Chapter 6

Analysing PIAAC Data with the IDB Analyzer (SPSS and SAS)

Andres Sandoval-Hernandez and Diego Carrasco

Abstract This chapter provides readers with a step-by-step guide to performing both simple and complex analyses with data from the Programme for the International Assessment of Adult Competencies (PIAAC) using the IEA International Database (IDB) Analyzer. The IDB Analyzer is a Windows-based tool that generates SPSS and SAS syntax. Using this syntax, corresponding analyses can be conducted in SPSS and SAS. The chapter presents the data-merging module and the analysis module. Potential analyses with the IDB Analyzer are demonstrated—for example, the calculation of percentages, averages, proficiency levels, linear regression, correlations, and percentiles.

This chapter describes the general use of the International Association for the Evaluation of Educational Achievement's (IEA) International Database Analyzer (IDB Analyzer) for analysing PIAAC data (IEA 2019). The IDB Analyzer provides a user-friendly interface to easily merge the data files of the different countries participating in PIAAC. Furthermore, it seamlessly takes into account the sampling information and the multiple imputed achievement scores to produce accurate statistical results (see Chap. 2 in this volume for details about PIAAC's complex sample and assessment design).

This chapter is subdivided into three main sections. In the first section, we will provide a brief overview of the software.¹ Sections 6.2 and 6.3 will be dedicated

¹Most of the information for this section is adapted from the last version of the Help Manual for the IDB analyzer (IEA 2019).

A. Sandoval-Hernandez (✉)
University of Bath, Bath, UK
e-mail: A.Sandoval@bath.ac.uk

D. Carrasco
Centro de Medición MIDE UC – Pontificia Universidad Católica de Chile, Santiago, Chile

to the Merge and Analysis modules of the IDB Analyzer, respectively. For each of these two sections, we will provide a description of the functionalities of the respective modules and examples to illustrate some of the capabilities of the IDB Analyzer (Version 4.0) to merge files and to compute a variety of statistics, including the calculation of percentages, averages, benchmarks (proficiency levels), linear regression, logistic regression, correlations, and percentiles.

6.1 The IDB Analyzer

Developed by IEA Hamburg, the IDB Analyzer is an interface that creates syntax for SPSS (IBM 2013) and SAS (SAS 2012). The IDB Analyzer was originally designed to allow users to combine and analyse data from IEA's large-scale assessments, but it has been adapted to work with data from most major large-scale assessment surveys, including those conducted by the Organisation for Economic Co-operation and Development (OECD), such as the Programme for the International Assessment of Adult Competencies (PIAAC), the Programme for International Student Assessment (PISA), and the Teaching and Learning International Survey (TALIS).

The IDB Analyzer generates SPSS or SAS syntax files that take into account information from the complex sampling design of the study to produce population estimates. In addition, the generated syntax makes appropriate use of plausible values for calculating estimates of achievement scores, combining both sampling variance and imputation variance. Considering PIAAC's complex sample and complex assessment design, using either SPSS or SAS to analyse PIAAC data without the IDB Analyzer would require the user to have programming knowledge in order to create their own macros. The IDB Analyzer automatically generates these macros (syntax files) in a user-friendly environment that allows their customisation according to the purposes of the intended analysis.

The IDB Analyzer consists of two modules: the merge module and the analysis module. These two modules are integrated and executed in one common application. When working with PIAAC data, the merge module is used to create analysis datasets by combining data files from different countries and selecting subsets of variables for analysis. The analysis module provides procedures for computing various statistics and their standard errors.

Once the IDB Analyzer application is launched,² the main window will appear, as shown in Fig. 6.1. Users have then the option of choosing either SPSS or SAS as their statistical software of choice. For the examples in this chapter, we will use the SPSS software. The main window also has options to select the 'Merge Module', the 'Analysis Module', the 'Help Manual' or to exit the application.

²The latest version of the IDB Analyzer (Version 4.0) and instructions to install it are available from the IEA website: <https://www.iea.nl/index.php/data-tools/tools>.

Fig. 6.1 IDB Analyzer main window

There are at least two ways to access guidance on how to use the IDB Analyzer: video tutorials made by IEA and the main ‘Help’ manual that accompanies this software installation. An easy way to get you started with the IDB Analyzer is to watch IEA video tutorials. These have been made available at the following link: https://www.iea.nl/training#IDB_Analyzer_Video_Tutorials.

These videos have been shared via YouTube; they cover step-by-step examples of how to estimate correlations, percentiles, percentages and means, logistic regression, linear regression, and benchmarks.

A second way to get help and guidance is to consult the ‘Help’ manual via the main menu in the IDB Analyzer. This official manual can be accessed by clicking on the third button present in the main menu. Figure 6.1 shows what this main menu looks like.

The IDB Analyzer will work on most IBM-compatible computers using current Microsoft Windows³ operating systems. The IDB Analyzer is licensed free of charge and may be used only in accordance with the terms of the licencing agreement. While the IDB Analyzer is free, the user must own a valid licence for at least one of the software packages used as statistical engine (i.e., SPSS Version 18 or later or SAS Version 9 or later). Additionally, the user should have a valid licence for Microsoft Excel 2003 or a later version (as outputs are also produced in this format). The IDB Analyzer licence expires at the end of each calendar year.

³Currently there is no stand-alone Mac version of the IDB Analyzer. However, the software can be used on Mac through a virtual machine and Windows installed on it. The current version was tested using Windows installed on Parallels Desktop for Mac (<http://www.parallels.com/products/desktop/>).

So, every year, users have to download and reinstall the most current version of the software and agree to the terms and conditions of the new licence.

6.2 Merging Files with the IDB Analyzer

PIAAC Public Use Files containing both responses to the background questionnaire and the cognitive assessment are available for downloading for each of the participant countries separately. The Merge Module of the IDB Analyzer allows users to combine datasets from more than one country into a single data file for cross-country analyses. For the purposes of this chapter, we will assume all data files have been copied within a folder named 'C:\Data\PIAAC'. PIAAC data files are available in both SPSS and SAS from the PIAAC website.⁴ Users should download the data files in the format of their preference.

The Merge Module recognises the data files for PIAAC by reading the file names in the selected directory and matching them to the file-naming convention pre-specified in the IDB Analyzer configuration files. For this reason, in order to ensure that the IDB Analyzer will correctly identify the different files contained in the PIAAC datasets, as well as the user-generated files:

- Users should not change the name of the files once downloaded from the PIAAC website.
- Users should not save the merged file in the same directory where the source files are located.
- Users should keep files from different studies and years in separate directories.

