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Report on the first data services

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CONTENTS

Introduction	5
Data Services	6
The Data Stewardship Wizard (DSW)	6
STM Neural Network	10
STM Metadata Explorer	12
SEM Classifier	14
References	15



INTRODUCTION

The present deliverable describes the initial set of data services made available to the NEP community. It is structured in four main sections: Section 1 describes the Data Stewardship Wizard (DSW) and its functionalities, Section 2 the STM Neural Network, Section 3 the STM Metadata Explorer and Section 4 the SEM Classifier.



DATA SERVICES

1. The Data Stewardship Wizard (DSW)

The Data Stewardship Wizard (DSW) is an open-source questionnaire tool that ensembles all the answers in a Data Management Plan (DMP). It is based on FAIR data stewardship, covering all the different aspects of data management and FAIR principles from data collection to publication and preservation. Through the smart questionnaire, users are guided by resources, recommendations and definitions when filling the answers.

A self-hosted DSW instance was deployed for the NEP project to support researchers on creating the DMP for their laboratories. The instance can be reached at: <https://dsw.nffa.eu/> and accessed using NFFA-Europe credentials and login as reported in Figure 1. It has three types of users: Admin, Data steward and Researcher.

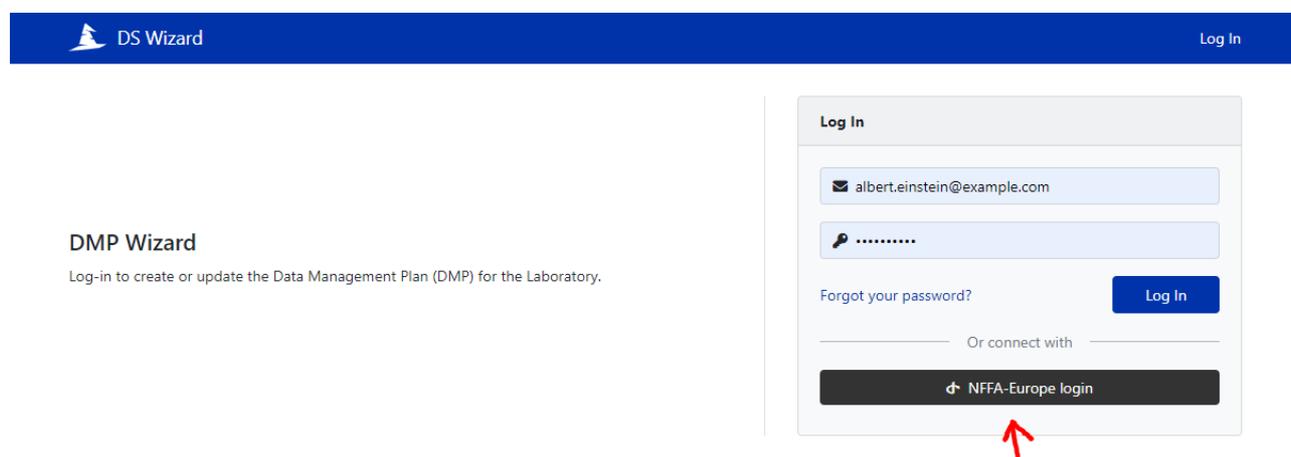


Figure 1. DS Wizard Login page

After login, the instance provides a dashboard page with a sidebar menu (Figure 2), containing:

- **Knowledge Models:** consisting of templates of DS Wizard and the personalized templates created for the NEP project
- **Projects:** contain the list of the DMPs created using the instance
- **Create new DMP:** by using the last version of the Knowledge Model here the user can start creating the DMP by answering the questionnaire
- **Quickstart Guide**¹: a pdf file with step by step instructions for using the tool
- **Video tutorial**²: a brief video demonstration on how to fill the DMP

¹ Quickstart Guide document: <https://datashare.nffa.eu/index.php/s/e534WHsR8wBLpyr>

