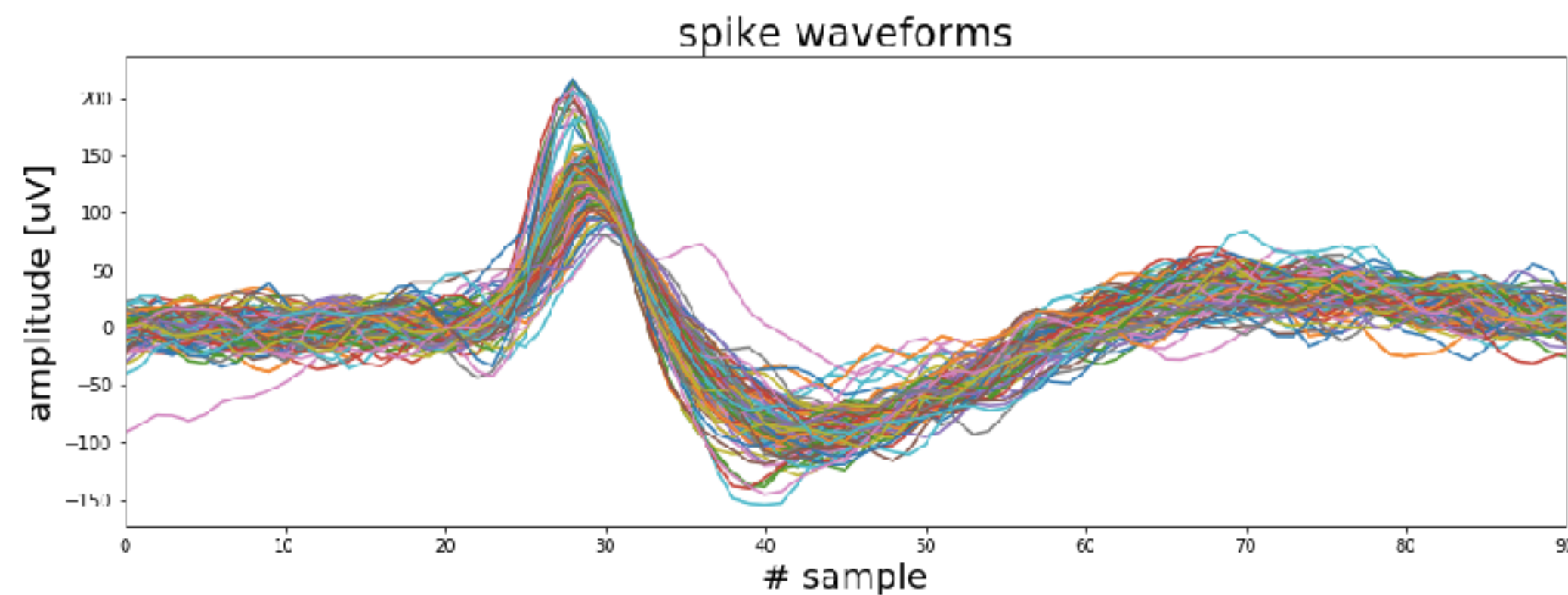


# **Understanding Machine Learning models**

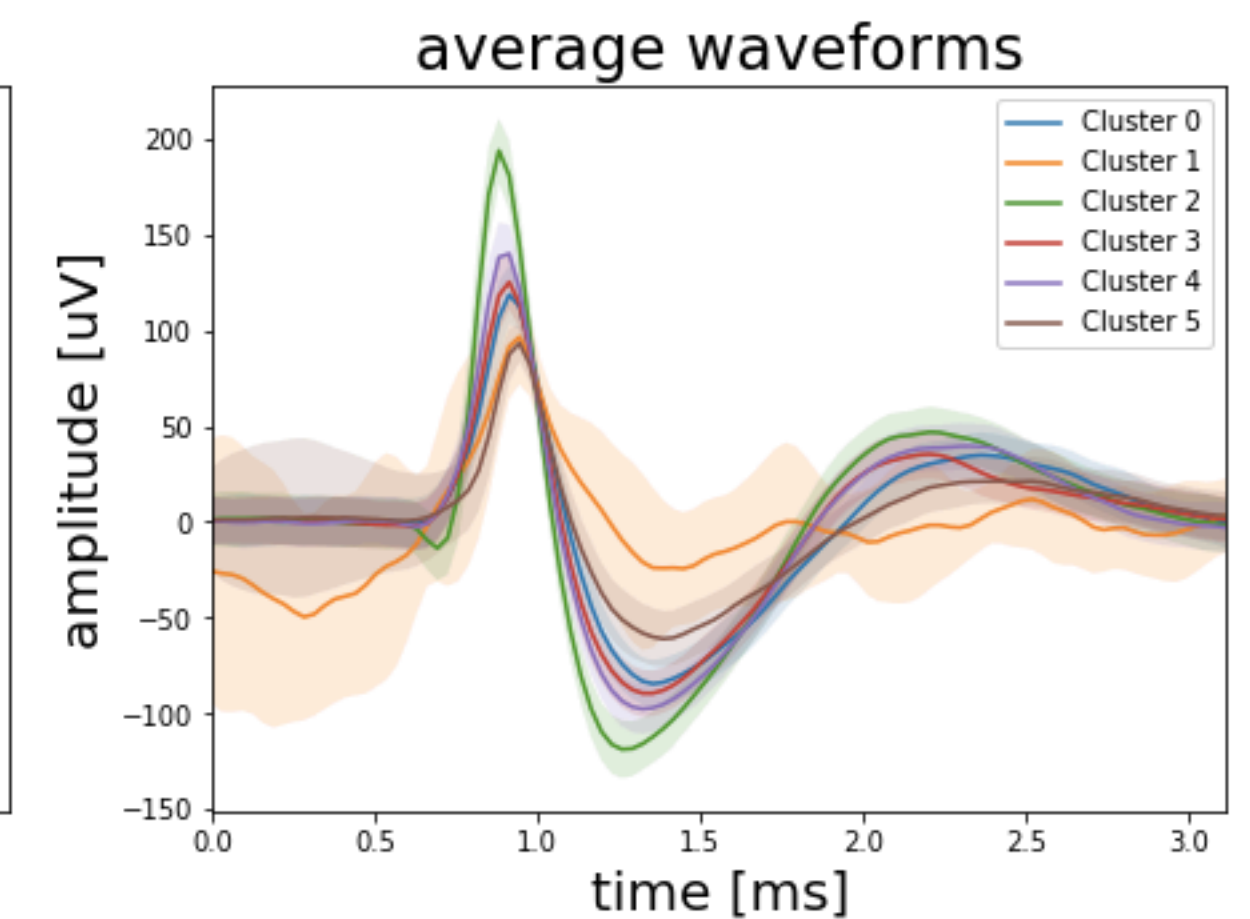
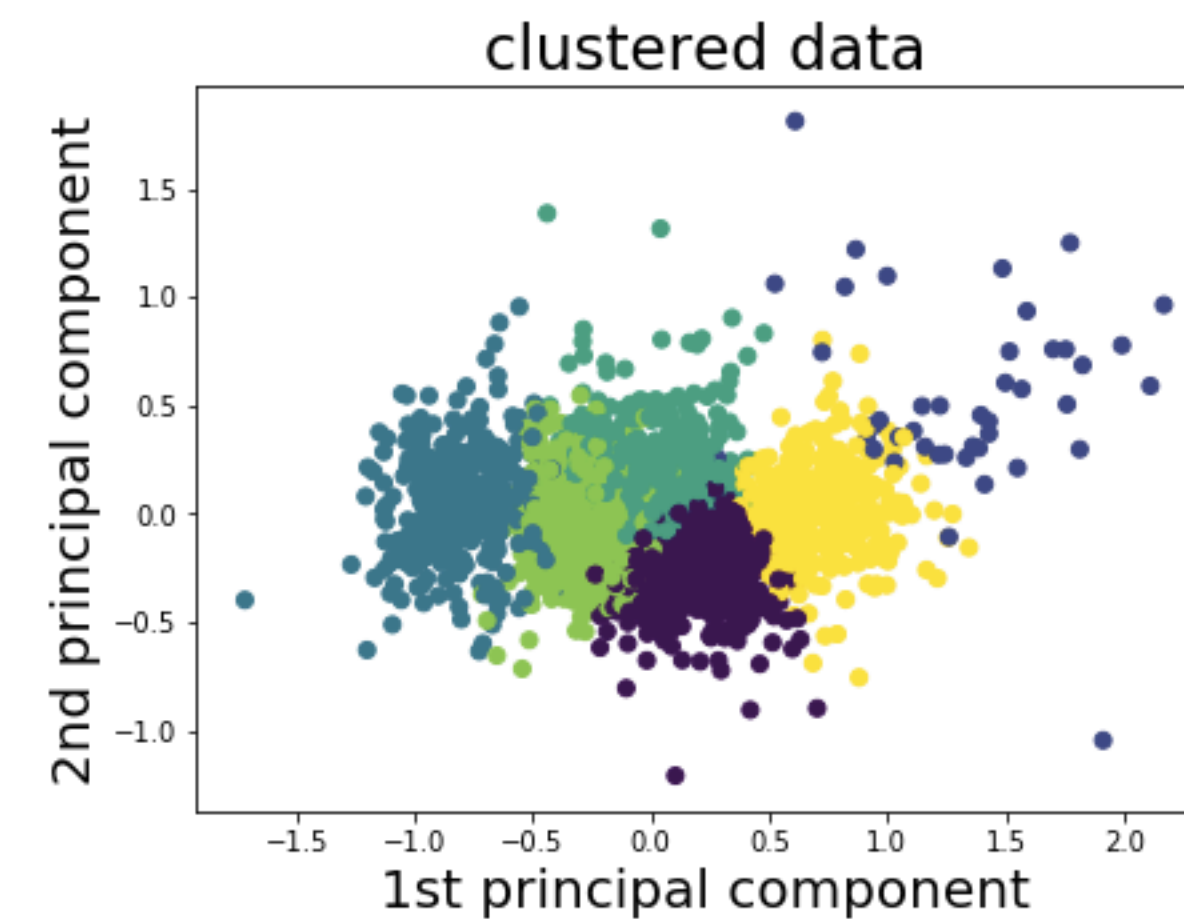
**Simulation & Analysis**

# inspiration & motivation

Epilepsy: find this seizure focus, recording electrodes are inserted into the patients brain with which the neural activity can be monitored in real time



