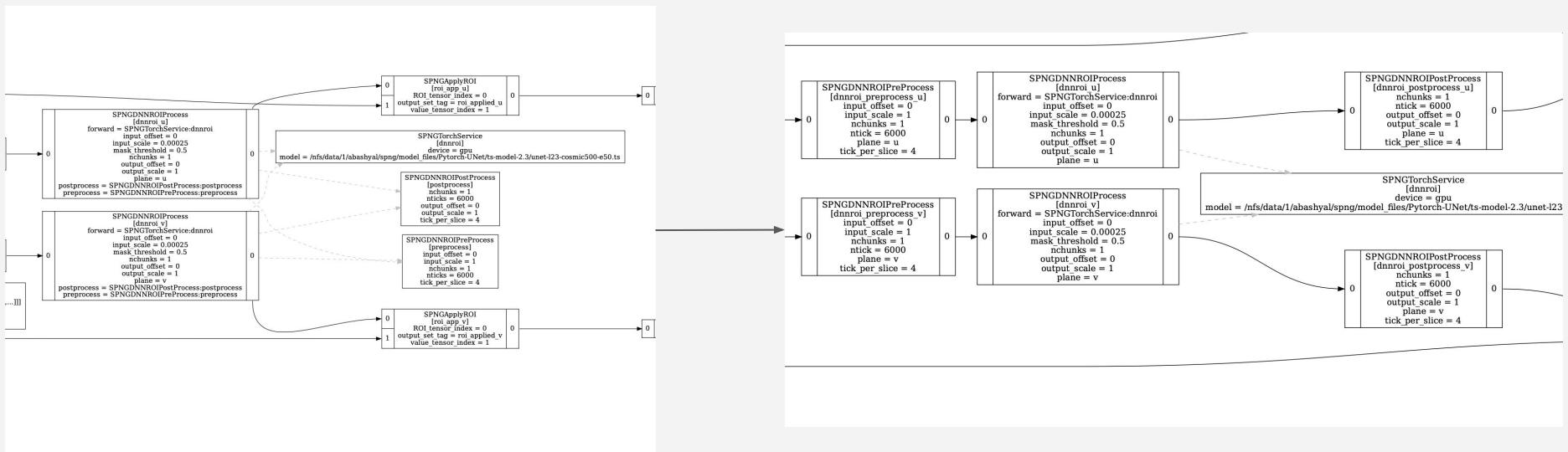


DUNE SPNG Updates

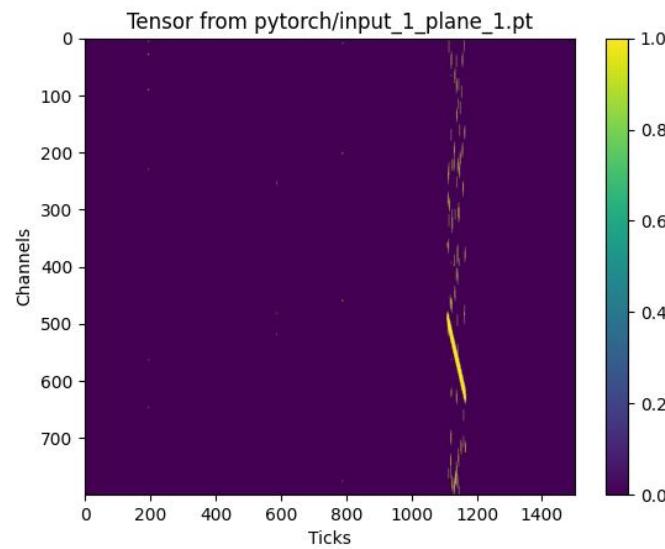
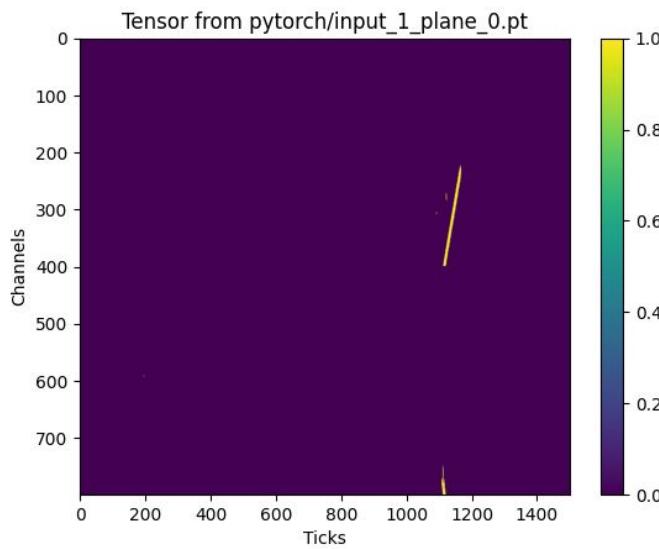
November 14, 2025

From Last Week



Pre and Post Processing of DNN-ROI changed from Service to Nodes

From Last Week

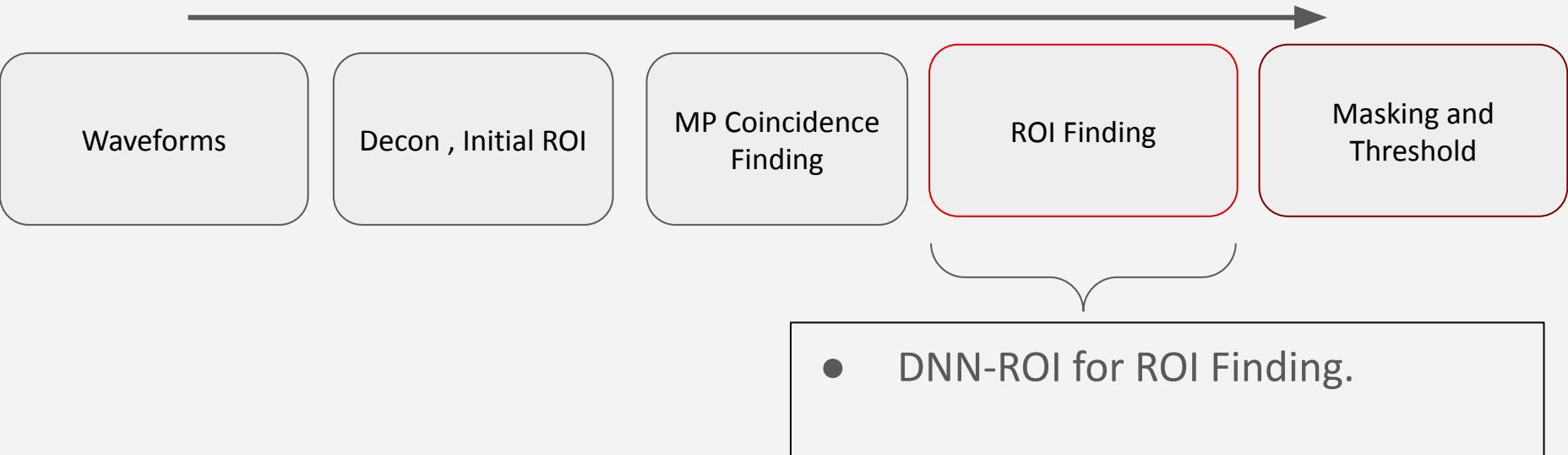


MP2 plot for the V plane (right) from OSP. Seems like this artifact is really there (?)

