USGS Western Ecological Research Station SFBE

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#### **Purpose/Objective:**

To classify, delineate, and digitize boundaries for key estuarine habitats using high resolution infrared aerial photographs and spatial software (ERDAS Imagine and ESRI ArcGIS). When repeated over time, changes to estuarine habitats can be quantified. Habitat and community composition data collected during vegetation surveys, as well as personal site knowledge can be used for ground-truthing and refinement of the spatial extent of each habitat classification.

These methods were written by Ben Gustafson (USGS) and applied to the Nisqually Aug 8<sup>th</sup>, 2010 aerial imagery, but the same general steps can be used to classify any imagery.

# These pages provide detailed information on the following:

Image Preparation Unsupervised Classification Supervised Classification Ground Truthing Classification Smoothing

Image Preparation Source Information Masking Resampling

Source Information: Company Name: Bergman Photographic Services Acquisition Date: August 8<sup>th</sup>, 2010 Time: unknown Altitude: unknown Scale: unknown Camera: unknown Focal Length: unknown Film/Digital: unknown

All images were received as 4-band .tif files with .823 x .823 foot pixel resolution. Individual tiles were provided as well as a mosaic.

#### Masking:

High spectral variability across an image will likely produce class confusion in the output file. One way to help reduce this





is to limit the number of pixels being considered during the classification process. We can do this by first creating a polygon mask that overlaps only the areas we are interested in classifying.

- 1. In ArcCatalog, create a new polygon shapefile or feature class with the same projection as the imagery you want to mask.
- 2. Start a new ArcMap session and add the mosaic and the masking polygon you created.
- 3. Create a polygon that covers the areas that you would like to include.
- Open the Extract by Mask tool (ArcToolbox > Spatial Analyst Tools > Extraction > Extract by Mask). NOTE: This tool requires the Spatial Analyst Extension to be turned on (Tools > Extensions > Spatial Analyst Checkbox).
- 5. Fill in the fields in the dialog box using the mosaic raster, your masking polygon, and an output raster name/location. **NOTE**: In order to retain all four bands in your output raster, you will need to choose the input mosaic raster file location from disk rather than the layer in the drop-down box. Using the layer name in the drop-down will result in a 3-band raster.

Click OK.

Þ	Extract by Mask		- 0	X	
	Input raster D:\Nisqually\NisquallyDelta_Aug_8_2010\Original_Images\usgs_nisqually_8Aug_mosaic.tif Input raster or feature mask data		•		*
	Mask_Polygon Output raster		•	2	
	D: Wisqually Wisqually Delta_Aug_8_2010 \Classifications \nisqually_8Aug_mosaic_mask.img			È	
					-
	OK Cancel E	nvironments	Show H	elp >>	·

6. You may need to add the output raster to the map. It should appear as if were clipped along the boundaries of the masking polygon you created.

**Resampling:** Another way to help reduce class confusion and make more spatially consistent classes is to resample high resolution imagery to a courser resolution (theoretically, fewer pixels should reduce spectral variability). However, before you choose a size to resample to, you need to determine your Minimum Mapping Unit (MMU). This will be the size of the



smallest feature you wish to capture in the classification. For the Nisqually imagery, we've decided our MMU will be the same as the vegetation surveys conducted in the restoration area: 0.5m x 0.5m.

- 1. Open the Resample tool (ArcToolbox > Data Management Tools > Raster > Raster Processing > Resample).
- 2. Fill out the fields in the Resample dialog box. In this case, we are resampling from 0.822906 feet to 1.64041 feet (.5 m = 1.64041 feet) so adjust the Output Cell Size value accordingly. Remember that the unit of measurement will be the same as the default unit of measurement of the raster's projection. Also, we will choose Cubic convolution as the Resampling Technique since satellite and aerial imagery are continuous data. NOTE: In order to retain all four bands in your output raster, you will need to choose the input mosaic raster file location from disk rather than the layer in the drop-down box. Using the layer name in the drop-down will result in a 3-band raster.