² Video tutorial: <https://datashare.nffa.eu/s/Tf9LRcBqqwFHQFW>

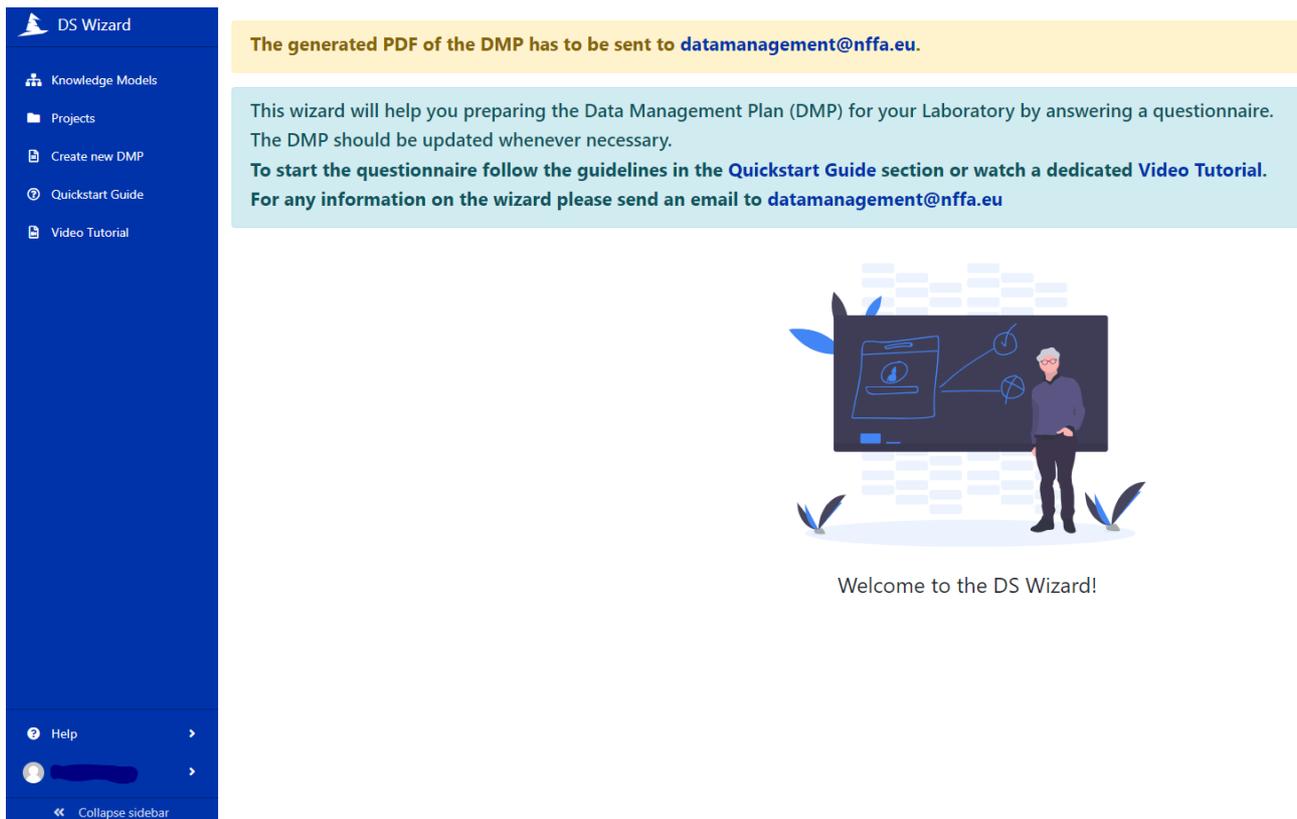


Figure 2. DS Wizard dashboard

The Knowledge model (KM) represents the content of the DSW that is used to create a Questionnaire that researchers fill in to generate a DMP. It contains different chapters with questions, external resources, definitions and recommendations.

Knowledge Model Editor: the DSW contains a system to maintain Knowledge Models and to adapt or create new ones from scratch. This functionality is available for Data Steward users. Using this system, a personalized Knowledge Model was customized for NEP Laboratories with specific questions, references and recommendations in line with FAIR principles and Open Access obligations (Figure 3). By using the KM editor, Knowledge Models can be edited directly in the wizard.

