

## A machine learning approach to estimate river bank erosion through multi-temporal LIDAR and spectral imagery

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Australian Rivers Institute, Griffith University

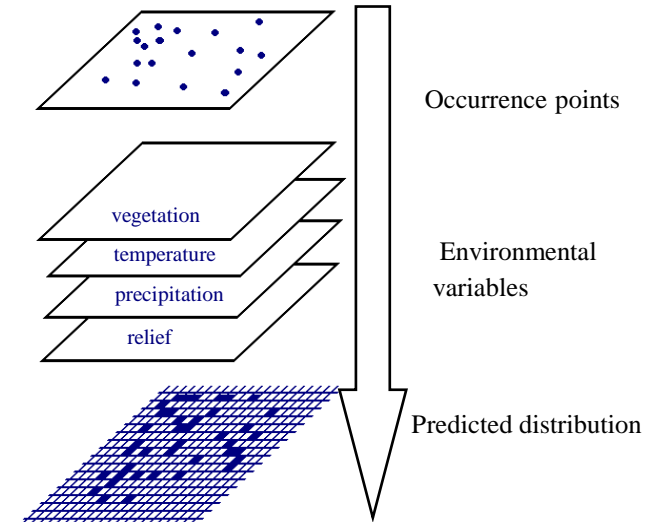
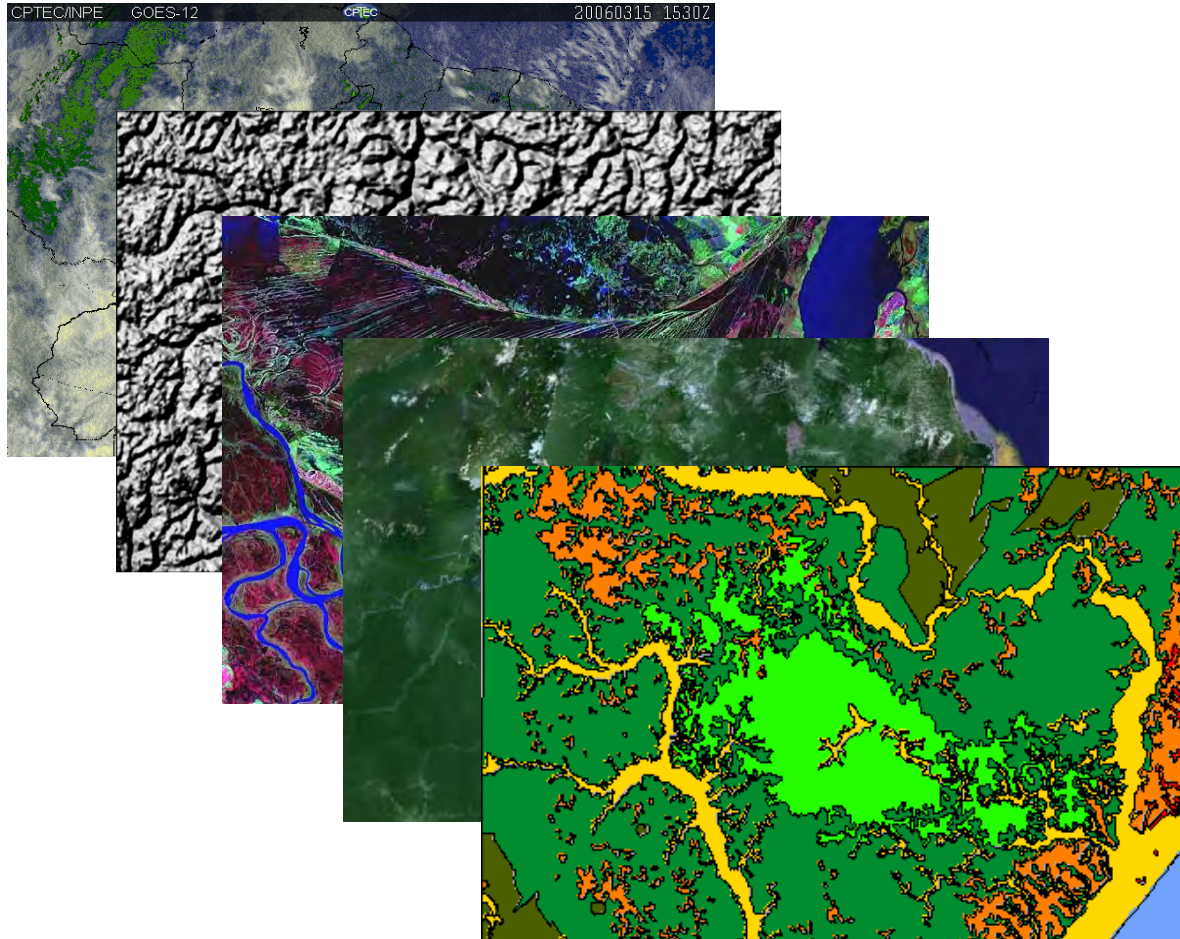
**Sediment Sinks Sources & Drivers  
in the Normanby Basin**

