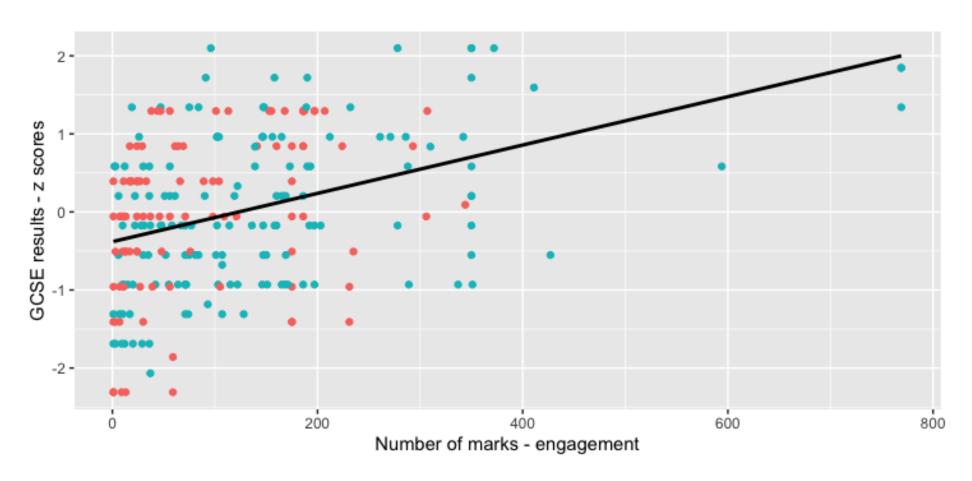
### **Moderator: Pupil Premium (0 – no, 1 - yes)**



#### **Model 2 – GCSE – z scores**

#### Predictor: number of marks attempted

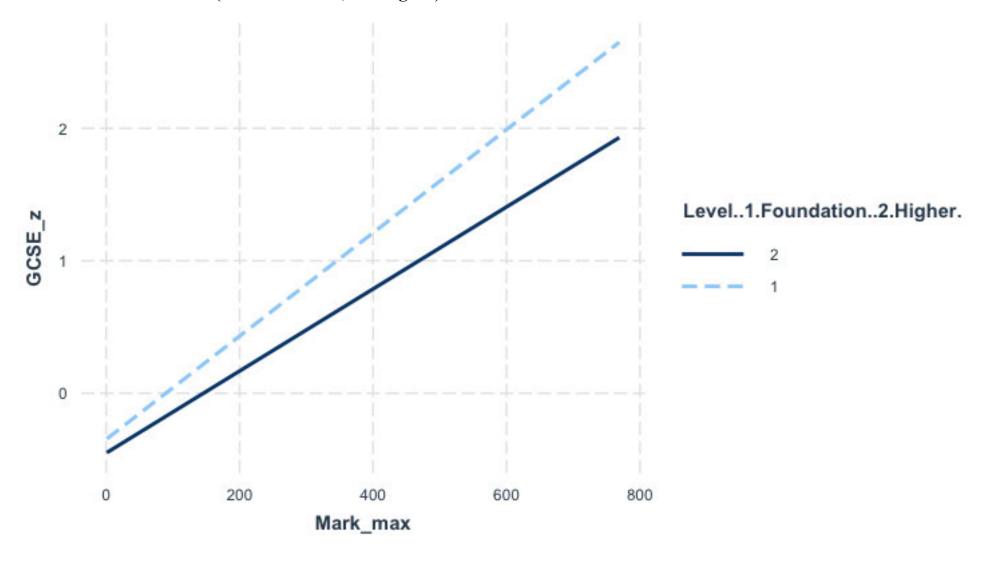
```
Call:
lm(formula = report$GCSE z ~ report$Mark max, data = report)
Residuals:
           10 Median 30
    Min
                                       Max
-2.11081 -0.67148  0.06054  0.72095  2.18239
Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.3811316  0.0825960  -4.614  6.45e-06 ***
report$Mark max 0.0030955 0.0004669 6.630 2.23e-10 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 0.9188 on 238 degrees of freedom
Multiple R-squared: 0.1559, Adjusted R-squared: 0.1523
F-statistic: 43.95 on 1 and 238 DF, p-value: 2.235e-10
>
```



#### **Moderator: Exam Level (1 – Foundation, 2 – Higher)**

```
Call:
lm(formula = GCSE_z ~ Level..1.Foundation..2.Higher. * Mark_max,
    data = report)
Residuals:
     Min
              1Q Median
-2.19132 -0.68336 0.07097 0.66829 2.25229
Coefficients:
                                      Estimate Std. Error t value Pr(>|t|)
(Intercept)
                                     -0.2448268 0.2915277 -0.840 0.4019
Level..1.Foundation..2.Higher.
                                     -0.1031349 0.1735632 -0.594 0.5529
Mark max
                                      0.0046995 0.0022494 2.089 0.0378 *
Level..1.Foundation..2.Higher.:Mark_max -0.0008016 0.0012151 -0.660 0.5101
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.9176 on 236 degrees of freedom
Multiple R-squared: 0.1652, Adjusted R-squared: 0.1546
F-statistic: 15.56 on 3 and 236 DF, p-value: 2.857e-09
SIMPLE SLOPES ANALYSIS
Slope of Mark_max when Level..1.Foundation..2.Higher. = 1.00 (1):
 Est. S.E. t val. p
------
 0.00 0.00 3.57 0.00
Slope of Mark max when Level..1. Foundation..2. Higher. = 2.00 (2):
 Est. S.E. t val.
------
 0.00 0.00 5.83 0.00
```

