



Unearthing Hidden Treasures: Detecting Critical Minerals from Historical Maps

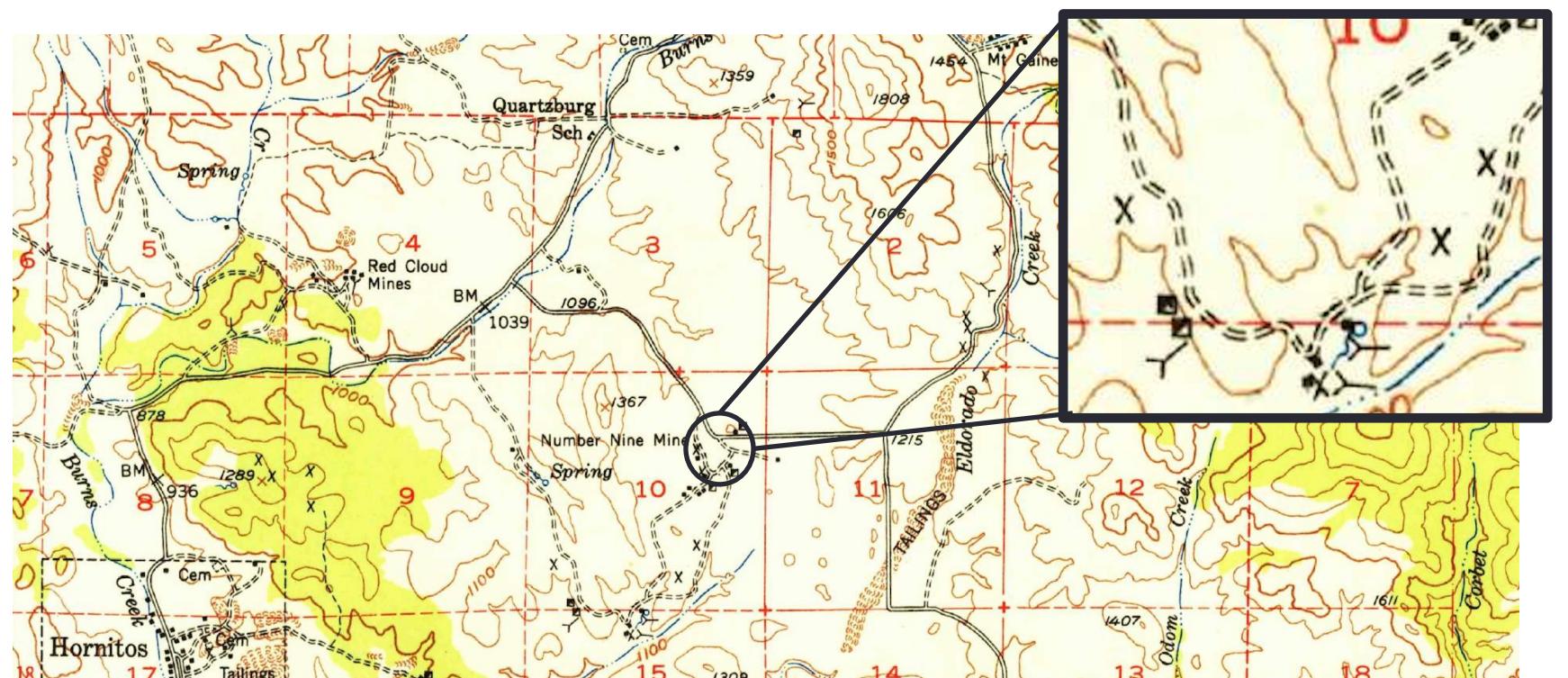


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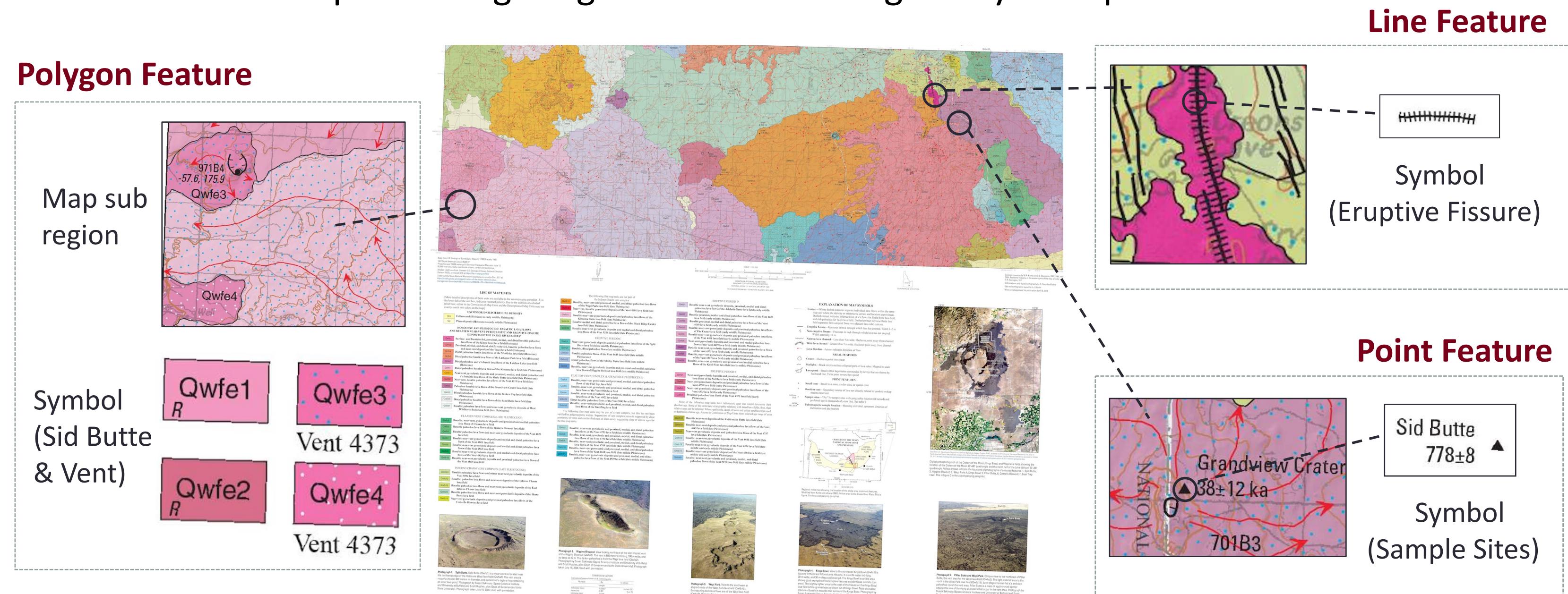
CRITICAL MINERAL DETECTION TASK

- The United States Geological Survey (USGS) collects large volumes of historical maps to **assess the availability of critical minerals**
 - Detect the mineral deposit on the maps (Figure 1)
 - However, manually reviewing these maps is time-consuming



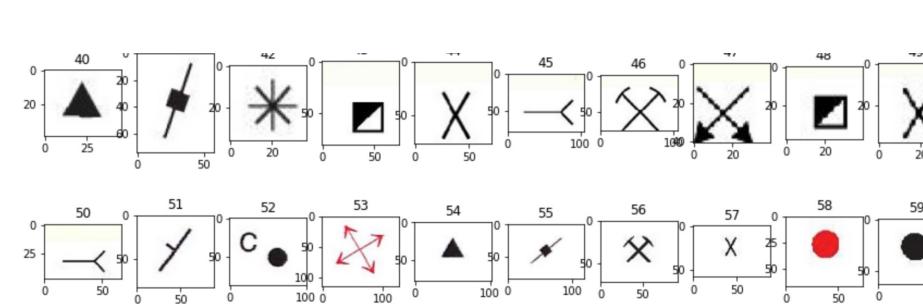
The locations of mineral deposit on a map

- Task:** Given a target symbol, automatically and accurately identify its appearances on the historical maps
 - Three types of features: lines, polygons, and points
 - The output is image segmentation indicating the symbol positions



CHALLENGES IN FEATURE DETECTION

- Computational challenge:** Need to build individual models for every symbol from the provided map scans
- Some **line features** are similar, easily causing false detection. Also, the detected lines need to be continuous in the segmentation results
- Polygon features** have various colors, texts, and textures. Simple color-based methods do not handle the symbols within the polygons. Some target symbols are hard to distinguish
- Point features** are suffering from lack of training data and large variations