**CAPE YORK WATER QUALITY**

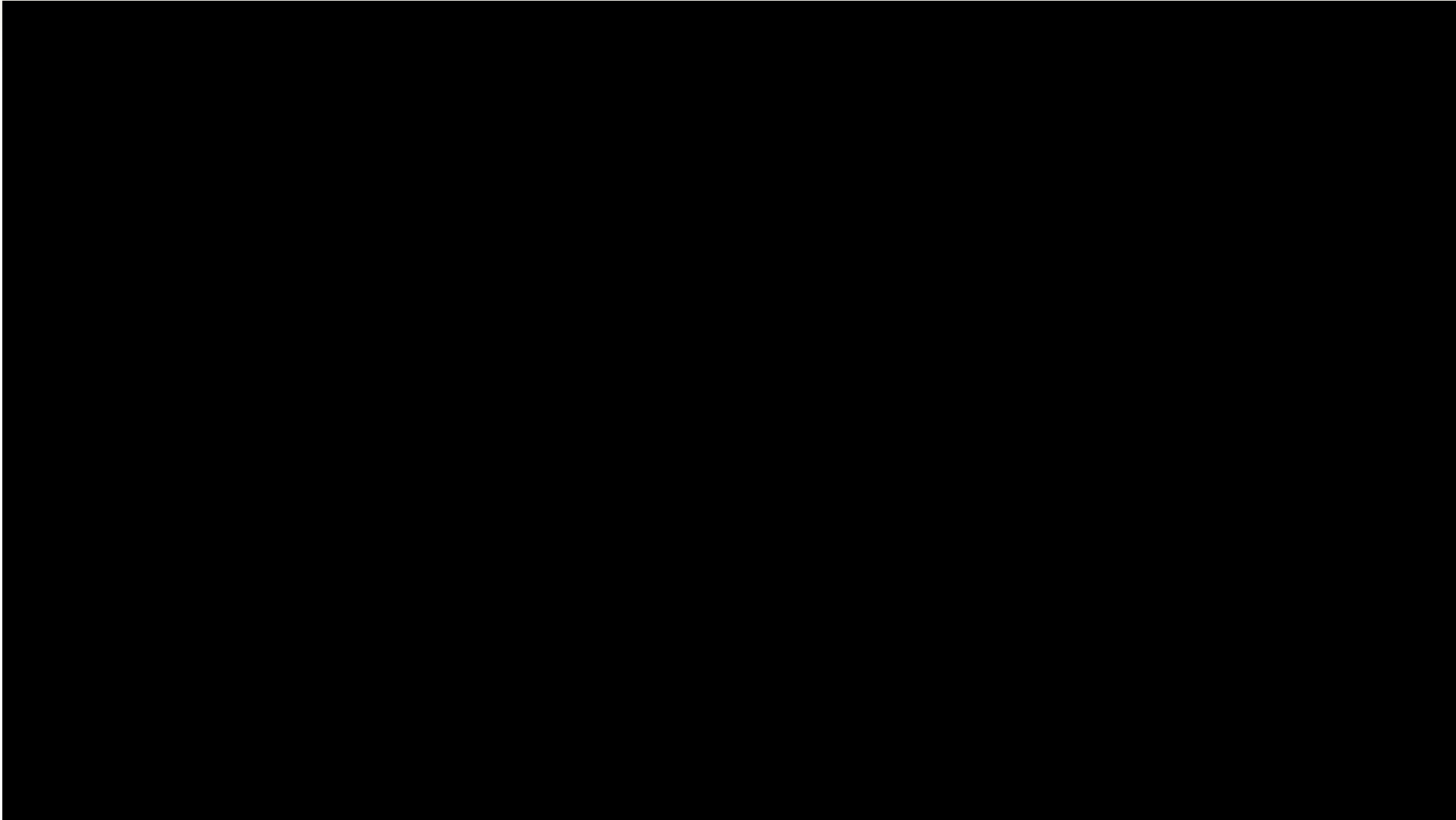


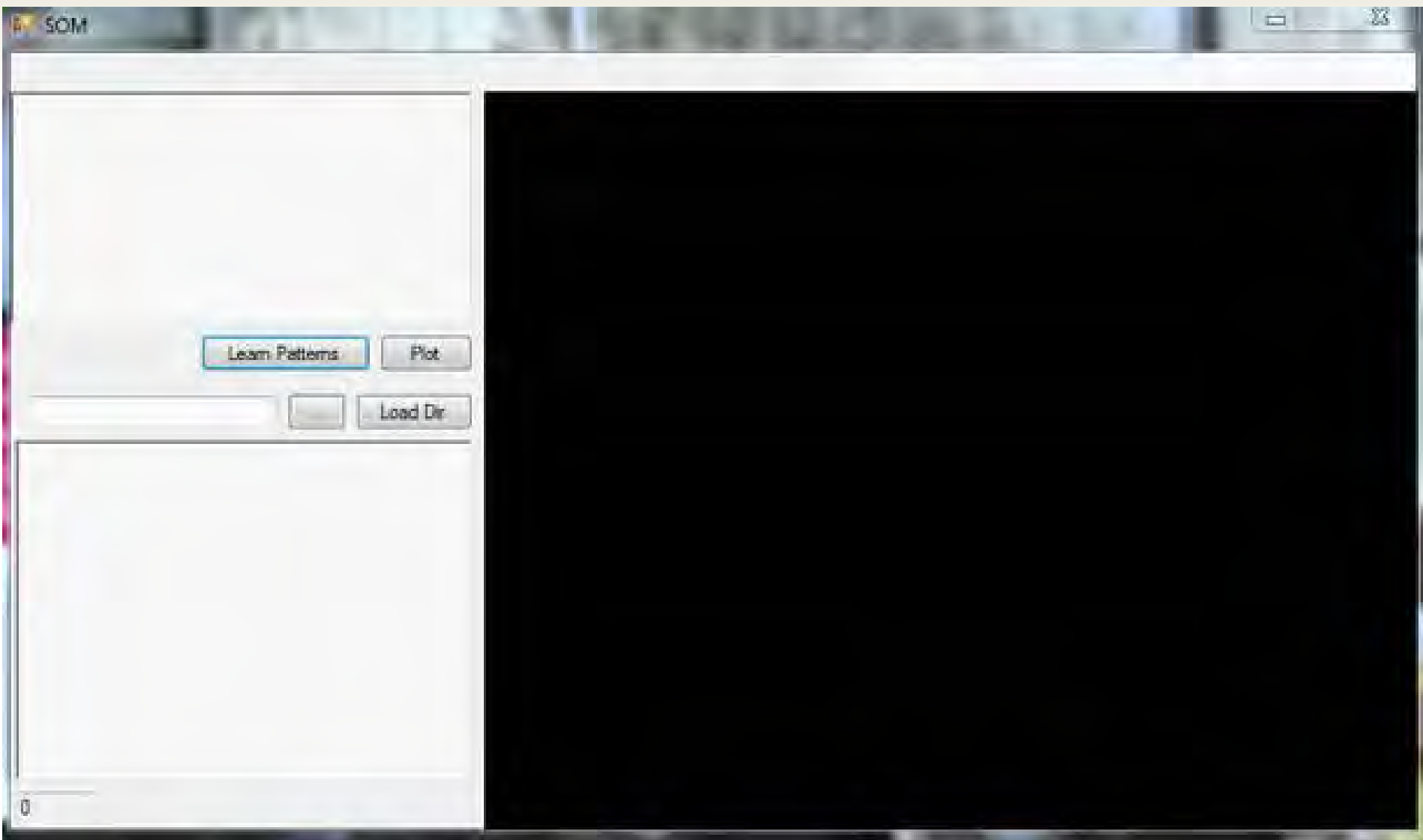
CARING  
FOR  
OUR  
COUNTRY

# Machine learning



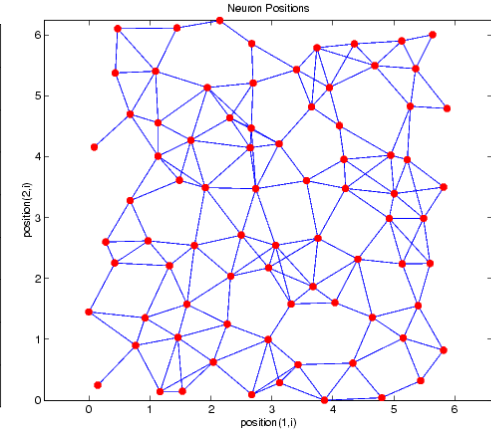
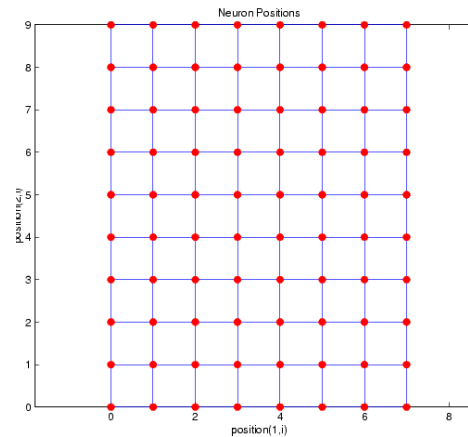
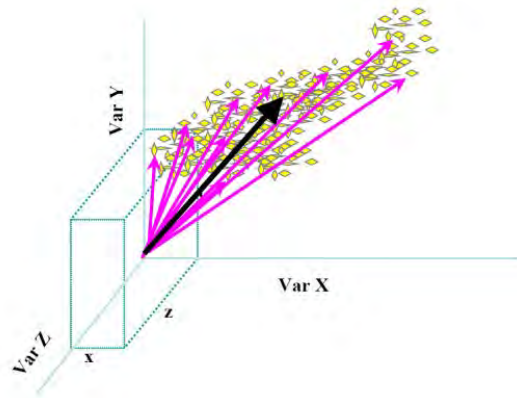
Modified from Scachett (2005)



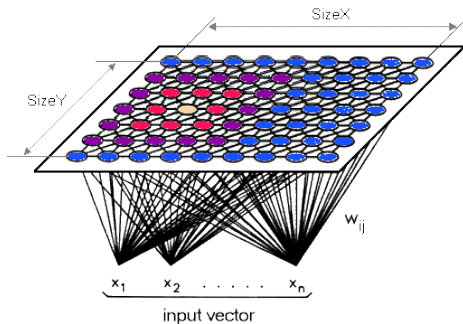


# Self organizing maps (SOM)

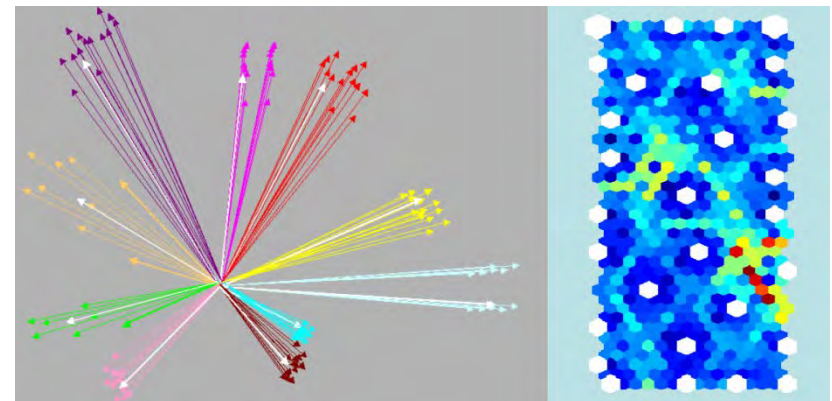
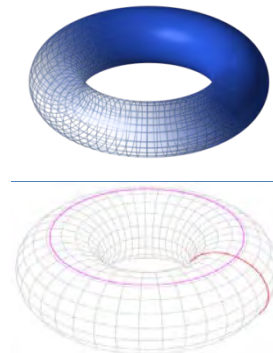
Each samples is associated to a vector



Positions of the samples change, the topology is preserved



Map is projected on a toroid



# Variables

Sensor	Metric	Resolution (m)
LIDAR	Erosion	1
	Elevation	30
SRTM	Slope	30
	Drainage area	30
	Range	30
	Ratio B5/B4	30
Landsat 5	Ratio B3/B5	30
	Ratio B3/B2	30
	Ratio B3/B4	30
	Ratio B7/B2	30
	Ratio B3/B1	30
	Irradiance B6	120
	NDVI	30
	Channel depth	
	Channe width	

# Landsat 5 - TM

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- Order on <http://earthexplorer.usgs.gov/>
- After 2003 “on demand”