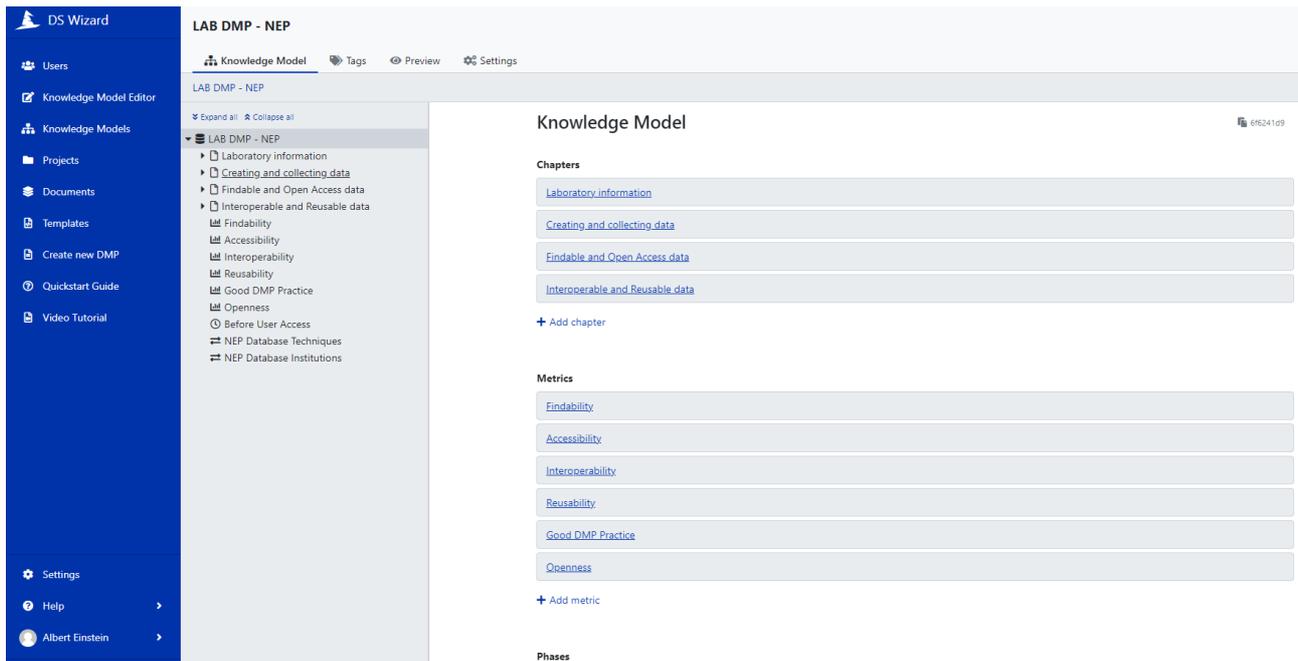


Figure 3. DS Wizard Knowledge Model Editor used to create Lab-NEP Knowledge Model

DS Wizard Questionnaire: contains questions and provides integrations in form of recommendations, definitions and resources. Most of the questions are closed, with options or lists of items, while a limited number are open questions (Figure 4).

