Progress on the peak finding algorithm

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Outline

- Peak finding with derivatives
- Peak finding with deep learning (preliminary)

Peak finding with derivatives

Peak finding with derivatives

• Low pass filter (smoothing)

- Filter out high frequency noises in the waveforms in order to improve the S/N ratio
- Finite impulse response (FIR) filter with a cutoff frequency (F): <u>FIR[i] = a0*x[n] + a1*x[n-1] + …</u>

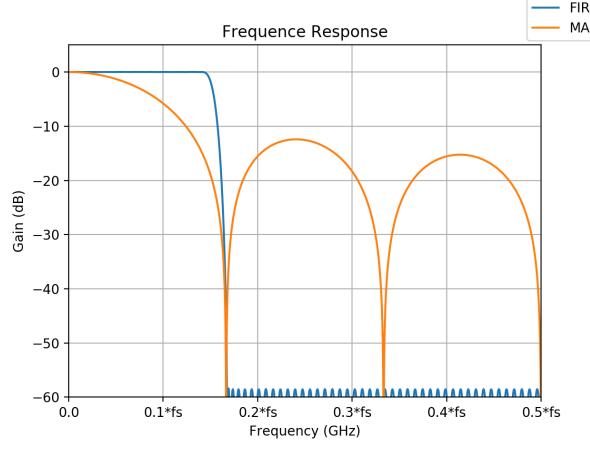
• Derivative (peak detection)

- First derivative (D1): <u>D1[i] = FIR[i] FIR[i G]</u>
- Second derivative (D2): <u>D2[i] = D1[i] D1[i G]</u>
- Hit detection: threshold passing (T)

Parameters:

- Cutoff frequency
- Derivative step
- Threshold

Smoothing with FIR



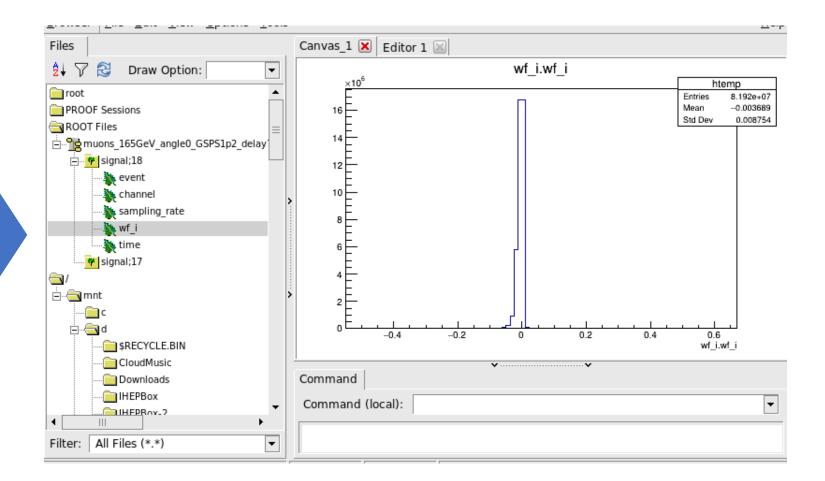
- <u>Digital filter</u>: Reduce impact from noises with high frequencies
- Moving average: Poor frequency response
- **Optimal filter** with Remez exchange algorithm
 - Fast roll-off
 - Good stopband attenuation

References

- 1 J. H. McClellan and T. W. Parks, "A unified approach to the design of optimum FIR linear phase digital filters", IEEE Trans. Circuit Theory, vol. CT-20, pp. 697-701, 1973.
- 2 J. H. McClellan, T. W. Parks and L. R. Rabiner, "A Computer Program for Designing Optimum FIR Linear Phase Digital Filters", IEEE Trans. Audio Electroacoust., vol. AU-21, pp. 506-525, 1973.

