STAT 8801 Group Mu Project

Abhishek Nandy, Heidi Sutter, Yanjia Yu, Li Zhong, Megan Heyman, Yoo Jeong Jang

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2 Part 2: Data Exploration



• Three teachers at a local high school conducted an experiment

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- Tactile, Kinesthetic, Auditory, Visual

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- New style help student learn better?
- Learning preference affect learning?
- How much does each new delivery method help?
- Use different style for different course material?

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- New style of teaching for two classes
- Traditional style for the control group
- Run order is randomized
- Record test scores before and after teaching each topic
- Record a higher learning score
- Record preference scores for new method

The dataset:

id	sex	clas	SS	p1	f1	s1	h1	p2	f2	s2	h2
s01	М	2		70	100	60	4	48	88	52	3
☎02	F	2		80	85	60	2	50	76	60	3
р3	f3	s3 h	13	t	k	а	v				
72	64	60	2	61	44	57	55				
70	72	60	1	35	44	57	34				

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id: Student identifier sex: Gender of Student class: Class: 1,2,3 Control group: 3,1,2 p1, p2, p3: Prestest score (out of 100) for UNIT = 1,2,3 f1, f2, f3: Posttest score (out of 100) for UNIT = 1,2,3 s1, s2, s3: Attitude score (out of 60) for UNIT = 1,2,3 h1, h2, h3: Higher learning test score for UNIT = 1,2,3 t,k,a,v: Learning Style Preference (60+ is strong preference)

Response variable: difference of test scores, higher learning test score

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Possible predictors: Teaching(categorical variable), preference, attitude, gender

Part 2: Data Exploration



Possible Response Variables

Pre-Test, Post-Test, Attitude, Higher Learning: Recorded on each unit







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Possible Response Variables (cont.)

Correlation in responses

	S	А	HL
S	1	0.05	0.42
A	-	1	0.10
HL	-	-	1

S=Score Change, A=Attitude, HL=Higher Learning

- Combined the test scores to measure change (post-pre)
 - Gives an idea how much students learn
 - Lose information about the high and low scores (100-70 is same as 70-40)
- Attitude: Mostly high scores with a few outliers. Higher variability in class 1.
- Further analysis conducted by our group only used change in score.

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Possible Response Variables (cont.)

Of those having an attitude score of 40 or less

- 14 are in class 1 and the other 3 are in class 2
- 13 are female
- 2 in Unit 1, 9 in Unit 2, 6 in Unit 3
- 2 students gave low attitude scores on all three units
 - One of these had failing scores on all 3 post tests
 - The other failed only one post test but had A's on the others.

Some Interesting Statistics

Strong Learning Style Preference (60+ on 1-100 scale)

	Т	Κ	А	V
	6	2	4	7
Т	-	2	4	4
K	-	-	1	0
Α	-	-	-	0
ΤK	-	-	1	4
AV	1	1	-	-
KAV	0	-	-	-

22 students did not have any strong preference.

Method vs. Change in Test overall



Class 3



High T, K, A, V vs. Change in Test









T Unit 2

K Unit 2

A Unit 2

V Unit 2



R P P New Control











K Unit 3



9

8

30

2

0

ę







Control

New



Unit and Gender with Score Change

Female, Unit 1

Male, Unit 1





Female, Unit 2

Male, Unit 2











Summaries Covariate by Class



	Male	Female
Class 1	9	12
Class 2	14	7
Class 3	11	10

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Part 3: Analysis (Section 1)



Linear Model for Part 1

- $m1 < -lm(y \sim sex * method * (t + k + a + v) + class + unit)$
- y: score change; sex: 1-Male, 0-Female; method: 1-New, 0-Traditional; t, k, a, v: 0-100; class, unit: block
- reasons:
- unit as a block (e.g.: gravitation, electrostatics, magnetism)
- *class* as a block (three teacher each teaches the same class through all three units)
- there is no significant interactions between t, k, a, v

Results for m1: ANOVA Table

> car::Anova(m1)									
Anova Table (Type II tests)									
Response: y									
	Sum Sq	Df	F value	Pr(>F)					
sex	69	1	0.3029	0.582877					
method	14180	1	62.3825	5.124e-13	***				
t	1321	1	5.8108	0.017115	*				
k	13	1	0.0580	0.809991					
a	52	1	0.2270	0.634449					
v	267	1	1.1757	0.279929					
class	778	2	1.7120	0.183939					
unit	490	2	1.0771	0.343145					
sex:method	108	1	0.4753	0.491602					
sex:t	5	1	0.0205	0.886250					
sex:k	99	1	0.4352	0.510430					
sex:a	31	1	0.1362	0.712620					
sex:v	466	1	2.0501	0.154237					
method:t	2430	1	10.6898	0.001331	**				
method:k	12	1	0.0542	0.816200					
method:a	502	1	2.2066	0.139479					
method:v	33	1	0.1462	0.702749					
<pre>sex:method:t</pre>	27	1	0.1172	0.732512					
<pre>sex:method:k</pre>	106	1	0.4664	0.495681					
sex:method:a	. 9	1	0.0392	0.843298					
<pre>sex:method:v</pre>	18	1	0.0796	0.778173					
Residuals	34779	153							

