

# Final Report: <u>Temporal Changes of La Parguera Reefs</u> <u>as Detected with Remote Sensing</u>

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#### Abstract

The purpose of this research is the use of Remote Sensing to determine changes through time of the distribution of the coral reefs of La Parguera. For this, three aerial photograph (1936, 1977, 1999) and two images (CASI, IKONOS) of the area were processed using IsoData, K-Means and Minimum Distance methods. IsoData and K-Means results were not the best. The Minimum Distance had the best results and changes through time in the distribution of the reefs are detected. Remote Sensing, then, can be used to detect changes of the distribution of the coral reefs of La Parguera.

### Introduction

Coral reefs are considered to be the rain forest of the oceans due to its biodiversity. Around 25 % of the world's marine species are located in these ecosystems (CNN, 1997). Due to anthropogenic and natural causes, coral reefs are in jeopardy. Human activities such as fishing, sailing, costal development, agriculture, deforestation, among others, are responsible for most of the coral reefs degradation (CoRIS, 2004).

Fishermen exploit coral reefs because of the diversity found in them. This over fishing causes the overgrowth of sea grass. Since there is no sufficient fish to eat it, coral reefs get covered with algae and can not receive enough sunlight needed for photosynthesis of the *Zooxanthellae* that live in them and provide color and additional food for the corals (Ocean World, 2004). Boats that pass by these ecosystems can affect them by breaking them apart and by spilling their fuels, polluting the area. Construction of hotels and tourist's developments causes runoffs, sedimentation and sewage spills. Eventually these end in the ocean bringing high

amounts of nutrients and turbidity, decreasing the sun light that gets into water and causing eutrophication (overgrowth of algae in water due to the excess of nutrients causing the go away of the *Zooxanthellae* from the coral because they do not receive sufficient light for their photosynthesis). Corals then bleach and can die (CoRIS, 2004).

Agriculture and deforestation have more or less the same effects on them. The chemicals needed for maximum cultivation go to the coasts when it rains and eutrophicate the areas. When there is a clear up of green areas, the land that used to be covered with grass, trees, plants, etc., is no more protected or stable. Rain falls and the unprotected terrain mixes with the water carrying sediments and nutrients to the coast.

Natural causes such as hurricanes, greenhouse gases, low tide, etc., also affect negatively the coral reefs. The drastic rains, involving hurricanes, bring huge quantities of sediments that end in the oceans and powerful waves, which can break corals apart. The greenhouse gases effect warms the water. *Zooxanthellae* can not live in such temperature conditions and they leave the corals. Then corals suffer from bleaching. Lowering in the tide rise can also cause bleaching. Coral reefs are exposed receiving direct rays of light that expel the Zooxanthellae (CoRIS, 2004).

Corals reefs around the world are affected by these stresses and Puerto Rico's coral reefs are no exception. In this research the area of study is La Parguera in Lajas. The coral reefs in this location are considered to be the healthiest of coastal Puerto Rico, because of their high abundance of living coral. This is caused by some factors (Bruckner, A., Carlo, M., Morelock, J., Ramírez, W., 2001).

Mangroves are natural barriers that prevent sediments from being carried by the run offs directly to reefs. There are no close rivers that could cause eutrophication and/or a lowering of

the sunrays into the water column due to sediments. These help to maintain certain healthy status in the coral reefs. Anyhow, they are being affected by the rapid urbanization that increases the sedimentation in rainy seasons, bringing nutrients from the terrain and the sewage systems, increasing the growth of sea grass. Hurricanes also have been contributing to their degradation due to the excess of sediments and their destructive waves (Bruckner, A., Carlo, M., Morelock, J., Ramírez, W., 2001). Having all these factors in mind it is possible to come up with some ideas for a research that could help to determine changes in the reefs of La Parguera.

Remote sensing can not determine which are exactly the reasons for the changes of the reefs, but it is possible to infer what may have cause it by establishing the relation of the urban development with the reefs changes. The sensors can detect construction and development of the area (land use and land cover changes); and we could find out if these are being increasing by comparing the images and photos. The urban development can cause run offs with sediments and nutrients eutrophicating the reefs. This eutrophication may cause changes in the distribution ecosystem.

# **Objectives**

This research is concentrated in the study of the coral reefs of La Parguera from 1930's to present. Images of these reefs taken by IKONOS, CASI (Compact Airborne Spectrographic Imager), and three aerial photos (1936, 1977 and 1999) were analyzed to see if there are visible and spectral changes in the reefs. It tries to determine changes through time of the distribution of the communities that compose the ecosystem. The research will find out if several remote sensors can easily capture these changes and which is the best sensor for it. This will help to select the right tool for future studies. Also, it pretends to determine whether or not there are

noticeable distribution differences in the spectral response in the reefs between the images and photos, and which of the unsupervised (IsoData and K-Means) methods used to detect the changes are similar to the supervised method. Therefore, the hypothesis of my work is that remote sensing techniques can be used to detect the distribution changes of coral reef's communities through time.

## Methodology

Three images of La Parguera, one from IKONOS without being pre-processed (Figure 5), the other one a de-glinted IKONOS (figure 6), CASI (Figure 4) and aerial photograph from 1999 (figure 3) were facilitated by the GERS (Geological and Environmental Remote Sensing Laboratory). Aerial photographs from 1936 (figure 1) and 1977 (figure 2) were provided by "La División de Fotogrametría de la Autoridad de Carreteras". ENVI (Environmental of Visualization Images) was the program used for the processing of the images and photos.

Т	able 1:IKONOS and C	CASI sensor description	ns
Sensor	Orbit	Resolution	Bands
IKONOS	Sun synchronous,	High spatial	Panchromatic,
	every 98 minutes	resolution (1m, 4m)	multispectral
CASI	Airborne combines	High spatial	Hyperspectral 288
	the best of aerial	resolution (5m)	channels (400- 915
	photography and		nm)
	satellite imagery.		



Figure 1: Aerial photograph of 1936



Figure 3: Aerial photograph of 1999



Figure 5: IKONOS image



Figure 2: Aerial photograph of 1977



Figure 4:CASI image



Figure 6: IKONOS de-glinted image

For the purpose of the investigation, IKONOS and CASI images with three aerial photos of La Parguera reefs were processed using ENVI. There was a need to use a de-glinted IKONOS image (figure 5) because the original one had sun glint problems that affected the outcome result of the unsupervised images, not letting to identify the real classification in the water. The aerial photos of different years (1936, 1977, and 1999) were compared in between and with the images (image interpretation) to determine visible changes of the reefs. Georeferencing and atmospheric corrections were performed. This is needed to have a better quality of the images and photos classification. The CASI image (figure 10) and the aerial photos (figure 7, 8 & 9) were georeferenced based on coordinates of the IKONOS image (figure 1), because it has the highest spatial resolution. The de-glinted image was not georeferenced, because it was already. Georeferencing consist in removing geometric errors and rectified to a real-world coordinate system (Decision Support, 2003). After the georeferenciation, they were re-sized using the subset application to have the same study area in all of them (figures: 11, 12, 13, 14, 15 & 16).

