



Advanced

Kitware Course Week, Clifton Park, NY

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Outline

- Introduction
- Custom Operators
- User Input for Operators
- External Pipelines
- More

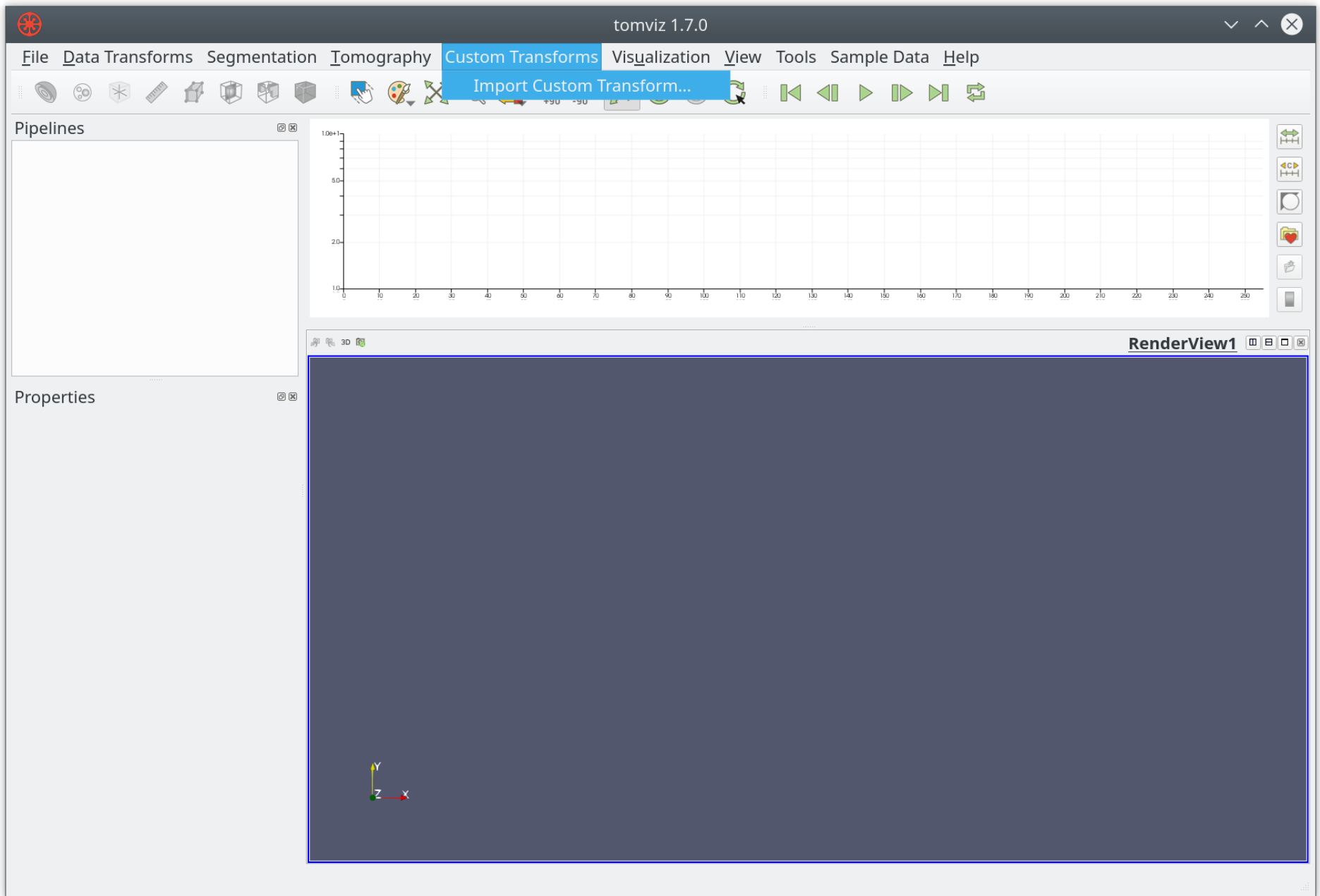
Introduction

- Tomviz can be extended at runtime
 - Custom operators and file formats
 - Primarily using Python scripts/JSON
- The processing pipeline is self-contained
 - Run in Docker for reproducibility
 - Soon external Python for CUDA, etc