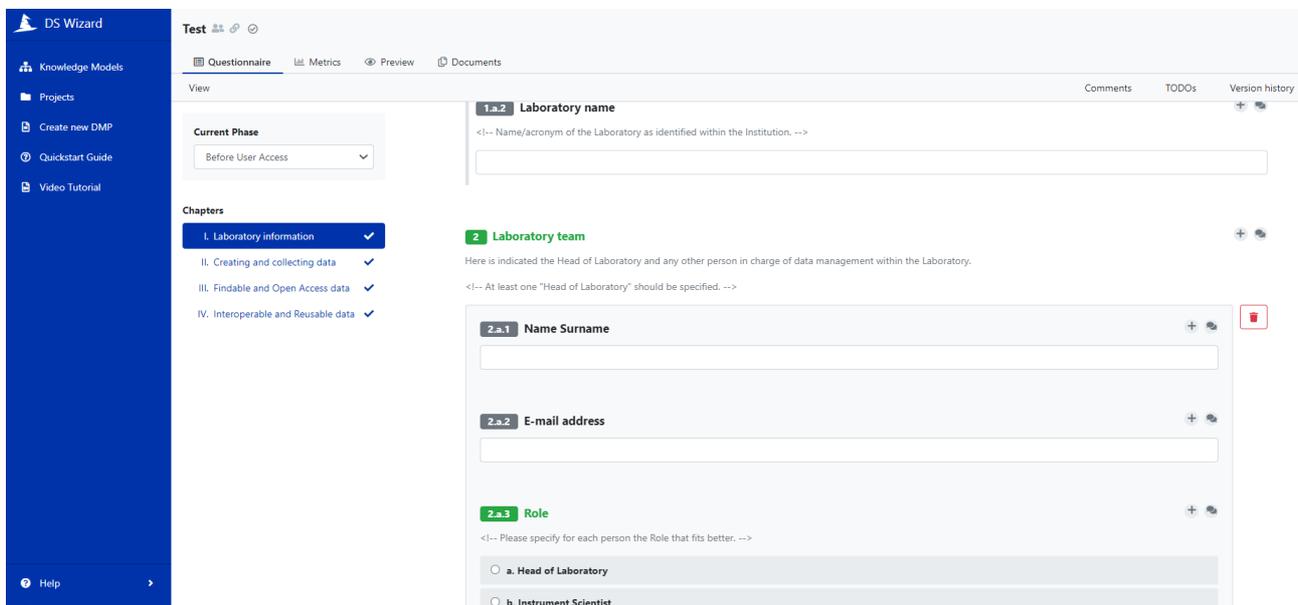


Figure 4. DS Wizard questionnaire based on the customized Knowledge Model created for the NEP project

For some questions, FAIR metrics were used as indicators to give an estimation of the FAIRness of data as by the summary report in Figure 5.

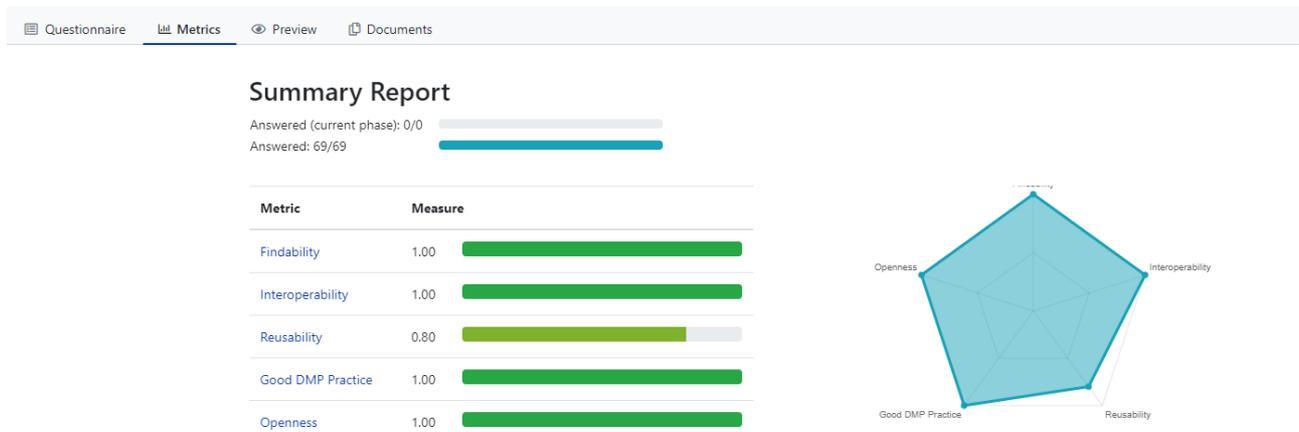


Figure 5. DS Wizard FAIR metrics

Data Management Plan (DMP): DSW assembles all the answers and generates a DMP document by using a selected template. The DMP can be exported in PDF, MS Word, HTML, LaTeX or JSON format, allowing the machine-actionability of the DMPs as reported in Figure 6. The DMP generated in the selected format can be downloaded and shared.

