

I will be presenting a method that we used in the Head Start on Science project.

The method is called Minimization, and it is an alternative to pure randomization. (Is the Gold Standard always the best? Are there other options?)

In the presentation I will go over some of the background regarding the project. Then I will go into issues with which we were confronted with and our approach to deal with them. And, finally I will demonstrate how this method was implemented with the chosen software.



The following are a few characteristics of our study. Each of these characteristics (among others that we will be discussing later in the presentation) influenced the selection of the method used in the allocation of classrooms into the two groups: Intervention and Control.

-The project is an intervention of Head Start teachers to improve science education.

-The sample will consist of two cohorts. Each of the cohorts will contain one-half of the total sample.

-The study design is a multi-site cluster randomized trial. More on these on the next slide.

We started our task with two questions:

1. What is the best research design for our study?

Is there a naturally embedded structure in our study.

2. How can we create balanced groups? (i.e., Intervention & Control)

Balanced allocation considering the research design and the research question of interest.

Why are these important?

- a. Understand the task at hand. Provides the bigger picture.
- b. Identify models for analyses.
- c. Ability to attribute the results to the impact of the intervention.

d. Better chance at avoiding competing explanations to our results.

e. Want to avoid any confounding variables that we might not be aware of and might become a nuisance to our study and our conclusions.

f. The properties of well balanced groups provide us with an opportunity to get as close as possible to causal inference regarding the effects of the teacher intervention.



Our study has a multilevel structure with programs (sites) that contain classrooms, and classrooms (clusters) that contain students.

It is important to point out that the intervention is at level 2. But the study is gathering information from teachers, students and parents. That is, The study will be gathering, studying, and analyzing effects at the teacher level and the child level.

We are using multiple sites in our study for each of the following reasons:

- 1. One program would not have enough classrooms for our study.
- 2. Multiple programs allow us to sample areas with different demographics.
- 3. It will give us a stronger argument to generalize our results.

D. Reyes-Gastelum, S.J. Pierce, L. Van Egeren



The intervention in our study happens at the classroom level, since it is a teacher intervention.

This informs us that we should think about the best way to allocate teachers into our two groups: Intervention and Control.

How can we do this?

What options do we have?

What is the best method to use?



Should we randomize? This is the gold standard.

If we had the money and personnel then we could use a large sample in our study. For example, we could take a sample of 100 programs and 1000 classrooms (ten classrooms per program), this is just an example. By randomly allocating classrooms to intervention/control group, flipping a 50/50 coin, this would probably balance out the two groups, that is, the two groups would be quite similar in important characteristics. This would be an ideal situation, but in most studies this is not the case.

What do we have in our study? SMALL SAMPLE Before we go over the sample size of our study.

Let us think about a simple example of why common randomization is NOT a good method for small samples.

If we just randomly assign classrooms, then many things could happen regarding the intervention/control groups.

Let us think about what can go wrong when we are working with small samples.

EXAMPLE: If we have 3 classrooms in one program, then we would have 8 possible group assignments (each classroom has two possibilities). And, two of these possibilities assign all three classrooms to the same group (all intervention or all control). So, in this case we have a 25% probability that all classrooms in the program would be allocated to the same group.

Why is this a problem? We would not be able to compare the intervention and control groups for this program (the one with all classrooms having the same allocation). We would not be able to conclude that the intervention had the same effect in this program as in the rest of the programs. There could be some special characteristics about this program that would be confounded in our

results, and this could present obstacles to our results at the conclusion of our study. NOT GOOD

Head Start on Science Project

Each cohort:

- 4 programs
- 9 classrooms (per program)
- Students in the classroom

Total per cohort:

36 classrooms

What do we have in our study? AND What do we do?

Based on a power analysis performed, we had that each cohort would have 4 programs.

And, that each program would have 9 classrooms (on the average).

Assuming that the study would have a balanced design. Of course, this is thinking that everything works out perfectly in real life.

What can we do with small samples?

Before we go further, we should ask ourselves the following questions:

- 1. What is the primary goal of the study/project?
- 2. What are the constrains/limitations with which we are going to be confronted with?
- NOTE: This implies that you need to talk to your colleagues involved in the project. And, that statistics is not a magical tool that you can use blindly (one method that works all the time).



