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Arc Hydro: Model Impervious Surfaces using Sentinel Imagery

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1.0 Introduction

Arc Hydro consists of a data model, toolset, and workflows developed over the years to support specific geographic information system (GIS) implementations in water resources. The initial implementation of Arc Hydro was in 2002 with the data model, Arc Hydro book published by Esri Press, and an initial set of about 30 tools. Since then, Arc Hydro has been used in many projects, and in the process, new tools and workflows have been developed. There are more than 300 Arc Hydro tools now, and they continue to be expanded based on work in specific implementations.

This document describes the workflow execute by the Model Impervious Surfaces Using Sentinel Imagery tool. At a high level, this tool applies a pre-trained deep learning model to a Sentinel image. The deep learning model uses its learned relationships and dependencies between the Sentinel bands and the CORINE land cover classes to predict the landcover type for each cell in the Sentinel image. The landcover predictions are then reclassified into areas of pervious, impervious, and open water. Unless a user has already gathered their own Sentinel imagery and version of the deep learning model, they must be logged into ArcGIS online before running the tool.

1.1 Document History

Table 1. Document Revision History

Version	Description	Date
1	First version (GLO)	5/2021
2	Link to Deep Learning Frameworks Installer (GLO)	6/2021

1.2 Recommended reading

This is one of several Arc Hydro “how to” documents. The following are the other suggested reading, in the order of importance:

1. [Arc Hydro - Project Development Best Practices](#)
2. [Arc Hydro - ArcGIS Pro Project Startup Best Practices](#)

Users should also refer to documentation for the CORINE landcover classes:

1. [Copernicus Technical Library](#)

2.0 Solution Overview

The Model Impervious Surfaces Using Sentinel Imagery tool reclassifies a set of landcover predictions into pervious, impervious, and open water areas. The initial landcover predictions are generated using a pre-trained deep learning model. This tool is in the Arc Hydro Tools Pro toolbox supporting ArcGIS Pro 2.7 and later.

The steps executed are:

1. Authenticates into ArcGIS online or ArcGIS enterprise using the ArcGIS Pro credentials.
 - a. If the user provides both the Input Sentinel Raster and CORINE Deep Learning Model inputs, authentication is not necessary and will be skipped. Steps 2 and 3 will also be skipped.
2. Downloads a Sentinel-2 image that covers the extent of the Input Boundary Feature.
 - a. The download will include a natural color representation of the scene (RGB bands) and a full image (all 13 Sentinel bands). The full image will render as fully black.
 - b. Images are acquired from the [Living Atlas Sentinel-2 Image Service](#).
 - c. The selected image will be the most recent and cloud free scene available at the time of tool execution. If the user wants to characterize the landscape from a different timeframe, they can manually acquire a Sentinel-2 image (with all 13 bands) from the desired timeframe and use this instead.
3. Downloads the Land Cover Classification (Sentinel-2) deep learning [model](#).
 - a. This pre-trained deep learning model is available from the Living Atlas.
 - b. The model is trained to generate CORINE Land Cover class predictions from Sentinel-2 imagery.
4. Generates CORINE land cover class predictions for each pixel in the Sentinel-2 image.
5. Reclassifies the CORINE classes into impervious, pervious, and open water classes.
 - a. Users can modify the default set of CORINE classes that will be considered impervious.

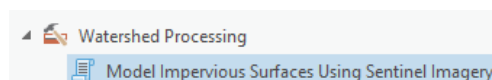


Figure 1. Model Impervious Surfaces Using Sentinel Imagery tool in the Watershed Processing toolbox of the Arc Hydro Pro toolbox.

2.1 Overview of Arc Hydro tools involved in Modeling Impervious Surfaces

This section presents a list of tools used in the extraction of impervious landcover using deep learning. Tools in this section are presented in the order they appear in the Arc Hydro Tools Python Toolbox and do not represent the order in which tools are used. The subsequent sections will explore specific workflows and which tools are used for them.

Toolset	Step	Tool	Description
(Core Software Tool) Image Analyst Tools / Deep Learning	Landcover predictions	Classify Pixels Using Deep Learning	This is an out-of-the-box tool that internally applies the pre-trained deep learning model to the Sentinel-2 image to predict the landcover classes. This tool requires that the deep learning frameworks are installed for ArcGIS Pro.
Watershed Processing	Reclassify the landcover predictions	Model Impervious Surfaces Using Sentinel Imagery	Characterizes a given watershed as pervious, impervious, or open water using deep learning and Sentinel-2 imagery.