The following steps will create an SPSS or SAS data file with data from multiple countries and/or multiple file types:

1. Open the IDB Analyzer.
2. Select the statistical software you want to work with (choose between SAS or SPSS).
3. Select the Merge Module of the IDB Analyzer.
4. Click the Merge Module button. The Merge Module interface is divided into two different tabs. In the first one, you can select the countries and edit country labels. In the second tab, you can select the variables you want to include in your analysis and specify the name of the merged file.
5. Under the 'Select Data Files and Participants' tab and in the 'Select Directory' field, browse to the folder where all data files are located. For example, in Fig. 6.2, all SPSS data files are located in the folder 'C:\Data\PIAAC'. The program will automatically recognise and complete the 'Select Study' and 'Select Cycle'

⁴<http://www.oecd.org/skills/piaac/>

Fig. 6.2 IDB Analyzer merge module: select data files and participants

fields and list all countries available in this folder as possible candidates for merging.

6. Click the countries of interest from the 'Available Participants' list and click the right arrow button ($\Rightarrow$) to move them to the 'Selected Participants' panel on the right. Individual countries can also be moved directly to the 'Selected Participants' panel by double-clicking on them. To select multiple countries, hold the CTRL-key of the keyboard when clicking on countries. Click the tab-right arrow button ($\Rightarrow$) to move all countries to the 'Selected Participants' panel. For this example, we selected all the countries available.
7. Click the 'Next >' button to proceed to the next step. The software will open the 'Select File Types and Variables' tab of the merge module (see Fig. 6.3), to select the file types and the variables to be included in the merged data file.
8. Select the files for merging by checking the appropriate boxes to the left of the window. For example, in Fig. 6.3, the 'General Response File' has been selected.⁵ Checking this box will automatically populate the 'Selected Variables' panel with the three scores available in PIAAC (i.e. Literacy Scale Score, Numeracy Scale Score, and Problem-Solving Scale Score), as well as with all the ID (e.g. Country ID) and sampling variables (e.g. sampling and replicate weights) needed for the corresponding analyses (Fig. 6.4).
9. Select the variables of interest from the 'Available Variables' list in the left panel. In SPSS, you can *right-click* on the variable names to open a menu with details about each of the available variables (i.e. variable name, label, measurement level, and value labels). Variables are selected by clicking on them

⁵With other studies such as PISA and TALIS, there are more options. In the case of PIAAC, there is only one option.

Fig. 6.3 IDB Analyzer merge module: selecting all countries

Fig. 6.4 IDB Analyzer merge module: select data files and participants

- and then clicking the right arrow (>) button. Clicking the tab-right arrow (⏏) button selects all variables (Fig. 6.5).
10. When selecting the variables, you can search variables by variable name or by variable label using the filter boxes (blue space between column header and list of variables) in the ‘Available Variables’ list and ‘Selected Variables’ list.
 11. Note that the IDB Analyzer assumes that files have the same structure and the variables have the same properties (e.g. variables, formats, labels) in each of these files. Any deviation from this can cause unexpected results. Should you want to modify the contents of a file for a country, or a set of countries, it is recommended to do this on the resulting merged file, after the merge is completed.

Fig. 6.5 IDB Analyzer merge module: selecting all variables

12. In the ‘Output Files’ field, click on the ‘Define’ button to specify the name for the merged data file and the folder where it will be saved. The IDB Analyzer will also create an SPSS syntax file (*.SPS) (or a SAS syntax file, *.SAS, if you are using this software) of the same name and in the same folder with the code necessary to perform the merge. In the example shown in Fig. 6.3, the merged data file ‘merge_piaac.sav’ and the syntax file ‘merge_piaac.sps’ both will be created and stored in the folder titled ‘C:\Data\’. The merged data file will contain all the variables listed in the ‘Selected Variables’ panel, and if all available variables were selected, the resulting merge file should be about 622 megabytes in size.
13. Click the ‘Start SPSS’ button to create the SPSS syntax file. An SPSS Syntax Editor window with the created syntax code will be automatically opened. The syntax file can be executed by opening the ‘Run’ menu of SPSS and selecting the ‘All’ menu option. Alternatively, you can also submit the code for processing with the keystrokes Ctrl+A (to select all), followed by Ctrl+R (to run the selection). In SAS, the syntax file can be executed by selecting the ‘Submit’ option from the ‘Run’ menu.

Once SPSS or SAS has completed its execution, it is important to check the SPSS output window or SAS log for possible warnings. If warnings appear, they should be examined carefully because they might indicate that the merge process was not performed properly and that the resulting merged data file might not include all the relevant variables or countries.

6.3 Example Analyses with the IDB Analyzer

In the following section, we will describe step-by-step instructions to produce means, percentiles, percentages, linear regressions, correlations, and benchmarks, using the latest PIAAC public-use data files. In each subsection, a sequence of steps will be included as a numbered list. These steps are reiterated for each analysis routine. In this way, each subsection is self-contained, and the reader does not need to consult any other part of the chapter to complete the steps she or he needs to follow to produce means, percentiles, percentages, linear regressions, correlations, or benchmarks.

6.3.1 Means with Plausible Values

In this section, we illustrate how to estimate the means of literacy scores by country. The first example contains a variable with plausible values. In PIAAC there are three variables with plausible values: the literacy scale scores, the numeracy scale score, and the problem-solving scale score. Each of these variables consists of ten different columns of values within the PIAAC dataset. For each test, plausible values are generated as random draws of the posterior distribution of the participant's proficiency (Wu 2005). To produce population estimates with these scores, the IDB Analyzer computes the results for each plausible value and combines these estimates using Rubin-Shaffer rules (Rutkowski et al. 2010). The following steps produce mean estimates of literacy proficiency by country, for females and males:

1. Open the IDB Analyzer.
2. Select the statistical software you want to work with (choose between SAS or SPSS).
3. Open the Analysis Module of the IDB Analyzer.
4. For this example, specify the data file 'merge_piaac.sav' as the Analysis File (see Sect. 6.2 in this chapter for details on how this file was created).
5. Select 'PIAAC (using final full sample weight)' as the Analysis Type.
6. Select 'Percentages and Means' as the Statistic Type.
7. Under the 'Plausible Values Options', select 'Use PVs'.
8. Click on the 'Separate Tables by' section at the right-hand side of the software window. This section will become active and highlighted in light yellow.
9. Go to the 'Select variables' section, and click on the 'GENDER_R' variable in the fourth row of the name list.
10. Drag the 'GENDER_R' variable to the 'Separate Tables by' section.
11. Click on the 'Plausible Values' section at the right-hand side of the software window. This section will become active and highlighted in light yellow.
12. Go to the 'Select variables' section and click on the 'PVLIT1–10' variable in the first row of the name list.
13. Drag the 'PVLIT1–10' variable to the 'Plausible Values' section.