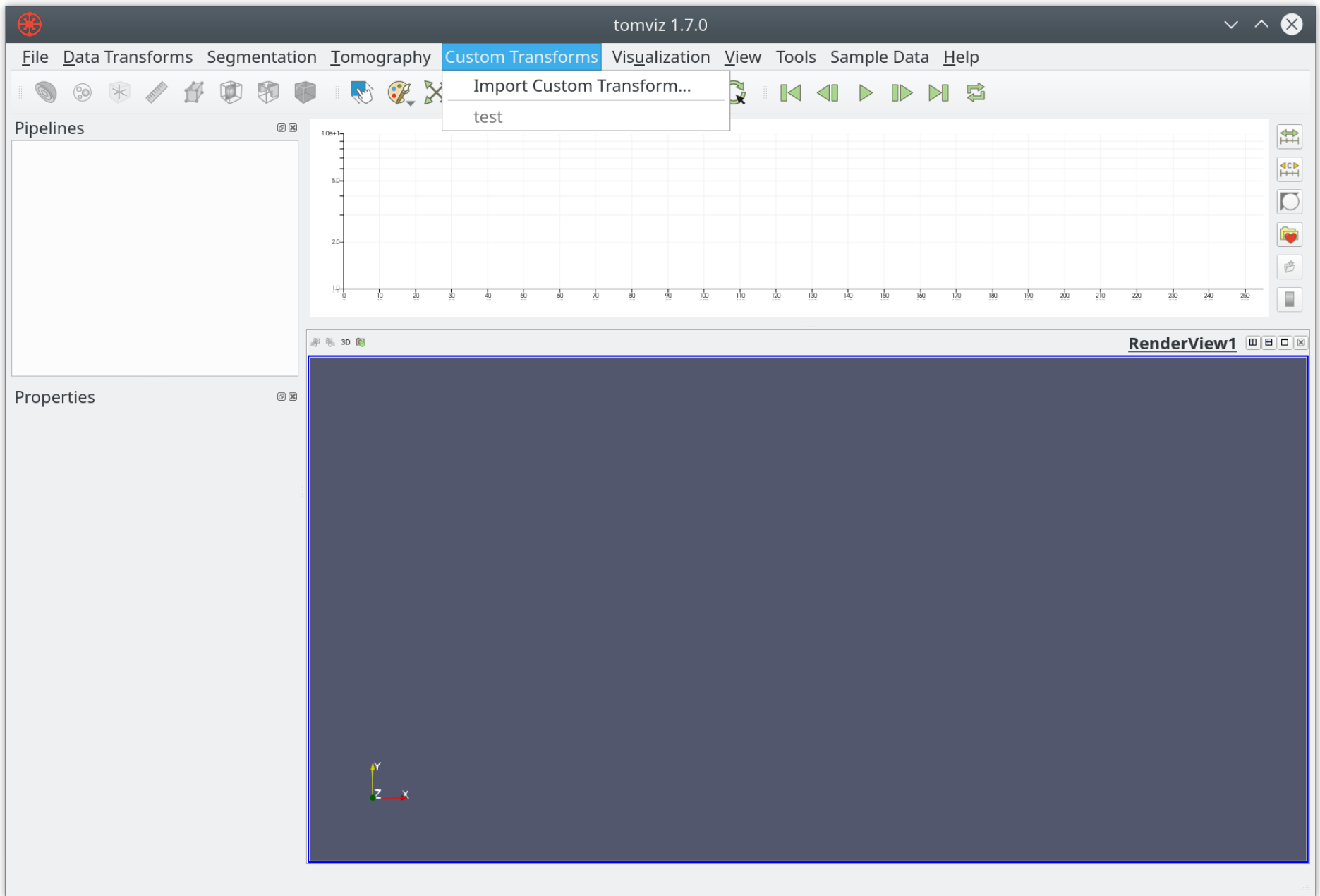
Custom Operators

- Initially just a single entry
 - Import custom transform
- Use import, or copy files over
 - Looks in ~/tomviz and ~/.tomviz
 - Adds suitable Python/JSON files