Compare outcomes between the classrooms in the intervention group and control group

This is based on the broader view of the main research question of the study and also based on our research design.

- Based on this goal and discussions from the project team, we identified important features that we should attempt to have in our allocation method. (What are these?)
- Is there more to the research question?



There is, in the analyses we want to be able to account for variation and correlations embedded in our design.

Are there any other desired characteristics that we would like for our allocation?

D. Reyes-Gastelum, S.J. Pierce, L. Van Egeren



After team discussions we identified the following seven features that we wanted our allocation method to have.

- 1. This feature increases statistical power to detect intervention effects.
- 2. To measure the effect of the intervention at the program level, and to assess the stability of the intervention effects across programs.

D. Reyes-Gastelum, S.J. Pierce, L. Van Egeren



We want our allocation to provide balanced groups on important characteristics.

Note:

The covariates (variables) were informed by research. These covariates are related to target outcomes, because having comparable groups at baseline allows for stronger inferences about intervention effects.



We would like to have a method that does not create a biased allocation.

(Selection bias refers to the bias arising from being able to predict the group to which a classroom would be allocated.)

And

We would like to be able to constrain allocations. What does this mean? In our case, this refers to the following: classrooms in the same building should be allocated to the same group, to prevent spill-over (contamination between intervention and control classrooms).



These are two competing features. Is this possible? We might need a few tradeoffs to (partially) satisfy these features. Desired key features for the allocation method:

- Allow sequential assignment.
- Method consistent with the Consolidated Standards for Reporting Trials (CONSORT).

Sequential Assignment:

We wanted to have the option of replacing classrooms. In the case of high attrition. (The power analysis accounted for attrition, so small attrition would not be a problem.)

CONSORT:

We wanted our method to be within the accepted practices for conducting randomized controlled trials described in the consolidated standards for reporting trials (CONSORT) statement.

Before going into our allocation method. Let us explain what CONSORT is and where you can find more information.



These standards were developed to improve the implementation and reporting of randomized controlled trials within the medical community.



IMPORTANT: Selected allocation method that made appropriate trade-offs. And Software that allowed us to control and implement all the desired features.



How does this method compare to the "Gold Standard".

Remember, true randomization (and most methods) only work for large samples

(to guarantee

equal group sizes and

good covariate balance).

MINIMIZATION:

Deterministic (Trade-off / Bad Feature):

How do we deal with this?

(We selected a Biased Coin minimization procedure to deal with the deterministic feature of the algorithm.)

How does minimization work?

Principles of Minimization:

- The first classroom is allocated to a group at random.
- For each subsequent classroom we determine which group allocation would lead to better balance between the groups in the variables of interest.

What are the big concepts associated with Minimization?

The method follows the two given principles.

- 1. It randomly assigns the first classroom. (Flip of a 50/50 coin.)
- 2. It creates a score for each group based on currently allocations, and based on this score it determines which assignment for the subsequent classroom would provide the best balance.

This has two consequences:

- 1. The order in which the classrooms are allocated matters.
- 2. Once the order is determined then the allocation is deterministic.

How can this be modified (or implemented) to get a better allocation (based on our desired) features?

Principles of Minimization:

- The first participant is allocated a treatment at random.
 - The order is randomized.
- For each subsequent participant we determine which group allocation would lead to better balance between the groups in the variables of interest.

- Introduce a biased coin.

What did we do to deal with this?

1.We randomly assigned the order.

2.We used a biased coin.

Did we satisfy our seven desired features?

First, we should mention that this method is consistent with CONSORT. We will go over the other desired features as we go through the software implementation.



The overall implementation is illustrated by this diagram:

- 1. We recruit classrooms.
- 2. We obtain some baseline data, and use this data to obtain our covariates.
- 3. Take the list of classrooms and the covariates, and we put them into the allocation algorithm.
- 4. Obtain the two groups: Intervention & control.



The covariates (variables) were informed by research. These covariates are related to target outcomes, because having comparable groups at baseline allows for stronger inferences about intervention effects.



So how did we implement this algorithm and what does the software allow us to do?

Note:

We applied the algorithm to each program separately.