3.0 Use Cases

3.1 Use Case 1: Modeling Impervious Surfaces with No Additional Inputs

1. Model Impervious Surfaces Using Sentinel Imagery

The screenshot shows the 'Geoprocessing' window for the tool 'Model Impervious Surfaces Using Sentinel Imagery'. The 'Parameters' tab is active, showing the 'Input Boundary Feature' set to 'Input Watershed Boundary'. The 'Impervious Classes' dropdown menu is open, displaying a list of classes: 11: Urban fabric, 12: Industrial, commercial and transport units, 13: Mine, dump and construction sites, and 33: Open spaces with little or no vegetation. A red arrow points from the 'Impervious Classes' dropdown to a list of default classes on the right. A purple arrow points from the 'Impervious Classes' dropdown to the text 'Default classes interpreted as impervious are shown'.

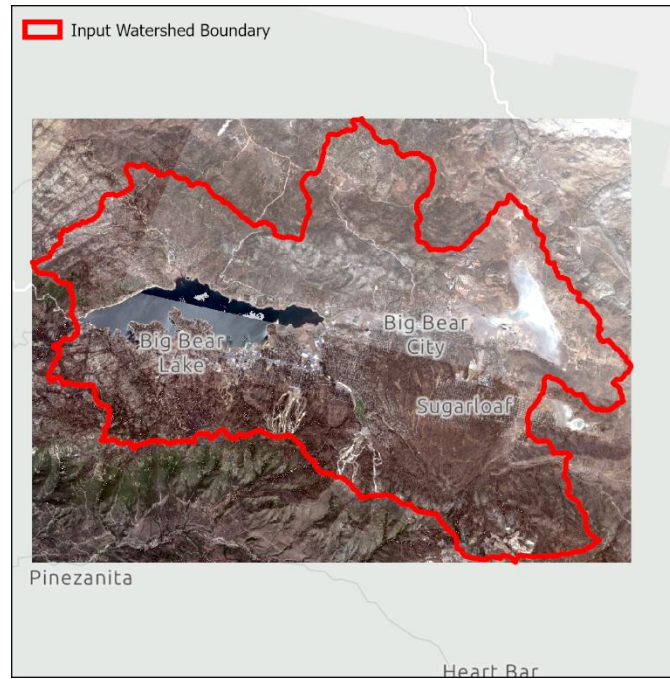
Choose the classes that will be reclassified to impervious

- 12: Industrial, commercial and transport units
- 13: Mine, dump and construction sites
- 14: Artificial, non-agricultural vegetated areas
- 21: Arable land
- 22: Permanent crops
- 23: Pastures
- 24: Heterogeneous agricultural areas
- 31: Forests
- 32: Scrub and/or herbaceous vegetation associations
- 33: Open spaces with little or no vegetation
- 41: Inland wetlands
- 42: Maritime wetlands

