## **Prepare Your Data for Randomization**

#### **REQUIRED**

At minimum, you must include a list of users with an individual identifier (ID) for each individual, class, or school that will be randomly assigned.

Option 1: If you are assigning individuals, your data set should include an ID for each student or teacher.

Option 2a: If you are assigning groups (classes or schools), your data set should include student and class or school IDs.

Option 2b: If you are assigning groups, but you do not yet have individual information or won't need it for your final analysis, your data set should include only class or school IDs.

AnonStudentID	AnonStudentID	SchoolID	SchoolID
159508	159508	100	100
694677	694677	100	101
807588	807588	100	102
482489	482489	100	103
555123	555123	100	104
124226	124226	200	201
232721	232721	200	202
834305	834305	200	203
490514	490514	300	301
573401	573401	300	302
275321	275321	300	303
288475	288475	300	304

Each row represents a single observation (student, teacher, school, and so on). Each observation must have an individual identifier. This is a unique code for each participant that will be assigned to the technology user group or the comparison group. The identifiers will enable you to determine who will use the technology, so participants can be notified. They will also enable you to combine (merge) data sets.

#### **RECOMMENDED**

The Coach recommends that you include in your data set pre-test data and background characteristics. You will use this information to make sure that the randomly assigned treatment and comparison groups are similar before you introduce the technology. If you include these additional variables, your data set will look like this:

Missing data is replaced with NA so that the Coach doesn't include a value for that observation.

Categorical variables must be converted to binary variables with two possible values: 1 instead of yes and 0 instead of no.

AnonStudentID	SchoolID	Grade	Fall_score	Female	EL	Low_SES	Black	White	Asian	Other
159508	100	3	320	1	1	1	1	0	.0	0
694677	100	3	450	.0	.0	0	.0	1	.0	0
807588	100	4	NA	1	.0	.0	1	.0	.0	0
482489	100	4	410	.0	1	0	.0	0	1	0
555123	100	4	534	1	1	1	.0	0	1	0
124226	200	3	604	.0	.0	0	1	.0	.0	0
232721	200	.3	378	1	.0	1	.0	.0	.0	1
834305	200	4	NA	1	1	1	.0	.0	1	0
490514	300	3	380	0	1	0	1	0	0	0
573401	300	3	468	.0	.0	1	.0	0	.0	1
275321	300	4	523	1	.0	0	.0	.0	1	0
288475	300	3	375	0	0	.0	.0	1	0	0

SES = socioeconomic status.

Each row represents a single observation (student, teacher, school, and so on).

Each variable has its own column

**NOTE:** You can work with data in a number of different programs. If you do not have access to statistical software, you can use Microsoft Excel to prepare your data. In the rest of this guide, we have included some tips that will help you manage your data in Excel.

#### STEP 1. IDENTIFYING DATA SOURCES

You should use several types of data to create this initial data set. A checklist and description of the data follows. Examples at the end of this guide show what each data set should look like.

- List of users who will be randomly assigned (required). Compile a list of all potential technology users. For example, if you plan to randomly select students to use the technology, you will need an identifier for every student who might be assigned to either the technology user group or the comparison group. If the students are in different schools or different classrooms, you should also include an identifier for the classroom or school. If you plan to randomly select teachers or schools, you will need an identifier for each teacher or each school. (Remember, larger sample sizes are better.)
- Pre-test data (recommended). If available, you should include an outcome measure from before the intervention (such as an assessment from the beginning of the school year). The Coach will double-check that the two groups are equivalent before giving you the final list of assignments. If the groups are not balanced, the Coach will rerandomize the list of users until balance is achieved.
- Background characteristics (recommended). Background characteristics provide data on
  observable traits for each participant. These could include gender, ethnicity, individualized
  education program status, English Learner (EL) status, socioeconomic status (SES), and more.
  If you include background traits in your data set, the Coach can make sure your intervention
  and comparison groups are well balanced before introducing the technology.

**CAUTION:** Some of these characteristics, such as EL status, can change over time. It is preferable to measure and record all background characteristics before introducing the educational technology.

#### STEP 2. PROCESSING YOUR DATA

When you have identified the data elements and data sources, the second step is to combine all data elements into one **tidy data**<sup>1</sup> set and prepare the variables that will be used for analysis. (We explain how to do this below.) We recommend generating tidy data sets not only because doing so is a requirement to use the RCE Coach and most statistical software packages for analyses, but because a tidy data set is easy to manipulate, model, and visualize. The data set at the beginning of this guide is an example of a tidy data set. This section will take you through a series of questions that will help you create your own.

A. Is each observation a row and each variable a column?

<sup>&</sup>lt;sup>1</sup> Having **tidy data** means that you've used a standardized way to structure your data set. Specifically: Each variable forms a column; each observation forms a row. If you want to learn more about tidy data you can refer to Wickham, H. "Tidy Data." Journal of Statistical Software, vol. 59. No. 10, 2014, pp. 1–23. doi:http://dx.doi.org/10.18637/jss.v059.i10

Exhibit 1. Example observation

AnonStudentID	Treatment	Fall_score	Gender
159508	1	320	F
694677	0	450	М
807588	0	999	F
482489	1	410	M

**NO:** Reorganize your data set so that each row represents one observation. Each variable you are interested in should be its own column.

YES: Continue on to B.

B. Do you have one data set that contains all of the variables you will need?

NO: You will have to merge your existing data sets into one complete data set. This will be easy to do using the unique identifiers. If you are using Excel to manage your data, you can do this using a VLOOKUP function.