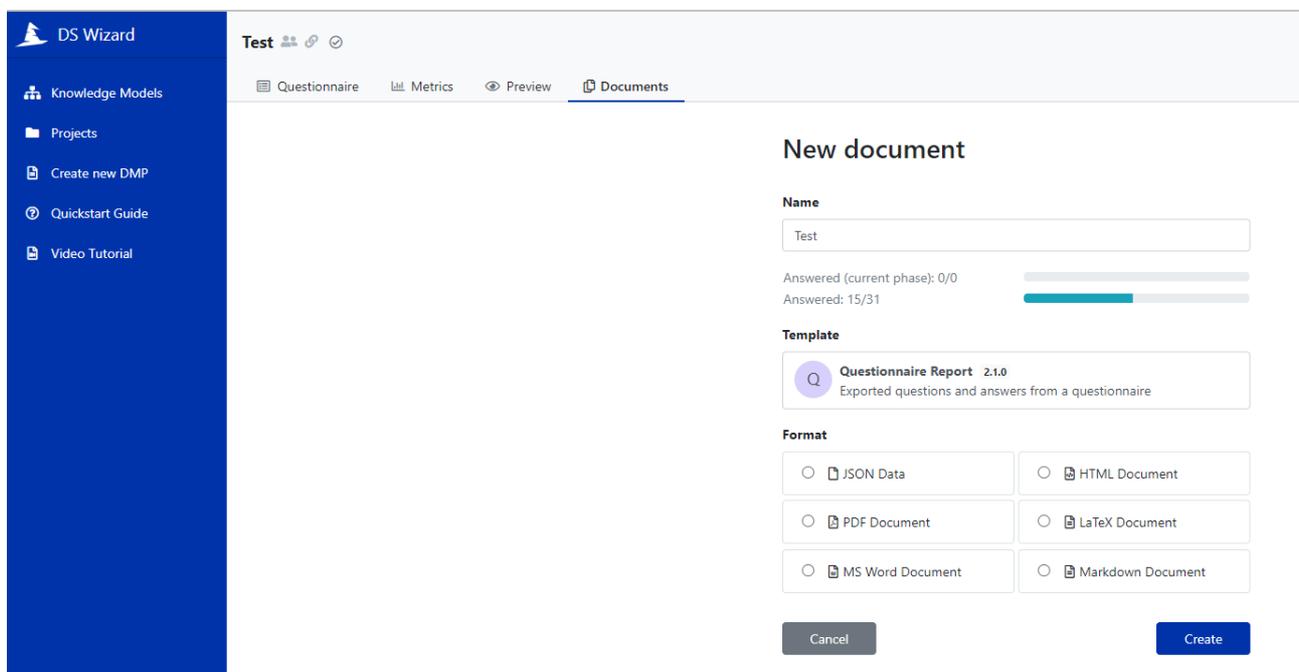


Figure 6. DS Wizard DMP generation formats

2. STM Neural Network

STM Neural Network is a machine learning based service for content-based image retrieval (CBIR). Content-Based Image Retrieval (CBIR) aims to find the similar images from a large scale dataset against a query image. The similarity between the representative features of the query image and dataset images is used to rank the images for retrieval. Convolutional Neural Networks (CNN) are especially suitable feature extractors for CBIR. Rather than using the classification output of the network, the information is taken from a layer prior to the output layer. These vectors contain image descriptive properties and are therefore used as a feature for CBIR.

STM Neural Network is used to retrieve a subset of Scanning Tunneling Microscopy (STM) images similar to an input STM image. Researchers use this service to efficiently query a dataset of STM images to quickly find images relevant for their purposes.

STM Neural Network is implemented in Python, uses FastAI Machine Learning library and spym [3] package to load, process and plot images, and FASTAPI for the creation of REST endpoints. Users can programmatically use the service within a jupyter notebook in a few lines of code, as shown in Figure 7.

```
# single image similarity example:
# 85404 is the image unique identifier, 24 is the number of similar images to extract
similar_images_df = get_similar_images_euclidean(img_repr_df, 85804, 24)

# plot similar images with input image in the center
show_similar_images(similar_images_df)
```

Figure 7: STM Neural Network example query

In the example query above, the image with a unique identifier of 85804 is used as input to search for 24 similar images in the STM dataset. Images retrieved are presented in a plot with the input image in the center (Figure 8).