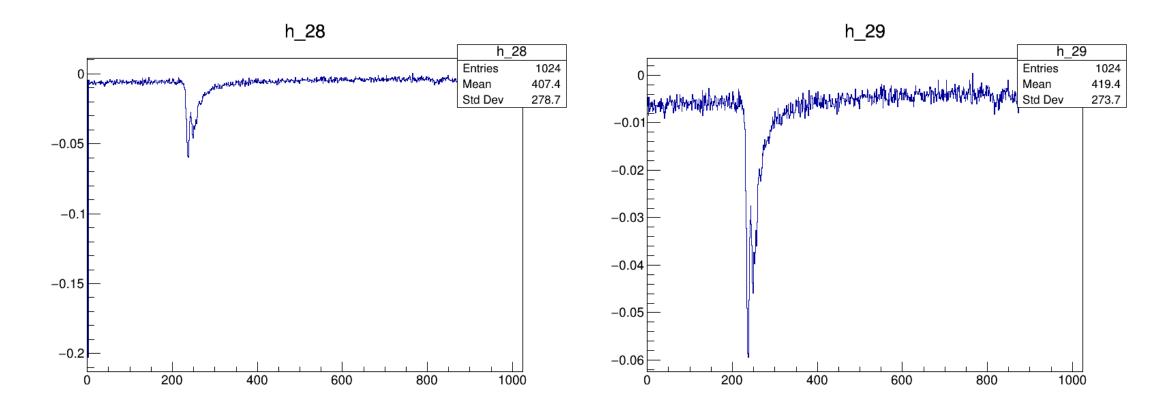
Data preprocessing: converter

| Word | Byte 0 | Byte 1 | Byte 2 | Byte 3 | Contents |
|------|----------------------|-------------|------------------------------|----------------------------|---|
| D | .D. | 'R' | 'S' | .8. | File header, Byte 3 = version |
| 1 | 'T' | т | 'M' | .E. | Time Header |
| 2 | 'B' | # | Board nun | nber | Board serial number |
| 3 | .C. | 10. | .0. | .0. | Channel 0 header |
| 4 | Time Bin Width #0 | | | | Effective time bin width in ns for channel 0 encoded in 4-Byte floating point format |
| 5 | Time Bin Width #1 | | | | |
| | | | | | |
| 1027 | Time Bin Width #1023 | | | | |
| 1028 | .C. | 10. | .0. | 11. | Channel 1 header |
| 1029 | Time Bin Width #0 | | | | Effective time bin width in ns for channel 1 encoded in 4-Byte floating point format |
| 1030 | Time Bin Width #1 | | | | |
| | | | | | |
| 2052 | Time Bin Width #1023 | | | | |
| 2053 | 'E' | н. | 'D' | 'R' | Event Header |
| 2054 | Event Seri | ial Number | | | Serial number starting with 1 |
| 2055 | Year | | Month | | Event date/time 16-bit values |
| 2056 | Day | | Hour | | |
| 2057 | Minute | | Second | | |
| 2058 | Millisecond | | Range | | Range center (RC) in mV |
| 2059 | .B. | # | Board nun | nber | Board serial number |
| 2060 | .C. | 101 | .0. | .0. | Channel 0 header |
| 2061 | Scaler #1 | | | Scaler for channel 0 in Hz | |
| 2062 | 'T' '#' Trigger cell | | Channel 0 first readout cell | | |
| 2063 | Voltage Bin #0 | | Voltage Bi | n #1 | Channel 0 waveform data encoded in 2-Byte integers. 0=RC-0.5V and 65535=RC+0.5V. RC see header. |
| 2064 | Voltage Bin #2 | | Voltage Bi | n #3 | |
| | | | | | |
| 2574 | Voltage Bin #1022 | | Voltage Bi | n #1023 | |
| 2575 | .C. | .0. | .0. | 11 | Channel 1 header |
| 2576 | Scaler #2 | | | | Scaler for channel 1 in Hz |
| 2077 | 'T' | '# ' | Trigger ce | | Channel 1 first readout cell |
| 2578 | Voltage Bi | in #0 | Voltage Bi | n #1 | Channel 1 waveform data |
| 2579 | Voltage Bin #2 | | Voltage Bi | n #3 | encoded in 2-Byte integers. 0=RC-0.5V and 65535=RC+0.5V. RC see header. |
| | | | ••• | | |
| 3089 | Voltage Bin #1022 | | Voltage Bi | n #1023 | |
| | | °H' | 'D' | 'R' | Next Event Header |



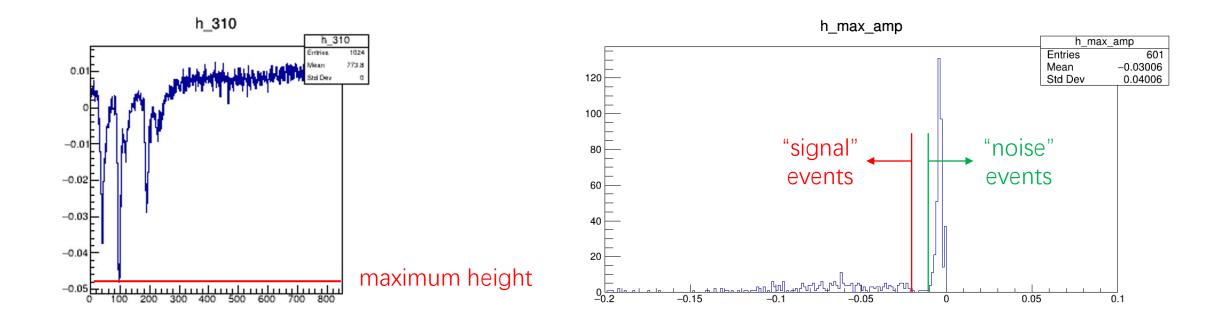
Convert the binaries to ROOT format that compatible to the counting algorithm