Sattings Groupy Vanishet Allocations Table Twit Trick UN to Report Description: Diversition Description: Diversition Sample Sac: Q I Simple Sac: Q I Randomization of Program-MoRgest in Cohort 1 Vanishiet: 1 CASS 2. Control: Athinute - Factor Analysis with Reduced Sit of Rems 1	0 8 a D Network Syr	Sample Size: 00 3	даа		es Allocations Table Balance Help	Settings Groups Variable:
Probability Method () Dutance Mesure () Trail Properties () Margin Laboration () Margin () Margi						Trial Title: ① No Repeat Description: ①
Probability Method @ Distance Messaure@ Teal Properties@ Walve Add Disase Coin Minimization @ Minipical Balance @ Name Disase Monthality @ 0.75 B @ Disase @ Overanie @ 2 Disase Probability @ 0.75 B @ Disase				1	n-NoRepeat in Cohort 1 r Analysis with Reduced Set of Items	Randomization of Program- Variables: 1. CLASS 2. Science Attitude - Factor
Probability Method (I) Dutance Mesure (I) Teil Properteis (I) Enter Con Menimization (I) Marginal Balance (I) Name (I) N				-		
Probability Method @ Distance Messaurc@ Trial Properties: @ Walve Add © Based Coin Minimization @ Minipiral Balance @ Name © Nave Minimization @ Sundrod Overation @ 2 © Savard Overation @ Sundrod Overation @ Detere © Variance: @ Data						
Probability Method 0 Distance Massawe 0 Trail Progenities 0 Vision Add Destination Add Destination Add Destination 0 Narrow Add Destination 0 Starce 0 Narrow 0 Distance 0 Destination 0				_		
C Nave Minimization C Range @ Range @ Determined Development @ Determin	bbA		Value	0	Distance Measure: ③ Trial Propertie ion ④ Marginal Balance: ④ Name	Probability Method: 3 Biased Coin Minimizatio
O Variance: O Date	Delete			2	Range: Range: Standard Deviation:	 Naive Minimization Base Probability:
	Date) -	🔿 Variance: 🗓	
			Value	2	Distance Measure: ① Trial Propertie ion ③ Marginal Balance: ① Name ○ Range: ① ③ Standard Devisition: ① ↓ Variance: ①	Probability Method: ① ③ Biased Coin Minimizatio ③ Naive Minimizatio Base Probability: ① 0.75

Since the fonts are small we will be focusing on these two areas of the screen.



- On the top you can see that the software has multiple tabs. We will be going over each one of these during the demonstration.
- This area allows you to enter any type of descriptions that will be useful later on, in case we have to go back and look at our process. This will remind us of what we did in the process.



- This section allows us to select the probability of the biased-coin. The number we chose was based on previous research.
- And, we can also select the desired distance measure. With this option we are selecting the formula that will be creating scores for the groups, and that will be used to balance the groups (Minimization).

Note:

We can see that the screen has many i's encapsulated in circles. What are these? Let us see at the one next to "Probability Method".

Related to our desired features:

We used the biased coin variation of the minimization algorithm. It prioritizes the balancing allocation, but still adds some uncertainty to the allocation process.



The software provides information regarding the options provided help us select the best options for our study.

File Tools Help Settingi Groups Variables Allocations Table Balance Help Group Name Tradement Tradement Control	Add -	
Settings Groups Variables Allocations Table Balance Help Group Name Treatment	Allocation Ratio	
Treatment		
Contor	1	
Settings Groups Variables Allocation	s Table Balance Help	
reatment		(I)

On the next tab, "Groups":

We assign names to our groups and provide the desired allocation ratio.

Related to our desired features:

- 1. This provides the desired balance in size (both groups will have similar number of classrooms).
- 2. Used the algorithm in each of the programs. So the balance in size applies to each program, and also overall.

variable	S			
000		HSQS Server		
Minimization Program [No Repeat] File Tools Help		÷.		- 8 💌
Settings Groups Variables Allocations	Table Balance Help	Add		
Variable Name CLASS	Weight	0	Levels Low - CLASS,High - CLASS	
Variable Name CLASS				
Science Attitude - Factor	Analysis		-	
		•		
	Weight			Levels
	1.000000			Low - CLASS,High - CLASS

The next tab, "Variables":

Here we can enter our covariates and the acceptable values for each of them. Important: You also have the option to assign weights to each of the covariates, this provides an opportunity to rank the importance of each them if needed (or desired).

