# 01 callbacks tmp

August 15, 2025

```
[1]: | # -*- coding: utf-8 -*-
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     #
     #
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```

## 1 CIL Callbacks How To

Callbacks are an essential tool in many optimization frameworks, providing a flexible way to monitor and control the execution of algorithms. Similar to popular packages such as keras, scipy, and pytorch, CIL utilizes callbacks that can be passed to the run method of an algorithm. These callbacks have access to the entire state of the algorithm, allowing them to perform a wide range of tasks—from logging and saving progress to implementing early stopping conditions. By integrating callbacks, users can enhance the flexibility, efficiency, and functionality of their algorithms.

In this demo, we explore the default behavior of callbacks in CIL and present a variety of example callbacks that can be customized to suit your needs. New callbacks are regularly added to CIL based on user requests, so be sure to check out the documentation for the latest updates.

#### 1.1 Install CIL and set some defaults

```
[2]: from cil.utilities.display import show2D from cil.recon import FDK from cil.processors import TransmissionAbsorptionConverter, Slicer
```

```
import numpy as np
import matplotlib.pyplot as plt
from cil.plugins.tigre import ProjectionOperator
from cil.optimisation.algorithms import FISTA
from cil.optimisation.functions import LeastSquares, TotalVariation
from cil.optimisation.utilities import callbacks
from cil.framework import DataContainer

# set up default colour map for visualisation
cmap = "gray"

# set the backend for FBP and the ProjectionOperator
device = 'gpu'
```

#### 1.2 Load Data

In this example, we utilize CIL's simulated sphere data. To accelerate computations in this note-book, we extract a 2D slice from the 3D dataset. Additionally, we select a subset of angles to create a limited-angle reconstruction scenario. We will then compare the ground truth data with a filtered back projection (FBP) reconstruction under these limited-angle conditions.

```
[3]: from cil.utilities import dataexample
# Load data
ground_truth = dataexample.SIMULATED_SPHERE_VOLUME.get()

data = dataexample.SIMULATED_CONE_BEAM_DATA.get()

# Consider just a single 2D slice
data = data.get_slice(vertical='centre')
ground_truth = ground_truth.get_slice(vertical='centre')

absorption = TransmissionAbsorptionConverter()(data)
absorption = Slicer(roi={'angle':(0, -1, 5)})(absorption)

ig = ground_truth.geometry

recon = FDK(absorption, image_geometry=ig).run()

show2D([ground_truth, recon], title = ['Ground Truth', 'FDK Reconstruction'],u
origin = 'upper', num_cols = 2)
New geometry: 2D Cone-beam tomography
```

System configuration:

Source position: [ 0., -20000.]

Rotation axis position: [0., 0.]

Detector position: [ 0., 60000.]

Detector direction x: [1., 0.]

Panel configuration:

Number of pixels: [128 1]

Pixel size: [64. 64.] Pixel origin: bottom-left

Channel configuration:

Number of channels: 1

Acquisition description:

Number of positions: 60

Angles 0-9 in degrees: [ 0., 6., 12., 18., 24., 30., 36., 42., 48.,

54.]

Angles 50-59 in degrees: [300., 306., 312., 318., 324., 330., 336.,

342., 348., 354.]

Full angular array can be accessed with acquisition\_data.geometry.angles

Distances in units: units distance

Shape out: (60, 128)

New geometry shape: (60, 128)

FDK recon

Input Data:

angle: 60

horizontal: 128

Reconstruction Volume:

horizontal\_y: 128
horizontal\_x: 128

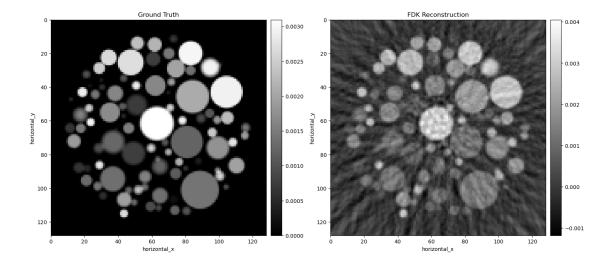
Reconstruction Options:

Backend: tigre Filter: ram-lak

Filter cut-off frequency: 1.0

FFT order: 8

Filter\_inplace: False



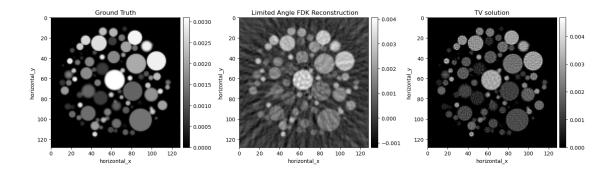
[3]: <cil.utilities.display.show2D at 0x7f607a3024b0>

#### 1.3 Default Behavior

0%1

When no callback is passed to the run method, a progress bar is automatically displayed. This progress bar provides useful information, including the time elapsed, estimated time remaining, iterations per second, and the current objective value. Keep in mind that the current objective value updates at intervals determined by the update\_objective\_interval parameter set in the algorithm.

| 0/500 [00:00<?, ?it/s]



[4]: <cil.utilities.display.show2D at 0x7f607a5e16a0>

#### 1.4 Available CIL Callbacks

Callbacks in CIL are provided as a list to the run method of an algorithm. In this example, we use two built-in CIL callbacks: callbacks.ProgressCallback() and callbacks.TextProgressCallback(). The first is the default callback that displays a progress bar, as demonstrated earlier. The second, callbacks.TextProgressCallback(), prints progress updates at intervals specified by update\_objective\_interval.

It's important to note that if you don't pass any callbacks, the default behavior is to display the progress bar. However, if you choose to include other callbacks and still want the progress bar, you must explicitly include callbacks.ProgressCallback() in your list.

Each callback is run at each iteration, after any update step is taken. In the following example, due to the design of the callback, the display is only updated and/or printed every update objective interval but the callback is still called at each iteration.