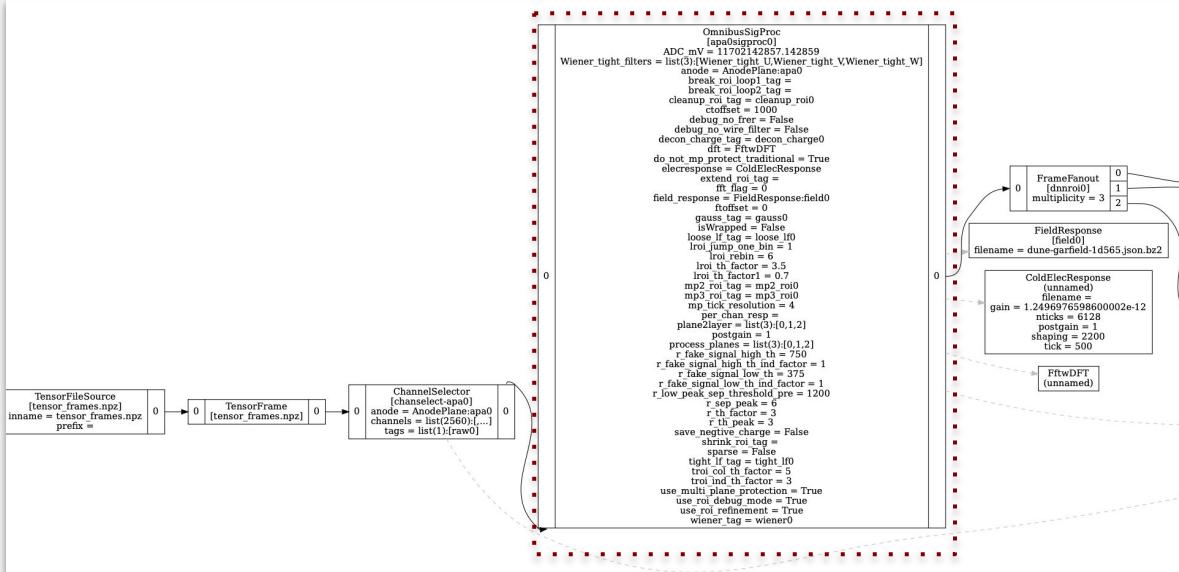
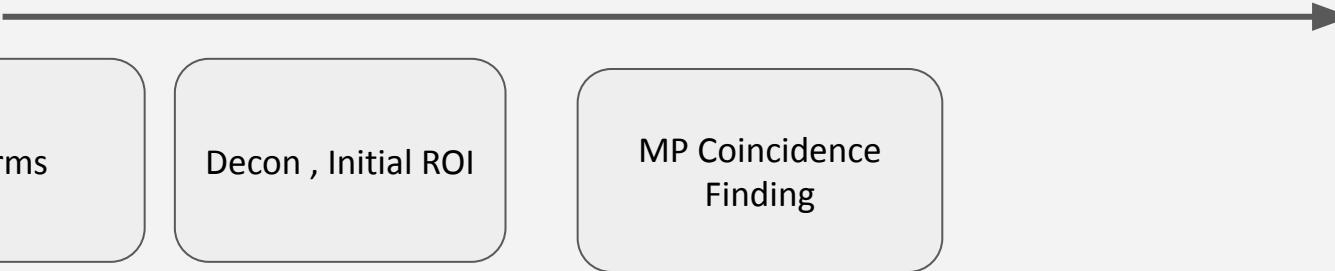
Validating the DNNROI implementation in the SPNG

- Validate the implementation of DNNROI by benchmarking against the OSP (Original Signal Processing) Steps

