# W241 Final Project | Decoding Game of Thrones

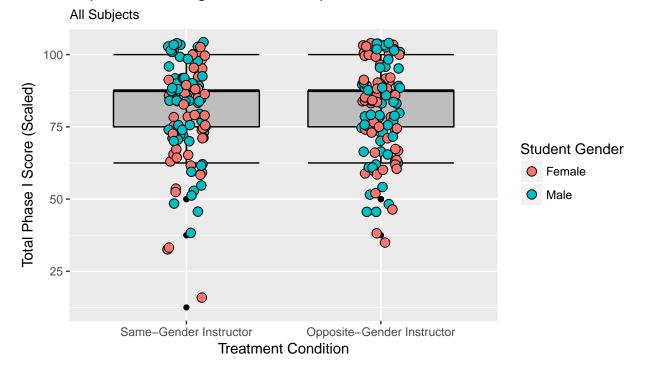
The Effect of Gender Expression on Information Uptake

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Individual Values Plot

## Experiment Design 1: Two-Group



#### Hypothesis Testing

## ##

n1 = 112

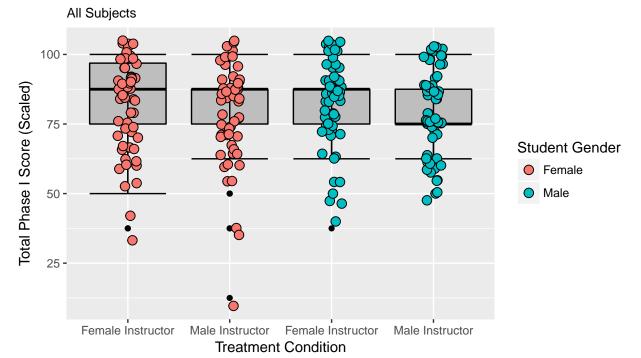
```
#calculate group distributions
same_gender_group <- e1[Same_Gen == 0, Total_Scaled, ]</pre>
mixed_gender_group <- e1[Same_Gen == 1, Total_Scaled, ]</pre>
t.test(same_gender_group, mixed_gender_group)
##
   Welch Two Sample t-test
##
## data: same_gender_group and mixed_gender_group
## t = -0.10149, df = 221.53, p-value = 0.9192
\mbox{\tt \#\#} alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -4.557389 4.110960
## sample estimates:
## mean of x mean of y
## 80.24554 80.46875
pwr.t2n.test(n1=length(same_gender_group), n2=length(mixed_gender_group), d=effect(same_gender_group,mix
##
##
        t test power calculation
```

```
## n2 = 112
## d = 0.01359298
## sig.level = 0.05
power = 0.05117589
## alternative = two.sided
```

### 3.1 | Primary Analysis 2: Four-Group

#### Individual Values Plot

# Experiment Design 2: Four-Group



#### Hypothesis Testing

```
#calculate group distributions
FF_group <- e1[FF == 1, Total_Scaled, ]</pre>
FM_group <- e1[FM == 1, Total_Scaled, ]</pre>
MF_group <- e1[MF == 1, Total_Scaled, ]</pre>
MM_group <- e1[MM == 1, Total_Scaled, ]</pre>
t.test(FF_group, FM_group, paired = TRUE, alternative = "two.sided")
##
   Paired t-test
##
## data: FF_group and FM_group
## t = 0.85695, df = 53, p-value = 0.3953
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -4.034115 10.052633
## sample estimates:
## mean of the differences
##
                  3.009259
pwr.t.test(n=length(FF_group), d=effect(FF_group, FM_group), type = "paired")
##
##
        Paired t test power calculation
##
##
                 n = 54
                 d = 0.1729021
##
##
         sig.level = 0.05
##
             power = 0.2387907
##
       alternative = two.sided
##
## NOTE: n is number of *pairs*
t.test(MF_group, MM_group, paired = TRUE, alternative = "two.sided")
##
## Paired t-test
##
## data: MF_group and MM_group
## t = 0.97569, df = 57, p-value = 0.3333
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -2.494825 7.236205
## sample estimates:
## mean of the differences
##
                   2.37069
pwr.t.test(n=length(MF_group), d=effect(MF_group, MM_group), type = "paired")
```