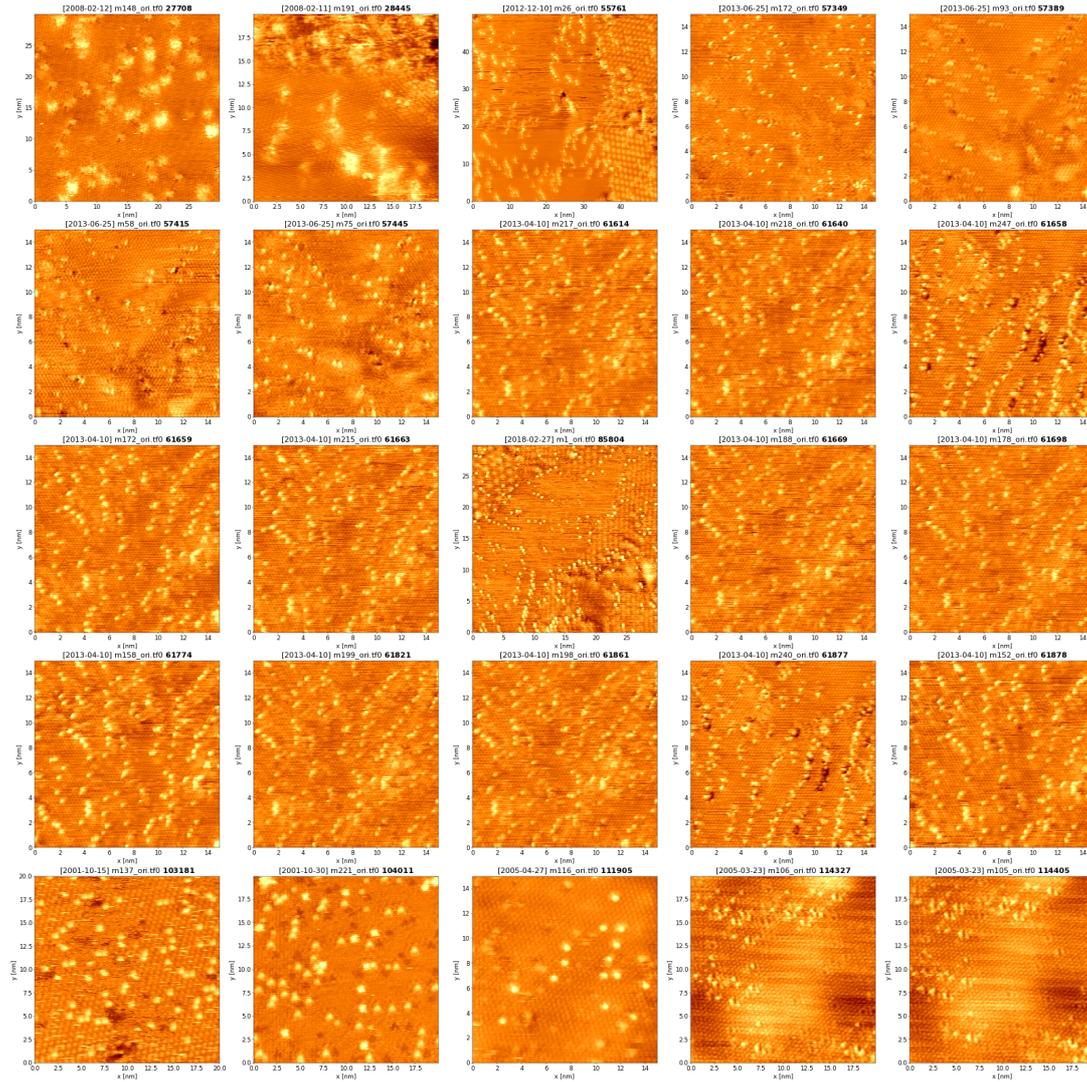


Figure 8: STM Neural Network extracted images from example query. The input image is the one in the center.