Atmospheric corrections were done to correct atmospheric attenuation and some scattering effects to obtain a close estimate of a surface spectral radiance (CIS, 2004). For these atmospheric corrections the dark subtract method was applied to each of the images and aerial photos, except for the de-glinted IKONOS that already was pre-processed with the de-glinted method. Then, they were classified using supervised and unsupervised methods. Supervised classification is defined as the processes using sample of known identity to classify pixels of unknown data. The analysts have control of the data and they select the regions of interest (ROIs), which are the areas that are going to be classified to distinguish one object from the other. Minimum distance was the method used for this classification. It classifies using the ROI's closest pixels.

Unsupervised classification is defined as the identification of natural groups, or structure, within multispectral data (Introduction to Remote Sensing, 2002). The researcher does not directly control this classification. ENVI gives classes according to the similarities of the spectral response. With this method one class may represent several objects, meaning that the classification can not distinguish between one another. IsoData and K-Means were the unsupervised methods used, because they were the only methods available in the ENVI program. Unsupervised and supervised classifications were compared. After image processing, the data were analyzed using statistical procedures.

Forty points were selected for each of the photos and images to evaluate the changes through time in the different classes.



Figure 7: Georefereced aerial photograph of 1936



Figure 8: Georefereced aerial photograph of 1977



Figure 9: Georefereced aerial photograph of 1999



Figure10: Georefereced CASI image



Figure 11: Re-sized aerial photograph of 1936



Figure 13: Re-sized aerial photograph of 1999



Figure 15: Re-sized IKONOS image



Figure 12: Re-sized aerial photograph of 1977



Figure 14: Re-sized CASI image



Figure 16: Re-sized IKONOS de-glinted image

### **Results and Discussion**

#### I. Description for the aerial photographs and images

Dark subtract was used to remove the effect of the atmosphere that affects negatively the outcome classification of the images and photos. Comparing these images and photos (figures: 17, 18, 19, 20 & 21) it is possible to see changes. The most noticeable change is found in the main land. Urban development largely increased since 1936 until 1977, and from this year to 2001 (IKONOS image) such changes are less dramatic. There is no floating house in the coastal line in 1936, but in 1977 until the present they are found. The second most noticeable change is located in the coral reefs were the sand had been reduced and taken by the sea grass community. In the coastline, just north of Magueyes Island there is a wetland area in 1936 that have been reduced, specifically their water content. The amount of mangroves has increase in this area. These wetlands have been covered with houses, buildings, and roads since 1977.

Mangroves in general are not well detected in 1936, mainly because of the quality of the aerial photograph. This demonstrates the limitation of using black and white photos for this type of work. Areas with sandy bottom in 1936 were covered by mangroves in 1977 and until present.



Figure 17: Dark Subtract aerial photograph of 1936



Figure 19: Dark Subtract aerial photograph of 1999



Figure 18: Dark Subtract aerial photograph of 1977



Figure 20: Dark Subtract CASI image



Figure 21: Dark Subtract IKONOS image

#### II. Supervised Classification

The supervised classification distributes the classes much better than the unsupervised methods. Although classes sometimes get confused with other classes, boundaries are better established. The small urban area of the 1936 photo can be distinguished and compared with the urban areas photos and images of the following years. The Minimum Distance method is adequate to determine the change in distribution of the urban development and it is clear that it have increased dramatically.

In almost all the photos and images the mangroves are well defined (figure: 22, 23, 24, 25, 26 & 27). The 1936 aerial photograph confronts problems distinguishing the mangroves areas, but the Minimum Distance method minimizes the problems and the mangroves are better identified. Mangroves are not good defined in the 1977 photo as it was with the unsupervised methods. For the 1999 photo mangroves tend to blend with shallow areas that were classified under the Mangroves class. The Minimum Distance is a good method to determine the changes in distribution for the mangroves of La Parguera reefs, because this classification shows that they have increased.

This classification does not significantly confuse sand with any other class as seen in the IsoData and K-Means methods. Comparing the classification it is clear that sand, sea grass and coral are well defined. Sand has been reduced; meanwhile the sea grass community has expanded. In the other hand, is difficult to determine if the coral community distribution has changed through time.



Figure 22: Minimum Distance of the aerial photograph of 1936



Figure 23: Minimum Distance of the aerial photograph of 1977



Figure 24: Minimum Distance of the aerial photograph of 1999



Figure 25: Minimum Distance CASI image



Figure 26: Minimum Distance IKONOS image



Figure 27: Minimum Distance IKONOSde-glinted image



Tabl	e 2:Color int	erpretation for a	erial photograj classifie	oh and images of cation	Minimum Distan	ce supervised
Class	Color	1936	1977	1999	CASI	IKONOS
Deep water	Red	Deep water	Deep water, coral reefs, sea grass and mangrove s	Deep water	Deep water and sea grass	Deep water
Shall ow water	Thistle 2	Shallow water, sea grass, coral reefs and mangroves	Shallow water, sea grass, coral reefs and mangrove s	Shallow water	Shallow water	Shallow water and sea grass
Sand	Yellow	Sand and urban areas	Sand and coral reefs	Sand and coral reefs	Sand	Sand
Sea grass	Yellow 3	Sea grass, coral reefs and shallow water	Sea grass, coral reefs and shallow water	Sea grass and shallow water	Sea grass and shallow water	Sea grass, coral reefs and shallow water
Mang roves	Green 1	Deep water and mangroves	Sea grass and mangrove s	Shallow water, coral reefs and mangroves	Mangroves and coral reefs	Mangroves
Coral	Coral	Coral reefs, shallow water and sea grass	Coral reefs, shallow water and sea grass	Coral reefs, shallow water, mangroves and sea grass	Coral reefs, shallow water, mangroves and sea grass	Coral reefs and urban areas
Urban areas	Orange 1	Coral reef and sand	Urban areas and sand	Sea grass and shallow water	Urban areas	Urban areas and sand

#### III. Unsupervised Classification

The IsoData method gave 7 classes (figures: 28, 29, 30, 31 & 32) while the K-Means method gave 5 classes (figures: 33, 34, 35, 36 & 37). These classes do not represent the same classes or colors used in the supervised classification. The distribution of classes using these methods depends on the spectral responses for each pixel in the photo or image.

