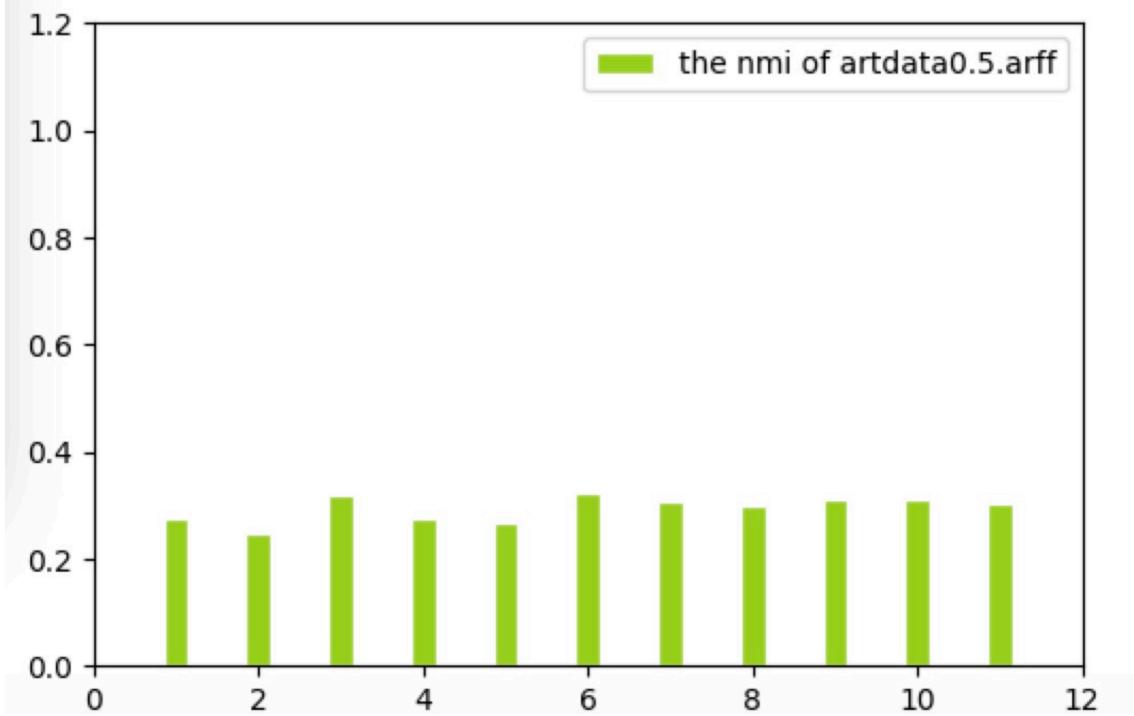
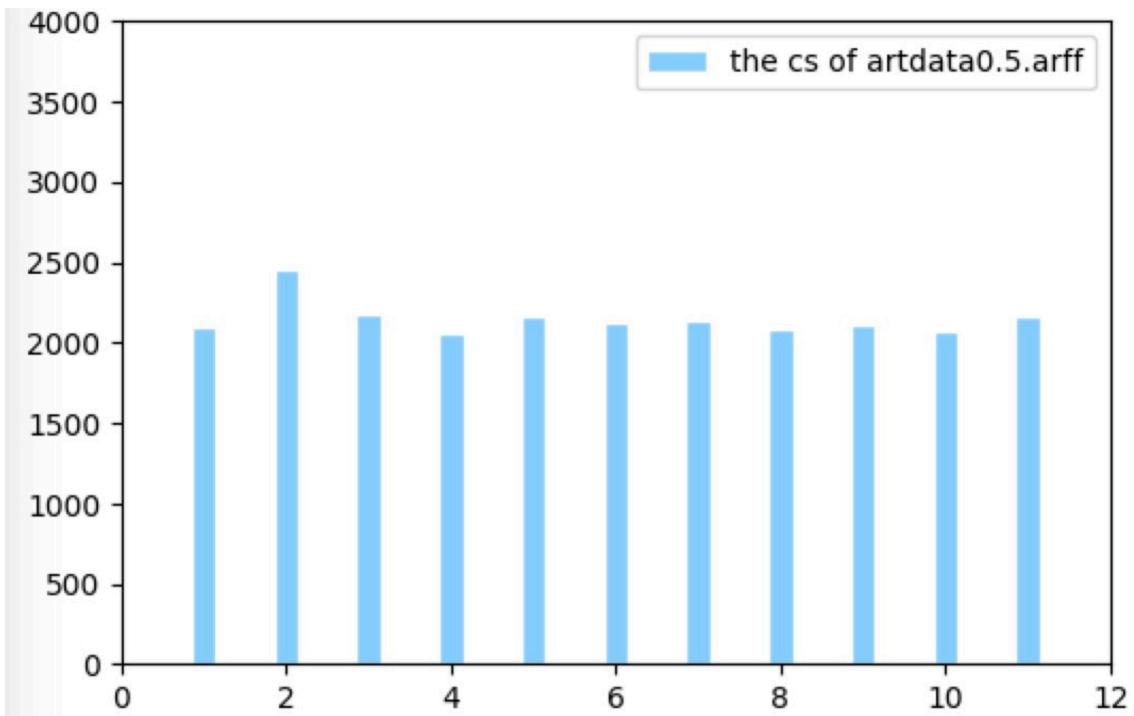
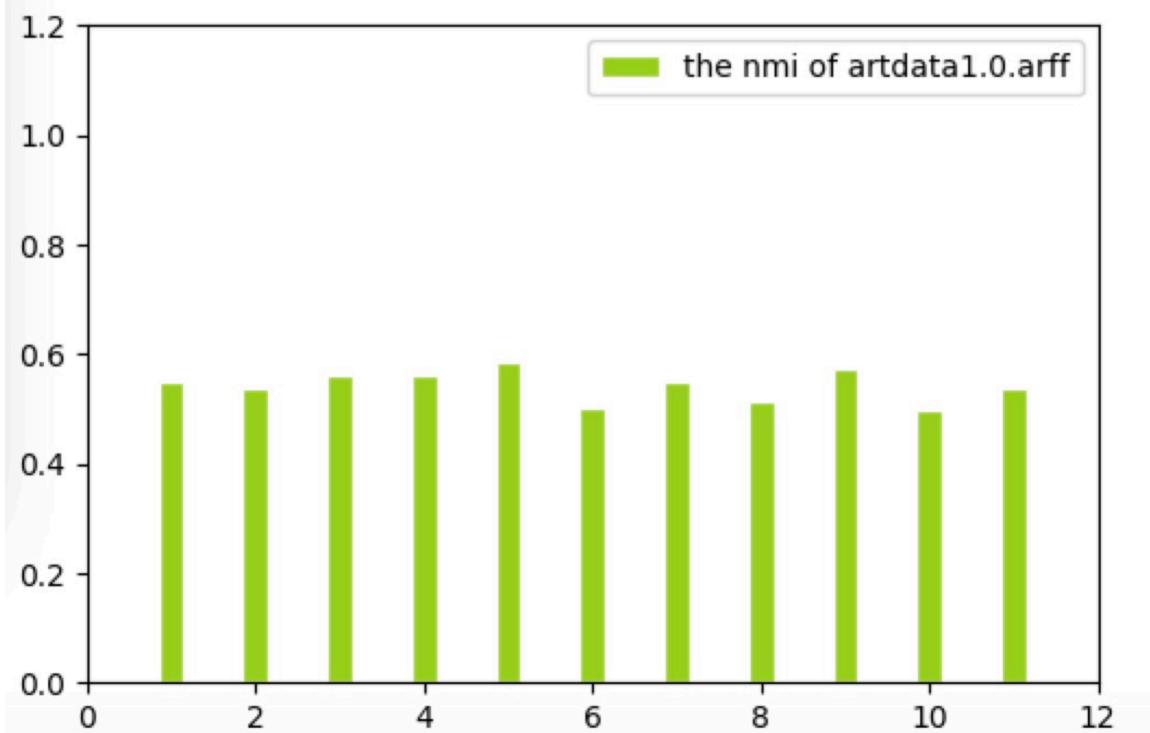
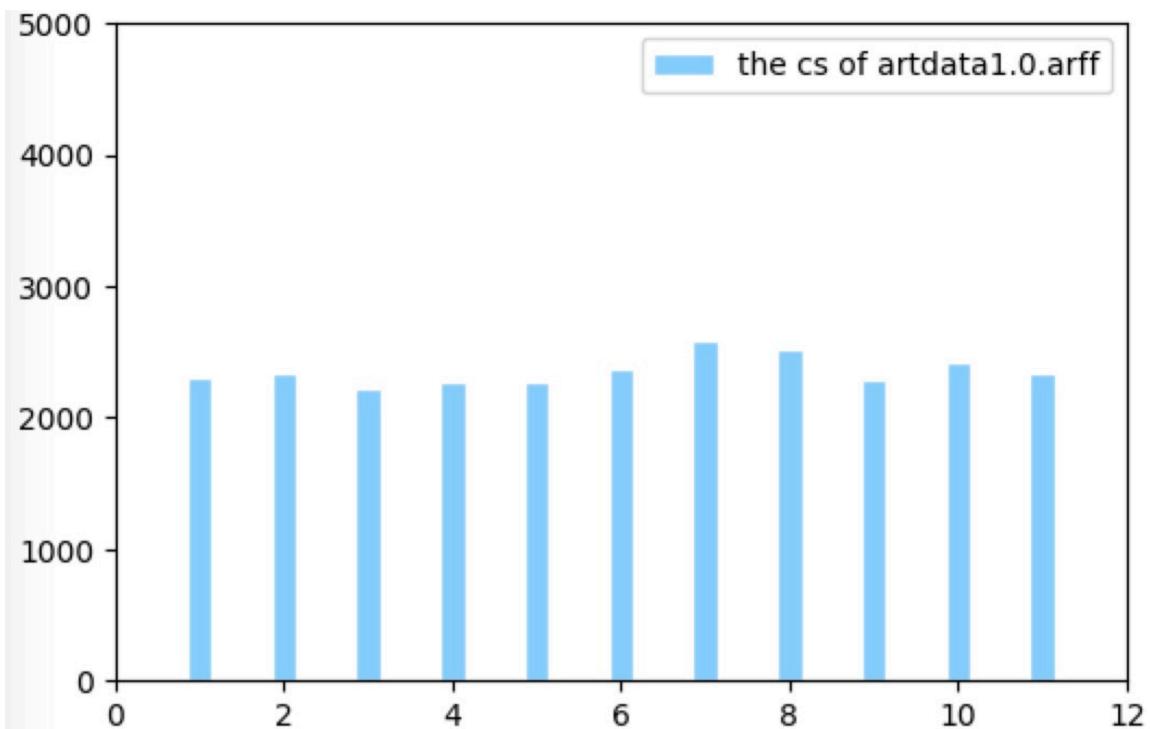