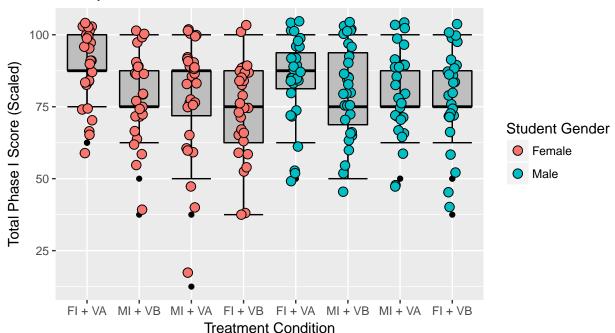
##

### 3.2 | Primary Analysis 3: Eight-Group

#### Individual Values Plot

## Experiment Design 3: Eight-Group





#### Hypothesis Testing

```
#calculate group distributions
FFA_group <- e1[FFA == 1, Total_Scaled, ]</pre>
FMB_group <- e1[FMB == 1, Total_Scaled, ]</pre>
FMA group <- e1[FMA == 1, Total Scaled, ]
FFB_group <- e1[FFB == 1, Total_Scaled, ]</pre>
MFA_group <- e1[MFA == 1, Total_Scaled, ]</pre>
MMB_group <- e1[MMB == 1, Total_Scaled, ]</pre>
MMA_group <- e1[MMA == 1, Total_Scaled, ]</pre>
MFB_group <- e1[MFB == 1, Total_Scaled, ]</pre>
Not_FFA_group <- e1[FFA != 1, Total_Scaled, ]</pre>
t.test(FFA_group, FMB_group, paired = TRUE, alternative = "two.sided")
##
##
   Paired t-test
##
## data: FFA_group and FMB_group
## t = 3.2022, df = 25, p-value = 0.003696
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
    3.774281 17.379565
## sample estimates:
## mean of the differences
##
                  10.57692
pwr.t.test(n=length(FFA_group), d=effect(FFA_group,FMB_group), type = "paired")
##
##
        Paired t test power calculation
##
##
                 n = 26
##
                 d = 0.6973988
         sig.level = 0.05
##
##
             power = 0.9273691
##
       alternative = two.sided
##
## NOTE: n is number of *pairs*
t.test(FMA_group, FFB_group, paired = TRUE, alternative = "two.sided")
##
## Paired t-test
## data: FMA_group and FFB_group
## t = 1.1214, df = 27, p-value = 0.272
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -3.333396 11.369110
```

```
## sample estimates:
## mean of the differences
                  4.017857
pwr.t.test(n=length(FMA_group), d=effect(FMA_group,FFB_group), type = "paired")
##
##
        Paired t test power calculation
##
                 n = 28
##
##
                 d = 0.2153292
         sig.level = 0.05
##
##
             power = 0.1957883
##
       alternative = two.sided
## NOTE: n is number of *pairs*
t.test(MFA_group, MMB_group, paired = TRUE, alternative = "two.sided")
##
## Paired t-test
##
## data: MFA_group and MMB_group
## t = 1.3975, df = 30, p-value = 0.1725
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -2.232504 11.909923
## sample estimates:
## mean of the differences
##
                   4.83871
pwr.t.test(n=length(MFA_group), d=effect(MFA_group,MMB_group), type = "paired")
##
##
        Paired t test power calculation
##
##
                 n = 31
                 d = 0.3100285
##
##
         sig.level = 0.05
##
             power = 0.3864208
##
       alternative = two.sided
##
## NOTE: n is number of *pairs*
t.test(MMA_group, MFB_group, paired = TRUE, alternative = "two.sided")
##
## Paired t-test
##
## data: MMA_group and MFB_group
## t = 0.13746, df = 26, p-value = 0.8917
```

```
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -6.460008 7.385934
## sample estimates:
## mean of the differences
##
                  0.462963
pwr.t.test(n=length(MMA_group), d=effect(MMA_group,MFB_group), type = "paired")
##
##
       Paired t test power calculation
##
##
                 n = 27
##
                 d = 0.03008682
##
         sig.level = 0.05
##
             power = 0.05260382
##
       alternative = two.sided
## NOTE: n is number of *pairs*
t.test(FFA_group, Not_FFA_group)
##
##
  Welch Two Sample t-test
##
## data: FFA_group and Not_FFA_group
## t = 3.4807, df = 36.706, p-value = 0.001308
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
   4.057041 15.367978
##
## sample estimates:
## mean of x mean of y
## 88.94231 79.22980
pwr.t2n.test(n1=length(FFA_group), n2=length(Not_FFA_group), d=effect(FFA_group,Not_FFA_group))
##
##
        t test power calculation
##
##
                n1 = 26
##
                n2 = 198
##
                 d = 0.5914583
##
         sig.level = 0.05
##
             power = 0.8059823
##
       alternative = two.sided
```