The IsoData method classified the areas according to their similarities in the spectral response or brightness. Apparently, for this classification, classs 7 (maroon color) includes urban areas, areas with small amounts of vegetation, sand and coral. Urban development and areas with small amounts of vegetation are difficult to distinguish because they are producing the same spectral response. Comparing the photos and images (figures: 28, 29, 30, 31 & 32), it is noticeable that class 7 has extended in the main land, but is difficult to know if such increment is due to the urban development, the amount of vegetation, or both. Consequently, the method does not help to determine increments through the years in the urban development of La Parguera. Similar results found with the sand and the corals. are They live together and the unsupervised method does not differentiate them. Therefore, if there is an increment in the distribution of the corals community or the sand, it is difficult to detect it with the IsoData classification. It is noticeable that the distribution of this class has declined through the years, indicating that maybe the sand, the corals or both have being decreasing since 1936 to the present. Comparing with non-processed images and photos it is possible to infer that classification results are mainly produced by changes in sand distribution, especially decreasing in the keys through the years.

Mangroves are not well seen in the aerial photograph of 1936, because of the limitation confronted with the black and white photos. Consequently, the IsoData method did not make a good classification for it. The best classified mangroves are located in the main land and the color green (class 2) and blue (class 3) represent them mostly. They get confused with the blue of the shallow water, therefore is difficult to know their distribution. For the aerial photograph of 1977 mangroves are more defined and they stand out from the rest for having a greenish color and for having a red (class 1) shadow next to it. Mangroves are not defined in the 1999 photo, they tend to get confused with the cyan (class 5) and yellow (class 4) that also represents shallow water. For the CASI image something similar happens. Mangroves are not distinguishable from the rest. They tend to form part of the classes 4 and 5 that represents shallow water. IKONOS image present very distinguishable mangroves represented mostly with the color green. Determining changes through time in their distribution with this unsupervised method is not easy to achieve due to problems with the classification where mangroves were not distinguishable and tended to be confused with other classes.

Cyan (class 5), yellow (class 4) and magenta (class 6) represent sea grass in the keys in all images and photos. Sea grass could be confused with shallow water, but they are away from the sea grass, except for the CASI image. Making a comparison we could determine that the sea grass community distribution has been increasing and occupying the sand's area. Thus, the most significant change in distribution through time provided by this classification in La Parguera is the changes in sea grass.



Figure 28: IsoData aerial photograph of 1936



Figure 30: IsoData aerial photograph of 1999



Figure 32: IsoData IKONOS image



Figure 29: IsoData aerialphotograph of 1977



Figure 31: IsoData CASI image



,	Table 3: Col	or interpretation	for aerial phot	ograph and imag	es of IsoData clas	sification
Class	Color	1936	1977	1999	CASI	IKONOS
1	Red	Deep water	Deep water	Deep water	What is not part of the image	Mask
2	Green	Deep water and mangroves	Deep water, sea grass and mangroves	Deep water	Deep water and corals	Mangroves
3	Blue	Shallow water and mangroves	Shallow water, mangroves and sea grass	Shallow water, sea grass and mangroves	Deep water, coral, sea grass and mangroves	Deep water and mangroves
4	Yellow	Shallow water, submerged areas in coastal line and sea grass	Shallow water, sediment and sea grass	Shallow water, mangroves and sea grass	Shallow water, mangroves, coral and sea grass	Deep water, sea grass, and mangroves
5	Cyan	Submerged areas in coastal line land, sea grass and shallow water	Sea grass and shallow water	shallow water sea, grass and mangroves	Shallow water, sea grass, coral and mangroves	Shallow water and sea grass
6	Magenta	Submerged areas in coastal line, shallow water and sea grass	Shallow water, mangroves, coral and sea grass	Shallow water, mangroves. Coral and sea grass	Shallow water, sea grass and mangroves	Shallow water, coral and sea grass
7	Maroon	Sand, coral, urban areas, submerged and little vegetation areas in main land	Sand, coral, urban areas and areas with little or without vegetation	Sand, coral, urban areas and areas with little or no vegetation	Sand and urban areas	Sand, urban areas, coral areas little or no vegetation

K-Means classification shows a pattern similar to the one found in the urban areas, in the areas covered with sand and corals of the IsoData classification. It confuses the urban areas, the sandy regions and the corals and gives them the same class (class 5, blue). It is not a good method to determine increments in the urban areas. This must be, due to the quality (spatial resolution) of the photos and images, because if all the photos and images had a high spatial resolution the classification must had an outcome such as the one for the IKONOS image. In this image the urban areas are well seen and distinguished.

Sand and corals are not differentiated and since they are next to the other is difficult to determine which of them had changed their distribution through the years. There has been a decrease in their distribution as the years have passed, but which of them had been reduced is not determined with this method.

Mangroves were not well classified in the 1936 due to the same reason for which the IsoData method did not classify it well. It is for the limitation of the black and white photos. Only in the 1977 photo and the IKONOS image provide a clear distribution of the mangroves represented by the class 2 (green), but if it has increased or decreased through the year can not be determined, because of the lack of well classified photos or images of other years.

Sea grass are represented by the classes 4 (yellow) and 3 (blue) and its distribution change can be determined because it does not get confused with any other mentioned above. The only problem notable is the confusion presented in the CASI image where the sea grass blends with shallow water represented by the class 1. According with this classification the sea grass distribution has been increasing through the years. This method is adequate to determine changes through years in the distribution of the sea grass of La Parguera reefs.



Figure 33: K-Means of the aerial photograph of 1936



Figure 35: K-Means of the aerial photograph of 1999



Figure 37: K-Means IKONOS image



Figure 34: K-Means of the aerial photograph of 1977



Figure 36: K-Means CASI image



Т	able 4: Colo	r interpretation f	or aerial photo	graph and image	s of K-Means cla	ssification
Class	Color	1936	1977	1999	CASI	IKONOS
1	Red	Deep water	Deep	Deep water	What is not	Mangroves
			water,		part of the	and mask
			mangrove		photo	
			s and sea			
			grass			
2	Green	Shallow	Shallow	Shallow	Deep water,	Mangroves
		water, sea	water, sea	water	sea grass,	
		grass and	grass and		coral reefs	
		mangroves	mangrove		and	
			S		mangroves	
3	Blue	Shallow	Shallow	Shallow	Shallow	Deep water,
		water, sea	water, sea	water, sea	water, sea	sea grass and
		grass,	grass and	grass and	grass, coral	mangroves
		mangroves	mangrove	mangroves	reefs and	
		and	S		mangroves	
		submerged				
		coastal line				
4	Yellow	Submerged	Shallow	Shallow	Shallow	Shallow
		coastal line	water,	water, coral	water, coral	water, sea
		and shallow	coral reefs	reefs sea	reefs, sea	grass coral
		water	sea grass	grass and	grass and	reefs and
			and	mangroves	mangroves	mangroves
			mangrove			
	~		S			
5	Cyan	Urban areas,	Urban	Urban areas,	Urban areas,	Urban areas,
		sand, coral	areas,	coral reefs	coral reefs	shallow
		reefs, sea	coral reefs	and sand	and sand	water, coral
		grass and	and sand			reefs, areas
		submerged				with little or
		coastal line				no vegetation
						and sand

#### Conclusion

Changes in time in the distribution or spectral response of the ecosystems of La Paraguera coral reefs can be dectected using Remote Sensing. These changes are detected using the Minimum Distance unsupervised classification. From the three classification used, this one detected more changes in distribution of the coral reefs of La Parguera. It detected urban areas increment, expansion in the sea grass communities and mangroves, and sand reduction. Though corals were distinguished, for this supervised method is difficult to detect changes through time in their distribution. The IsoData and K-Means only detected expansion in the sea grass communities. Therefore, none of these unsupervised methods are similar to the Minimum Distance method.

