Reimagining the noise reduction algorithm

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Summary

- Motivation;
- Thought solution;
- The new algorithm;
- Results;
- Conclusion



Motivation



- As commented last meeting, pre-processing could be a more cumbersome process than clusterization itself;
- We've found that the noise reduction algorithm was the bottleneck for full resolution images (45~60s in google colab);
 - This part is important because it reduces the number of elements to be sent to the clustering algorithm;
- Then we've searched for faster methods applied to noise reduction.

Thought solution



- The original process is done through a recursive removal of noise pixels, being necessary to sweep every pixel while removing noise.
- A similar process could be done with a convolution of the whole image with a unit kernel of size 3.

The new algorithm



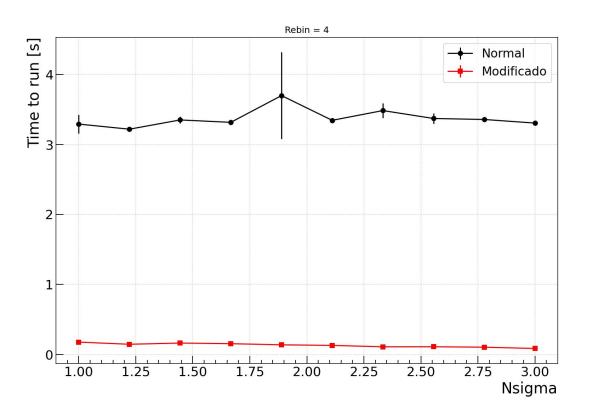
- As the Noise_reductor is a recursive process and the convolution is done for the whole image, to get a similar result we should filter more than once;
- We used nsigma as a parameter to set the number of recursions;
- All tests were done for the rebined image, as both filters were calibrated for it.

The new algorithm



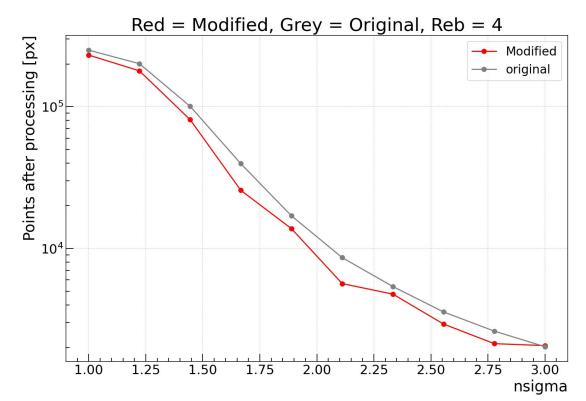
```
1 noise_reductor.py +
    def noisereductor_mod(edges,min_energy=0.35,sigma=3):
        tpx = 10
        for x in range(int(10 - 1.8*sigma)):
           mpx = uniform_filter(edges,size=3,mode='constant')
           mpx = (mpx*9.-edges)/8.
            edges[np.abs(edges - mpx) > tpx] = mpx[np.abs(edges - mpx) > tpx]
           mpx = uniform_filter(edges,size=3,mode='constant')
           mpx = (mpx \star 9.-edges)/8.
           neighbours = np.copy(edges)
           neighbours[neighbours>0] = 1
            neighbours = uniform filter(neighbours.size=3,mode='constant')*9
            edges[neighbours < x] = 0</pre>
            edges[mpx < (min_energy + x/70)] = 0
17
        return edges
        noise_reductor.py + <st/noise_reductor.py 100% 17:17</pre>
                                                                          [python]
NORMAL
```





As time was our biggest concern, our first look will be on runtime of each process.





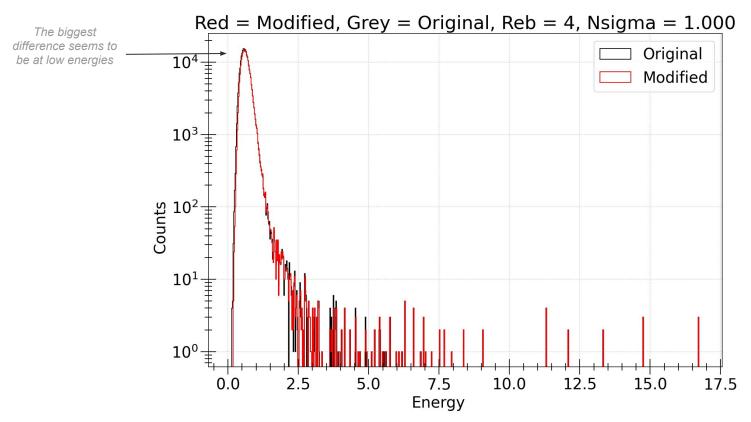
Then we verified the number of points remaining for each process.