Data preprocessing: Outlier removal



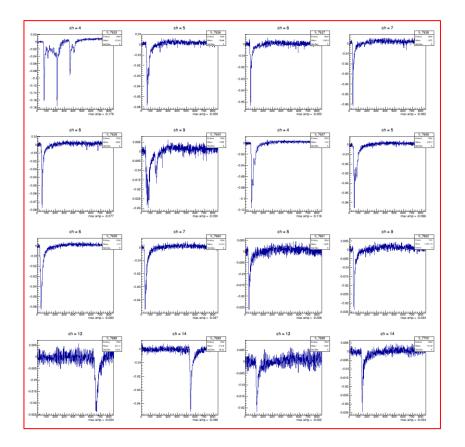
Some outliers appear at the very beginning or ending of the waveform. Detect and remove them automatically by an algorithm.

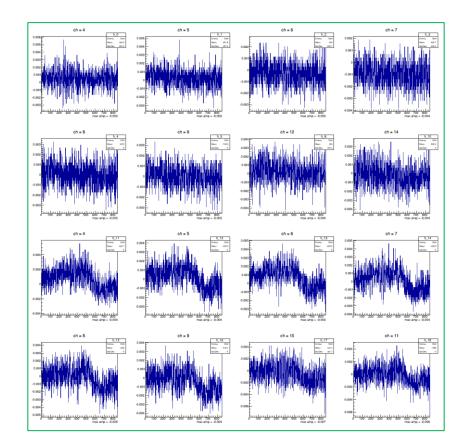
Data preprocessing: classification



Classify waveforms to "signals" and "noises"

Data preprocessing: classification (II)



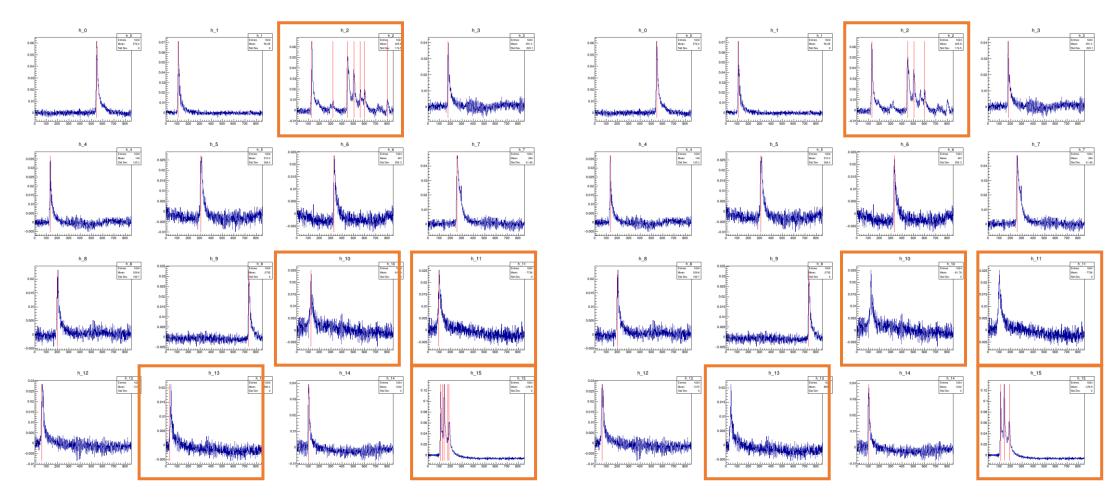


Further optimize the peak finding algorithm by using these classified events. Scan the parameters and control the fake rate with the "noise" events. Look at the peak finding with the "signal" events