The best sensor for this work was the IKONOS because of its high spatial resolution. This make easier to differentiate from one class to the other. Camera sensor can also be used, but details are lost due to its low spatial resolution. Black and white photos are not the best option for this kind of work because of their limitation of color that confuses and loses information. CASI is not the best sensor from them all, because frequently, due to its low spatial resolution, details of the area were hard to see.

Although, processing the images with the classifications already mentioned, can detect changes through time in the distribution in La Pargueras coral reefs, the truth is that these changes are easily detected without the processing. Thus, classifying the images can help to determine the changes, but these changes can be detected comparing the original aerial photographs and images.

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# Appendix

	Appendix 1: Pixel description for the IsoData unsupervised classification													
Pix	Pixel	Coordinates			Aerial p	hotograph				In	ages			
el #	n		19	36	1	.977	19	999	C	ASI	Ik	KONOS		
			class	color	class	color	class	color	class	color	class	color		
1	1588, 4719	17°57'36.43 <sup>°</sup> N, 67°02'52.46 <sup>°°</sup> W	2	green	1	red	2	green	3	blue	4	yellow		
2	619, 4180	17°57'53.85` N, 67°03'25.45" W	1	red	1	red	2	green	3	blue	4	yellow		
3	1658, 4643	17°57'38.91 N, 67°02'50.09" W	2	green	1	red	2	green	3	blue	4	yellow		
4	905,45 40	17°57'42.18 N, 67°03'15.69" W	2	green	1	red	2	green	3	blue	4	yellow		
5	2066, 4337	17°57'48.91 N, 67°02'36.25'' W	2	green	3	green	2	green	5	cyan	5	cyan		
6	1744, 3941	17°58'1.75"N, 67°02'47.24" W	3	blue	2	green	3	blue	5	cyan	3	blue		
7	623, 4358	17°57'48.06''N, 67°03'25.29 W	3	blue	1	red	3	blue	4	yellow	3	blue		
8	2513, 3403	17°58'19.33` N, 67°02'21.17'' W	4	yellow	5	cyan	4	yellow	5	cyan	5	cyan		
9	2182, 4393	17°57'47.10`N, 67°02'32.31'' W	3	blue	2	green	2	green	5	cyan	4	yellow		
10	548, 4097	17°57'56.55` N, 67°03'27.87" W	3	blue	1	red	2	green	4	yellow	5	Cyan		
11	1749, 5182	17°57'21.39` N, 67°02'46.93` W	7	maroon	7	maroon	6	magent a	7	maroon	7	maroon		
12	2027, 5032	17°57'26.30 N, 67°02' 37.50" W	5	cyan	4	yellow	5	cyan	4	yellow	6	magenta		
13	2436, 4581	17°57'41.01 N, 67°02'23.65'' W	7	maroon	4	yellow	5	cyan	6	magent a	7	maroon		
14	2869, 4309	17°57'49.90` N, 67°2'8.97" W	7	maroon	7	maroon	4	yellow	7	maroon	6	magenta		
15	2081, 5018	17°57'26.75` N, 67°02'37.81" W	7	maroon	5	cyan	5	cyan	4	yellow	5	cyan		
16	2438, 4746	17°57'39.82` N, 67°02'29.29 W	4	yellow	6	magent a	5	cyan	5	cyan	5	cyan		
17	1268, 5325	17°57'23.42` N, 67°03'3.15" W	7	maroon	7	maroon	7	maroon	7	maroon	7	maroon		
18	1828, 5092	17°57'24.32` N, 67°02'44.26''W	7	maroon	7	maroon	5	cyan	5	cyan	6	magenta		
19	279 <del>6</del> , 4423	17°57'46.19 N, 67°02'44.26" W	7	maroon	7	maroon	4	yellow	6	magent a	5	cyan		
20	3000, 4266	17°57'51.32` N, 67°02'4.52"W	7	maroon	7	maroon	4	yellow	7	maroon	6	magenta		