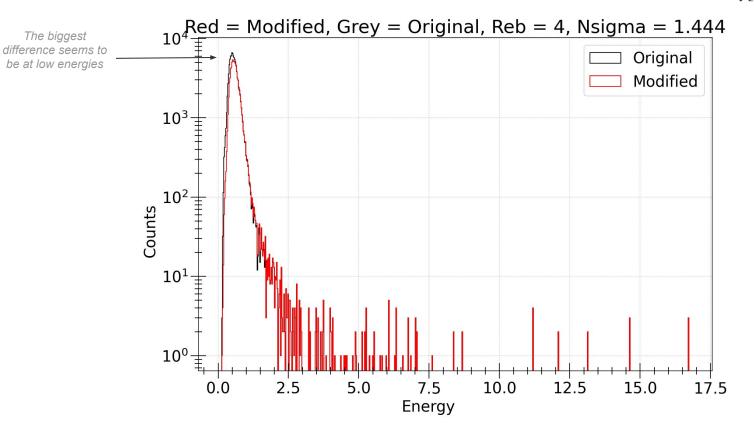
• Then we've looked at the energy histogram of the remaining points to compare both processes in a more measurable way.

Results - Energy distribution (*nsigma* = 1)

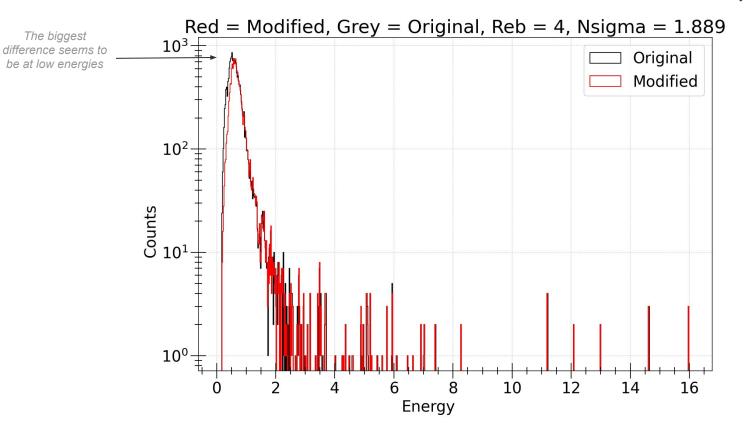




Results - Energy distribution (nsigma = 1.44)

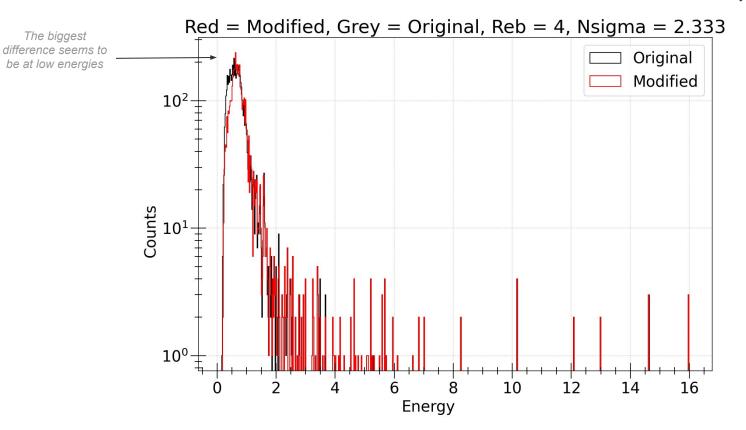


Results - Energy distribution (nsigma = 1.89)



Results - Energy distribution $(nsigma = 2.33)_{FE}$

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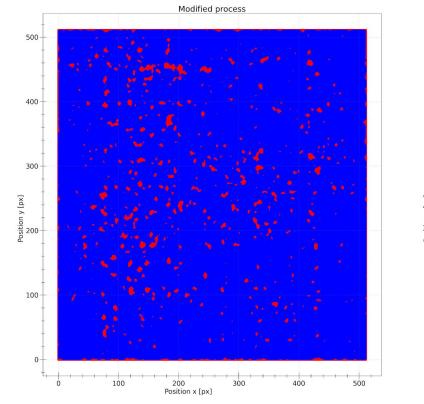


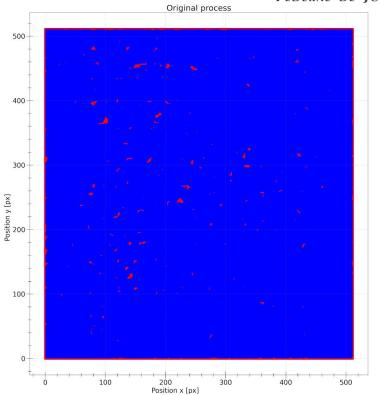


- Then we verified which points were being removed by each process;
- The next slides will show a scatter plot for removed points during the noise reduction algorithm;
- Red represents removed points while blue shows which points were kept.