Fig. 6.7 Analysis of mean by group output

Table 6.1 Generated files by an analysis of means

Generated files	File type	Content
means_with_pv.sps	SPSS	Syntax to run the means computations
means_with_pv.spv	SPSS	Output of the means computations
means_with_pvGENDER_R.sav	SPSS	Contains the means estimates and their standard errors
means_with_pvGENDER_R.xlsx	Excel	
means_with_pv_PVLIT_by_GENDER_R_Sig.sav	SPSS	Contains a group within-country comparison for the estimated means and percentages, providing t-statistics for these comparisons
means_with_pv_PVLIT_by_GENDER_R_Sig.xlsx	Excel	

the percentage of missing cases in the variables involved in the analysis (‘PVLIT1-10’ and ‘GENDER_R’).

The IDB Analyzer creates six files after an analysis of means with plausible values is complete. Table 6.1 details these files and their content.

Using the results provided in the file ‘means_with_pvGENDER_R.xlsx’, we created Table 6.2 to present the computed results. Means are presented and their standard errors are included in parenthesis.

The IDB Analyzer produces a ‘Table Average’, which contains an overall mean between all countries, with its standard error. These estimates are presented in Table 6.2 in the last row, in the second column. The illustrated routine can be

Table 6.2 Means of literacy scores for female and males in each country

Country	Female	Male	Country	Female	Male
Austria	267.39 (0.93)	271.53 (1.04)	Korea, Republic of	269.43 (0.87)	275.72 (0.75)
Belgium	272.81 (1.08)	278.09 (0.97)	Lithuania	268.47 (1.20)	264.97 (1.32)
Canada	272.19 (0.78)	274.49 (0.86)	Netherlands	280.92 (0.94)	287.06 (1.08)
Chile	216.36 (2.77)	223.94 (2.48)	New Zealand	280.69 (1.06)	280.66 (1.20)
Cyprus	269.60 (0.97)	267.99 (1.18)	Norway	276.43 (0.91)	280.34 (0.97)
Czech Republic	272.32 (1.30)	275.68 (1.26)	Poland	270.08 (0.86)	263.66 (0.97)
Denmark	271.00 (0.80)	270.58 (1.03)	Russian Federation	277.37 (2.88)	272.90 (2.98)
Estonia	276.64 (0.81)	275.06 (1.09)	Singapore	253.89 (1.01)	261.42 (0.98)
Finland	289.15 (0.99)	285.96 (1.21)	Slovak Republic	274.22 (0.82)	273.47 (0.86)
France	262.23 (0.69)	262.05 (0.87)	Slovenia	257.67 (0.99)	255.17 (1.08)
Germany	267.21 (1.19)	272.35 (1.17)	Spain	249.45 (1.04)	254.11 (1.00)
Greece	256.25 (1.23)	251.44 (1.54)	Sweden	277.54 (1.10)	280.88 (1.08)
Ireland	265.43 (1.10)	267.71 (1.17)	Turkey	220.89 (1.35)	231.98 (1.56)
Israel	255.04 (0.96)	255.45 (1.14)	United Kingdom	271.03 (1.29)	273.90 (1.37)
Italy	250.61 (1.32)	250.36 (1.50)	United States	269.47 (1.33)	270.16 (1.21)
Japan	294.69 (1.01)	297.78 (0.88)	Table Average	266.34 (0.22)	267.96 (0.23)

replicated with the numeracy scale scores and with the problem-solving scores present in PIAAC study.

6.3.2 Means with Other Variables

In the following example, which is simpler than its previous counterpart, we compute the mean of total years of schooling in each country. In the PIAAC study, total years of schooling was derived using different responses of participants regarding their educational participation during their lifetime. These values can be found in the ‘YRSQUAL_T’ variable. Using the IDB Analyzer, we need to follow the next steps:

1. Open the IDB Analyzer.
2. Select the statistical software you want to work with (choose between SAS or SPSS).
3. Open the Analysis Module of the IDB Analyzer.
4. For this example, specify the data file ‘merge_piaac.sav’ as the Analysis File (see Sect. 6.2 in this chapter for the details of how this file was created).
5. Select ‘PIAAC (using final full sample weight)’ as the Analysis Type.
6. Select ‘Percentages and Means’ as the Statistic Type.
7. Under the ‘Plausible Values Options’, select ‘None Used’.
8. Click on the ‘Analysis Variables’ section at the right-hand side of the software window. This section will become active and highlighted in light yellow.

9. Go to the 'Select variables' section, and under the 'Description' heading click on it, and type in 'total years'. This action would look for all the variables containing 'total' and 'year' in their description field.
10. Specify the variable YRSQUAL_T as the analysis variable by clicking the 'Analysis Variables' field to activate it. Select YRSQUAL_T from the list of available variables present in the 'Select Variables' section and move it to the 'Analysis variables' by clicking the right arrow button in this section.
11. The Weight Variable is automatically selected by the software. SPFTWT0 is selected by default; this variable contains the final sampling weight.
12. Specify the name and the folder of the output files in the 'Output Files' field by clicking the Define/Modify button. For this example, we use the term 'mean'.

After all these steps, the reached setup should look similar to Fig. 6.8:

13. Then, click the 'Start SPSS' button. This will create an SPSS syntax file and open it in an SPSS editor window.
14. To start the computations, one needs to press the following key combinations: CTRL+A first, to select the entire generated code present in the syntax window, and then CTRL+R to run these commands. The output of these analyses is depicted in Fig. 6.9.