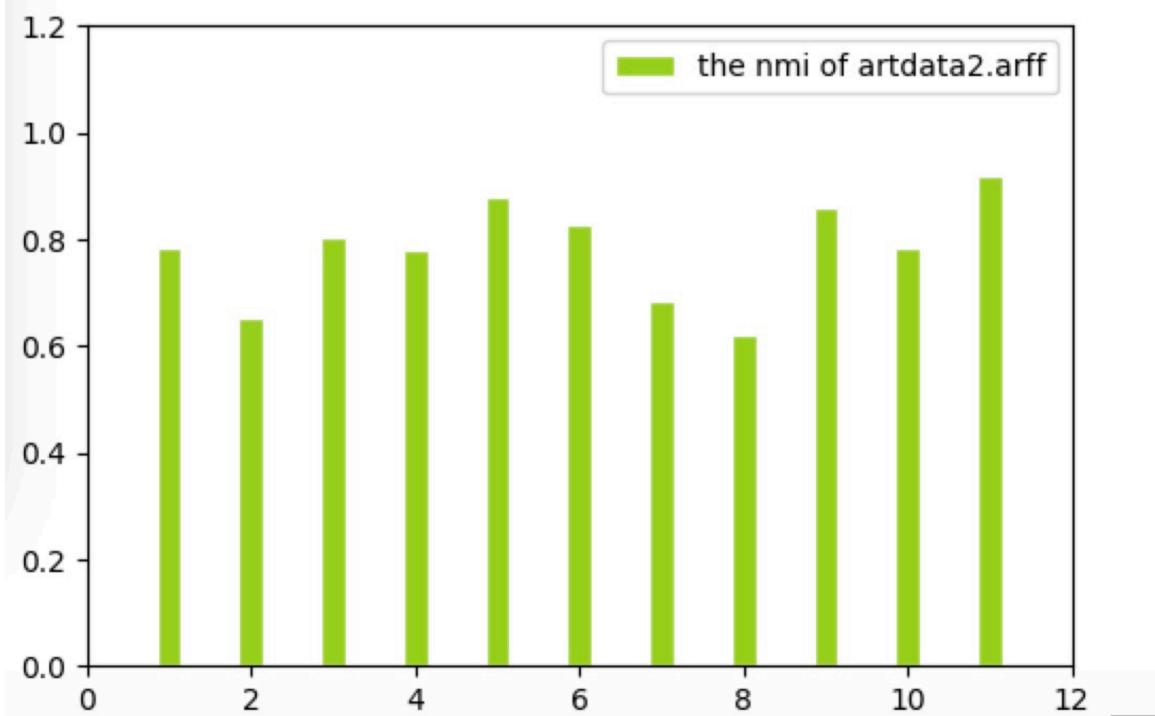
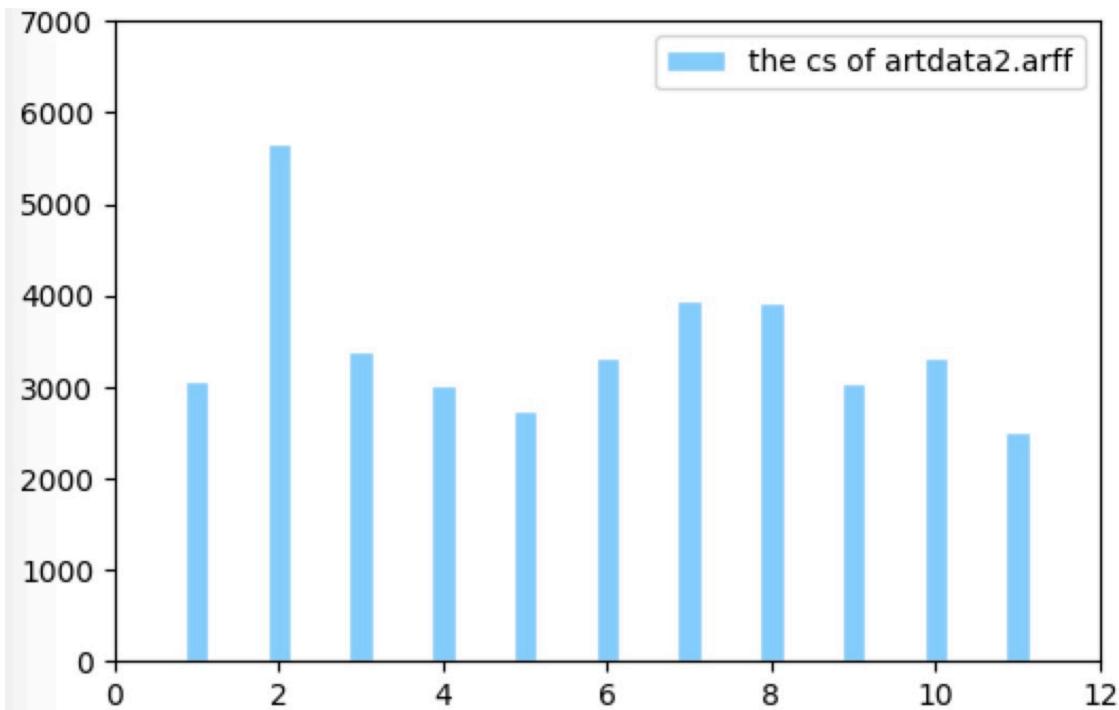
Project 3