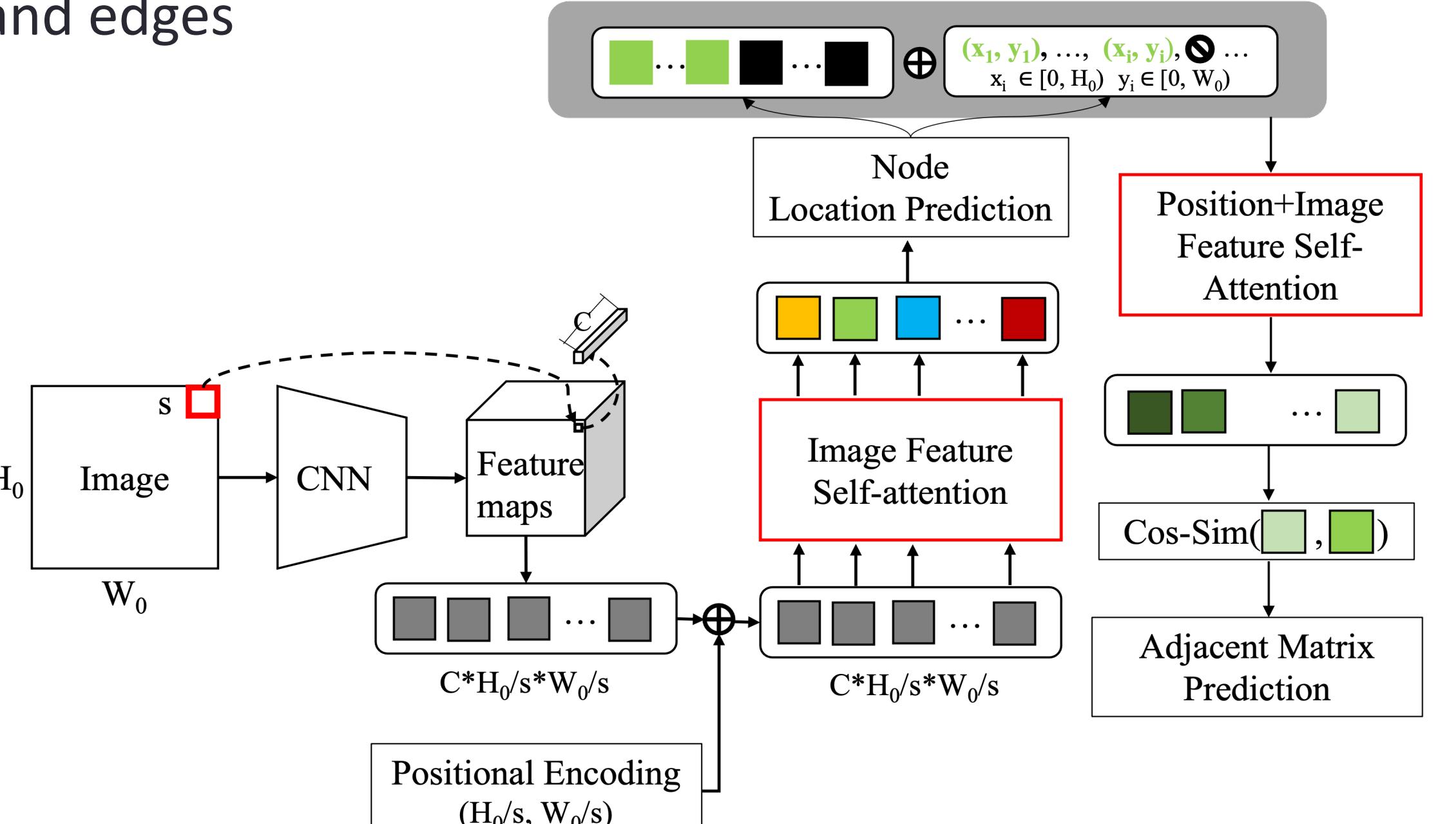
Legend: X Prospect Pit
On map: Color mismatch