Regression Analysis

```
# fit regression models
m3c1 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e1)
m3c2 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA + IFE_Scaled, data = e1)
m3c3 <- lm(Total_Perf ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e1)
# output results
reg_rse_3(m3c1, m3c2, m3c3, "Experiment Design 3: Eight-Group")</pre>
```

Table 1: Experiment Design 3: Eight-Group

	Dependent variable:		
	$Total\_Scaled$		Total_Perf
	(1)	(2)	(3)
FFA	9.776**	6.035**	5.976
	(4.103)	(2.410)	(4.103)
FMB	-0.801	$-4.542^{*}$	2.569
	(4.466)	(2.420)	(4.466)
FMA	-1.042	3.093	-4.842
	(5.044)	(2.519)	(5.044)
FFB	-5.060	-0.925	-1.690
	(4.567)	(2.537)	(4.567)
MFA	5.511	3.147	2.711
	(4.167)	(2.572)	(4.167)
MMB	0.672	-1.692	0.672
	(4.369)	(2.506)	(4.369)
MMA	0.463	0.463	-2.337
	(4.308)	(2.530)	(4.308)
IFE_Scaled		0.986***	
		(0.060)	
Constant	79.167***	0.396	-0.363
	(3.189)	(5.283)	(3.189)
Type of Std. Error	Robust	Robust	Robust
Observations	224	224	224
$\mathbb{R}^2$	0.065	0.681	0.038
Adjusted R <sup>2</sup>	0.034	0.669	0.006
Residual Std. Error	16.136 (df = 216)	9.444  (df = 215)	16.136  (df = 216)
F Statistic	$2.135^{**} (df = 7; 216)$	$57.411^{***} (df = 8; 215)$	1.205 (df = 7; 216)

```
# fit regression models
m3f1 <- lm(Total_Scaled ~ FFA, data = e1)
m3f2 <- lm(Total_Scaled ~ FFA + IFE_Scaled, data = e1)
m3f3 <- lm(Total_Perf ~ FFA, data = e1)

# output results
reg_rse_3(m3f1, m3f2, m3f3, "Experiment Design 3: Female Students/Female Instructor/Video A")</pre>
```

Table 2: Experiment Design 3: Female Students/Female Instructor/Video A

	Dependent variable:		
	Total	_Scaled	${\bf Total\_Perf}$
	(1)	(2)	(3)
FFA	9.713***	6.089***	6.351**
	(2.837)	(1.800)	(2.835)
IFE Scaled		0.971***	
		(0.059)	
Constant	79.230***	1.590	-0.739
	(1.177)	(5.103)	(1.172)
Type of Std. Error	Robust	Robust	Robust
Observations	224	224	224
$\mathbb{R}^2$	0.036	0.661	0.016
Adjusted R <sup>2</sup>	0.032	0.658	0.011
Residual Std. Error	16.159 (df = 222)	9.606 (df = 221)	16.095 (df = 222)
F Statistic	$8.303^{***} (df = 1; 222)$	215.308*** (df = 2; 221)	$3.578^* \text{ (df} = 1; 222)$
Note:	*p<0.1; **p<0.05; ***p<0.01		

## 4.1 | Secondary Analyses

### Student Subgroup

```
# create student subset
e_student <- subset(e1, e == 1)
dim(e_student)

## [1] 120 59

# fit regression models
m4a1 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_student)
m4a2 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA + IFE_Scaled, data = e_student)
m4a3 <- lm(Total_Perf ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_student)
# output results
reg_rse_3(m4a1, m4a2, m4a3, "Subgroup Analysis: Students")</pre>
```