```
[5]: algo=FISTA(initial=ig.allocate(0), f=F, g=G, update_objective_interval=10)
cb1=callbacks.ProgressCallback()
cb2=callbacks.TextProgressCallback()
algo.run(500, callbacks=[cb1, cb2])
show2D([ground_truth, recon, algo.solution], title = ['Ground Truth', 'FDK_
Reconstruction', 'TV solution'], origin = 'upper', num_cols = 3)
```

```
0%1
             | 0/500 [00:00<?, ?it/s]
   0/500
                ?it/s
  10/500
            26.45it/s, objective=+8.586e+01
  20/500
            26.33it/s, objective=+9.047e+00
  30/500
            26.36it/s, objective=+2.640e+00
  40/500
            26.38it/s, objective=+1.188e+00
  50/500
            26.07it/s, objective=+6.928e-01
  60/500
            26.15it/s, objective=+4.585e-01
```

```
70/500
              26.24it/s, objective=+3.451e-01
    80/500
              26.24it/s, objective=+2.741e-01
    90/500
              26.27it/s, objective=+2.337e-01
   100/500
              26.23it/s, objective=+2.062e-01
   110/500
              26.29it/s, objective=+1.870e-01
   120/500
              26.38it/s, objective=+1.729e-01
   130/500
              26.45it/s, objective=+1.623e-01
   140/500
              26.55it/s, objective=+1.541e-01
              26.54it/s, objective=+1.476e-01
   150/500
   160/500
              26.45it/s, objective=+1.421e-01
   170/500
              26.49it/s, objective=+1.375e-01
   180/500
              26.47it/s, objective=+1.336e-01
   190/500
              26.33it/s, objective=+1.303e-01
  200/500
              26.30it/s, objective=+1.275e-01
              26.29it/s, objective=+1.250e-01
  210/500
  220/500
              26.29it/s, objective=+1.228e-01
  230/500
              26.01it/s, objective=+1.208e-01
   240/500
              26.02it/s, objective=+1.190e-01
              25.83it/s, objective=+1.173e-01
   250/500
   260/500
              25.92it/s, objective=+1.159e-01
Exception in thread Thread-5:
Traceback (most recent call last):
  File
"/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/threading.py",
line 1075, in _bootstrap_inner
    self.run()
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/tqdm/_monitor.py", line 84, in run
    instance.refresh(nolock=True)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/tqdm/std.py", line 1347, in refresh
    self.display()
  File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/tqdm/notebook.py", line 157, in display
    pbar.value = self.n
  File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
```

packages/traitlets/traitlets.py", line 716, in \_\_set\_\_

```
self.set(obj, value)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/traitlets/traitlets.py", line 706, in set
    obj._notify_trait(self.name, old_value, new_value)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/traitlets/traitlets.py", line 1513, in _notify_trait
    self.notify change(
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipywidgets/widgets/widget.py", line 700, in notify_change
    self.send_state(key=name)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipywidgets/widgets/widget.py", line 586, in send_state
    self._send(msg, buffers=buffers)
  File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipywidgets/widgets/widget.py", line 825, in _send
    self.comm.send(data=msg, buffers=buffers)
  File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/comm/base_comm.py", line 144, in send
    self.publish_msg(
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipykernel/comm/comm.py", line 42, in publish_msg
    parent=self.kernel.get_parent(),
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipykernel/kernelbase.py", line 783, in get_parent
    return self._shell_parent.get()
LookupError: <ContextVar name='shell_parent' at 0x7f62b66b67a0>
   270/500
              25.59it/s, objective=+1.145e-01
   280/500
              25.86it/s, objective=+1.134e-01
  290/500
              26.02it/s, objective=+1.123e-01
  300/500
              26.06it/s, objective=+1.113e-01
  310/500
              26.19it/s, objective=+1.104e-01
  320/500
              26.30it/s, objective=+1.096e-01
              26.23it/s, objective=+1.088e-01
  330/500
  340/500
              26.23it/s, objective=+1.081e-01
   350/500
              26.26it/s, objective=+1.075e-01
   360/500
              26.24it/s, objective=+1.068e-01
  370/500
              25.99it/s, objective=+1.063e-01
```

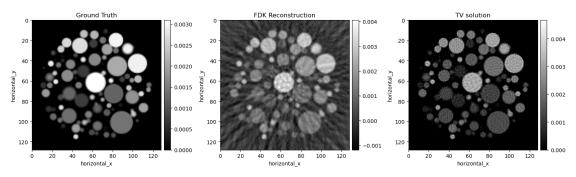
26.07it/s, objective=+1.057e-01

26.01it/s, objective=+1.052e-01

380/500

390/500

