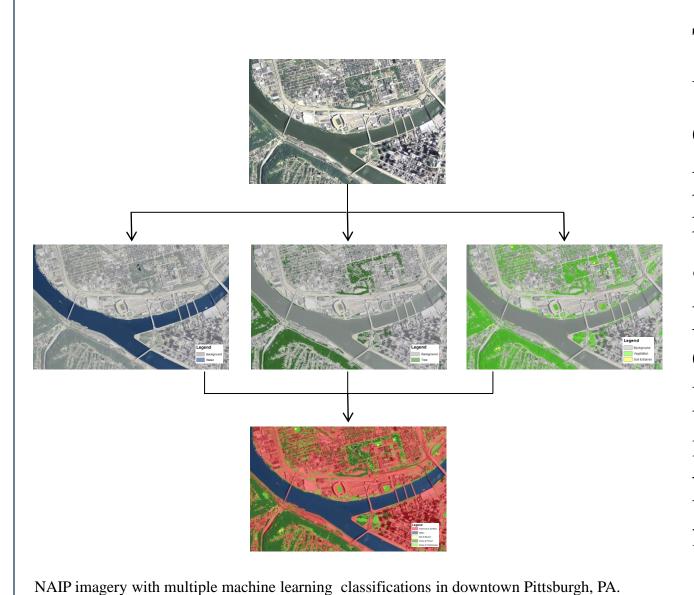
EnviroAtlas

www.epa.gov

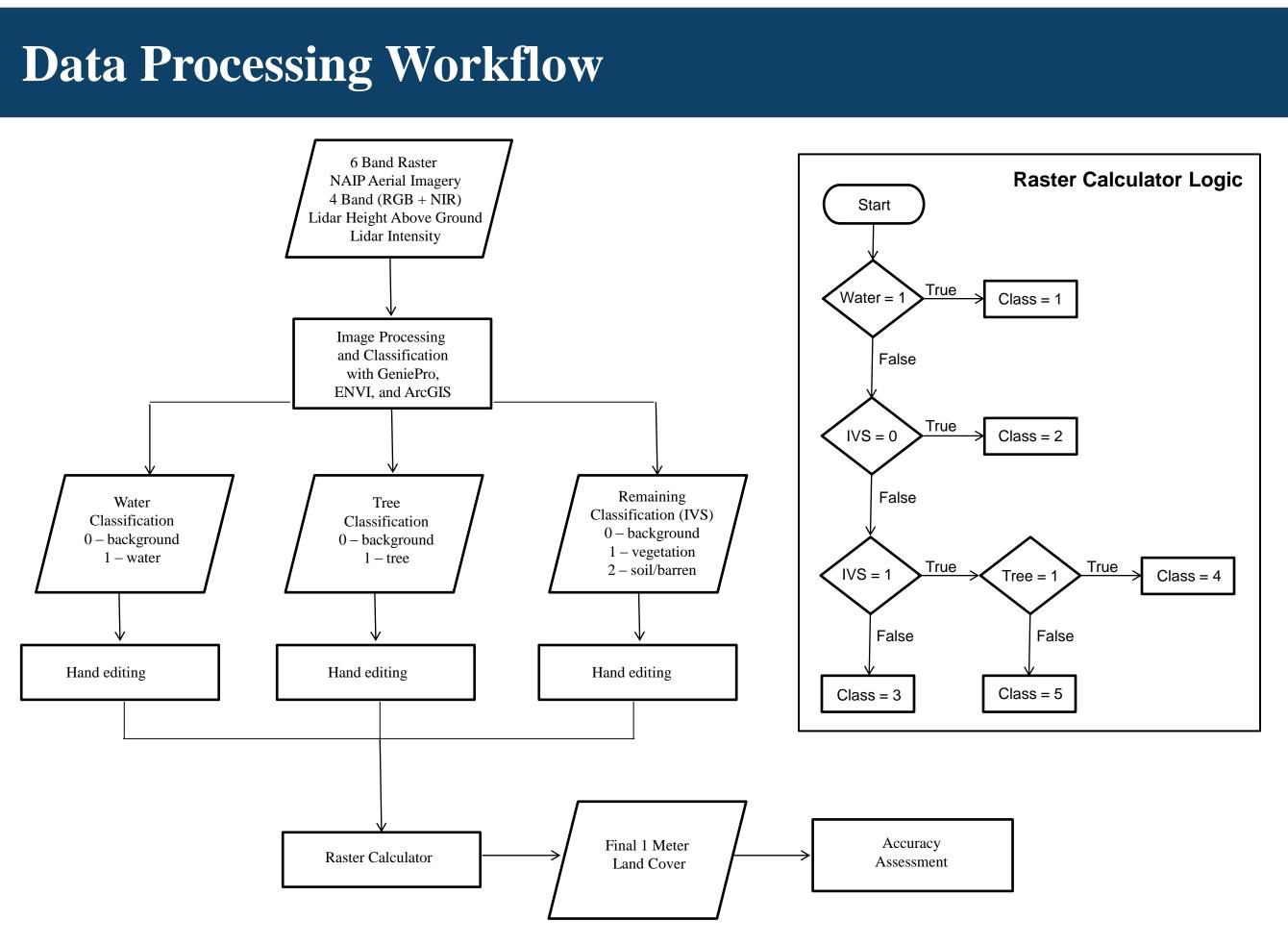
Background

Neighborhood parks and street trees are a vital part of urban green spaces that often occur at a sub-pixel scale in the 30 meter National Land Cover Dataset (NLCD). In order to include these features in urban sustainability and ecosystem service analyses, there is a need for finer spatial resolution land cover (LC) data. The National Agriculture Imagery Program (NAIP) collects 4-band (RGB+NIR) aerial imagery at 1 meter resolution on a one to three year cycle for most of the conterminous US. Multiple supervised machine learning classifications were created from NAIP imagery, each to identify a low number of classes. Using a hierarchical image processing workflow, high resolution urban land cover maps were created with these classes: water, impervious surface, soil & barren, trees & forest, grass & herbaceous, agriculture, and wetlands. The high resolution land cover maps serve as input for many of the EnviroAtlas data layers.