Signal Processing Step in Rough for illustration

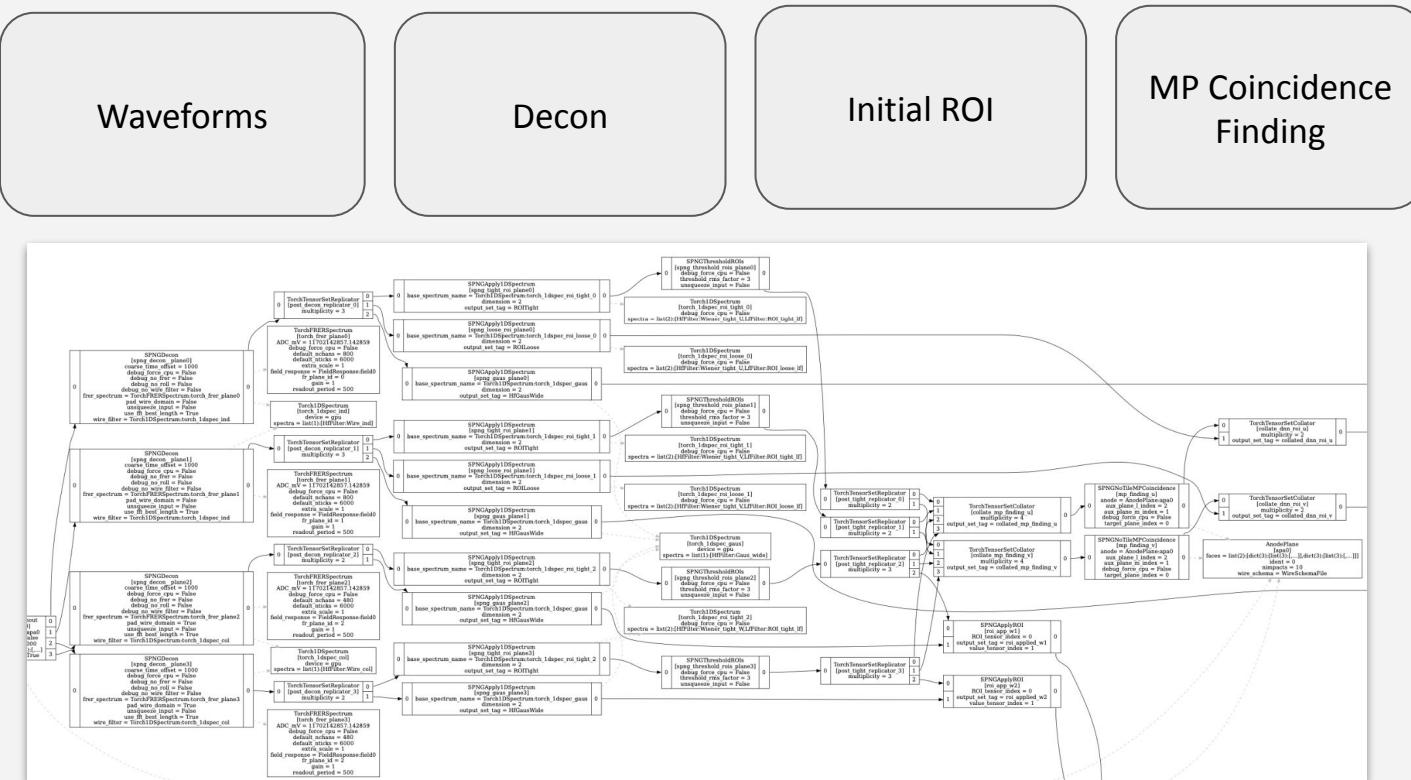


Signal Processing Step OSP



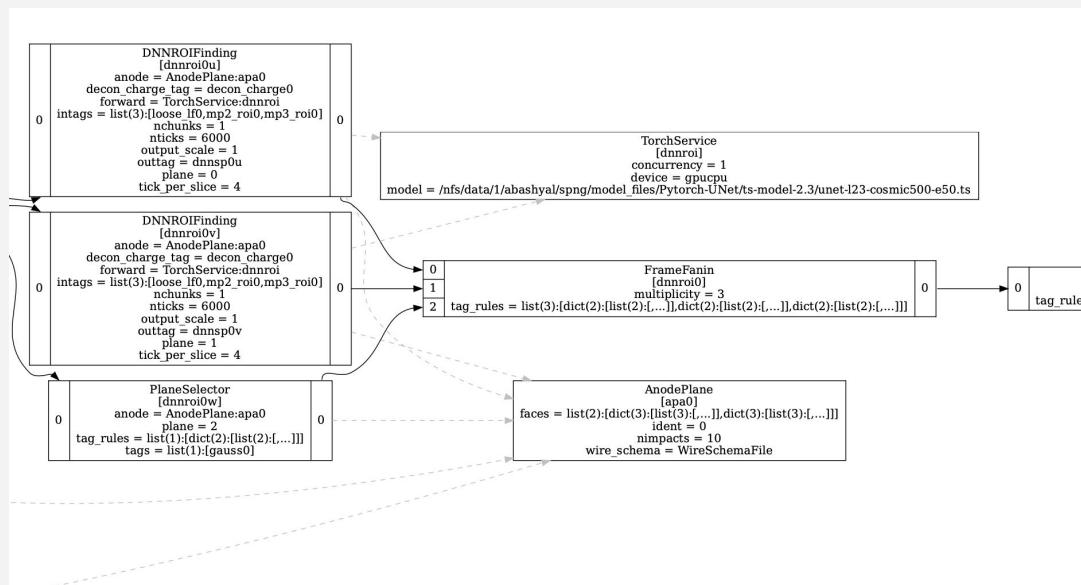
Most of the heavy lifting
done by
OmnibusSigProc.

Signal Processing Step SPNG



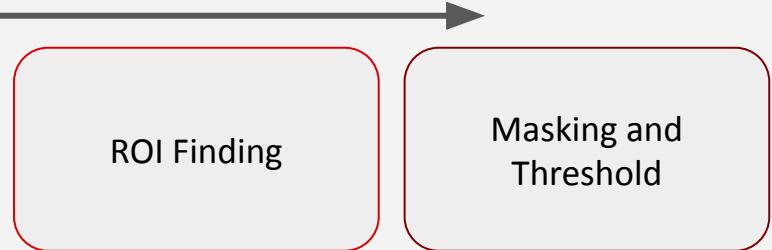
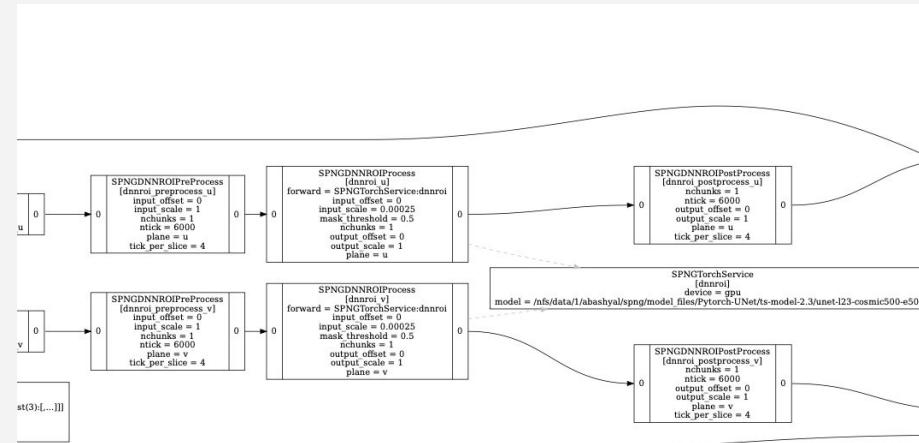
In SPNG, split into
Decon, filters,
MP2/MP3
co-incidence
finding.

(DNN) ROI Finding (OSP)



- **DNNROIFinding**
 - Preprocessing, inference (via TorchService) and Postprocessing)
 - Masking and Threshold

(DNN) ROI Finding (SPNG)



- Pre-Processing [DNNROIPreProcess]
- Inference [DNNROIProcess]
- Post-Processing [DNNROIPostProcess]

Pre-Processing

- Pre-Process Target Tensor [u,v], MP2 and MP3 coincidence tensors
 - Scaling and offsetting (User defined scale and offset factor)
 - Rebin Ticks [6000 ticks → user defined [1500 ticks] (**Downsampling**)
 - Shape consistency → [Batch, Channel, Time]
 - Stack pre-processed tensors (target, mp2,mp3)
 - Consistency in stacking

Processing

- Forward stacked Tensors for inference using TorchService
- Retrieve Back the post inference Tensor (Target tensor only)