Similar to the previous example, the generated output presents several columns. The first column is the list of countries. In the second column is the nominal sample size of each country. Notice that Austria and Germany do not have observations for this variable and present 100% of missing. The third column contains the sum of survey weights, which represents the survey population size (Heeringa et al. 2009), and in the fourth column, the IDB Analyzer includes the standards errors of the survey population size. In the 'Percent' column, the proportion of the survey population size is depicted. For example, the United States projects its number of cases (4286) to a survey population of more than 166 million people, and its resulting proportion in the table is of '24,13', whereas Canada has a larger

Fig. 6.8 Analysis of means setup

Fig. 6.9 Analysis of means output

Table 6.3 Generated files by an analysis of means

Generated files	File type	Content
mean.sps	SPSS	Syntax to run the means computations
mean.spv	SPSS	Output of the means computations
mean.sav	SPSS	Contains the means estimates and their standard errors
mean.xlsx	Excel	
mean_YRSQUAL_T_by_CNTRYID_Sig.sav	SPSS	Contains a country-by-country comparison for the estimated means, providing t-statistics
mean_YRSQUAL_T_by_CNTRYID_Sig.xlsx	Excel	

nominal sample of 26,472 cases, yet projected to a survey population of more than 23 million people, and hence its proportion in the table is of ‘3,37’. These percentages are accompanied by their standard errors included in the sixth column. In the seventh column, the estimates of interest are included: the mean of total years of schooling per country, under the heading ‘YRSQUAL_T (Mean)’. Next to it, in the eighth column, we find the standard errors of these estimates, below the heading ‘YRSQUAL_T (s.e.)’. The ‘Std.Dev’ column contains the standard deviations of the analysis variable, and the ‘Std.Dev (s.e.)’ contains the standard deviations standard errors. The last column of the table presents the percentage of missing values of the analysed variable.

When the analysis of means is complete, the IDB Analyzer generates six files. Table 6.3 details these files and their content.

Using the results provided in the file ‘mean.xlsx’, we created Table 6.4 to present the computed results.

Table 6.4 Means of lifetime years of schooling

Country	Mean	Standard Error	Country	Mean	Standard Error
Belgium	12.34	0.03	Lithuania	13.07	0.03
Canada	13.21	0.01	Netherlands	13.12	0.03
Chile	11.46	0.19	New Zealand	13.72	0.04
Cyprus	12.26	0.02	Norway	13.94	0.03
Czech Republic	12.95	0.02	Poland	12.48	0.04
Denmark	12.47	0.02	Russian Federation	13.35	0.03
Estonia	12.06	0.03	Singapore	11.68	0.01
Finland	12.24	0.03	Slovak Republic	12.94	0.04
France	11.18	0.02	Slovenia	10.31	0.00
Greece	11.77	0.01	Spain	11.30	0.02
Ireland	14.48	0.02	Sweden	12.01	0.02
Israel	12.61	0.02	Turkey	8.44	0.02
Italy	10.51	0.02	United Kingdom	13.00	0.03
Japan	12.94	0.01	United States	13.27	0.03
Korea, Republic of	12.60	0.02	Table Average	12.34	0.01

Considering that the population average might not be the most informative location parameter to describe the variable's distribution, in the next section we describe how to obtain percentiles of a continuous variable.

6.3.3 Percentiles

Means and percentiles are different location parameters in a distribution (Wilcox 2017). The arithmetic mean is the expected location of the value with the least difference to the rest of the values within a distribution. In contrast, percentiles are any location under which there is a certain proportion of cases. Means are informative for symmetric distributions, such as the normal distribution. However, when distributions depart from normality, medians (50th percentile) or other location parameters could be of interest. For the following example, we choose the 25th, 50th, and 75th percentile for the same variable. We will repeat Steps 1–3 from the previous routine, but we will change the statistic type.

1. Open the Analysis Module of the IDB Analyzer.
2. For this example, specify the data file 'merge_piaac.sav' as the Analysis File (see Sect. 6.2 in this chapter for the details of how this file was created).
3. Select 'PIAAC (using final full sample weight)' as the Analysis Type.
4. Select 'Percentiles' as the Statistic Type.
5. Under the 'Plausible Values Options', select 'None Used'.
6. Click on the 'Analysis Variables' section on the right-hand side of the software window. This section will become active and highlighted.

- 7. Go to the ‘Select variables’ section, and under the ‘Description’ heading, click on it, and type in ‘years’. This action would look for all the variables containing ‘years’ in their description field.
- 8. Specify the variable YRSQUAL_T as the analysis variable by clicking the ‘Analysis Variables’ field to activate it. Select YRSQUAL_T from the list of available variables present in the ‘Select Variables’ section, and move it to the ‘Analysis variables’ by clicking the right arrow button in this section. In this step, it is also possible to select more than one variable in this routine. However, for the sake of simplicity, in this example, we are including only one variable.
- 9. In the ‘Percentiles’ section, type in ‘25 50 75’, all separated by a space.
- 10. Specify the name and the folder of the output files in the ‘Output Files’ field by clicking the ‘Define/Modify’ button. For this example, we use the term ‘percentile’.

The generated setup should be similar to the screenshot presented in Fig. 6.10.

11. Afterwards, click the ‘Start SPSS’ button, run the syntax, and wait for the results to appear in the output window. The output from this routine is presented in Fig. 6.11

The generated output presents nine columns. The first is the list of countries; the second is the nominal sample size for each country; and in the third column, we find the sum of survey weights, which represents survey population size (Heeringa et al. 2009). In the ‘Percent’ column, the IDB Analyzer includes the proportion that the survey population size represents within the output table. Then, for each requested percentile (p25, p50, p75), we can find the point estimates, and its standard error on its right (p25_se, p50_se, p75_se).

For the computation of percentiles, the IDB Analyzer generates four files. Table 6.5 details these files and their content.

Fig. 6.10 Percentile setup

Fig. 6.11 Percentile output

Table 6.5 Generated files by an analysis of percentiles

Generated files	File type	Content
percentile.sps	SPSS	Syntax to run the means computations
percentile.spv	SPSS	Output of the means computations
percentile.sav	SPSS	Contains the means estimates and their standard errors
percentile.xlsx	Excel	

Using the results provided in the file ‘percentile.xlsx’, we created Table 6.6 to present the computed results. The estimated percentiles are included for each country, alongside their standard errors in parenthesis.

From the generated results, we notice that most of the participating countries have a median lifetime of schooling of 12 years. Ireland, the Netherlands, and Norway reach at least 14 years of schooling for half of their population of participants. At the lower end, Italy and Turkey presented a median schooling lifetime of 9 years.