Table 3: Subgroup Analysis: Students

	Dependent variable:			
	Tota	al Scaled	Total_Perf	
	(1)	(2)	(3)	
		<u> </u>	. , ,	
FFA	8.929* (4.649)	7.103** (3.033)	5.129 $(4.649)$	
	(4.049)	(5.055)	(4.049)	
FMB	-4.464	-6.290**	-1.094	
	(5.154)	(3.040)	(5.154)	
	,	,	,	
FMA	-5.469	1.116	-9.269	
	(7.214)	(3.758)	(7.214)	
PPD	<b>7</b> 001	0.440	0.001	
FFB	-7.031	-0.446	-3.661	
	(5.586)	(3.740)	(5.586)	
MFA	0.000	0.830	-2.800	
1711 11	(6.121)	(3.484)	(6.121)	
	(**==)	(0.101)	(0:)	
MMB	-2.679	-1.849	-2.679	
	(5.760)	(3.424)	(5.760)	
200	o ===.			
MMA	0.781	0.781	-2.019	
	(5.266)	(3.405)	(5.266)	
IFE_Scaled		0.992***		
II L_Scared		(0.096)		
		(0.000)		
Constant	81.250***	0.295	1.720	
	(3.727)	(8.438)	(3.727)	
Type of Std. Error	Robust	Robust	Robust	
Observations	120	120	120	
$\mathbb{R}^2$	0.080	0.693	0.054	
Adjusted $\mathbb{R}^2$	0.022	0.671	-0.005	
Residual Std. Error	16.205 (df = 112)	9.405 (df = 111)	16.205 (df = 112)	
F Statistic	1.390 (df = 7; 112)	$31.302^{***} (df = 8; 111)$	0.919 (df = 7; 112)	

#### Under 30 Subgroup

```
# create under 30 subset
e_young <- subset(e1, b == 1 | b == 2)
dim(e_young)

## [1] 90 59

# fit regression models
m4b1 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_young)
m4b2 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA + IFE_Scaled, data = e_young)
m4b3 <- lm(Total_Perf ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_young)
# output results
reg_rse_3(m4b1, m4b2, m4b3, "Subgroup Analysis: Under 30 Years Old")</pre>
```

#### Over 30 Subgroup

```
# create over 30 subset
e_old <- subset(e1, b == 3 | b == 4 | b == 5)
dim(e_old)

## [1] 134 59

# fit regression models
m4c1 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_old)
m4c2 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA + IFE_Scaled, data = e_old)
m4c3 <- lm(Total_Perf ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_old)

# output results
reg_rse_3(m4c1, m4c2, m4c3, "Subgroup Analysis: Over 30 Years Old")</pre>
```

#### **Highly Educated Subgroup**

```
# create highly educated subset
e_edu <- subset(e1, f == 6 | f == 7)
dim(e_edu)

## [1] 148 59

# fit regression models
m4d1 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_edu)
m4d2 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA + IFE_Scaled, data = e_edu)
m4d3 <- lm(Total_Perf ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_edu)

# output results
reg_rse_3(m4d1, m4d2, m4d3, "Subgroup Analysis: Highly Educated")</pre>
```

Table 4: Subgroup Analysis: Under 30 Years Old

	Dependent variable:		
	${\bf Total\_Scaled}$		Total_Perf
	(1)	(2)	(3)
FFA	4.773	3.750	0.973
	(6.426)	(4.145)	(6.426)
FMB	-7.727	-8.750**	-4.357
	(6.895)	(4.145)	(6.895)
FMA	-11.250	-3.750	-15.050
	(10.623)	(5.397)	(10.623)
FFB	-8.750	-1.250	-5.380
	(8.618)	(5.397)	(8.618)
MFA	-2.857	-0.714	-5.657
	(6.364)	(3.958)	(6.364)
MMB	-6.429	-4.286	-6.429
	(7.015)	(3.958)	(7.015)
MMA	-5.000	-5.000	-7.800
	(7.607)	(4.497)	(7.607)
IFE_Scaled		1.000***	
		(0.111)	
Constant	85.000***	2.500	5.470
	(5.122)	(10.031)	(5.122)
Type of Std. Error	Robust	Robust	Robust
Observations	90	90	90
$\mathbb{R}^2$	0.072	0.708	0.064
Adjusted R <sup>2</sup>	-0.007	0.679	-0.016
Residual Std. Error	17.643  (df = 82)	9.960 (df = 81)	17.643  (df = 82)
F Statistic	0.914  (df = 7; 82)	$24.545^{***} (df = 8; 81)$	0.805 (df = 7; 82)