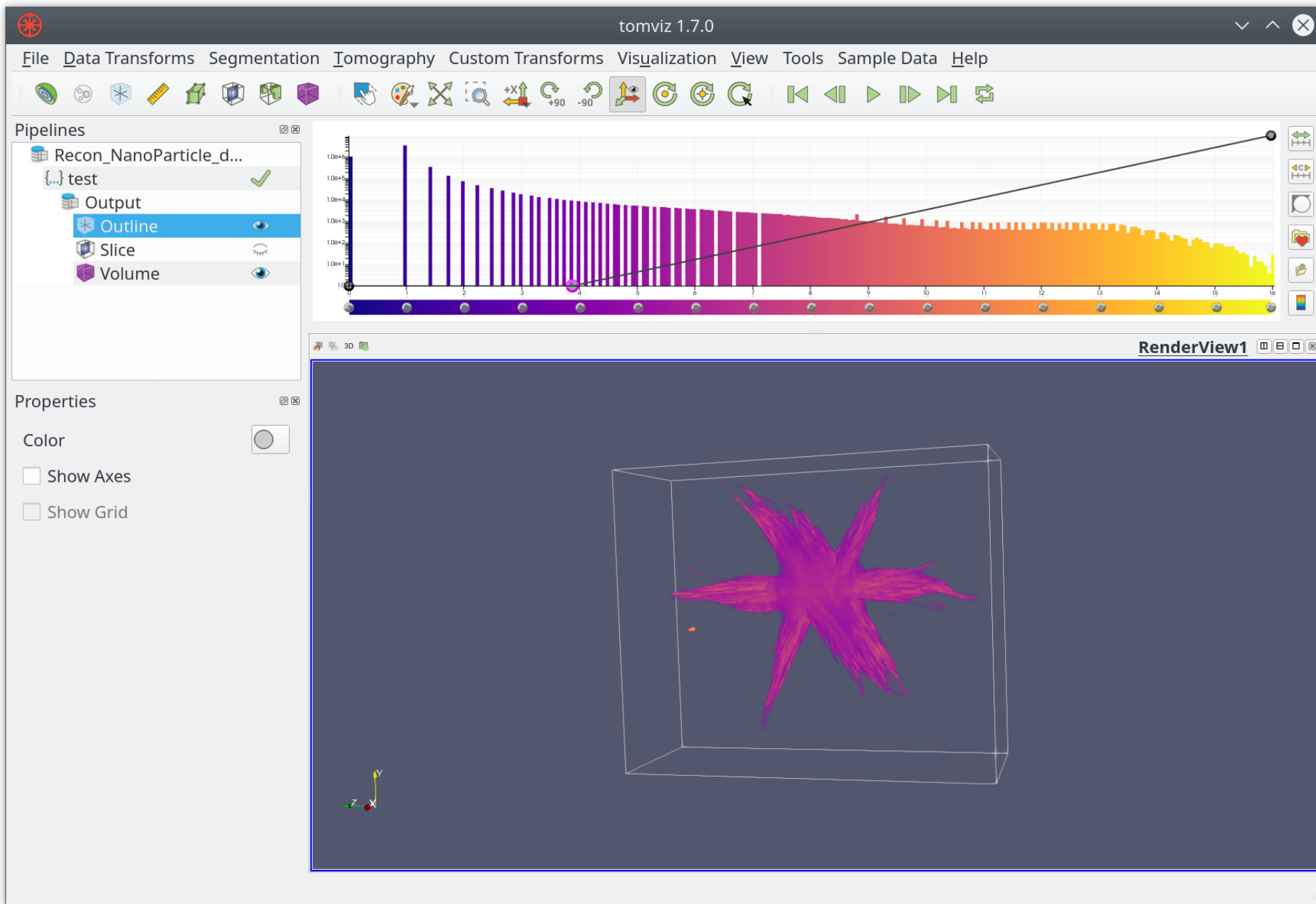
Custom Transforms



Custom Transforms—Test



Custom Transforms—Test



Anatomy of an Operator

```
def transform_scalars(dataset):  
  
    from tomviz import utils  
    import numpy as np  
  
    # Get the current volume as a NumPy array.  
    array = utils.get_array(dataset)  
  
    # Operate on your data, here we square root it.  
    result = np.sqrt(array)  
  
    # Set the transformed data, displayed in Tomviz.  
    utils.set_array(dataset, result)
```

Classifier Operators—Outline

```
import tomviz.operators

class SquareRootOperator(tomviz.operators.CancelableOperator):

    def transform_scalars(self, dataset):
        from tomviz import utils
        import numpy as np
        self.progress.maximum = 10
        scalars = utils.get_scalars(dataset)
        # Process dataset in chunks, user can cancel.
        result = np.float32(scalars)
        step = 0
        # Processing loop goes here...
        utils.set_scalars(dataset, result)
```

Classifier Operators—Loop

```
def transform_scalars(self, dataset):
    # ...other stuff from previous slide
    scalars = utils.get_scalars(dataset)
    # Process dataset in chunks, user can cancel.
    result = np.float32(scalars)
    for chunk in np.array_split(result, 10):
        if self.canceled:
            return
        np.sqrt(chunk, chunk)
        step += 1
        self.progress.value = step

    utils.set_scalars(dataset, result)
```

Classifier Operators—Complete

```
import tomviz.operators
CHUNKS = 10

class SquareRootOperator(tomviz.operators.CancelableOperator):

    def transform_scalars(self, dataset):
        from tomviz import utils
        import numpy as np
        self.progress.maximum = CHUNKS
        scalars = utils.get_scalars(dataset)
        # Process dataset in chunks, user can cancel.
        result = np.float32(scalars)
        step = 0
        for chunk in np.array_split(result, CHUNKS):
            if self.canceled:
                return
```

User Input for Operators

- You have a custom operator now
 - How can we change the chunk size?
 - Edit the code?
- Generating user interface
 - Inject input into operators

JSON for Increased Control

```
{
  "name": "Fancy Square Root",
  "label": "Classy Square Root",
  "description": "This is the fanciest square root operator, it does",
  "parameters": [
    {
      "name": "number_of_chunks",
      "label": "Number of Chunks",
      "type": "int",
      "default": 10,
      "minimum": 1,
      "maximum": 1000
    }
  ]
}
```

