Processing TB2020 SRS raw data

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APV 25

- 128 channels
- 192 samples analog pipeline
- 40 MHz sampling/RO frequency
- 50ns nominal shaping time
- ~ 2000 ENC @ 50pF
- 100mV / 25000 electrons



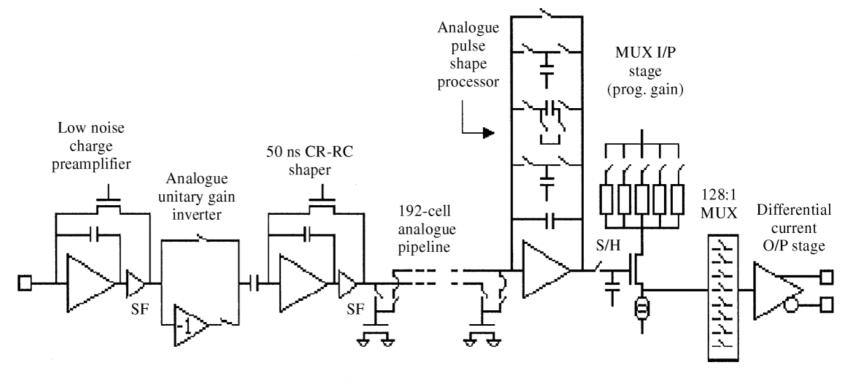


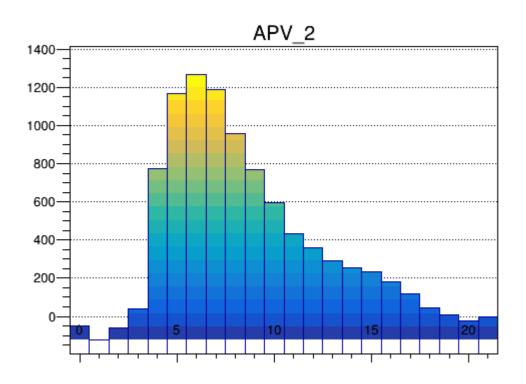
Fig. 1. Block diagram of one channel of the APV25.

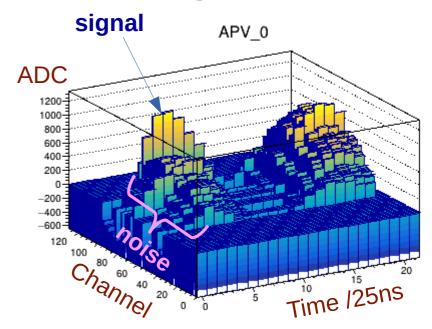
APV 25 sampled signal

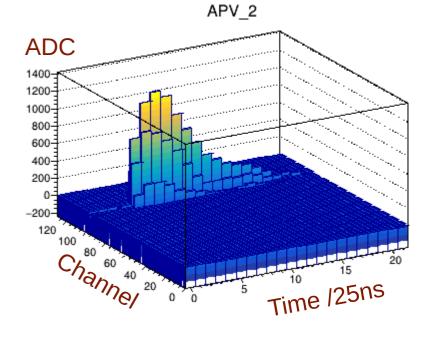
SRS Run 16 5 GeV, Energy scan study, SRS + Telescope

Example plots: APV 0,2 (Master)

Signal shape can be used to reject the noise



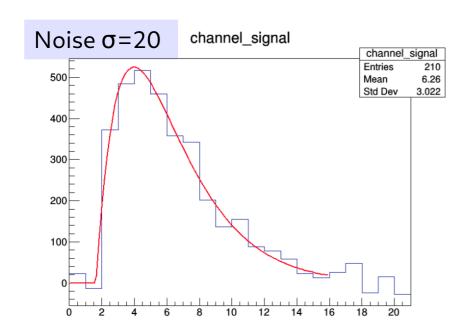


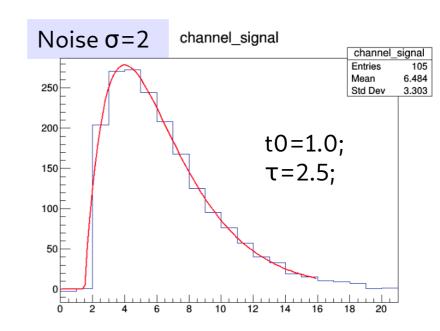


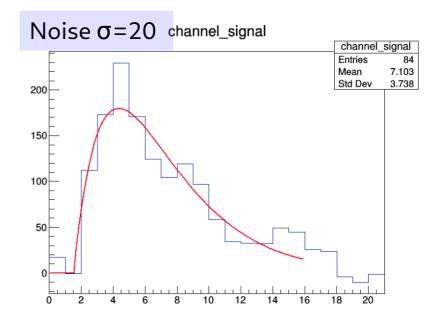
Signal Extraction Optimization

Fit signal with time response function of CR-RC filter:

$$S(t) = A \frac{t - t_0}{\tau} e^{-\frac{t - t_0}{\tau}} \Theta(t - t_0)$$



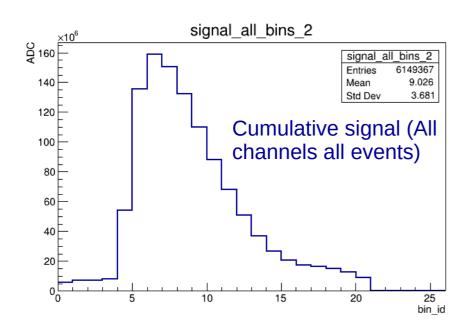


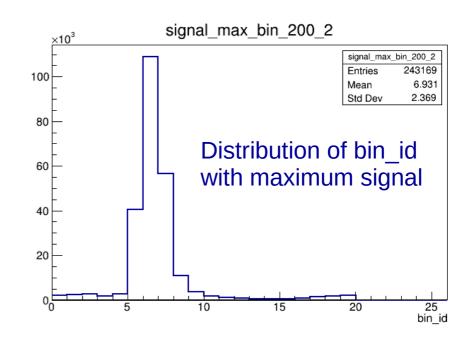


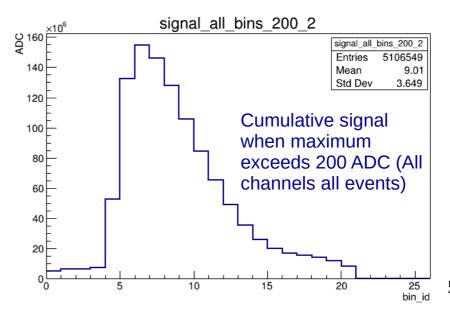
Time bin with signal maximum

SRS Run 16 5 GeV, Energy scan study, SRS + Telescope

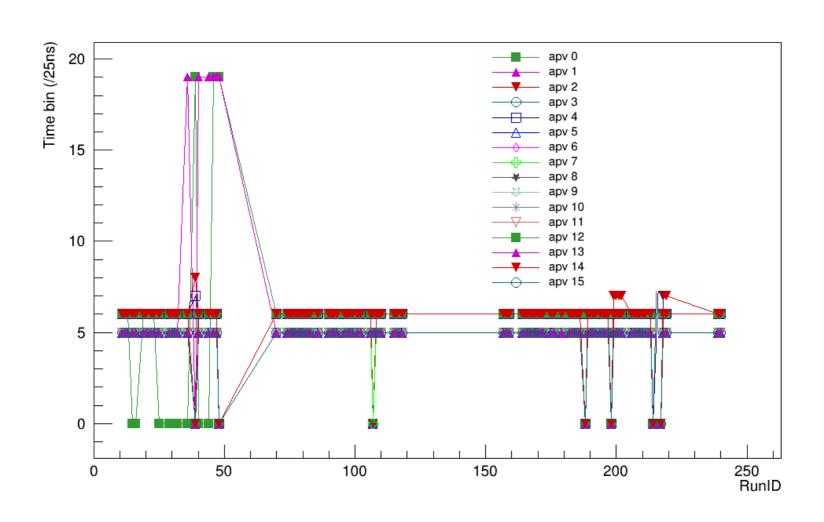
Example plots: APV 2 (Master)