$ \begin{array}{c c c c c c c c c c c c c c c c c c c $		Appendix 1: Pixel description for the IsoData classification continuation													
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Pix	Pixel	Coordinates			Aerial	photograp	h			In	nages			
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	ei#	n			1936	1	977		1999	0	CASI	Iŀ	KONOS		
21         3075, 3360         17'58'20.79' N, 67'02'2.07' W         3         blue         2         green         3         blue         4         yellow         2         green           22         2961, 3382         17'58'20.79' N, 67'02'2.07' W         3         blue         1         red         4         yellow         4         yellow         3         blue           23         3407         67'02'2.07' W         3         blue         1         red         4         yellow         4         yellow         3         blue           23         1983, 17'58'18.79' N, 4073         5         cyan         7         maroon         7         green         4         yellow <th></th> <th></th> <th></th> <th>class</th> <th>color</th> <th>class</th> <th>color</th> <th>class</th> <th>color</th> <th>class</th> <th>color</th> <th>class</th> <th>color</th>				class	color	class	color	class	color	class	color	class	color		
21       3075, 3360       17*58*20.79' N, 67'02'2.07" W       3       blue       2       green       3       blue       4       yellow       2       green         22       2961, 3382       17*58*20.03 N, 67'02'5.94" W       3       blue       1       red       4       yellow       4       yellow       3       blue         23       1983, 3417       67'02'5.94" W       5       cyan       7       maroon       7       for 0''''''''''''''''''''''''''''''''''''															
3360         67'02'2.07' W         C <thc< th=""> <thc< th=""> <thc< th="">         &lt;</thc<></thc<></thc<>	21	3075,	17°58'20.79 N.	3	blue	2	green	3	blue	4	vellow	2	green		
22         2961, 3382         17'58'20.03' N, 67'02'5.94' W         3         blue         1         red         4         yellow         4         yellow         3         blue           23         1983, 3417         17'58'18.79' N, 67'02'2.9.15' W         5         cyan         7         maroon         7		3360	67°02'2.07" W				C				5		U		
3382         67'02'5.94" W         C <thc< th=""> <thc< th=""> <thc< th="">         &lt;</thc<></thc<></thc<>	22	2961,	17°58'20.03 N,	3	blue	1	red	4	yellow	4	yellow	3	blue		
23       1983, 3417       17*58*18.79*N, 67*02'29.15**W       5       cyan       7       maroon       7		3382	67°02'5.94" W												
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	23	1983,	17°58'18.79 N,	5	cyan	7	maroon	7	maroon	7	maroon	7	maroon		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		3417	67°02'29.15" W												
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	24	2463,	175757.54 N,	4	yellow	1	red	I	red	2	green	4	yellow		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	25	2718	17°58'7.06 N	4	vellow	4	vellow	6	magenta	4	vellow	2	green		
261261, 534617*57*16.0° N, 67*03'3.50" W5cyan2green4yellow3blue5cyan271918, 526517*57*18.71 N, 67*02'4118" W4yellow3blue7maroon5cyan7maroon281879, 524017*57*18.71 N, 67*02'4118" W4yellow3blue7maroon5cyan7maroon281879, 524017*57*36.36 N, 67*02'42.51" W7maroon2green7maroon5cyan7maroon292493, 472417*57*36.36 N, 67*02'21.70" W3blue7maroon7maroon7maroon7maroon302964, 438617*57*47.41 N, 67*02'5.73" W4yellow5cyan6magenta5cyan7maroon312960, 317917*58*26.56 N, 67*02'4.26" W7maroon7maroon7maroon7maroon7maroon7maroon321952, 317917*58*26.56 N, 67*02'40.26" W7maroon7maroon7maroon7maroon7maroon7maroon331565, 317917*58*25.22 N, 67*02'40.26" W7maroon7maroon7maroon7maroon7maroon7maroon7maroon331565, 317917*58*25.21 N, 67*02'40.26" W	25	3781	67°02'14 16" W	-	yenow	-	yenow	0	magema	-	yenow	2	green		
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	26	1261.	17°57'16.00 N,	5	cyan	2	green	4	vellow	3	blue	5	cyan		
$\begin{array}{cccccccccccccccccccccccccccccccccccc$		5346	67°03'3.50" W		5		C		5				5		
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	27	1918,	17°57'18.71 N,	4	yellow	3	blue	7	maroon	5	cyan	7	maroon		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		5265	67°02'4118" W												
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	28	1879,	17°57'19.51 N,	7	maroon	2	green	7	maroon	5	cyan	7	maroon		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		5240	67°02'42.51" W												
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	29	2493,	17°57'36.36 N,	3	blue	7	maroon	7	maroon	7	maroon	7	maroon		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	20	4/24	6/ 02 21./0 W	4	vallary	5	a1/0 P	6	maganta	5	0110.00	7	maganta		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	30	2904, 1386	17 37 47.41 N, $67^{\circ}02'5 73'' W$	4	yenow	3	cyan	0	magenta	3	cyan	/	magenta		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	31	2960	17°58'24 03 N	3	blue	7	maroon	4	vellow	7	maroon	7	maroon		
32         1952, 3179         17°58'26.56' N, 67'02'40.26''W         7         maroon	51	3260	67°02'5.99" W	5	orae	,	muroon		y en o n	,	marcon	,	marcon		
3179         67°02'40.26"W         maroon         7          7	32	1952,	17°58'26.56 N,	7	maroon	7	maroon	7	maroon	7	maroon	7	maroon		
33 1565, 17°58'25.22 N, 7 maroon 7 maroon 7 maroon 7 maroon 7 maroon 7 maroon		3179	67°02'40.26"W												
2210 67°02'52 41" W	33	1565,	17°58'25.22 N,	7	maroon	7	maroon	7	maroon	7	maroon	7	maroon		
5217 0/ 02 53.41 W		3219	67°02'53.41" W												
34 1983, 175813.55 N, 2 green 7 maroon 7 maroon 7 maroon 7 maroon 7 maroon	34	1983,	17°58'13.55 N,	2	green	7	maroon	7	maroon	7	maroon	7	maroon		
<u>35/9 6/0236.16 W</u> <u>35 1504 17°58'226' N 2 blue 7 mercen 9 mercen 7 mercen 9 mercen </u>	25	3579	6/ 02/36.16" W	2	blue	7	maraan	7	maraan	7	maraan	7	maraan		
33 1394, $173822.08$ N, $5$ blue / matoon / matoon / matoon / matoon	33	3297	67°02'52 41" W	3	blue	/	maroon	/	maroon	/	maroon	/	maroon		
36 1970 1758218 88 N 5 cvan 7 marcon 7 marcon 7 marcon 7 marcon 7	36	1970	17°58'18 88 N	5	cvan	7	maroon	7	maroon	7	maroon	7	maroon		
3415 67°02'39.62" W	50	3415	67°02'39.62" W	<sup>c</sup>	eyun	,	muroon	,	inditotii		marcon	,	marcon		
37 2965, 17°58'20.13 N, 3 blue 1 red 3 blue 4 yellow 3 blue	37	2965,	17°58'20.13 N,	3	blue	1	red	3	blue	4	yellow	3	blue		
3380 67°02'5.81" W		3380	67°02'5.81" W												
38         1625,         17°58'23.60 N,         6         magenta         7         maroon         7 <th7< th=""> <th7< th=""></th7<></th7<>	38	1625,	17°58'23.60 N,	6	magenta	7	maroon	7	maroon	7	maroon	7	maroon		
3269 67°02'51.36" W		3269	67°02'51.36" W												
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	39	2611,	17°58'23.34 N,	6	magenta	7	maroon	7	maroon	7	maroon	7	maroon		
<u>3280 0/021/.85 W</u>	40	3280	0/021/.85"W		mag t-	7		7		7	magr	7			
40 1022, 1/36 24.0/18, 0 magenta / maroon / maroon / maroon / maroon / maroon	40	3236	1/ 38 24.0/ N, 67°02'51 47" W	0	magenta	/	inaroon	/	maroon	/	maroon	/	maroon		