Click OK.

Input Raster	^
D:\Nisqually\NisquallyDelta_Aug_8_2010\Classifications\nisqually_8Aug_mos	aic_mask.img 🗾 🖻
Output Raster Dataset	
D: Wisqually WisquallyDelta_Aug_8_2010 \Classifications \nisqually_8Aug_mosaic_mas	k_resample.img 🗃
Output Cell Size (optional)	
1.64041	🗃
Resampling Techinque (optional)	
CUBIC	<b>▼</b>
	*
0	Cancel Environments Show Help >>

3. The output resampled raster should have the cell size you specified. To check this, add the file to your map, right-click on it in the table of contents and choose Properties. On the Source tab, look for the Cellsize (X,Y) property and the associated value.

#### Unsupervised Classification:

One method of creating a landcover classification is to let Erdas Imagine group the image pixels into categories automatically with only a few parameters. Although this method can produce poor results, it can provide you with an idea of which classes are being easily confused and you can then put more focus on those areas in the supervised classification. If your imagery contains a high contrast between the landcover types you would like to classify, the results of the unsupervised classification may be sufficient for your needs (or close enough to manually recode confused areas).

- 1. In Erdas Imagine, open the masked, resampled mosaic you created in the viewer.
- 2. On the main Erdas Imagine toolbar (not the viewer toolbar), click the Classifier button, then click the Unsupervised Classification button on the Classification menu that appears.
- 3. In the Unsupervised Classification dialog box, fill out the fields for the input raster, the output cluster layer (the output raster), and the output signature file.
- 4. Clustering Options: A couple of these values may need to be determined through a little trial and error. To begin with, make sure Initialize from Statistics is selected with the Number of Classes set to 15 (we are looking for classes that fall into water, mudflat, trees, lowland vegetation, bare earth, and

🖌 Unsupervised Classification (Isodata)						
Input Raster File: (*.img) nisqually_8aug_mosa 👻 🚰	Input Signature File: (*.sig)					
✓ Output Cluster Layer Filename: (*.img) nisqually_8aug_mosa ↓	<ul> <li>✓ Output Signature Set</li> <li>Filename: (*.sig)</li> <li>nisqually_8aug_mosa → 😹</li> </ul>					
Clustering	g Options:					
<ul> <li>Initialize from Statistics</li> <li>Use Signature Means</li> <li>Number of Classes:</li> </ul>						
Initializing Options	Color Scheme Options					
Processin	g Options:					
Maximum Iterations: 2 Convergence Threshold: 0 Classify zeros	Skip Factors:           .980         ×           Y:         1           Y:         1					
OK Batch AOI Cancel Help						

impervious, but some of these could easily be lumped together so we ask for more classes to determine this manually). The Initializing Options should be set to Diagonal Axis and 2 Std. Deviations. The Color Scheme Options should be set to Approximate True Color.

- 5. Processing Options: Again, these values may require some trial and error to determine the best settings. To start, we'll set it to run 20 iterations with a 0.980 Convergence Threshold. Then click OK. You should see the Process List appear with a progress meter for each iteration. This process can take several minutes or longer depending on the number of classes, iterations, and the convergence threshold specified.
- 6. Add the output classification raster you created to the Viewer. You'll notice that the colors used for the classes are similar to those of the mosaic. We will want to change these so we can analyze how accurate the classes are. Choose Raster > Attributes form the Viewer window to open the Raster Attribute Editor for the classification.
- 7. In the Raster Attribute Editor dialog, change the opacity of all of the items in the list to 0 except but one class. Highlight the one class left with an opacity of 1 by clicking on its Row number, then click on the color for that class. Choose yellow for the color. In the Viewer, you should see only the pixels associated with that class in yellow over the mosaic raster.