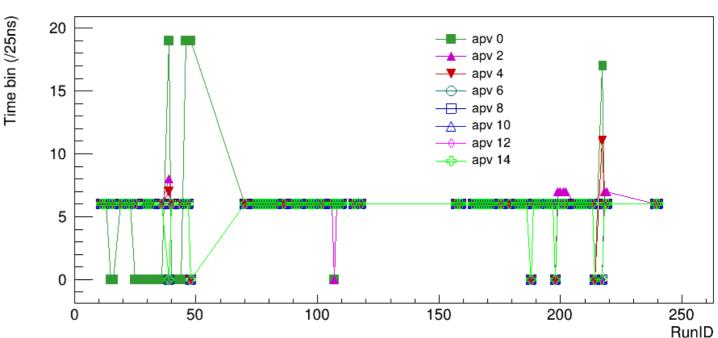
Position of the signal maximum

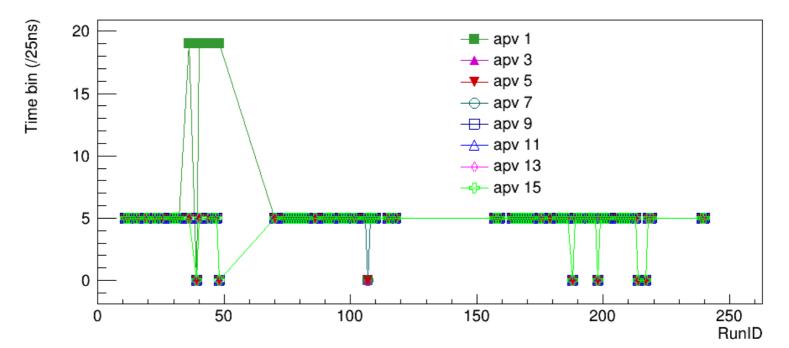


Position of the signal maximum

Max(max_bin(signal_array))

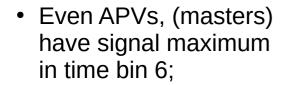
- Even APVs, (masters)
 have signal maximum
 in time bin 6;
- Odd APVs, (slaves)
 have signal maximum
 in time bin 5;



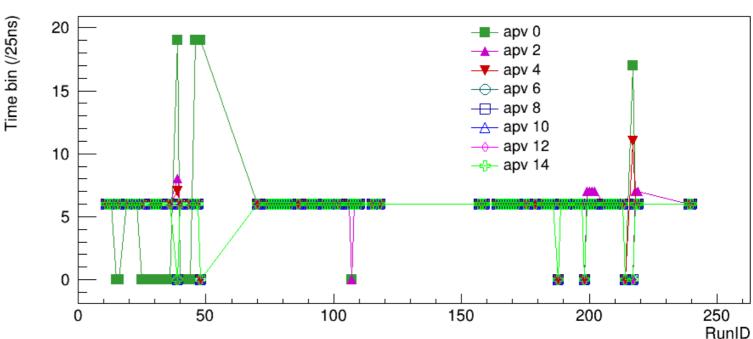


Position of the signal maximum

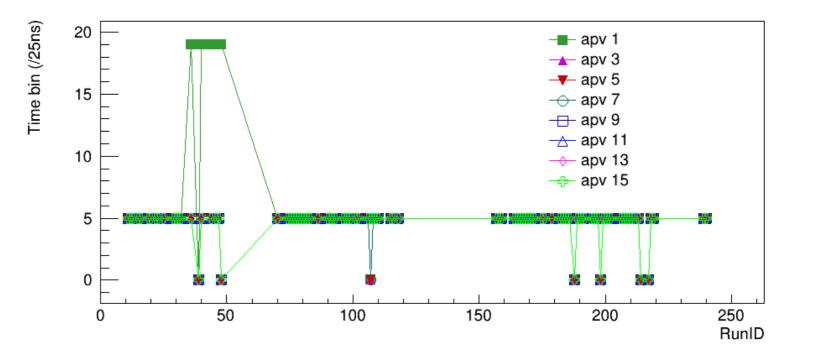
Max(cumulative_signal_array)



