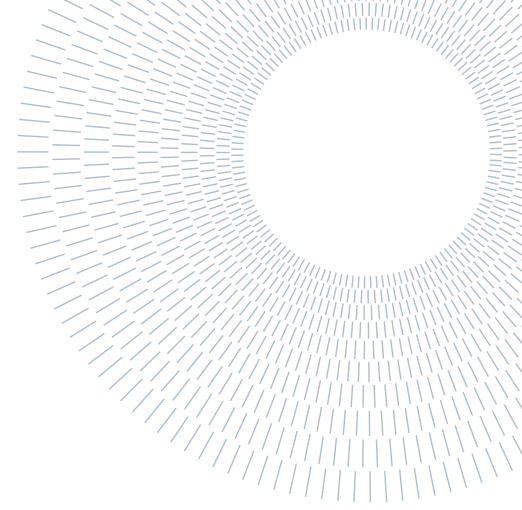




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CPA/Template attacks output viewer

PROJECT FOR THE EMBEDDED SYSTEMS COURSE
COMPUTER SCIENCE AND ENGINEERING

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Abstract: The Abstract is a concise summary of the content of the thesis (single page of text) and a guide to the most important contributions included in your thesis. The Abstract is the very last thing you write. It should be a self-contained text and should be clear to someone who hasn't (yet) read the whole manuscript. The Abstract should contain the answers to the main research questions that have been addressed in your thesis. It needs to summarize the motivations and the adopted approach as well as the findings of your work and their relevance and impact.

1. Introduction

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2. Implementation

2.1. Libraries used

The implementation is mainly build around existing libraries. Data handling and mathematical operations (such as peak detection) are carried out by NumPy, a fundamental package for scientific computing in Python. The GUI is build from PyQt5 which is the Python equivalent of the C++ Qt framework. It uses a slot mechanism to communicate between items and make the software easily re-usable. Finally the plotting is using PyQtGraph which a data visualisation package build from PyQt.

2.2. Required input files

The input files are stored in the `data/input/csv` folder. Each file is representing one the 16 bytes. The format of the files is a two dimensional array where the rows are the time instants (up to a few thousands rows) and the columns are the correlation values (256 columns). The values are in the range $[-1, 1]$ and separated by a coma and are representing correlations values. These are useful to determine the success of an attack.

2.3. Pre-processing steps

Before running the main script for the first time, it is necessary to execute two pre-processing steps in this specific order:

1. `csv_to_npy_conversion.py`, this script is converting the `csv` files to `npv`. This is carried out by a built-in function of the NumPy package. The `npv` format is the standard binary file format in NumPy to save a single arbitrary NumPy array on disk. The format stores all of the shape and dtype information necessary to reconstruct the array correctly even on another machine with a different architecture.
2. `peak_detection.py`, this script is executed once the `npv` files are generated. The aim of this step is to detect the peak value of each byte. The sample number of the peak value, the peak value and the byte value (i.e. column number in the corresponding `csv` file) are stored and written to a dedicated file (`data/output/csv/peaks.csv`) which is then used in the main program.

2.4. Implemented features

2.4.1 Mouse hover

The movement of the mouse cursor is generating signals (emitted by built-in slots of `PyQt` and `PyQtGraph`). This signal is connected to callback function written by us. The position of the mouse is caught by the callback and mapped from pixels coordinates to the plot coordinates. These values are then printed on the graph window to help the user in the data visualisation and a crosshair symbol is plotted to show the cursor position on the graph widget.

2.4.2 Signal generation

If the user wants to study a specific value, he has the possibility to left-click on the data point. This will automatically emit a custom signal named `message` with 3 arguments: `x`, `y`, `line`. The argument `x` is an `int` corresponding to the sample number, `y` is a `float` corresponding to the correlation values and `line` is a `str` with the number of the line (i.e. column number in the corresponding `csv` file). The user can use this emitted signal and implement his own callback function with the features he wants. Currently, as a proof of concept, this signal is caught and the informations are simply printed to the console.

2.4.3 Zoom features

The user can zoom using different commands. Scrolling up and down is zooming and de-zooming the whole plot centered on the cursor position. Holding the right click pressed and dragging horizontally is (de-)zooming the x axis. A similar behaviour on the y axis is achieved by dragging the right click vertically. The final possibility to zoom is to right click once anywhere on the plot and select `Mouse Mode > 1 button` from the contextual menu. Then dragging with the left click is selecting a rectangular region and zooming to fit that region of interest to the window.