8. Now we want to try to determine which class the yellow pixels correspond to on the image. Zoom into an area on the image where the yellow pixels are present. Choose View > Arrange Layers from the Viewer menu to open the Arrange Layers window. Click on the classification layer to drop it below the mosaic layer and click Apply. Then click on the mosaic layer to drop it below the classification layer and click Apply. You can do this repeatedly to toggle one layer over the other and compare the class with the image and determine the landcover type it is associated with.

Alternately, you can also choose Utility > Swipe from the Viewer menu to open up the Swipe dialog. This allows you to swipe back and forth between the two layers and compare them that way.



9. Evaluate the class throughout the image. If the pixels appear to correspond to a single class type throughout, you can enter that as the class name in the Raster Attribute Editor and assign an appropriate color. If the class seems to correspond to multiple landcover types, you can enter "confused" as the class name or more specific information such as "bare earth mudflat confused."

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€ D	🔛 117 🖿	🔁 Laye	r Number: 1					
Row	Histogram	Color	Red	Green	Blue	Opacity	Class_Names	
0	29045339		0	0	0	0	Unclassified	
1	2447412		0.24	0.05	0	0	Class 1	
2	3476618		0.07	0.25	0.17	0	Class 2	
3	3467764		0.14	0.44	0.38	0	Class 3	
4	5067894		0.32	0.31	0.26	0	Class 4	
5	1533708		0.62	0.02	0.04	0	Class 5	
6	6127750		0.46	0.39	0.38	0	Class 6	
7	8074958		0.33	0.59	0.52	0	Class 7	
8	7722397		0.47	0.63	0.61	0	Class 8	
	5042598		1	1	0	1	Mudflat Bare Earth Impervious Confused 3	
10	3869039		0	1	0	0	Vegetation Grass Shrubs 4	
11	3337084		0	1	0	0	Vegetation Grass Shrubs 3	
12	6007946		0	1	0	0	Vegetation Grass Shrubs 2	
13	3145641		0	1	0	0	Vegetation Grass Shrubs 1	
14	3160315		1	1	0	0		
15	1838655		1	1	0	0	Mudflat Bare Earth Inmpervious Confused 1	

10. Change the opacity for the class you just evaluated to 0 and set a different one to 1. Evaluate this class using the method described above. Repeat this process for each class until you have determined the landcover type (or class confusion) for all of them. Be sure to save your changes to the attributes. NOTE: If the accuracy of the classes appear to be consistently wrong, you can go back and re-run the unsupervised classification with different parameters. For example, you might consider adding more classes to see if Erdas Imagine can find more spectral separation and hopefully reduce the amount of confusion.

11. Using the unsupervised classification settings above, it appears that Mudflat, Bare Earth, and Impervious features are consistently confused. These are areas that we now know we will need to focus on next in a Supervised Classification.

12. **NOTE**: Another method you can use to try to reduce class confusion is to run an unsupervised classification on the confused classes only. You can do this by creating an AOI (Area of Interest) file of the confused classes and adding it to the unsupervised classification settings, telling Erdas Imagine to create a subset that looks only at the pixels of the confused classes. The output



signature file may have better separation among those areas that were initially confused. This signature file can then be appended to the initial one you created to create a cleaner classification.

In the case of this imagery, creating an AOI subset only produced a couple of cleaner classes, so we will move on to the Supervised Classification.

#### Supervised Classification:

This method requires the user to manually create training sites throughout the imagery for each class. Erdas Imagine uses these training sites as representative spectral signatures to determine the classes for the rest of the image. Although this method is the most time-consuming, it can also produce the best overall results in the output file.

- 1. In Erdas Imagine, open the resampled, masked, mosaic you created in the viewer.
- 2. On the main Erdas Imagine toolbar, click the Classifier button, then click the Signature Editor button on the Classification menu that appears. You should see the Signature Editor window appear.

