## Table 13 Event Based – Flood Plumes

	DATA OPTION 1: Quickbird 2	DATA OPTION 2: MODIS
Spatial Dimensions		
Area to cover	12 km x 12 km per scene	Up to 1000 km <sup>2</sup>
Mapping unit	068m panchromatic	0.5m – 5m
Positional accuracy	4.0 m multi-spectral Dependent on georef- erencing process	Dependent on Geo-referencing process
Temporal Dimensions		
When	Approx 10.45 am	User defined
How often	Minimum every 4 days	User defined (can be < 1 day)
Variable to map	Extent of plume and concentrations of sediments	Extent of plume and concentrations of sediments
Environmental / Sensor Restrictions	For sub-tidal vegetation to depth limited by water clarity. Inter-tidal and supra-tidal vegetation can have water on top. Not possible for turbid water	For sub-tidal vegetation to depth limited by water clarity. Inter-tidal and supra-tidal vegetation can have water on top. Not possible for turbid water Clouds, strong winds and
	Clouds, strong winds and breaking waves.	breaking waves.
Processing technique	Image classification and	Image classification and
(Output)	application empirical or analytical models to estimate sediment concentrations.	application empirical or analytical models to estimate sediment concentrations.
Resources – Hardware	PC	PC Image processing software
and Software	Image processing software GIS with image classification module (e.g. ARCGIS Image Analyst)	with Hyper-spectral analysis capabilities, including sub- pixel mapping techniques.
Resource – Personnel	Trained in image classification Experience with high	Trained in hyper-spectral data processing.

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	spatial resolution data	
References:	No specific examples but	Tralli et al. (2005), Shi and
Note these are some	should be possible. See	Wang (2009), Jiang et al.
example references	Curran and Novo (1988)	(2009)

Curran, P. and E. Novo, (1988). "The relationship between suspended sediment concentration and remotely sensed spectral radiance: a review." Journal of Coastal Research 4(3): 351-368.

Jiang, L., Yan, X., and Klemas, V. (2009). "Remote sensing for the identification of coastal plumes: case studies of Delaware Bay." International Journal of Remote Sensing 30: 2033-2048.

Shi, W. and Wang, M. (2009). "Satellite observations of flood driven Mississippi River plume in the spring of 2008." <u>Geophysical Research Letters</u> 36: L07607.

Tralli, D. M., Blom, R. G., Zlotnicki, V., Donnellan, A., and Evans, D. L. (2005). "Satellite remote sensing of earthquake, volcano, flood, landslide and coastal inundation hazards." <u>ISPRS Journal of Photogrammetry</u> and Remote Sensing 59: 185-198.

## Table 14 Event Based – Ship groundings

	DATA OPTION 1: Quickbird 2	DATA OPTION 2: Airborne hyper-spectral data
Spatial Dimensions		
Area to cover	12 km x 12 km per scene	Up to 1000 km <sup>2</sup>
Mapping unit	068m panchromatic	0.5m – 5m
Positional accuracy	4.0 m multi-spectral	Dependent on
	Dependent on georef- erencing process	Geo-referencing process
Temporal Dimensions		
When	Approx 10.45 am	User defined
How often	Minimum every 4 days	User defined (can be < 1 day)
Variable to map	Benthic cover type	Benthic cover type
Environmental / Sensor Restrictions	For sub-tidal vegetation to depth limited by water clarity. Inter-tidal and supra-tidal vegetation can have water on top. Not possible for turbid water Clouds, strong winds and breaking waves.	For sub-tidal vegetation to depth limited by water clarity. Inter-tidal and supra-tidal vegetation can have water on top. Not possible for turbid water Clouds, strong winds and breaking waves.
(Output)	feature detection using segmentation and classification	feature detection
Resources – Hardware and Software	PC Image processing software GIS with image classification module (e.g. ARCGIS Image Analyst)	PC Image processing software with Hyper-spectral analysis capabilities, including sub- pixel mapping techniques.
Resource – Personnel	Trained in image classification Experience with high	Trained in hyper-spectral data processing. Knowledge of area to be

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	spatial resolution data Knowledge of area to be mapped	mapped
References: Note these are some example references	No specific examples but should be possible	No specific examples but should be possible
	See Chabanet et al. (2005) for background information	See Chabanet et al. (2005) for background information

Chabanet, P., Adjeroud, M., Andréfouët, S., Bozec, Y., Ferraris, J., Garcìa-Charton, J., and Schrimm, M., (2005). "Human-induced physical disturbances and their indicators on coral reef habitats: A multi-scale approach." <u>Aquatic Living Resources</u> 18: 215-230.