2.5. Separate the logic from the GUI

An effort was made to use as much as possible `ui` files generated by `QtCreator` which is an IDE to create applications.

2.6. Nice to have improvements

Some safety measures are implemented in order to guarantee the right course of actions of the application. Our implementation makes an extensive use of file (both for reading and writing data). If by any chances one of the file is not in the correct folder or not present at all, an error message is printed to the console and the script exits. In addition, some correlation values in the files might be corrupted. The pre-processing scripts are checking for non conform values and replace them by a correlation of zero. This is the reason why executing these pre-processing scripts are primordial and an error will be thrown if they are not run or run in the wrong order.

3. Faced problems and solutions

During the development of the application, we had to face up some problems, mainly due to the size of the dataset to handle. We think it can be useful to share here some insights about our decision-making process as the developement was moving forward, and the solutions we came up with.

Starting from the very beginning, our goal is plotting the content of a .csv file, while maintaining the possibility of moving around inside the plot (i.e., zoom-in, zoom-out, change the X-range).

Our initial approach was plotting all the file at once.

The problem that immediately arose was that handling over 7.5 million points (every .csv file has about 30 thousand rows and 256 columns) was definitely too resource-demanding. That led to an evidently unusable GUI due to its poor performance, not only when moving around, but also for the loading time of the plot itself.

Therefore, we started experimenting looking for some possible improvements to load data into memory in an efficient way using Python.

A first available option we found was converting the .csv file into a .hdf5 one. HDF5 is a file format built for fast I/O processing and storage, whose potential can be fully exploited through a handy Python library. However, the result we obtained was still very poor and the HDF5 option had to be discarded.

Since we handled all the data as numpy arrays, we also came across the .npy file format. It is a simple format for saving numpy arrays, in an optimized way, with no loss of information. After converting the .csv file into a .npy one, we plotted it in the same way as before. The loading time was significantly lower and the usability of the GUI improved, but still far from an acceptable result.

Having understood that .npy was a far better choice over .csv and .hdf5, we decided to keep that conversion step and start tweaking around to gain as much performance as possible, this time in the PyQtGraph plotting process.

A special tweak worth mentioning is the way we subsample data in order to be displayed in the plot. PyQtGraph provides three downsampling modes: subsample, mean, and peak.

Despite being the slowest one, we picked the last one ("peak") since it downsamples by drawing a saw wave that follows the min and max of the original data. This is the only way to make sure that the peak of our plot does not get lost while displaying the points.

Back to the fact that the GUI responsiveness was far from being acceptable, we came to the conclusion that showing all the 7.5 M points at the same time was not feasible in any way.

At this point we tried to change our approach. Here was (and currently is) our new one.

Since every file contains 256 rows, and each row corresponds to a curve in the plot, we can preprocess every file and identify for it which is the peak and what row it shows up in. Then, save the results in a new file. When the user opens a plot, now the program reads the content of such a file to display by default only the curve with the peak. In addition, each one of the other curves can be shown/hidden in the plot by using a dedicated checkbox.

This new approach leads to a very low loading time and to a GUI which is not only usable, but which has high performance in every zoom/scroll operation.

Despite the high performance increase intrinsic to the new way of displaying data, we still decided to keep the conversion step from .csv to .npy. In fact, the same file stored as .npy takes more than 40% less space on disk than if stored as .csv. As we will cover in more detail in the next section, this leads to both faster loading times and lower memory usage at runtime.

4. Experimental results

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5. Conclusions

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6. Guide (to be removed)

Figures, Tables and Algorithms have to contain a Caption that describes their content, and have to be properly referred in the text.

6.1. Figures

Include figures (.png, .jpg, .eps) with the following command:

```
\includegraphics[options]{filename.xxx}
```

Thanks to the `\subfloat` command, a single figure, such as Figure ??, can contain multiple sub-figures with their own caption and label, e.g. Figure ?? and Figure ??.

6.2. Tables

How to insert itemized lists:

- first item;
- second item.

How to write numbered lists:

1. first item;
2. second item.



Figure 1: Caption of the Figure.



Figure 2: Caption of the Figure.

Example of Table (optional)

	column1	column2	column3
row1	1	2	3
row2	α	β	γ
row3	alpha	beta	gamma

Table 1: Caption of the Table.

	column1	column2	column3	column4	column5	column6
row1	1	2	3	4	5	6
row2	a	b	c	d	e	f
row3	α	β	γ	δ	ϕ	ω
row4	alpha	beta	gamma	delta	phi	omega

Table 2: Highlighting the columns