

Report on the Automation of Data Loading from the RT13m in a Python App

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1 Introduction

This report describes the work carried out to automate the process of loading data from the RT13m into a Python application (`FS_Message_Form`) through the use of graphical interface buttons. The main objective of this task was to improve the usability of the application, reduce manual intervention, decrease the time required to complete the process, and make the workflow more efficient and intuitive for the user.

Before this improvement, the process of loading and managing data involved several manual steps, which increased the risk of user errors and made the application less convenient to use. To address this issue, a button-based interaction model was implemented, allowing users to execute data-loading actions directly from the interface in a simpler and more controlled way.

2 Initial Workflow

Initially, the application relied on a manual method for entering or loading data. The user had to search through a very large log file to find all the required information, copy it into the application, and, in some cases, calculate up to 16 averages from large sets of data.

This made the application cumbersome to use, as it required significant effort and attention from the user to complete all the fields correctly. In addition, the probability of making errors during the process was high.

3 Implemented solution

To solve this problem, the first step was to extract all the necessary information from the RT13m logs. For this purpose, a Python script was developed to parse the entire log file using regular expressions and store the extracted data in JSON format. Some of the extracted data include:

- Session name
- Dot, Dot2Gps, and Dot2Pps values
- Session start and end times
- Temperature, pressure, and humidity
- Coordinates
- Sky conditions (see Section 3.1)

Among many other fields, the complete data structure can be found in Appendix 1.

Then, several functions were implemented within the application itself to process the extracted data and place each value into its corresponding field. These functions were responsible for reading the generated JSON file, selecting the relevant information for each tab, and automatically populating the interface fields with the appropriate values.

3.1 Sky Conditions

To obtain information about sky conditions, a Python module called `skyConditions` was developed. Using the `meteostat` library, it retrieves the measured value of `cldc` (cloud cover) for given coordinates and date. Then, by applying the function shown in Appendix 2, an approximation of the sky conditions during the session is obtained.

Once all the information has been extracted, a JSON file is generated. This format is much easier to handle from the software point of view and allows the different tab fields to be filled in automatically.

4 Final Workflow

The application workflow is shown in Figure 4.1. First, the parser is executed by pressing the Load button in the Ready tab. This generates the corresponding JSON file and automatically fills in the Ready page. Once this step is completed, the user can press the Load buttons in the Start and Stop tabs, and the fields are then populated directly from the generated JSON file.

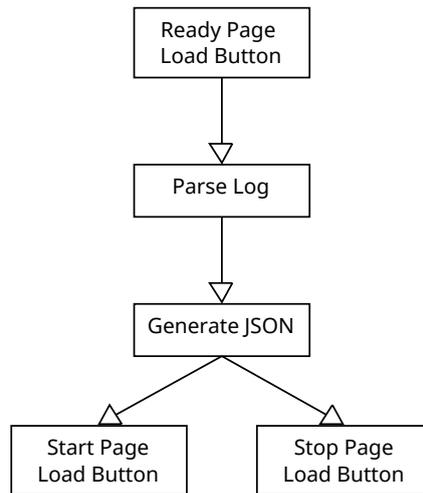


Figure 4.1: Application workflow

5 Ready Page

To automate the loading of information, a Load button was added to the interface (see Figure 5.1). This button allows the user to select a log file from the local file system, parse it automatically, and generate a JSON file containing all the relevant information.