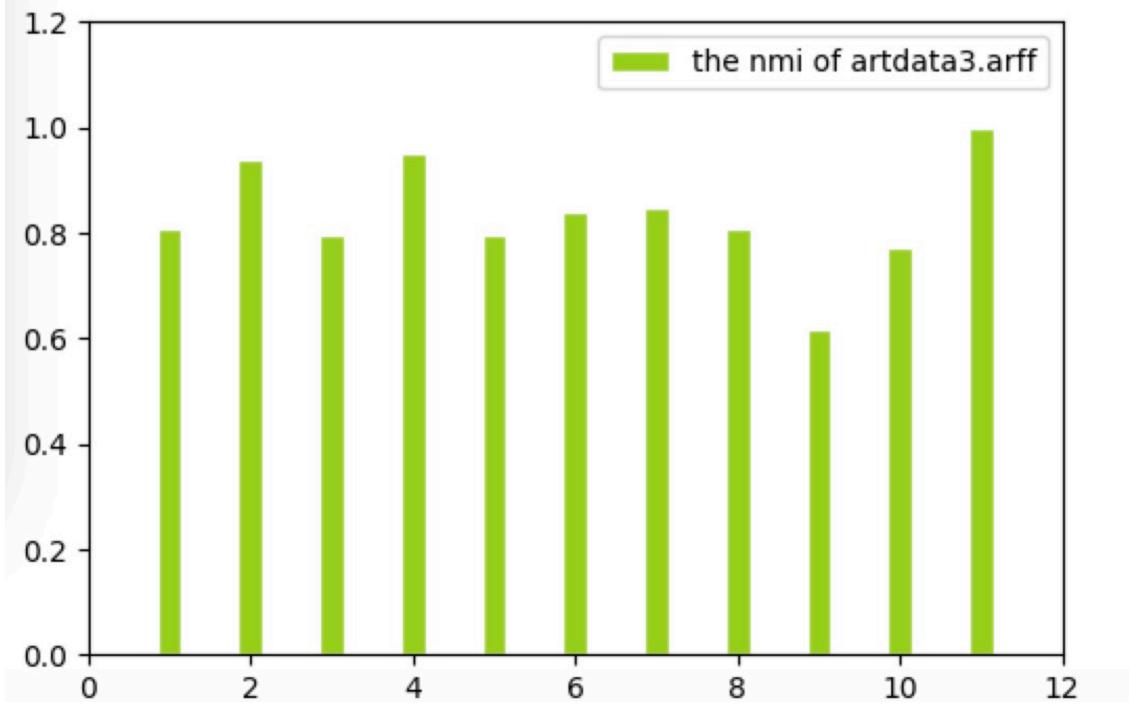
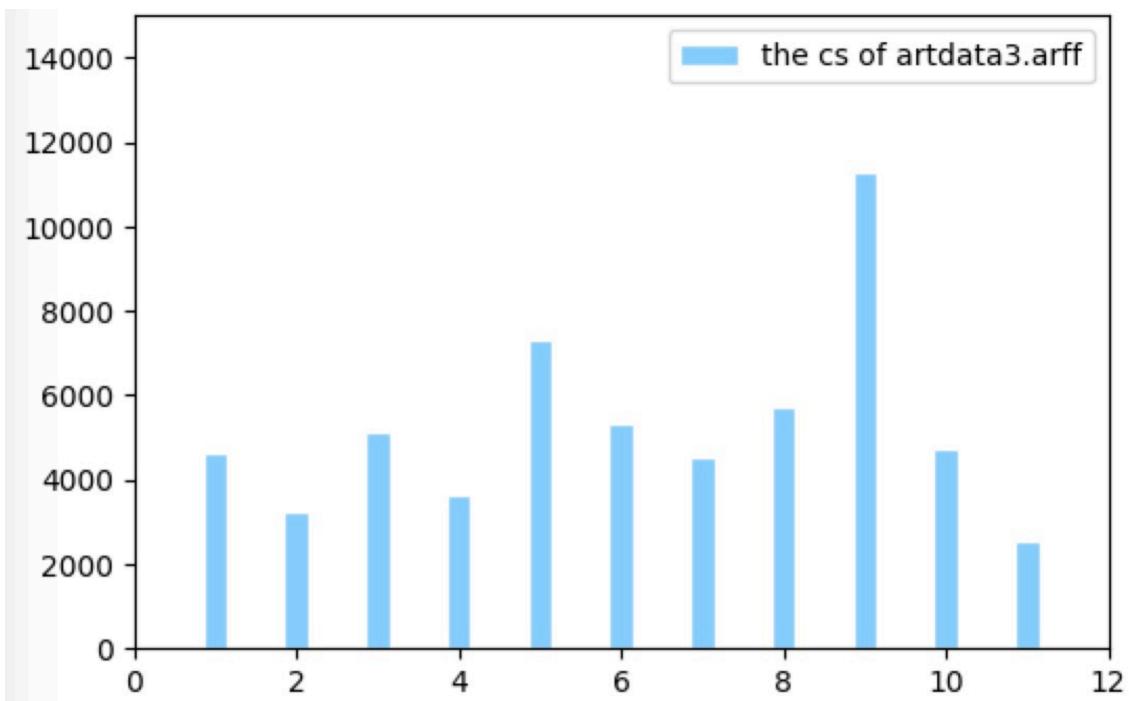
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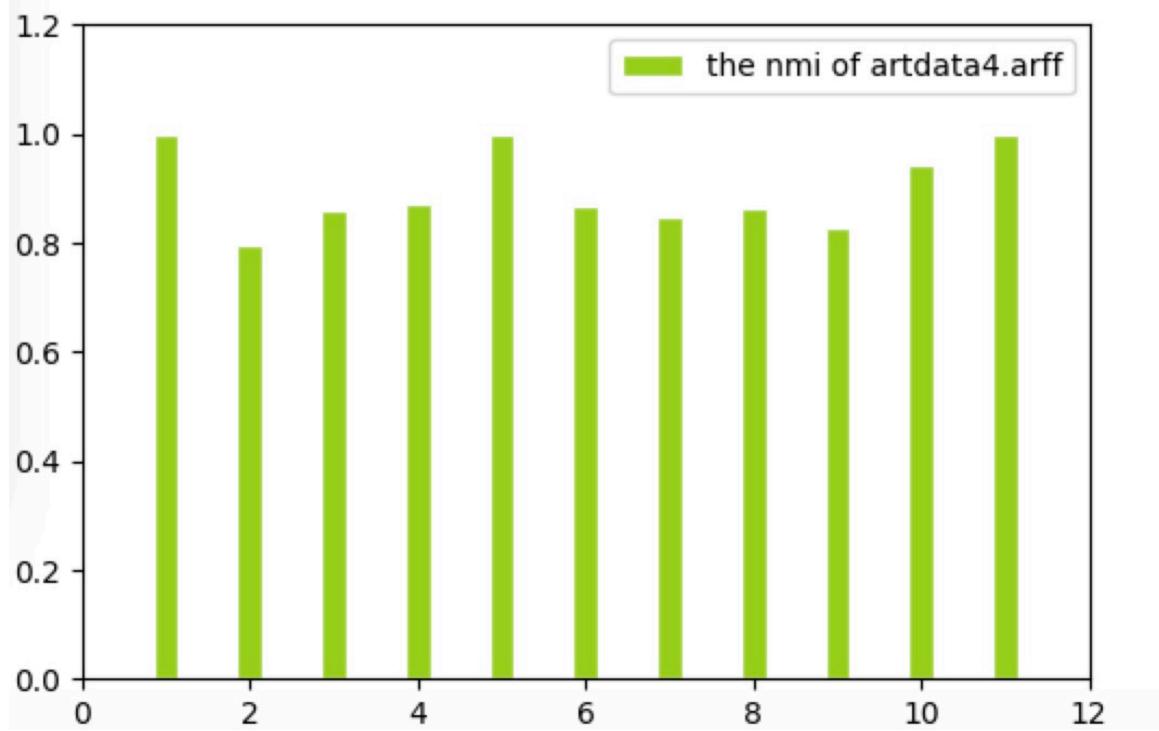
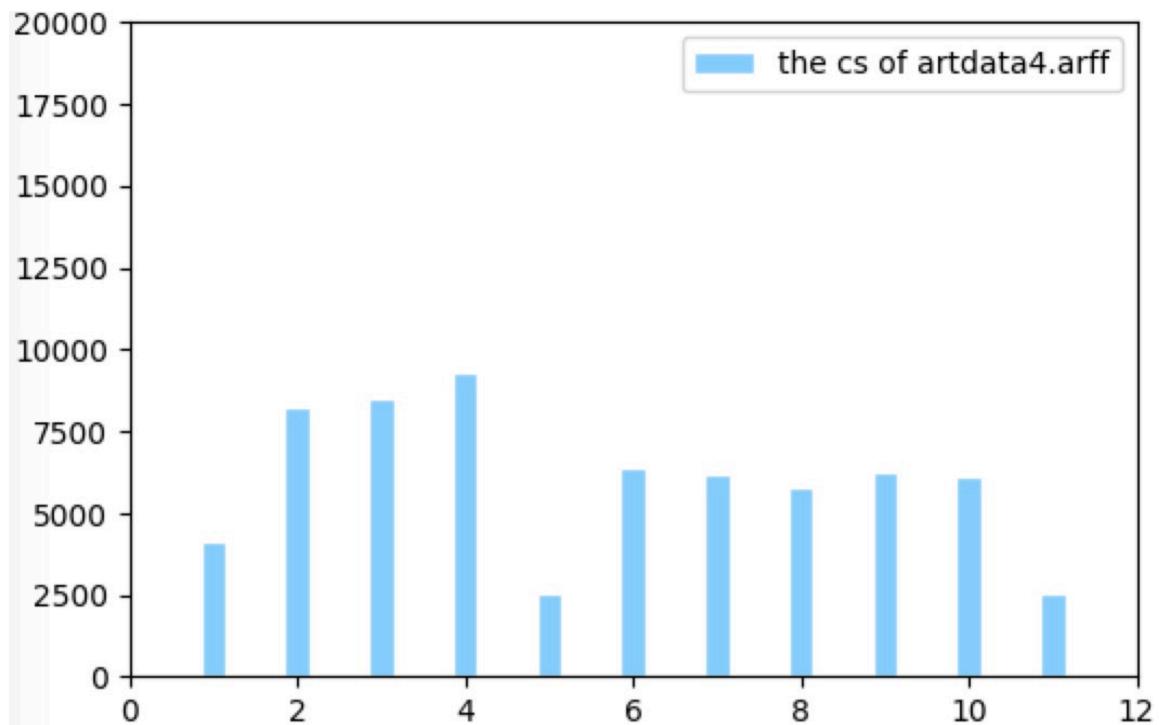
—. Calculate and plot the cs and nmi of 8 datasets.