Machine Learning



The Genie Pro software (*www.observera.com*) uses machine learning algorithms to classify any number of classes in a single pass based on user-defined training regions. However, sequentially identifying a single class against a background tends to produce better results. Using a series of classification steps, in which one to three LC classes were mapped (rather than six or more at once), we implemented a hierarchical workflow to determine the final classification for each pixel.



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A Hierarchical, Machine Learning Approach to Meter-scale Urban Land Cover Mapping

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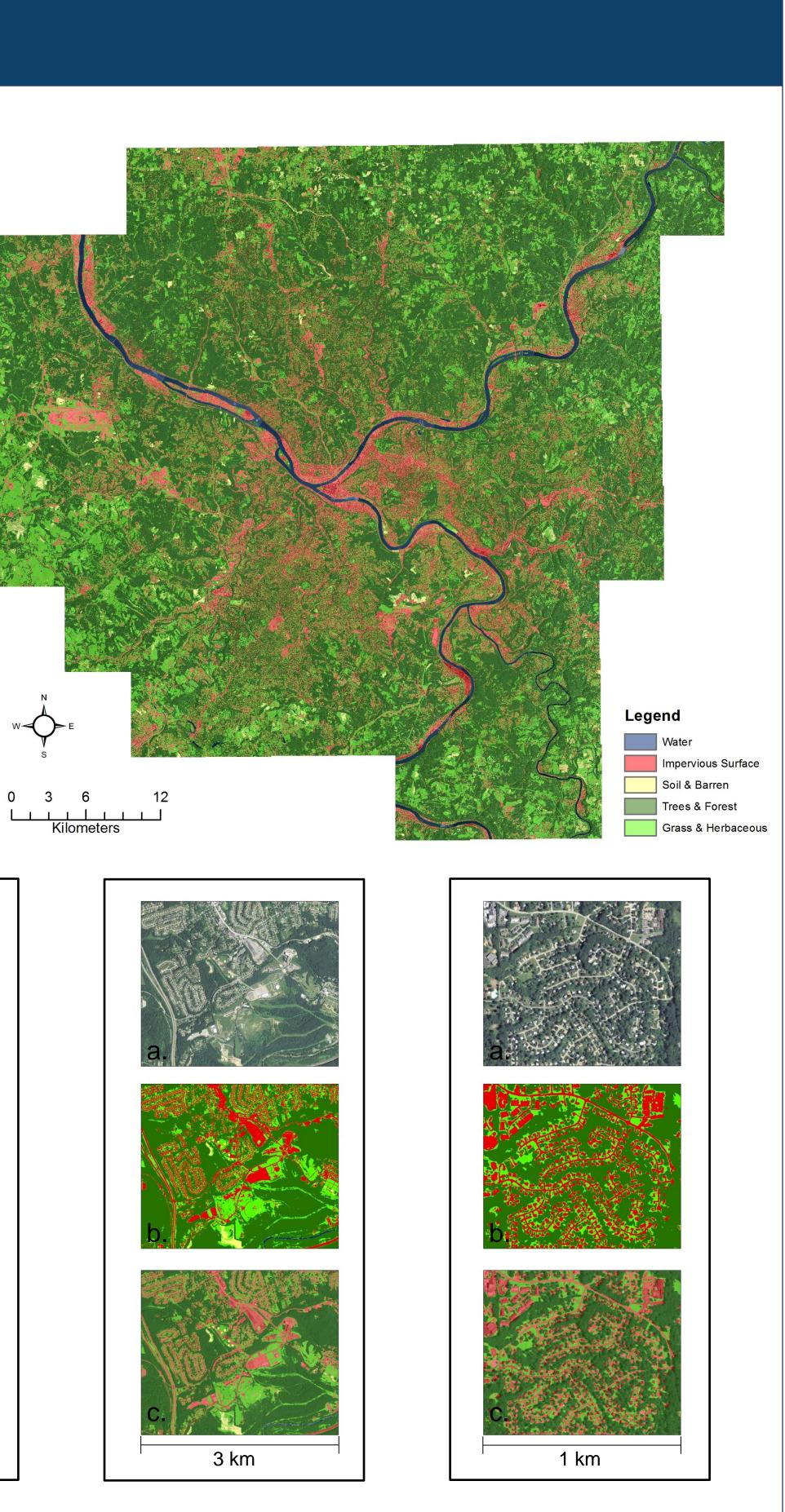
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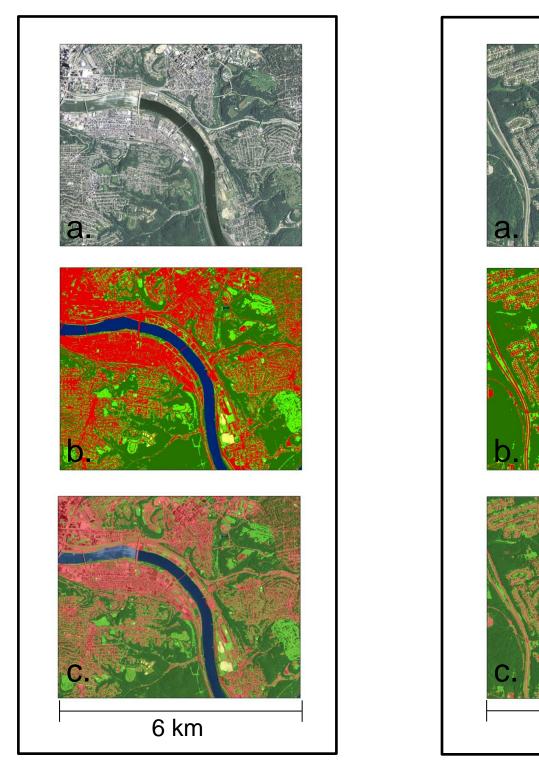
Results