Peak finding

FIR

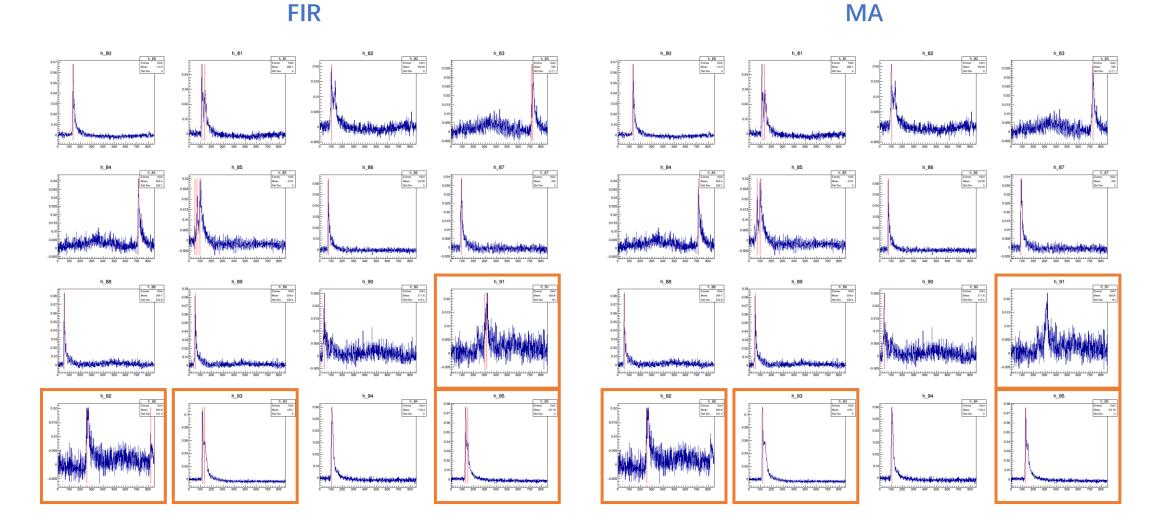




muons_165GeV_angle0_GSPS1p2_delay725ns_7Nov_0321



MA



muons_165GeV_angle0_GSPS1p2_delay725ns_7Nov_0321

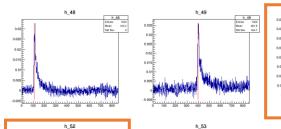
Peak finding (III)

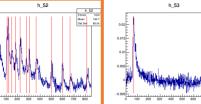
FIR

h_50

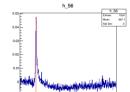
300 400 500 600

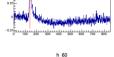
h_54

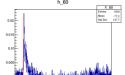


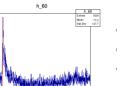


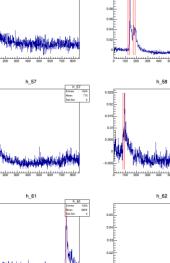
0.03



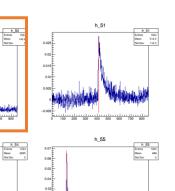


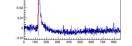


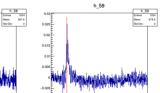




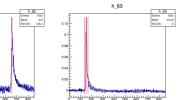
h_53 Entries 1024 Mean -454.2 Skil Dev 0



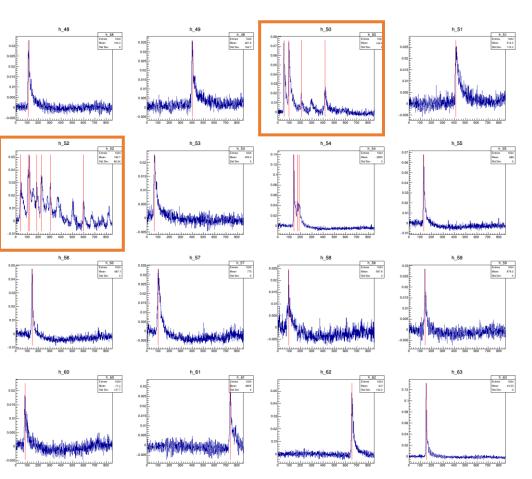












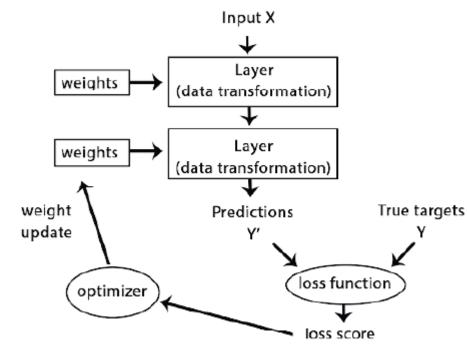
Conclusion

- Using FIR filter instead of MA, as the former one gives much better frequency response.
- Apply the peak finding algorithms with derivatives to the beam test data
 - Can detect most of the major peaks
 - Will focus on the small peaks. Need to discriminate them from noises
- Open question:
 - Why most waveforms only has one "big" peak?