## Table 15 Event Based – Cyclone

	DATA OPTION 1: Quickbird 2	DATA OPTION 2: LANDSAT 7 ETM
Spatial Dimensions		
Area to cover	12 km x 12 km per scene	185 km x 185 km per scene
Mapping unit	068m panchromatic	Heron Reef
Positional accuracy	4.0 m multi-spectral Dependent on georef- erencing process	15 m panchromatic 30 m multi-spectral Dependent on Geo-
	cremeing process	referencing process
Temporal Dimensions		
When	Approx 10.45 am	Approx 9.45 am
How often	Minimum every 4 days	every 16 days
Variable to map	Benthic cover type	Reef Province Reef Type Geomorphic Zones
Environmental / Sensor Restrictions	For sub-tidal vegetation to depth limited by water clarity. Inter-tidal and supra-tidal vegetation can have water on top. Not possible for turbid water Clouds, strong winds and breaking waves.	For sub-tidal vegetation to depth limited by water clarity. Inter-tidal and supra-tidal vegetation can have water on top. Not possible for turbid water Clouds, strong winds and breaking waves.
Processing technique (Output)	Image classification or feature detection using segmentation and classification. Object based analysis or manual delineation	PC Image processing software GIS with image classification module (e.g. ARCGIS Image Analyst), Object based analysis software if using this technique.
Resources – Hardware and Software	PC Image processing software GIS with image classification module (e.g. ARCGIS Image Analyst) When object based analysis then experience	Trained in image classification Experience with Landsat data Knowledge of reef geomorphology to be mapped

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	with object based analysis software (e.g. Ecognition)	When object based analysis then experience with object based analysis software (e.g. Ecognition)
Resource – Personnel	Trained in image classification Experience with high spatial resolution data Knowledge of area to be mapped	PC Image processing software GIS with image classification module (e.g. ARCGIS Image Analyst), Object based analysis software if using this technique.
References: Note these are some example references	Scopélitis et al (2009)	Klemas (2009)

Klemas, V. (2009). "The Role of Remote Sensing in Predicting and Determining Coastal Storm Impacts." Journal of Coastal Research 25: 1264-1275

Scopélitits J., Andréfouët S., Phinn S., Chabanet P., Naim O., Tourrand C., Done T. (2009) "35 years of coral community change on Saint-Leu Reef (la Réunion, Indian Ocean) from in situ and remote sensing assessment." <u>Estuarine and Coastal Shelf Science</u> 84: 342-352.

## Table 16 Event Based – Oil Spills

	DATA OPTION 1:	DATA OPTION 2:
	Quickbird 2	Radarsat or Terrasar
Spatial Dimensions		
Area to cover	12 km x 12 km per	Up to 3600 km <sup>2</sup>
	scene	
Mapping unit		5m -60m
	068m panchromatic	
	4.0 m multi-spectral	
Positional accuracy	Den en dent en meant	Dependent on
	Dependent on georef-	Geo-referencing
Tomporal	erencing process	processGeo-referencing
Temporal Dimensions		
Dimensions		
When	Approx 10.45 am	User defined
How often	Minimum every 4 days	User defined (can be < 1
		day)
Variable to map	Extent of oil spill on water	Extent of oil spill on water
-	surface	surface
Environmental / Sensor	Cloud cover, strong winds	Strong winds and breaking
Restrictions	and breaking waves.	waves.
Processing technique	Image classification or	Image classification or
	feature detection using	feature detection using
(Output)	segmentation and	segmentation and
	classification	classification
Resources –	PC	PC
Hardware	Image processing	Image processing software
and Software	software	with radar image analysis
	GIS with image	capabilities, including sub-
	classification module (e.g.	pixel mapping techniques.
	ARCGIS Image Analyst)	
Resource – Personnel	Trained in image	Trained in radar data
	classification	processing.
	Experience with high	Knowledge of area to be
	spatial resolution data	mapped
	Knowledge of area to be	
	mapped	
References:	Wettle et al (2009)	Brekke and Solberg (2005)
Note these are some	Hese and Schmullius	Ferraro et al (2010)
example references	(2009)	

Brekke, C. & Solberg, A. H. S. (2005). Oil spill detection by satellite remote sensing. *Remote Sensing of Environment*, 95, 1-13.

Ferraro, G., Baschek, B., de Montpellier, G., Njoten, O., Perkovic, M., and Vespe, M. (2010). "On the SAR derived alert in the detection of oil spills according to the analysis of the EGEMP." <u>Marine Pollution Bulletin</u> 60: 91-102.

Hese, S. and Schmullius, C., (2009). "High spatial resolution image object classification for terrestrial oil spill contamination mapping in West Siberia." <u>International Journal of Applied Earth Observation and</u> <u>Geoinformation</u> 11: 130-141.

Wettle, M., Daniel, P., Logan, G., and Thankappan, M. (2009). "Assessing the effect of hydrocarbon oil type and thickness on a remote sensing signal: A sensitivity study based on the optical properties of two different oil types and the HYMAP and Quickbird sensors." <u>Remote Sensing of Environment</u> 113: 2000-2010.