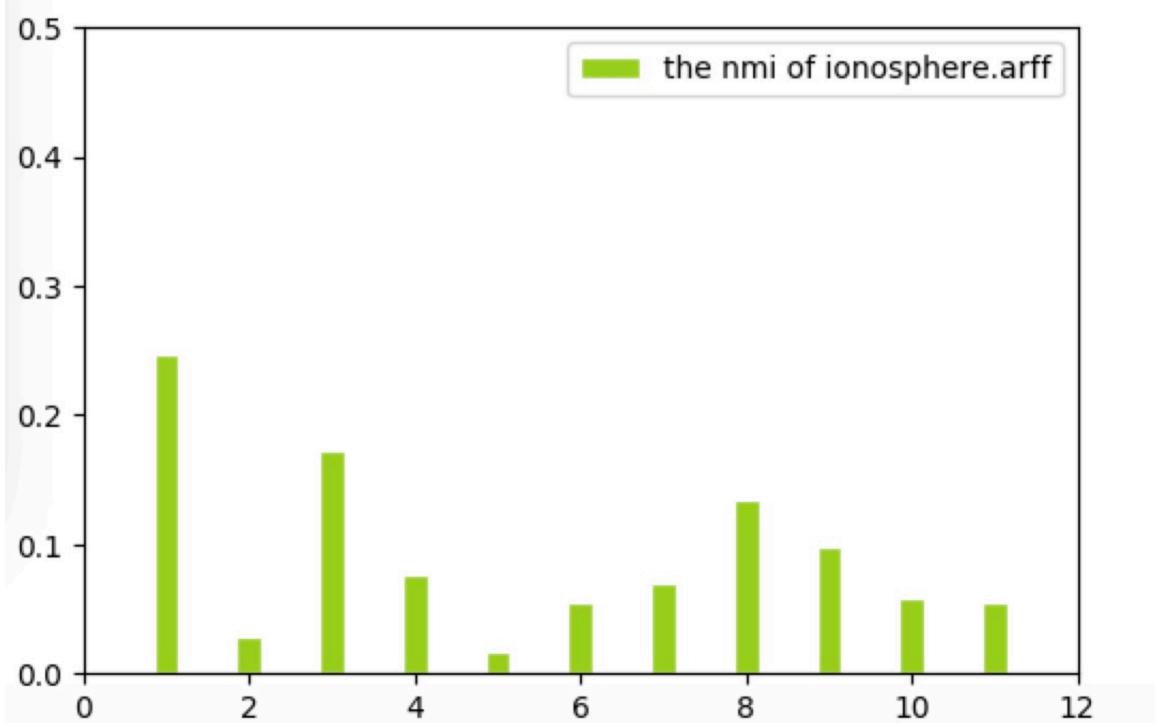
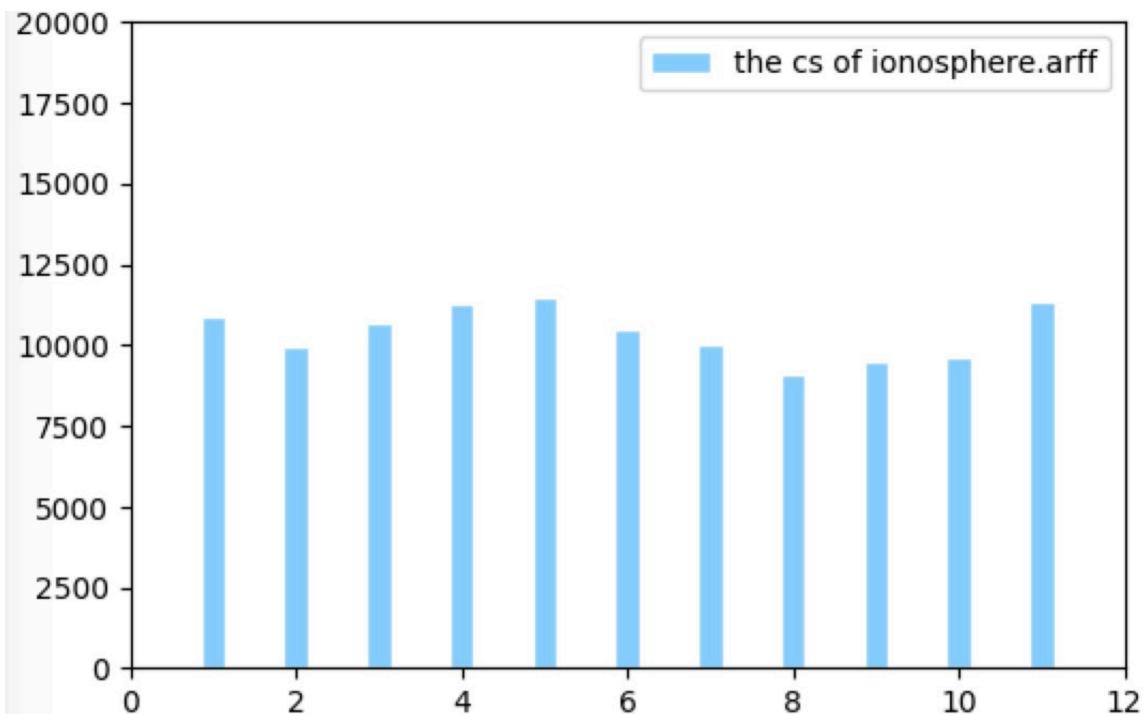