3. STM Metadata Explorer

The STM Metadata Explorer is an easy-to-use and interactive web service integrated within the Trieste Advanced Data Services (TriDAS <https://tridas.nffa.eu/>). By selecting a small subset of metadata that provides significant information about the measurement settings, it allows to visually explore the STM images through interactive and downloadable plots.

The workflow is summarized in Figure 9 and explained in detail in a guidelines document [1]. Users can visually explore images metadata through a quantile plot for a single metadata field or a scatter plot showing the distribution of images between two chosen metadata fields. In these interactive plots, some information about the metadata pop up by hovering on top of them. In particular, from scatter plots it is possible to retrieve a table containing metadata fields for each image in the subset, which can be selected, ordered and filtered based on their metadata values. These features are useful for a first exploration of the dataset but images can also be downloaded for further analysis and image visualization: by clicking on the unique identifier, the corresponding STM image is shown, allowing to download the related data, metadata and provenance.

The web service workflow on the TriDAS website

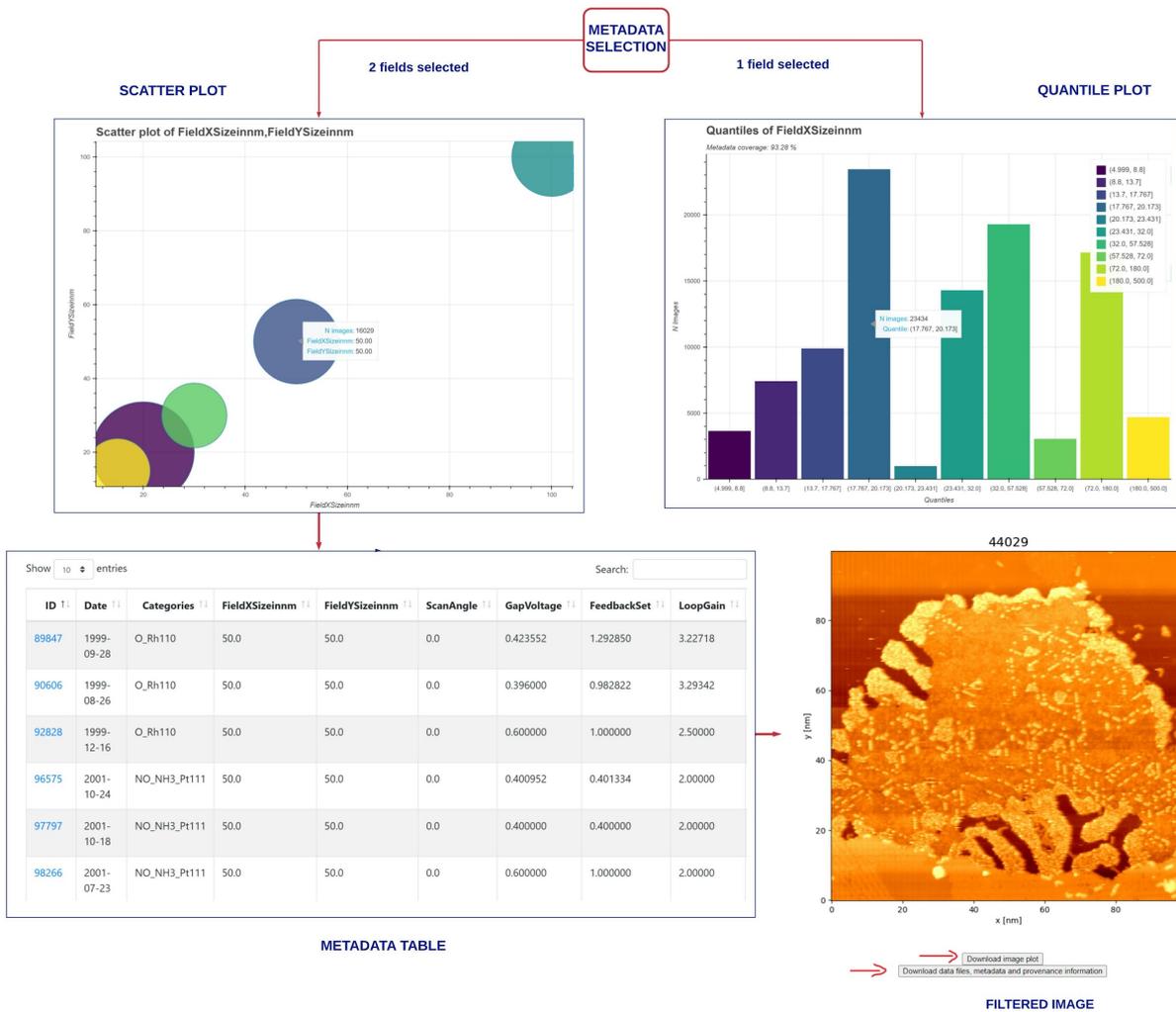


Figure 9. The workflow of the webservice on the TriDAS website

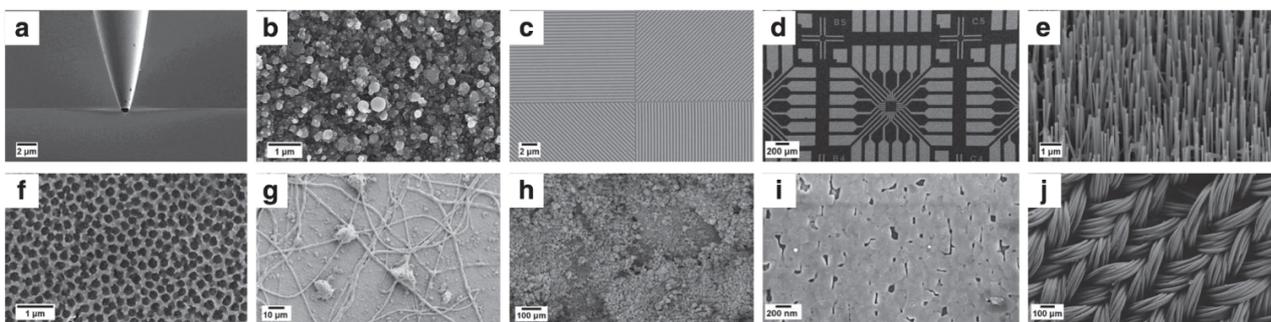
TriDAS is implemented in Python, and uses Bokeh [2] for data visualization, spym [3] package to process and plot images, and Flask [4] framework as backend. The dataset of STM images together with their provenance description [5] and all the source code of the software developed [6] are publicly available.



4. SEM Classifier