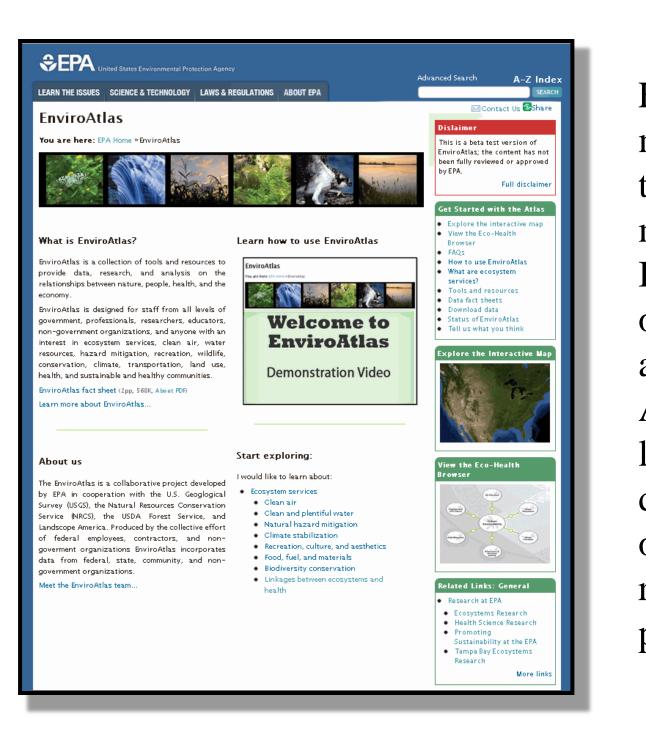
Pittsburgh, PA Land Cover 2010

Right: The large LC map covers the full study area as defined by the 2010 Census Urban Area.

Below: 3 areas at various scales to show the level of detail. For each group: a) NAIP imagery; b) classification; c) transparent classification over imagery.







EnviroAtlas is an easy to use multi-faceted decision support tool. In includes an interactive map, the Eco-Health Relationship Browser, research and information on ecosystem services, and analysis and statistical tools. Additionally EnviroAtlas provides links to other environmental decision support tools, as well as other resources on the relationships between ecosystems, people, and well-being.

Shadows

Shadows cast by vegetation and structures add significant noise to imagery at one meter resolution. Shadows typically appear at the edges of tall buildings and trees, and mottled within tree canopies, and are commonly misclassified as water or impervious surface. Using a binary step of classifying vegetation versus non-vegetation reduced these errors. The image on the right shows shadowed areas correctly classified as impervious (red) or vegetation (green).

Conclusions

- is availabe at little or no cost.
- dates posed challenges for classification of large regions.
- processing software.

- improved results, especially with tree canopy.

Acknowledgements

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• NAIP imagery was effective for this very high resolution land cover mapping, and

• Variable image radiometry (brightness) due to multiple flightlines and collection

• Vegetated and impervious surfaces not in shadow were easily separated.

• Shadows over impervious surfaces were distinguished from shadow over vegetated surfaces when using a simplified classification scheme and sophisticated image

• Manual correction of misclassified pixels was considerably quicker and easier to perform on multiple binary classifications than on a single multi-class layer. • Bare soil and impervious surfaces were difficult to separate.

• Adding LiDAR height-above-ground and intensity to the four NAIP bands