6.3.4 Percentages

In the next example, we will create a new variable, not present in the merged files, to then retrieve percentage estimates at the population level for each country. We will use PIAAC data to estimate the proportion of the population in each participating country that has reached at least upper secondary education. To do this, we first need to recode a derived variable present in the public use file of the study. We will recode variable EDCAT8 into a dummy variable. EDCAT8 contains codes from the

Table 6.6 Percentiles (25th, 50th, and 75th) for total years of schooling by country

Country	P25	P50	P75
Belgium	12.00 (0.00)	12.00 (0.00)	15.00 (0.00)
Canada	12.00 (0.00)	13.00 (0.00)	16.00 (0.00)
Chile	8.00 (0.00)	12.00 (0.00)	14.00 (4.47)
Cyprus	12.00 (0.00)	12.00 (0.00)	14.00 (0.00)
Czech Republic	12.00 (0.00)	13.00 (0.00)	13.00 (0.00)
Denmark	10.00 (0.00)	12.00 (0.00)	15.00 (0.00)
Estonia	10.00 (0.00)	12.00 (0.00)	14.00 (0.00)
Finland	11.00 (0.00)	12.00 (0.00)	14.00 (0.00)
France	9.00 (0.00)	11.00 (0.00)	14.00 (0.00)
Greece	9.00 (0.00)	12.00 (0.00)	14.00 (0.00)
Ireland	11.00 (0.00)	14.00 (0.00)	18.00 (0.00)
Israel	12.00 (0.00)	12.00 (0.00)	15.00 (0.00)
Italy	8.00 (0.00)	8.00 (0.00)	13.00 (0.00)
Japan	12.00 (0.00)	12.00 (0.00)	14.00 (0.00)
Korea, Republic of	12.00 (0.00)	12.00 (0.00)	16.00 (0.00)
Lithuania	12.00 (0.00)	13.00 (0.00)	15.00 (0.00)
Netherlands	11.00 (0.00)	14.00 (0.00)	16.00 (0.00)
New Zealand	12.00 (0.00)	13.00 (0.00)	16.00 (0.00)
Norway	12.00 (0.00)	14.00 (0.00)	16.00 (0.00)
Poland	11.00 (0.00)	12.00 (0.00)	15.00 (0.00)
Russian Federation	11.00 (0.00)	12.00 (0.00)	18.00 (0.00)
Singapore	10.00 (0.00)	12.00 (0.00)	15.00 (0.00)
Slovak Republic	12.00 (0.00)	13.00 (0.00)	13.00 (0.00)
Slovenia	9.00 (0.00)	11.00 (0.00)	11.00 (0.00)
Spain	10.00 (0.00)	12.00 (0.00)	14.00 (0.00)
Sweden	11.00 (0.00)	12.00 (0.00)	14.00 (0.00)
Turkey	5.00 (0.00)	8.00 (0.00)	12.00 (0.00)
United Kingdom	11.00 (0.00)	11.00 (0.00)	16.00 (0.00)
United States	12.00 (0.00)	13.00 (0.00)	16.00 (0.00)
Table Average	10.66 (0.00)	12.03 (0.00)	14.69 (0.15)

International Standard Classification of Education (ISCED) to express the highest level of formal education of the participants (OECD 2015).

Using the following syntax code (see Table 6.4), we can create a dummy variable that differentiates between the participants who hold an upper secondary qualification (coded as one) and the participants who present lower educational qualifications, such as a primary school qualification or an incomplete secondary school qualification (coded as zero).

To include this new variable in the generated merged file, the user needs to open the merged file in SPSS. Then, open a new syntax window; type in the syntax code included in Code 6.1; press CTRL+A and CTRL+R to create this variable. Click

Fig. 6.12 Selecting a newly generated variable

on the window with the merged data, and press CTRL+S to save this variable in the merged file.

Code 6.1: Recoding Highest Educational Level to a Dummy Variable

```
if (EDCAT8 <= 2) edu_usl = 0 .
if (EDCAT8 >= 3) edu_usl = 1 .
execute .
```

```
VARIABLE LABELS edu_usl 'Population with upper secondary
education (1=yes, 0=no)'.
```

```
VALUE LABELS edu_usl
0 'No'
1 'Yes' .
```

With the merged file closed, one can open the IDB Analyzer and use this new variable for further analysis. In the next example, we will estimate what proportion of the population of the participant countries has at least upper secondary educational qualifications. Similarly to previous examples, we start by opening the IDB Analyzer:

1. Open the Analysis Module of the IDB Analyzer.
2. For this example, specify the data file 'merge_piaac.sav' as the Analysis File (see Sect. 6.2 in this chapter for the details on how this file was created).
3. Select 'PIAAC (using final full sample weight)' as the Analysis Type.
4. Select 'Percentages only' as the Statistic Type.
5. Click on the 'Grouping Variables' section.
6. Go to the 'Select variables' section, and under the 'Description' heading, click on it, and type in 'upper'. This action would look for all the variables containing 'upper' in their description field. This is presented in Fig. 6.12.
7. Drag the variable 'edu_usl' to the 'Grouping variable' section. By clicking the 'Analysis Variables' field to activate it, select 'edu_usl' from the list of available

Fig. 6.13 Percentage setup

Fig. 6.14 Percentage output

- variables present in the ‘Select Variables’ section, and move it to the ‘Grouping Variables’ field by clicking the right arrow button in this section.
- Specify the name and the folder of the output files in the ‘Output Files’ field by clicking the ‘Define/Modify’ button. In this example, we will use the term ‘percentage’. This setup is presented in Fig. 6.13.
 - Click the ‘Start SPSS’ button to create the SPSS syntax file and open it in an SPSS editor window.
 - After the user has executed the generated syntax, by pressing the sequence of keys CTRL+A and CTRL+R, the IDB Analyzer will start to run their macros to compute the requested percentages.

Once the calculations are finished, the SPSS output window would present the following results (see Fig. 6.14).

Table 6.7 Generated files by percentage analysis

Generated files	File type	Content
percentage.sps	SPSS	Syntax to run the percentage computations
percentage.spv	SPSS	Output of the percentage computations
percentage.sav	SPSS	Contains the percentage estimates and their standard errors
percentage.xlsx	Excel	
percentage__by_EDU_USL_Sig.sav	SPSS	Contains a country-by-country comparison for the estimated percentages
percentage__by_EDU_USL_Sig.xlsx	Excel	

Similarly to the procedure of means estimation, the procedure to estimate percentages produces six files as outputs. These files and their contents are described in Table 6.7.