*Note:* \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table 5: Subgroup Analysis: Over 30 Years Old

	Dependent variable:		
	Total_Scaled		Total_Perf
	(1)	(2)	(3)
FFA	12.598**	7.310**	8.798
	(5.472)	(3.098)	(5.472)
FMB	3.431	-1.857	6.801
	(6.053)	(3.137)	(6.053)
FMA	4.820	7.024***	1.020
	(5.446)	(2.583)	(5.446)
FFB	-2.819	-0.615	0.551
	(5.449)	(2.703)	(5.449)
MFA	11.029**	5.670	8.229
	(5.627)	(3.731)	(5.627)
MMB	5.147	-0.212	5.147
	(5.697)	(3.553)	(5.697)
MMA	3.676	3.676	0.876
	(5.346)	(3.122)	(5.346)
IFE_Scaled		0.972***	
		(0.068)	
Constant	75.735***	-0.369	-3.795
	(4.055)	(5.926)	(4.055)
Type of Std. Error	Robust	Robust	Robust
Observations	134	134	134
$\mathbb{R}^2$	0.094	0.678	0.051
Adjusted R <sup>2</sup>	0.043	0.658	-0.002
Residual Std. Error	15.318 (df = 126)	9.162 (df = 125)	15.318  (df = 126)
F Statistic	$1.859^* \text{ (df} = 7; 126)$	$32.945^{***} (df = 8; 125)$	0.970 (df = 7; 126)

Table 6: Subgroup Analysis: Highly Educated

	Dependent variable:		
	$Total\_Scaled$		Total_Perf
	(1)	(2)	(3)
FFA	7.341	4.924	3.541
	(5.562)	(3.080)	(5.562)
FMB	1.091	-1.326	4.461
	(5.915)	(3.084)	(5.915)
FMA	-6.548	1.395	-10.348
	(7.006)	(3.212)	(7.006)
FFB	-5.853	2.089	-2.483
	(5.603)	(3.200)	(5.603)
MFA	2.307	1.530	-0.493
	(5.445)	(2.901)	(5.445)
MMB	1.786	1.009	1.786
	(5.500)	(2.898)	(5.500)
MMA	1.786	1.786	-1.014
	(6.135)	(3.328)	(6.135)
IFE_Scaled		0.995***	
		(0.075)	
Constant	79.464***	-1.339	-0.066
	(4.429)	(6.694)	(4.429)
Type of Std. Error	Robust	Robust	Robust
Observations	148	148	148
$\mathbb{R}^2$	0.066	0.688	0.068
Adjusted R <sup>2</sup>	0.019	0.671	0.021
Residual Std. Error	16.241 (df = 140)	9.413  (df = 139)	16.241 (df = 140)
F Statistic	1.414 (df = 7; 140)	$38.401^{***} (df = 8; 139)$	1.455 (df = 7; 140)

#### Native English Speaker Subgroup

```
# create native English speaker subset
e_eng <- subset(e1, d == 1)
dim(e_eng)

## [1] 164 59

# fit regression models
m4e1 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_eng)
m4e2 <- lm(Total_Scaled ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA + IFE_Scaled, data = e_eng)
m4e3 <- lm(Total_Perf ~ FFA + FMB + FMA + FFB + MFA + MMB + MMA, data = e_eng)

# output results
reg_rse_3(m4e1, m4e2, m4e3, "Subgroup Analysis: Native English Speakers")</pre>
```