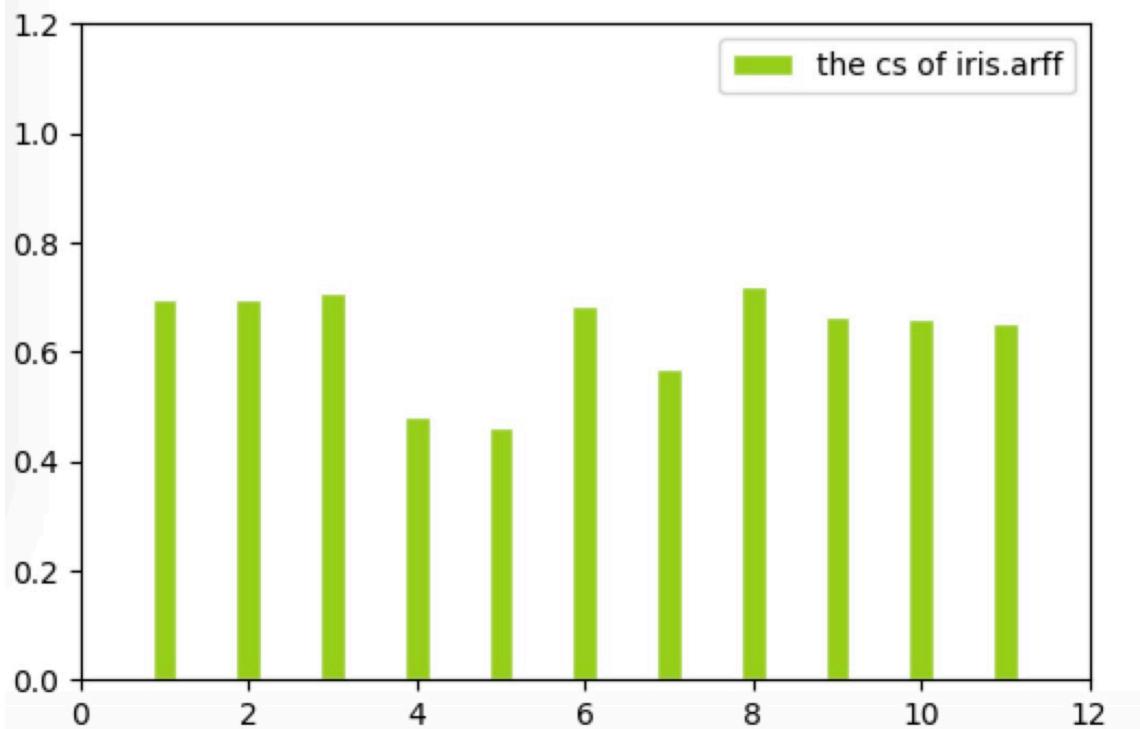
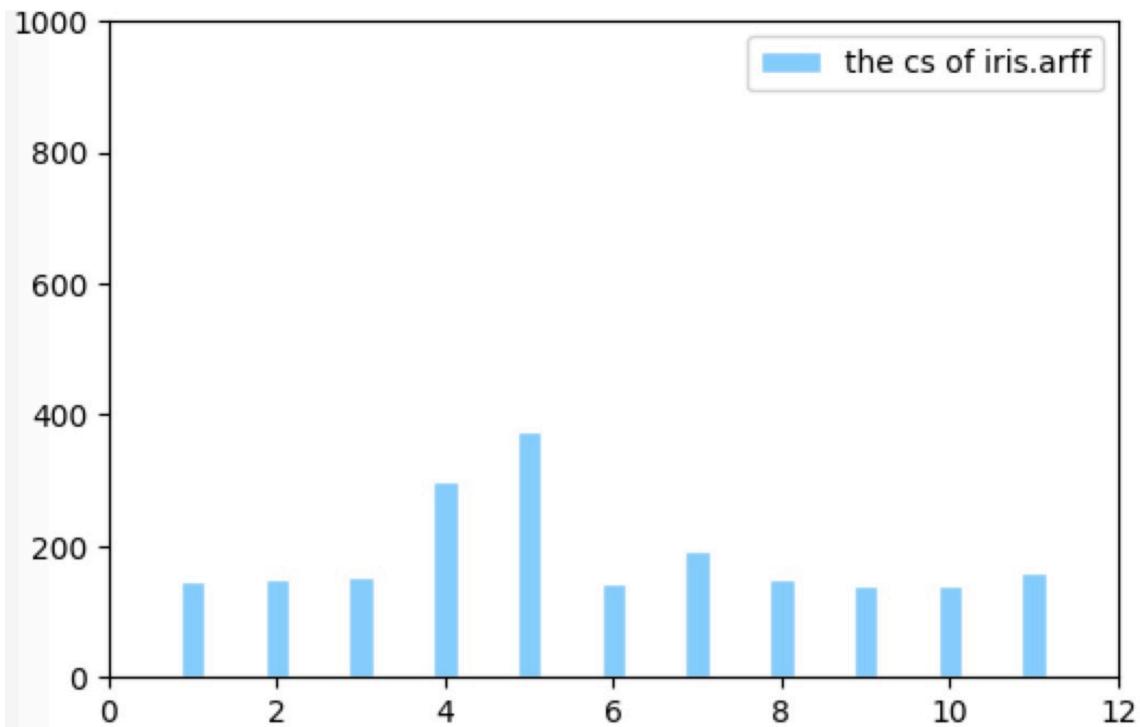


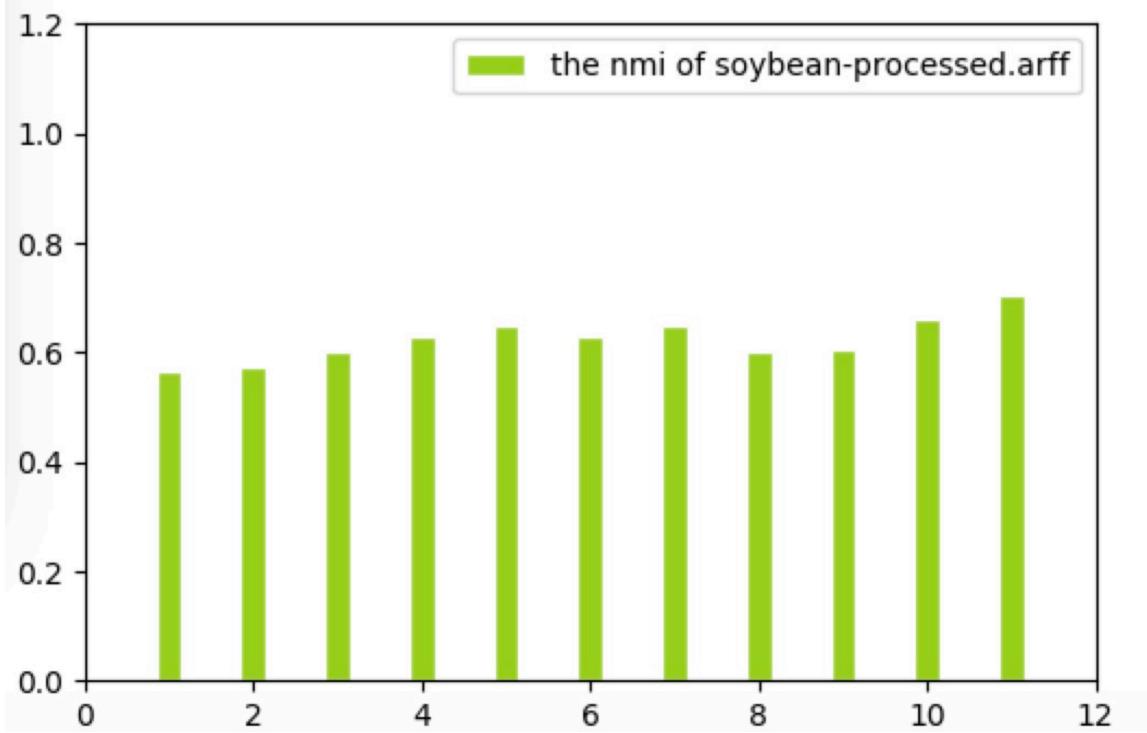
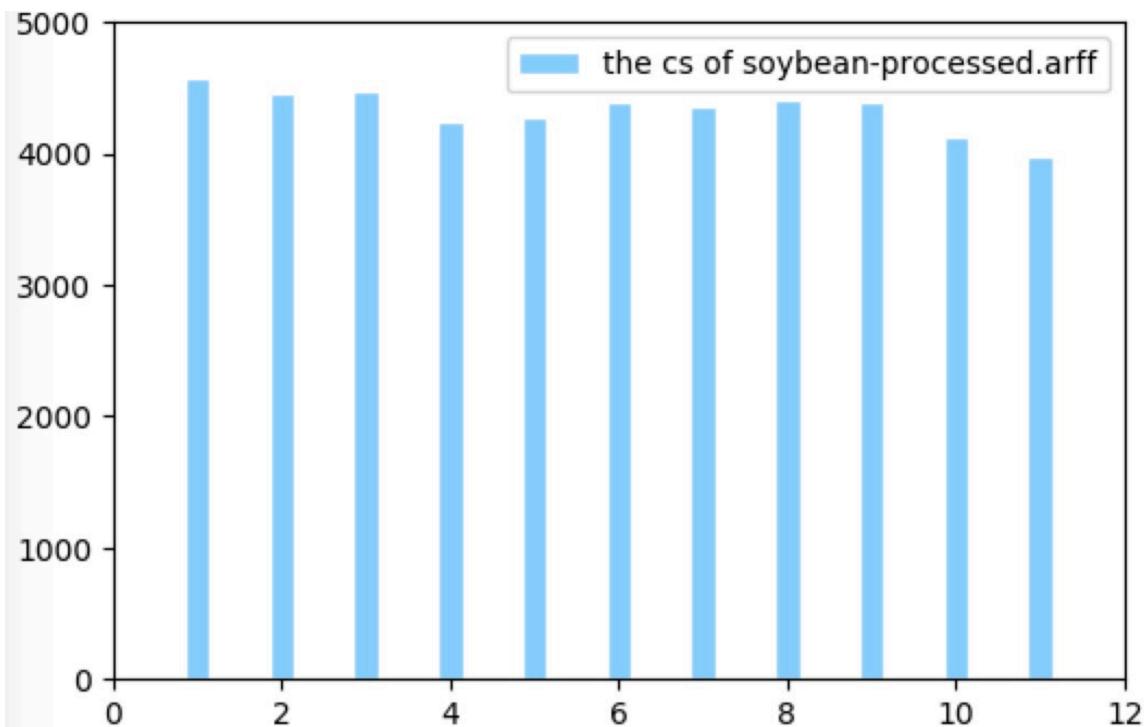












