endmember extraction algorithms, SVDSS and ModPPI. The unmixing results using PPI and SVDSS are quite consistent and related to the fact that we used simulated noise free imagery. We expect more differences with more realistic simulated data.

SVDSS was the best in extracting the four endmembers present in the simulated data but requires prior knowledge of the number of endmembers. However, ModPPI was capable of extracting all component signatures, and it was able to determine the number of endmember from the data. Work is still needed on an approach to select the threshold to determine the number of endmembers in ModPPI.

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