```
400/500
           26.04it/s, objective=+1.048e-01
410/500
           26.15it/s, objective=+1.043e-01
420/500
           26.10it/s, objective=+1.039e-01
430/500
           26.16it/s, objective=+1.035e-01
           26.09it/s, objective=+1.031e-01
440/500
450/500
           26.14it/s, objective=+1.028e-01
460/500
           26.19it/s, objective=+1.025e-01
470/500
           26.21it/s, objective=+1.022e-01
           26.30it/s, objective=+1.019e-01
480/500
490/500
           26.40it/s, objective=+1.016e-01
500/500
           26.38it/s, objective=+1.013e-01
```



## [5]: <cil.utilities.display.show2D at 0x7f607a344f20>

When you call the algorithm's run method again, it resumes from where it left off. To start fresh, you should redefine both the algorithm and the callbacks.

# [6]: algo.run(100, callbacks=[cb1, cb2])

510/?	18.41it/s,	objective=+1.011e-01
520/?	20.31it/s,	objective=+1.008e-01
530/?	21.90it/s,	objective=+1.006e-01
540/?	23.07it/s,	objective=+1.004e-01
550/?	24.00it/s,	objective=+1.002e-01
560/?	24.68it/s,	objective=+1.000e-01
570/?	24.79it/s,	objective=+9.986e-02
580/?	25.27it/s,	objective=+9.969e-02

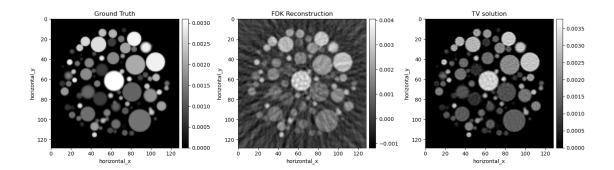
```
590/? 25.67it/s, objective=+9.953e-02
600/? 25.91it/s, objective=+9.938e-02
```

# 1.5 Early Stopping (Custom Callback Example)

To create your own callback, define a class with a \_\_call\_\_ method. It's a good practice to subclass the CIL callbacks.Callback class. The \_\_call\_\_ method should take self and algorithm as arguments, where algorithm is an initialized CIL algorithm. This allows the callback to access and utilize any properties stored within the algorithm.

In this example, the callback raises a **StopIteration** exception if a specified stopping criterion is met, forcing the algorithm to terminate early. In this basic case, it stops if the algorithm objective is less than 0.2. You can see that the algorithm does not run for the full 500 iterations.

```
0/500
               ?it/s
 10/500
           26.74it/s, objective=+8.586e+01
20/500
           26.46it/s, objective=+9.047e+00
30/500
           26.55it/s, objective=+2.640e+00
 40/500
           26.31it/s, objective=+1.188e+00
           26.48it/s, objective=+6.928e-01
50/500
 60/500
           26.51it/s, objective=+4.585e-01
70/500
           26.63it/s, objective=+3.451e-01
           26.68it/s, objective=+2.741e-01
80/500
           26.69it/s, objective=+2.337e-01
90/500
100/500
           26.71it/s, objective=+2.062e-01
           26.67it/s, objective=+1.870e-01
110/500
110/500
           26.60it/s
```



### [7]: <cil.utilities.display.show2D at 0x7f604b3311f0>

In another early stopping example, the algorithm terminates when the current solution approaches a given reference image. This example demonstrates a callback that accepts arguments upon initialization, allowing for more flexible and customized stopping criteria. It terminates when the mean square difference between the current solution and a provided reference image is equal to or less than a provided tolerance.

```
[8]: class EarlyStoppingReference(callbacks.Callback):
            Terminates the algorithm when the mean square difference between the
      \hookrightarrowcurrent solution and a provided reference image is equal to or less than a_\sqcup
      \negprovided tolerance.
            Parameters
              _____
            ref_img: ImageData
                  The image to compare the iterates to. The algorithm will terminate \sqcup
      _{\ominus} if the current solution satisfies :math: `\frac{1}{N}//x-ref_img//_2^2<tol`,__
      →where :math:`N` is the number of pixels in the image.
            tolerance: float, default 1e-8
                  A small value which determines the sensitivity of this stopping_
      ocriterion. The algorithm will terminate if the current solution satisfies :
      \negmath: `\frac{1}{N}//x-ref_img//_2^2<tol`, where :math: `N` is the number of \Box
      \negpixels in the image.
            11 11 11
           def __init__(self, ref_img, tolerance=1e-8):
                  self.ref=ref_img
                  self.tol=tolerance
            def __call__(self, algorithm):
                  if np.mean((algorithm.solution.array-self.ref.array)**2) <= self.</pre>
      ⇔tol:
                        raise StopIteration
```

```
algo=FISTA(initial=ig.allocate(0), f=F, g=G, update_objective_interval=10)
cb=EarlyStoppingReference(ground_truth, 3e-8)
algo.run(500, callbacks=[callbacks.TextProgressCallback(),cb])
show2D([ground_truth, recon, algo.solution], title = ['Ground Truth', 'FDK_L