### **Moderator:** Exam Level (1 – Foundation, 2 – Higher)



#### **Moderator: Gender (0 – Male, 1 - Female)**

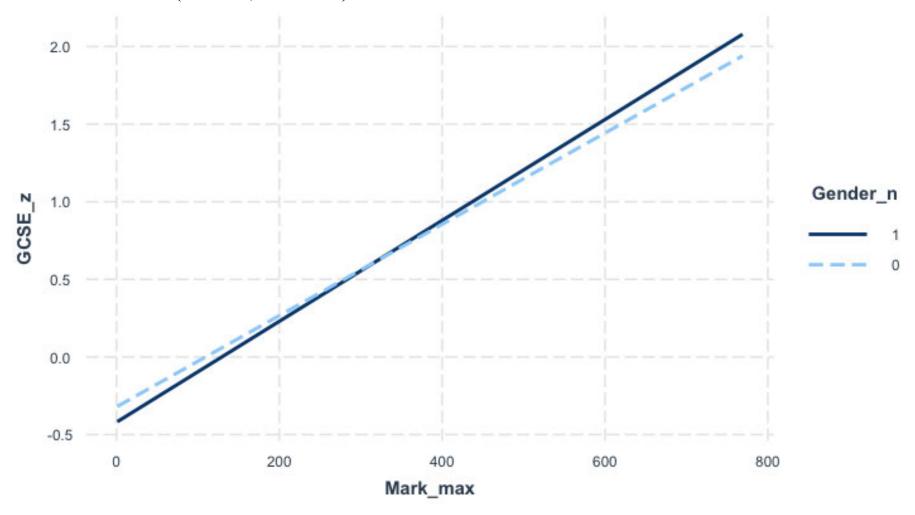
Est. S.E. t val.

4.67 0.00

0.00 0.00

```
Call:
lm(formula = GCSE_z ~ Gender_n * Mark_max, data = report)
Residuals:
    Min
              1Q Median
                               3Q
                                       Max
-2.08092 -0.67419 0.04996 0.70880 2.20659
Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.3214235 0.1319315 -2.436 0.0156 *
             -0.0987008 0.1717597 -0.575 0.5661
Gender n
Mark max
               0.0029402 0.0006362 4.621 6.28e-06 ***
Gender n: Mark max 0.0003095 0.0009428 0.328 0.7430
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.922 on 236 degrees of freedom
Multiple R-squared: 0.1571, Adjusted R-squared: 0.1464
F-statistic: 14.66 on 3 and 236 DF, p-value: 8.694e-09
SIMPLE SLOPES ANALYSIS
Slope of Mark_max when Gender_n = 0.00 (0):
 Est. S.E. t val. p
 0.00 0.00 4.62 0.00
Slope of Mark max when Gender n = 1.00 (1):
```

### **Moderator: Gender (0 – Male, 1 - Female)**



#### Moderator: Pupil Premium (0 - no, 1 - yes)

```
Call:
lm(formula = GCSE_z ~ Pupil.premium_n * Mark_max, data = report)
Residuals:
     Min
              1Q Median
                              3 Q
 -2.12693 -0.66642 0.04789 0.70512 2.16603
Coefficients:
                       Estimate Std. Error t value Pr(>|t|)
(Intercept)
                     -0.3654200 0.0962083 -3.798 0.000185 ***
                   -0.0136641 0.2023781 -0.068 0.946227
Pupil.premium_n
Mark max
                       0.0031023 0.0004988 6.220 2.24e-09 ***
Pupil.premium_n:Mark_max -0.0009053  0.0018569 -0.488  0.626336
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '. 0.1 ' 1
Residual standard error: 0.9216 on 236 degrees of freedom
Multiple R-squared: 0.1578, Adjusted R-squared: 0.1471
F-statistic: 14.74 on 3 and 236 DF, p-value: 7.895e-09
SIMPLE SLOPES ANALYSIS
Slope of Mark max when Pupil.premium n = 0.00 (0):
 Est. S.E. t val.
 0.00 0.00 6.22 0.00
Slope of Mark_max when Pupil.premium_n = 1.00 (1):
 Est. S.E. t val.
 0.00 0.00 1.23 0.22
```

## **Moderator: Pupil Premium (0 – no, 1 - yes)**

