

# Supplementary Information for: “An Object- and Topology-Based Analysis (OTBA) Method for Mapping Rice-Crayfish Fields in South China”

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**Table S1.** Classification features for extracting RCFs using RGB image.

Feature Category	Selected Features	Equations	Parameters	References
Spectral features	G-index (Greenness index)	$G - index = \frac{Green}{Red}$	<i>Blue, Green, Red</i> = surface reflectance values of Blue, Green and Red bands.	[1]
	VIgreen (Green vegetation index)	$VIgreen = \frac{Green - Red}{Green + Red}$	# <i>P<sub>v</sub></i> = total number of pixels contained in the object <i>P<sub>v</sub></i> .	[2]
	Brightness	$\bar{c}(v) = \frac{1}{3} \sum_{k=1}^3 \bar{c}_k(v)$	<i>c<sub>k</sub>(x, y, z, t)</i> = the image layer intensity value at pixel <i>(x, y, z, t)</i> .	[3]
	Mean Blue, Green, Red	$\bar{c}_k(v) = \frac{1}{\#P_v} \sum_{(x,y,z,t) \in P_v} c_k(x, y, z, t)$	<i>N<sub>v</sub><sup>B</sup></i> = the darker direct neighbor to <i>v</i> , with $N_v^B \{u \in N_v : \bar{c}_k(u) < \bar{c}_k(v)\}$ .	
	Rel. border to brighter objects Green	$\sum_{u \in N_v^B} \frac{b(v, u)}{b_v}$	<i>b(v, u)</i> = the length of common border between <i>v</i> and <i>u</i> .	
Geometric features	Area	$A_v = \#P_v \times u^2$	<i>u</i> = the pixel size in coordinate system units.	
	Border length	$b_v = b_o + b_i$	<i>b<sub>o</sub></i> = the length of outer border. <i>b<sub>i</sub></i> = the length of inner border.	[3]
	Length/Width	$\gamma_v = \min \gamma_v^{EV}, \max \gamma_v^{BB}$		

	Shape index	$s_v = \frac{b_v}{4\sqrt{A_v}}$	$\gamma_v^{EV}$ = the ratio length of $v$ of the eigenvalues. $\gamma_v^{BB}$ = the ratio length of $v$ of the bounding box.
Textural features	GLCM Mean	$\mu_{i,j} = \frac{\sum_{i,j=0}^{N-1} P_{i,j}}{N^2}$	$i$ = the row number.
	GLCM Std Dev	$\sigma_{i,j} = \sqrt{\sum_{i,j=0}^{N-1} P_{i,j}(i,j - \mu_{i,j})^2}$	$j$ = the column number.
	GLCM Entropy	$e_{i,j} = \sum_{i,j=0}^{N-1} P_{i,j}(-\ln P_{i,j})$	$P_{i,j}$ = the normalized value in the cell $i, j$ .
	GLCM Homogeneity	$h_{i,j} = \sum_{i,j=0}^{N-1} \frac{P_{i,j}}{1 + (i - j)^2}$	$N$ = the number of rows or columns.
	GLCM Contrast	$c_{i,j} = \sum_{i,j=0}^{N-1} P_{i,j}(i - j)^2$	

[3]

**Table S2.** Accuracy assessment about classification results derived by using images with different spectral bands.

	<b>Image with Three Bands (R, G, B)</b>		<b>Image with Four Bands (R, G, B, NIR)</b>	
	Producer's Accuracy	User's Accuracy	Producer's Accuracy	User's Accuracy
RCF	75.59%	84.96%	90.32%	93.33%
Non-RCF	85.71%	76.69%	93.28%	93.33%
Overall Accuracy	80.49%		91.77%	

**Table S3.** Accuracy assessment about classification results derived by using images in different phases.

	Flooding Phase		Rice Growth Phase	
	Producer's Accuracy	User's Accuracy	Producer's Accuracy	User's Accuracy
RCF	83.64%	89.32%	90.32%	93.33%
Non-RCF	88.42%	82.35%	93.28%	93.33%
Overall Accuracy	85.85%		91.77%	

## References

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