Legend: □ Mine Shaft
On map: Extreme Size Variation

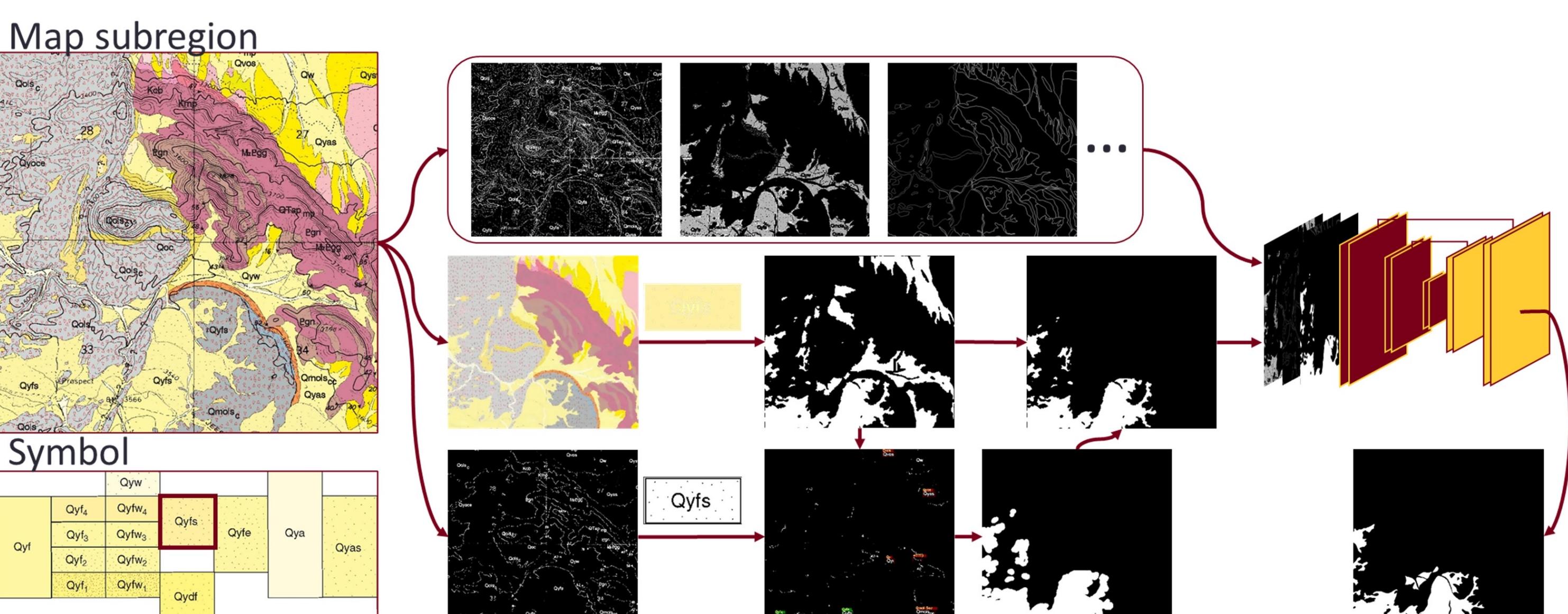
Samples of point feature symbols

SYMBOL DETECTION

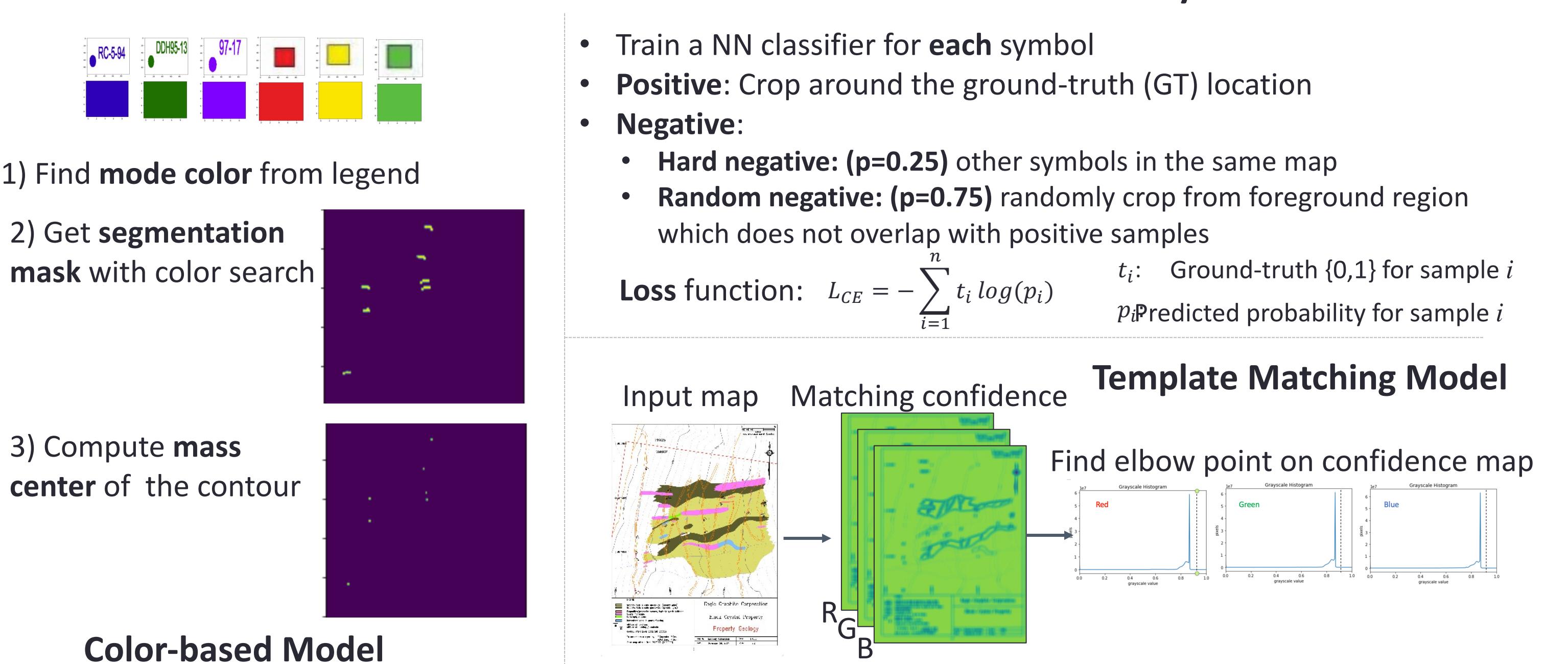
- The **line feature detection** model takes a map image as an input, and leverages attention mechanism to predict the line vectors, consisting of nodes and edges



- The **polygon feature detection** model synergizes the color, pattern of text, and map texture for extracting polygons



- The **point feature detection** module involves three models to handle large variations of symbols: a color-based model, a deep neural network (DNN) model, and a shape-based template matching model.