🔏 Signature Editor (No File)	1000			
File Edit View Evaluate Feature Classif	/ Help			
🖨 ⊡ +५ +→ ≣५ Σ ∖∖ 🕍 🖣	7 🔺			
Class # > Signature Name Color	Red Gree	n Blue Value	Order Count Prob. P I H	A FS
				E

- 3. Open the AOI tools by choosing AOI > Tools on the Viewer menu.
- 4. Zoom in to area of the imagery that has an example one of the landcover types you are interested in classifying.

- 5. From the AOI toolbar, choose the Create Polygon AOI 🔟 tool.
- 6. On the image, draw a polygon large enough to overlap several pixels of a single landcover type. Double-click to finish the AOI polygon.
- 7. On the Signature Editor window, click on the Create New Signature(s) from AOI button. This adds the signature to the list based on the pixels within the AOI polygon you created in the image. Edit the Signature name to reflect the landcover type for which it corresponds.
- 8. Gather several training sites per landcover type (preferably 10 to 30 each). The goal is to capture as much of the spectral variability within each class as possible. If there are landcover types that you know will have spectral similarities, make extra training sites for each. Make a few shadow classes to try and separate between shadow types (shadow trees, shadow impervious, etc.). Be sure to save the signature file often.



- 9. Once you have completed the training sites, the signature file can be used to create a supervised classification. On the Classification menu, click the Supervised Classification button. In the dialog window that appears, select the mosaic image and the signature file you created for the inputs. Designate and output file name and leave the other default settings (None for Non-parametric Rule and Maximum Likelihood for the Parametric Rule). Click OK.
- 10. Add the output raster to the Viewer display and open the Raster Attribute Editor (Raster > Attributes).
- 11. In the Raster Attribute Editor dialog, change the opacity of all of the items in the list to 0 except but one class. Highlight the class with the opacity of 1 by clicking on its Row

K Supervised Classification	<b>X</b>				
Input Raster File: (*.img)	Input Signature File: (*.sig)				
nisqually_8aug_mosaic_r 👻 🖨	nisqually_8aug_mosaic_ 👻 🗃				
Classified File: (*.img) isqually_Baug_mosaic_:	Distance File Filename: (*.img)				
Fuzzy Classification	2 🚔 Best Classes Per Pixel				
Dec	ision Rules:				
Non-parametric Rule:	None				
Overlap Rule:	Parametric Rule 🔹				
Unclassified Rule:	Parametric Rule 💌				
Parametric Rule:	Maximum Likelihood 🔹				
Classify zeros 🔲 Use Probabilities					
OK Batch	AOI Cancel Help				

number, then click on the color for that class. Choose yellow for the color. In the Viewer, you should see only the pixels associated with that class in yellow over the mosaic raster.

12. Evaluate the accuracy of the class throughout the image. If the class appears to be classified correctly, give it an appropriate color and note that it is good. If the class seems to correspond to multiple landcover types, you can enter "confused" after the class name or more specific information such as "Mudflat 13 confused some Impervious."

🔀 Raster	Raster Attribute Editor - nisqually_8aug_mosaic_super02.img(:Layer_1)								
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Row	Histogram	Color	Red	Green	Blue		Class_Names	-	
0	35294941		0	0	0		0 Unclassified		
1	159635		1	1	0		1 Mudflat 13 confused some Impervious		
2	157279		0.47	1	0.98		0 Impervious 2		
3	700		0.83	1	1		0 Impervious 3		
4	13464		1	1	1		0 Impervious 4		
5	40860		0.69	1	1		0 Impervious 5		
6	208305		0.57	0.98	0.9		0 Impervious 6		
7	160585		0.47	0.93	0.85		0 Impervious 7	-	
•								F.	

13. Change the opacity for the class you just evaluated to 0 and set a different one to 1. Evaluate this class using the method described above. Repeat this process for each class until you have determined how well the image was classified. It is likely that you will have some classes that have poor separation, such as shadow areas.

