



The EBA Stress Test data set

Guide for data exploitation

As a result of the 2018 EU-wide Stress Test Exercise, the EBA has published bank-by-bank data contained in 10 Transparency templates for a sample of 48 banks.



2018 EU-wide Stress Test

Bank Name	
LEI Code	
Country Code	

Cover	TRA_SUM	TRA_CR_IRB	TRA_CR_STA	TRA_CR_SEC	TRA_REA	TRA_CAP	TRA_P&L	TRA_CAPMEAS	TRA_NPE	TRA_FORB
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The EBA has developed a range of practical tools that aim to facilitate the use of the stress test data. These include interactive maps and excel aggregation tools, as well as the complete stress test dataset in CSV format, which can be imported in any analytical software for analysis purposes.

The stress test dataset is stored in two different CSV files and includes all the bank-by-bank data contained in transparency templates. Each CSV file contains a specific stress test data category that reflects the content of one or more transparency templates as shown in the table below:

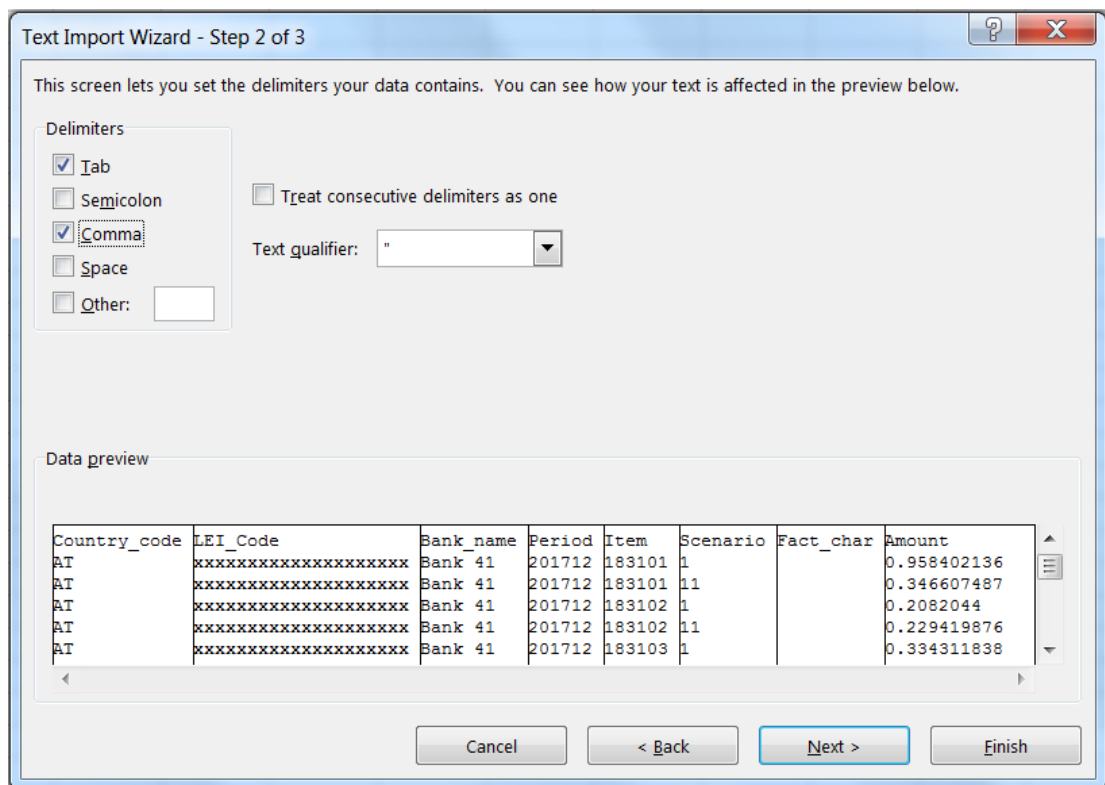
CSV Name	Stress Test category	Transparency Template
TRA_CR.csv	Credit risk	TRA_CR_STA TRA_CR_IRB TRA_CR_SEC TRA_NPE TRA_FORB
TRA_OTH.csv	Summary results, Capital, Risk exposure amount, P&L	TRA_SUM TRA_CAP TRA_CAPMEAS TRA_P&L TRA_REA

Along with the CSV, users will find the data dictionary table and the metadata table that are needed for understanding the database structure of each file (the two databases have a different structure) as well as for setting up the queries to extract the data.

An example will be useful to understand how to use and query the EBA Stress test database (bear in mind that **the figures below show fake data**). In the example below, the files have been converted into excel files in order to use standard analytical tools embedded in excel.

Capital: CET1 Ratio – fully loaded - for each bank by scenario using a pivot table

- i) Once the CSV file containing data on *Capital* is downloaded (TRA_OTH.csv), we import it in excel using the text import wizard:
- ii)





iii) The database structure turns to be the following:

	A	B	C	D	E	F	G	H	I
1	Country_code	LEI_Code	Bank_name	Period	Item	Scenario	Fact_char	Amount	
2	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183101	1		0.958402136	
3	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183101	11		0.346607487	
4	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183102	1		0.2082044	
5	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183102	11		0.229419876	
6	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183103	1		0.334311838	
7	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183103	11		0.42031039	
8	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183104	1		0.396678857	
9	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183104	11		0.588999415	
10	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183105	1		0.759536289	
11	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183105	11		0.814771079	
12	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183106	1		0.933118076	
13	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183106	11		0.085313677	
14	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183107	1		0.044178602	
15	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183107	11		0.735674431	
16	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183108	1		0.301016246	
17	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183108	11		0.72074073	
18	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183109	1		0.57812756	
19	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183109	11		0.190605442	
20	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183110	1		0.191584982	
21	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183110	11		0.348755495	
22	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183111	1		0.546963318	
23	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183111	11		0.998915705	
24	AT	xxxxxxxxxxxxxxxxxxxx	Bank 41	201712	183112	1		0.480153236	

iv) The database structure is explained in a metadata file in which you one can find a description of all the values that each column can assume. For *Capital*, the database has 8 columns:

- *Country_code*: code of the country of the Bank
- *LEI_code*: a bank identifier
- *Bank_Name*: name of the bank
- *Period*: time period
- *Item*: code of each variable
- *Scenario*: code of the scenario
- *Fact_char*: value that the string variable assumes
- *Amount*: value that the variable assumes

Users can find decoding information either in the metadata file (.xlsx) and/or in the data dictionary file (Data dictionary.xlsx).



For instance, in the sheet “Scenario” of the Metadata file, one can see that the dimension Scenario can only assume values equal to 1, 11, 2 or 3 and find the corresponding explanation in it.

Scenario	Label
1	Actual figures
11	Restated figures
2	Baseline scenario
3	Adverse scenario

- v) For identifying the item code associated with the financial concept “CET1 Ratio – fully loaded”, users can look for the name of the item in the column *Label* of the Data dictionary file and they will find that the item code is 183765.

The screenshot shows an Excel spreadsheet with the following data in the first row:

Collection	Template	Item	Category	Label
ST2018	TRA_CAP	183765	Transparency - CAP	Common Equity Tier 1 Capital ratio (fully loaded)

A context menu is open over the 'Label' column, showing a search filter for 'Common Equity Tier 1 Capital ratio (fully loaded)'. The search results list several items, with the selected item checked.

- vi) Now we click on “Pivot table” and select the entire dataset (or a subsample if you already filtered the data you need) as the pivot table range. We set up the pivot table structure, dragging in the box *Row Label* the variable *Bank_name* while in the columns we want the *Period* and the *Scenario*. We drag in the box *Values* the variable *Amount* where the variables’ values are stored and we aggregate them by sum. Finally, via the *Design* tab, we switch off the Subtotals and Grand Totals for both columns and rows.



vii) Final result turns to be the following:

Item	201712	201712	201812	201912	202012	201812	201912	202012
Bank 41	58.08%	14.42%	41.37%	11.42%	37.18%	35.93%	24.43%	27.08%
Bank 19	49.56%	88.76%	50.17%	50.93%	0.17%	30.27%	16.43%	36.22%
Bank 27	98.54%	49.53%	90.35%	52.79%	42.23%	97.23%	77.10%	22.71%
Bank 10	62.94%	46.40%	95.50%	60.97%	11.07%	1.14%	26.79%	81.23%
Bank 18	81.04%	80.30%	63.43%	82.87%	15.94%	35.21%	33.52%	42.69%
Bank 29	70.07%	29.78%	12.26%	62.94%	17.65%	61.25%	34.69%	1.14%
Bank 16	33.63%	70.73%	11.43%	29.00%	15.04%	28.83%	29.10%	95.34%
Bank 14	97.40%	66.36%	66.94%	69.36%	1.51%	96.73%	79.96%	94.48%
Bank 33	19.41%	50.32%	89.57%	24.71%	18.79%	30.18%	60.72%	87.75%
Bank 9	35.99%	56.12%	7.66%	12.22%	6.48%	79.62%	63.77%	31.55%
Bank 30	15.59%	7.59%	98.47%	25.78%	51.66%	40.30%	85.85%	72.35%
Bank 35	12.20%	81.71%	1.20%	35.92%	37.47%	1.94%	54.04%	48.81%
Bank 15	32.75%	14.58%	23.80%	14.49%	8.27%	69.15%	68.28%	89.91%
Bank 26	98.79%	12.17%	87.75%	90.48%	8.05%	14.53%	17.69%	10.13%
Bank 36	26.35%	45.69%	77.45%	90.93%	57.62%	17.60%	98.49%	34.68%
Bank 6	58.26%	18.47%	56.41%	52.44%	37.03%	36.04%	34.60%	46.88%
Bank 3	56.96%	20.56%	73.41%	10.01%	50.05%	1.02%	47.65%	8.16%
Bank 12	60.28%	35.17%	23.06%	56.26%	36.83%	4.96%	17.21%	64.52%
Bank 5	91.35%	31.48%	47.19%	18.86%	90.28%	37.13%	33.04%	13.94%
Bank 37	6.30%	4.04%	77.02%	93.59%	63.34%	1.03%	72.48%	41.33%
Bank 11	53.05%	82.41%	48.84%	74.69%	93.90%	88.38%	49.46%	74.64%
Bank 22	7.73%	90.52%	20.07%	87.19%	79.36%	27.14%	72.26%	32.19%
Bank 43	12.13%	40.18%	43.96%	54.93%	43.86%	12.88%	67.18%	96.82%
Bank 20	41.50%	39.38%	2.82%	46.40%	27.74%	11.10%	59.90%	53.16%
Bank 21	16.87%	56.39%	10.02%	64.02%	4.07%	66.30%	35.18%	43.11%
Bank 28	85.49%	57.38%	34.28%	26.40%	44.64%	65.31%	54.45%	22.10%
Bank 8	76.49%	94.88%	74.29%	37.45%	43.80%	82.73%	88.87%	9.12%
Bank 31	98.99%	63.82%	52.94%	74.46%	35.21%	64.40%	20.74%	5.08%