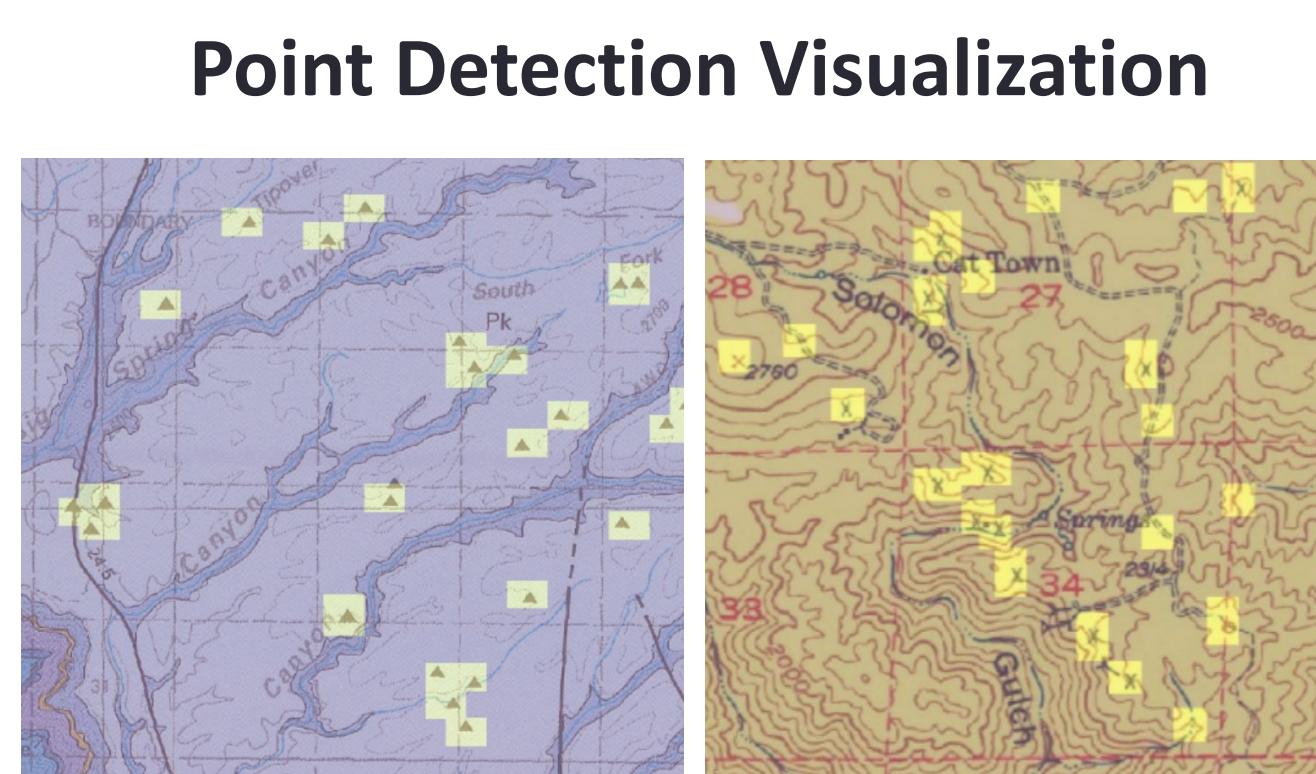
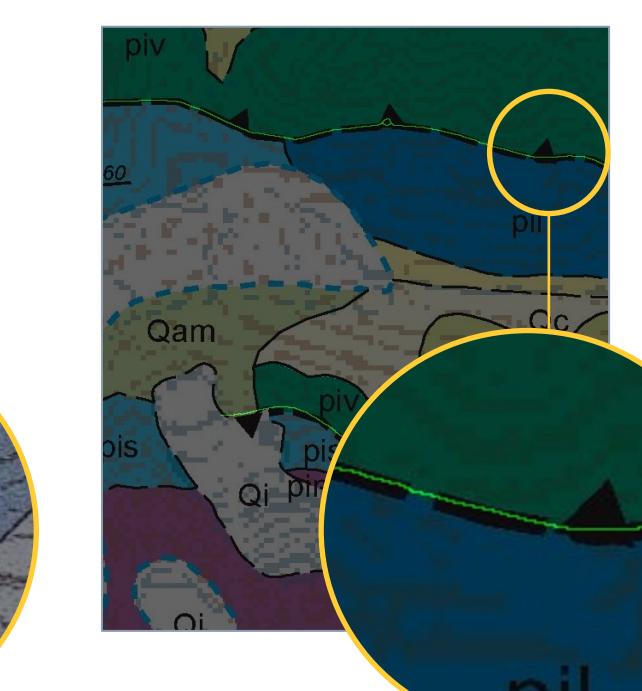
MSI HPC RESOURCES FOR TASKS

- We store the provided large-scale scanned maps (283 in total with a size of ~100G) and large volumes of intermediate results on the **MSI High Performance Storage**.
- We leverage the **MSI high-performance machines** to preprocess these maps in parallel, e.g., image cropping and generating training samples
- We take advantage of the **GPU resources** (a100 machines) on MSI for training various deep neural networks.
- We **train models** for more than 30 types of line features and 15 types of point features, respectively.

EVAL METRICS & RESULTS

Polygon Detection	Median Precision	Median Recall	Median Macro F-1
Color	0.680	0.971	0.737
Color + Text	0.747	0.967	0.780
Color + Text + Texture	0.866	0.937	0.823

Line Detection Visualization (green lines)



CONCLUSION & ACKNOWLEDGEMENT

- Our team (ISI-UMN) won the First Place in DARPA Map Feature Extraction Challenge (<https://criticalminerals.darpa.mil/>)
- The proposed system automatically detects line, polygon, point features on the historical map scans, which helps critical mineral assessments
- We thank USGS and DAPAR for providing the high-resolution historical maps and organizing the competition
- We acknowledge MSI and Research Computing for providing powerful computational resources that significantly benefit the research



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