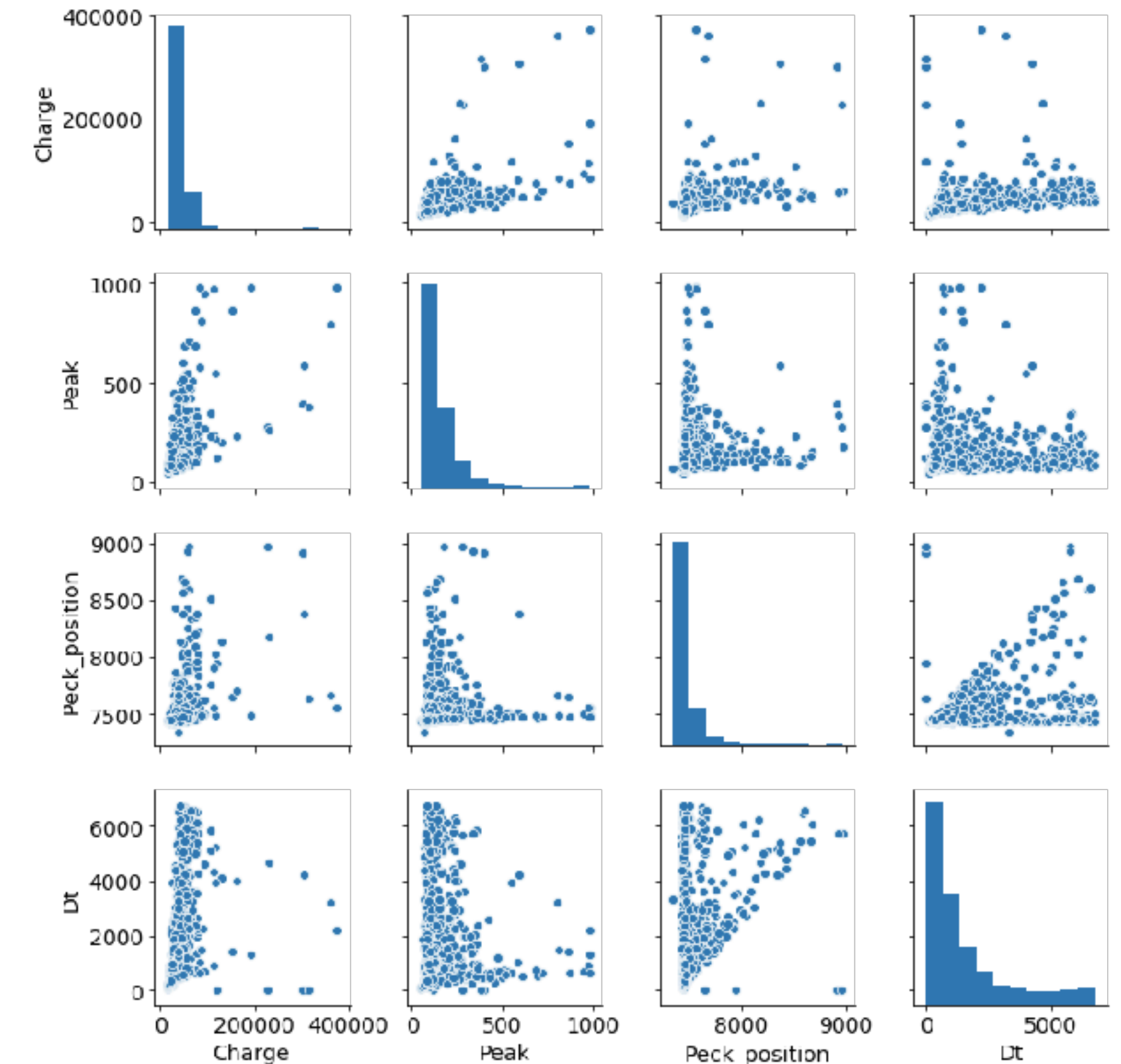
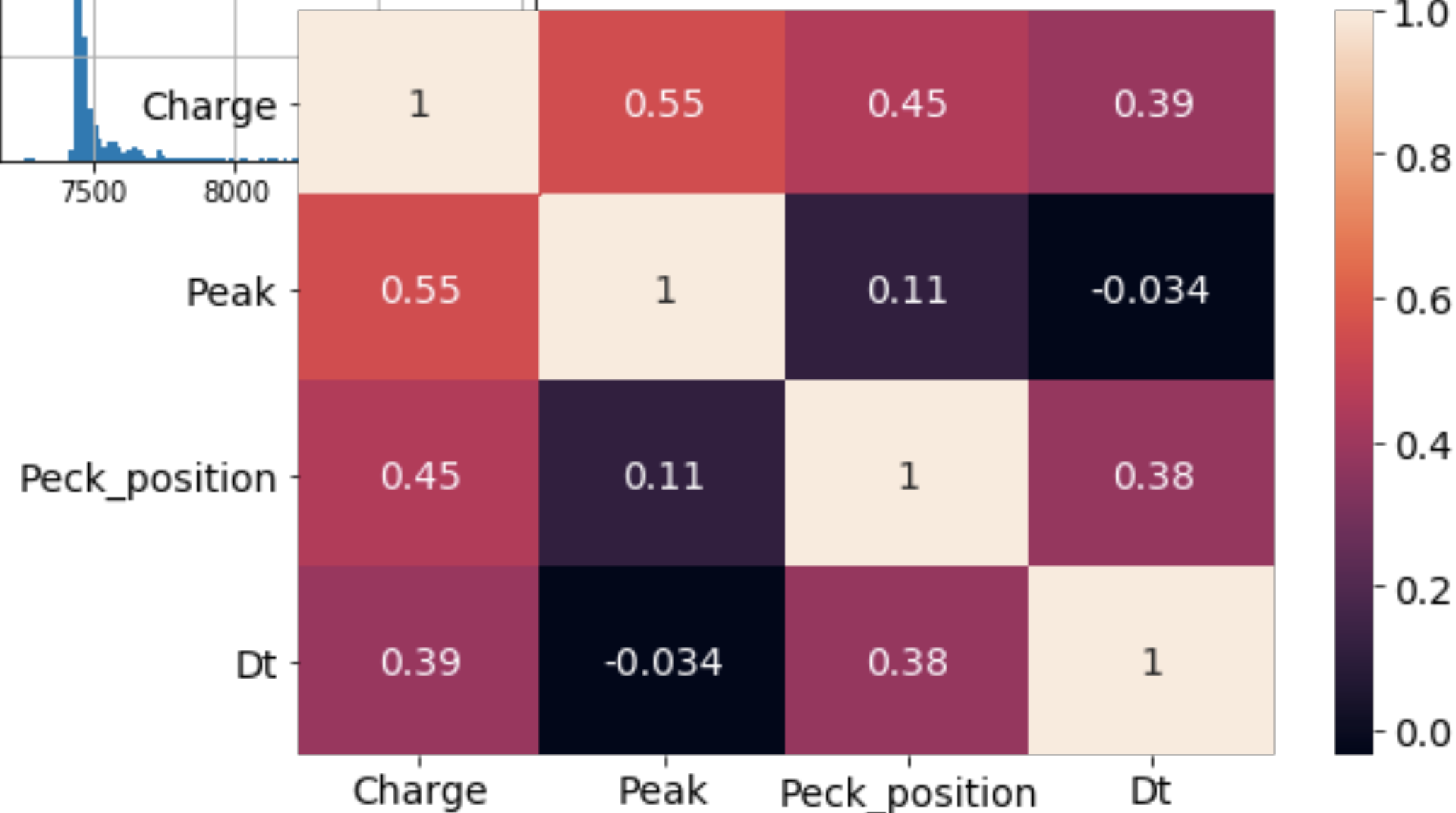
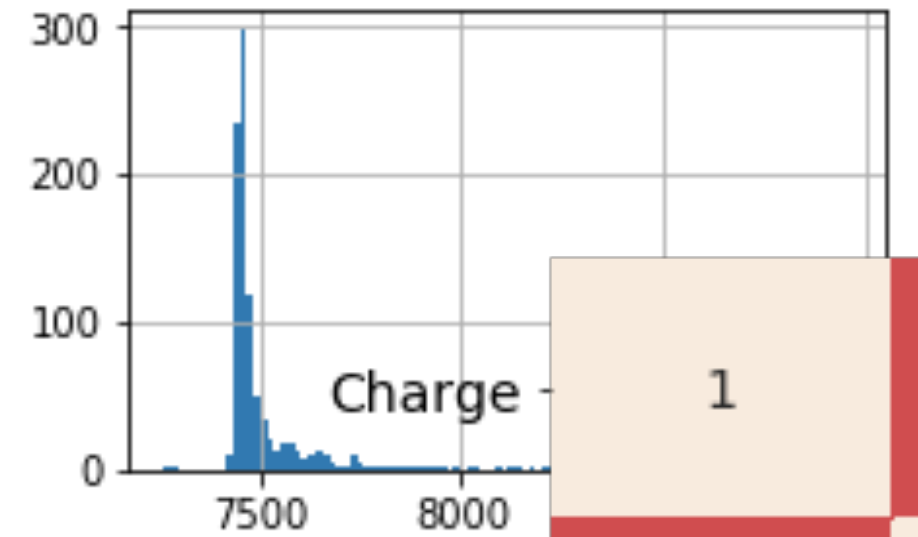
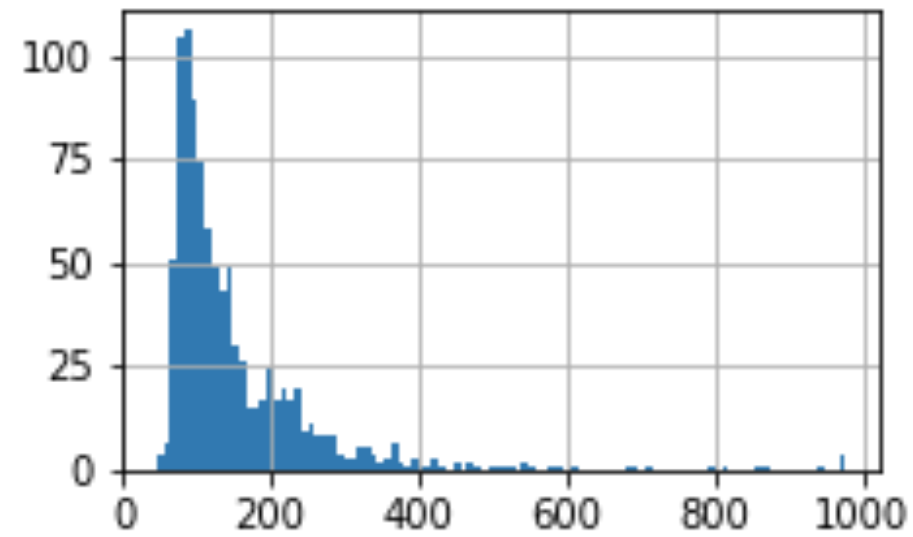
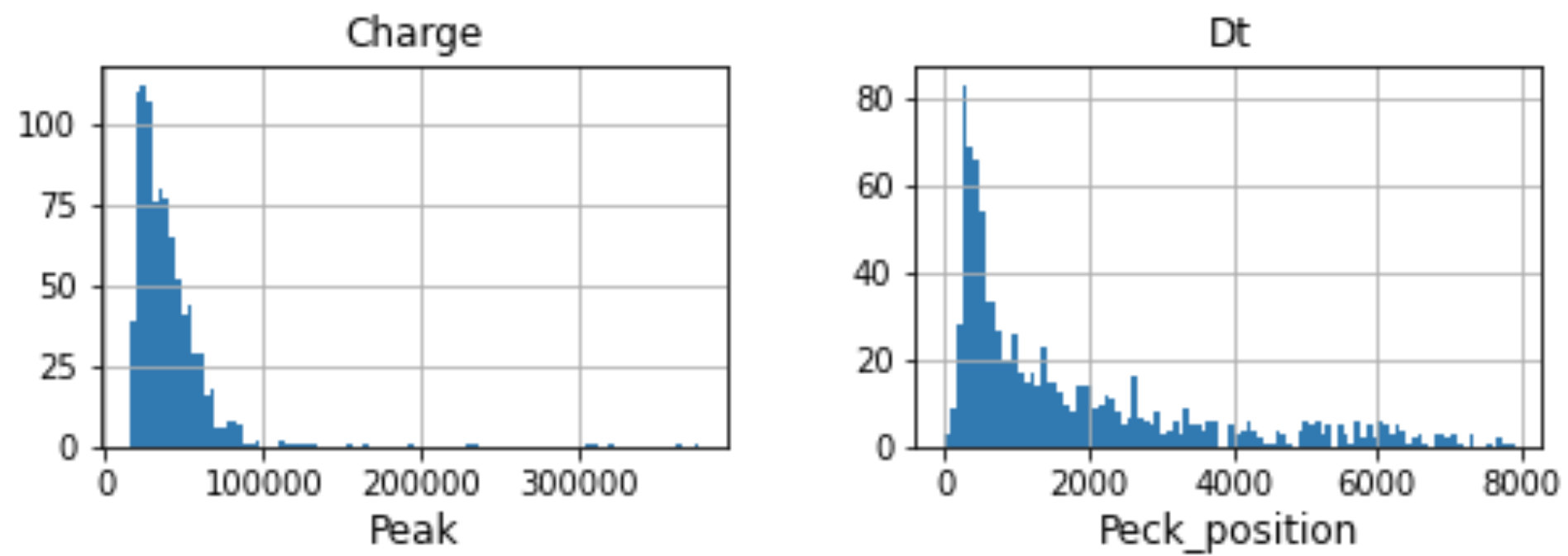
raw data



