

Supplementary materials

Application of Lithological Mapping Based on Advanced Hyperspectral Imager (AHSI) Imagery Onboard Gaofen-5 (GF-5) Satellite

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Some parameters of M3D-DCNN, HybridSN, SSUN, and SVM-RBF methods are shown in Table S1-S8. Table S1 and Table S2 are about M3D-DCNN method; Table S3 and Table S4 are about HybridSN method; Table S5-S7 are about SSUN method; and Table S8 is for SVM-RBF model.

Table S1. Parameters of convolutional layers in M3D-DCNN method [1].

| Kernel Name | Kernel Number | Kernel Size (H, W, B) ¹ | Kernel Stride Δ (H, W, B) |
|-------------|---------------|---------------------------------------|-------------------------------------|
| conv1 | 16 | 3, 3, 11 | 1, 1, 3 |
| conv2_1 | | 1, 1, 1 | |
| conv2_2 | | 1, 1, 3 | |
| conv2_3 | 16 | 1, 1, 5 | 1, 1, 1 |
| conv2_4 | | 1, 1, 11 | |
| conv3_1 | | 1, 1, 1 | |
| conv3_2 | | 1, 1, 3 | |
| conv3_3 | 16 | 1, 1, 5 | 1, 1, 1 |
| conv3_4 | | 1, 1, 11 | |
| conv4 | 16 | 2, 2, 3 | 1, 1, 1 |
| pooling | - | 2, 2, 3 | 2, 2, 3 |

¹ H, W, B represent the size of kernel along spatial and spectral dimensions respectively.

Table S2. Other parameters of M3D-DCNN method [1].

| Name | Value | Name | Value |
|--------------------|-----------------------|--------------|--------------------------|
| Training algorithm | AdaGrad algorithm [2] | Dropout | 0.6 |
| Base learning rate | 0.01 | Weight Decay | 0.01 |
| Batch size | 40 | Input Size | $7 \times 7 \times$ Band |

Table S3. Parameters of layers in HybridSN method [3].

| Convolutional Layers | | | |
|----------------------|---------------|-----------------------|------------------------------|
| Kernel Name | Kernel Number | Kernel Size (H, W, B) | Activation |
| Conv3d_1 | 8 | 3, 3, 7 | |
| Conv3d_2 | 16 | 3, 3, 5 | Rectified linear unit (ReLU) |
| Conv3d_3 | 32 | 3, 3, 3 | |

| | | | |
|-------------------------------|-------|------------|---------|
| Conv2d_1 | 64 | 3, 3 | |
| Fully Connected Layers | | | |
| Name | Units | Activation | Dropout |
| Dense_layer1 | 256 | ReLU | 0.4 |
| Dense_layer2 | 128 | ReLU | 0.4 |
| Output_layer | 14 | softmax | - |

Table S4. Other parameters of HybridSN method [3].

| Name | Value | Name | Value |
|---------------|-------|------------|-------|
| Optimizer | Adam | Batch size | 256 |
| Learning Rate | 0.001 | Epochs | 100 |
| Decay | 1e-06 | - | - |

Table S5. Parameters of LSTM layers in SSUN method [4].

| Name | Units | L2 Regularization | Activation |
|---------------|-------|-------------------|------------|
| LSTM_Spectral | 128 | 0.0001 | - |
| LSTM_Dense | 128 | - | ReLU |
| LSTM_SOFTMAX | 14 | - | softmax |

Table S6. Parameters of MSCNN layers in SSUN method [4].

| Kernel Name | Function | Kernel Number | Kernel Size (H, W) | Activation | Padding |
|-------------|--------------|---------------|--------------------|------------|---------|
| CONV1 | Conv2D | 32 | 3, 3 | ReLU | same |
| POOL1 | MaxPooling2D | - | 2, 2 | - | - |
| CONV2 | Conv2D | 32 | 3, 3 | ReLU | same |
| POOL2 | MaxPooling2D | - | 2, 2 | - | - |
| CONV3 | Conv2D | 32 | 3, 3 | ReLU | same |
| POOL3 | MaxPooling2D | - | 2, 2 | - | - |
| DENSE1 | DENSE | 128 | - | ReLU | - |
| DENSE2 | DENSE | 128 | - | ReLU | - |
| DENSE3 | DENSE | 128 | - | ReLU | - |
| CNN_SOFTMAX | DENSE | 14 | - | softmax | - |

Table S7. Other parameters of SSUN method [4].

| Name | Value | Name | Value |
|---------------|-------|--------------------|-------|
| Optimizer | Adam | Batch size | 64 |
| Learning rate | 1e-4 | Epochs | 500 |
| Beta_1 | 0.9 | Time steps in LSTM | 3 |
| Beta_2 | 0.999 | Amsgrad | False |

Table S8. Parameters of SVM-RBF method.

| Name | Value |
|------------------------------|---|
| Gamma in the kernel function | inverse of the number of bands (1/number of bands) |
| Penalty parameter | 100.00 |

References

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