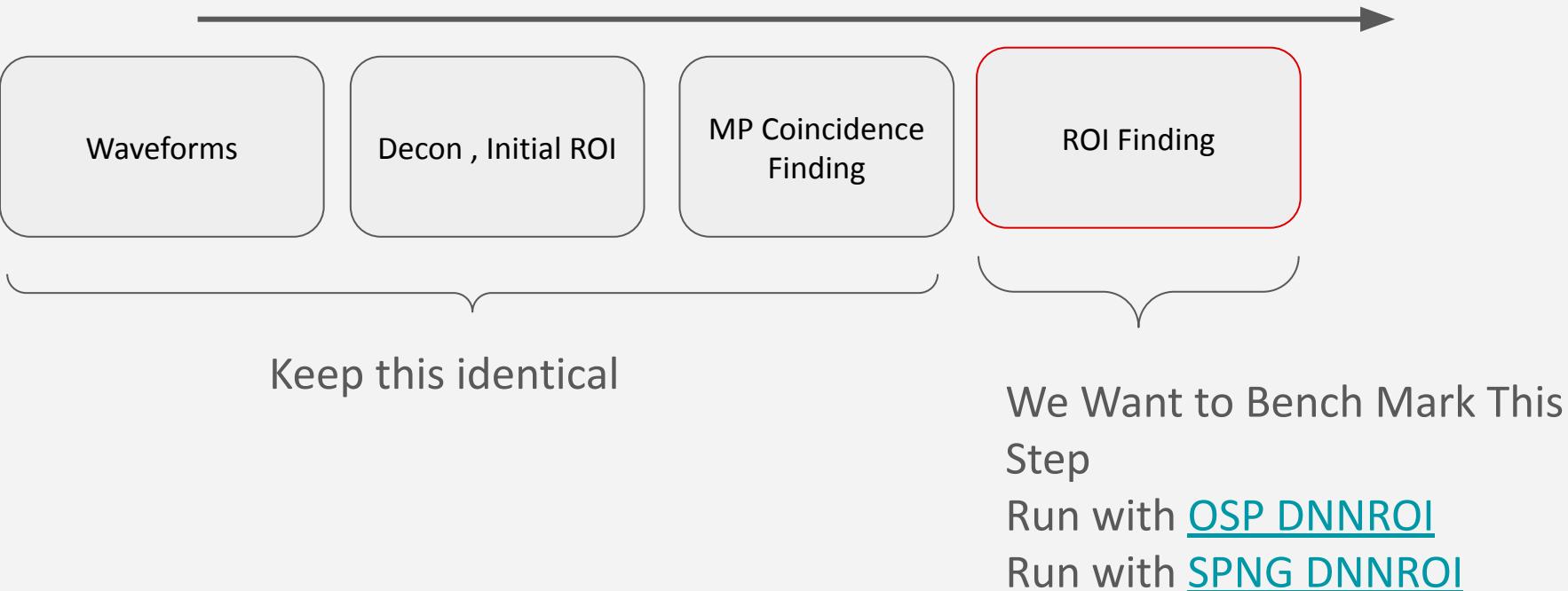
Post Processing

- Upsample [1500 ticks → 6000 ticks] post inference tensor
- Dimension consistency → 3 D tensor
- OSP
 - Masking and Thresholding
- SPNG
 - Marking and Threshold in another stage

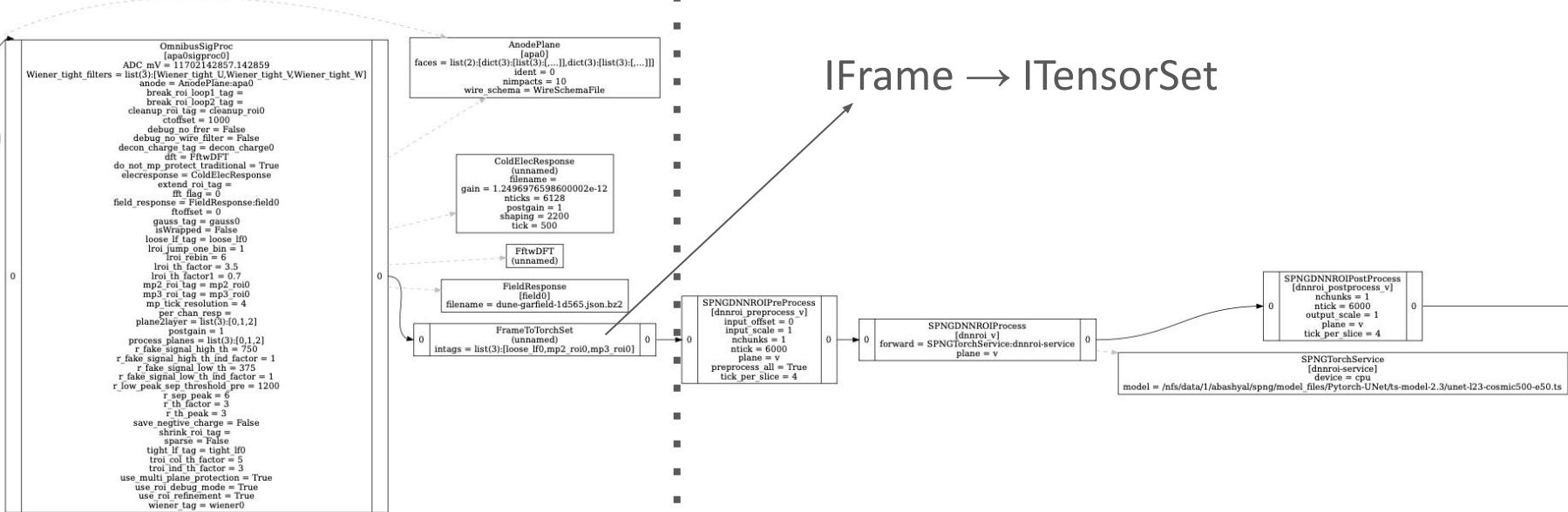
Benchmarking SPNG DNNROI against OSP

- Downsample
 - OSP: Use of WireCell::[Array](#) library library on Eigen Arrays
 - SPNG: torch libraries (slicing, average-pool)
- Inference
- Upsampling
 - OSP: Use of WireCell::[Array](#) library library on Eigen Arrays
 - SPNG: Use of torch libraries

Benchmarking



Hybrid Workflow

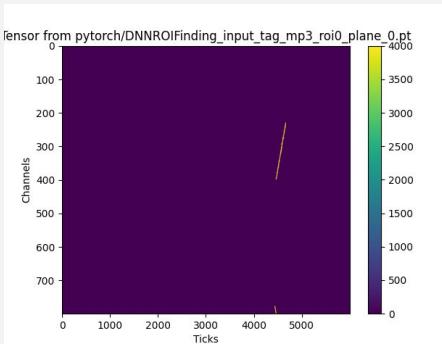
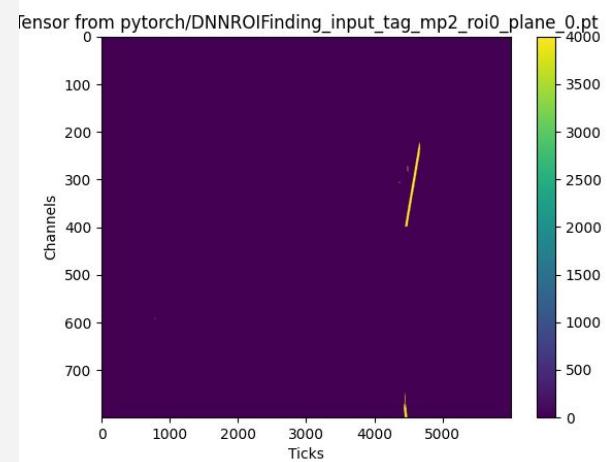
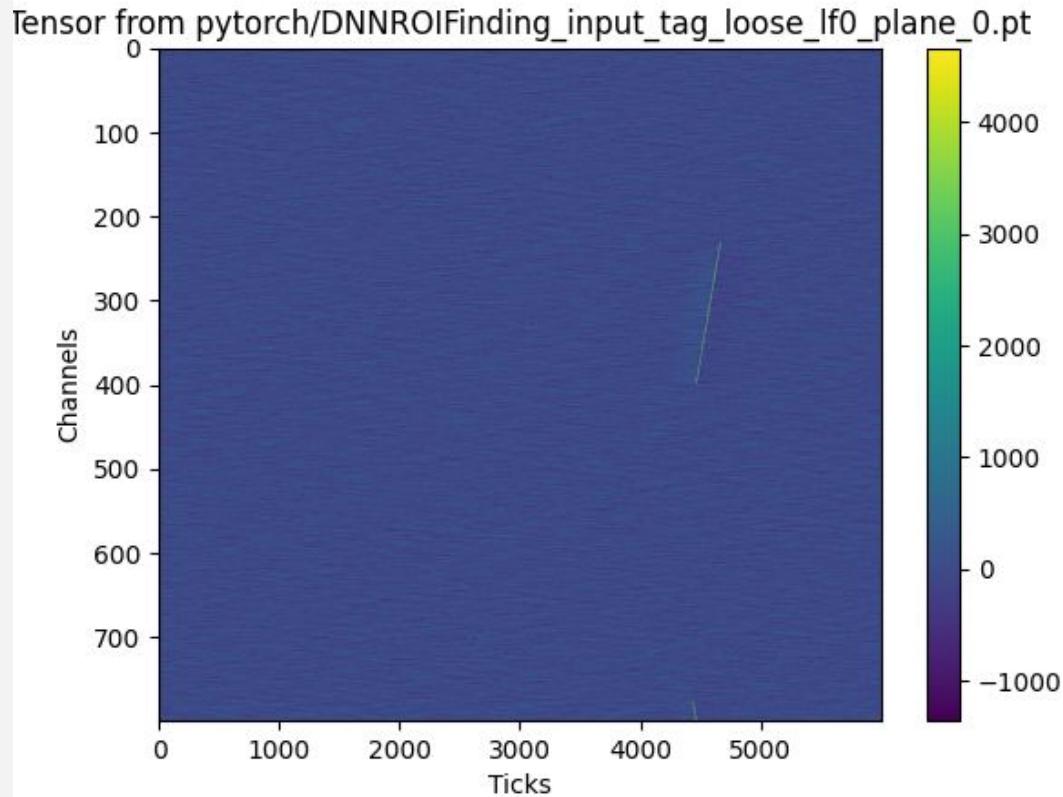