features analysis

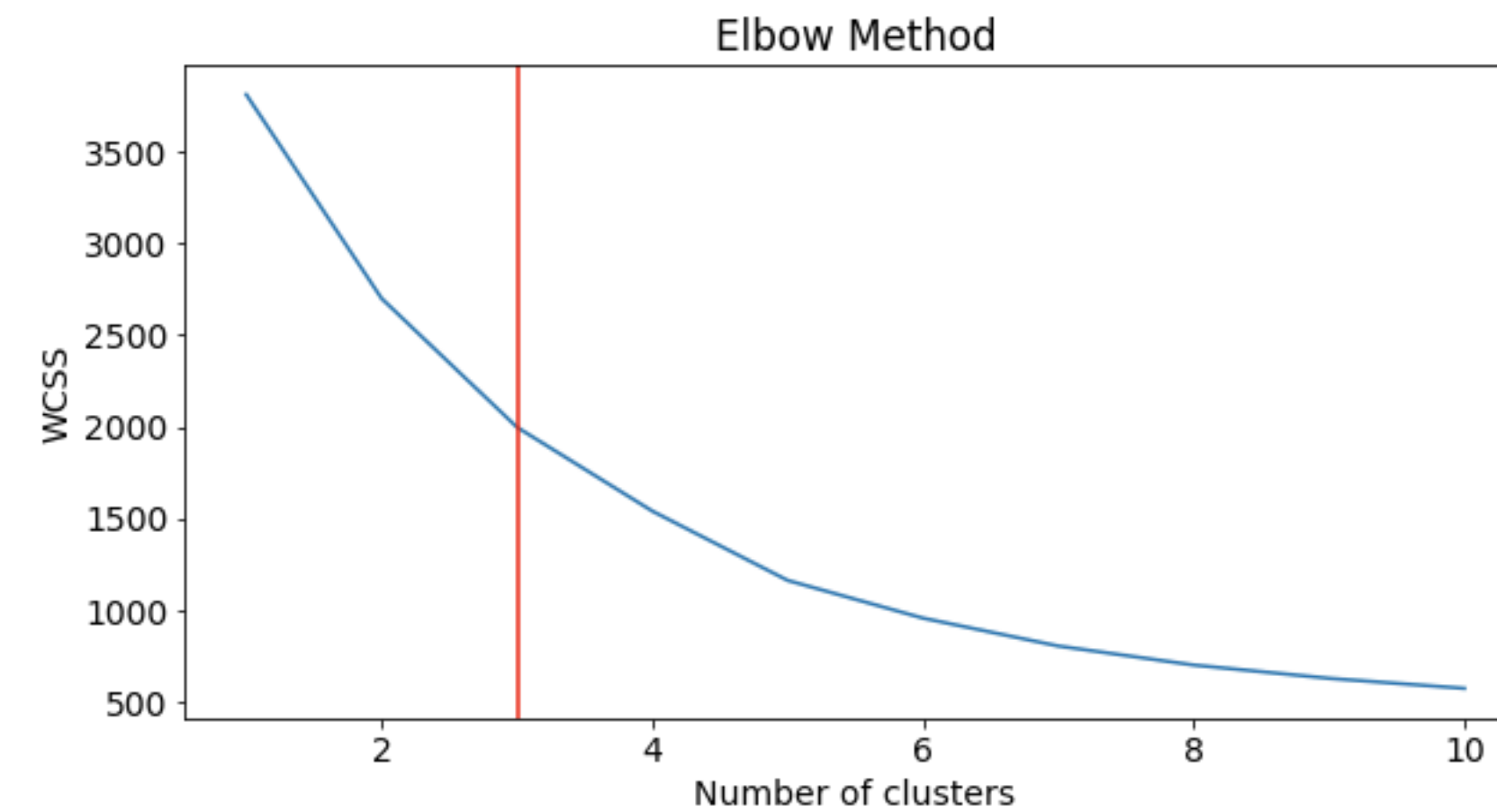
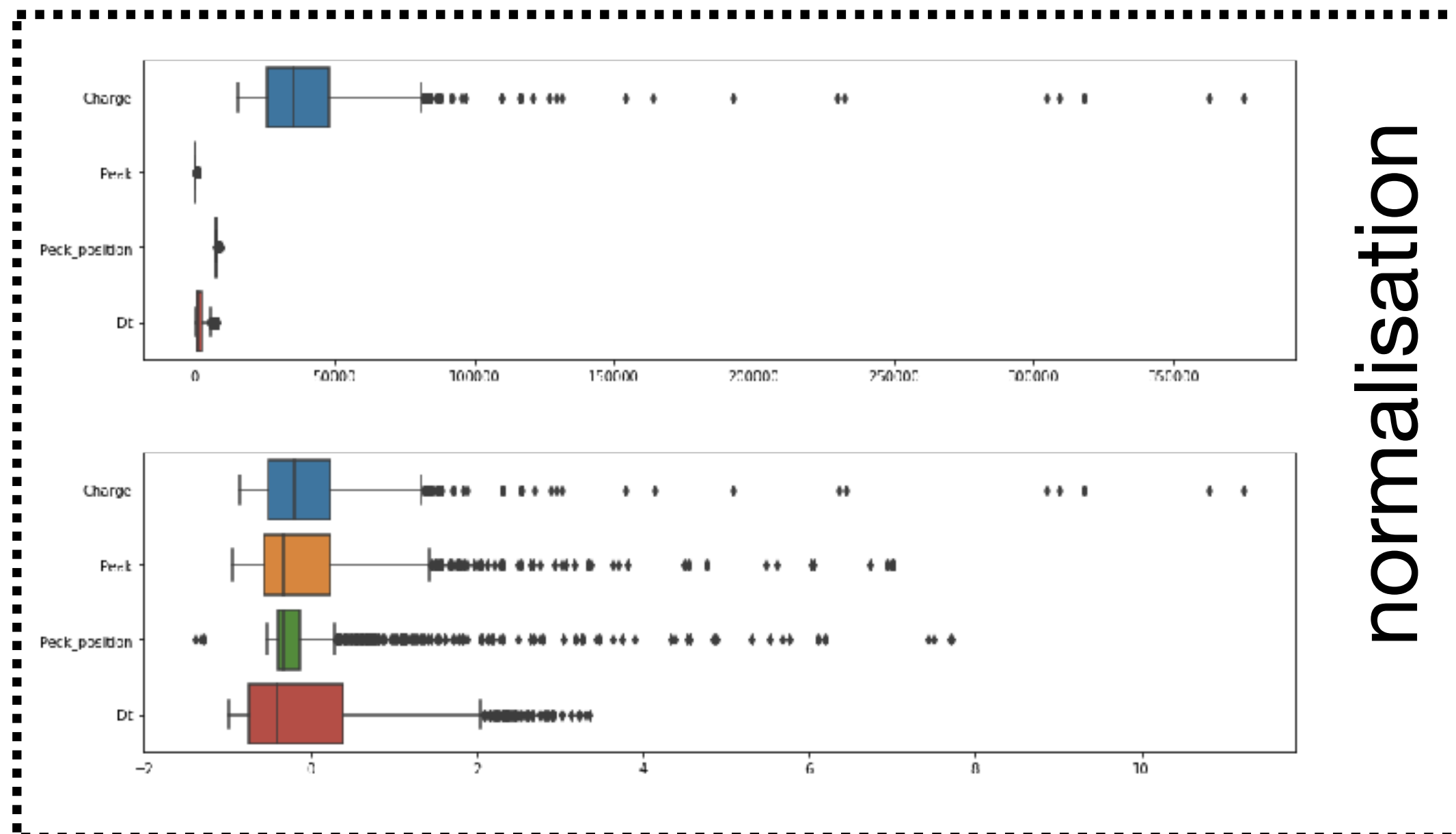
# Waveform & Clustering

features approach: charge, DT, Peak, Peak\_position (RUN2098-AmBe)

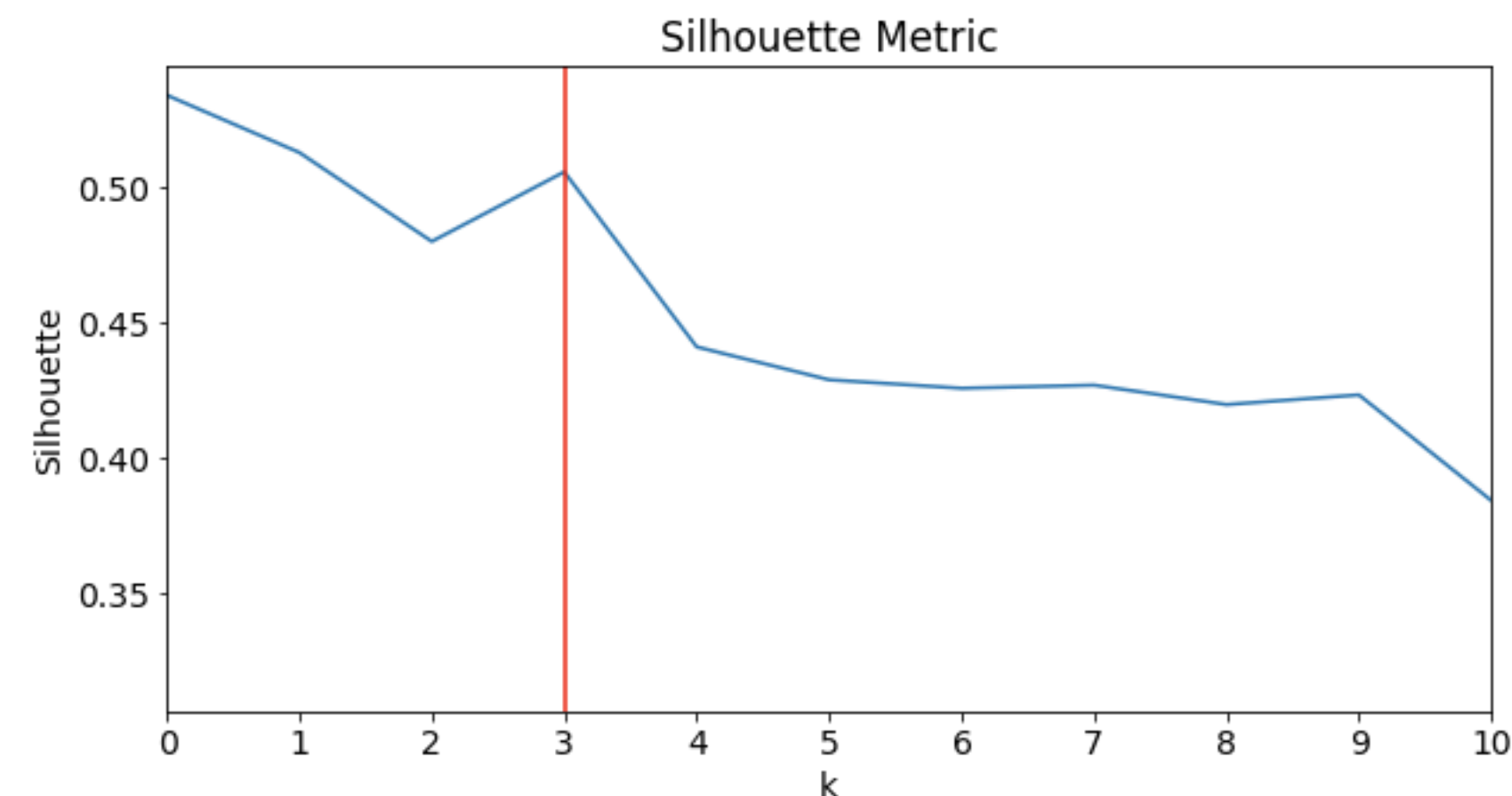


# Waveform & Clustering

features approach: charge, DT, Peak, Peak\_position (RUN2098-AmBe)



the elbow method is a heuristic approach in determining the number of clusters in a data set

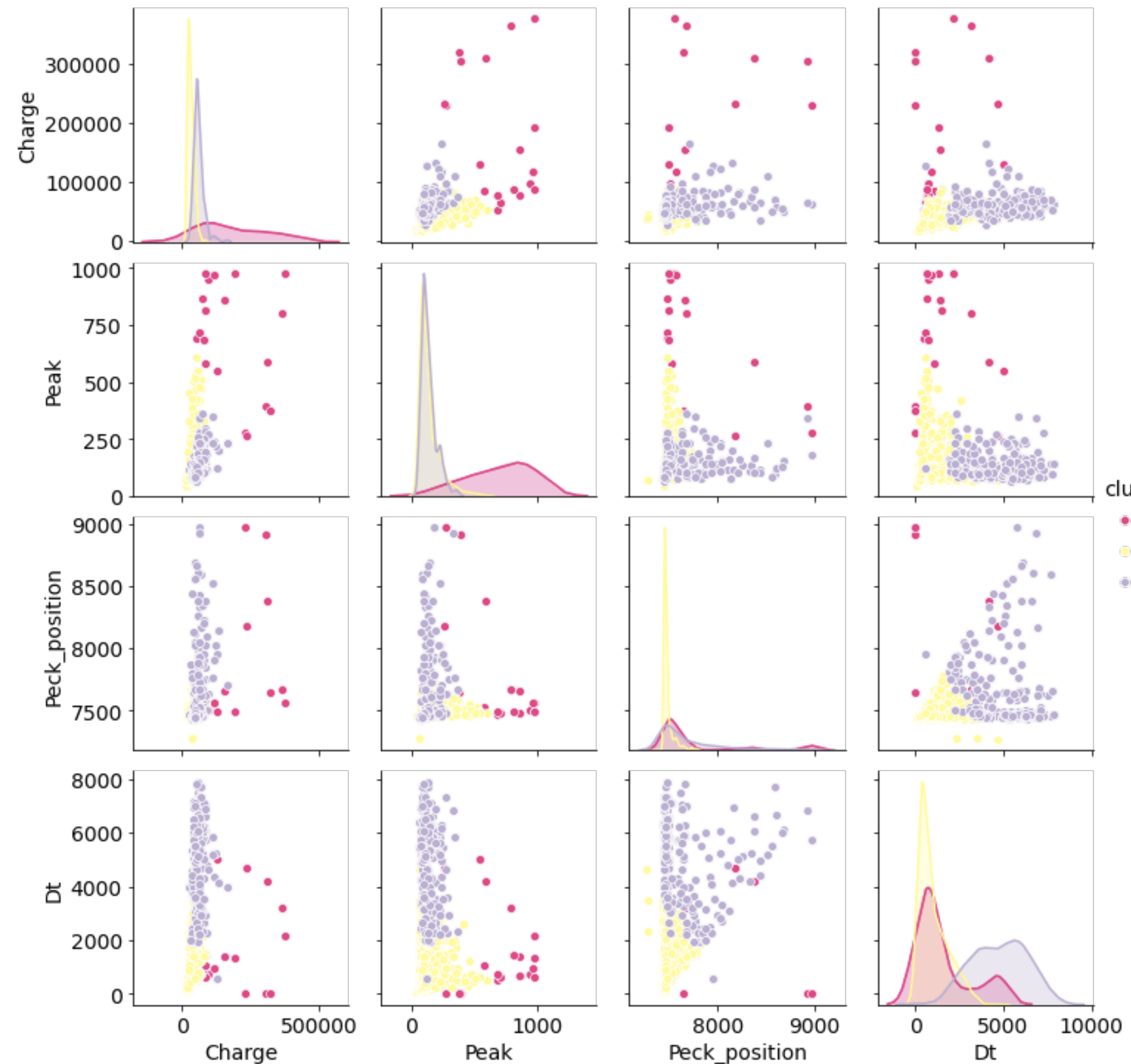


Silhouette: mean intra-cluster distance (a) and the mean nearest-cluster distance (b) for each sample