Peak finding with deep learning

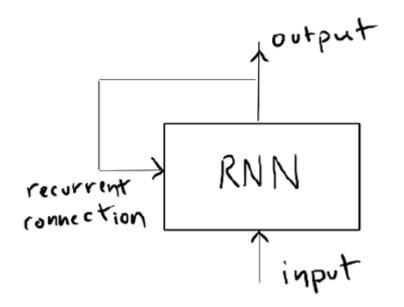
Peak finding with deep learning

For time-sequence problem



Densely-connected network:

• No memory among sequence elements



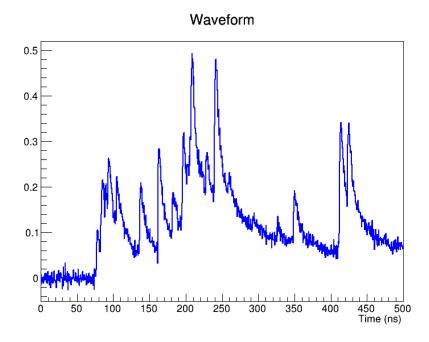
Recurrent Neural Network (RNN):

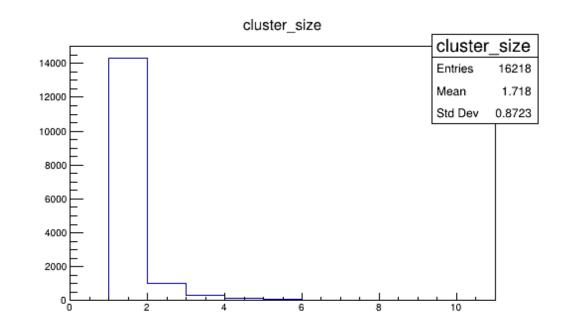
- Internal loops over sequence elements. Has "memories"
- Powerful to handle time-sequence problems

Dataset

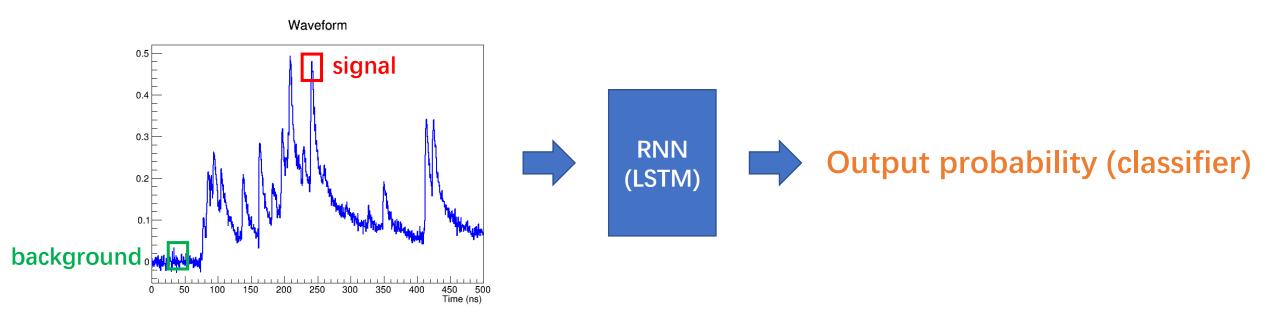
• Toy Monte Carlo samples

- ~20 primary ionizations per waveform (~90% cluster size == 1)
- ~10% noise level
- 10000 samples





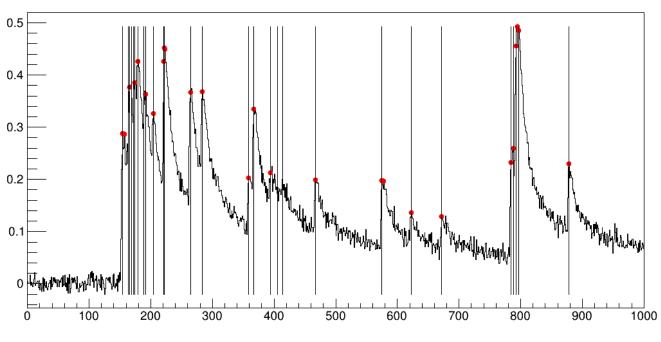
Model



- Using slices of the waveform as the inputs
- Using LSTM (a variation of RNN) to train the dataset
- Preliminary results show good performances (training has not been fully optimized yet)

