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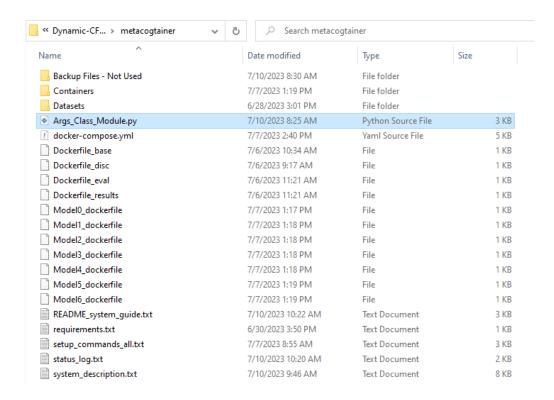
System Usage Overview:

Setup Overview:

- First, you'll need to install Docker Desktop which you can find here: https://www.docker.com/
 - This will allow you to manage created Images and Containers easier.
- Then, you'll need to install Docker in your terminal, you can do that easily with: pip install docker
 - This will allow you to easily create and run Containers.
 - If you are unable to run commands from the terminal, try starting Docker Desktop if you have not already.
- Lastly, you'll need the code. Download the Dynamic-CFAR GitHub folder to your computer.
 - Only copy the required data file in the ./Datasets folder
 (All files will be copied into the Docker file system, thus doubling the storage size)
- Open a terminal, such as Bash or Powershell, and navigate to the Dynamic-CFAR/metacogtainer folder
 - To make the code run faster, you can build the Docker Images before running the code
 - These commands can be found in the README system guide.txt
- Run the following command: docker compose up

- The output will appear in the terminal after all the containers have run.
 - For explanations on how to run containers individually, please view setup_commands_all.txt
- To close the system run the following command: docker compose down

Code Overview:



To begin with a major change, the argument parser has been moved into a separate class file. This means that optional parameters can not be passed through the command line anymore, and are instead set in a configuration file.

This can be found in 'Args Class Module.py' in the /Dynamic-CFAR/metacogtainer folder.

```
class Args_Class
      def __init__(self,
exp_index=None,
      exp_type=None,
      label='Test'
      results_path='./results',
      algorithm='amf',
      N=16.
      P_fa=1/np.power(10,4),
      sample_len=16,
      PRI=1e-8,
      f_d=2e7
      target=False,
      max_test=50,
      discriminator='./classifier/ordered a Dense200 50 drop 0 100 LR 0 000100 model',
     model_thresh4='./classifier/P_14_Dense1000_200_50_drop_0_100_LR_0_000100_model'
model_thresh5='./classifier/P_44_Dense1000_200_50_drop_0_100_LR_0_000100_model'
model_thresh6='./classifier/P_84_Dense1000_200_50_drop_0_100_LR_0_000100_model'
model_thresh1='./classifier/K_54_Dense1000_200_50_drop_0_100_LR_0_000100_model'
      model_thresh2='./classifier/K_14_Dense1000_200_50_drop_0_100_LR_0_000100_model
      model_thresh3='./classifier/K_44_Dense1000_200_50_drop_0_100_LR_0_000100_model'
model_final='./classifier/G_GF4_Dense500_50_drop_0_100_LR_0_000100_model',
      alg='glrt',
```

A few more arguments have also been added for further customization, these include:

- run ideal
- show output.

run_ideal determines if the system will calculate the Ideal False Alarm rate, which is used for testing.

- When false, the system will not display output relating to this metric.

The show_output variable sets whether or not each container will output its calculations when finished.

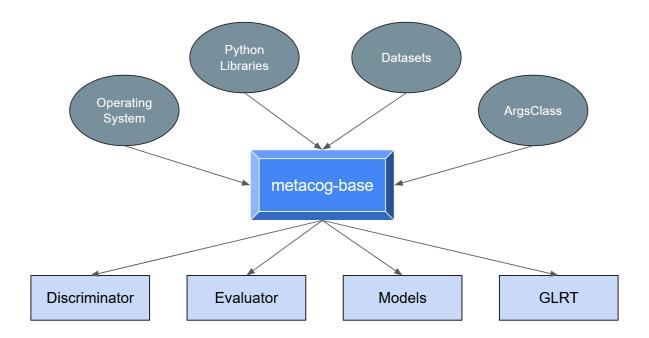
This is a lot of text and will include the probability distributions for each data point from the discriminator and the highest probability from the evaluator.