Discussion:

1. Are CS and NMI stable across multiple random initializations:

From the above graphs, the answer is no, For example, we can see the the cs and nmi value changes across different initialization.

2. Are their quality judgements in agreement?

Yes, if we see the trend of CS and NMI for the same sample, we can see NMI is higher and its CS will lower.

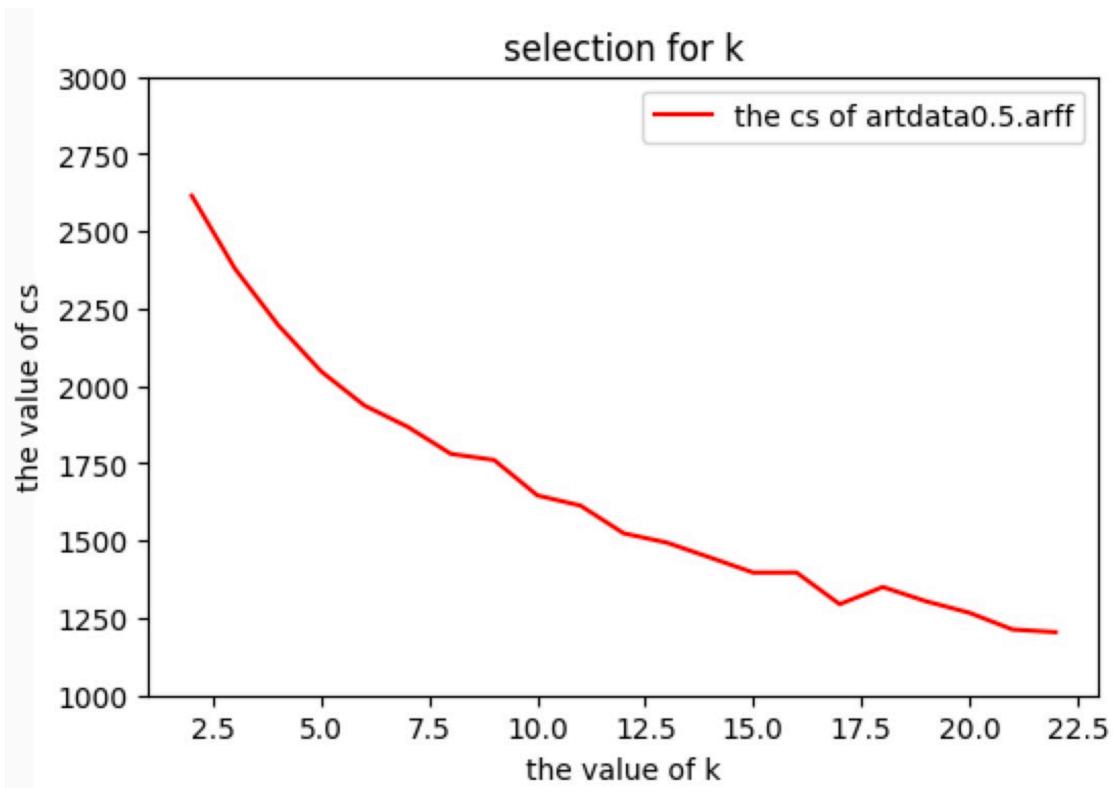
3. How does the smart initialization fare relative to the best random initialization.

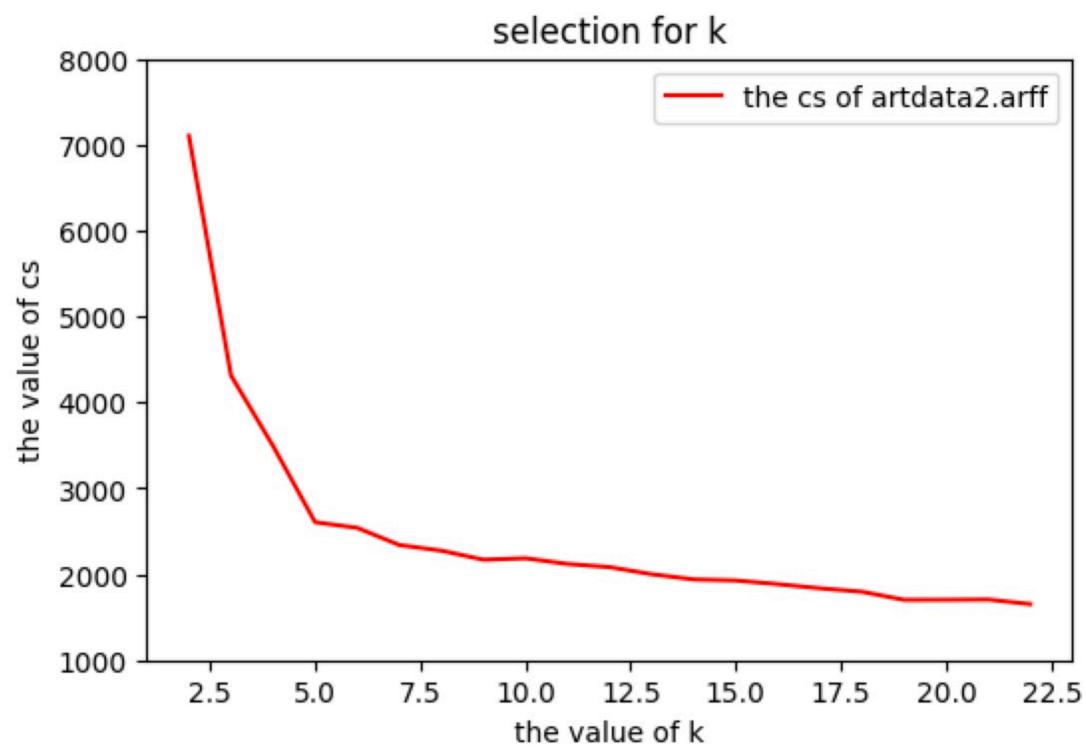
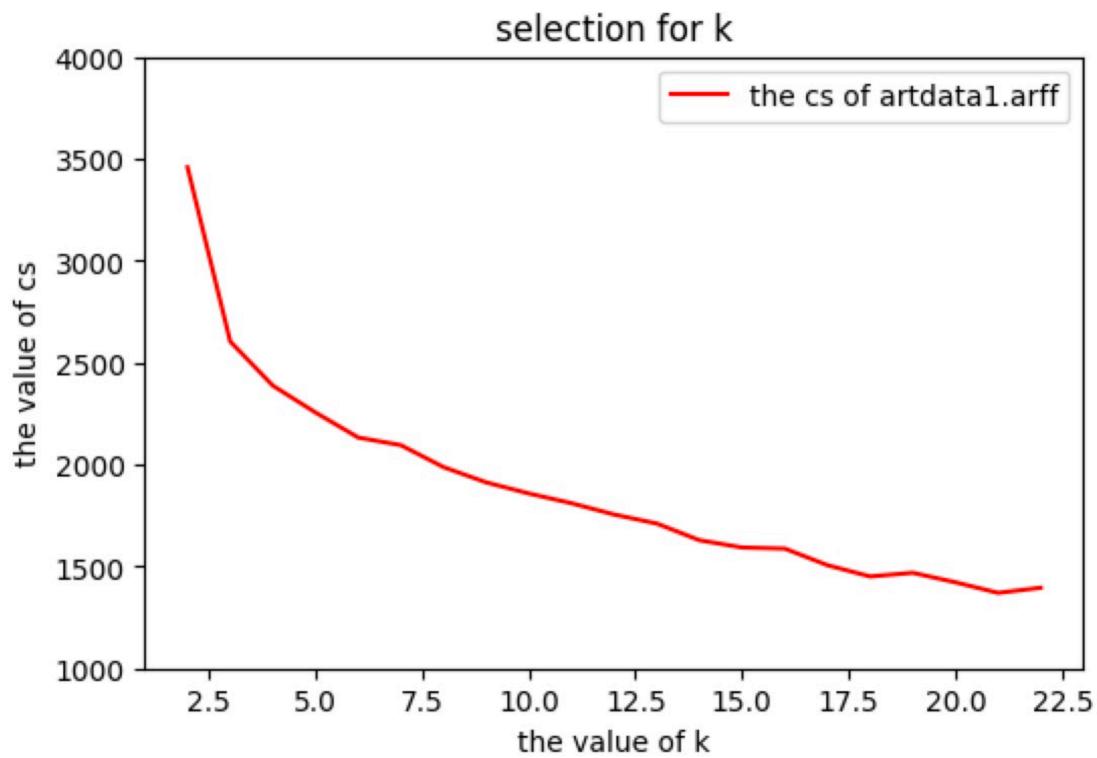
CS of smart initialization sometimes is higher than best random initialization and sometimes is lower than best random initialization. Compared random initialization, smart initialization cannot get the best result always but most of time, its result will be better than random initialization.