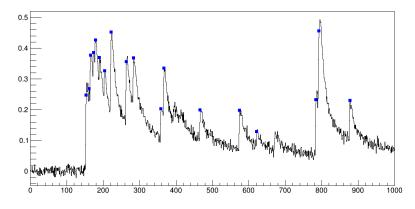
Peak finding

RNN (LSTM)



Black line: truth (primaries and secondaries)

Derivative



Peak finding (II)

RNN (LSTM) 0.3 0.25 0.2 0.15 0.1 1W . ЧM 0.05 0 -0.05 500 200 300 600 700 1000 100 400 800 900

Derivative 0.3 0.25 0.2 0.15 0.1 0.05 -0.05 200 300 1000 100 600 900 400 500 800

Black line: truth (primaries and secondaries)

Conclusion

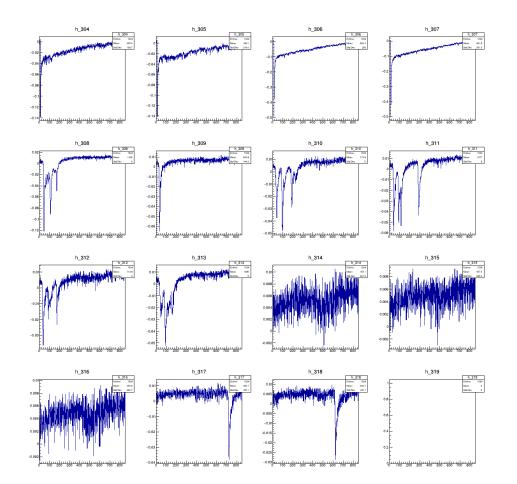
- Peak finding with the LSTM model show good performance for MC samples
- Next to do
 - Need further tunning the model
 - Will make more detailed comparisons to the derivative-method
 - Consider to classify primary peaks and secondary peaks

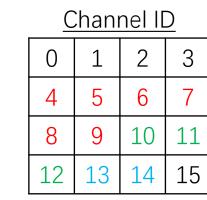
Open question:

• How to apply the model to data? As differences exist between data and MC.