	Appendix 2: Pixel description for the K-Means unsupervised classification													
Pix	Pixel	Coordinates			Aerial	photograp	h			Ima	iges			
el#	n		1	936	1	977	19	999	C	ASI	IKO	ONOS		
			class	color	class	color	class	color	class	color	class	color		
1	1588,	17°57'36.43 N,	1	red	1	red	1	red	3	blue	3	blue		
	4719	67°02'52.46" W												
2	619,	17°57'53.85 <sup>°</sup> N,	1	red	1	red	2	green	3	blue	3	blue		
2	4180	6/°03°25.45″ W	1		1		1		2		2	1-1		
3	1658,	17 57 58.91 N, 67°02'50 09" W	1	rea	1	rea	1	rea	2	green	3	blue		
4	905.45	17°57'42.18 N.	1	red	1	red	2	green	2	green	3	blue		
	40	67°03'15.69" W						0		0	_			
5	2066,	17°57'48.91 N,	2	green	2	green	2	green	3	blue	4	yellow		
	4337	67°02'36.25" W												
6	1744,	17°58'1.75"N,	3	blue	2	green	2	green	3	blue	4	yellow		
7	623	07 02 47.24 W	2	green	1	red	2	green	3	blue	4	vellow		
'	4358	67°03'25.29 W	2	Breen	1	icu	2	green	5	orae	-	yenow		
8	2513,	17°58'19.33 N,	3	blue	3	blue	3	blue	3	blue	4	yellow		
	3403	67°02'21.17" W										-		
9	2182,	17°57'47.10` N,	2	green	2	green	2	green	3	blue	4	yellow		
10	4393	67°02'32.31" W	-		1	1				11	4	11		
10	548, 4007	1/ 5/ 50.55 N, 67°03'27 87" W	2	green	1	rea	2	green	3	blue	4	yellow		
11	1749	17°57'21.39 N	5	cvan	5	cvan	4	vellow	5	cvan	5	cvan		
	5182	67°02'46.93` W	-	• ) ••••	-	• ) ••••	-	<i>j</i> ====	-	• ) ••••	-	• ) ••••		
12	2027,	17°57'26.30 N,	4	yellow	3	blue	4	yellow	3	blue	4	tellow		
	5032	67°02' 37.50" W												
13	2436,	17°57'41.01 N,	5	cyan	3	blue	4	yellow	5	cyan	5	cyan		
14	4581	6/ 02 23.65 W	5	ovan	5	ovan	3	blue	5	ovan	4	vellow		
14	4309	67°2'8.97" W	5	Cyan	5	Cyan	5	onuc	5	Cyall	4	yenow		
15	2081,	17°57'26.75 N,	2	green	2	green	4	yellow	5	cyan	4	yellow		
	5018	67°02'37.81" W		-		-		-		-		-		
16	2438,	17°57'39.82 N,	5	cyan	3	blue	5	cyan	5	cyan	5	cyan		
17	4746	67°02'29.29 W	2	11	2	11	2	1.1	2		2	11		
1/	1268,	1/ 5/ 25.42 N, 67°03'3 15" W	3	blue	3	blue	5	blue	2	green	5	blue		
18	1828.	17°57'24.32` N	5	cvan	5	cvan	3	blue	3	blue	4	vellow		
-	5092	67°02'44.26"W		. ,	-	. ,	-					J		
19	2796,	17°57'46.19 N,	5	cyan	5	cyan	3	blue	4	yellow	4	yellow		
	4423	67°02'44.26" W	<u> </u>											
20	3000,	17°57'51.32 N,	5	cyan	5	cyan	3	blue	5	cyan	4	yellow		
	4266	67 02 4.52 W												