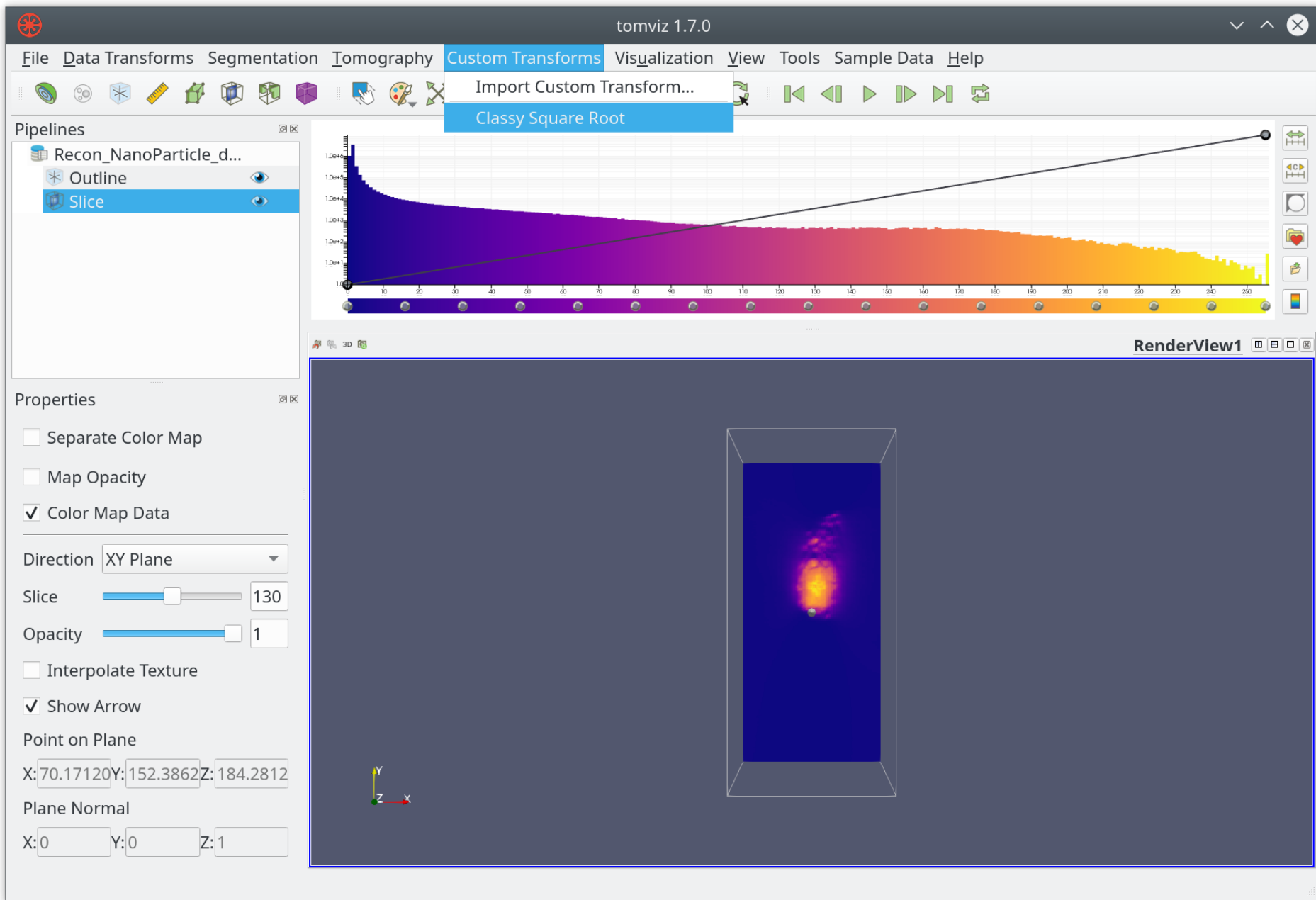

Classifier Operators—Complete

```
import tomviz.operators

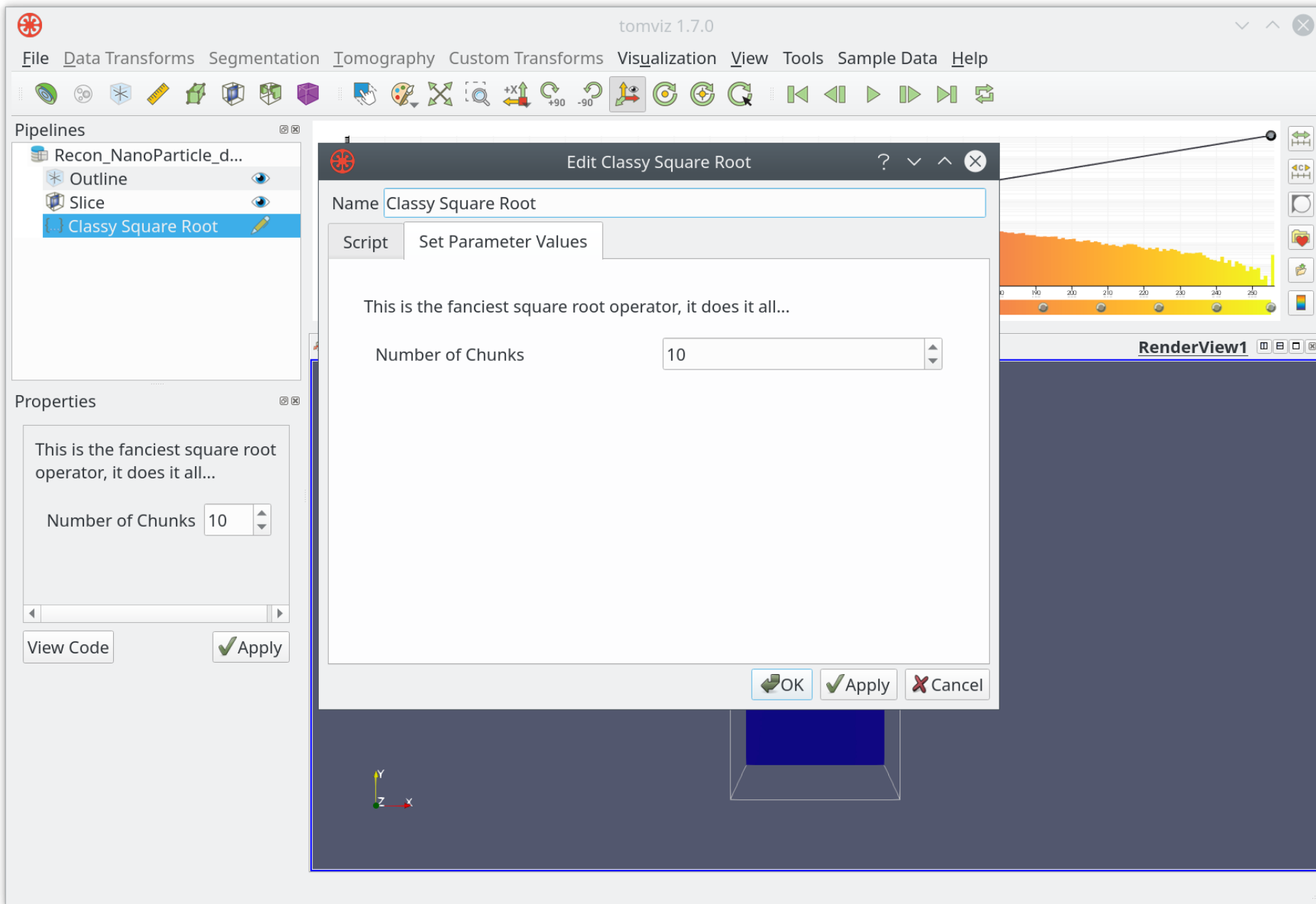
class SquareRootOperator(tomviz.operators.CancelableOperator):

    def transform_scalars(self, dataset, number_of_chunks = 10):
        from tomviz import utils
        import numpy as np
        self.progress.maximum = number_of_chunks
        scalars = utils.get_scalars(dataset)
        # Process dataset in chunks, user can cancel.
        result = np.float32(scalars)
        step = 0
        for chunk in np.array_split(result, number_of_chunks):
            if self.canceled:
                return
            # result[chunk_start:chunk_end] = np.sqrt(chunk)
```