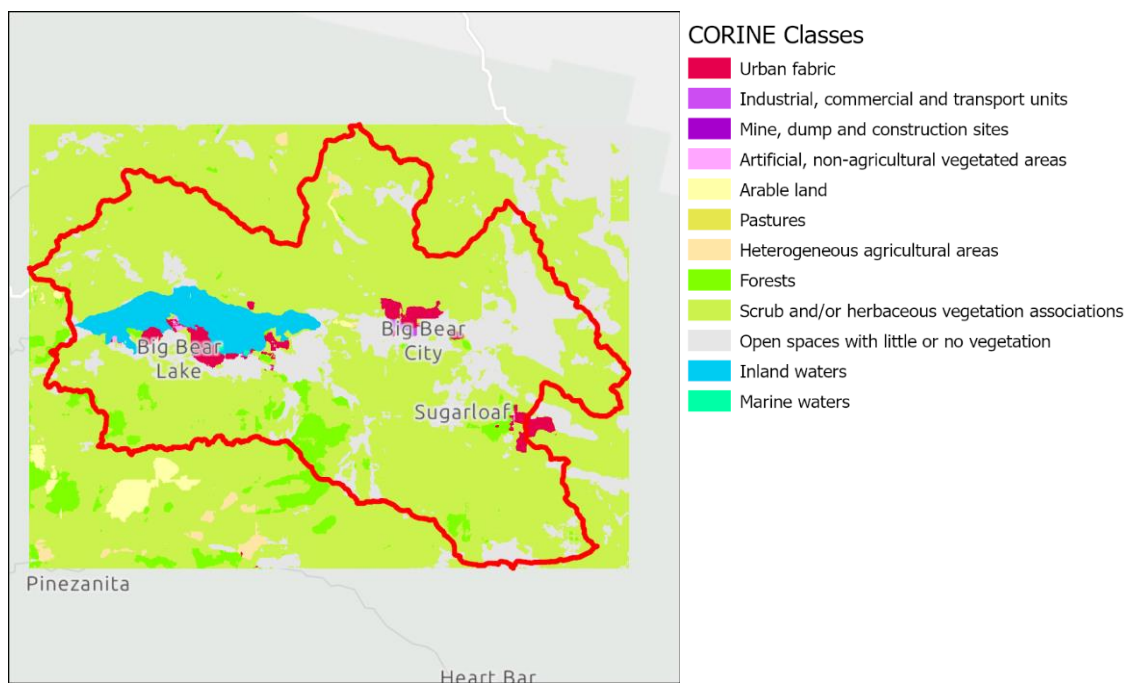
Default classes interpreted as impervious are shown

Outputs for this tool include:

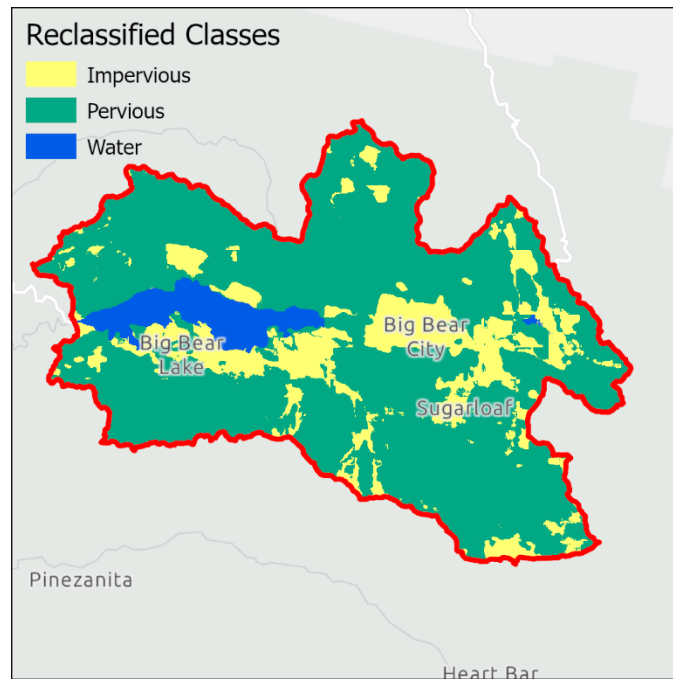
1. Natural color representation of the Sentinel-2 image downloaded for the boundary feature (full sentinel image output not shown).



2. CORINE Land Cover classifications predicted by the deep learning model.



3. Reclassification of the CORINE Land Cover classifications.



From Python:

```
import modelimpervioussurfaces

watershed = r'C:\Documents\ArcGIS\Projects\ModelImpervious\ModelImpervious.gdb\watershed'

out_lc_classes = r'C:\Documents\ArcGIS\Projects\ModelImpervious\Layers\CORINE_class.tif'

out_impervious = r'C:\Documents\ArcGIS\Projects\ModelImpervious\Layers\reclass.tif'

impervious_classes = [11, 12, 13, 33]

model_impervious = modelimpervioussurfaces.ModelImperviousSurfaces()


model_impervious.executeAH(watershed, impervious_classes, out_lc_classes, out_impervious, None,
None)
```


3.2. Use Case 2: Modeling Impervious Surfaces with Optional Inputs

1. Manually download the CORINE Land Cover deep learning model.

ArcGIS Pricing Map Scene Help Search Sign In

Land Cover Classification (Sentinel-2) Overview

 This deep learning model is used to perform land cover classification using Sentinel-2 imagery. The imagery will be classified into 16 classes based on a modified CLC Class Level II classification system as used by the CORINE Land Cover inventory.