$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$		Appendix 2: Pixel description for the K-Means classification continuation													
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	Pix	Pixel	Coordinates			Aerial	photograp	h			In	nages			
Lam         class         color         class         c	ciπ	n		1	.936	1	977	19	999	C	ASI	IF	KONOS		
1 $1$ <th></th> <th></th> <th></th> <th>class</th> <th>color</th> <th>class</th> <th>color</th> <th>class</th> <th>color</th> <th>class</th> <th>color</th> <th>class</th> <th>color</th>				class	color	class	color	class	color	class	color	class	color		
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$															
3360         67'02'2.07" W         C <thc< th=""> <thc< th="">         C</thc<></thc<>	21	3075.	17°58'20.79` N.	2	green	2	green	3	blue	3	blue	2	green		
$\begin{array}{cccccccccccccccccccccccccccccccccccc$		3360	67°02'2.07" W	_	8	_	8	-		-		_	8		
3382         67 02 5 94" W	22	2961,	17°58'20.03 N,	3	blue	4	yellow	3	blue	4	yellow	3	blue		
23         1983, 3417         17'Sr'18.79' N, 67'02'29.15" W         4         yellow         5         cyan         5         cyan <t< td=""><td></td><td>3382</td><td>67°02'5.94" W</td><td></td><td></td><td></td><td>-</td><td></td><td></td><td></td><td>-</td><td></td><td></td></t<>		3382	67°02'5.94" W				-				-				
3417         67'02'29.15" W         4         yellow         5         cyan         5         cyan <t< td=""><td>23</td><td>1983,</td><td>17°58'18.79` N,</td><td>4</td><td>yellow</td><td>5</td><td>cyan</td><td>5</td><td>cyan</td><td>5</td><td>cyan</td><td>5</td><td>cyan</td></t<>	23	1983,	17°58'18.79` N,	4	yellow	5	cyan	5	cyan	5	cyan	5	cyan		
24       2463, 17*57*57.54 N, 4       yellow       5       cyan       5		3417	67°02'29.15" W												
40/3       6/0/02/2/9       70 <td>24</td> <td>2463,</td> <td>17°57'57.54 N,</td> <td>4</td> <td>yellow</td> <td>5</td> <td>cyan</td> <td>5</td> <td>cyan</td> <td>5</td> <td>cyan</td> <td>5</td> <td>cyan</td>	24	2463,	17°57'57.54 N,	4	yellow	5	cyan	5	cyan	5	cyan	5	cyan		
23       2118, 3781       67 02 12.16 W       5       blue       5       cyan       5       cyan       5       cyan       5       cyan         26       1261, 3346       17'57'16.00 N, 5265       4       yellow       5       cyan	25	40/3	6/ 02 <sup>-22</sup> ./9 <sup>-7</sup> W	2	1-1	2	1.1	4	11	2	1-1	2			
26       1761       0702       171.600       N       4       yellow       5       cyan	25	2718,	1/ 58 /.00 N, 67°02'14 16'' W	3	blue	3	blue	4	yenow	3	blue	2	green		
10       11 <t< td=""><td>26</td><td>1261</td><td>17°57'16.00 N</td><td>4</td><td>vellow</td><td>5</td><td>cyan</td><td>5</td><td>cvan</td><td>5</td><td>cyan</td><td>5</td><td>cyan</td></t<>	26	1261	17°57'16.00 N	4	vellow	5	cyan	5	cvan	5	cyan	5	cyan		
27       1918, 5265       17*57'18.71'N, 67'02'4118" W       3       blue       2       green       5       cyan       4       yellow       5       cyan         28       1879, 5240       17*57'19.51'N, 67'02'42.51" W       1       red       1       red       1       red       2       green       3       blue         29       2493, 4724       17*57'36.36'N, 67'02'42.51" W       3       blue       5       cyan       5	20	5346	67°03'3.50" W		yenow	5	eyun	5	eyun	5	eyun	5	eyun		
5265 $67'02'4118"$ W $1$ $red$ $1$ $red$ $1$ $red$ $1$ $red$ $1$ $red$ $2$ $green$ $3$ $blue$ 28 $17'57'19.51$ N, $2240.51"$ W $1$ $red$ $1$ $red$ $1$ $red$ $1$ $red$ $2$ $green$ $3$ $blue$ 29 $2493$ , $4724$ $67'02'21.70"$ W $3$ $blue$ $5$ $cyan$ $5$ $cyan$ $5$ $cyan$ $5$ $cyan$ 30 $2964$ , $4386$ $17'57'36.36$ N, $67'02'5.73" W3blue4yellow5cyan3blue5cyan312960,326017'58'24.03 N,67'02'5.99" W3blue5cyan3blue5cyan5cyan321952,17'58'25.20 N,32195cyan5cyan5cyan5cyan5cyan341983,17'58'13.55 N,32693blue4yellow5cyan5cyan5cyan351594,327917'58'22.68 N,67'02'36.16" W3blue4yellow5cyan5cyan5cyan361970,341567'02'3.61'' W3blue5cyan5cyan5cyan5cyan372965,3264017'58'23.60 N,3360$	27	1918,	17°57'18.71 N,	3	blue	2	green	5	cyan	4	yellow	5	cyan		
28       1879, 5240       17'57'19.51'N, 67'02'42.51'W       1       red       1       red       1       red       1       red       2       green       3       blue         29       2493, 4724       17'57'36.36'N, 67'02'21.70'W       3       blue       5       cyan       5		5265	67°02'4118'' W				c		5		5		2		
5240 $67'02'42.51"W$ $17'57'36.36"N$ , $4724$ $3$ $10uc$ $5$ $cyan$ $30$ $2964$ , $4386$ $17'58'24.03"N$ , $67'02'5.99"W$ $3$ $blue$ $5$ $cyan$ $3$ $blue$ $5$ $cyan$ $3$ $blue$ $5$ $cyan$ $5$ $cyan$ $5$ $cyan$ $32$ $1952$ , $17'58'26.56"N,321'95cyan5cyan5cyan5cyan5cyan5cyan331565,17'58'25.22"N,321'95cyan5cyan5cyan5cyan5cyan341983,17'58'13.55"N,320'16''W3blue4yellow5cyan5cyan5cyan3517'58'23.68"N,320'92green5cyan5cyan5cyan5cyan341983,17'58'23.68"N,320'92green5cyan5cyan5cyan5$	28	1879,	17°57'19.51 N,	1	red	1	red	1	red	2	green	3	blue		
29       2493, 4724       17*57*36.36 N, 67*02*21.70* W       3       blue       5       cyan       5<		5240	67°02'42.51" W												
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	29	2493,	17°57'36.36 N,	3	blue	5	cyan	5	cyan	5	cyan	5	cyan		
30       2964, 4386       67'02'5.73" W       3       blue       4       yellow       5       cyan       3       blue       5       cyan         31       2960, 3260       17'58'24.03 N, 67'02'5.99" W       3       blue       5       cyan       3       blue       5       cyan       5       cya	20	4724	67°02'21.70" W	2	11	4	11	~			1.1	~			
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	30	2964, 4286	1/5/4/.41 N,	3	blue	4	yellow	5	cyan	3	blue	5	cyan		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	31	2960	07 02 5.75 W	3	blue	5	evan	3	blue	5	evan	5	evan		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	51	3260	67°02'5 99" W	5	onuc	5	Cyan	5	onue	5	Cyan	5	Cyan		
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	32	1952,	17°58'26.56 N.	5	cyan	5	cyan	5	cyan	5	cyan	5	cyan		
$\begin{array}{cccccccccccccccccccccccccccccccccccc$		3179	67°02'40.26"W		5		5		5		5		5		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	33	1565,	17°58'25.22 N,	5	cyan	5	cyan	5	cyan	5	cyan	5	cyan		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		3219	67°02'53.41" W												
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	34	1983,	17°58'13.55 N,	3	blue	4	yellow	5	cyan	5	cyan	5	cyan		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	25	35/9	6/ 02/36.16" W	2		5		5		5		5			
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	33	1394,	17 38 22.08 N, $67^{\circ}02'52 41'' W$	2	green	3	cyan	3	cyan	3	cyan	3	cyan		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	36	1970	17°58'18 88 N	3	blue	5	cyan	5	cyan	5	cyan	5	cvan		
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	50	3415	67°02'39.62" W	5	orac	Ũ	eyun	U	eyun		eyun	Ũ	eyun		
3380       67°02'5.81" W       Image: Constraint of the second se	37	2965,	17°58'20.13 N,	3	blue	1	red	3	blue	3	blue	2	green		
38       1625, 3269       17°58'23.60° N, 67°02'51.36° W       4       yellow       5       cyan       5       cyan <td< td=""><td></td><td>3380</td><td>67°02'5.81" W</td><td></td><td></td><td></td><td></td><td></td><td></td><td></td><td></td><td></td><td>-</td></td<>		3380	67°02'5.81" W										-		
3269         67°02'51.36" W         Image: Constraint of the second secon	38	1625,	17°58'23.60 N,	4	yellow	5	cyan	5	cyan	5	cyan	5	cyan		
39       2611, 3280       17/58/23.34 N, 67°02'17.85" W       4       yellow       5       cyan	26	3269	67°02'51.36" W	.		-			ļ			-			
3280         0/02 1/.85 W         4         yellow         5         cyan         5	39	2611,	17°58'23.34 N,	4	yellow	5	cyan	5	cyan	5	cyan	5	cyan		
40 1022, 17.50.24.07 N, 4 yenow 5 cyan 5	40	3280	0/ 02 1/.85 W	4	vallow	5	ovon	5	oven	5	avan	5	avan		
	40	3236	1/ 38 24.0/ IN, 67°02'51 47" W	4	yenow	3	cyan	3	cyan	3	cyan	3	cyan		