Upstream stages from OSP to ensure both OSP and SPNG DNNROI get identical input

Input Files

- Only showing the U planes
- Except for one slide, comparison shows processing done in the CPU

Input Files

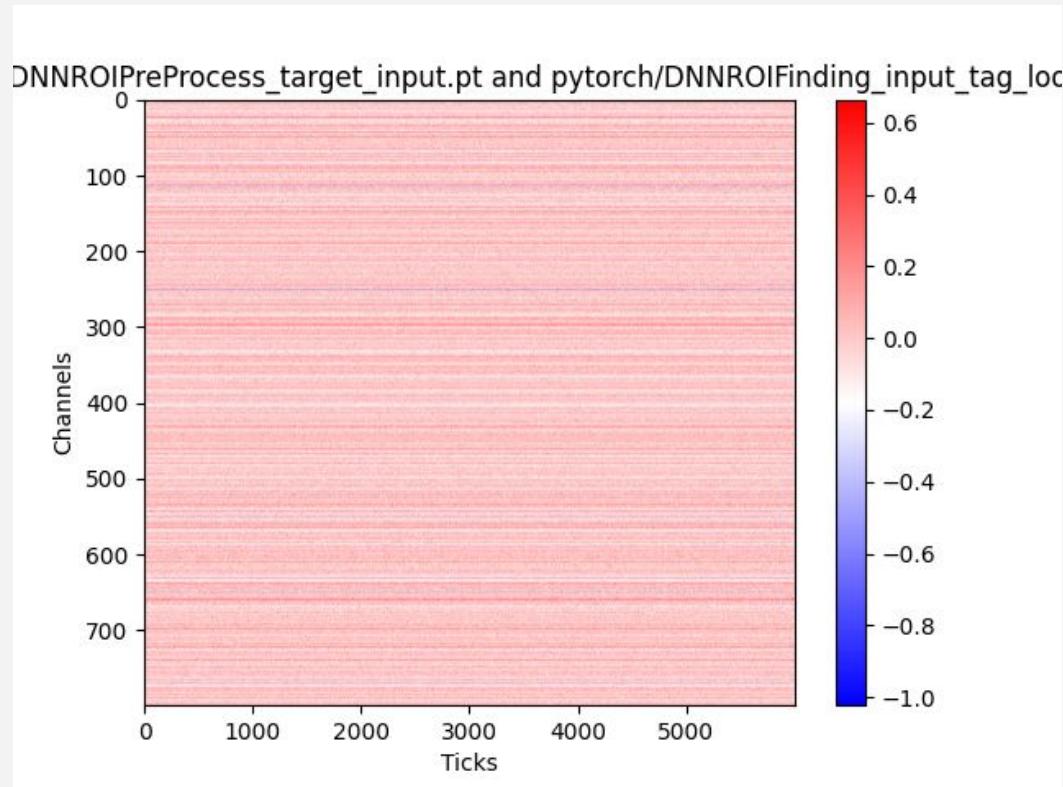


Overall Summary

Stage	OSP	SPNG	MSE
Input File			
Input Target	DNNROIFinding_in	DNNROIPreF	0
Input MP2	DNNROIFinding_in	DNNROIPreF	0
Input MP3	DNNROIFinding_in	DNNROIPreF	0
Downsampled			
Downsampled Target	DNNROIFinding_in	DNNROIPreF	0
Downsampled MP2	DNNROIFinding_in	DNNROIPreF	0
Downsampled MP3	DNNROIFinding_in	DNNROIPreF	0
After Inference			
Target	DNNROIFinding_cl	DNNROIProc	4.57E-11
After Upsampling			
Target	DNNROIFinding_u	DNNROIPost	0
Masking			
Threshold	NA		

Mean Squared Error as a metric to quantify the difference between the OSP and SPNG tensors