### Appendix

Exploratory Data Analysis 1

```
summary(e1$Total_Perf)

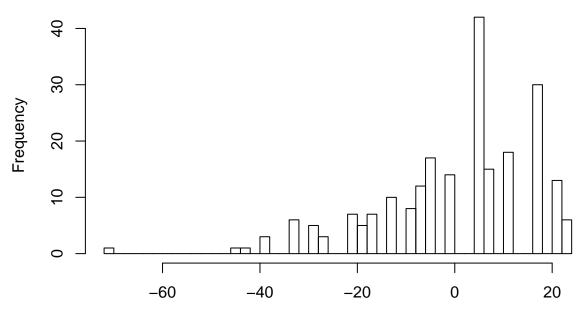
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -70.83000 -8.33000 4.17000 -0.00152 11.34000 23.84000

hist(e1$Total_Perf, breaks = 50,
    main = "Total Performance",
    xlab = "Performance Relative to Mean")
```

Table 7: Subgroup Analysis: Native English Speakers

	Dependent variable:			
	Total_Scaled		Total_Perf	
	(1)	(2)	(3)	
FFA	8.260 (5.296)	6.703** (3.046)	4.460 (5.296)	
	,	,	` '	
FMB	-0.950	-2.508	2.420	
	(5.503)	(3.050)	(5.503)	
FMA	-3.770	4.390	-7.570	
	(6.479)	(3.121)	(6.479)	
FFB	-7.341	0.818	-3.971	
	(5.846)	(3.156)	(5.846)	
MFA	3.299	3.901	0.499	
	(5.335)	(3.031)	(5.335)	
MMB	0.694	1.297	0.694	
	(5.472)	(2.956)	(5.472)	
MMA	2.778	2.778	-0.022	
	(5.524)	(3.347)	(5.524)	
IFE_Scaled		0.991***		
_		(0.069)		
Constant	80.556***	-1.337	1.026	
	(4.188)	(6.376)	(4.188)	
Type of Std. Error	Robust	Robust	Robust	
Observations	164	164	164	
$\mathbb{R}^2$	0.065	0.701	0.042	
Adjusted R <sup>2</sup>	0.023	0.686	-0.001	
Residual Std. Error	17.006 (df = 156)	9.641  (df = 155)	17.006 (df = 156)	
F Statistic	1.551 (df = 7; 156)	$45.517^{***} (df = 8; 155)$	0.987 (df = 7; 156)	

# **Total Performance**

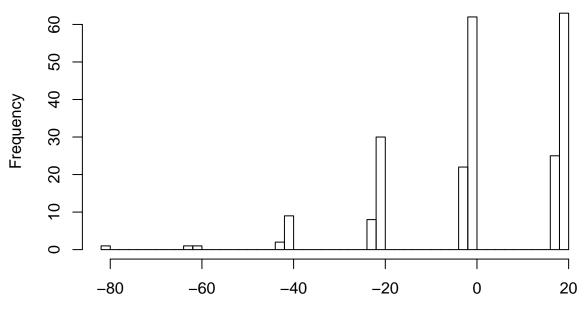


Performance Relative to Mean

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -81.07000 -3.45000 -1.07000 -0.07982 18.62000 19.26000

hist(e1$Recall_Perf, breaks = 50,
    main = "Recall Performance",
    xlab = "Performance Relative to Mean")
```

# **Recall Performance**

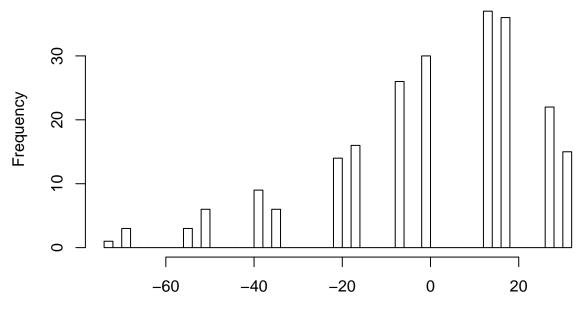


Performance Relative to Mean

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -72.99000 -17.24000 -1.85000 0.00071 16.09000 31.48000

hist(e1$Comp_Perf, breaks = 50,
    main = "Comprehension Performance",
    xlab = "Performance Relative to Mean")
```