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Variable Selection

```
> library(MASS)
> stepAIC(lm(v<sup>*</sup>sex*method*(t+k+a+v)+class+unit),
         scope = list(upper= v~sex*method*(t+k+a+v)+class+unit,
+
                      lower= v~1+class+unit).direction="backward")
Call:
lm(formula = v ~ sex + method + t + a + v + class + unit + sex:v +
   method:t + method:a)
Coefficients:
(Intercept) sex1
                           method1
                                              t1
                                                           а
   11.3810
            16.3395
                           -10.5335
                                         -6.2979
                                                     -0.2237
            class2
                          class3
                                           unit2
                                                       unit3
         v
    0.2725 -2.3364
                             3.0217
                                          1.9050
                                                     -2.2453
    sex1:v method1:t1
                          method1:a
   -0.3464
                20.7686
                             0.4128
```

So, we get the updated model m2:

 $y \sim sex + method + t + a + v + class + unit + sex : v + method : t + method : a$

Results for m2: ANOVA Table

> car::Anova(m2)							
Anova Tab	le (Type	II	tests)				
Response:	У						
	Sum Sq	Df	F value	Pr(>F)			
sex	69	1	0.3215	0.571468			
method	14180	1	66.0148	1.028e-13	***		
t	2105	1	9.8010	0.002065	**		
a	30	1	0.1382	0.710560			
v	280	1	1.3029	0.255349			
class	709	2	1.6513	0.194981			
unit	495	2	1.1521	0.318525			
sex:v	624	1	2.9036	0.090275			
method:t	3843	1	17.8920	3.878e-05	***		
method:a	583	1	2.7118	0.101520			
Residuals	35228	164					

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Results for m2: Summary

```
> summary(m2)
Call
lm(formula = v ~ sex + method + t + a + v + class + unit + sex:v +
   method:t + method:a)
Coefficients:
          Estimate Std. Error t value Pr(>|t|)
(Intercept) 11.3810 15.2375 0.747 0.4562
sev1
           16.3395 10.6097 1.540 0.1255
method1
          -10.5335 13.6946 -0.769 0.4429
t1
          -6.2979 4.0654 -1.549 0.1233
a
           -0.2237 0.2170 -1.031
                                      0.3041
           0.2725
                      0.1414 1.927
                                      0.0557 .
v
           -2.3364
                      2.7666 -0.845
                                      0.3996
class2
class3
           3.0217
                      2.8180 1.072
                                      0.2852
unit2
           1,9050
                      2.7282
                              0.698
                                      0.4860
unit3
           -2.2453
                      2.7383 -0.820
                                      0.4134
sex1:v
           -0.3464
                      0.2033 -1.704
                                      0.0903 .
method1:t1
           20.7686
                      4.9100
                              4.230 3.88e-05 ***
           0.4128
                      0.2507
                              1.647
                                      0.1015
method1:a
```

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Model Diagnostic



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Conclusion about the Influence of New Method

- On average, the new method improves students' performance of score change.
- For students with a high value on t or a, they benefit more than other students.

Some More Analysis Using SPSS

		Unstand Coeffic	lardized cients	Standardized Coefficients			95.0% Col Interval	fidence for B
Model		в	Std. Error	Beta	t	Sig.	Lower Bound	Upper Bound
9	(Constant)	14.526	3.518	20	4.129	.000	7.482	21.571
	taught by new teaching method for unit 1	13.074	4.272	.376	3.060	.003	4.518	21.629
2	(Constant)	16.472	4.742		3.473	.001	6.941	26.002
	taught by new teaching method for unit 1	3.751	5.891	.108	.637	.527	-8.087	15.589
	preference for learning style t	-11.220	8.048	334	-1.394	.170	-27.393	4.954
	preference for learning style k	11.675	10.171	.280	1.148	.257	-8.765	32.115
	preference for learning style a	1.138	11.539	.028	.099	.922	-22.050	24.326
	preference for learning style v	7.748	12.866	.207	.602	.550	-18.107	33.603
	interaction between t.type an intervention 1	21.868	9.604	.572	2.277	.027	2.568	41.168
	interaction between k.type an intervention 1	-13.376	12.072	282	-1.108	.273	-37.636	10.883
	interaction between a type an intervention 1	5.726	12.870	.132	.445	.658	-20.138	31.590
	interaction between v.type an intervention 1	748	13.977	019	054	.958	-28.836	27.340

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Some More Analysis Using SPSS, cont.