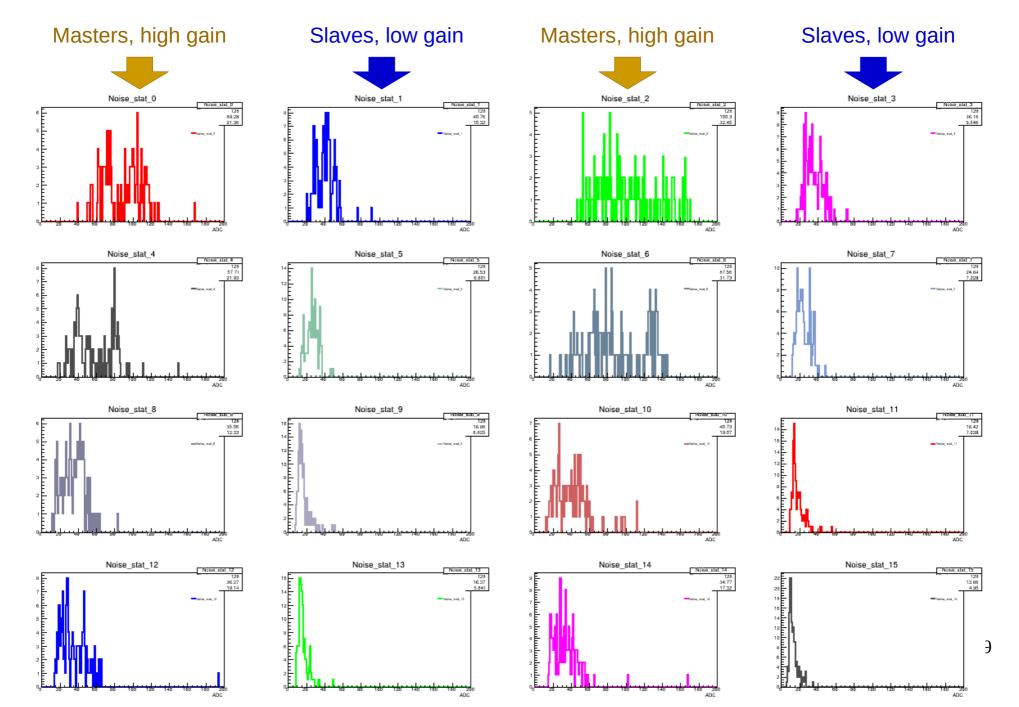
Odd APVs, (slaves)
 have signal maximum
 in time bin 5;



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APVs noise run 182



NN Training

MPV: 50 ADC

• Noise: Mean: 15, sigma: 5

tarining sig sim land 50 noise 15 to 3 15k.txt

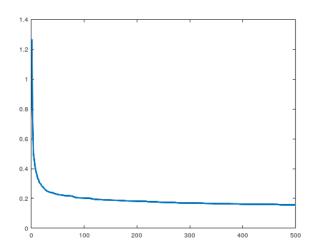
Training Neural Network... Iteration 500 | Cost: 1.564154e-01 Program paused. Press enter to continue.

Visualizing Neural Network...

Training Set Accuracy: 98.614286

Cross validation Set Accuracy: 98.300000

Test Set Accuracy: 97.775000



MPV: 50 ADC

Noise: Mean: 30, sigma: 10

tarining sig sim land 50 noise 30 to 4 15k.txt

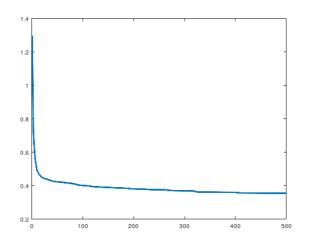
Training Neural Network... Iteration 500 | Cost: 3.551756e-01 Program paused. Press enter to continue.

Visualizing Neural Network...

Training Set Accuracy: 94.700000

Cross validation Set Accuracy: 92.825000

Test Set Accuracy: 92.425000



MPV: 180 ADC

• Noise: Mean: 30, sigma: 10

tarining sig sim land 180 noise 30 to 4 15k.txt

Training Neural Network... Iteration 500 | Cost: 9.489138e-02 Program paused. Press enter to continue.

Visualizing Neural Network...

Training Set Accuracy: 99.985714

Cross validation Set Accuracy: 100.000000

Test Set Accuracy: 100.000000

MPV: 180 ADC

Noise: Mean: 60, sigma: 20

tarining sig sim land 180 noise 60 to 4 15k.txt

Training Neural Network... Iteration 500 | Cost: 1.998111e-01 Program paused. Press enter to continue.

Visualizing Neural Network...