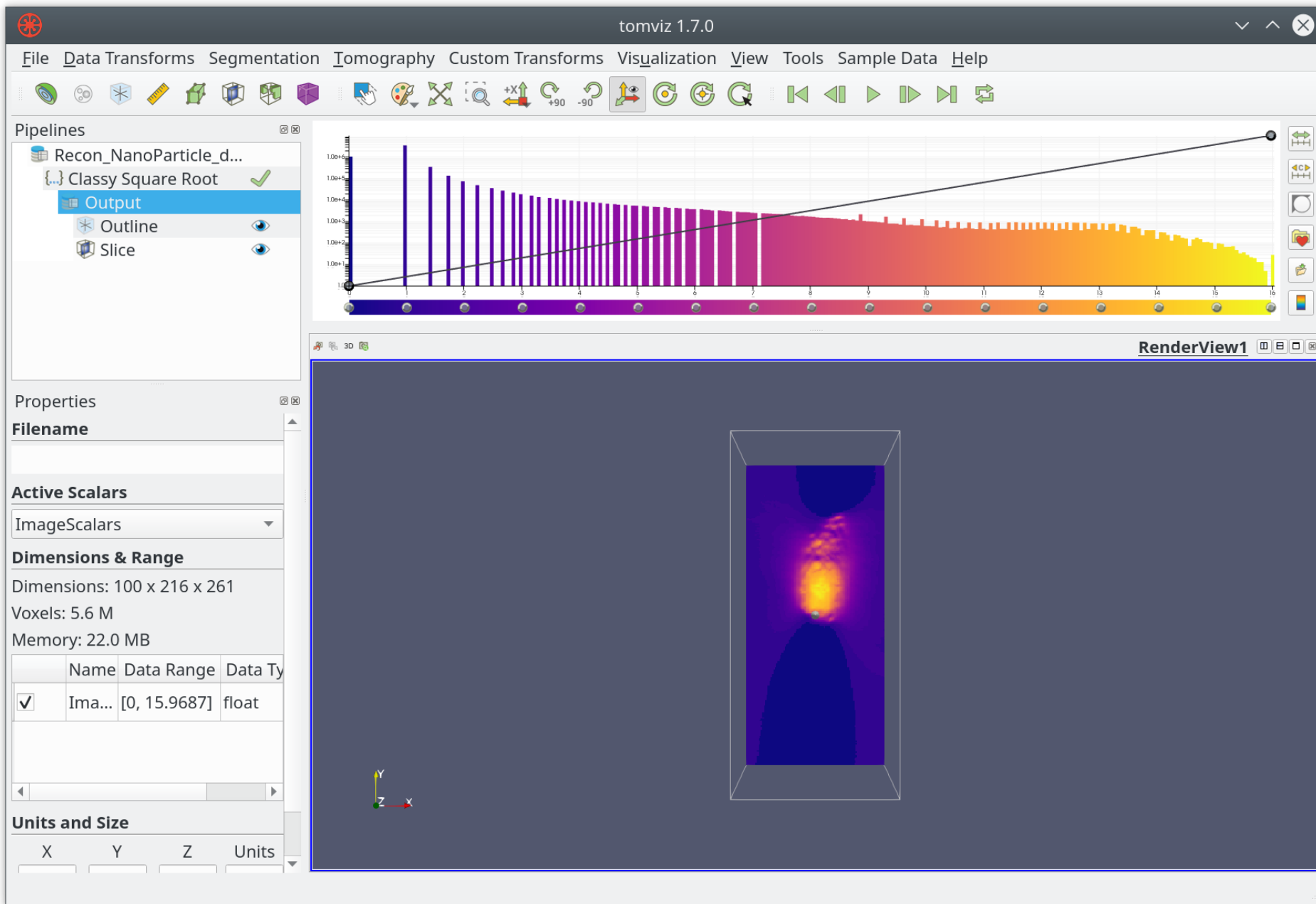
Custom Transforms—JSON Menu



Custom Transforms—User Input



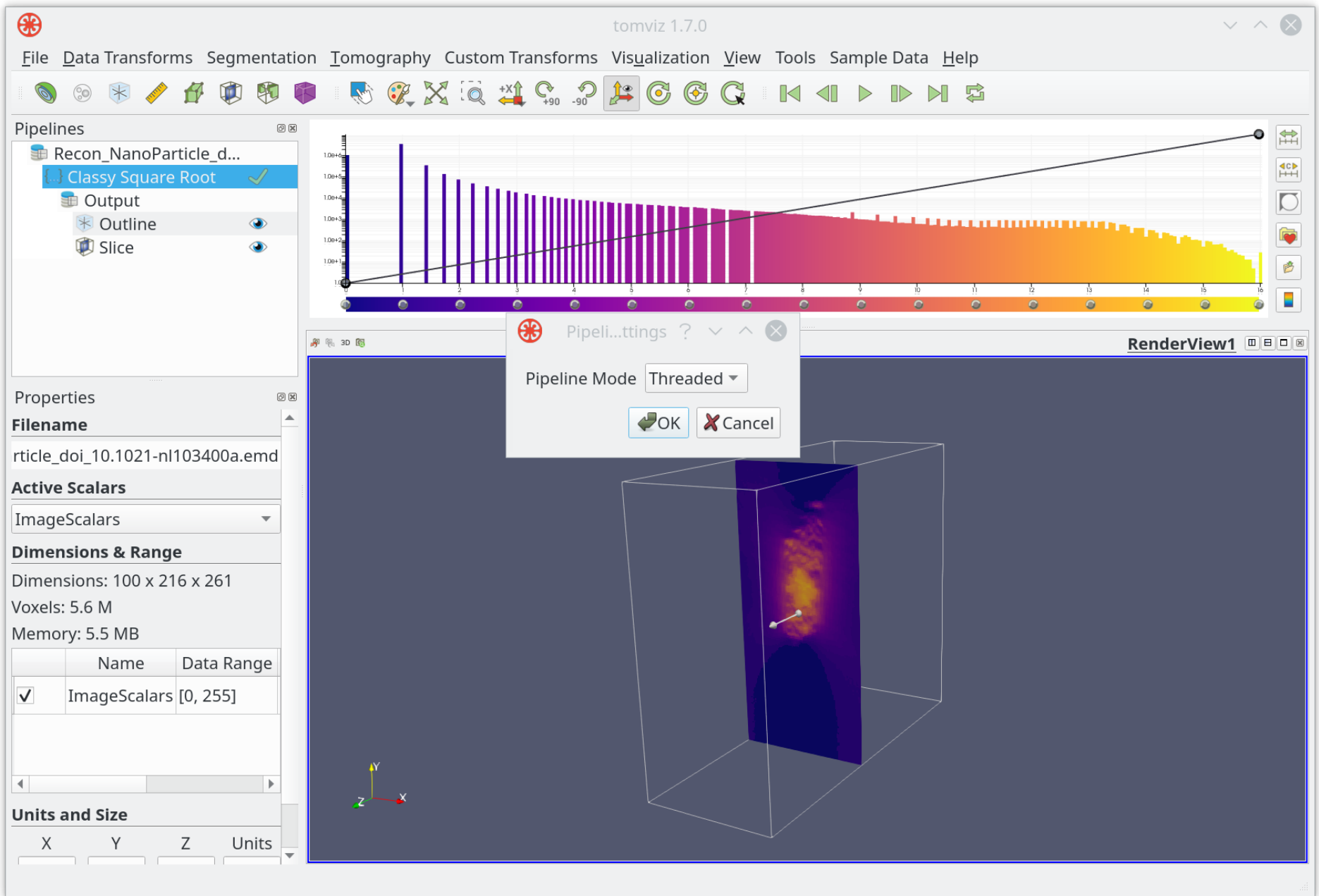
Custom Transforms—Result



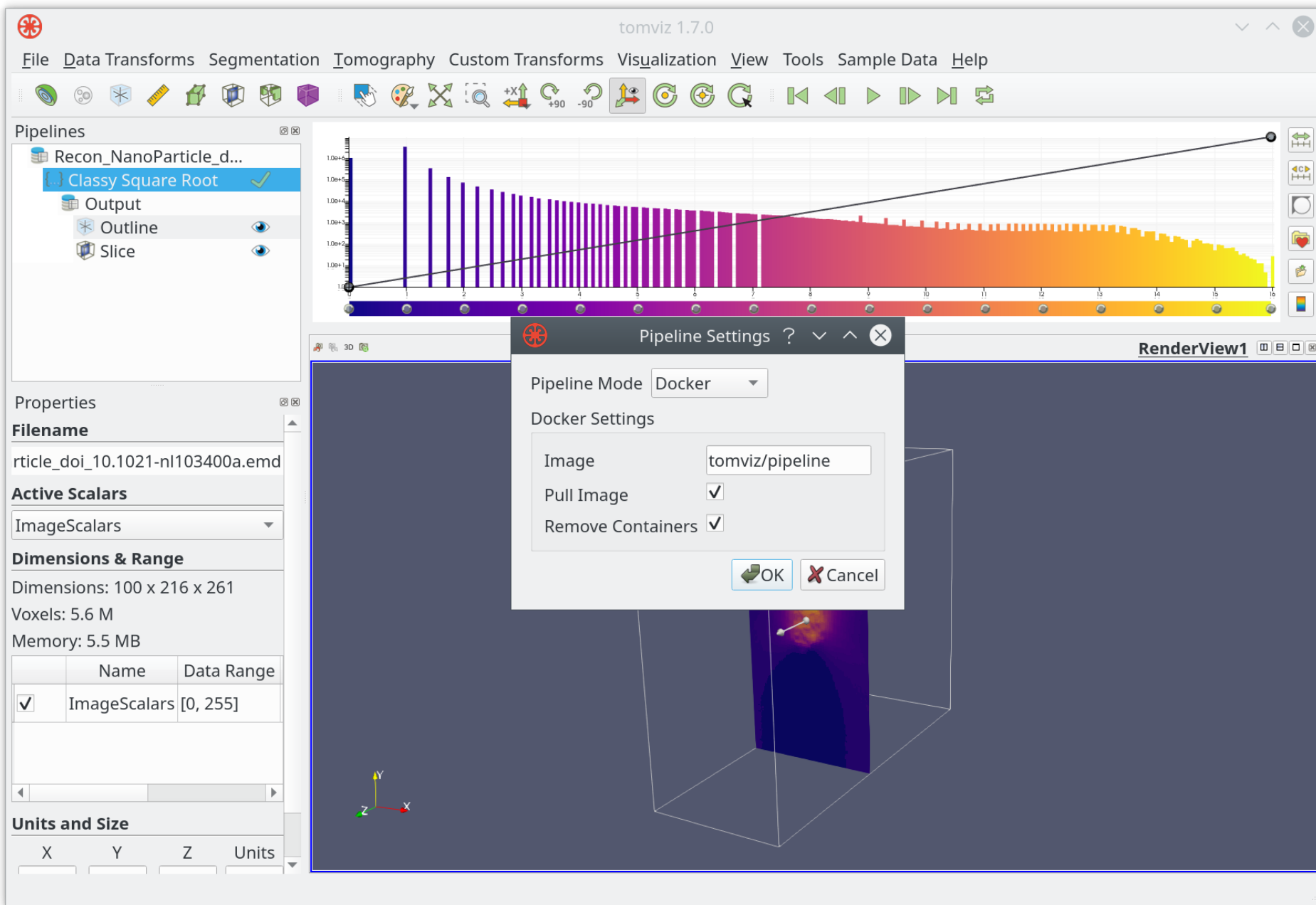
External Pipelines

- Tomviz usually runs pipelines in a background thread
 - This can be changed in Pipeline Settings
 - Threaded (default) or Docker
- Working on making this more flexible in the future