Red = Removed, Blue = Kept, Reb = 4, Nsigma = 1.000

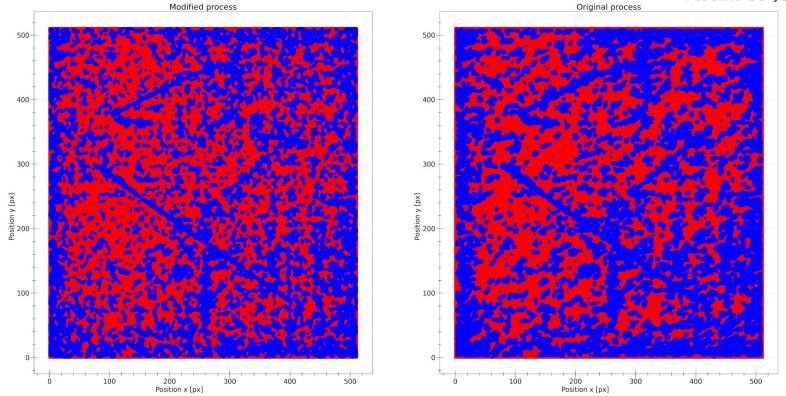






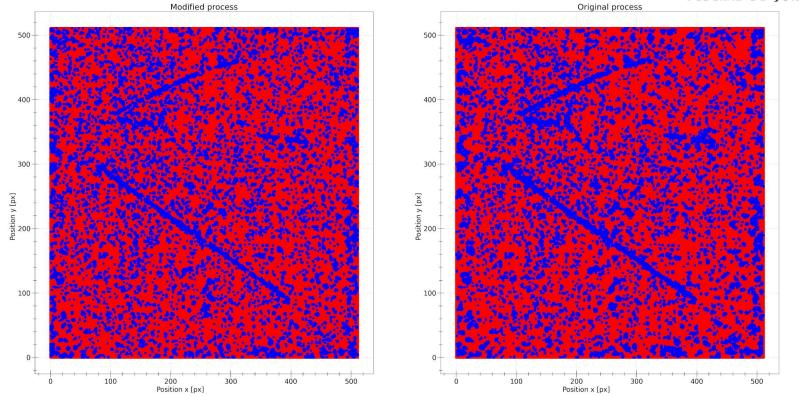
Red = Removed, Blue = Kept, Reb = 4, Nsigma = 1.444





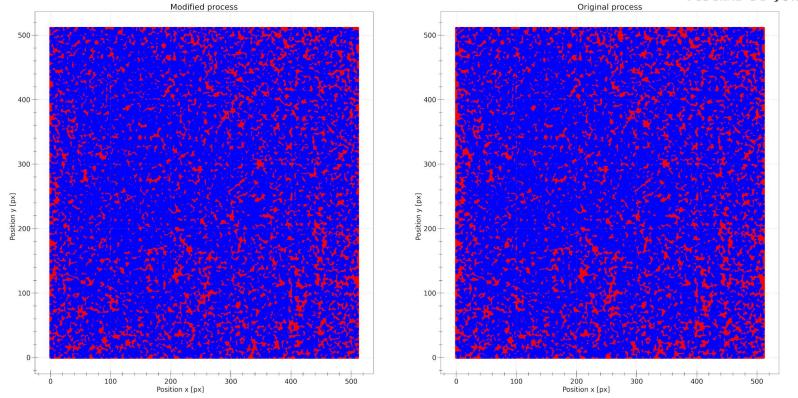
Red = Removed, Blue = Kept, Reb = 4, Nsigma = 1.889





Red = Removed, Blue = Kept, Reb = 4, Nsigma = 2.333



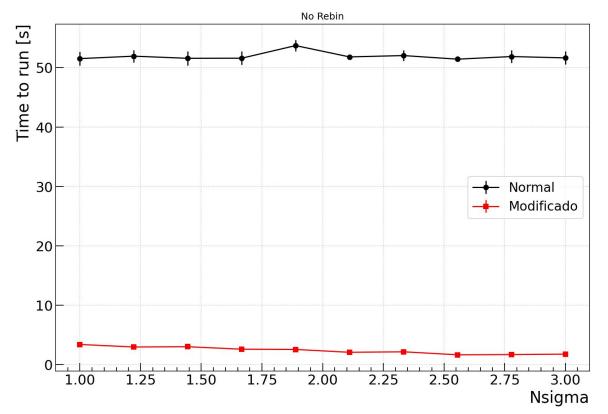




 This modification was made with the full resolution images in mind, so we had to test how long does it take to complete the process for it.

Results - time to run at full resolution





Conclusion



- Given how crucial the noise reduction algorithm is for the pre-processing, we've realized that it's one of the most time consuming steps, as it's recursive;
- If the collaboration intends to process images in full resolution, the noise reductor should to be redesigned as a faster filter;
- In order to help, a kernel-based filter was designed, with similar response and smaller processing time.
- Adiciona aqui que ainda esta em fase de testes...

Conclusion



• Any thoughts?