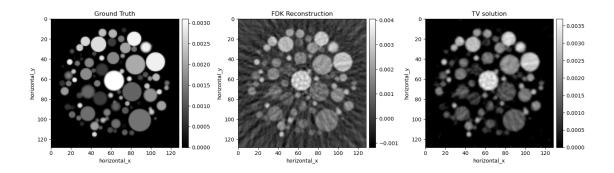
AReconstruction', 'TV solution'], origin = 'upper', num_cols = 3)
```

```
0/500 ?it/s

10/500 26.64it/s, objective=+8.586e+01

20/500 26.79it/s, objective=+9.047e+00

22/500 26.80it/s
```



#### [8]: <cil.utilities.display.show2D at 0x7f604b16a150>

# 1.6 Calculating Data Discrepancy at Each Iteration (Custom Callback Example)

In this example, a custom metric —specifically a least squares data discrepancy calculation— is computed at the end of each iteration and stored within the callback. We demonstrate how to initialize two callbacks to save the results from two different methods: FISTA with TV, with and without non-negativity enforced. The results are then plotted for comparison.

```
[9]: class LeastSquaresCallback(callbacks.Callback):

"""

Parameters

------

A: Operator

The forward operator for the least squares calculation
data: DataContainer

Acquisition data for the least squares calculation
```

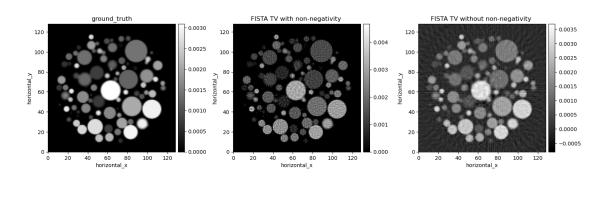
```
Properties
    save_values: list of floats
      The saved least squares calculation, one per iteration
    11 11 11
    def __init__(self, A, data):
        self.f = LeastSquares(A, data)
        self.save_values=[]
    def __call__(self, algorithm):
        self.save_values.append(self.f(algorithm.get_output()))
mycallback_FISTA_lower_bound= LeastSquaresCallback(A, absorption)
algo1=FISTA(initial=ig.allocate(0), f=F, g=alpha*TotalVariation(lower=0), u
  →update_objective_interval=10)
algo1.run(500, callbacks=[mycallback_FISTA_lower_bound])
mycallback FISTA no lower bound= LeastSquaresCallback(A, absorption)
algo2=FISTA(initial=ig.allocate(0), f=F, g=alpha*TotalVariation(), u
  →update_objective_interval=10)
algo2.run(500, callbacks=[mycallback_FISTA_no_lower_bound])
show2D([ground_truth, algo1.get_output(), algo2.get_output()],__
 ⇔title=['ground_truth', 'FISTA TV with non-negativity ', 'FISTA TV without⊔
  →non-negativity '], num_cols=3)
show2D([absorption, A.direct(algo1.get_output())-absorption, A.direct(algo2.
 oget_output())-absorption], title=['ground_truth', 'Data error FISTA TV with_
 ⇔non-negativity', 'Data error FISTA TV without non-negativity'], ⊔

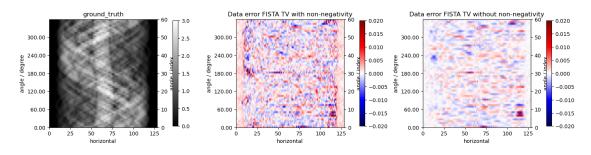
→fix_range=[[0,3], [-0.02, 0.02], [-0.02, 0.02]], cmap=['gray', 'seismic', |