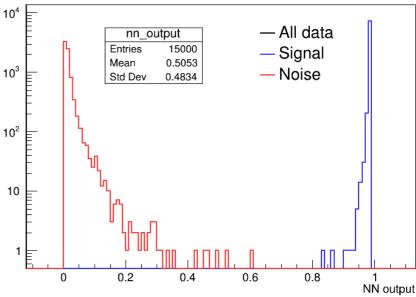
Training Set Accuracy: 99.428571

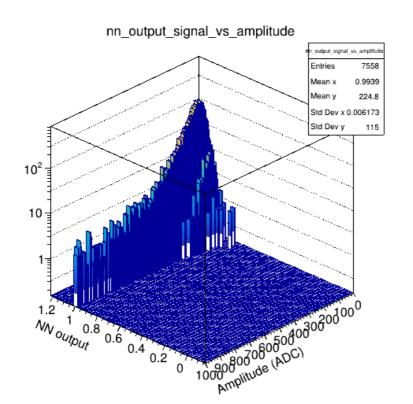
Cross validation Set Accuracy: 97.825000

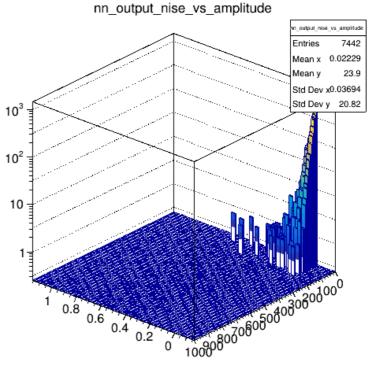
Test Set Accuracy: 98.200000

NN performance

tarining_sig_sim_land_180_noise_30_t0_4_15k.txt







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NN performance

tarining_sig_sim_land_180_noise_60_t0_4_15k.txt

For the threshold at 0.7

Correct predictions:

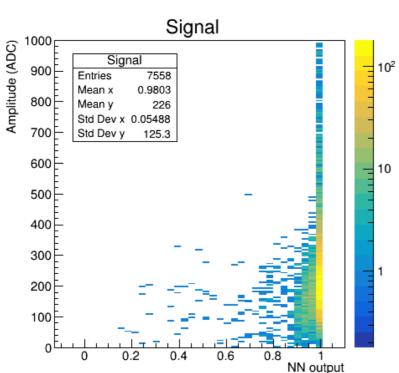
n/n_data: 14816 / 15000 = 98.7733%

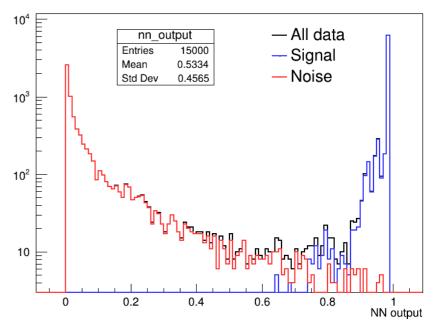
Correct signal:

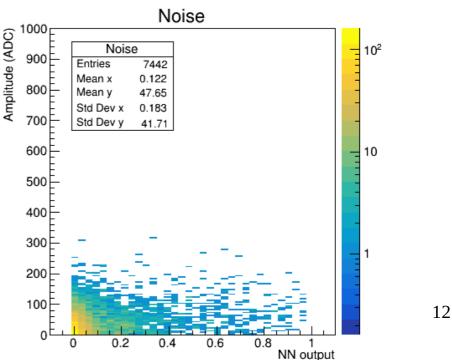
n/n_data: 7492 / 7558 = 99.1268%

Correct noise:

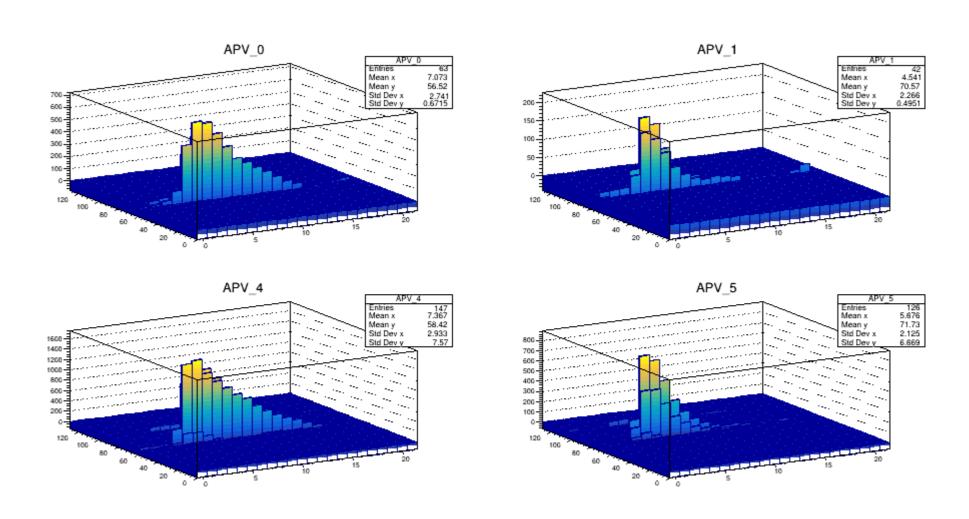
 n/n_data : 7324 / 7442 = 98.4144%



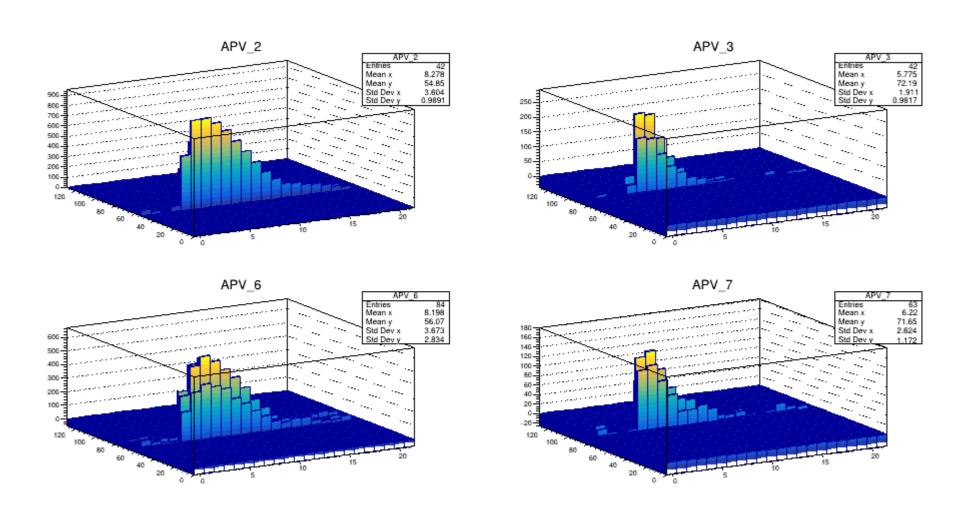




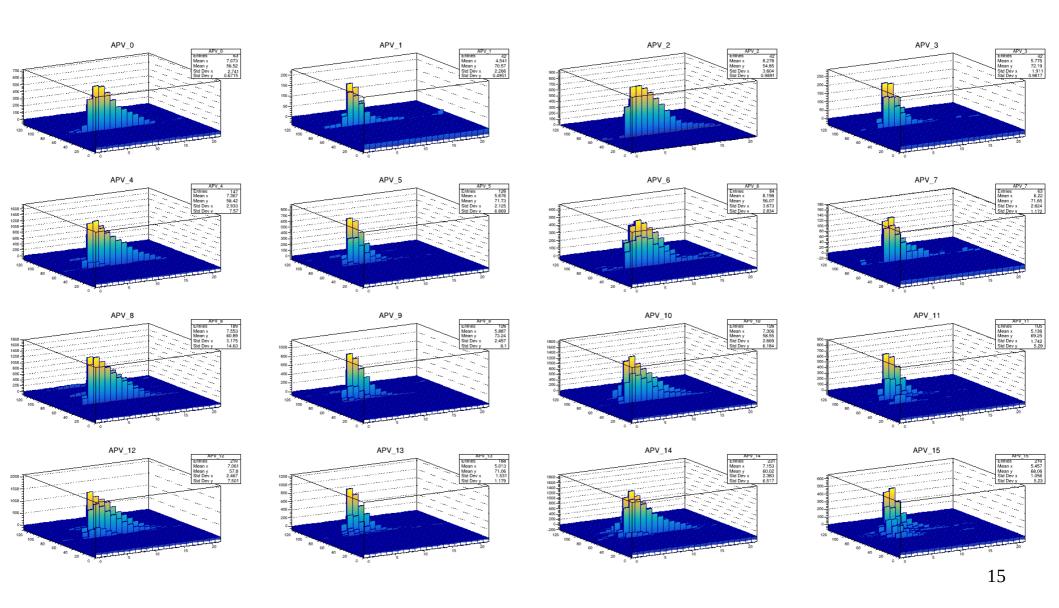
Run 77 event 10



Run 77 event 10



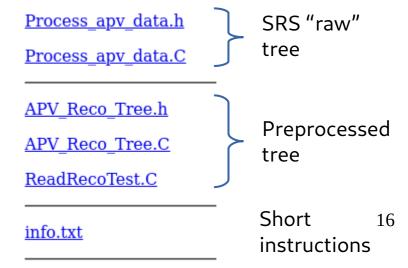
Run 77 event 10



Preprocessed tree

```
p recotree->Branch("apv evt",
                                 &m apv evt);
p recotree->Branch("apv time s",
                                 &m apv time s);
                                                                 std::vector <UInt_t>
                                                                                              m_apv_ch;
p recotree->Branch("apv time us", &m apv time us);
p recotree->Branch("apv fecNo", &p m apv fec);
                                                                 std::vector <Double_t> m_apv_signal_maxbin;
p recotree->Branch("apv id",
                               &p m apv id);
p recotree->Branch("apv ch",
                               &p m apv ch);
                                                                 ......
                                                                    // maximum value of the bins in signal range
p recotree->Branch("apv signal maxbin",
                                         &p m apv signal maxbin);
p recotree->Branch("apv signal bint1",
                                         &p m apv signal bint1);
                                                                    // value of time bin 10
p recotree->Branch("apv signal bint2",
                                         &p m apv signal bint2);
                                                                     // value of time bin 11
p recotree->Branch("apv max bin", &p m apv maxbin);
                                                         // number of the time bin with maximum signal
p recotree->Branch("apv bint1",
                                  &p m apv bint1);
                                                          // number of the fixed time bin for this apv and this run
                                                                    // maximum of the fit with RC-CR response function
p recotree->Branch("apv signal maxfit",
                                         &p m apv signal maxfit);
p recotree->Branch("apv fit t0",
                                         &p m apv fit t0);