# **Comprehension Performance**



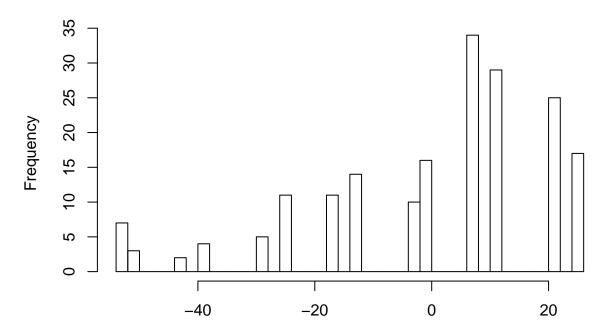
Performance Relative to Mean

```
summary(e2$Ret_Perf)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -53.19150 -13.29790 7.97870 -0.00002 11.70210 25.00000
```

```
hist(e2$Ret_Perf, breaks = 50,
    main = "Retention Performance",
    xlab = "Performance Relative to Mean")
```

### **Retention Performance**



Performance Relative to Mean

```
# fit regression models
mOa <- lm(Total_Perf ~ G_s, data = e1)
mOb <- lm(Total_Perf ~ G_i, data = e1)
mOc <- lm(Total_Perf ~ G_s + G_i, data = e1)
mOd <- lm(Total_Perf ~ G_s*G_i, data = e1)
# output results
reg_rse_4(mOa, mOb, mOc, mOd, "Preliminary Models")</pre>
```

#### Primary Analysis 1: Two-Group Regression Anlysis

```
# fit regression models
m2a1 <- lm(Total_Scaled ~ Same_Gen, data = e1)
m2a2 <- lm(Total_Scaled ~ Same_Gen + IFE_Scaled, data = e1)
m2a3 <- lm(Total_Perf ~ Same_Gen, data = e1)

# output results
reg_rse_3(m2a1, m2a2, m2a3, "Experiment Design 1: Two Groups")</pre>
```

```
# fit regression models
m2b1 <- lm(Recall_Scaled ~ Same_Gen, data = e1)
m2b2 <- lm(Recall_Scaled ~ Same_Gen + IFE_Scaled, data = e1)
m2b3 <- lm(Recall_Perf ~ Same_Gen, data = e1)

# output results
reg_rse_3(m2b1, m2b2, m2b3, "Experiment Design 1: Two Groups (Recall)")</pre>
```

Table 8: Preliminary Models

	Dependent variable:				
	${f Total\_Perf}$				
	(1)	(2)	(3)	(4)	
$G_s$	-0.004 (2.187)		-0.004 (2.189)	-1.649 (6.736)	
G_i		-2.707 (2.170)	-2.707 (2.180)	-4.372 (7.222)	
G_s:G_i				1.097 $(4.397)$	
Constant	0.004 $(3.593)$	4.058 $(3.330)$	4.064 $(4.586)$	6.562 (10.910)	
Type of Std. Error Observations R <sup>2</sup> Adjusted R <sup>2</sup> Residual Std. Error	Robust 224 0.000 -0.005 16.224 (df = 222)	Robust $224$ $0.007$ $0.003$ $16.167 (df = 222)$	Robust $224$ $0.007$ $-0.002$ $16.204 (df = 221)$	Robust $224$ $0.007$ $-0.006$ $16.238 \text{ (df} = 220)$	
F Statistic	0.00000  (df = 1; 222)	1.569 (df = 1; 222)	0.781  (df = 221)	0.540  (df = 3; 220)	

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table 9: Experiment Design 1: Two Groups

	Dependent variable:		
	Tota	al_Scaled	${\bf Total\_Perf}$
	(1)	(2)	(3)
Same_Gen	0.223	0.223	0.451
	(2.209)	(1.322)	(2.178)
IFE Scaled		0.984***	
_		(0.060)	
Constant	80.246***	1.140	-0.227
	(1.598)	(5.236)	(1.603)
Type of Std. Error	Robust	Robust	Robust
Observations	224	224	224
$\mathbb{R}^2$	0.00005	0.647	0.0002
Adjusted $\mathbb{R}^2$	-0.004	0.644	-0.004
Residual Std. Error	16.458 (df = 222)	9.803 (df = 221)	16.223 (df = 222)
F Statistic	0.010  (df = 1; 222)	$202.381^{***} (df = 2; 221)$	0.043  (df = 1; 222)