 plt.plot(range(10,501), mycallback_FISTA_lower_bound.save_values[10:],
  →label='FISTA TV with non-negativity ')
plt.plot(range(10, 501), mycallback_FISTA_no_lower_bound.save_values[10:],_
 →label='FISTA TV without with non-negativity')
plt.yscale('log')
plt.ylabel('Data discrepancy')
plt.xlabel('Iteration')
plt.legend()
Exception in thread Thread-4:
Traceback (most recent call last):
 File
"/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/threading.py",
```

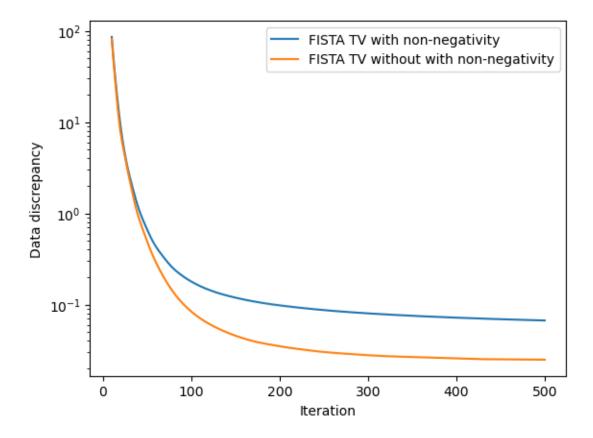
```
line 1075, in _bootstrap_inner
    self.run()
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/tqdm/_monitor.py", line 84, in run
    instance.refresh(nolock=True)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/tqdm/std.py", line 1347, in refresh
    self.display()
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/tqdm/notebook.py", line 171, in display
   rtext.value = right
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/traitlets/traitlets.py", line 716, in __set__
    self.set(obj, value)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/traitlets/traitlets.py", line 706, in set
    obj._notify_trait(self.name, old_value, new_value)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/traitlets/traitlets.py", line 1513, in _notify_trait
    self.notify change(
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipywidgets/widgets/widget.py", line 700, in notify_change
    self.send_state(key=name)
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipywidgets/widgets/widget.py", line 586, in send_state
    self._send(msg, buffers=buffers)
  File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipywidgets/widgets/widget.py", line 825, in _send
    self.comm.send(data=msg, buffers=buffers)
  File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/comm/base_comm.py", line 144, in send
    self.publish_msg(
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipykernel/comm/comm.py", line 42, in publish msg
    parent=self.kernel.get_parent(),
 File "/home/bgb37495/miniconda3/envs/cil_test_demos/lib/python3.12/site-
packages/ipykernel/kernelbase.py", line 783, in get_parent
    return self._shell_parent.get()
```

LookupError: <ContextVar name='shell\_parent' at 0x7f62b66b67a0>





[9]: <matplotlib.legend.Legend at 0x7f607a4e7da0>



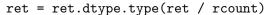
We see that the without the non-negativity, the reconstruction overfits to the noisy absorption data.

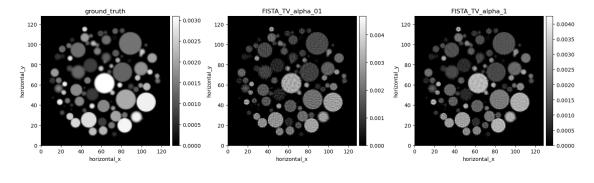
# 1.7 Calculating a Noise Approximation for Each Iteration (Custom Callback Example)

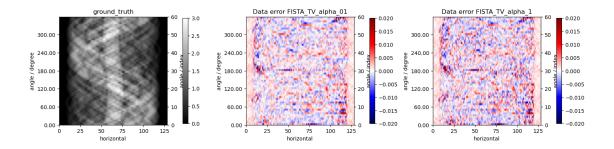
In this example, we define a custom callback that saves the values of a wavelet-based estimator of Gaussian noise standard deviation, provided by skimage, at each iteration. Using TV regularization and a FISTA optimization algorithm with non-negativity, we compare how the noise level in the solution changes for two different regularization parameters: a small value (0.1) and a larger value (1).

```
[10]: import skimage
      class SigmaEstimateCallback(callbacks.Callback):
          Properties
          save_values: list of floats
            The saved sigma calculation, one per iteration
          11 11 11
          def __init__(self):
              self.save_values=[]
          def __call__(self, algorithm):
              self.save_values.append(skimage.restoration.estimate_sigma(algorithm.
       ⇔get_output().as_array()))
      mycallback_FISTA_TV_alpha_01= SigmaEstimateCallback()
      algo1=FISTA(initial=ig.allocate(0), f=F, g=0.1*TotalVariation(lower=0), __
       →update_objective_interval=10)
      algo1.run(500, callbacks=[mycallback_FISTA_TV_alpha_01])
      mycallback_FISTA_TV_alpha_1= SigmaEstimateCallback()
      algo2=FISTA(initial=ig.allocate(0), f=F, g=1*TotalVariation(lower=0), __
       →update_objective_interval=10)
      algo2.run(500, callbacks=[mycallback_FISTA_TV_alpha_1])
      show2D([ground_truth, algo1.get_output(), algo2.get_output()],__
       otitle=['ground_truth', 'FISTA_TV_alpha_01', 'FISTA_TV_alpha_1'], num_cols=3)
```

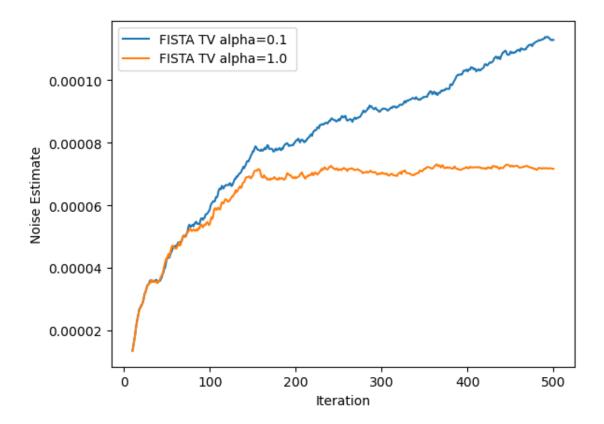
/home/bgb37495/miniconda3/envs/cil\_test\_demos/lib/python3.12/site-packages/numpy/core/fromnumeric.py:3504: RuntimeWarning: Mean of empty slice. return \_methods.\_mean(a, axis=axis, dtype=dtype, /home/bgb37495/miniconda3/envs/cil\_test\_demos/lib/python3.12/site-packages/numpy/core/\_methods.py:129: RuntimeWarning: invalid value encountered in divide







[10]: <matplotlib.legend.Legend at 0x7f60401ff9e0>



We see with a larger regularisation parameter, the resulting image is less noisy.

# 1.8 Image metric callbacks (complex custom callback example)

We now move on to some more complex callbacks. In this callback, a dictionary of metrics are calculated and printed each print\_interval. For more information on the metrics in CIL see the documentation: https://tomographicimaging.github.io/CIL/nightly/utilities/#module-cil.utilities.quality\_measures.