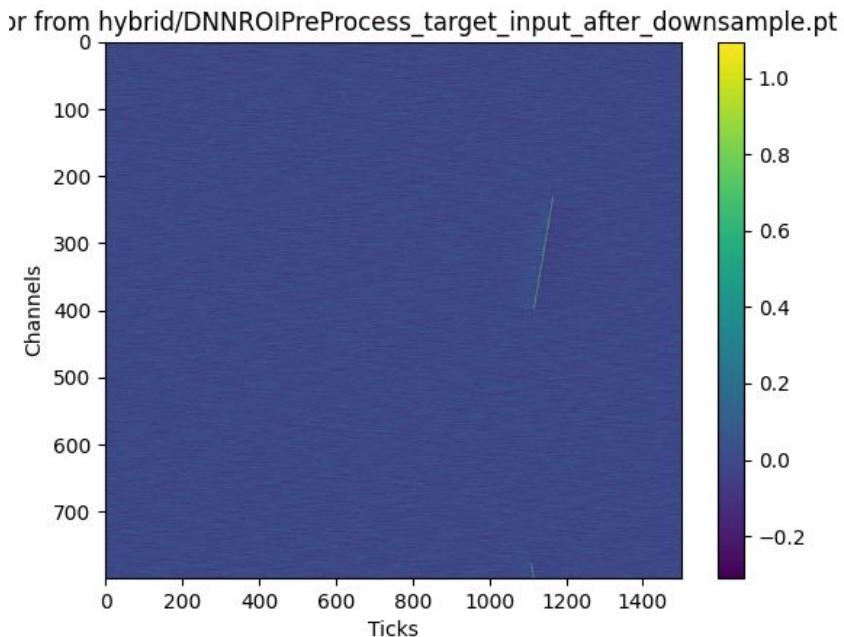
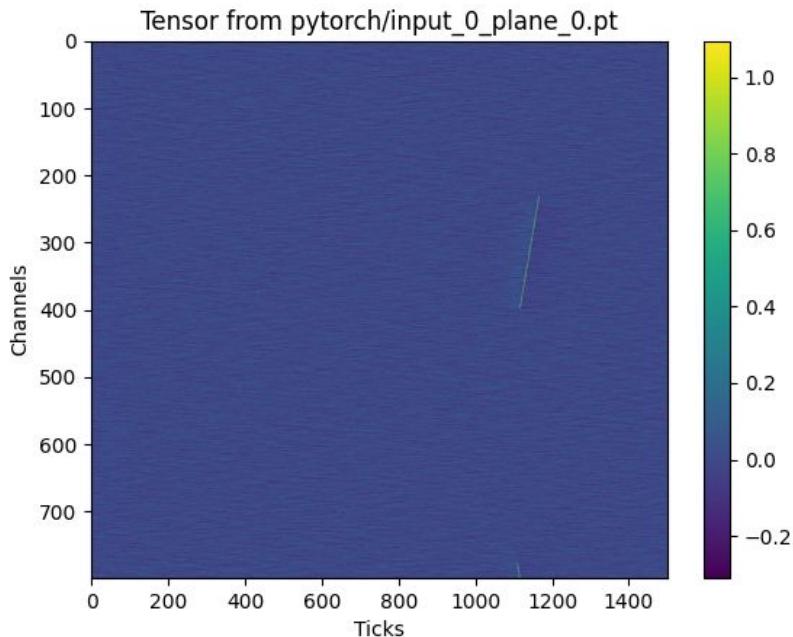
Difference of the input files (CPU- GPU)



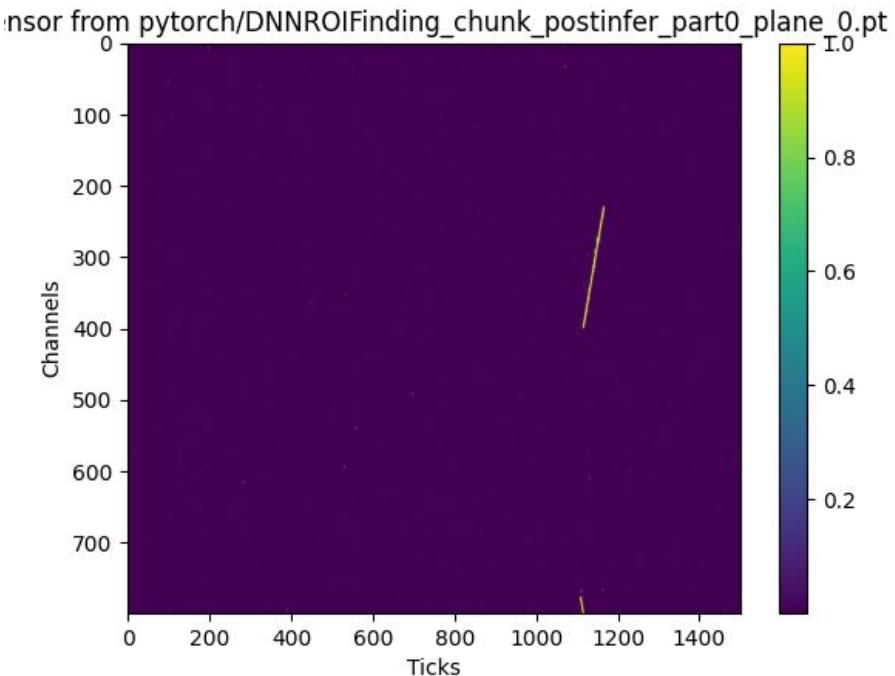
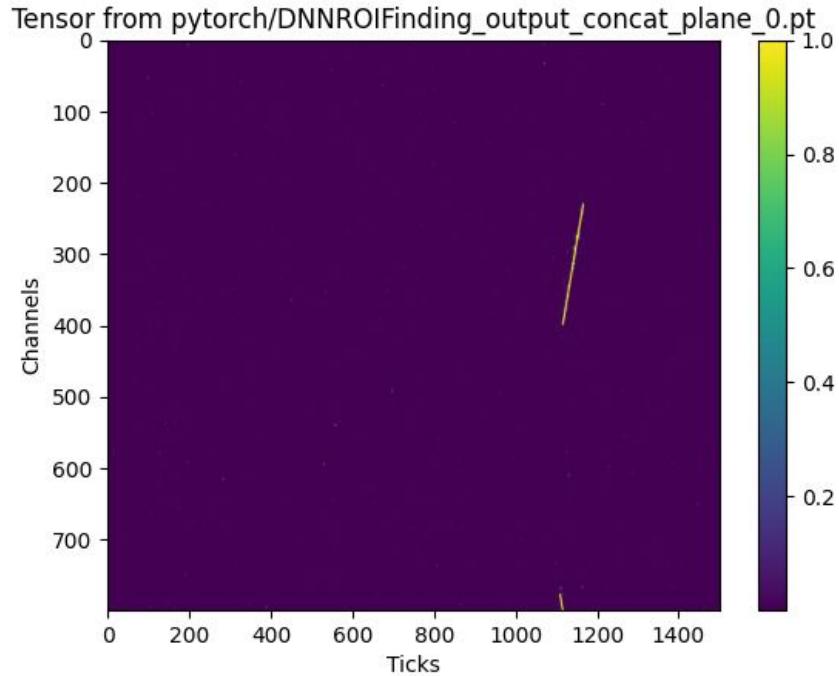
Mean squared error (MSE):
~0.014

Difference between the output
of Post Signal Processing in
GPU and CPU.