Inspecting the generated output file in excel format, ‘percentage.xlsx’, we can filter and order the results to produce Table 6.8 and display the proportions of participants without upper secondary education for each participating country in PIAAC.

In the following section, we will use the dummy variable we have created, ‘edu_usl’, and estimate its relation to literacy scores in the population of each country.

6.3.5 Linear Regression

Apart from descriptive estimates such as means, percentiles, and percentages, the IDB Analyzer can also estimate regression models and logistic regression models (IEA 2019). In the following example, we will estimate the relationships between educational qualifications and literacy in each country. Specifically, we will estimate the gap in literacy scores between those who have at least an upper secondary education and the rest of the population. Although this gap can be obtained with a mean comparison, we want to retrieve more estimates than the mean differences between the two groups. We will use the linear regression routine for these purposes and get this difference as a standardised effect while also retrieving a measure of explained variance. These results can answer the question of ‘how much difference in literacy skills is there between those with and without upper secondary education?’. To estimate a regression analysis, we need to follow the next steps in the IDB Analyzer:

1. Open the Analysis Module of the IDB Analyzer.
2. For this example, specify the data file ‘merge_piaac.sav’ as the Analysis File (see Sect. 6.2 in this chapter for the details of how this file was created).
3. Select ‘PIAAC (using final full sample weight)’ as the Analysis Type.
4. Select ‘Regression’ as the Statistic Type.
5. Under the ‘Plausible Values Options’, select ‘Use PVs’.

Table 6.8 Proportion of participants without upper secondary education

Country	Estimate	Standard Error
Turkey	64.51	0.17
Italy	53.78	0.08
Spain	47.46	0.05
Greece	32.31	0.12
Chile	32.18	1.99
Netherlands	31.02	0.63
Ireland	28.48	0.08
France	27.91	0.40
Norway	27.43	0.53
Denmark	26.36	0.47
United Kingdom	24.09	0.59
Sweden	23.74	0.39
Slovenia	23.63	0.20
Austria	22.84	0.27
New Zealand	22.63	0.72
Korea, Republic of	21.66	0.49
Cyprus	21.64	0.33
Slovak Republic	20.61	0.62
Belgium	20.04	0.52
Finland	19.65	0.43
Singapore	18.92	0.17
Israel	17.81	0.43
Czech Republic	15.61	0.34
Poland	15.34	0.42
Japan	14.78	0.40
United States	14.74	0.28
Lithuania	11.93	0.47
Russian Federation	7.03	0.78
Table Average	25.29	0.11

6. On the right-hand side of the window, click on the area of ‘Dependent variables’. This will become highlighted once it is clicked.
7. Then, select ‘Plausible Values’ in the right-hand side window.
8. Move the cursor to the left-hand side of the window and click on the ‘PVLIT1–10’ variable to select the literacy scores.
9. Go back to the right-hand side and click on the right arrow to move the ‘PVLIT1–10’ variables, to the ‘Dependent Variables’ section.
10. Move the cursor to the ‘Independent Variables’ section, and click on the ‘Categorical Variables’ to activate this section.
11. Move the cursor to ‘Select Variables’ section on the left. Just right before the variable list, in the first row under the description section, type in ‘upper’. This will filter all present variables from the merged file.

Fig. 6.15 Regression setup

12. Select the variable ‘edu_usl’, and move it to the right-hand side, by clicking on the right arrow, under ‘Independent Variables’, specifically using the right arrow from the ‘Categorical Variables’ subsection.
13. Specify the name and the folder of the output files in the ‘Output Files’ field by clicking the ‘Define/Modify’ button. In this example, we will name the syntax file as ‘regression’.

Once all these steps have been completed, the regression setup should look like Fig. 6.15.

14. Click the ‘Start SPSS’ button. This action will open SPSS and create the syntax to run the regression model.
15. To execute the generated syntax, select all the written commands in the syntax editor, and run these commands using the ‘Run Selection’ button. Alternatively, press CTRL+A, to select all the commands, and then press CTRL+R to execute the syntax. This action would make SPSS run the regression analysis.

Because this analysis involves plausible values, it may take considerably longer in comparison with examples without plausible values in their calculations. This is because the regression analysis needs to be computed for each plausible value once, and then these results are synthetically presented using Rubin–Shaffer rules (Rutkowski et al. 2010). As such, this routine may take ten times longer than a regression analysis without the use of plausible values.

Once the regression analysis is done, SPSS will present the results in its output window. Figure 6.16 depicts how these results are displayed.

Once the analysis is concluded, the IDB Analyzer will generate eight files. These files include the syntax, the output, the model fit, the coefficients of the regression, and the descriptives of the included variables in the model. Table 6.9 lists the eight generated files and provides a description of their contents.

Fig. 6.16 Regression output

Table 6.9 Files generated by a regression analysis

Generated files	File type	Content
regression.sps	SPSS	Syntax to run the regression analysis
regression.spv	SPSS	Output of the regression analysis
regression_Model.sav	SPSS	Contains R-squared of the regression model and its adjusted R-squared
regression_Coeff.sav	SPSS	Contains unstandardised and standardised regression coefficients and their standard errors
regression_Coeff.xlsx	Excel	
regression_Desc.sav	SPSS	Contains descriptive statistics for all the variables included in the regression model. These include means, standard deviations, and variances
regression_Desc.xlsx	Excel	

Using the estimates present in ‘regression_Coeff.xlsx’ and in ‘regression_Model.xlsx’, we created Table 6.10, to show at a glance the general results of the fitted model. These results are ranked in descending order using the R^2 , a measure of explained variance (see, e.g. Field 2013, for more information about regression analysis).

Inspecting the regression coefficients present in ‘regression_Coeff.xlsx’ and their t-values, we can conclude that all estimated differences are above the sampling error and all beta.t are larger than two. Thus, in all countries, those who have at least upper secondary education obtain higher literacy scores in the PIAAC test. The average difference of all participating countries is 0.32 standard deviations of literacy scores. The estimated gap varies between countries. For example, in Singapore, Spain, and Chile, it is larger than 0.45 standard deviations. In contrast, in Lithuania, the Russian Federation, and Poland, this difference is less than or equal to 0.16 standard deviations of literacy scores.