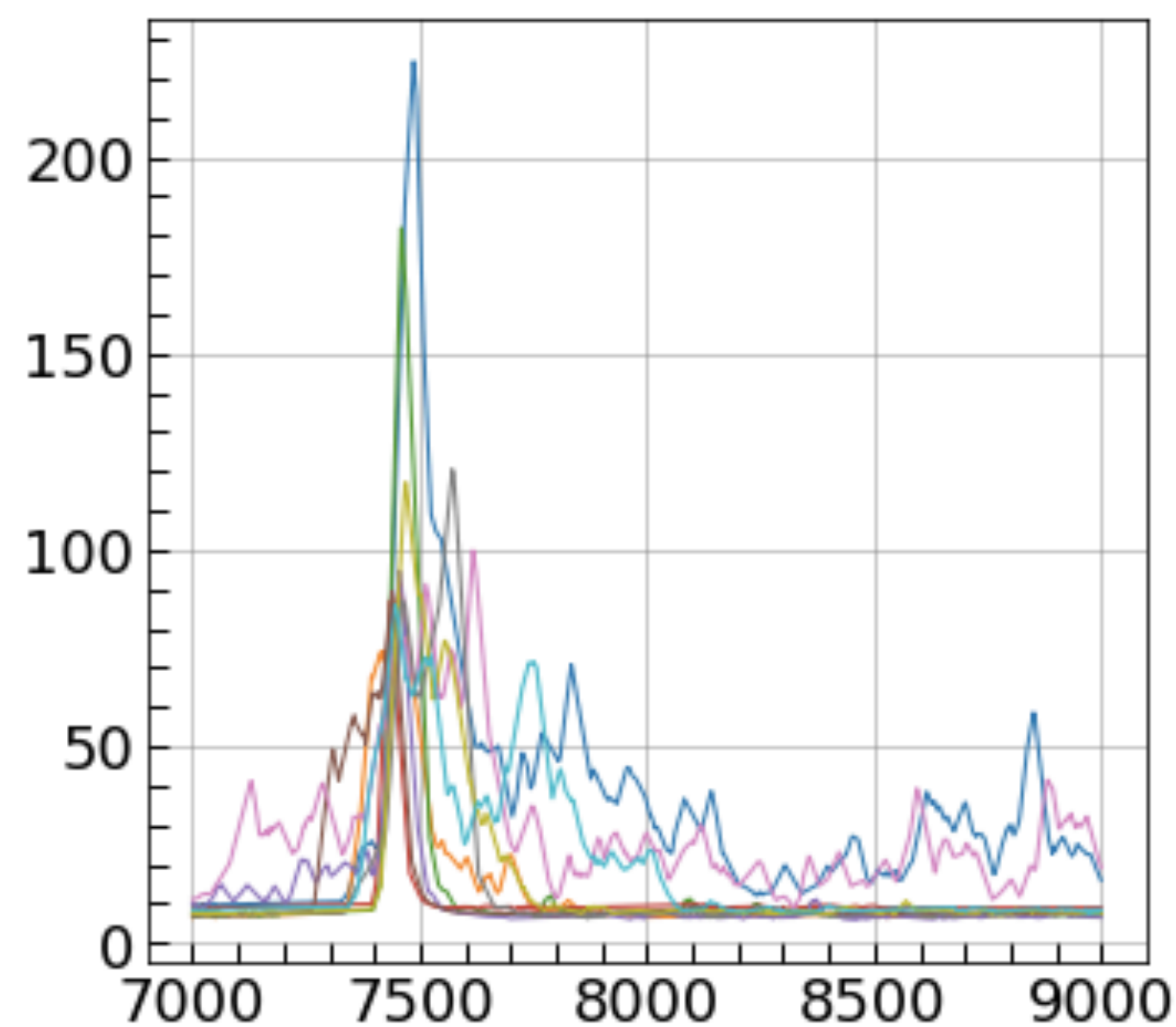
# Waveform & Clustering

features approach: charge, DT, Peak, Peak\_position (RUN2098-AmBe)

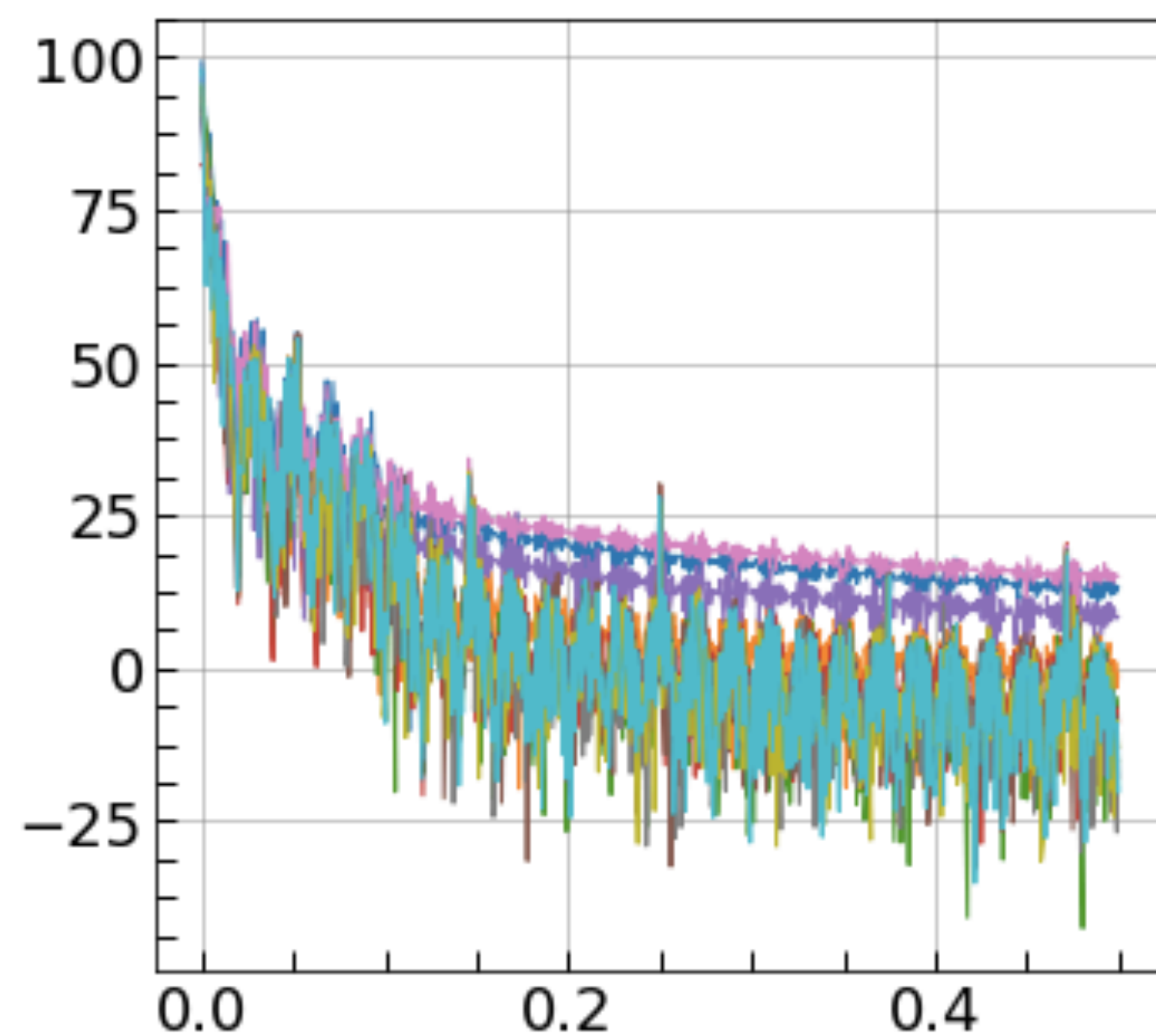


# Waveform

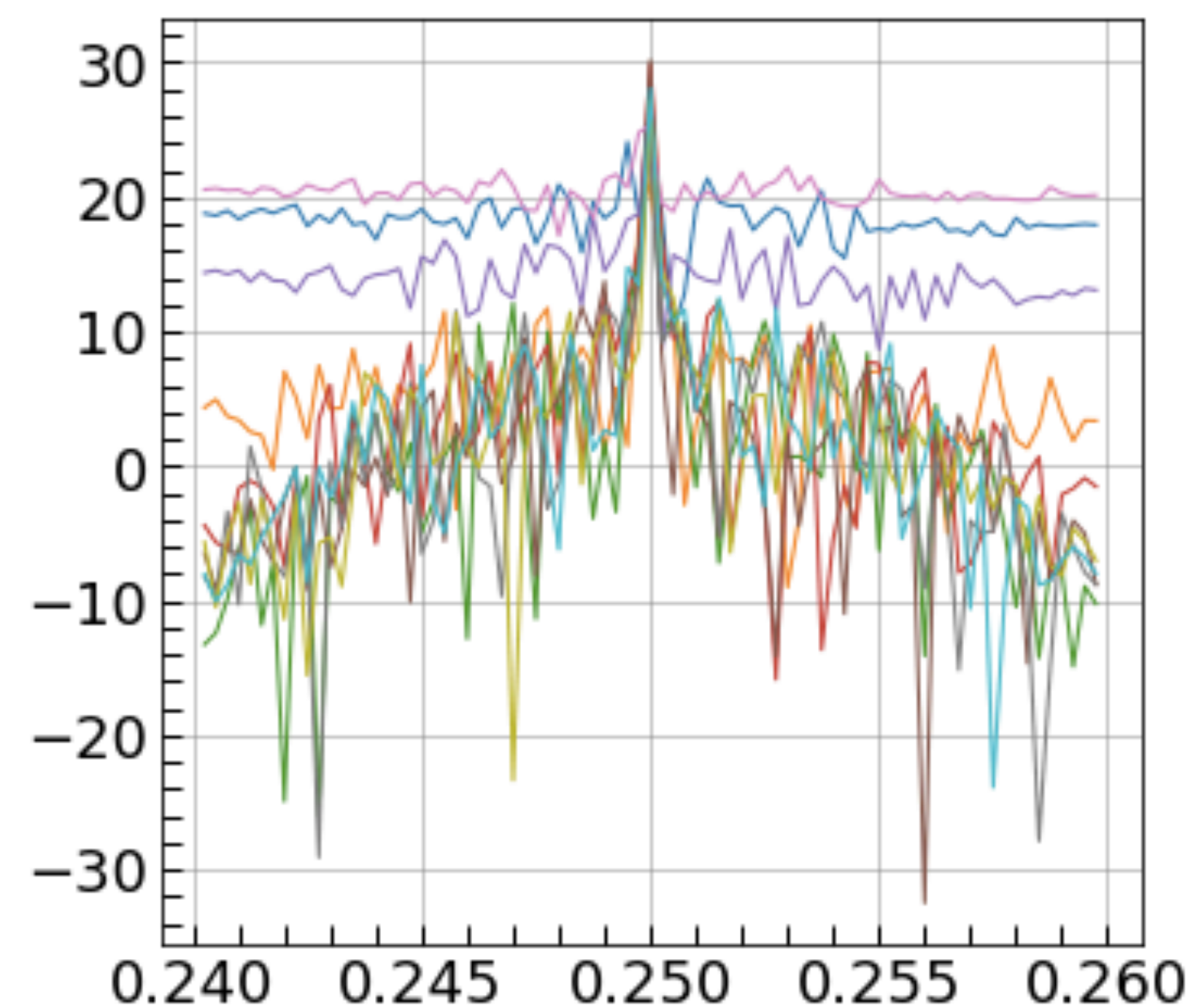
(RUN2098-AmBe) raw data example - 10 waveform