Scanning Electron Microscopy (SEM) classifier is a data service web service integrated within the Trieste Advanced Data Services (TriDAS <https://tridas.nffa.eu/>) that allows researchers to automatically classify and tag SEM images in a uniform way. After the classification done in collaboration with SEM researchers, 10 categories that cover a broad range of SEM images were identified: Biological, Fibres, MEMS devices and electrodes, Particles, Porous Sponge, Tips, Films and Coated Surfaces, Nanowires, Patterned surfaces, Powder (Figure 10). The classification is performed using convolutional neural networks (CNN) [7].

Figure 10. Representative images for each of the ten categories



The researcher can upload an image or a folder and after a few seconds per image can obtain the resulting categories, together with the rate (a measure of “certainty”).

Effective nanoimages recognition

Welcome to the [NFFA-EUROPE](#) online SEM classifier. Use this service to automatically classify and tag your SEM images. If you are not satisfied with the result, you can manually insert a new category. Help us improving! When you have finished, you can download the complete summary of your results.

Login with your NextCloud or NFFA-EUROPE account to immediately store your images on your personal NextCloud on [datashare-iom](#) or to publish to the [IDRP](#).

The classification is performed by an [Inception-v3](#) deep convolutional neural network, trained on [this](#). Interested in details? Read [our paper](#)

Drag images file here or browse from your computer

Figure 11. The SEM Classifier home page

The researcher can confirm and download the result or change it by adding new categories.

The SEM classifier is implemented in Python, uses TensorFlow Machine Learning library and FASTAPI for the creation of REST endpoints.

References

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