Because all the containers copy the same common file from the main directory, changes only need to be made once to affect all containers.

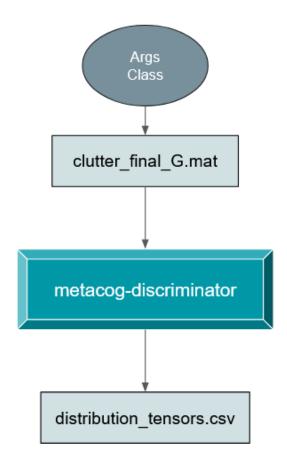
**However, when a change is made to a container, it needs to be built again for the changes to have any effect.

This is because containers are isolated systems and are not connected to your host machine.

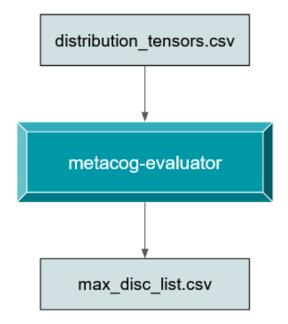
Next, the code is segmented into various parts within different containers:



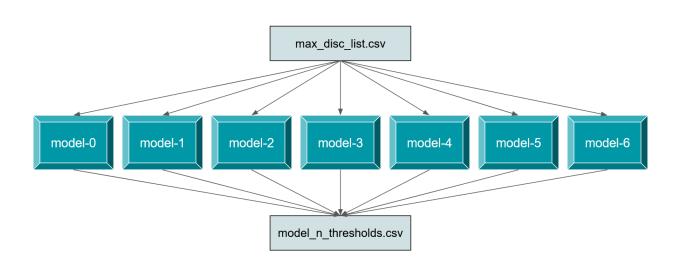
- (1) The first container, **metacog-base**, is key to the whole system.
- It installs Ubuntu as well as the required libraries found in ./requirements.txt
- This is vital as it ensures that all of this data only needs to be installed once.
- Base also copies the dataset and configuration file (ArgsClass) to the shared storage volume
- The following containers all inherit from this base image, thus the code is not reinstalled.
- Due to this efficiency, building the containers is now much faster.
- Although this base does not execute any commands for the algorithm, it is essential to initialize common dependencies.



- (2) The second container, **metacog-discriminator**, runs the initial processing of the specified data.
- Currently, this data file is found in metacogtainer/Datasets/clutter_final_G.mat
- metacog-discriminator currently calls a modified version of the 'runDet' function found in the original code.
- This simply calls the Discriminator model and captures the output of the Softmax classification before it is evaluated.
- This is a probability distribution for each data point.
- Because of difficulty saving this data from a Tensor to a csv file, the data is converted to a Numpy array before saving.
- This output is saved to 'distribution_tensors.csv'



- (3) The third container, **metacog-evaluator**, processes these probability distributions.
- Currently, it simply loads the data, and calculates the argmax for each array.
- Argmax finds the element in the array with the highest value.
- In this scenario, it finds the element with the highest probability.
- After doing so for each data point, the data is saved as a Numpy array with the argmax of each.
- This output is saved to 'max_disc_list.csv'



(4) The fourth through tenth container (seven total) are the different threshold models used to predict specific thresholds.

The image names are as follows:

model-0, model-1, model-2, model-3, model-4, model-5, model-6.

In each container the output file from the evaluator is loaded.

For each data point in this file, there is a check to see if the selected model number (from the evaluator) is equal to the current model number.

If so, the corresponding model is called to predict a corresponding threshold.

For each data point where calculations are used, the current index and the predicted threshold are appended to an array.

At the end of the loop through all data points, the list of index and thresholds is saved to the file: model_n_thresholds.csv

This structure was designed so that all models can be run in parallel and still hold efficacy for mixed clutter and retain grouping of specific data points with their specific thresholds while isolated.

model_n_thresholds.csv

metacog-glrt

false_alarms.csv

(5) The last / eleventh container, metacog-glrt, gathers the thresholds and index from the seven model containers then calculates the glrt and outputs the number of false alarms to the terminal and saves to a results file.