# Digital Number to Spectral Reflectance

$$L_{\lambda} = G_{\text{rescale}} \times Q_{\text{cal}} + B_{\text{rescale}}$$

$$\rho_P = \frac{\Pi \cdot L_{\lambda} \cdot d^2}{\text{ESUN}_{\lambda} \cdot \cos \theta_s}$$

$$T = \frac{K2}{\ln \left( \frac{K1}{L_{\lambda}} + 1 \right)}$$

Spectral Radiances, LMIN <sub>λ</sub> and LMAX <sub>λ</sub> in W/(m <sup>2</sup> .sr. μm)								
Processing Date	From March 1, 1984				After May 5, 2003			
	To May 4, 2003							
Band	LMIN <sub>λ</sub>	LMAX <sub>λ</sub>	G <sub>rescale</sub>	B <sub>rescale</sub>	LMIN <sub>λ</sub>	LMAX <sub>λ</sub>	G <sub>rescale</sub>	B <sub>rescale</sub>
1	-1.52	152.10	0.602431	-1.52	-1.52	193.0	0.762824	-1.52
2	-2.84	296.81	1.175100	-2.84	-2.84	365.0	1.442510	-2.84
3	-1.17	204.30	0.805765	-1.17	-1.17	264.0	1.039880	-1.17
4	-1.51	206.20	0.814549	-1.51	-1.51	221.0	0.872588	-1.51
5	-0.37	27.19	0.108078	-0.37	-0.37	30.2	0.119882	-0.37
6	1.2378	15.303	0.055158	1.2378	1.2378	15.303	0.055158	1.2378
7	-0.15	14.38	0.056980	-0.15	-0.15	16.5	0.065294	-0.15