Note:

Table 10: Experiment Design 1: Two Groups (Recall)

	$Dependent\ variable:$				
	Reca	all_Scaled	$Recall\_Perf$		
	(1)	(2)	(3)		
Same_Gen	2.143	2.143	2.075		
	(2.560)	(1.862)	(2.556)		
IFE Scaled		0.980***			
_		(0.080)			
Constant	80.536***	1.805	-1.117		
	(1.948)	(7.084)	(1.950)		
Type of Std. Error	Robust	Robust	Robust		
Observations	224	224	224		
$\mathbb{R}^2$	0.003	0.479	0.003		
Adjusted R <sup>2</sup>	-0.001	0.474	-0.002		
Residual Std. Error	19.074 (df = 222)	13.825 (df = 221)	19.043 (df = 222)		
F Statistic	0.707  (df = 1; 222)	$101.458^{***} (df = 2; 221)$	0.665  (df = 1; 222)		
Note:		*p<0.1: **p<0.05: ***p<0.01			

*Note:* \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

```
# fit regression models
m2c1 <- lm(Comp_Scaled ~ Same_Gen, data = e1)
m2c2 <- lm(Comp_Scaled ~ Same_Gen + IFE_Scaled, data = e1)
m2c3 <- lm(Comp_Perf ~ Same_Gen, data = e1)

# output results
reg_rse_3(m2c1, m2c2, m2c3, "Experiment Design 1: Two Groups (Retention)")</pre>
```

#### Primary Analysis 2: Four-Group Regression Analysis

```
# fit regression models
m2c1 <- lm(Total_Scaled ~ FF + FM + MF, data = e1)
m2c2 <- lm(Total_Scaled ~ FF + FM + MF + IFE_Scaled, data = e1)
m2c3 <- lm(Total_Perf ~ FF + FM + MF, data = e1)

# output results
reg_rse_3(m2c1, m2c2, m2c3, "Experiment Design 1: Four Groups")</pre>
```

Table 11: Experiment Design 1: Two Groups (Retention)

	Dependent variable:		
	Com	p_Scaled	${\rm Comp\_Perf}$
	(1)	(2)	(3)
Same_Gen	-2.977	-2.977	-2.245
	(3.322)	(2.812)	(3.158)
IFE Scaled		0.992***	
_		(0.111)	
Constant	79.763***	0.032	1.123
	(2.221)	(9.538)	(2.182)
Type of Std. Error	Robust	Robust	Robust
Observations	224	224	224
$\mathbb{R}^2$	0.004	0.293	0.002
Adjusted $\mathbb{R}^2$	-0.001	0.287	-0.002
Residual Std. Error	24.750 (df = 222)	20.894 (df = 221)	23.525 (df = 222)
F Statistic	0.810  (df = 1; 222)	$45.820^{***} (df = 2; 221)$	0.510  (df = 1; 222)
Note:	*p<0.1; **p<0.05; ***p<0.01		

Table 12: Experiment Design 1: Four Groups

	Dependent variable:			
	$Total\_Scaled$		${\bf Total\_Perf}$	
	(1)	(2)	(3)	
FF	1.509	$3.113^*$	2.730	
	(3.087)	(1.852)	(2.965)	
FM	-1.501	0.104	-0.545	
	(3.218)	(1.862)	(3.265)	
MF	2.371	2.371	2.178	
	(2.908)	(1.819)	(2.898)	
IFE_Scaled		0.985***		
		(0.060)		
Constant	79.741***	-0.215	-1.092	
	(2.053)	(5.188)	(2.063)	
Type of Std. Error	Robust	Robust	Robust	
Observations	224	224	224	
$\mathbb{R}^2$	0.008	0.654	0.007	
Adjusted $R^2$	-0.005	0.647	-0.006	
Residual Std. Error	16.466 (df = 220)	9.751 (df = 219)	16.238 (df = 220)	
F Statistic	0.597 (df = 3; 220)	$103.362^{***} (df = 4; 219)$	0.540  (df = 3; 220)	
Note:	*p<0.1; **p<0.05; ***p<0.01			

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