raw data



FFT

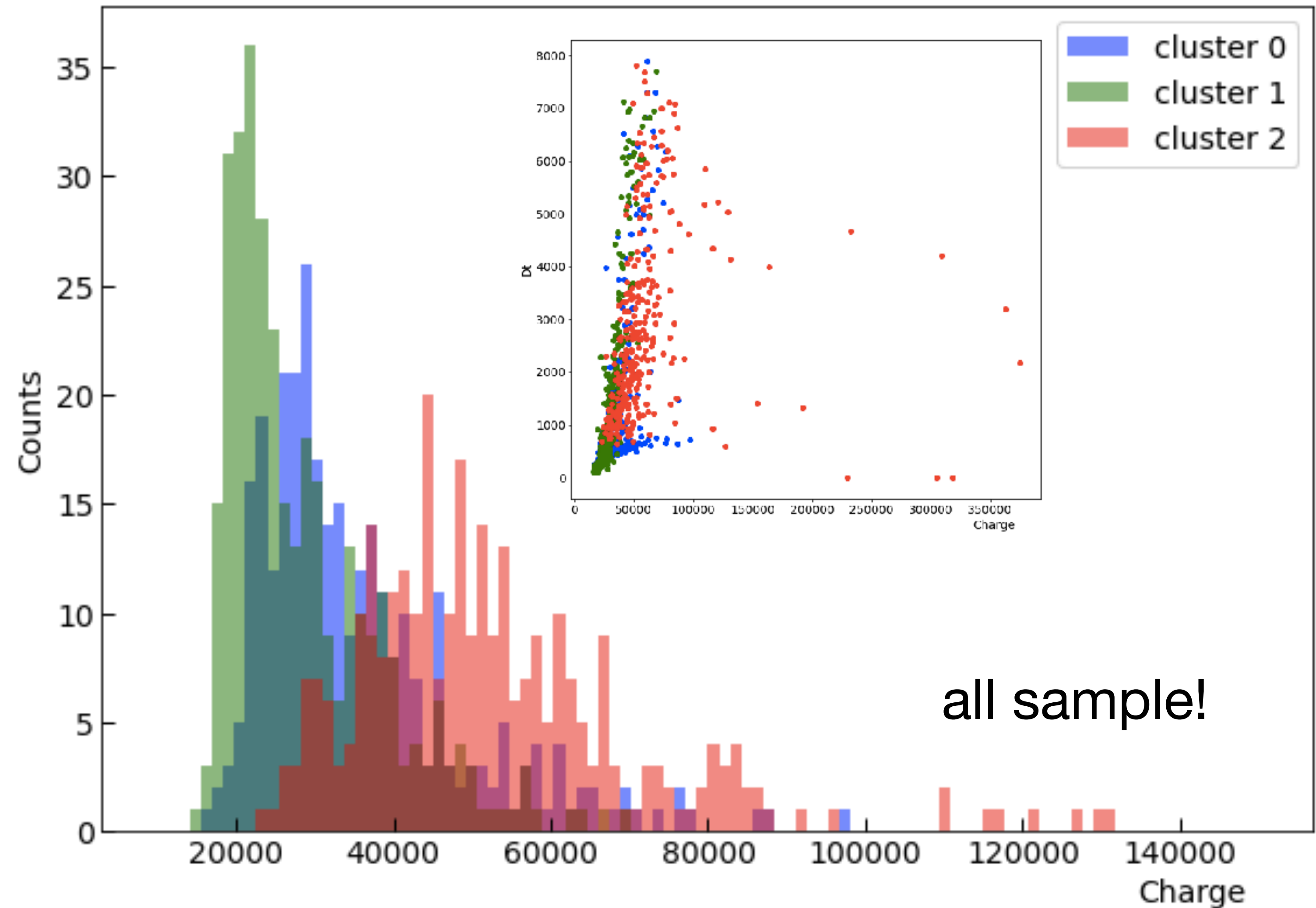
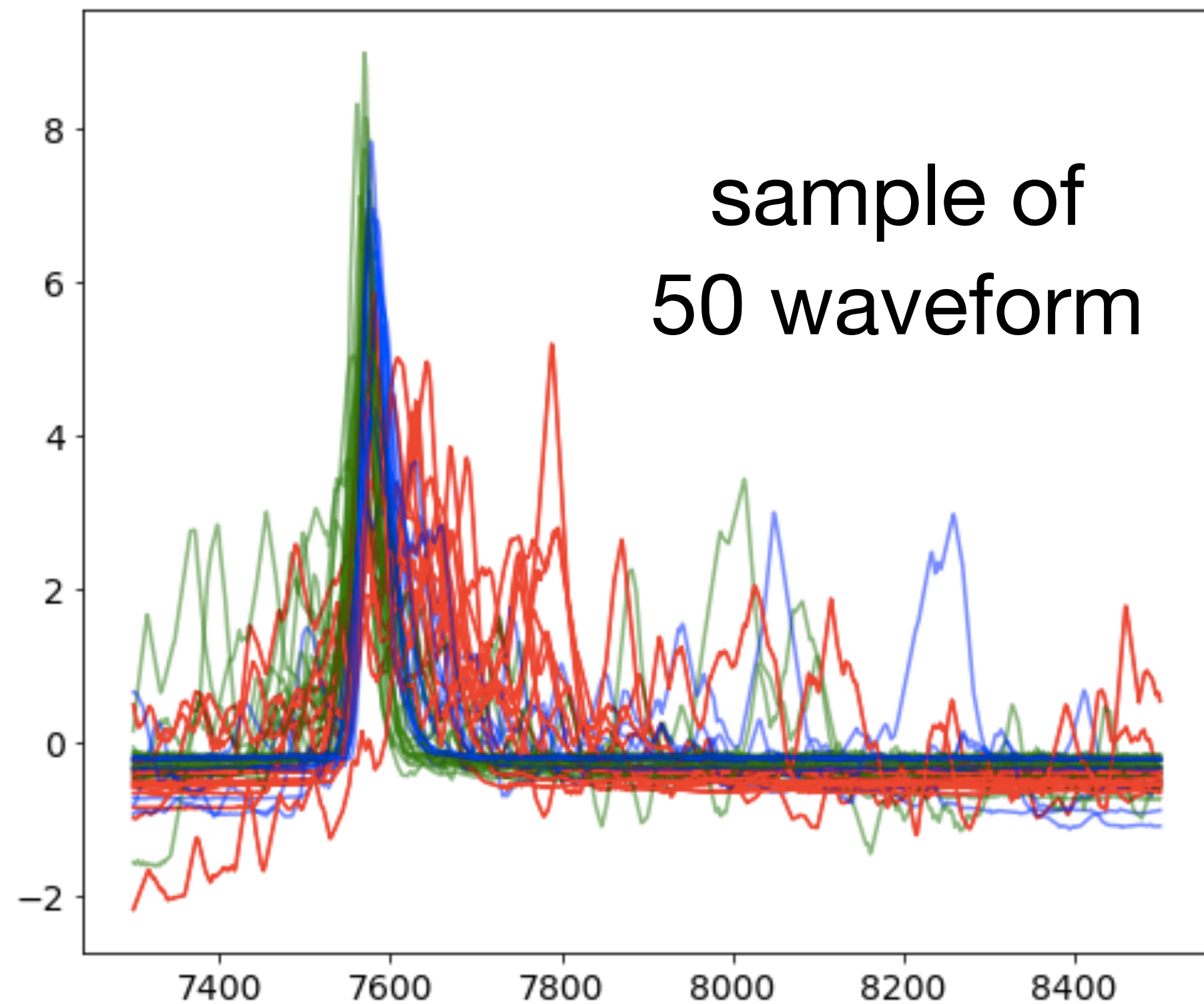