DOY	Distance	DOY	Distance	DOY	Distance
1	0.9832	121	1.0076	242	1.0092
15	0.9836	135	1.0109	258	1.0057
32	0.9853	152	1.014	274	1.0011
46	0.9878	166	1.0158	288	0.9972
60	0.9909	182	1.0167	305	0.9925
74	0.9945	196	1.0165	319	0.9892
91	0.9993	213	1.0149	335	0.986
106	1.0033	227	1.0128	349	0.9843
<b>DOY- Day of Year (Julian Day)</b>				365	0.9833

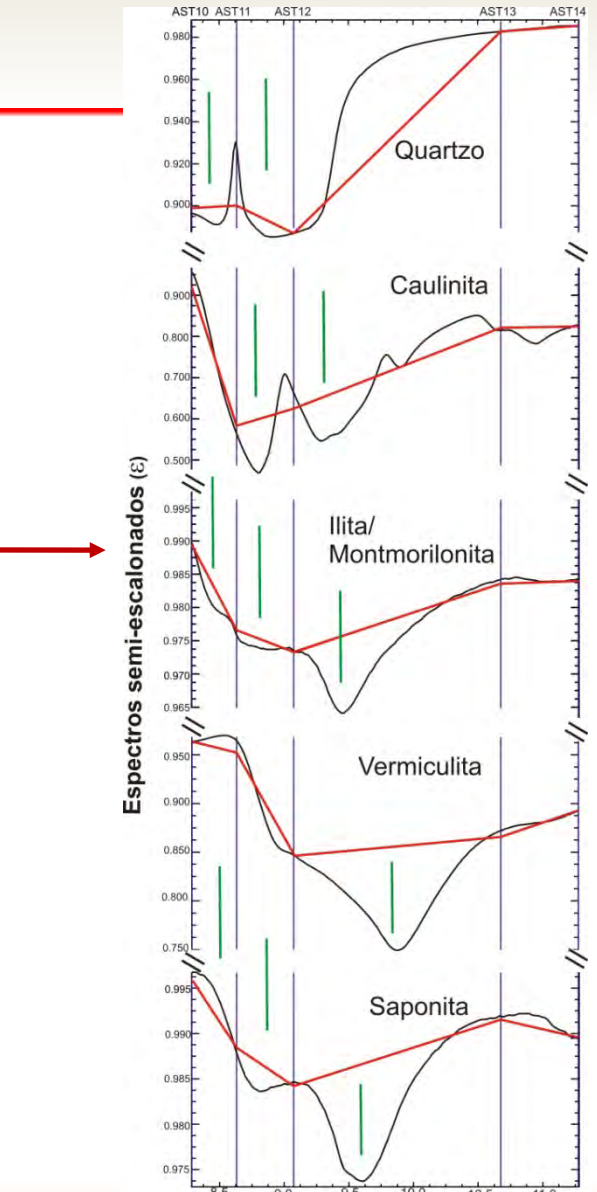
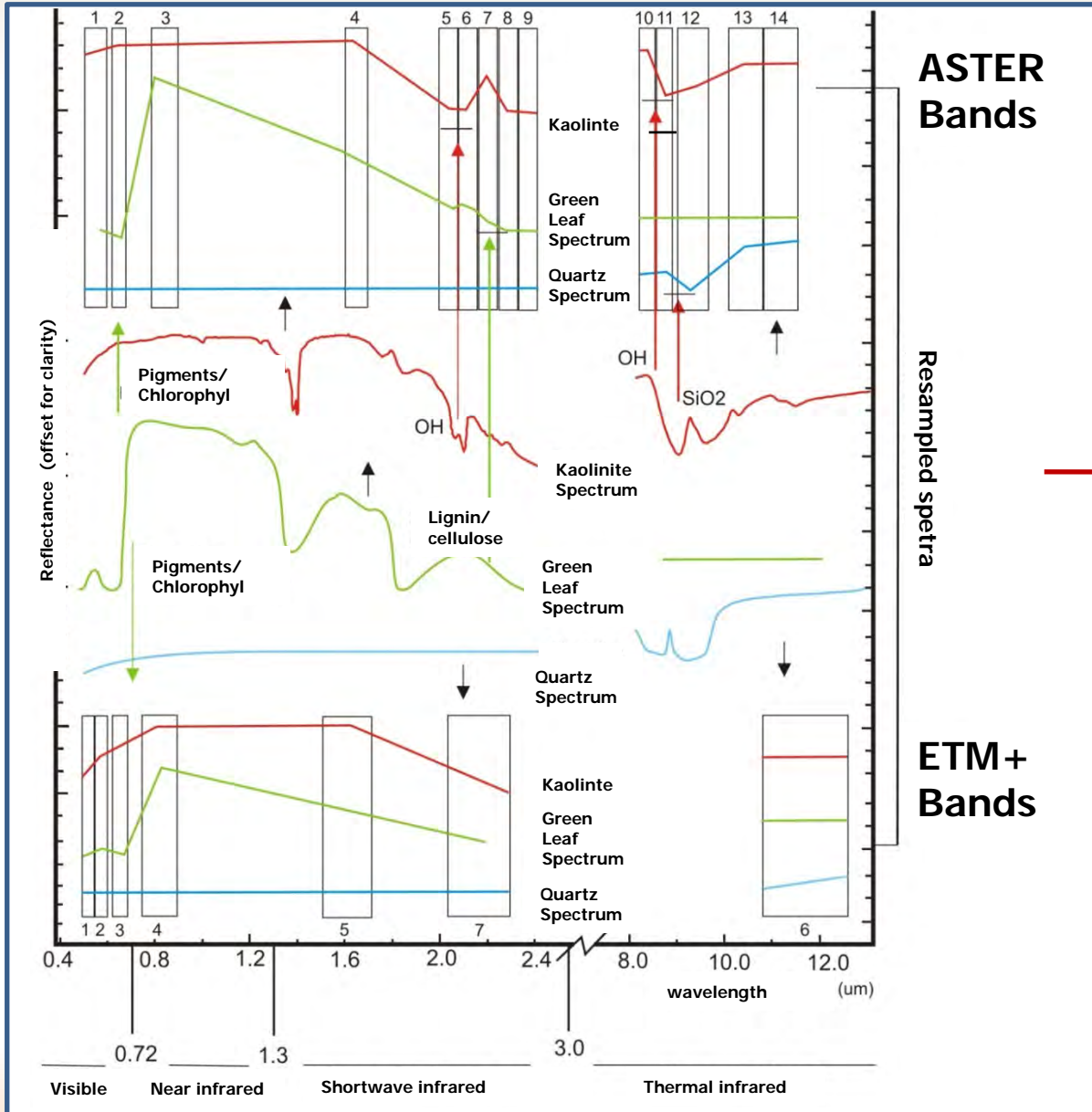
Units: ESUN = W/(m <sup>2</sup> . μm)		
Model:	Chance Spectrum CHKUR	
Band	Landsat 4	Landsat 5
1	1957	1957
2	1825	1826
3	1557	1554
4	1033	1036
5	214.9	215.0
7	80.72	80.67