```
print_interval: positive integer
      The results are calculated and printed every `print_interval` number of \Box
\hookrightarrow iterations
  HHHH
  def __init__(self, reference_image, metrics_dict, print_interval=1):
      self.reference_image = reference_image
      self.metrics_dict = metrics_dict
      self.computed_metrics = []
      self.print_interval=print_interval
      super(MetricsDiagnostics, self).__init__()
  def __call__(self, algo):
      for metric_name, metric_func in self.metrics_dict.items():
           if not hasattr(algo, metric_name):
               setattr(algo, metric_name, [])
          metric_list = getattr(algo, metric_name)
          metric_value = metric_func(self.reference_image, algo.get_output())
          metric_list.append(metric_value)
          self.computed_metrics.append(metric_value)
      if algo.iteration == 0:
          print (self.callback_header())
      print(self.callback_iteration())
  def callback_header(self):
      return " ".join("{:>20}".format(metric_name) for metric_name in self.
→metrics_dict.keys())
  def callback_iteration(self):
      if isinstance(self.computed_metrics, list):
           # Handle list of metrics
          return " ".join("{:>20.5e}".format(metric) for metric in self.
⇔computed_metrics[-len(self.metrics_dict):])
```

```
else:

# Handle single metric

return "{:>20.5e}".format(self.computed_metrics)

from cil.utilities.quality_measures import mae, psnr, mse

metric_callback= MetricsDiagnostics(ground_truth, {'MSE':mse, 'MAE':mae, 'PSNR':

psnr})

algo=FISTA(initial=ig.allocate(0), f=F, g=G, update_objective_interval=10)