Table 6.10 Standardised regression coefficients and model fit

Countries	Standardised Estimate	Standard Error	R ²
Singapore	0.55	0.01	0.30
Spain	0.46	0.01	0.21
Chile	0.45	0.02	0.20
Netherlands	0.42	0.02	0.18
Ireland	0.39	0.02	0.15
United Kingdom	0.39	0.01	0.15
Turkey	0.39	0.02	0.15
France	0.38	0.01	0.15
Italy	0.37	0.02	0.14
New Zealand	0.37	0.02	0.13
Korea, Republic of	0.36	0.02	0.13
Belgium	0.35	0.02	0.12
Sweden	0.35	0.01	0.12
United States	0.33	0.01	0.11
Slovak Republic	0.33	0.02	0.11
Denmark	0.31	0.01	0.10
Austria	0.30	0.02	0.09
Norway	0.30	0.01	0.09
Slovenia	0.29	0.02	0.09
Japan	0.28	0.02	0.08
Finland	0.27	0.02	0.07
Israel	0.26	0.02	0.07
Greece	0.25	0.03	0.06
Cyprus	0.22	0.02	0.05
Czech Republic	0.19	0.02	0.04
Poland	0.16	0.02	0.03
Russian Federation	0.15	0.03	0.02
Lithuania	0.06	0.02	0.00
Table Average	0.32	0.00	0.11

6.3.6 Correlations

In the PIAACe study, literacy, numeracy, and problem solving in technology-rich environments were measured. How are these different skills related to each other? In other words, to what extent do these three variables fluctuate together? In the OECD (2016a) report, ‘Skills Matter’, these variables were reported to be highly and positively correlated, with correlations of 0.86 for literacy and numeracy for the OECD partners (see, e.g. Field 2013, for more information about correlation analysis). In the following example, we will estimate the correlation between proficiency in literacy, numeracy, and problem solving in technology-rich environments. To compute these correlations, we need to follow the next steps:

1. Open the Analysis Module of the IDB Analyzer.

2. For this example, specify the data file ‘merge_piaac.sav’ as the Analysis File (see Sect. 6.2 in this chapter for the details of how this file was created).
3. Select ‘PIAAC (using final full sample weight)’ as the Analysis Type.
4. Select ‘Correlations’ as the Statistic Type.
5. Under the ‘Plausible Values Options’, select ‘Use PVs’.
6. Under the ‘Missing Data’ option, select ‘Pairwise’.
7. Click on the ‘Plausible Values’ section on the right-hand side of the software window.
8. Go to the ‘Select variables’ section, and select the three plausible values variables.
9. Move all the selected variables, by clicking the right arrow in the right-hand side window under the ‘Plausible Values’ subsection.
10. Specify the name and the folder of the output files in the ‘Output Files’ field by clicking the ‘Define/Modify’ button. In this example, we define the syntax as ‘correlation’.

The final setup should resemble the setup presented in Fig. 6.17.

11. Then, click the ‘Start SPSS’ button. This will create an SPSS syntax file and open it in an SPSS editor window.
12. To start the computations, one needs to press the following key combinations: CTRL+A first, to select the entire generated code present in the syntax window, and then CTRL+R to run these commands. The output of these analyses is depicted in Fig. 6.18.

Because these computations involve the plausible values of the three proficiency scores, its estimation will take longer compared with correlations between variables with no plausible values. When the computations are done, six files are generated. These files are described in Table 6.11.

The output of these computations is displayed in Fig. 6.18.

These results match those shown in Table A2.7 of the report ‘Skills Matter: Further Results from the Survey of Adult Skills’ (OECD 2016a). In Table 6.12, we include only the matching countries from the OECD report, and the countries

Fig. 6.17 Correlation setup

Fig. 6.18 Correlation output

Table 6.11 Files generated by a correlation analysis

Generated files	File type	Content
correlation.spv	SPSS	Syntax to run the correlation analysis
correlation.sps	SPSS	Output of the correlation analysis
correlation_Corr.sav	SPSS	Contains the correlation estimates and their standard errors
correlation_Corr.xlsx	Excel	
correlation_Desc.sav	SPSS	Contains descriptives for all the variables included in the correlation analysis. These include means, standard deviations, and variances
correlation_Desc.xlsx	Excel	

present in the current merged file. Thus, the correlations from Australia, Northern Ireland, and Jakarta (Indonesia) are excluded in the present table.

6.3.7 Proficiency Levels

The PIAAC study presents proficiency levels—that is, segments of scores used to describe the skills of literacy, numeracy, and problem solving in technology-rich environments at different levels of ability. These are ranges of scores to describe in qualitative terms what participants can do at different levels of proficiency. In general terms, those participants with higher scores in each domain are more likely to resolve more difficult tasks than their counterparts with lower scores (OECD 2016a).

Table 6.12 Correlations between literacy and numeracy scale scores

	OECD estimates	IDB estimates
Austria	0.86	0.86
Belgium	0.87	0.87
Canada	0.87	0.87
Chile	0.84	0.84
Cyprus	0.81	0.81
Czech Republic	0.80	0.80
Denmark	0.88	0.88
Estonia	0.83	0.83
Finland	0.86	0.86
France	0.86	0.86
Germany	0.87	0.87
Greece	0.81	0.81
Ireland	0.87	0.87
Israel	0.86	0.86
Italy	0.83	0.83
Japan	0.86	0.86
Korea, Republic of	0.88	0.88
Lithuania	0.84	0.84
Netherlands	0.89	0.89
New Zealand	0.87	0.87
Norway	0.89	0.89
Poland	0.85	0.85
Russian Federation	0.79	0.79
Singapore	0.93	0.93
Slovak Republic	0.85	0.85
Slovenia	0.88	0.88
Spain	0.89	0.89
Sweden	0.89	0.89
Turkey	0.85	0.85
United Kingdom	0.87	0.87
United States	0.89	0.89

Literacy scale scores have six proficiency levels. These proficiency levels are briefly described in Table 6.13; more details can be found in ‘The Survey of Adults Skills: Reader’s Companion’ (OECD 2016b).