TM THERMAL BAND CALIBRATION CONSTANTS

Units	W/(m <sup>2</sup> .sr. μm)	Kelvin
Constant	K1	K2
Landsat 4	671.62	1284.30
Landsat 5	607.76	1260.56

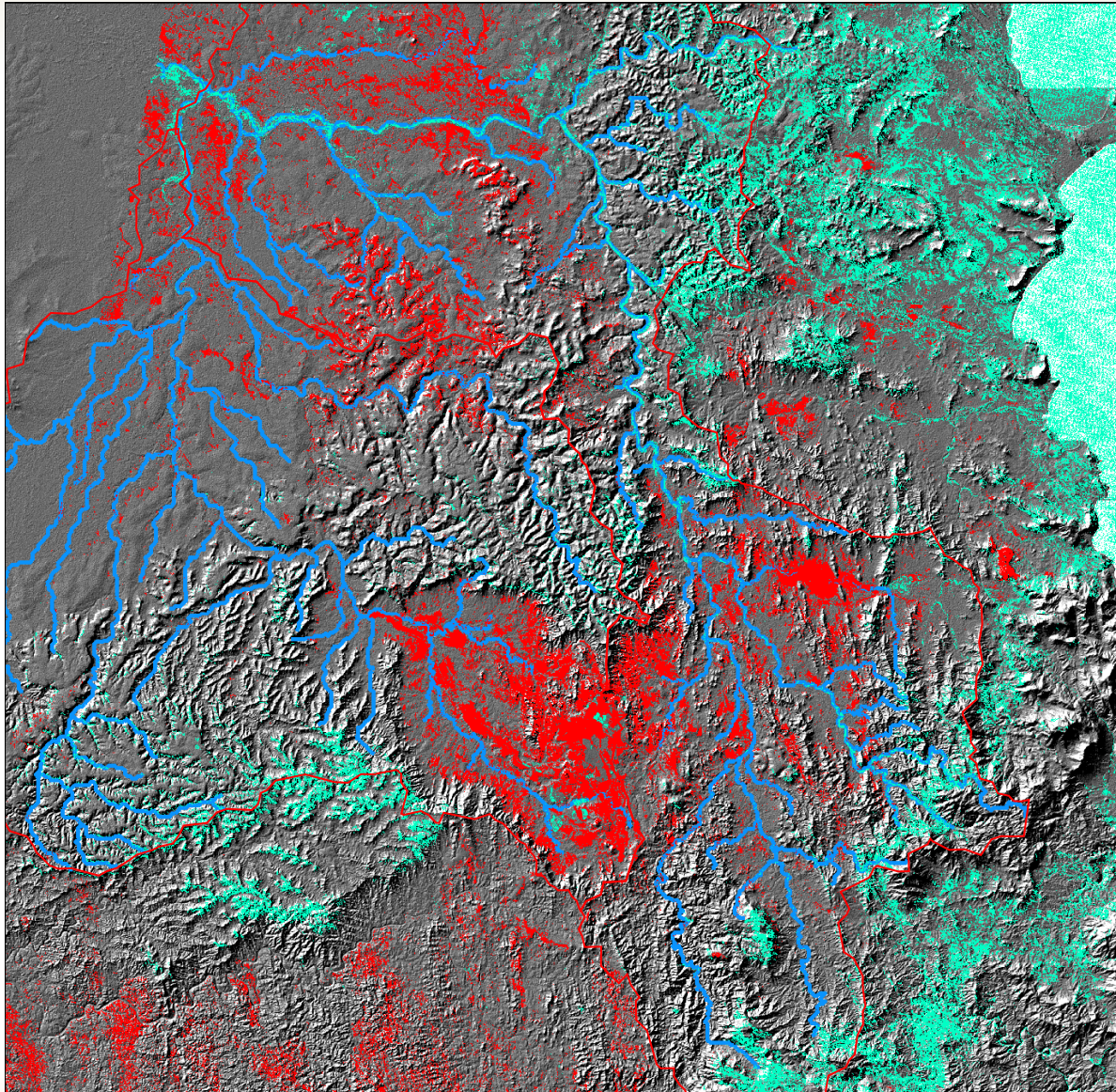