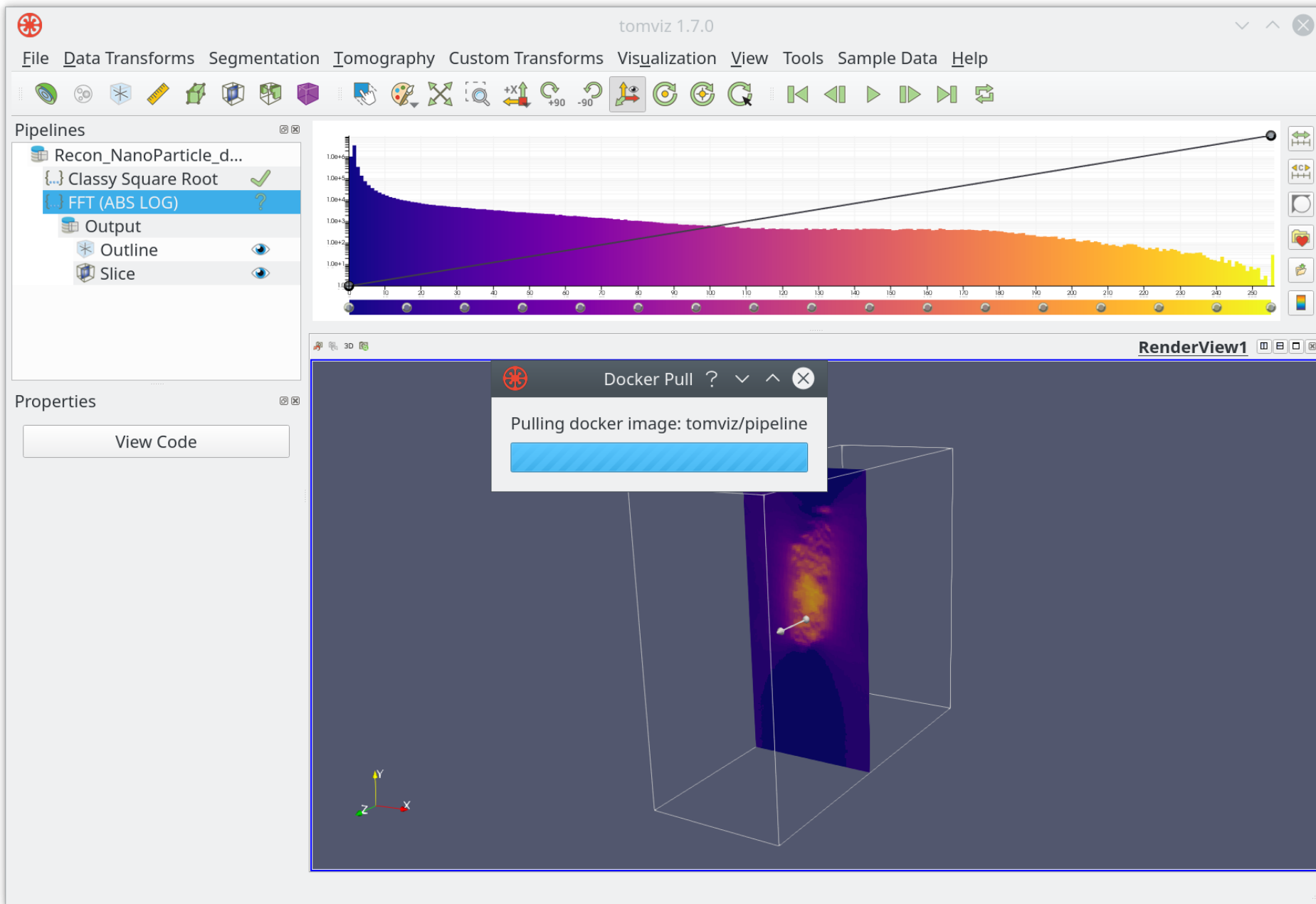
Pipeline Settings



Pipeline Settings—Docker



Pulling Docker Image



Running Pipelines

- Create a virtual environment, install

```
$ git clone --recursive git://github.com/openchemistry/
$ cd tomviz/tomviz/python
$ mkvirtualenv tomviz
$ pip install -e .
```

- Use Tomviz to build a pipeline
 - Save the state file (JSON)

Running Pipelines

- Run the pipeline in the virtual environment

```
$ tomviz-pipeline -s state.tvsm -d data.emd -o output.emd  
[2019-07-23 14:14:59,647] INFO: Executing 'Invert Data' op  
[2019-07-23 14:14:59,963] INFO: Writing transformed data.
```

- Input must be an EMD file right now
- Operators are executed in sequence
- Output is written to an EMD file

Summary

- Adding custom operators is easy
 - NumPy centric approach
 - Runs in a background thread
 - Options to customize input, etc
- The pipeline can be run externally
 - A few bugs we will fix in the next release
 - Docker out of the box, Python batch possible
- File formats and more are in the works