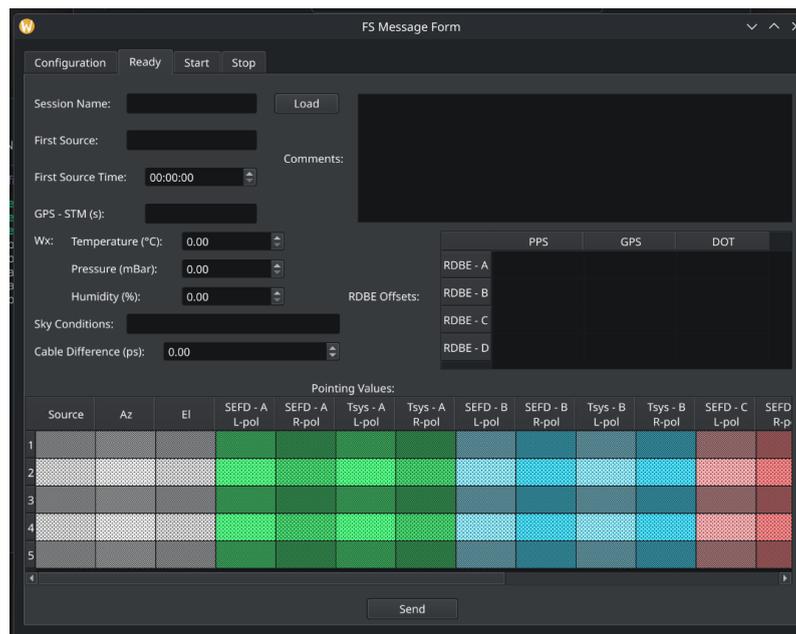


Figure 5.1: Ready page with the Load button before selecting a log file

Once the file has been generated, the page is filled automatically, as shown in Figure 5.2. This saves operators a considerable amount of time compared to manual data entry and significantly simplifies the overall process.

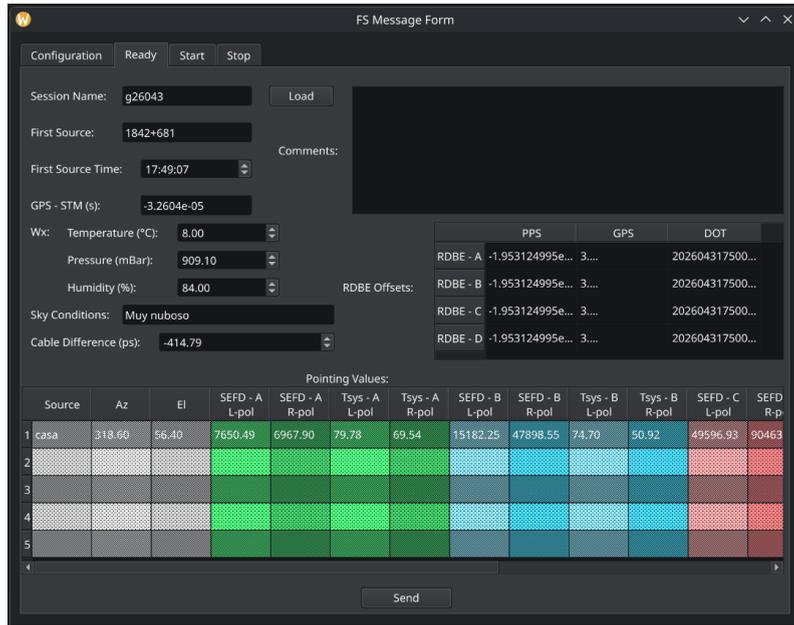


Figure 5.2: Ready page after parsing the log file

6 Start Page

Once the JSON file has been generated, the Load button in this tab (see Figure 6.1) simply accesses it and extracts the relevant information required to fill in the page automatically.

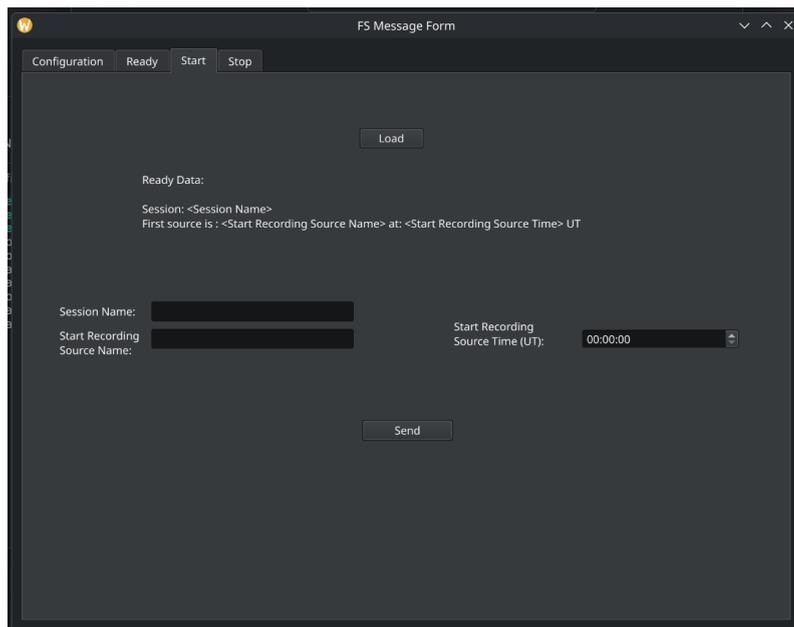


Figure 6.1: Start page before loading the data

In this page, the session name, the time of the first recording, and the name of the first source observed by the radio telescope at the moment the recording starts are filled in automatically, as shown in Figure 6.2.