# Band ratios



# Ratio B5B4

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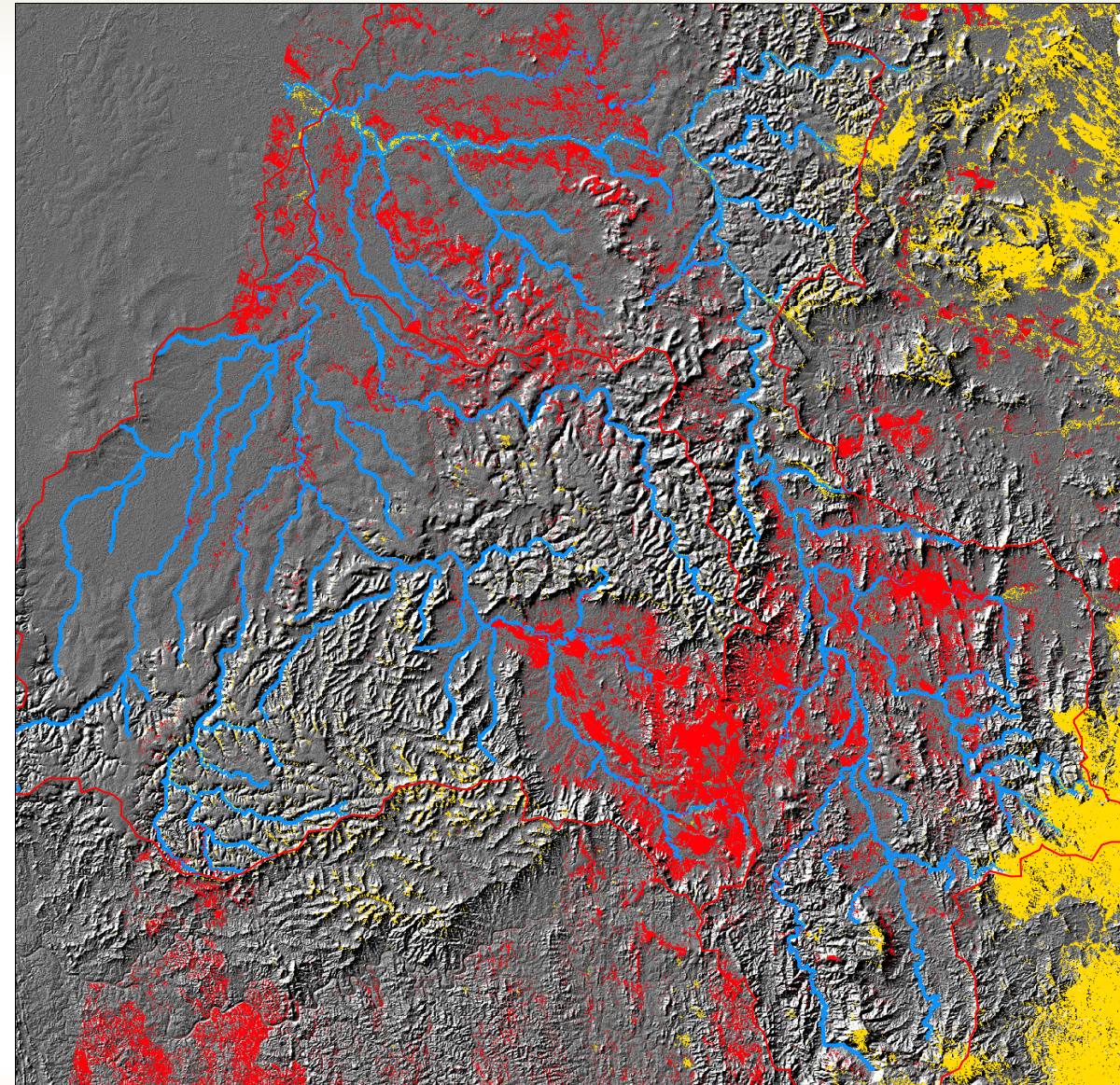


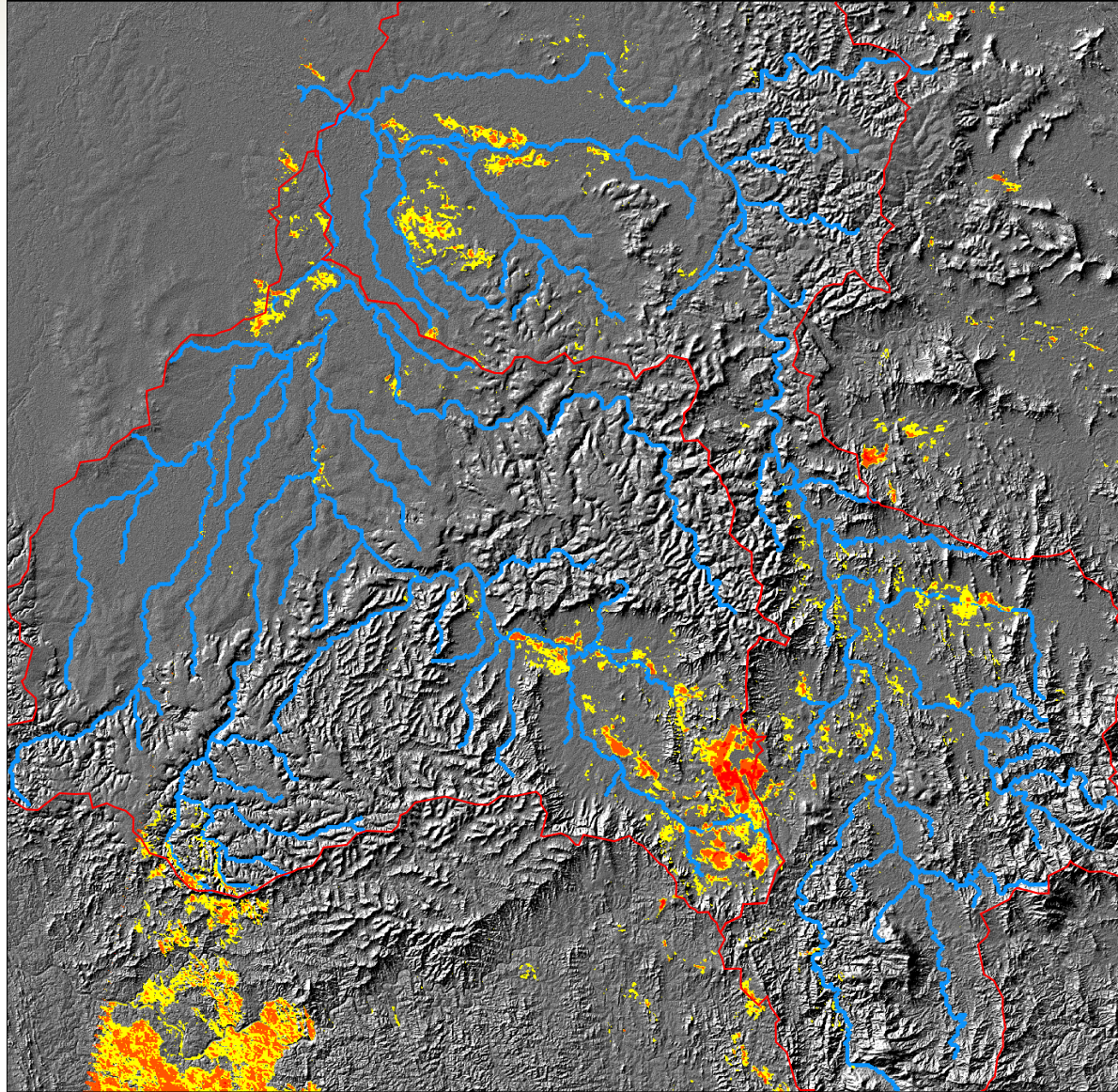
soil moisture,  
water body,  
vegetation,  
barren lands,