4. What other observations can you make from the results?

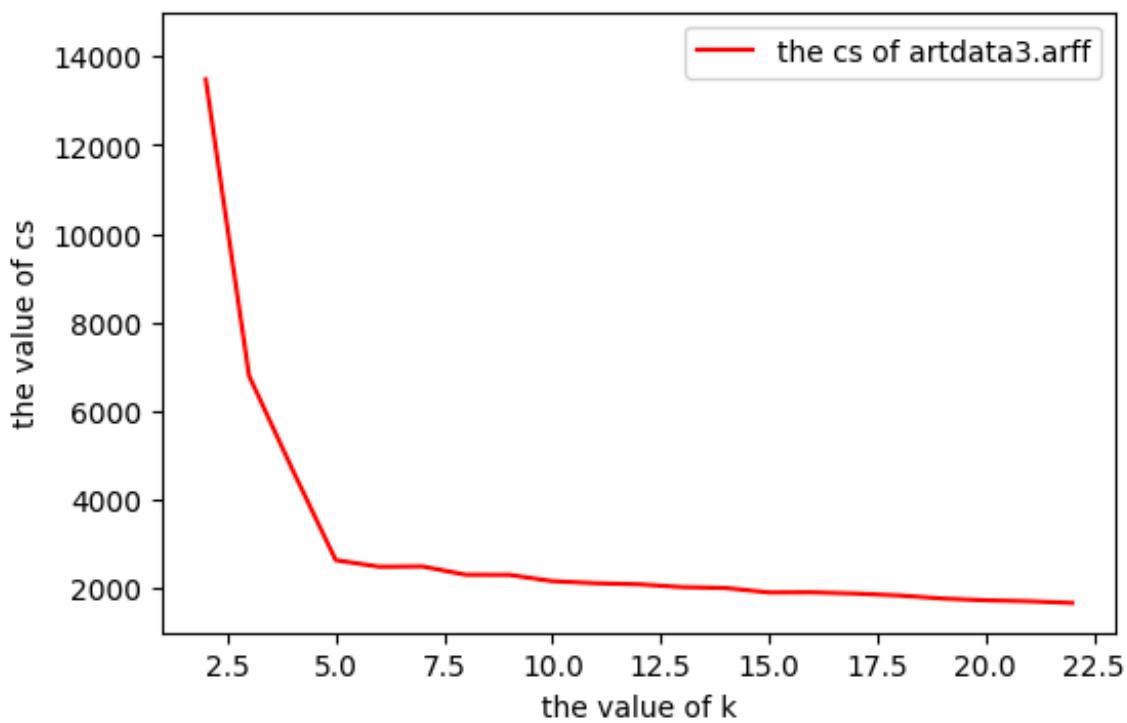
All these datasets were generated using 5 classes each corresponding to a Gaussian sphere, the main difference between these datasets relates to how close the spheres are to one another relative to their width. The closer they are, the harder it is to cluster the dataset. So compared with different artificial data, the variation of CS of the different initialization on the same sample increase with the sequence of artificial dataset(0.5,1,2,3,4) because the width of sphere to create data is smaller for 0.5 than that of 4.

二。 Calculate and plot CS value as a function of k(2 to 22).