In the Nisqually imagery, classification improved in some areas, but there was still a lot of confusion between mudflat and impervious features. Environmental conditions at the time that the imagery was acquired (position of the sun, high reflectance, cloud cover, etc.) can cause some



landcover types to have similar spectral values. In order to get a high quality classification, you will need to refine the output.

14. One way to improve the output of the supervised classification is to add more training sites to the signature file and re-run the classification. However, if you already have a large number of training sites per landcover type, the output may not improve.

Another option is to manually recode areas of confusion. This is especially good for areas with only a slight confusion. For example, in the masked version of the Nisqually image, there are very few impervious features that exist. By going back to the Mudflat classes that had some Impervious pixels, we can go back and select those pixels and recode them as Impervious. This will in turn remove the confused Impervious cells from the other classes, making the classification more accurate.

- 15. Set the opacity of one of the classes with slight confusion to 1. Do the same for a good class that is same type as the one with the slight confusion (not the predominate class type). For example, I am starting with a Mudflat class with slight Impervious confusion and a good Impervious class.
- 16. Zoom into an area with where class confusion is occurring. In this case, I am zooming into a building that has some mudflat class pixels over it. This is the area that you are going to recode to match the non-confused class. Choose AOI > Tools to bring up the AOI toolbar and Raster > Tools to bring up the Raster toolbar.
- 17. From the AOI toolbar, click on the Region Grow AOI tool. Choose AOI > Seed Properties from the Viewer menu. In the Region Growing Properties dialog box, set the Area to a very high number, such as 1,000,000. This will allow the Region Grow AOI tool to select a larger number of connected pixels at a time. Close the Region Growing Properties dialog and then click on an area of confused pixels that you want to reclassify. The Region Grow AOI should produce a selection of all the adjacent or connected pixels from where you clicked that are

the same class. For large, contiguous areas (such as roads or structures) you can also use the Create Polygon AOI 201 to delineate sections to recode.

18. Once you've created an AOI selection, click on

the Fill Area 论 tool on the Raster toolbar. In the Area Fill dialog that opens, change the Fill With field to the same number as the Row of the class you want to reclassify the selected area as. In this example, I am setting the value to match the good Impervious class. Click Apply. You can leave this dialog open and make another selection with the Region Grow AOI or Create Polygon AOI tools and repeat this process until you have changed all the confused areas that you

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Function: Constant 🗸	
	Apply
Fill With 3.000	Close
	Help
Exclude Value: 0.000	Preview

want to reclassify. If this eliminates the class confusion, rename the class in the Raster Attribute Editor to reflect this. Repeat this process for any other slightly confused classes.



Cleaning up the classification can take a considerable amount of time. If some of the spectral values are too similar between classes, it may be impossible to make a perfect classification. You will need to decide at what point the classification has reached an acceptable level of accuracy.

- 19. Once you have finished cleaning up the classification, you will probably want to merge the classes that are of the same type. You can do this in Erdas Imagine using the Recode tool (Interpreter > GIS Analysis > Recode), however the Reclassify tool in ArcMap seems to provide a better output.
- 20. Start ArcMap and add your classification to the map. Write down your final class types (Water, Vegetation, Impervious, etc.) and assign a number value to each (1, 2, 3, etc.).
- 21. Open the Reclassify tool (ArcToolbox > Spatial Analyst Tools > Reclass > Reclassify). In the dialog window, set the classification as the input raster and the Reclass field to Value.