FFT zoom



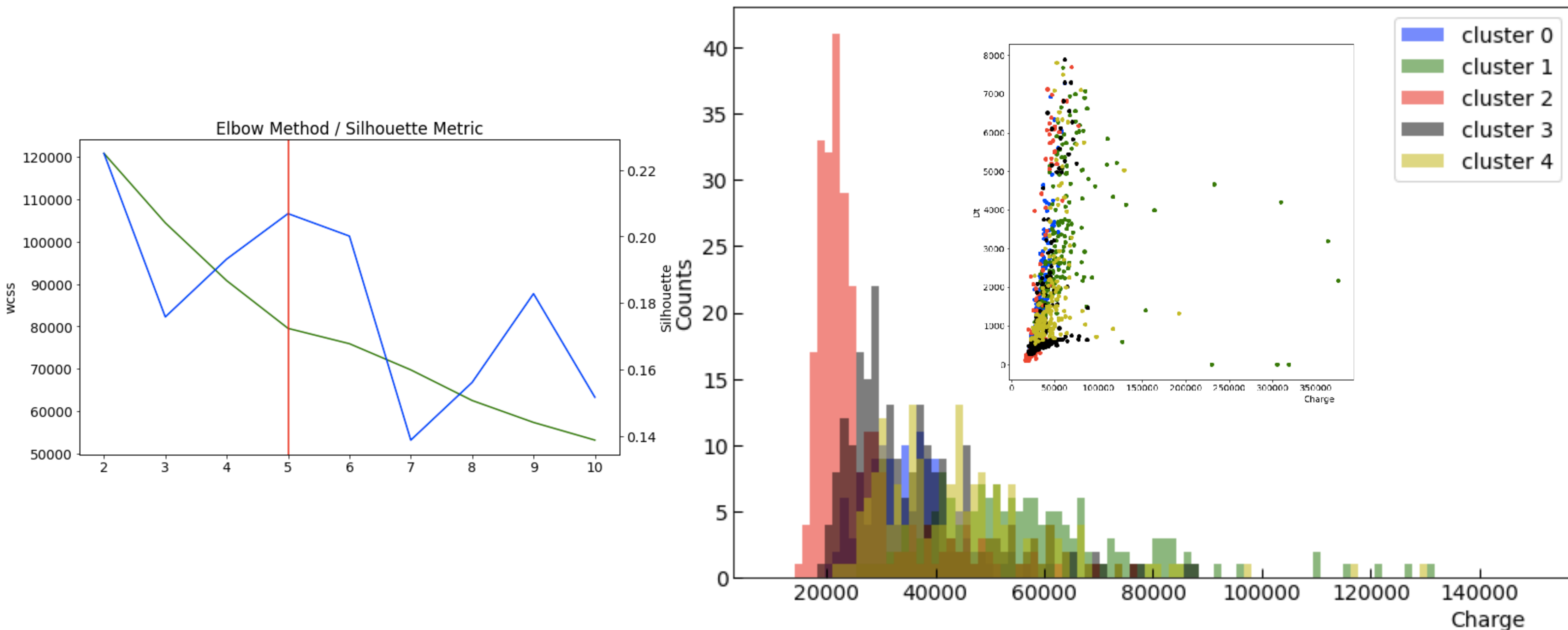
# Waveform

(RUN2098-AmBe) Kmeans clustering in 3 families



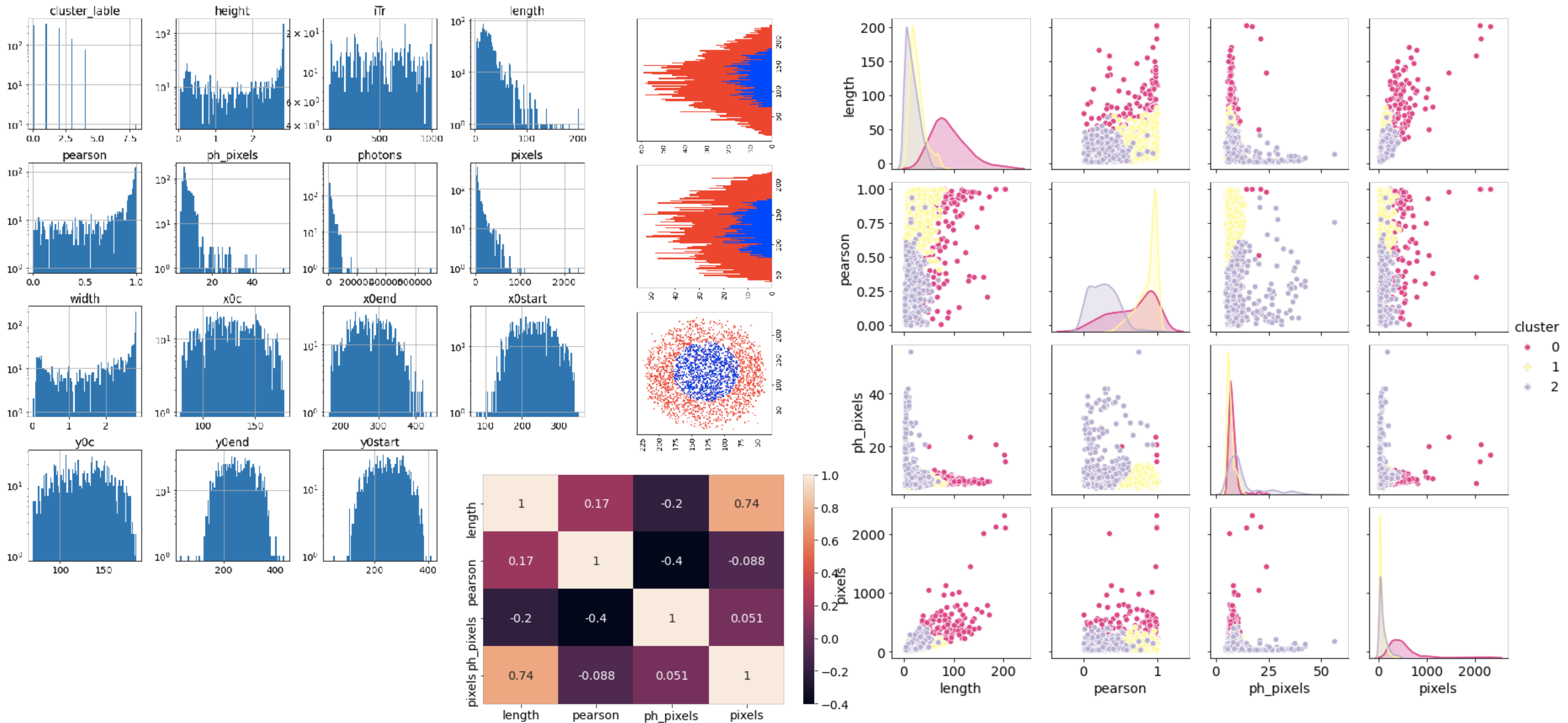
# Waveform

## (RUN2098-AmBe) Kmeans clustering in 5 families

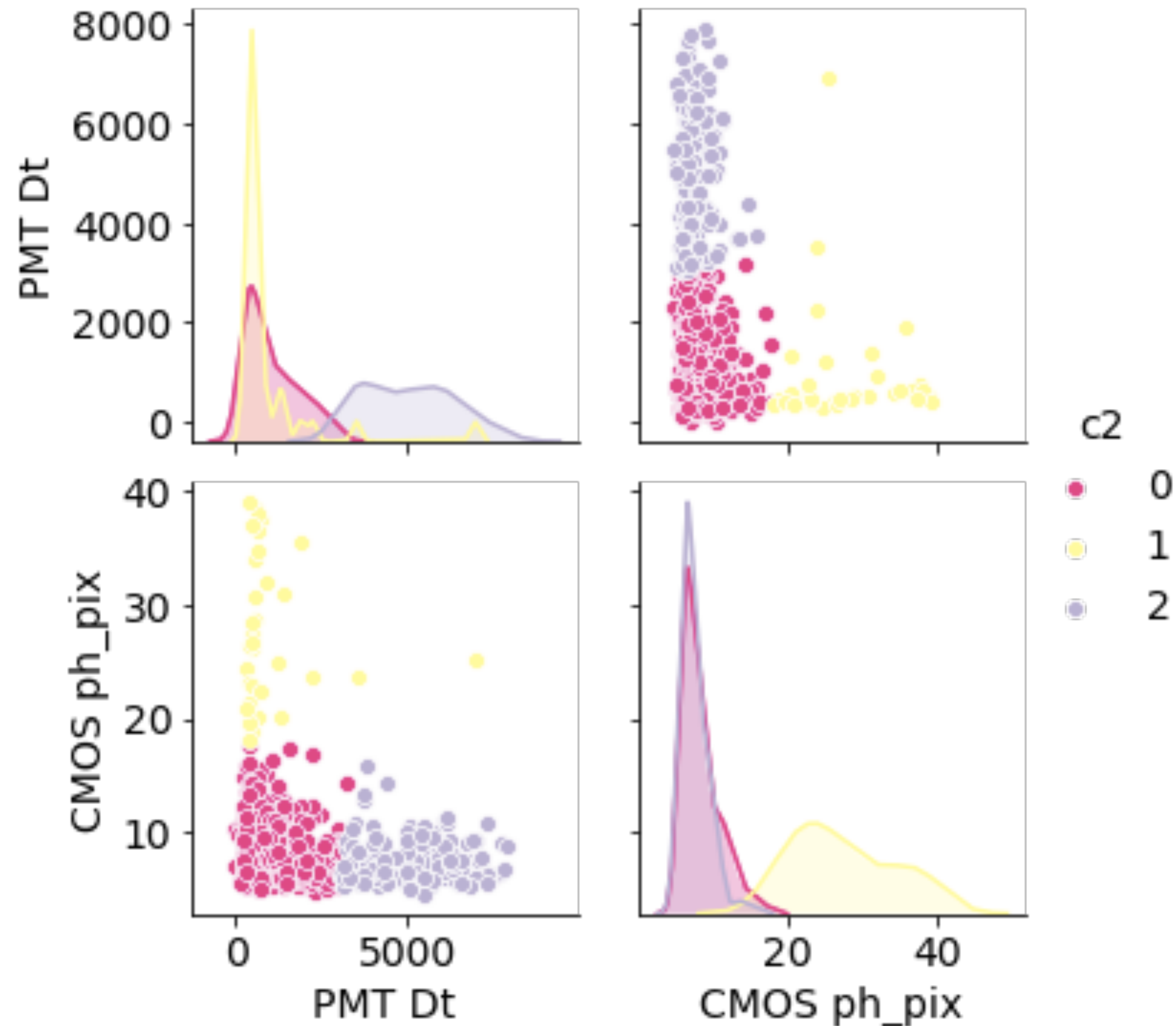




# DBSCAN futures



# Conclusion 1/2

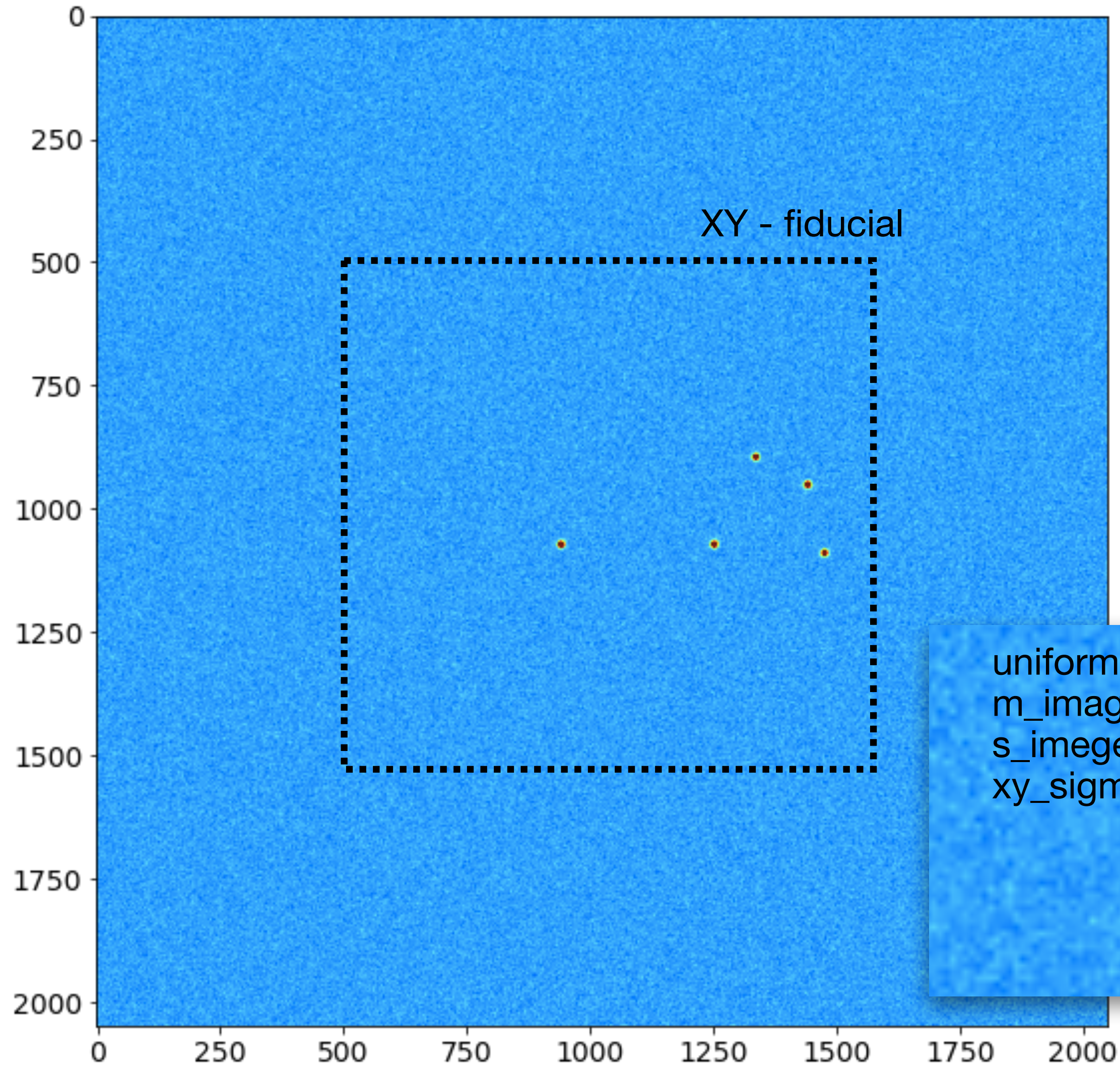


- looks that a clustering algorithms work fine on PMT waveform
- the FFT could also be use to improve features
- what and how to correlate DBSCAN clusters data with PMT sum of many different contribution?

# Simulation



# Fe work around sklearn

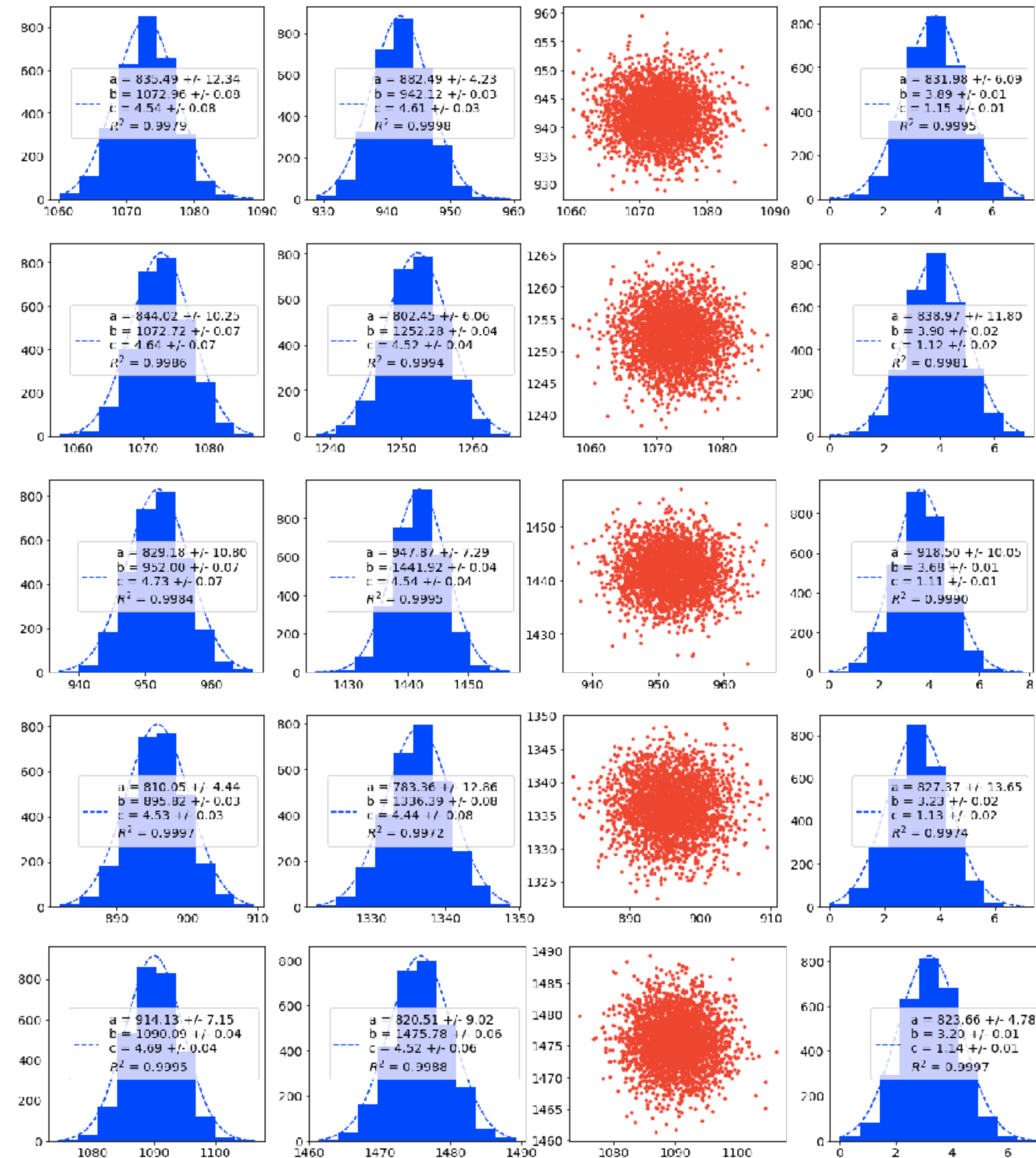


uniform dist. back  
 m\_image = 99.5  
 s\_image = 3.8  
 xy\_sigma = 4.0

XY - norm dist. (Px=Py)

XY - tilt

ph norm dist.