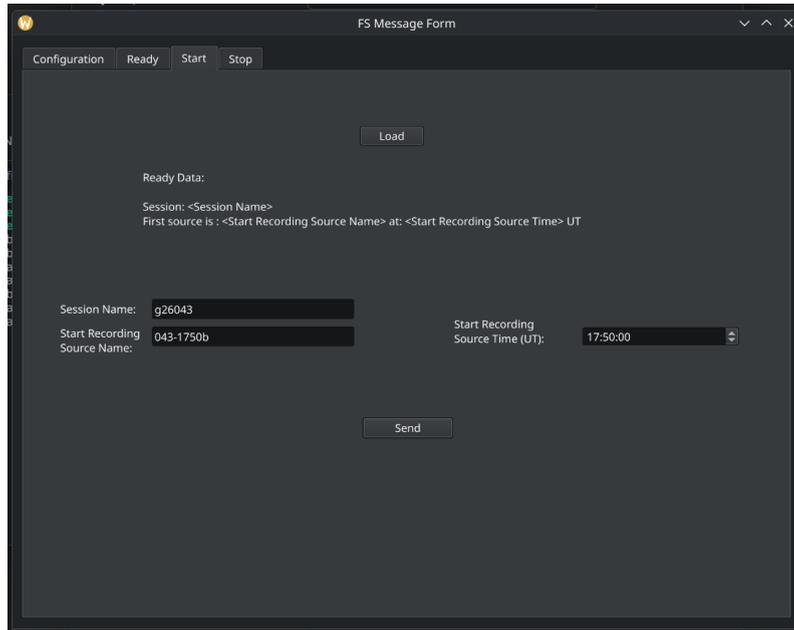


Figure 6.2: Start page after loading the data

7 Stop Page

The Stop page follows the same automated loading logic as the previous tabs. Once the JSON file has been generated, the Load button in this tab (see Figure 7.1) retrieves the required information and uses it to automatically populate the corresponding fields.

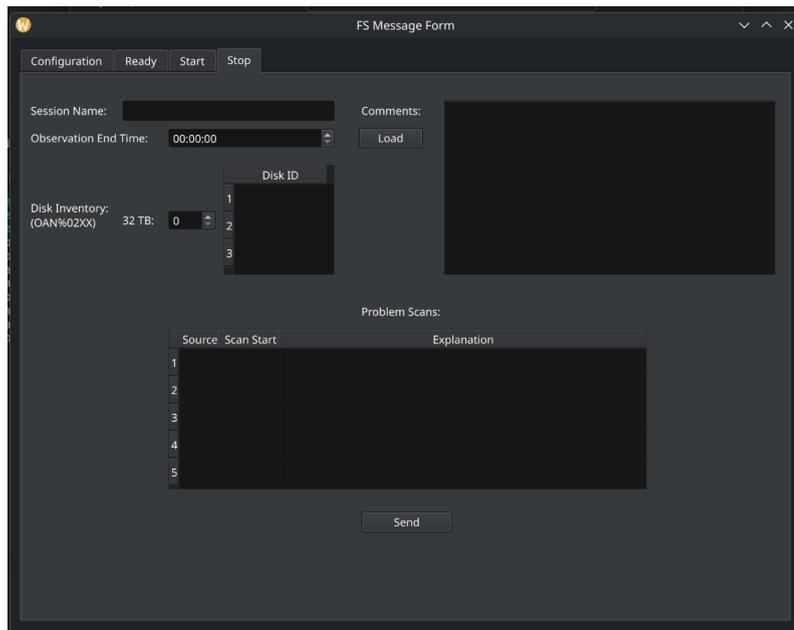


Figure 7.1: Stop page before loading the data

In particular, this page is completed with the session name and the information corresponding to the last disk record. The final result after loading the data can be seen in Figure 7.2.