Related to our desired features:

1. The covariates were used as balancing criteria.

000			11505 5		
Minimization Program [No	o Repeat]		insus server		- 6 ×
File Tools Help			🚔 👻		
Settings Groups Variables	Allocations Table Balance He	10	-		
Add Delete	Refresh		ω		
#	• uī	4 Group	 CLASS 	Science Attitude - Factor Analysis	•
1	2	Control	Low - CLASS High - CLASS	Low - Sci Att	
2	3	Treatment	High - CLASS	High - Sci Att	
3	4	Control	Low - CLASS	High - Sci Att	
4	6	Control	Low - CLASS	Low - Sci Att	
5	1	Control	Low - CLASS	Low - Sci Att	
7	0	Treatment	Low - CLASS	Low - Sci Att	
			1		
		Ū	¥		
1 0			¥		
 Group 		CLASS	¥	 Science Attitude - Fa 	ctor Analy
 Group Treatment 		CLASS Low -	CLASS	 Science Attitude - Fa Low - Sci Att 	ctor Analy
Group Treatment Control		 CLASS Low - High - 	CLASS CLASS	 Science Attitude - Fa Low - Sci Att Low - Sci Att 	ctor Analy
Group Treatment Control Treatment		CLASS Low - High - High -	CLASS CLASS CLASS CLASS	 Science Attitude - Fa Low - Sci Att Low - Sci Att High - Sci Att 	ctor Analy
Group Treatment Control Treatment Control		CLASS Low - High - High - Low -	CLASS CLASS CLASS CLASS CLASS CLASS	 Science Attitude - Fa Low - Sci Att Low - Sci Att High - Sci Att High - Sci Att High - Sci Att 	ctor Analy
Group Treatment Control Treatment Control Control		CLASS Low - High - High - Low - Low -	CLASS CLASS CLASS CLASS CLASS CLASS CLASS	 Science Attitude - Fa Low - Sci Att Low - Sci Att High - Sci Att High - Sci Att Low - Sci Att 	ctor Analy
Group Treatment Control Treatment Control Control Control		CLASS Low - High - High - Low - Low - Low -	CLASS CLASS CLASS CLASS CLASS CLASS CLASS CLASS	 Science Attitude - Fa Low - Sci Att Low - Sci Att High - Sci Att High - Sci Att Low - Sci Att Low - Sci Att Low - Sci Att 	ctor Analy
Group Treatment Control Treatment Control Control Control Treatment		CLASS Low - High - High - Low - Low - Low - Low - Low -	CLASS CLASS CLASS CLASS CLASS CLASS CLASS CLASS CLASS	 Science Attitude - Fa Low - Sci Att Low - Sci Att High - Sci Att High - Sci Att Low - Sci Att Low - Sci Att Low - Sci Att Low - Sci Att 	ctor Analy

The "Allocation" tab:

We can see the allocation of the classrooms that have already been entered. This current allocations are used to determine the allocation for the next entered classroom.

Related to our desired features:

- 1. The software allows us to force the allocation (override if needed), and then based on the current groups the next classroom should be allocated following the algorithm.
- 2. The algorithm works in a sequential manner.

000					
Minimization Program	[No Repeat]		HSOS Server		
File Tools Help					
Settings Groups Variabl	les Allocations Table Balance Help		Add		
Group	Low - CLASS	High - CLASS	Low - Sci Att	High - Sci Att	Total
Treatment	3	1	3	1	4
Control	3	1	3	1	4
	X				
Settings	Groups Variable	s Allocations	Table Balance H	Help	
Settings	Groups Variable	s Allocations	Table Balance H - CLASS	Help	High - CLAS
Settings Group Treatment	Groups Variable	s Allocations Low 3	Table Balance H - CLASS	Help	High - CLAS
Settings Group Treatment Control	Groups Variable	s Allocations Low 3 3	Table Balance H - CLASS	Help	High - CLAS 1 1

On this tab, "Table": We obtain a breakdown of our covariates by group.