	Appendix 3: Pixel description for the Minimum Distance supervised classification													
Pix	Pixel	Coordinates			Aerial	photograp	h			Ima	ages			
CI TT	n		1	936	1977		19	1999		CASI		ONOS		
			class	color	class	color	class	color	class	color	class	color		
1	1588,	17°57'36.43 N,	DW	red	DW	red	DW	red	SW	thistle	DW	red		
	4719	67°02'52.46" W								2				
2	619,	17°57'53.85` N,	DW	red	DW	red	SW	thistle 2	DW	red	DW	red		
2	4180	67°03°25.45″ W	DW		DW		DW		DW		DW			
3	1658, 4643	1/ 5/ 58.91 N, 67°02'50 09" W	Dw	rea	Dw	red	Dw	rea	Dw	rea	Dw	rea		
4	905 45	17°57'42.18`N	DW	red	DW	red	DW	red	DW	red	DW	red		
	40	67°03'15.69" W	2	ieu	2	icu	2.0	ica	2	icu	2	icu		
5	2066,	17°57'48.91 N,	М	green	SW	thistle	SW	thistle	SW	thistle	SW	thistle		
	4337	67°02'36.25" W				2		2		2		2		
6	1744,	17°58'1.75"N,	SW	thistle	SW	thistle	SW	thistle	SW	thistle	SW	thistle		
7	3941	67°02′47.24″ W	CW	2	DW	2	CW	2	CW	2	CW	2		
/	023, 4358	17 57 48.06 N, 67°03'25 29`W	5 W	2	Dw	red	5 W	2	5 W	2	5 W	2		
8	2513	17°58'19 33` N	SG	Yellow	С	coral	М	green	SG	Yellow	SW	thistle		
0	3403	67°02'21.17" W	50	3	C	corur	101	Breen	50	3	511	2		
9	2182,	17°57'47.10 N,	М	green	SW	thistle	DW	red	SW	thistle	SW	thistle		
	4393	67°02'32.31" W		-		2				2		2		
10	548,	17°57'56.55 N,	SW	thistle	DW	red	SW	thistle	SW	thistle	SW	thistle		
11	4097	67°03′27.87″ W	G	2	G		C	2	G	2	G	2		
11	1749, 5182	175721.39 N, $67^{\circ}02^{\prime}46.93$ W	5	yenow	5	yenow	C	corai	5	yellow	5	yenow		
12	2027	17°57'26 30 N	С	coral	SG	Yellow	SG	Yellow	SG	Yellow	SG	Yellow		
12	5032	67°02' 37.50" W	C	colui	50	3	50	3	50	3	50	3		
13	2436,	17°57'41.01 N,	С	coral	SG	Yellow	SG	Yellow	SG	Yellow	SG	Yellow		
	4581	67°02'23.65" W				3		3		3		3		
14	2869,	17°57'49.90 N,	С	coral	S	Yellow	SG	Yellow	С	coral	SG	Yellow		
15	4309	6/2'8.9/" W	cw/	thistle	<b>CW</b>	thistle	80	3 Vallaw	C	aaral	SC	3 Vallaw		
15	2081, 5018	17 37 20.73 N, 67°02'37 81" W	5 W	3	5 W	3	50	1 enlow	C	corar	50	r enow		
16	2438.	17°57'39.82 N.	С	coral	С	coral	С	coral	С	coral	S	sand		
-	4746	67°02'29.29 W			_		-		_					
17	1268,	17°57'23.42` N,	SG	Yellow	SG	Yellow	М	green	М	green	С	coral		
	5325	67°03'3.15" W		3		3								
18	1828,	17°57'24.32 N,	S	yellow	S	yellow	SG	Yellow	SG	Yellow	SG	Yellow		
10	5092 2704	6/02'44.26''W	c	vallarr	ç	vallarr	80	3 Vollow	80	3 Vollow	80	5 Vallaw		
19	2790, 4423	67°02'44 26'' W	5	yenow	5	yenow	50	1 enow	50	3 s	50	1 enow		
20	3000,	17°57'51.32 N.	S	vellow	S	vellow	SG	Yellow	SG	Yellow	SG	Yellow		
-	4266	67°02'4.52''W		<u> </u>		J		3		3		3		

DW= deep water, SW= shallow water, S= sand, M= mangroves, C= corals, SG= sea grass, UA= urban areas

	Appendix 1: Pixel description for the Minimum Distance supervised classification continuation													
Pix	Pixel	Coordinates			Aerial	photograp	h			Ima	ages			
CI TT	n		1	1936 1977			19	99	C	ASI	IKONOS			
			class	color	class	color	class	color	class	color	class	color		
21	3075,	17°58'20.79 N,	SW	thistle	SW	thistle	М	green	М	green	М	green		
	3360	67°02'2.07" W		2		2		Ũ		C		C		
22	2961,	17°58'20.03` N,	SW	thistle	DW	red	М	green	М	green	М	green		
	3382	67°02'5.94" W		2										
23	1983,	17°58'18.79 N,	C	corals	S	yellow	UA	Orange	UA	Orange	UA	Orange		
24	3417	67°02°29.15″ W	00	X7 11	DW	1	DIU		DW	1	м	1		
24	2403, 4073	1/ 5/ 5/.54 N, 67° 02'22 79" W	50	r ellow	Dw	red	Dw	rea	Dw	rea	IVI	green		
25	2718	17°58'7.06 N	SG	Yellow	SG	Yellow	SG	Yellow	М	green	М	green		
20	3781	67°02'14.16" W	50	3	50	3	50	3		Breen		8.000		
26	1261,	17°57'16.00 N,	С	coral	С	coral	С	coral	С	coral	С	coral		
	5346	67°03'3.50" W												
27	1918,	17°57'18.71 N,	SG	Yellow	SW	Thistle	С	coral	SG	Yellow	S	yellow		
• •	5265	67°02'4118'' W	~	3		3	~		~~	3	~			
28	1879, 5240	17°57′19.51 N,	S	yellow	SW	thistle	С	coral	SG	thistle	S	yellow		
20	5240 2402	67 02 42.51 W	C	aaral	c	3 vollow	TTA	Orongo	C	3	S	vallary		
29	2493, 4724	67°02'21 70" W	C	corar	5	yenow	UA	1	C	corar	3	yenow		
30	2964.	17°57'47.41 N.	С	coral	С	coral	С	coral	SG	thistle	С	coral		
	4386	67°02'5.73" W	_		_		-			2	_			
31	2960,	17°58'24.03` N,	SW	thistle	UA	Orange	М	green	С	coral	М	green		
	3260	67°02'5.99" W				1								
32	1952,	17°58'26.56 N,	C	coral	UA	Orange	UA	Orange	UA	Orange	UA	Orange		
22	3179	67°02′40.26″W	0	1	TTA	1	TTA	1	TTA	1	TTA	1		
33	1565,	1/ 38 25.22 N, 67°02'52 41" W	C	corai	UA	Orange	UA	Orange	UA	Orange	UA	Orange		
3/	1983	17°58'13 55 N	М	oreen	ΠA	Orange	ΠA	Orange	ΠA	Orange	ΠA	Orange		
54	3579	67°02'36.16" W	141	Breen	011	1	011	1	011	1	011	1		
35	1594,	17°58'22.68 N,	С	coral	UA	Orange	UA	Orange	UA	Orange	UA	Orange		
	3297	67°02'52.41" W				1		1		1		1		
36	1970,	17°58'18.88` N,	С	coral	UA	Orange	UA	Orange	С	coral	С	coral		
	3415	67°02'39.62" W				1		1						
37	2965,	17°58'20.13 N,	SW	thistle	DW	red	М	green	М	green	M	green		
38	3380	0/ 02 5.81° W	C	2	IIA	Orange	11A	Orango	C	coral	LIA	Orange		
20	3269	67°02'51 36" W		corar	UA	1	UA	1		corar	UA	1		
39	2611	17°58'23.34 N	С	coral	UA	Orange	С	coral	UA	Orange	С	coral		
57	3280	67°02'17.85" W	Ũ		0	1	e		0	1	Ũ	corar		
40	1622,	17°58'24.67` N,	С	coral	UA	Orange	UA	Orange	UA	Orange	UA	Orange		
	3236	67°02'51.47" W				1		1		1		1		

DW= deep water, SW= shallow water, S= sand, M= mangroves, C= corals, SG= sea grass, UA= urban areas