Backup

Waveform display: An example from an event

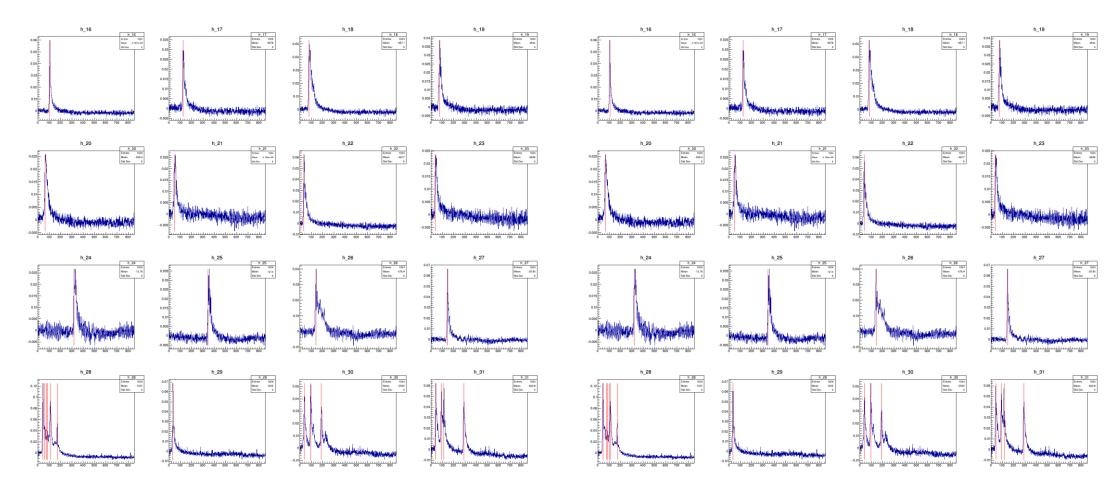




ch0-3: trigger ch4-9: 1 cm ch10-12: 2 cm ch13-14: 3 cm ch15: NULL

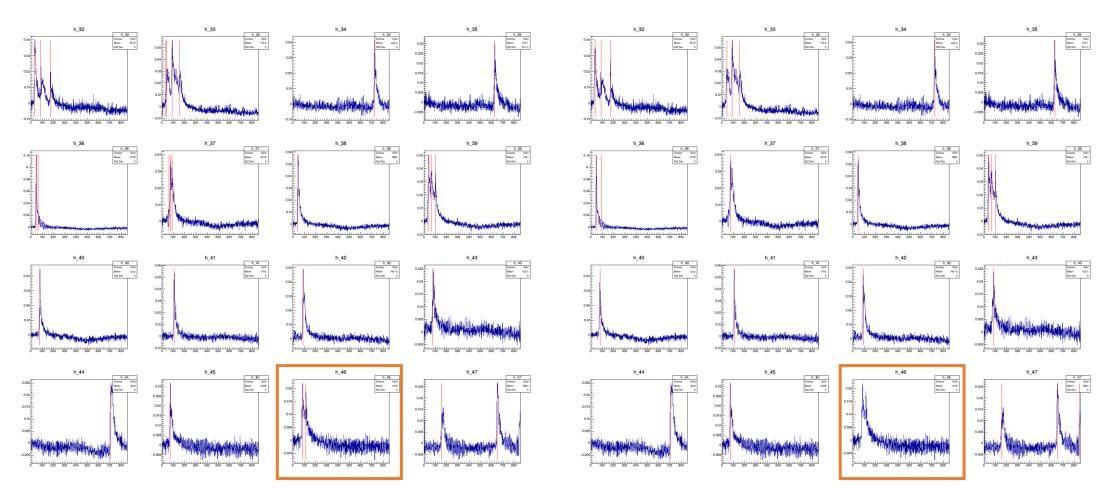
FIR





FIR

MA 8

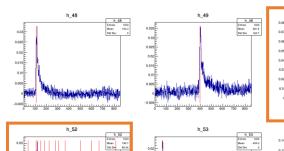


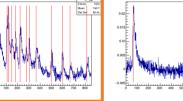
FIR

h_50

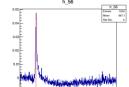
300 400 500 600

h_54

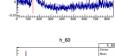


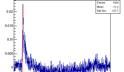


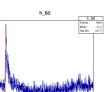