algo.run(100, callbacks=[metric_callback])
```

MSE	MAE	PSNR
1.07888e-06	5.48145e-04	9.48530e+00
5.85316e-07	6.22034e-04	1.21411e+01
5.05844e-07	5.72563e-04	1.27749e+01
4.31374e-07	5.19819e-04	1.34665e+01
3.64704e-07	4.67054e-04	1.41956e+01
3.06416e-07	4.16492e-04	1.49519e+01
2.56388e-07	3.70092e-04	1.57261e+01
2.14156e-07	3.28810e-04	1.65077e+01
1.78987e-07	2.92725e-04	1.72868e+01
1.50022e-07	2.60981e-04	1.80535e+01
1.26383e-07	2.33361e-04	1.87981e+01
1.07187e-07	2.09652e-04	1.95136e+01
9.16141e-08	1.89309e-04	2.01954e+01
7.89449e-08	1.72049e-04	2.08418e+01
6.85910e-08	1.57283e-04	2.14524e+01
6.00884e-08	1.44610e-04	2.20271e+01
5.30737e-08	1.33746e-04	2.25662e+01
4.72670e-08	1.24414e-04	2.30695e+01
4.24393e-08	1.16364e-04	2.35374e+01
3.84176e-08	1.09415e-04	2.39697e+01
3.50543e-08	1.03451e-04	2.43676e+01
3.22300e-08	9.82989e-05	2.47324e+01
2.98493e-08	9.39156e-05	2.50657e+01
2.78304e-08	9.01341e-05	2.53698e+01
2.61075e-08	8.68680e-05	2.56474e+01
2.46249e-08	8.40164e-05	2.59013e+01
2.33423e-08	8.14809e-05	2.61336e+01
2.22266e-08	7.92211e-05	2.63463e+01
2.12462e-08	7.72332e-05	2.65422e+01
2.03792e-08	7.54337e-05	2.67232e+01
1.96080e-08	7.38151e-05	2.68907e+01
1.89173e-08	7.23520e-05	2.70464e+01
1.82934e-08	7.10307e-05	2.71921e+01

1.77264e-08	6.98001e-05	2.73288e+01
1.72101e-08	6.86725e-05	2.74572e+01
1.67384e-08	6.76756e-05	2.75779e+01
1.63068e-08	6.67997e-05	2.76913e+01
1.59109e-08	6.59966e-05	2.77981e+01
1.55498e-08	6.52429e-05	2.78978e+01
1.52208e-08	6.45565e-05	2.79907e+01
1.49199e-08	6.39012e-05	2.80774e+01
1.46448e-08	6.32730e-05	2.81582e+01
1.43935e-08	6.26837e-05	2.82334e+01
1.41640e-08	6.21308e-05	2.83032e+01
1.39533e-08	6.16084e-05	2.83683e+01
1.37602e-08	6.11234e-05	2.84288e+01
1.35827e-08	6.06740e-05	2.84852e+01
1.34200e-08	6.02613e-05	2.85375e+01
1.32710e-08	5.98831e-05	2.85860e+01
1.31342e-08	5.95366e-05	2.86310e+01
1.30086e-08	5.92132e-05	2.86727e+01
1.28936e-08	5.89066e-05	2.87113e+01
1.27882e-08	5.86154e-05	2.87469e+01
1.26929e-08	5.83402e-05	2.87794e+01
1.26069e-08	5.81077e-05	2.88090e+01
1.25294e-08	5.79025e-05	2.88357e+01
1.24593e-08	5.77139e-05	2.88601e+01
1.23964e-08	5.75408e-05	2.88821e+01
1.23400e-08	5.73717e-05	2.89019e+01
1.22899e-08	5.72179e-05	2.89196e+01
1.22457e-08	5.70800e-05	2.89352e+01
1.22065e-08	5.69476e-05	2.89491e+01
1.21716e-08	5.68220e-05	2.89616e+01
1.21399e-08	5.67079e-05	2.89729e+01
1.21121e-08	5.66082e-05	2.89828e+01
1.20881e-08	5.65168e-05	2.89914e+01
1.20672e-08	5.64386e-05	2.89990e+01
1.20490e-08	5.63735e-05	2.90055e+01
1.20338e-08	5.63197e-05	2.90110e+01
1.20213e-08	5.62742e-05	2.90155e+01
1.20117e-08	5.62335e-05	2.90190e+01
1.20049e-08	5.61994e-05	2.90214e+01
1.20006e-08	5.61720e-05	2.90230e+01
1.19991e-08	5.61517e-05	2.90236e+01
1.19998e-08	5.61385e-05	2.90233e+01
1.20029e-08	5.61309e-05	2.90222e+01
1.20088e-08	5.61240e-05	2.90201e+01

1.20170e-08	5.61242e-05	2.90171e+01
1.20275e-08	5.61325e-05	2.90133e+01
1.20408e-08	5.61499e-05	2.90085e+01
1.20565e-08	5.61750e-05	2.90028e+01
1.20747e-08	5.62071e-05	2.89963e+01
1.20954e-08	5.62404e-05	2.89888e+01
1.21182e-08	5.62744e-05	2.89806e+01
1.21432e-08	5.63137e-05	2.89717e+01
1.21702e-08	5.63569e-05	2.89620e+01
1.21990e-08	5.64026e-05	2.89518e+01
1.22294e-08	5.64532e-05	2.89410e+01
1.22611e-08	5.65052e-05	2.89297e+01
1.22934e-08	5.65577e-05	2.89183e+01
1.23272e-08	5.66136e-05	2.89064e+01
1.23621e-08	5.66716e-05	2.88941e+01
1.23983e-08	5.67352e-05	2.88814e+01
1.24357e-08	5.68040e-05	2.88683e+01
1.24743e-08	5.68758e-05	2.88549e+01
1.25140e-08	5.69482e-05	2.88411e+01
1.25548e-08	5.70229e-05	2.88269e+01
1.25965e-08	5.71005e-05	2.88125e+01
1.26388e-08	5.71802e-05	2.87980e+01
1.26821e-08	5.72615e-05	2.87831e+01
1.27264e-08	5.73452e-05	2.87680e+01

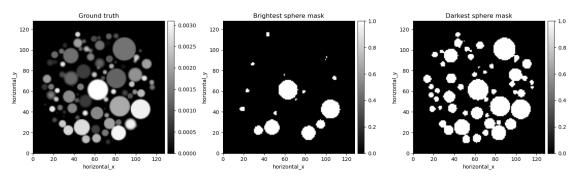
# 1.9 Image metric callbacks with region of interest (complex custom callback example)

For another complex example. This callback takes a reference image, region of interest mask and a dictionary of metrics and a dictionary of statistics to evaluate at each iteration. We define some regions of interest and then can plot the metrics for the different regions of interest.