                                                                    // to of the fit with RC-CR response function
                                                                    // tau of the fit with RC-CR response function
p recotree->Branch("apv fit tau",
                                         &p m apv fit tau);
                                         &p m apv fit chi2);
p recotree->Branch("apv fit chi2",
                                                                    // chi2 of the fit with RC-CR response function
p recotree->Branch("apv nn output",
                                         &p m apv nn output);
                                                                    // neural network output
p recotree->Branch("apv cm",
                                   &p m apv cm);
                                                 // commomn mode noise
```

https://www.desy.de/~oborysov/tb2020/index.html



SRS "raw" tree

```
p rawtree->SetBranchAddress("apv evt",
                                p m apv evt);
p rawtree->SetBranchAddress("time s",
                                p m time s);
p rawtree->SetBranchAddress("time us",
                               p m time us);
p rawtree->SetBranchAddress("apv fecNo", &p m apv fec);
p_rawtree->SetBranchAddress("apv_id",
                                &p m apv id);
p_rawtree->SetBranchAddress("apv_ch",
                                &p m apv ch);
                               &p_m_mm id);
p_rawtree->SetBranchAddress("mm_id",
p rawtree->SetBranchAddress("mm readout", &p m mm readout);
p rawtree->SetBranchAddress("mm strip",
                               &p m mm strip);
```

Summary

- SRS TB2020 and TB2019 data were preprocessed using the procedure similar to TB2016.
- Several signal estimations are available:
 - Value from the expected maximum time bin (6 for master and 5 for slave);
 - Maximum within the range around expected value (time bins 1-15);
 - Amplitude of the fit with CR-RC response function;
- Output of neural network trained separately for different gain and noise level;
- Parameter of CR-RC response function: t_0 , τ and Chi².
- There is significant difference in signal shape between modified and non-modified APV. Can be taken into account for NN training when confirmed and understood.
- Deconvolution can be also used for signal estimation.