		Coe	fficients					
		Unstand	ardized	Standardized Coefficients			95.0% Co Interval	nfidence I for B
Model		в	Std. Error	Beta	t	Sig.	Lower Bound	Upper Bound
1	(Constant)	13.667	4.018	10.07	3.401	.001	5.620	21.713
	taught by new teaching method for unit 2	18.412	5.007	.438	3.677	.001	8.386	28.439
2	(Constant)	16.834	5.576		3.019	.004	5.629	28.039
	taught by new teaching method for unit 2	7.823	7.049	.186	1.110	.273	-6.343	21.989
	preference for learning style t	4.503	10.201	.108	.441	.661	-15.998	25.003
	preference for learning style k	-2.679	10.880	052	246	.807	-24.543	19.186
	preference for learning style a	-12.699	10.227	254	-1.242	.220	-33.251	7.852
	preference for learning style v	-2.964	10.552	064	281	.780	-24.170	18.242
	interaction between t.type an intervention 2	10.883	11.931	.245	.912	.366	-13.093	34.859
	interaction between k.type an intervention 2	9.309	13.466	.150	.691	.493	-17.751	36.369
	interaction between a type an intervention 2	14.268	12.772	.229	1.117	.269	-11.399	39.935
	interaction between v.type an intervention 2	727	12.574	014	058	.954	-25.995	24.540

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Some More Analysis Using SPSS, cont.

		Coel	ficients ^a					
		Unstand Coeffi	lardized cients	Standardized Coefficients			95.0% Col Interval	nfidence for B
Model		В	Std. Error	Beta	t	Sig.	Bound	Bound
1	(Constant)	2.737	3.085		.887	.379	-3.440	8.914
	taught by new teaching method for unit 3	25.538	3.746	.670	6.817	.000	18.036	33.040
2	(Constant)	3.356	4.210		.797	.429	-5.105	11.817
	taught by new teaching method for unit 3	20.132	4.835	.528	4.163	.000	10.415	29.849
	preference for learning style t	-6.967	5.261	- 189	-1.324	.192	-17.539	3.605
	preference for learning style k	8.052	7.114	.176	1.132	.263	-6.244	22.349
	preference for learning style a	1.956	5.994	.044	.326	.746	-10.090	14.001
	preference for learning style v	1.119	5.463	.027	.205	.839	-9.858	12.097
	interaction between t.type an intervention 3	24.644	6.579	.574	3.746	.000	11.422	37.866
	interaction between k.type an intervention 3	3.762	8.530	.068	.441	.661	-13.380	20.904
	interaction between a type an intervention 3	-10.707	7.629	- 194	-1.403	.167	-26.038	4.624
	interaction between v.type an intervention 3	-9.655	7.301	- 175	-1.322	.192	-24.327	5.017

a. Dependent Variable: change score for unit 3

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Conclusion about the Influence of New Method

- On average, the new method improves students' performance of score change.
- For students with a high value on t, they benefit more than other students.

Part 3: Analysis (Section 2)





Does the new teaching method especially help any of the learning types?

Variables of interest:

Learning Types - Binary Variable (1 if 60+, 0 if less than 60): Tactile Kinesthetic Auditory Visual Teaching Method - Binary (1 if New, 0 if Control)

Score Difference between Pre- and Post-Tests for each Unit

Question 2: Type of Analysis

Linear Models

For each unit: Score diff \sim T + K + A + V + Method + two-way interactions

Two-Sample T-Tests

For each unit and learning type, compare the average score difference for the new and control methods:

 $H_0: \mu_{new} = \mu_{control}$ $H_1: \mu_{new} > \mu_{control}$

Question 2: Two-Sample T-Test Assumptions

Independent Samples

- Students independently took pre- and post-tests
- Possible within-class correlation

Normality Assumption (Shapiro-Wilks Test)

- 20 of the samples met the normality assumption
- 2 samples failed to meet the assumption
- 2 samples had fewer than 3 data points

Variance

Samples had unequal variances, which we accounted for when running the t-tests

Question 2: Two-Sample T-Test

Bonferroni Adjusted P-Values

	Т	K	А	V
Unit 1	0.0028	0.7444	0.9480	1.0000
Unit 2	0.0001	0.2326	0.0423	0.9994
Unit 3	0.0000	0.0002	0.1716	0.0259

Conclusions: The new teaching method is helpful for..

- Strong tactile learners, for all three units.
- Strong auditory learners for Unit 2.
- Strong kinesthetic and visual learners for Unit 3.