1. Open the Analysis Module of the IDB Analyzer.
2. For this example, specify the data file ‘merge_piaac.sav’ as the Analysis File (see Sect. 6.2 in this chapter for the details of how this file was created).
3. Select ‘PIAAC (using final full sample weight)’ as the Analysis Type.
4. Select ‘Benchmarks’ as the Statistic Type.
5. Under the ‘Benchmarks Options’ select ‘Discrete’. This option will retrieve what proportion of the population falls within each proficiency level. Other options include ‘Cumulative’, which computes the proportion of people at

Table 6.13 Levels of proficiency

Level	Cut scores	Brief descriptions of more likely resolved tasks
Below level 1	Below 176	Basic reading comprehension with a basic vocabulary
Level 1	From 176 to below 226	Reading tasks resolved at this level include the integration of information, using identical or synonymous terms
Level 2	From 226 to below 276	Reading tasks from this level require the integration of information of similar meaning, via low inference or paraphrase, and discerning competing information
Level 3	From 276 to below 326	Reading tasks require the participant to read through larger pieces of text and construct meaning across paragraphs
Level 4	From 326 to below 376	Reading tasks from this level require complex inferences and application of background knowledge. The participants need to evaluate subtle evidence-claims or persuasive discourse relationships
Level 5	At or above 376	Tasks may require the respondent to search for information and integrate information of similar and contrasting ideas, points of view, or evaluate evidence-based arguments. Evaluating the reliability of evidentiary sources and selecting key information is frequently a requirement

or above the cut score, and ‘Discrete with analysis variables’, which permits the user to calculate the mean of an analysis variable for those within each proficiency level. For this example, we will use the ‘Discrete’ option.

6. Click on the ‘Plausible Values’ section on the right-hand side of the software window.
7. Move the cursor to the left-hand side of the window and click on the ‘PVLIT1–10’ variable to select the literacy scores.
8. Move the selected variable, by clicking on the right arrow in the right-hand side window, under the ‘Plausible Values’ subsection.
9. Under the ‘Achievement Benchmarks’ section, select the corresponding scores for the literacy scores; these are ‘176 226 276 326 376’.
10. Specify the name and the folder of the output files in the ‘Output Files’ field by clicking the ‘Define/Modify’ button. Here we define the syntax as ‘benchmark’.

The setup of this analysis is depicted in Fig. 6.19.

11. Click the ‘Start SPSS’ button. This action will open SPSS and create the syntax to run the regression model.
12. To execute the generated syntax, press CTRL+A to select all the commands, and then press CTRL+R to execute the syntax. Now, SPSS will compute the proportion of cases at each benchmark.

Results are displayed in Fig. 6.20, as they appear in SPSS.

Fig. 6.19 Benchmark setup

Fig. 6.20 Benchmark output

What do these results mean? We need to consider the procedure the benchmark routine is doing to explain this output. Each cut score is the lower bound value for each defined range (IEA 2019). We used the following cut scores: ‘176 226 276 326 376’. Thus, it computes all the cases below 176 points, all the cases between 176 and 226, between 226 and 276, between 276 and 326, between 326 and 376, and finally all the cases above 376. In the last two columns of the output, the estimates are the percentage of cases and their standard errors that fall into the specified ranges. This procedure generates the following files in the specified location (see Table 6.14).

Using the information contained in ‘benchmark.xlsx’, we created Table 6.15. This table displays the proportions of participants who performed below Level 1 in the literacy proficiency scale.

Table 6.14 Files generated by a correlation analysis

Generated files	File type	Content
benchmark.spv	SPSS	Syntax to run the benchmark computations
benchmark.sps	SPSS	Output of the benchmark computations
benchmark.sav	SPSS	Contains the benchmark estimates and their standard errors
benchmark.xlsx	Excel	
benchmark_by_CNTRYID_Sig.sav	SPSS	Contains a country-by-country comparison for the percentages at each benchmark
benchmark_by_CNTRYID_Sig.xlsx	Excel	

Table 6.15 Percentage of participants below Proficiency Level 1

Country	Estimate	Standard error	Country	Estimate	Standard error
Chile	20.37	1.39	United Kingdom	3.32	0.38
Turkey	12.92	0.85	Norway	3.05	0.3
Singapore	10.21	0.39	Belgium	2.9	0.28
Israel	8.25	0.41	Finland	2.66	0.23
Spain	7.27	0.47	Netherlands	2.62	0.27
Slovenia	6.01	0.41	New Zealand	2.57	0.28
Italy	5.56	0.57	Austria	2.5	0.32
France	5.37	0.31	Lithuania	2.26	0.38
Greece	4.95	0.53	Korea, Republic of	2.23	0.2
Ireland	4.3	0.43	Estonia	2.02	0.19
United States	4.09	0.47	Slovak Republic	1.89	0.24
Poland	3.94	0.32	Cyprus	1.89	0.28
Canada	3.87	0.24	Russian Federation	1.56	0.54
Denmark	3.83	0.29	Czech Republic	1.54	0.32
Sweden	3.68	0.33	Japan	0.57	0.15
Germany	3.33	0.39	Table Average	4.57	0.08

In Chile, 20% are below Level 1, whereas in Japan, Czech Republic, the Russian Federation, Cyprus, and the Slovak Republic, there are fewer than 2% below Level 1.

6.4 Concluding Remarks

In this chapter, we demonstrated how to perform both simple and complex analysis with PIAAC data using the IEA International Database (IDB) Analyzer. We showed examples of how to combine datasets from more than one country into a single data file for cross-country analyses. We also described and illustrated in a step-by-step

fashion how to run descriptive statistical analyses, including means, percentiles, and percentages, as well as inferential analyses, such as correlations and regressions.

Although the examples included in this chapter used data from OECD's PIAAC study, it is important to mention that the IDB Analyzer can be used to analyse not only PIAAC data but many other international large-scale assessments, such as the OECD's PISA and TALIS studies, as well as, for example, IEA's Trends in International Mathematics and Science Study (TIMSS), Progress in International Reading Literacy Study (PIRLS), and International Civic and Citizenship Study (ICCS).

The IDB Analyzer is certainly not the only tool available to obtain correct estimates when analysing PIAAC data, but it is probably the most user-friendly one. As mentioned before, the IDB Analyzer is a Windows-based tool that creates SAS code or SPSS syntax to perform analyses with PIAAC data. The code or syntax generated by the IDB Analyzer automatically takes into account the complex sample (e.g. sampling weights, replicate weights) and complex assessment design (e.g. plausible values) of PIAAC to compute analyses with the correct standard errors. It enables researchers to test statistical hypotheses in the population without having to write any programming code.

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