Deep Learning Package by [esri_analytics](#)

Created: Feb 17, 2021 Updated: Apr 29, 2021 Number of items: 1

[Living Atlas](#)

Download

Download the deep learning model manually from Living Atlas in your web browser.

Description

This deep learning model is used to create a land cover product from Sentinel-2 imagery. The classified land cover will have the same classes as the [CORINE Land Cover 2018](#). Land cover describes the surface of the earth. The resultant land cover maps are useful for understanding urban planning, resource management, change detection, agriculture, and a variety of other applications where information related to earth surface is required.

Details

Size: 147 MB
★★★★★

[f](#) [t](#) [s](#)

This generic model is expected to work across Europe. It has been trained on the CORINE Land Cover (CLC) 2018 with the same Sentinel-

2. Download a Sentinel image for the boundary feature area. Recall that the full Sentinel image is required (i.e., all 13 bands). Options to do so include:

- Filter for and export an image from the [Living Atlas](#)
 - Imagery from the past 14 months is available here
- Use [EarthExplorer](#)

We filtered the Sentinel-2 Views in ArcGIS Pro to obtain a Sentinel image from a winter month. This will allow us to compare results outside of the growing season. To export the Sentinel image with all 13 bands, we followed this [Esri Community post](#) and queried an image from the month of January in 2019 (see image below).

Layer Properties: Sentinel-2 Views

General
Metadata
Source
Elevation
Selection
Display
Cache
Definition Query
Time
Range
Mosaic
Processing Templates

Definition Queries + New definition query

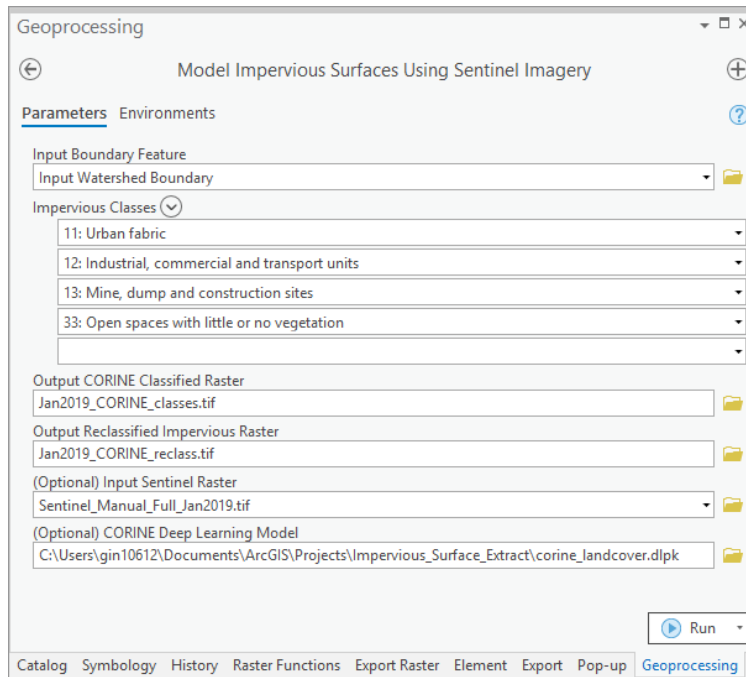
Query 1
datatype_format = 'Cloned' Or datatype_format IS NULL Edit

Query 2
acquisitiondate >= timestamp '19-1-1 17:14:45' And acquisitiondate <= timestamp '19-1-31 17:15:4' Edit