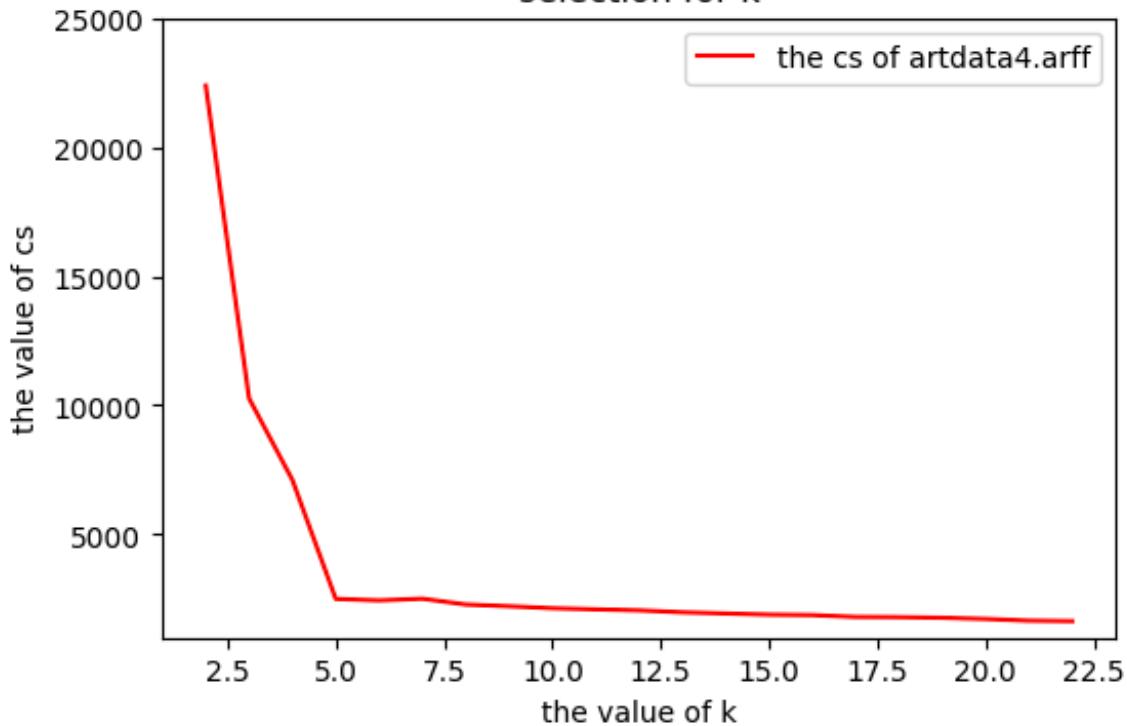


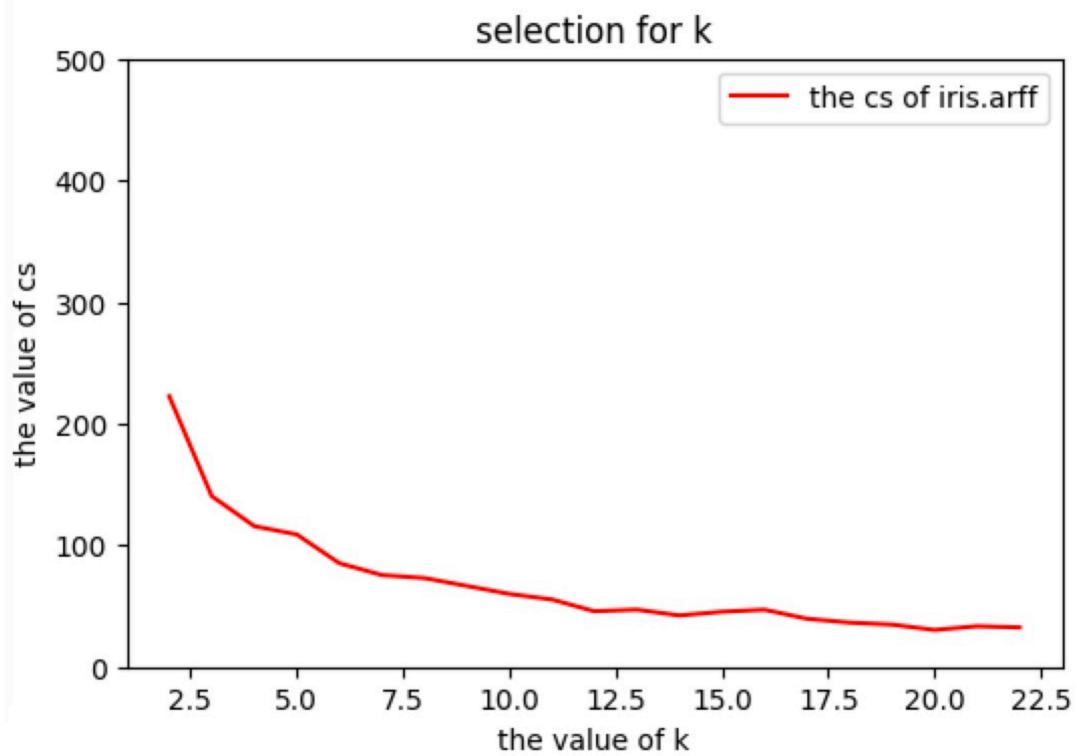
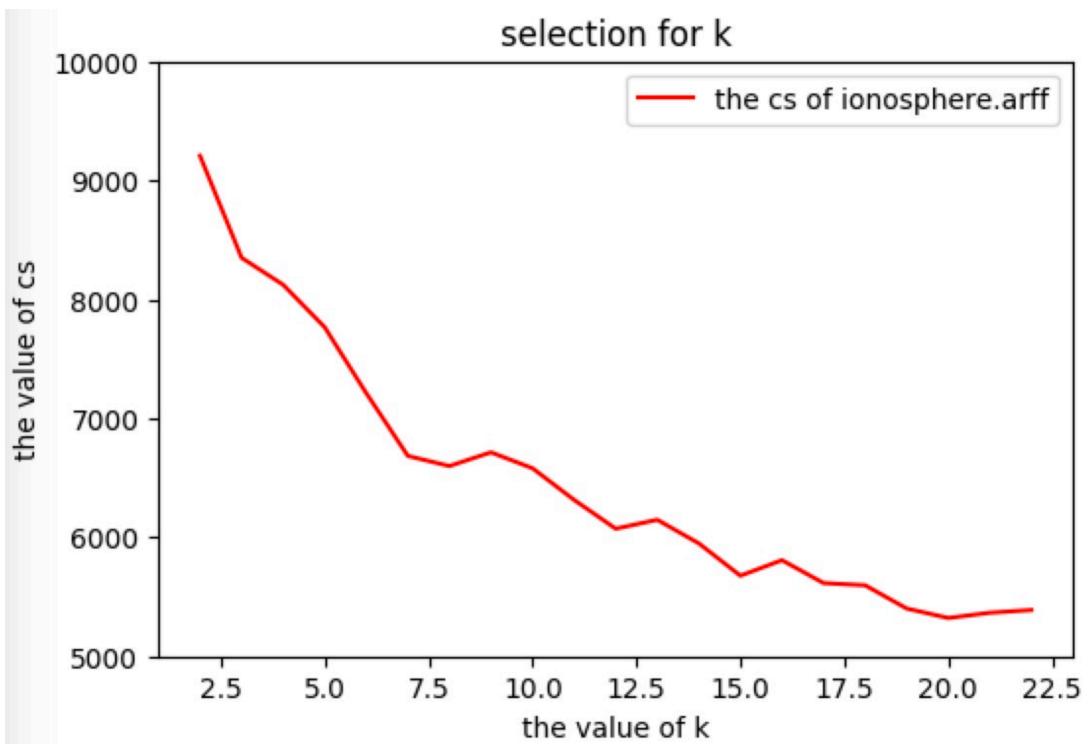


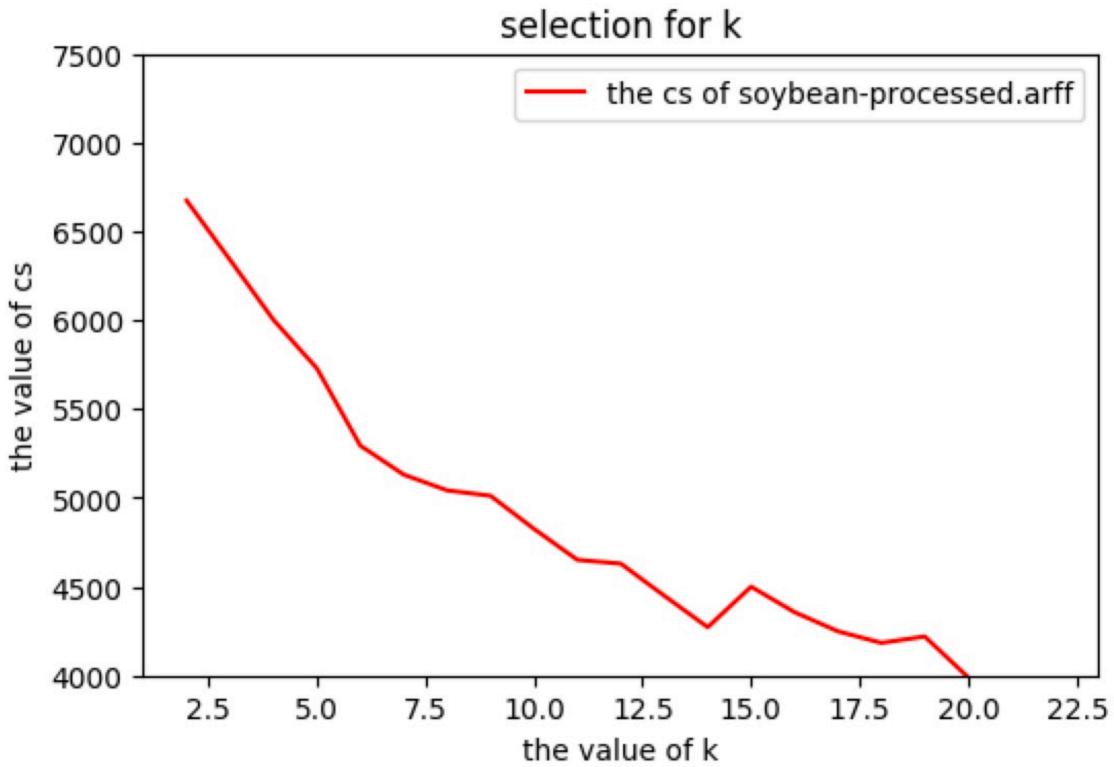
selection for k



selection for k







Discussion:

1, Is there visual evidence suggesting that CS can be used as a criterion for selecting k?

It depends, sometimes yes, sometimes no, it depends on data, if graph have knee criterion which mean we can use it to choose k.

2. What can you conclude from the relative performance on the artificial datasets?

The first artificial sample (0.5, 1) haven't knee criterion, others have the knee criterion. The reason why the first artificial sample haven't might be the data generated related to too close the spheres are relative to their width because all these datasets were generated using 5 classes each corresponding to a Gaussian sphere, the main difference between these datasets relates to how close the spheres are to one another relative to their width. The closer they are, the harder it is to cluster the dataset. The Gaussian distribution of sample (0.5, 1) is too close to cluster so there is no any knee criterion. Others' knee criterion have the same knee criterion when k is equal to 5 because all of data generated by five Gaussian sphere.

4. Are the results consistent across the datasets? If not, what observations can you make from the results?

No, Most artificial sample have knee criterion, but no any real dataset have knee criterion, it looks like this method is hard to be used in real world.