# Ratio B7B2

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- soil moisture, water body, vegetation, barren lands,





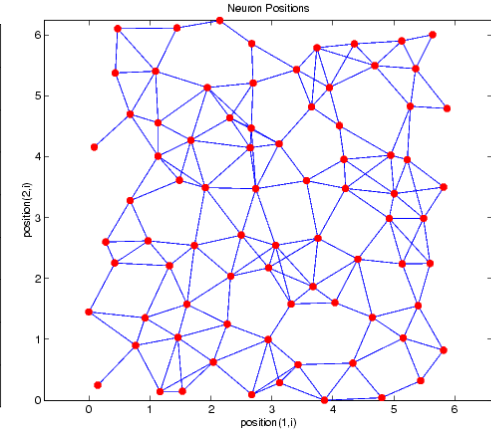
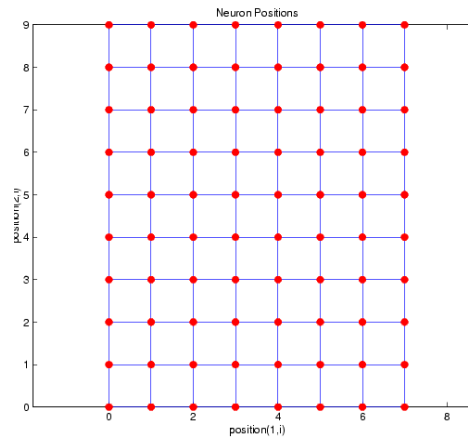
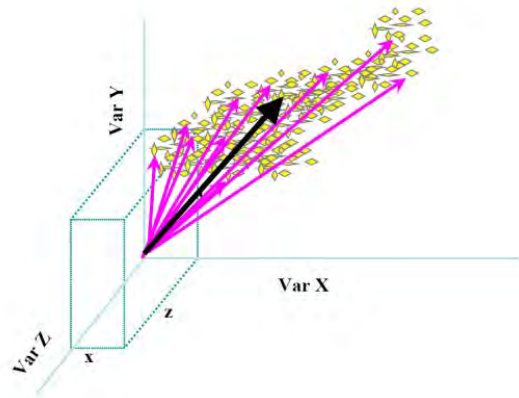
# Band 6

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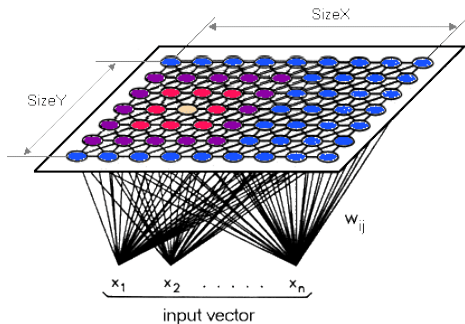
Thermal infra-red  
Irradiance

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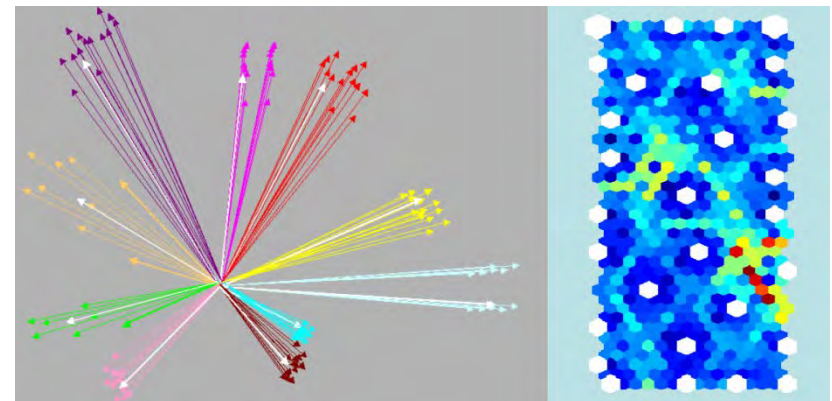
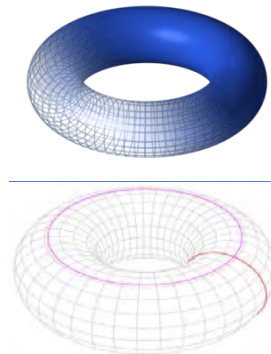
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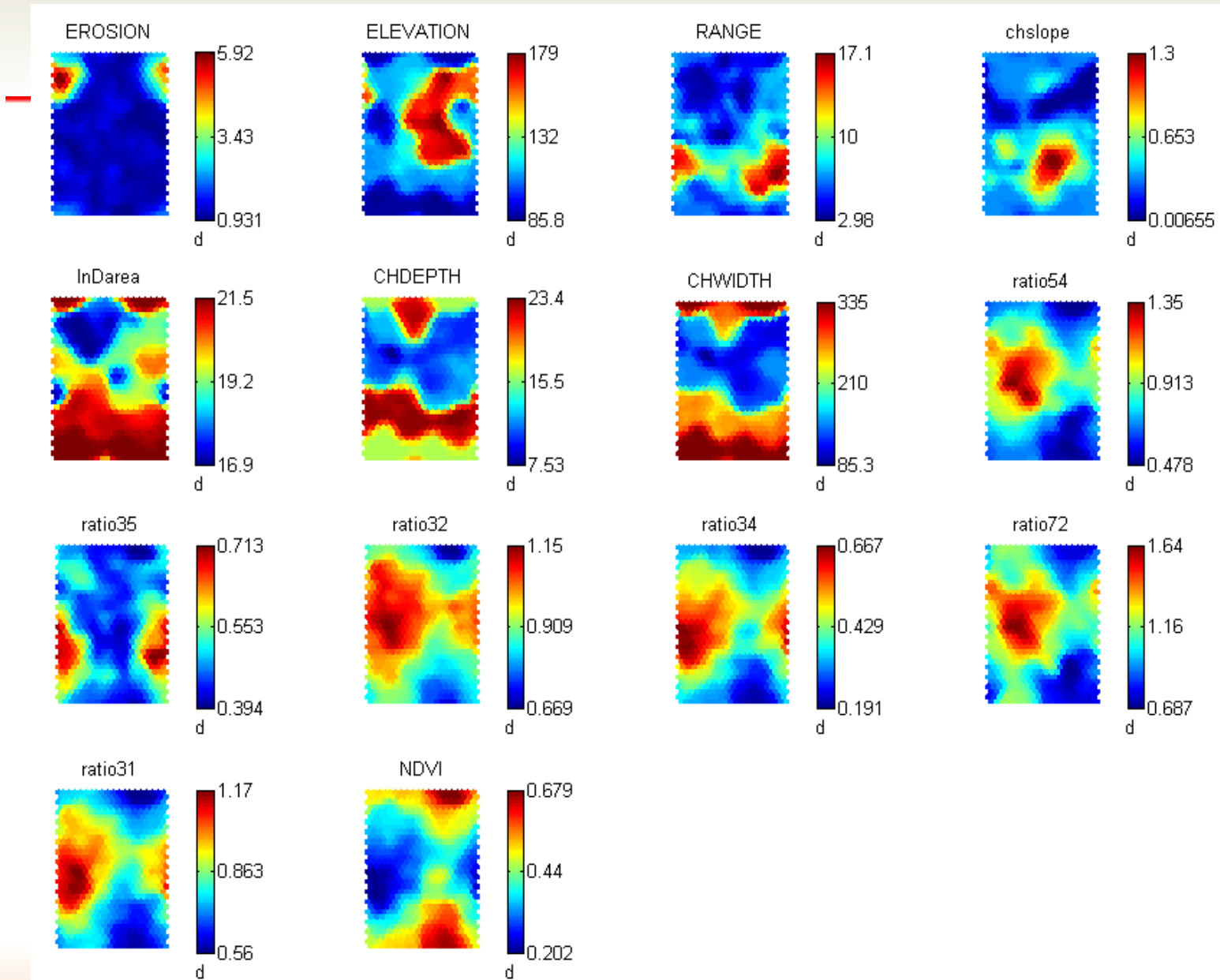