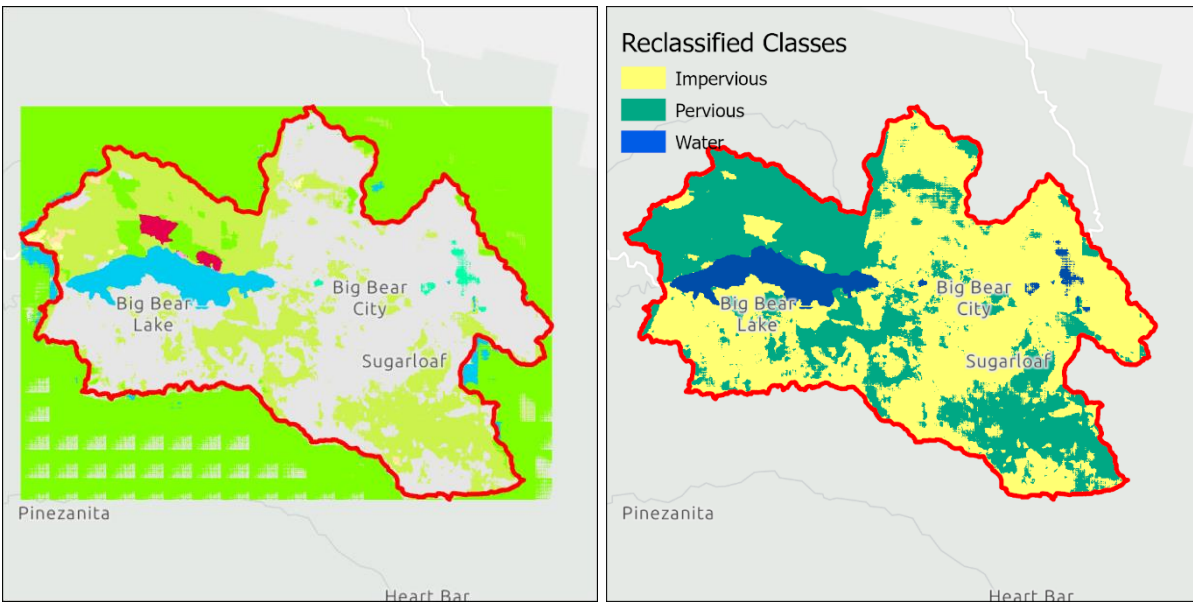
2 Queries Active definition query: Query 1

OK Cancel

3. Model Impervious Surfaces Using Sentinel Imagery



Results for this Sentinel scene are shown below. The deep learning model predicted a much larger impervious area for the winter scene. Looking at the intermediate CORINE output (left), we see this is due to the relatively larger area predicted as open space with little or no vegetation (grey pixels). This is likely a reflection of the reduced vegetative cover during January.



From Python:

```
import modelimpervioussurfaces

watershed = r'C:\Documents\ArcGIS\Projects\ModelImpervious\ModelImpervious.gdb\watershed'
out_lc_classes = r'C:\Documents\ArcGIS\Projects\ModelImpervious\Layers\Jan2019_CORINE_classes.tif'
out_impervious = r'C:\Documents\ArcGIS\Projects\ModelImpervious\Layers\Jan2019_CORINE_reclass.tif'
impervious_classes = [11, 12, 13, 33]

in_sentinel_image =
r'C:\Documents\ArcGIS\Projects\ModelImpervious\Layers\Sentinel_Manual_Full_Jan2019.tif'

in_dl_pkg = r'C:\Documents\ArcGIS\Projects\ModelImpervious\Layers\corine_landcover.dlpk'

model_impervious = modelimpervioussurfaces.ModelImperviousSurfaces()

model_impervious.executeAH(watershed, impervious_classes, out_lc_classes, out_impervious,
in_sentinel_image, in_dl_pkg)
```

4.0 Appendix

4.1 Processing Times

The information in this appendix is provided for reference only. The actual performance will vary greatly depending on the size and complexity of the underlying input data and function parameters. The numbers provided here are to be used as a relative measure of performance – which functions are “faster” and which are “slower”. Remember that raster processing tends to be non-linear for large rasters and that doubling the size of the raster will usually result in more than doubling the time it takes to process it.

The hardware used for processing will also have an impact on the performance. Fast hard-drive (HD) subsystem (preferably SSD) can double the improvement in performance compared to traditional HD. Make sure that both the scratch and permanent storage are pointing to the fast HD. For other processing suggestions, refer to “Arc Hydro - Project Development Best Practices” document.

The following results were collected using a machine with 16 GB of RAM. The largest extent of the testing dataset, located in inland southern California, included:

- Sentinel-2 image (10m resolution) with 2000 columns, 1500 rows, and 13 bands

Tool	Runtime
Model Impervious Surfaces Using Sentinel Imagery (Use Case 1)	10 minutes
Model Impervious Surfaces Using Sentinel Imagery (Use Case 2)	5 minutes, 30 seconds