In the Reclassification fields, change each of the New values to match the class types you wrote down above. This can be a little tricky, so I find it helps to have the attribute window of the raster open at the same time to refer to (you can also add a Reclass field to the attribute table and put the new values there first for reference).

		x
Input raster D:\Nisqually\NisquallyDelta_Aug_8_ Reclass field Value	2010\Classifications\nisqually_8aug_mosaic_super01.img	* *
Reclassification       Old values       90       91       92       93       94       95       NoData       Output raster       D: WisquallyWisquallyDelta_Aug_8_20:       Change missing values to NoData (or content of the second seco	W values       Classify         7       Unique         4       Unique         5       Add Entry         5       Delete Entries         NoData       Delete Entries         Reverse New Values       Precision         0\Classifications\nisqually_8aug_mosaic_super01_reclassify.img       2         xtional)       2	3
	OK Cancel Environments Show Hel	

22. Once the reclassified image has been created, right click on it in the Table of Contents and choose Properties. On the Symbology tab, set the Show field to Unique Values and the Value Field to VALUE. You should see a Count column in the main section for each class value. This is the count of pixels that comprise each of the classes. We can use this to find the area of area of each class type by multiplying the count value by the area of each pixel (the count

value is also in the attribute table). In the case of the August 2010 Nisqually imagery, the pixel area is:  $1.64041 \times 1.64041 = 2.6909$  square feet.

#### **Ground-Truthing**:

Classification accuracy is subject to the interpretation of the analyst and the settings used to calculate classes. Ideally, classifications should be checked against the real ground conditions. Random areas, or areas where accuracy is most important, should be inspected on the ground and compared to the classification values. This is especially important in situations where more specific landcover types are being classified, such as different kinds of vegetation. Ground-truthing should be done as close to the time of image acquisition as possible.

#### **OPTIONAL**

#### **Classification Smoothing**:

To reduce salt and pepper effects and improve the overall appearance of your images, you can run a majority filter on your classifications. This will use a moving neighborhood window to smooth out the cell values throughout the image.

- 1. From the main Erdas Imagine menu, choose Interpreter > GIS Analysis > Neighborhood.
- 2. In the Neighborhood Functions window, set the input and output file names. Choose a neighborhood size in the Neighborhood Definition section (3x3, 5x5, or 7x7). The larger the window size, the more smoothing that will occur. Make sure the Function is set to Majority and click OK.

Neighborhood Functions		٢					
Input File: (*.img) Output I	File: (*.img) Neighborhood Definition:						
d:/nisqually/nisquallyd 👻 📸 🛛 saic_super01_f	filter.imd 🗸 🖨 🖉 k1' 🖉 k2' 🖉 k3' 🖉 k4' 🖉 k5' 🖉 k6' 🖉	k7 <sup>.</sup>					
Coordinate Type: Data Type:	♥ k1: ♥ k2: ♥ k3: ♥ k4: ♥ k5: ♥ k6: ♥	k7:					
Map     Input: Unsigned 8 bit	♥ k1: ♥ k2: ♥ k3: ♥ k4: ♥ k5: ♥ k6: ♥	k7:					
File     Output: Unsigned 8 bit	▼ k1· ▼ k2· ▼ k3· ▼ k4· ▼ k5· ▼ k6· ▼	k7·					
Subset Definition: From In	nquire Box	k7!					
UL X: 1082943.99 📑 LR X: 1098478	8.67 🔹 🖉 k11 🖉 k21 🖉 k31 🖉 k41 🖉 k51 🖉 k61 🖉	k7I					
UL Y: 656555.67 🚔 LR Y: 640386.	.14 🔹 🖉 k1' 🖉 k2' 🖉 k3' 🖉 k4' 🖉 k5' 🖉 k6' 🦉	k7'					
Function Definition:	Size: 7x7 💌 🗹 include 🗌 exclude						
Function: Majority	Ignore Zero in Stats.						
Use all values in computation	Apply function at all values						
OK Batch View	w AOI Cancel Help						





- 3. The resulting output raster should contain more solid landcover types throughout the image. Keep in mind that although the image may appear cleaner, you are probably lowering the overall accuracy of your classification.
- August 8, 2010

**Nisqually Delta Supervised Classification** 

#### Landcover Area (Square Meters) 4,145,563.25 Water 4,747,288.00 Vegetation 942,374.00 Trees 37,929.00 Shadow 5,196,432.00 Mudflat 69,696.50 Impervious

941,337.25 Bare Earth