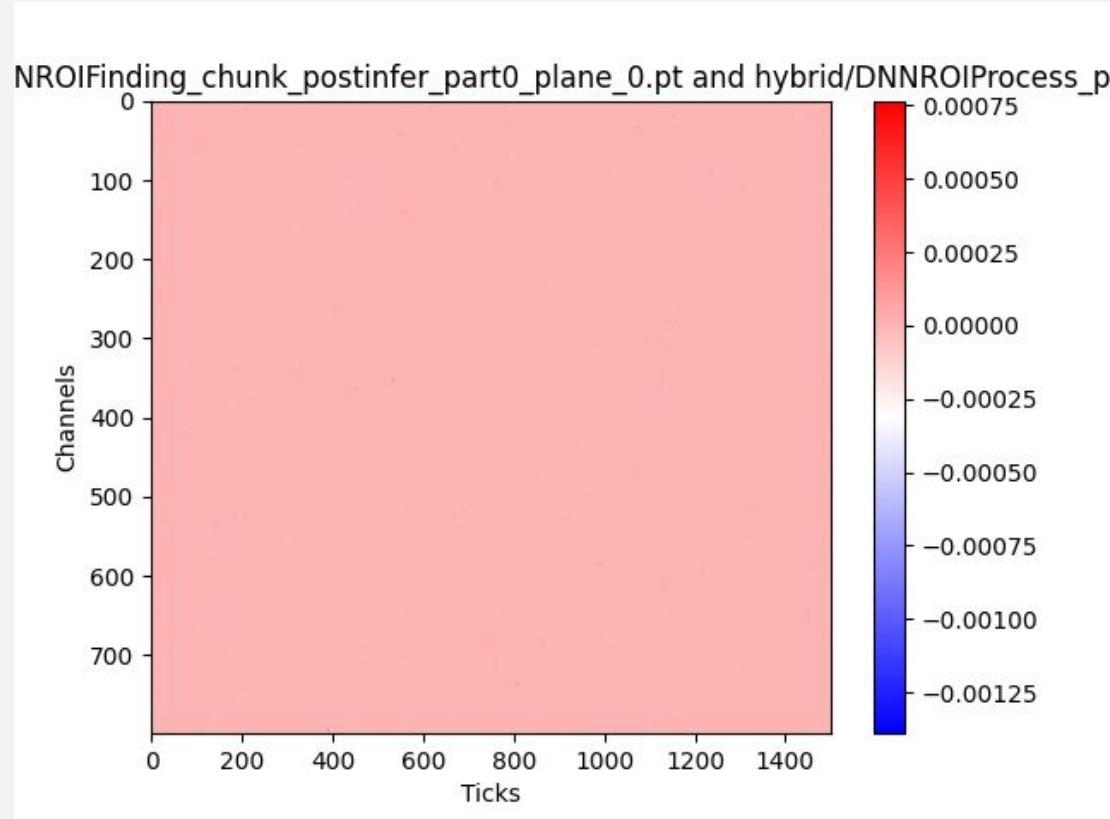
OSP vs SPNG (Downsample)



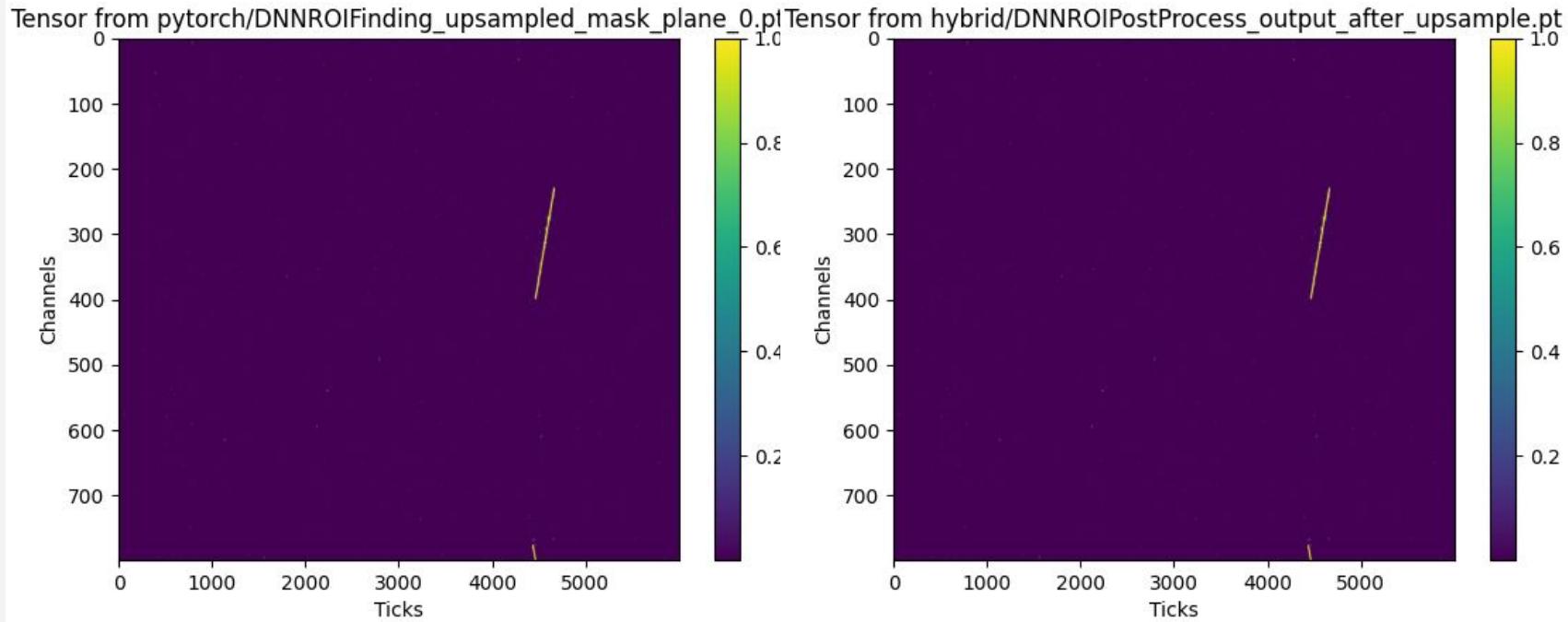
OSP vs. SPNG (Post Inference)



