



Status report on **DNNROI sigproc**

Hokyeong Nam
Chung-Ang University

Following up Xin's progress

- Lots of progress made by Xin for DNN ROI studies including new techniques and bug fix
- All the scripts and documents are in private git repo: https://github.com/lastgeorge/DNN_ROI_SP
- Key changes
 - Create pre-processed training data separately from DNN training
 - Flexible DNN inputs: [loose_lf, mp2_roi, mp3_roi] + [**tight_lf, gauss, decon_charge**]
 - New techniques: Knowledge Distillation (KD), Quantization-Aware Training (QAT), INT8
 - Field response file for PD-VD is updated: protodunevd_FR_imbalance3p_260501.json.bz2
 - For PD-VD, training samples now include TDE data
 - DNN ROI threshold is changed from 0.5 to 0.2
 - Wire-Cell-Toolkit now supports DNNROIFindingMultiPlane
- Notes
 - Training time decreased (pre-processed training sample + upload full sample to CPU memory)
 - 6ch input + new techniques showed better performance than baseline (3ch + MobileU-Net V3) in terms of ML metrics
 - PD-VD model trained on BDE + TDE showed better performance than baseline (BDE only)

Following up Xin's progress

- Additional bug fix required for sp-filters.jsonnet in cfg/experiment/protodunevd/

```
97      hf('Wiener_tight_U_b', { sigma: 0.148788 * wc.megahertz, power: 3.76194 }),
98      hf('Wiener_tight_U_t', { sigma: 0.148788 * wc.megahertz, power: 3.76194 }),
99      hf('Wiener_tight_V_b', { sigma: 0.1596568 * wc.megahertz, power: 4.36125 }),
100     hf('Wiener_tight_V_t', { sigma: 0.1596568 * wc.megahertz, power: 4.36125 }),
101     hf('Wiener_tight_W_b', { sigma: 0.13623 * wc.megahertz, power: 3.35324 }),
102     hf('Wiener_tight_W_t', { sigma: 0.13623 * wc.megahertz, power: 3.35324 }),
```

- Parameters are from PD-SP
- Optimized filter parameters for PD-VD BDE
- Which one is used for recent training samples?



```
hf('Wiener_tight_U', {
  sigma: 0.286735 * wc.megahertz,
  power: 8.07074,
}),
hf("Wiener_tight_V", {
  sigma: 0.289041 * wc.megahertz,
  power: 9.94129 }),
hf('Wiener_tight_W', {
  sigma: 0.207127 * wc.megahertz,
  power: 4.13956,
}),
```

Following up Xin's progress

- Trained baseline models with different inputs 3ch/6ch and converted into TorchScript files
- Training data: /nfs/data/1/xqian/toolkit-dev/DNN_ROI_SP/data/pdvd/processed/pdvd_trainval_1200ev_6ch.h5
- Updated to_torchscript.py to support input-channels (e.g. 3 or 4 or 5 or 6)

```
[ckpt] checkpoints/pdvd_base_mbv3_3ch_1200ev/CP25.pth (epoch 25)
[data] test slice [960, 1200) x=(240, 3, 952, 1600) y=(240, 952, 1600)
[model] mobilenetv3 input_channels=3
[eval] all n= 240 loss=0.0175 dice=0.7376 eff_pix=0.6949 pur_pix=0.7995 eff_roi=0.7242 pur_roi=0.8249
[eval] A (crp1,crp2) n= 60 loss=0.0128 dice=0.7657 eff_pix=0.7326 pur_pix=0.8085 eff_roi=0.7443 pur_roi=0.8237
[eval] B (crp3,crp4) n= 60 loss=0.0213 dice=0.6759 eff_pix=0.6314 pur_pix=0.7430 eff_roi=0.6634 pur_roi=0.7681
```

metric	all	A (crp1,crp2)	B (crp3,crp4)
loss	0.0175	0.0128	0.0213
dice	0.7376	0.7657	0.6759
eff_pix	0.6949	0.7326	0.6314
pur_pix	0.7995	0.8085	0.7430
eff_roi	0.7242	0.7443	0.6634
pur_roi	0.8249	0.8237	0.7681
n	240	60	60

```
[ckpt] checkpoints/pdvd_base_mbv3_6ch_1200ev/CP11.pth (epoch 11)
[data] test slice [960, 1200) x=(240, 6, 952, 1600) y=(240, 952, 1600)
[model] mobilenetv3 input_channels=6
[eval] all n= 240 loss=0.0172 dice=0.7548 eff_pix=0.7084 pur_pix=0.8092 eff_roi=0.7292 pur_roi=0.8333
[eval] A (crp1,crp2) n= 60 loss=0.0123 dice=0.7686 eff_pix=0.7225 pur_pix=0.8222 eff_roi=0.7450 pur_roi=0.8315
[eval] B (crp3,crp4) n= 60 loss=0.0212 dice=0.6976 eff_pix=0.6533 pur_pix=0.7505 eff_roi=0.6717 pur_roi=0.7772
```

metric	all	A (crp1,crp2)	B (crp3,crp4)
loss	0.0172	0.0123	0.0212
dice	0.7548	0.7686	0.6976
eff_pix	0.7084	0.7225	0.6533
pur_pix	0.8092	0.8222	0.7505
eff_roi	0.7292	0.7450	0.6717
pur_roi	0.8333	0.8315	0.7772
n	240	60	60

Back Up