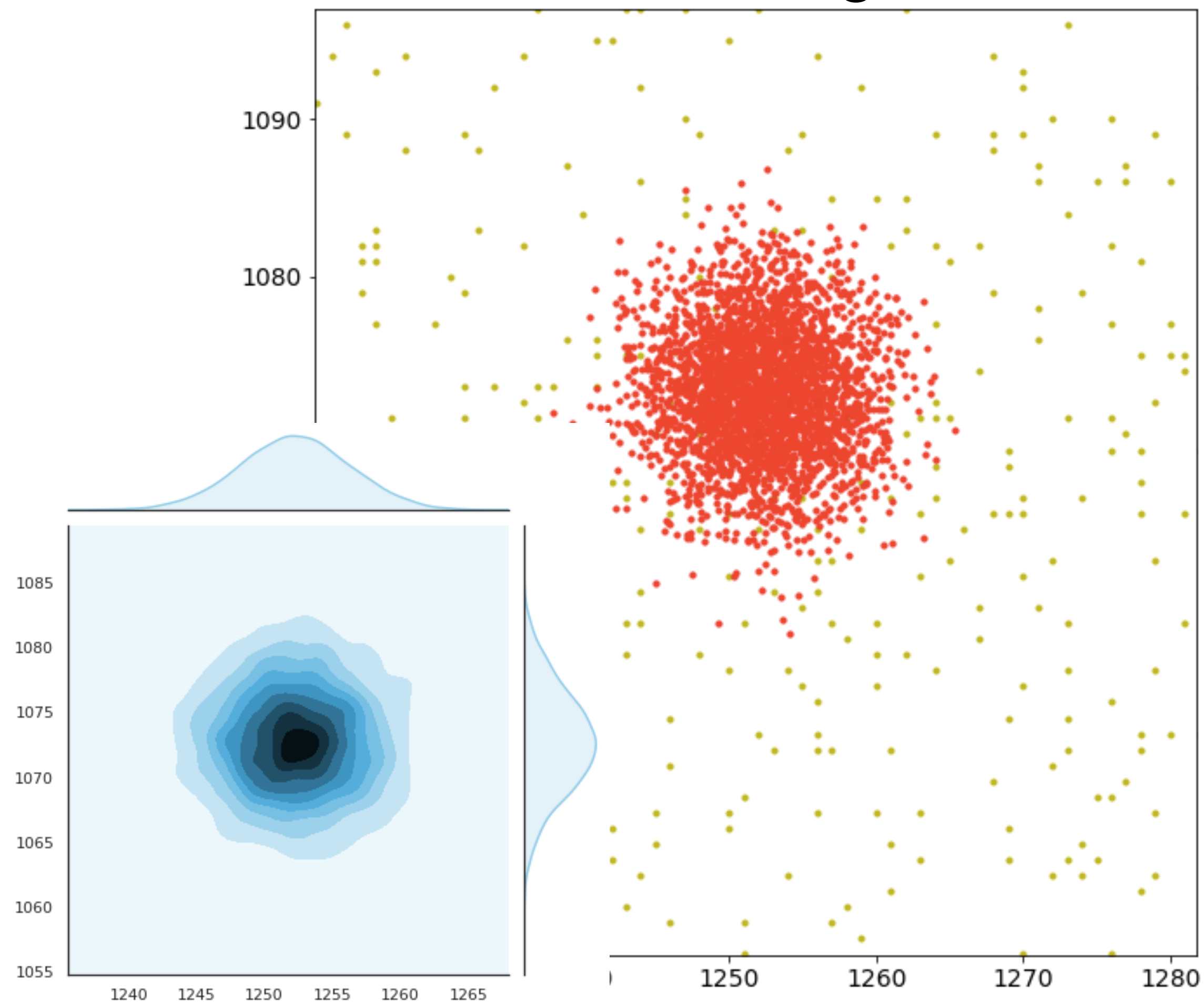




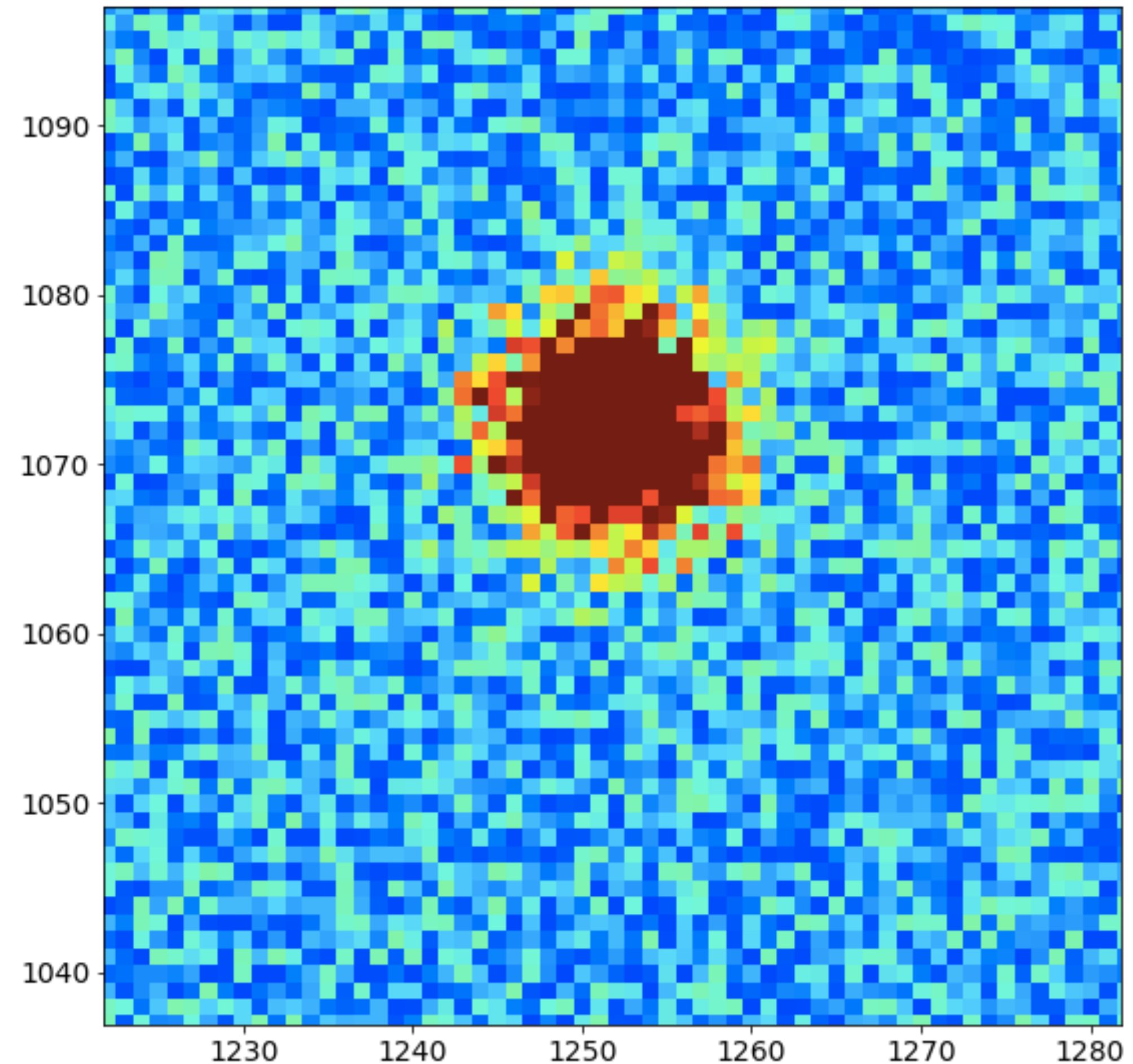
# imaging

5 spot \* 100 image ~ 17 s on colab

noise-m\_image > 3.5



vmin=98,vmax=110



! all float

# DBSCAN

SIM(OUT)

	iTr	image
95	95	[[99.73538564343858, 101.77535204841473, 102.6...
96	96	[[99.84488393284403, 102.52351488762105, 99.76...
97	97	[[102.54013475607381, 101.47048471208953, 102....
98	98	[[99.57021351709082, 103.16033324534018, 100.4...
99	99	[[101.88468871854188, 102.51683772727921, 103....

plk2root (Emanule)

	iTr	photons	blobs_points
495	99.0	2452.577832	0.0
496	99.0	2604.912876	1.0
497	99.0	2660.719968	2.0
498	99.0	2529.341997	3.0
499	99.0	2724.850270	4.0