**CAUTION:** Some observations could be present in some data sets but not in others. Therefore, when merging these data sets you might introduce some missing data. For example, a student in a data set consisting of test scores might not exist in another data set and therefore could have missing data for other variables (such as background characteristics) after combining the data sets.

YES: Continue on to C.

C. Are all of the categorical variables that will be used for your analysis numeric?

**NOTE:** If you are using a statistical software package, you will want to make sure that all of these variables are recognized as numeric values and not string or character values.

**NO:** Convert all of your categorical variables, or variables that include names or labels, into numbers. This might mean you have to change a variable into a binary, or dummy, variable. A dummy variable uses 1 to indicate yes or that a condition was met and 0 to indicate no or that a condition was not met.

For example, if your background characteristics include gender as a variable, you might have male or female, or M or F, in each cell of that column. Instead, you should change the variable from Gender to Female, and change each cell that indicates the participant is a female to 1, and each cell that indicates the participant is male to 0. You can do this for every variable that is non-numeric.

## Exhibit 2. Example categorical variables

Student ID	Gender	Student ID
159508	F	159508
694677	М	694677
807588	F	807588

**NOTE:** If your categorical variable contains more than two options, such as Race—where the options are (1) Asian, (2) Black, (3) White, and (4) Other—you will have to create a binary or dummy variable for each option. For example, Asian would be one column (with 0 representing non-Asians and 1 representing Asians), Black would be a second column, White would be a third column, and Other would be a fourth column (Exhibit 6).

YES: Continue on to D.

D. Are all missing data coded consistently in your data set?

NO: If you have merged data sets and data are missing, make sure you are consistently coding that as NA (not available) to ensure that the Coach can analyze your data. You want to be extra careful to make sure missing data have not been given a numerical designation, such as 0 or 999. These values will get incorporated into the analysis. If you are using your own statistical software, such as SAS or Stata, it will be helpful to code the missing data as a period (.)

**YES:** Congratulations! You have a tidy data set!

#### **Step 3.** Checking the quality of your data

After constructing your data file and converting your variables, the final step is to check the quality of your data. You can run the following checks to identify potential data issues that warrant additional investigation:

Check the minimum and maximum values of variables. This check can help to identify extremely low or high values that are outliers in your distribution or that signal a special missing code that must be converted to a missing value. You might want to check with someone who is familiar with the data to confirm the value range makes sense.

**NOTE:** If you are working in Excel, you can use MIN and MAX functions to easily find these values.

Consider the impact of missing data. The Coach, and some statistical software packages, will automatically drop observations that contain missing data. You should try to understand why data are missing and how excluding students with incomplete data will affect your results.

**NOTE:** If you are working in Excel, you can sort and filter your data to view missing values. To determine exactly how many values are missing for a single variable you can use the COUNTIF function; to determine how many observations have at least one missing value, you can use a nested COUNTIF with OR function.

### **Example data sets**

## Exhibit 3. Data set 1a: List of participants

Student_name	AnonStudentID	School	SchoolID
Guillermo Gonzalez	159508	Alan Elementary	100
Robert Rice	694677	Alan Elementary	100
Sophia Smith	807588	Alan Elementary	100
Patricia Pacheco	482489	Alan Elementary	100

You will need identifiers for each participant. If you are randomizing groups of participants (classes or schools), you must also include a group ID.

# **Exhibit 4.** Data set 1b: List of participants without personally identifiable information

Eliminate personally
identifiable information,
such as student and
school names, from the
data that you upload to
the Coach.

AnonStudentID	SchoolID
159508	100
694677	100
807588	100
482489	100

## Exhibit 5. Data set 2: Test scores

AnonStudentID	Fall_score	
159508	320	
694677	450	These are
807588	NA	missing values.
482489	NA NA	values.

**Exhibit 6.** Background characteristics (with non-numeric and numeric categorical variables)

AnonStudentID	SchoolID	Gender	EL status	SES	Race
159508	100	F	EL	Low	Black
694677	100	M	Not EL	Medium	White
807588	100	F	Not EL	High	Black
482489	100	M	EL	High	Asian
555123	100	F	EL	Low	Asian
124226	200	М	Not EL	Medium	Black
232721	200	F	Not EL	Low	Other
834305	200	F	EL	Missing	Missing

Data set 3a:
Background
characteristics
(with nonnumeric
categorical
variables)

Data set 3b:
Background
characteristics
(with numeric
categorical
variables)

EL = English learner; SES = socioeconomic status.

AnonStudentID	SchoolID	Female	EL	Low_SES	Black	White	Asian	Other
159508	100	1	1	1	1	0	0	0
694677	100	0	0	0	0	1	0	0
807588	100	1	0	0	1	0	0	0
482489	100	0	1	0	0	0	1	0
555123	100	1	1	1	0	0	1	0
124226	200	_ 0	0	0	1	0	0	0
232721	200 /	_ 1	0	1	0	0	0	1
834305	200	1	1	NA	NA	NA	NA	NA

EL = English learner; SES = socioeconomic status.

The Coach requires that all categorical variables are converted to binary or dummy variables with values of 0 and 1. In the case of SES, you can choose to group medium and high SES together (1 = low and 0 = medium or high) if you are only interested in the effect of low SES on your outcome of interest.

**Note:** Research shows that students of higher SES often have an academic advantage over students of lower SES because of differences in early education access, home enrichment, levels of stress, food access, and many other factors. Therefore, if possible, it is good to check that the two randomly assigned groups are balanced on SES. However, some schools or districts might not have access to SES measures or might not be able to use them in this evaluation because of privacy concerns.

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