Table	Settings Groups Variables	Allocations	Table Balance	Help		
lable	Group	Lo	w - CLASS		High	n - CLASS
	Treatment	3			1	
	Control	4			0	
	Total	7			1	
	CLASS		0.571429	1.000000	0.500000	0.707107
Dulunioc	E CLASS		0.571429	1.000000	0.500000	0.707107
	Low - CLASS		0.142857	1.000000	0.500000	0.707107
	High - CLASS		1.000000	1.000000	0.500000	0.707107
	Science Attitude - Fact	or Analysis	0.266667	1.000000	0.500000	0.707107
	Low - Sci Att		0.333333	1.000000	0.500000	0.707107
	High - Sci Att		0.200000	1.000000	0.500000	0.707107
	Mean		0.419048	1.000000	0.500000	0.707107
			1 000000	1 000000	0 500000	0 707107

- This is a different example, in this program the variable did not balance out perfectly (due to randomness and the number of classrooms with values in each of the covariates).
- On the "Balance" tab, we can see the results of the "Marginal Balance" minimization formulas.

Sample Descriptive	Characteristics (Cohort 1)
--------------------	----------------------------

	Intervention N (%)	Control N (%)	Total N (%)
Classrooms	18 (51)	17 (49)	35 (100)
Teachers	35 (51)	34 (49)	69 (100)
Teacher type		· · · ·	
Lead teacher	18 (51)	17 (50)	35 (51)
Teaching assistant	17 (49)	17 (50)	34 (49)
Gender			
Female	34 (97)	33 (97)	67 (97)
Male	1 (3)	1 (3)	12 (3)
Race			
African American	6 (17)	7 (21)	13 (19)
White	24 (69)	21 (62)	45 (65)
Other	4 (11)	5 (15)	9 (13)
Missing	1 (3)	1 (3)	2 (3)
Education			
Master's degree	3 (9)	1 (3)	4 (6)
Bachelor's degree	11 (31)	13 (38)	24 (35)
Associate degree	13 (37)	7 (21)	20 (29)
High school degree/GED	8 (23)	13 (38)	21 (30)
	M(SD)	M(SD)	M (SD
Experience in early childhood programs	9.9 (6.6)	11.4 (8.6)	10.6 (7.6)
(years)			

This table illustrates what happened to all the other variables we were not controlling for in the minimization algorithm.

References

- Chen, H.-F., & Green, K. E. (2010). Minimization as an alternative to unrestricted randomization in educational research. *Educational Research Quarterly*, *34*(1), 3-17.
- Moher, D., Hopewell, S., Schulz, K. F., Montori, V., Gotzsche, P. C., Devereaux, P. J., . . . Altman, D. G. (2010). CONSORT 2010 explanation and elaboration: Updated guidelines for reporting parallel group randomized trials. *Journal of Clinical Epidemiology*, *63*(8), e1-e37.
- Pianta, R. C., La Paro, K. M., & Hamre, B. K. (2008). Classroom assessment scoring system (CLASS) manual, pre-k. Baltimore, MD: Paul H. Brookes Publishing Co., Inc.
- Pocock, S. J. (1983). *Clinical trials: A practical approach*. New York, NY: John Wiley & Sons.

References.

D. Reyes-Gastelum, S.J. Pierce, L. Van Egeren

MICHIGAN STATE

- Saghaei, M., & Saghaei, S. (2011). Implementation of an open-source customizable minimization program for allocation of patients to parallel groups in clinical trials. *Journal of Biomedical Science and Engineering*, 4(11), 734-739.
- Schulz, K. F., Altman, D. G., & Moher, D. (2010). CONSORT 2010 Statement: Updated guidelines for reporting parallel group randomised trials. *Journal of Clinical Epidemiology*, 63(8), 834-840.

Scott, N. W., McPherson, G. C., Ramsay, C. R., & Campbell, M. K. (2002). The method of minimization for allocation to clinical trials: a review. *Controlled Clinical Trials*, 23(6), 662-674.

- Taves, D. R. (2010). The use of minimization in clinical trials. *Contemporary Clinical Trials, 31*(2), 180-184.
- Torgerson, C. J., & Torgerson, D. J. (2007). The use of minimization to form comparison groups in educational research. *Educational Studies*, 33(3), 333-337.

References.



Co-PIs and project information.



Contact Information.