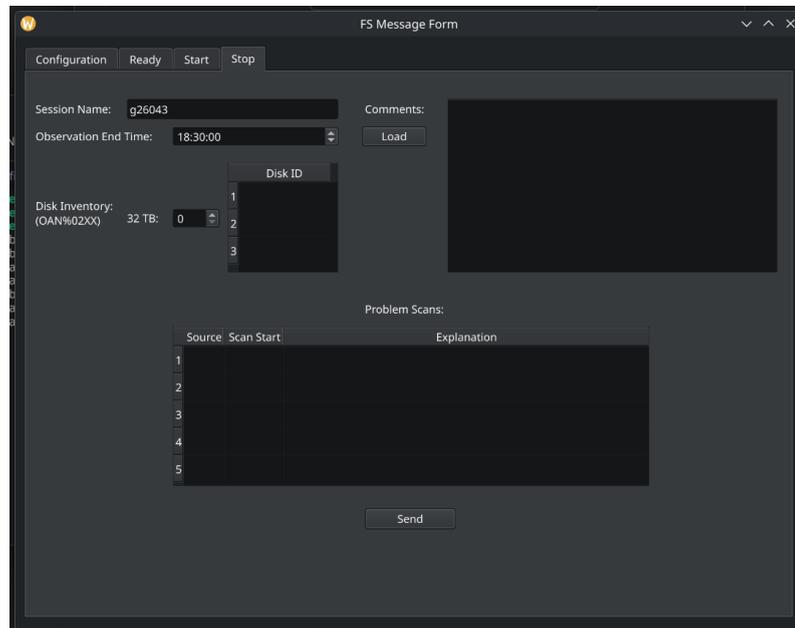


Figure 7.2: Stop page after loading the data

8 Results and Benefits

The implemented solution significantly reduced the amount of manual work required to fill in the application fields. Tasks that previously required searching through the log files, copying values manually, and performing repetitive calculations can now be completed automatically through the interface.

This improvement reduced the probability of human error, simplified the workflow for operators, and made the application faster and more convenient to use in practice. In addition, the use of an intermediate JSON structure made the solution more modular and easier to maintain or extend in the future.

A Appendix: Codes

Listing 1: Initialization of the session data structure

```

1 session_data = {
2     "session_name": session_name,
3     "sky_conditions": None,
4     "first_source": None,
5     "first_source_time": None,
6     "first_source_exper_initi": None,
7     "first_source_time_exper_initi": None,
8     "end_source": None,
9     "end_source_time": None,
10    "cryogenic_prev": [],
11    "cryogenic_post": [],
12    "location": [],
13    "gps_stm": [],
14    "wx": [],
15    "cdms": [],
16    "rdtca_dot": [],
17    "rdtca_dot2gps": [],
18    "rdtca_dot2pps": [],
19    "rdtcb_dot": [],
20    "rdtcb_dot2gps": [],
21    "rdtcb_dot2pps": [],
22    "rdtcc_dot": [],
23    "rdtcc_dot2gps": [],
24    "rdtcc_dot2pps": [],
25    "rdtcd_dot": [],
26    "rdtcd_dot2gps": [],
27    "rdtcd_dot2pps": [],
28    "disk_record": [],
29    "onoff_pre_initi": [],
30    "onoff_post_initi": [],
31    "scans": [],
32    "_pending_scan": [],
33    "onsource_blocks": [],
34    "_pending_onsource": None
35 }
```

Listing 2: Function used to estimate sky conditions from cloud cover

```

1 def sky_from_cldc(cldc):
2     """Devuelve un string con el estado del cielo a partir de la nubosidad media (cldc)
3     Args:
4         cldc: Nubosidad media (cldc) en octas (0-8)
5     Returns:
6         Un string con el estado del cielo
7     """
8     if pd.isna(cldc):
9         return None
10    if cldc <= 1:
11        return "Despejado"
12    elif cldc <= 3:
13        return "Poco nuboso"
14    elif cldc <= 5:
15        return "Intervalos nubosos"
16    elif cldc <= 7:
17        return "Muy nuboso"
18    else:
19        return "Cubierto"
```