Machine learning based augmentation for LISS-4

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Resourcesat-2/2A



Fig: Resourcesat 2/2A 3-tier imaging

Resourcesat-2/2A

- LISS-4, LISS-3 and AWiFS sensors image the scene simultaneously.
- LISS-4 at 5.0 m resolution has 3 bands Green, Red and NIR ; no SWIR band
- LISS-3 at 24.0 m resolution has 4 bands Green, Red, NIR and SWIR

Objective:

• To augment LISS-4 by generating SWIR band data at 5.0 m spatial resolution



SWIR band data

- Can penetrate smoke and thin clouds
- Can differentiate between snow and cloud
- Helpful for proper delineation of water bodies
- Applications in mineral exploration, wildfire response, soil moisture studies..



R2A-L3-107-052-29OCT2023 (SWIR,NIR,R)



R2A-L3-107-052-29OCT2023 (NIR,R,G)

Generative Al

- Buzz word of recent times...
- Large language models... Chat GPT.....Dall E...
- How to generate SWIR band data ?
 Make use of Generative AI techniques...
 Generate by super-resolution techniques









Super-resolution

- Enhancing the resolution of an image from low resolution to high resolution
- In remote sensing, typical approaches are:
 Pan-sharpening
 Inverting an explicit imaging model
 Machine learning based methods
- Machine learning based methods:
 - Relation between the low resolution and high resolution data is not explicitly specified
 - Learns from data
 - Can capture complex relationships in data
- Different types:
 - Single image super-resolution
 - Multi-image super-resolution



Low resolution

High resolution

R2A-L4-098-061-D-10FEB2023(SWIR,NIR,R)

Super-resolution for LISS-4

- LISS-3 and LISS-4 are acquired simultaneously
- Make use of LISS-3 SWIR at 24.0 m to generate SWIR at 5.0 m
- Different approaches:



Supervised approach

- Super-resolve SWIR at 24.0 m to 5.0 m
- Assumption spectral correlation of image texture is self-similar over a (limited) range of scales
- Training data downsampled images of LISS-4 and LISS-3 SWIR
- Preserve spectral information of SWIR band
- An end-to-end mapping from low resolution to high resolution
- Texture transferred from LISS-4 bands
- Learns a prior on the high resolution structure of SWIR, conditioned on known high resolution structure of LISS-4 data
- Loss function L1
- Trained weight used to super-resolve LISS-3 SWIR, guided by LISS-4 data





Addition Output

Guided Super-resolution

- A unifying framework where
- the inputs are low resolution source image and high resolution guide image from different domain and the target is the high resolution version of the source.
- Formulate as a pixel-to-pixel mapping of guide image(LISS-4) to the domain of source image(LISS-3 SWIR)
- Pixel-wise mapping as a machine learning model.
- Weights learned by minimizing the discrepancy between source image and down-sampled target image
- Unsupervised method, using only the specific source and guide images to fit the mapping







Sambhar lake (RS2A-94-52d-17Oct2022)



	B2	B3	B4	B5
RMSE	5.63	10.05	7.85	8.33
NMSE	0.0488	0.0828	0.040	0.0432
SSIM	0.959	0.946	0.9655	0.968
SAM	2.205	3.230	1.765	2.438



Sambhar lake (RS2A-94-52d-10Mar2023)



R2A-102-63-A-19APR2023



Part of Penna river (L3)

Part of Penna river (L4)

106-52-B-16dec2022 (Parts of Sikkim)



Part of Lhonak lake(L3)

Part of Lhonak lake(L4)

R2-101-060-C-22DEC2023



(R2A-098-061-D-10FEB2023)



NRSC Shadnagar campus (R2-099-061-B-27FEB2023)



Conclusion

- A method to super-resolve LISS-3 SWIR band to generate SWIR band data at 5.0 m resolution is developed
- State of the art quantitative measures such RMSE, NMSE, SSIM and SAM are used to evaluate the generated data and shows the effectiveness of the method.
- Scatter plots for different features such as water, ocean, vegetation and show close match with LISS-3 data.
- Visual analysis also shows high spectral fidelity.
- A new product with simulated SWIR band is realized. Soon it will be released to the users.