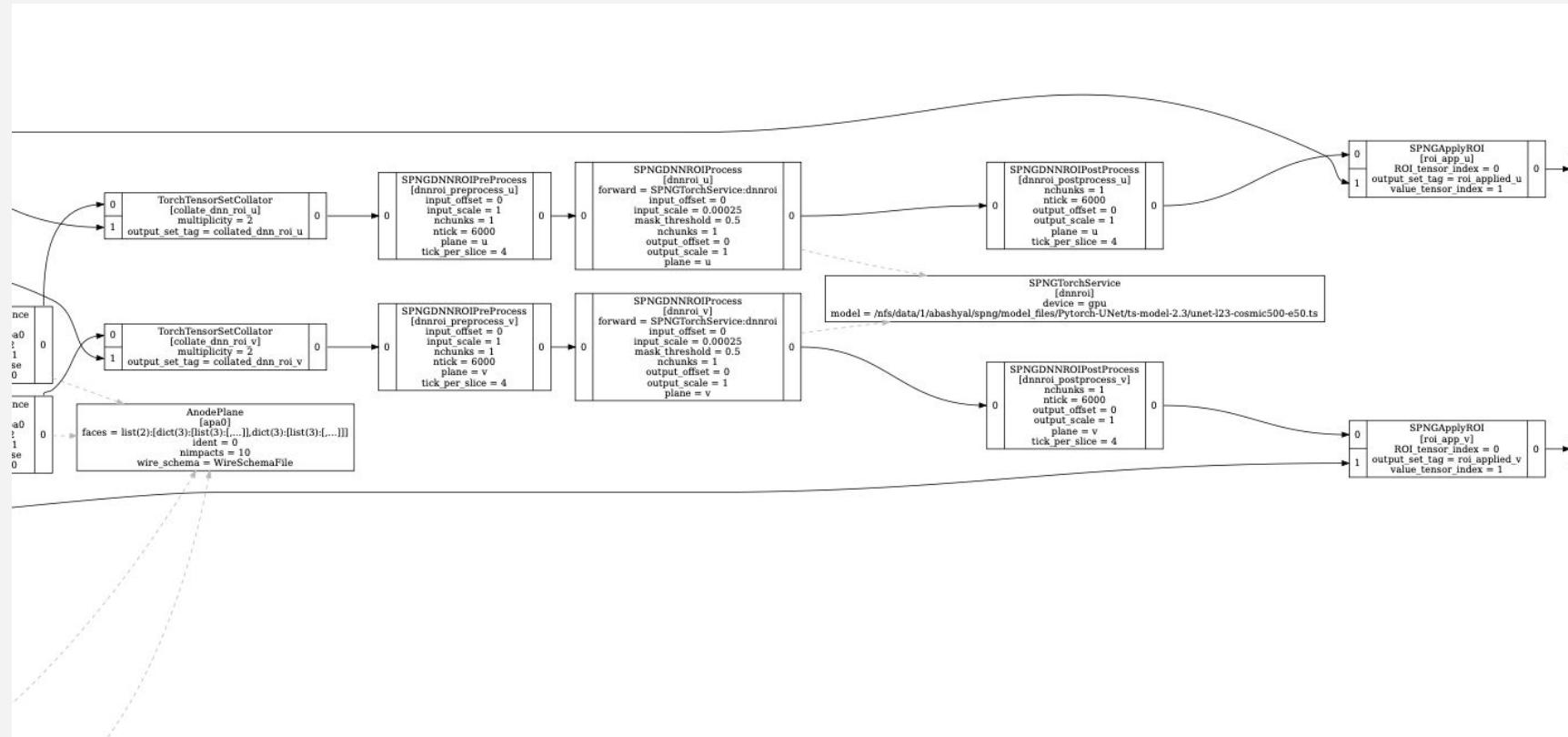
Difference Plot



OSP vs SPNG Upsampling

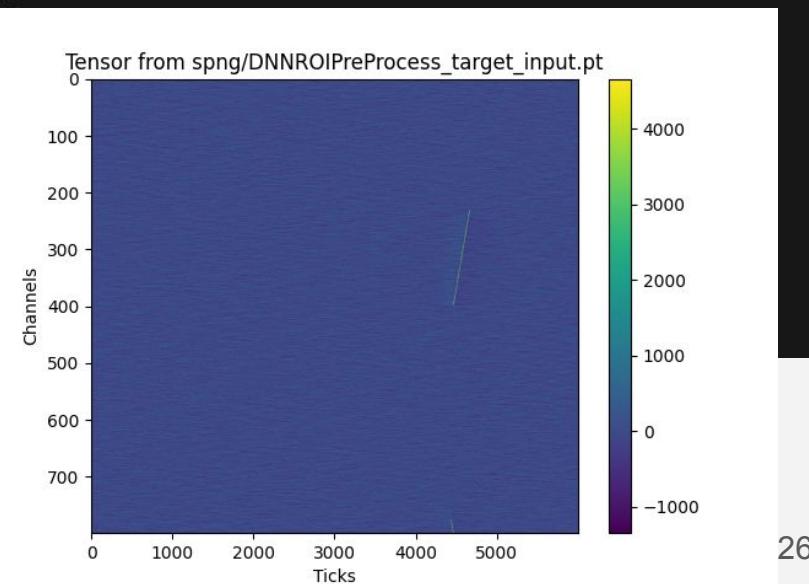


Changing Pre and Post Processing steps to Nodes from service



Input tensor comparison (spng/OSP) target Tensor

```
abashyal@wgpu0:~/nts/data/1/abashyal/spng/spng_dev_09302025/spng_dev/wire-cell-toolkit/spng/test/pytools$ tpython compare_ten
IPreProcess_target_input.pt hybrid/DNNR0IPreProcess_target_input.pt
Parameter from spng/DNNR0IPreProcess_target_input.pt has shape torch.Size([1, 800, 6000]) and dtype torch.float64
Parameter from hybrid/DNNR0IPreProcess_target_input.pt has shape torch.Size([1, 800, 6000]) and dtype torch.float32
Mean Squared Error between the two tensors: 82.09602838303353
Difference plot saved to ./comparison_DNNR0IPreProcess_target_input.pt)_DNNR0IPreProcess_target_input.pt.png
First 20 values of the first tensor:
[-282.031036 -384.61549069 -325.45189639 -128.70842339  122.32225584
 323.26104144  394.27375499  312.05754028  115.30507905 -116.27476158
-298.08968777 -375.39473374 -341.9319162 -235.48779622 -113.86681631
-23.87934457  20.60210069  40.0647697   70.00898544  132.7779269
First 20 values of the second tensor:
[-288.83716 -391.31946 -332.2404   -135.69762   115.13656   315.975
 387.0088   304.89624   108.27026  -123.21338  -304.99573  -382.344
-348.9874  -242.6676  -121.12116  -31.09877   13.53353   33.193
 63.264835  125.99377 ]
First 20 values of the difference:
[6.8061222  6.70396732  6.78849057  6.98919318  7.18569731  7.2854555
 7.26496592  7.16130005  7.03481538  6.93861733  6.90603977  6.94935196
 7.05548004  7.17980632  7.2543461   7.21942557  7.06857046  6.87081141
 6.74415008  6.78416015]
```



We start with different tensors....

Post Inference we end up with different tensors

```
Mean Squared Error between the two tensors: 0.00032544127316214144
```

```
Difference plot saved to ./comparison_DNNROIProcess_post_forward_tensor.pt)_DNNROIProcess_post_forward_tensor.pt.png
```

```
First 20 values of the first tensor:
```

```
[0.1673405  0.23058009 0.10084029 0.0187975  0.00861831 0.02837664  
 0.01102848 0.02291899 0.01905112 0.08326053 0.03201512 0.01032963  
 0.1302534  0.05322308 0.00793222 0.05788688 0.0517725  0.02288031  
 0.00573107 0.00473321]
```

```
First 20 values of the second tensor:
```

```
[0.09786492 0.12614973 0.07085852 0.0160213  0.00960159 0.01806434  
 0.00729345 0.00891055 0.00560344 0.02399396 0.01196703 0.00594258  
 0.0657875  0.03202351 0.00668285 0.03213002 0.01989424 0.00787695  
 0.00487487 0.00441002]
```

```
First 20 values of the difference:
```

```
[ 0.06947558 0.10443036 0.02998176 0.00277621 -0.00098327 0.01031  
 0.00373503 0.01400844 0.01344769 0.05926657 0.02004809 0.00438  
 0.0644659 0.02119957 0.00124937 0.02575687 0.03187826 0.01500  
 0.0008562 0.00032319]
```

