

## Machine learning for SMART mineral mapping using coupled XRF-XRD

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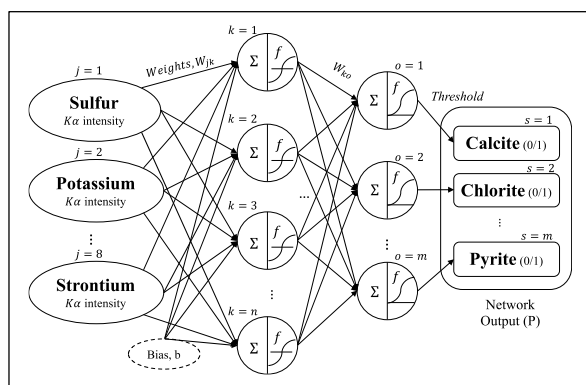
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A Synchrotron-based Machine learning Approach for RasTer (SMART) mineral mapping was developed to train a mineral classifier that can interpret raster-scanned millimeter-sized areas of rock thin sections with micron-sized resolution. Training is done using Artificial Neural Networks (ANN) (Figure 1) with coupled micro X-ray fluorescence ( $\mu$ XRF) intensities, which provide information about element abundances, and micro X-ray diffraction ( $\mu$ XRD) patterns, which provide information about mineral identity.

The resulting SMART mineral mapper can identify minerals using only micro X-ray Fluorescence ( $\mu$ XRF) data. The value of this approach comes from the fact that  $\mu$ XRF data are relatively fast to collect and interpret whereas the  $\mu$ XRD data take longer to collect and much longer to interpret.



**Figure 1.** Artificial Neural Network (ANN) for mineral classification from XRF data.

In application to a 14 mm<sup>2</sup> area of a thin section of Eagle Ford shale scanned with 2  $\mu$ m resolution, 26 minerals were detected using 8 elements (S, K, Ca, Ti, Mn, Fe, As, Sr). In the testing phase, all but two of 76 testing data points were correctly classified.