This container appends all the threshold/index arrays from the model containers.

Then it loops through each data point in the arrays and calculates detection using the index of each data point alongside its corresponding threshold.

After looping over all the data, it outputs the number of false alarms to the terminal and saves to:

File Structure Overview:

• The main directory contains a containers folder, a datasets folder, the Args_Class_Module.py file, the image configuration files (Dockerfiles), and system guides.

The containers folder contains files needed by specific containers.

The datasets folder contains the common data file used by the containers.

The Args_Class_Module.py is the common configuration file that has replaced the parser for command line argument.

The Dockerfiles define how each Image is built before Containers are run.

The system guide provides documentation on how the architecture works, how the code has changed, and also specific commands to run individual containers.

The Base Image copies the common ./Datasets data file (clutter_final_G.mat currently), and the common Args_Class_Module.py file, thus all the following Images inherit these files.

Each Image is built by copying the corresponding files from ./Containers.

The container files contain the python scripts needed to run individual code segments, such as the discriminator, evaluator, models, and glrt.

Architecture Overview:

This system of containers is partially sequential.

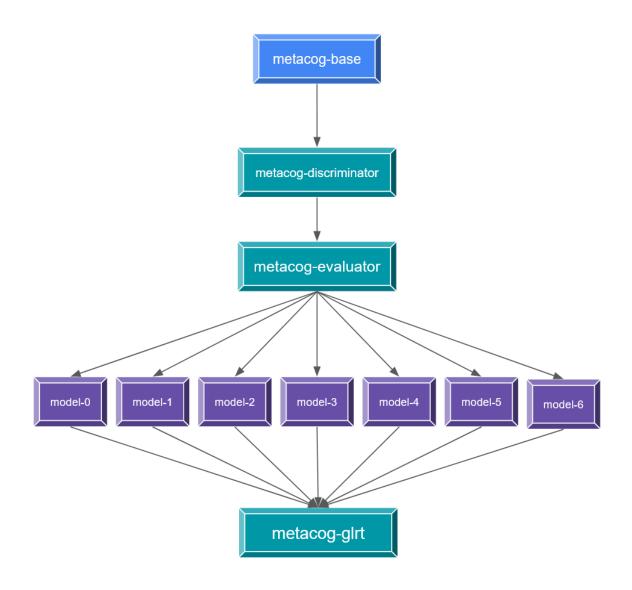
First the metacog-base is built.

Once this has finished, the metacog-discriminator is run to generate probability distributions.

Then once this has been generated, the metacog-evaluator is run to process this data and generate argmax values.

Now, with the argmax array, each model is run at the same time in parallel to process each data point and predict thresholds.

Once all seven models have finished running, the metacog-glrt container is run to gather thresholds and data points, then calculate GLRT.



System Usage Overview:

Everything can be done by simply running the following command in the terminal: docker compose up

Note: If you do not have the images built before running this command it will build them for you. However, this method is much slower.

It is recommended to build the images manually from the terminal prior to running the system for efficiency.

The commands to build the images can be found in the metacogtainer/setup commands all.txt file (the second block of code)

(base) PS C:\Users\Ryan\Documents\GitHub\Dynamic-CFAR\metacogtainer> docker compose up

```
metacogtainer-results-1 | CD: 0.0 / GLRT: 0.0 / Ideal: 2.0 metacogtainer-results-1 exited with code 0
```

```
(base) PS C:\Users\Ryan\Documents\GitHub\Dynamic-CFAR\metacogtainer> docker compose down
Container metacogtainer-results-1
Container metacogtainer-model-5-1
Container metacogtainer-model-6-1
Container metacogtainer-model-4-1
Container metacogtainer-model-3-1
                                          Removed
Container metacogtainer-model-0-1
                                         Removed
Container metacogtainer-model-2-1
Container metacogtainer-model-1-1
Container metacogtainer-evaluator-1
Container metacogtainer-discriminator-1 Removed
Container metacogtainer-base-1
Network metacogtainer_default
(base) PS C:\Users\Ryan\Documents\GitHub\Dynamic-CFAR\metacogtainer> _
```