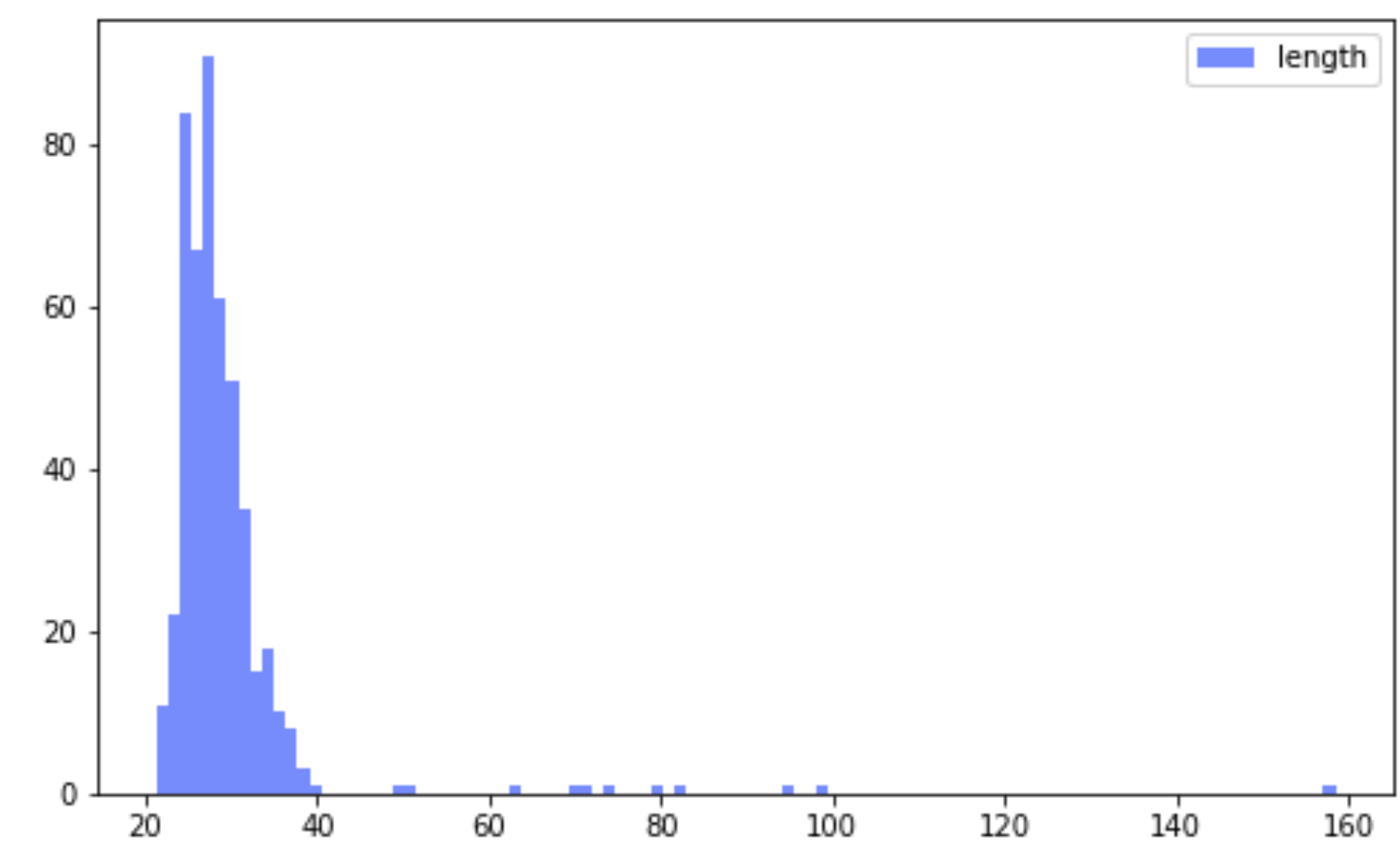
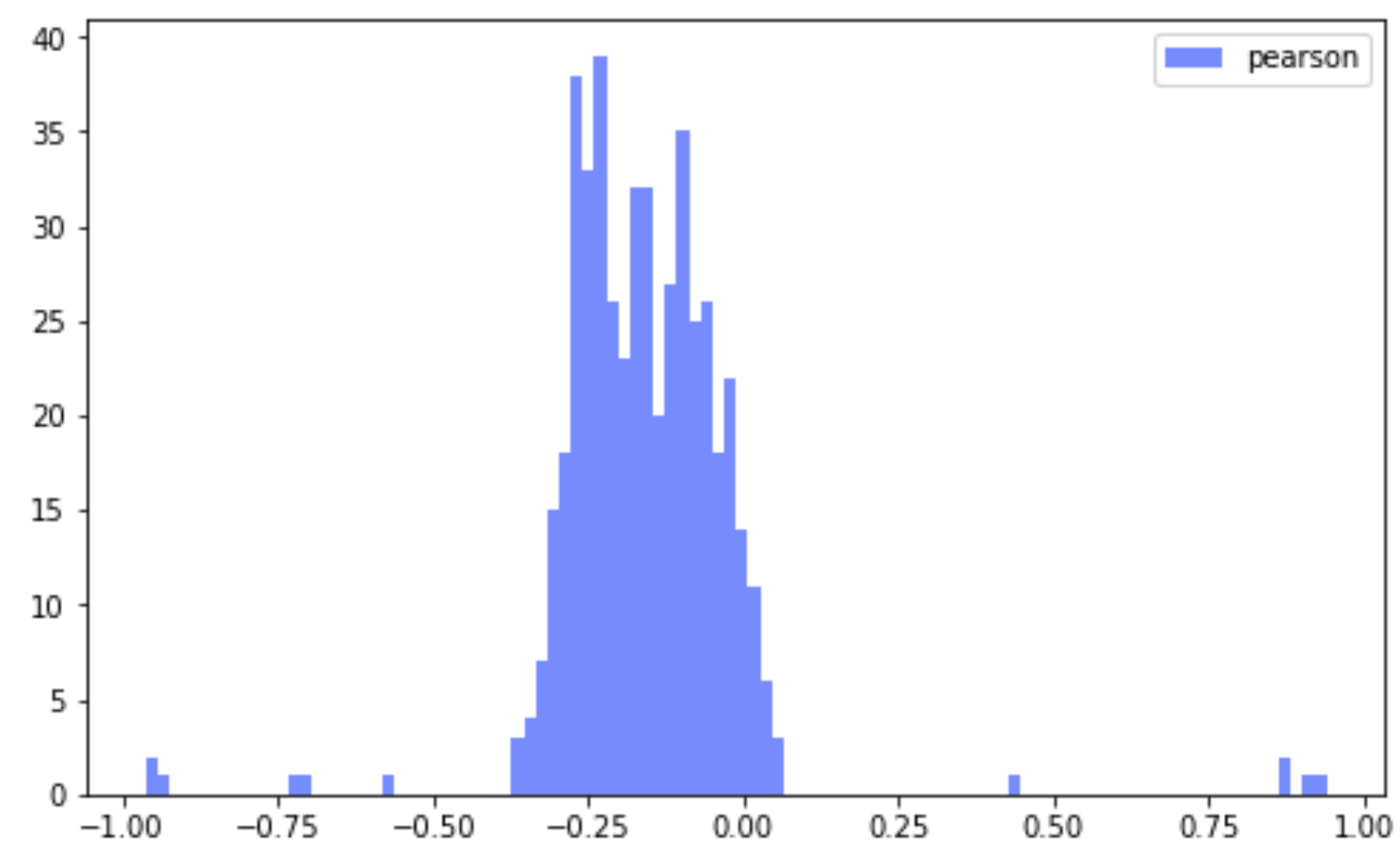
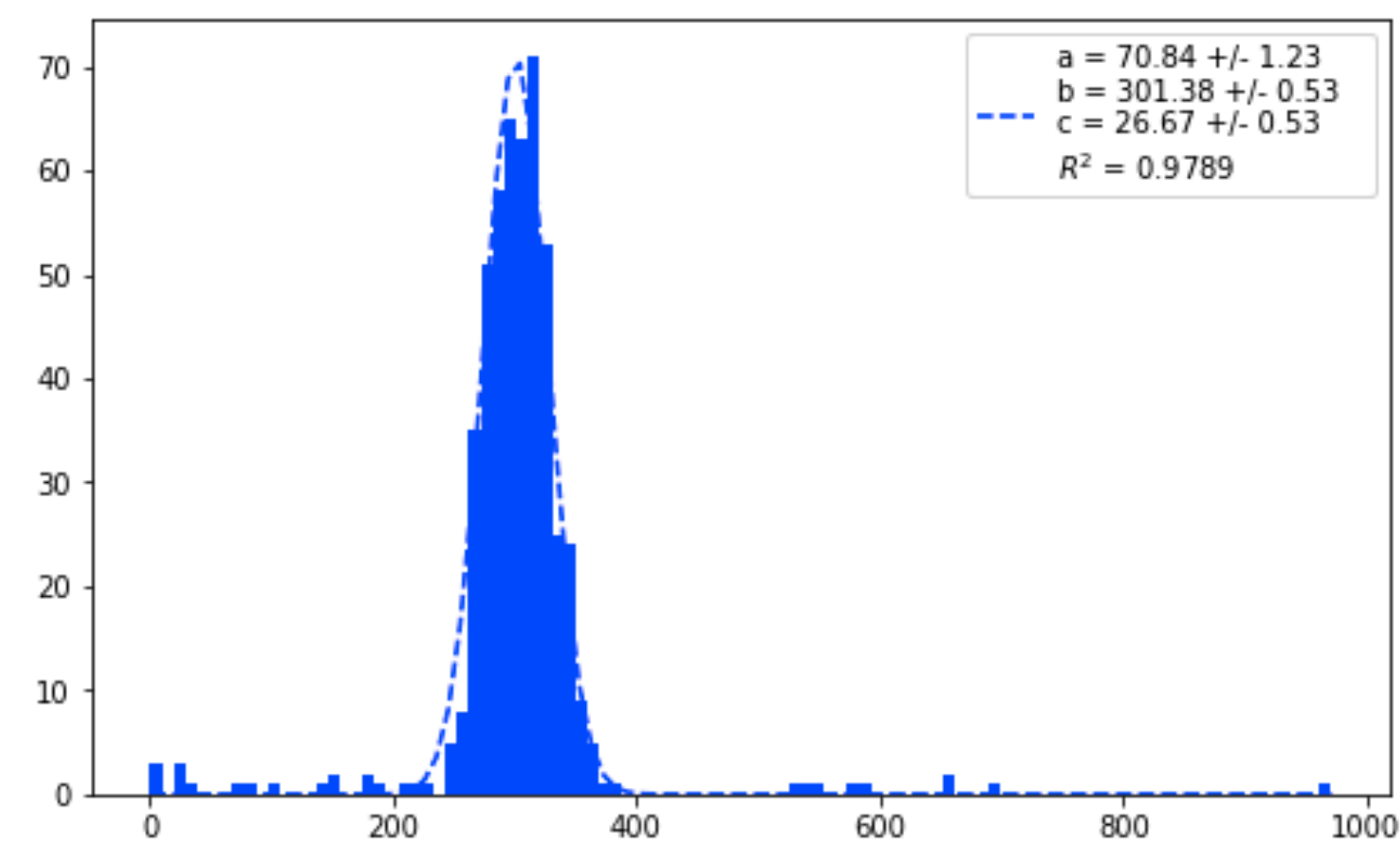
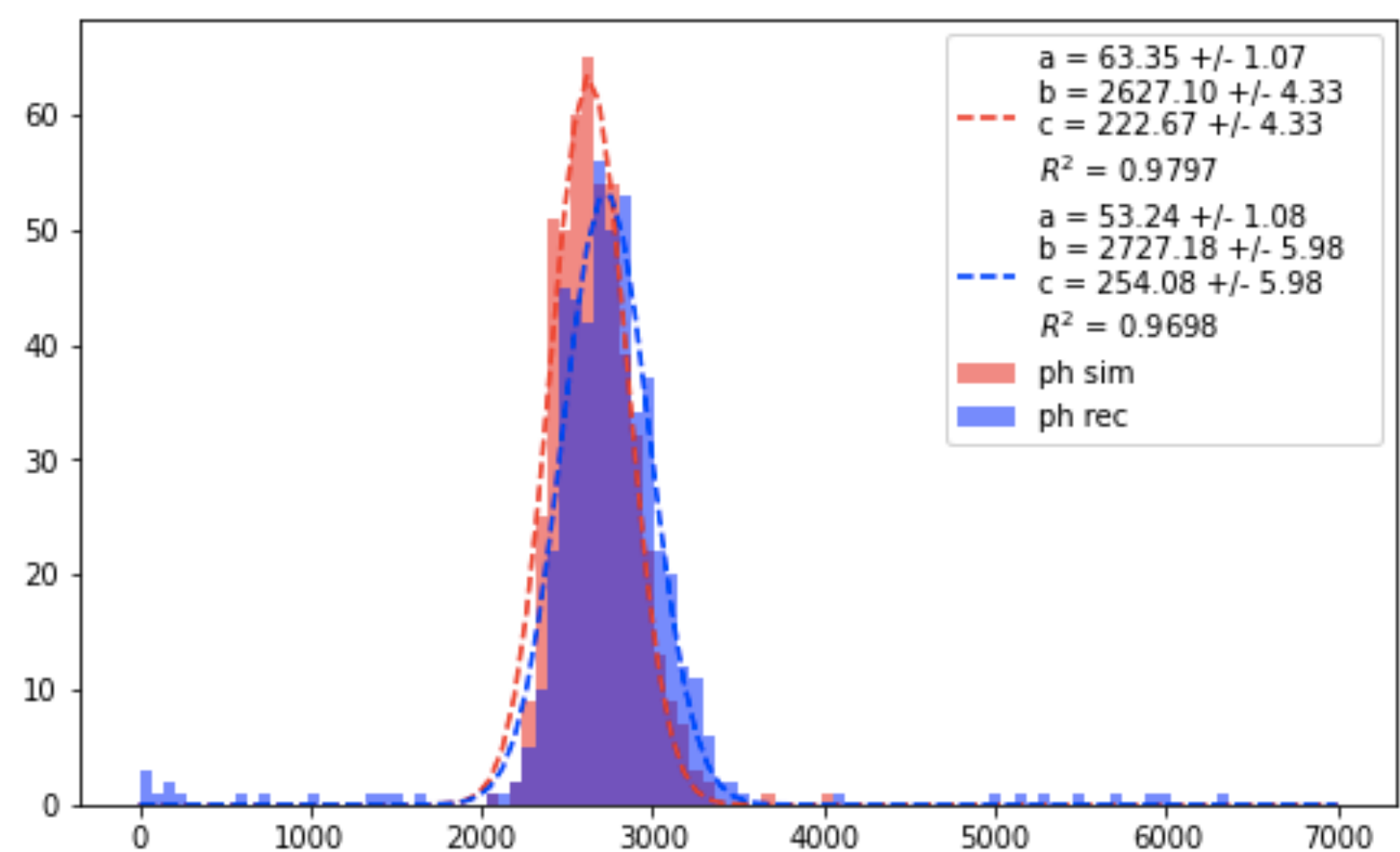
plk2CYGNODB (??)

DBSCAN  
(OUT)

	iTr	cluster_lable	pixels	photons	ph_pixels	x0start	y0start	x0end	y0end	width	height	pearson	length	x0c	y0c
481	98.0	4.0	325.0	2523.933871	7.765950	1666.0	712.0	1698.0	706.0	1.584169	2.343162	-0.372602	32.557641	1682.0	709.0
482	99.0	0.0	491.0	5305.053363	10.804589	576.0	1248.0	628.0	1282.0	2.780871	0.516483	0.933311	62.128898	602.0	1265.0
483	99.0	1.0	233.0	2668.744005	11.453837	780.0	571.0	804.0	582.0	2.189313	1.790784	0.198273	26.400758	792.0	576.5
484	99.0	2.0	245.0	2776.820931	11.333963	903.0	1014.0	925.0	1020.0	2.146689	1.841665	0.152068	22.803509	914.0	1017.0
485	99.0	3.0	227.0	2462.578364	10.848363	1184.0	1035.0	1207.0	1042.0	2.177378	1.805276	0.185244	24.041631	1195.5	1038.5



# DBSCAN (2D)



# Conclusion 2/2

## simulation

- round numbers, realistic ph distribution
- set correct simulation parameter
- ! COMS noise (Telegraph)/radioactivity and cosmic
- produce large statics (parallelise production/analysis)
- select a list of interesting input/output