```
metrics_dict : dictionary of lambda functions f(x,y) mapping
     two 1-dimensional numpy arrays x and y to a scalar value or a
     numpy.ndarray. Optional - default is None.
       x and y can be the voxel values of the whole images or the values of
       voxels in a ROI such that the metric can be computed on the whole
       images and optionally in the ROIs separately.
       E.g. f(x,y) could be MSE(x,y), PSNR(x,y), MAE(x,y)
   statistics\_dict: dictionary of lambda functions f(x) mapping a
     1-dimensional numpy array x to a scalar value or a numpy.ndarray.
\neg Optional - default is None
       E.g. mean(x), std deviation(x) that calculate global and / or
      ROI mean and standard deviations.
  Properties
  metric_store: dictionary
     The keys of the dictionary are "global_"+metric keys or roi key +'_'+_{\sqcup}
\hookrightarrowmetric key. Stored under these keys is a list of calculations of the metric\sqcup
⇔ for that roi, one per iteration.
  stat_store: dictionary
     The keys of the dictionary are "global_"+statistic keys or roi key +'_'+_{\square}
⇔statistic key. Stored under these keys is a list of calculations of the ⊔
⇒statistic for that roi, one per iteration.
     11 11 11
  def __init__(self, reference_image,
                roi_mask_dict=None,
                metrics_dict=None,
                statistics_dict=None
                ):
       # the reference image
       self.reference_image = reference_image
      self.roi_indices_dict = {}
      self.roi store = []
      self.roi_mask_dict = roi_mask_dict
      self.metrics dict = metrics dict
       self.metrics_store = {}
       for key in self.metrics_dict:
           self.metrics_store['global_'+key] = []
```

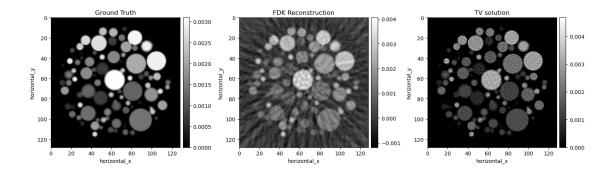
```
if roi_mask_dict is not None:
            for roi_name in roi_mask_dict:
                self.metrics_store[roi_name+'_'+key] = []
    self.statistics_dict = statistics_dict
    self.stat_store = {}
    for key in self.statistics dict:
        self.stat_store['global_'+key] = []
        if roi mask dict is not None:
            for roi_name in roi_mask_dict:
                self.stat_store[roi_name+'_'+key] = []
def __call__(self, algorithm):
    if self.metrics_dict is not None:
        for metric_name, metric in self.metrics_dict.items():
            ans = metric(self.reference_image, algorithm.x)
            self.metrics_store['global_'+metric_name].append(ans)
            for roi_name, roi in self.roi_mask_dict.items():
                ans = metric(self.reference_image, algorithm.x, mask=roi)
                self.metrics_store[roi_name+'_'+metric_name].append(ans)
    if self.statistics_dict is not None:
        for statistic name, stat in self.statistics dict.items():
            ans = stat(algorithm.x.array, np._NoValue)
            self.stat_store['global_'+statistic_name].append(ans)
            for roi_name, roi in self.roi_mask_dict.items():
                ans = stat(algorithm.x.array, roi.array.astype('bool'))
                self.stat_store[roi_name+'_'+statistic_name].append(ans)
```

For this data we take one region of interest to be the brightest spheres, another region of interest the darkest spheres. We define the masks for this in the next cell:



### [13]: <cil.utilities.display.show2D at 0x7f604021e090>

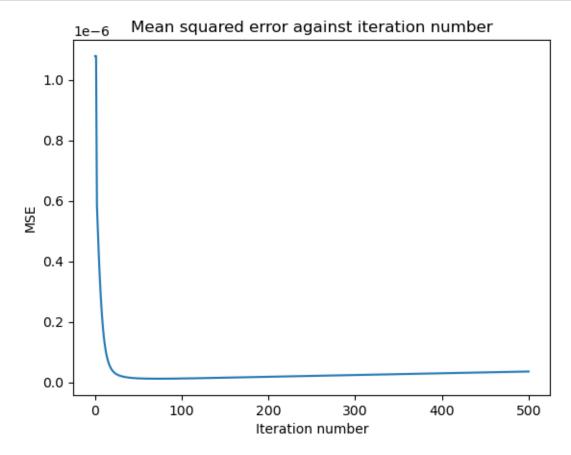
```
[15]: algo=FISTA(initial=ig.allocate(0), f=F, g=G, update_objective_interval=10)
algo.run(500, callbacks=[img_qual_callback])
show2D([ground_truth, recon, algo.solution], title = ['Ground Truth', 'FDK_\subseteq
Acconstruction', 'TV solution'], origin = 'upper', num_cols = 3)
```



[15]: <cil.utilities.display.show2D at 0x7f60403c3d40>

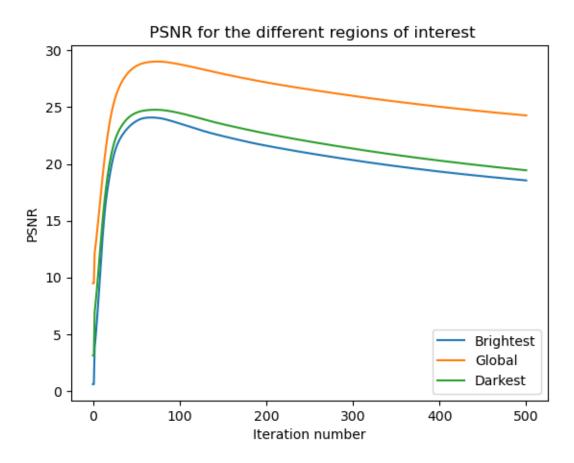
We see that over the whole image, the mean squared error initially decreases, but then eventually starts to overfit to the noise in the data.

```
[16]: plt.plot(range(501), img_qual_callback.metrics_store['global_MSE'])
    plt.ylabel('MSE')
    plt.xlabel('Iteration number')
    plt.title('Mean squared error against iteration number')
    plt.show()
```



Similarly PSNR for the whole image, and for each of the regions of interest, initially increases to a peak at about 60-70 iterations then begins to decrease again. We can see that the PSNR over the whole image is greater than the regions of interest, suggesting that the algorithm is fitting the background well but not reconstructing the spheres as well.

[17]: <matplotlib.legend.Legend at 0x7f604021fda0>



[]:[