0.03

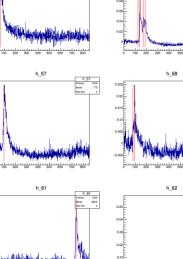


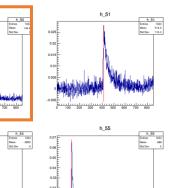
h_56

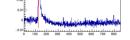


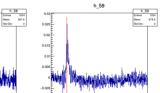




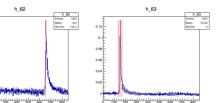




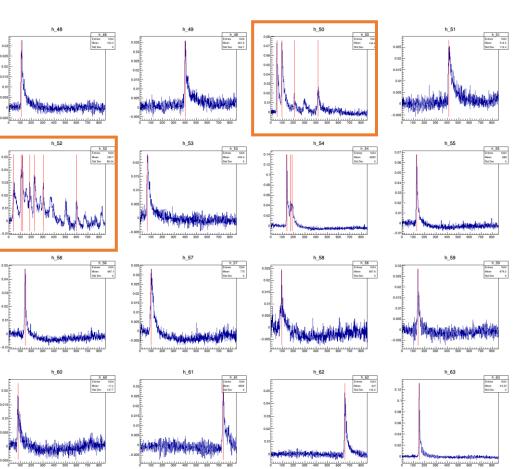






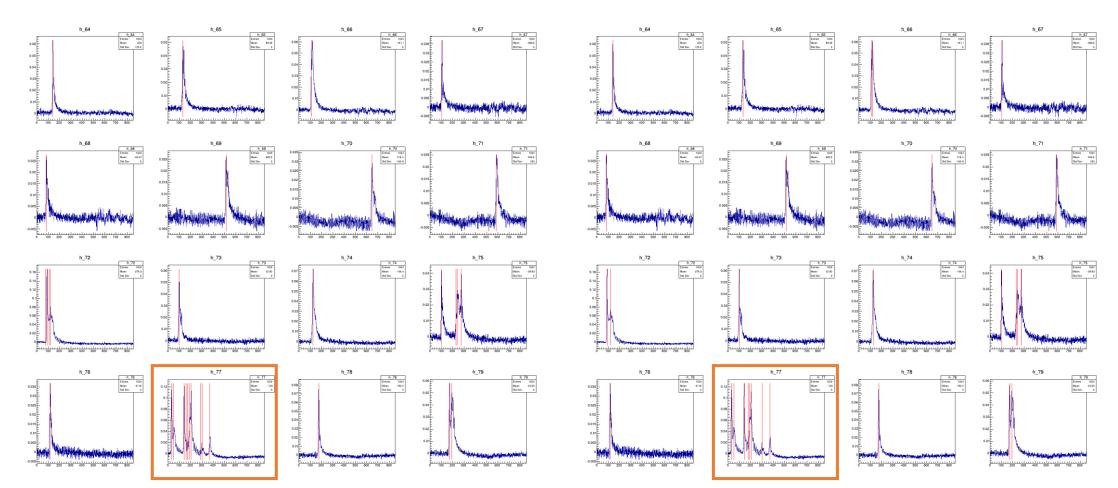






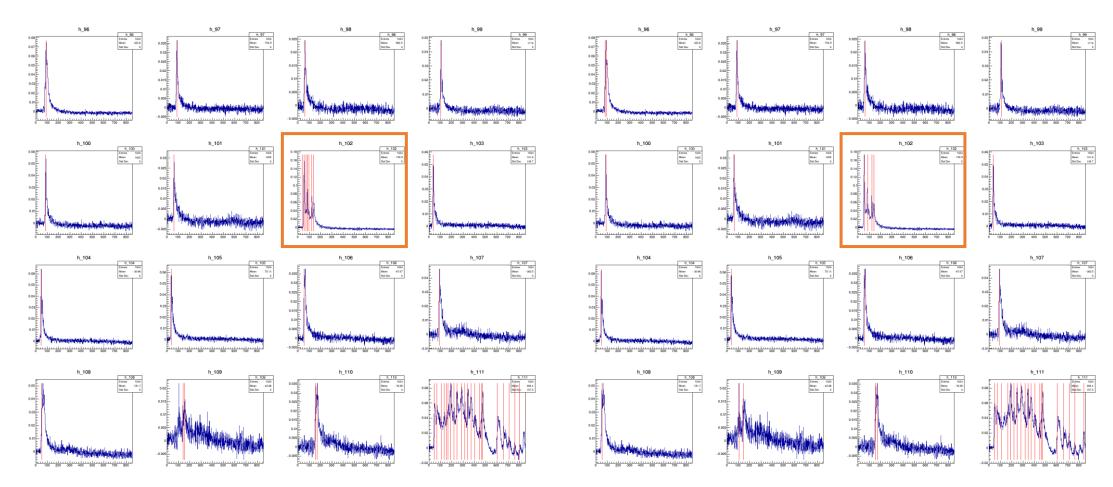
FIR

MA 8

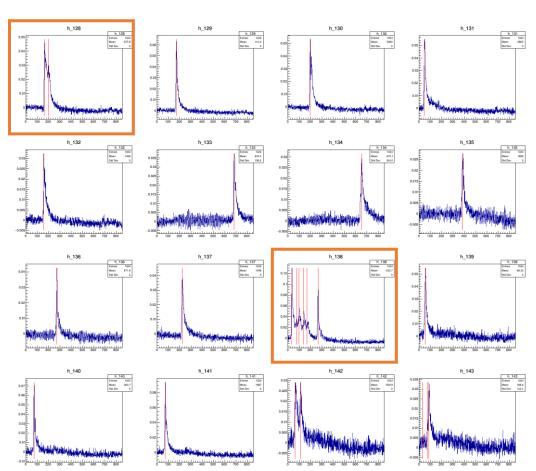


FIR

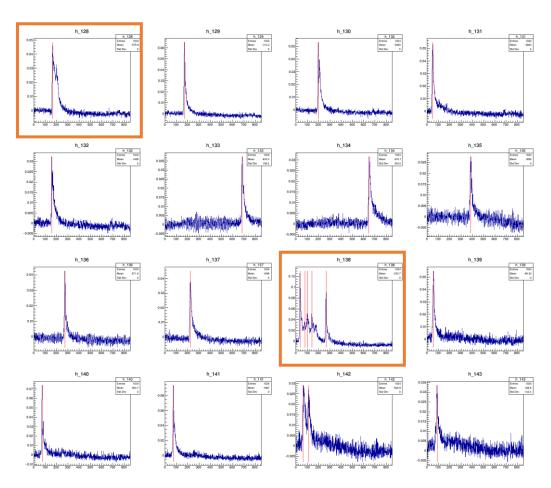




FIR



MA 8



FIR

MA 8