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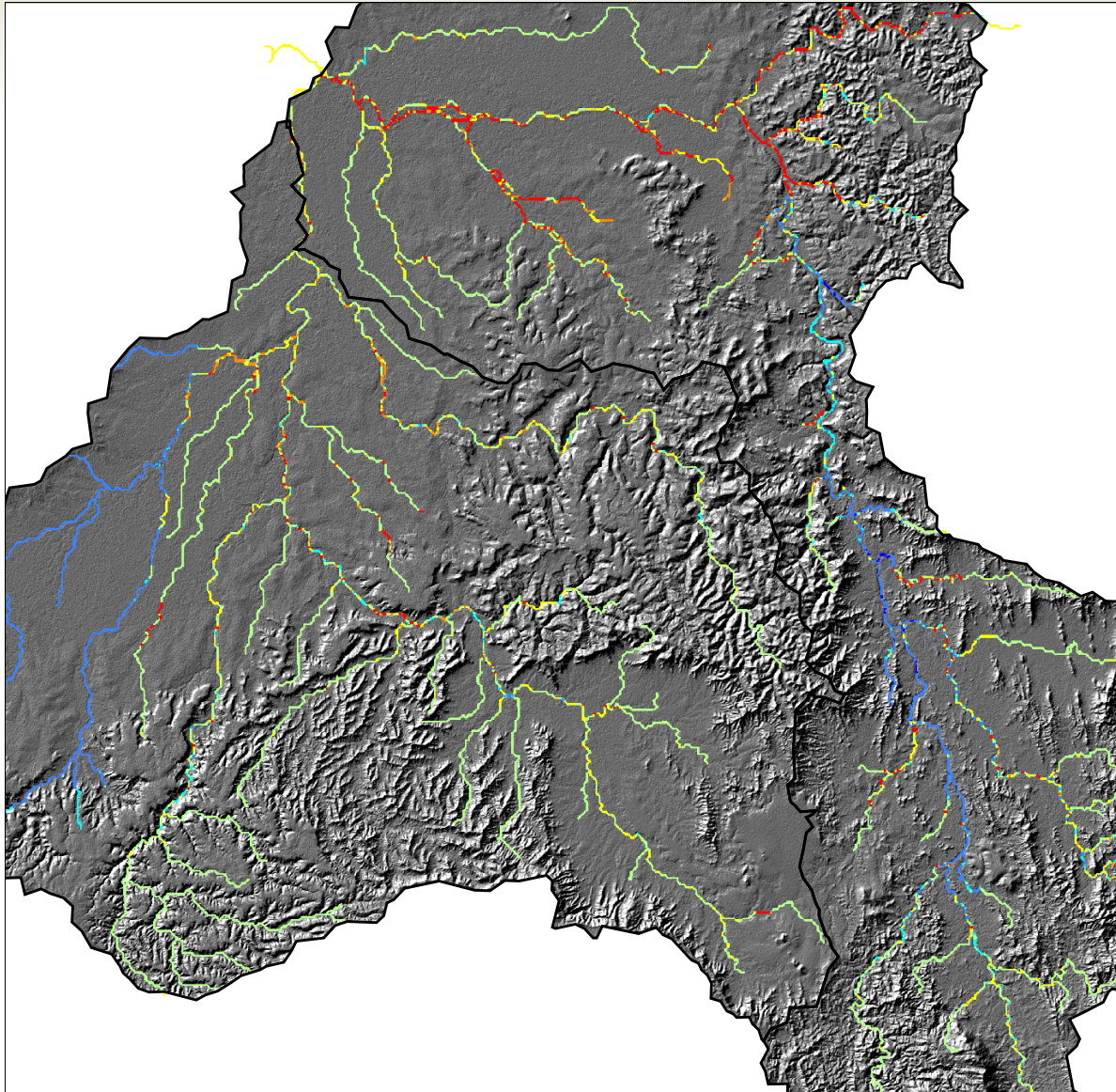
Map is projected on a toroid





# Erosion

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# Conclusions

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- Results show an increasing trend of erosion from the headwaters to the lower reaches associated with bank retreat due to mass failures.
- Anomalous high values of erosion in the upper basin are likely related to alluvial gully erosion as observed during our fieldwork.