





ENHANCING AND DETECTING CO₂ PLUMES IN SATELLITE IMAGES USING COMPUTER VISION DENOISING, INPAINTING, AND RIDGE TRACING

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Discussion & summary

Background



Copernicus Carbon Dioxide Monitoring (CO2M) satellite constellation

- To be launched in 2025
- 2×2 km² resolution
- Measures CO₂ [$\sigma_{VEG50} \approx 0.7$ ppm] and NO₂ [$\sigma_{ref} \approx 10^{15}$ molecules/cm²] images

Purpose of this study

- Evaluate potential of established computer-vision techniques on satellite images
- Focus on **detecting plumes in future CO2M satellite images** Used input data:
- Synthetic data generated in SMARTCARB project [Kuhlmann et al., 2021]





Denoising – example 1: hard thresholding





Denoising – example 2: Wiener filtering





Denoising – BM3D



[Dabov et al. (2007)]



BM3D applied to satellite data



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BM3D applied to satellite data



Enhancing and Detecting CO2 Plumes in Satellite Images Using Computer Vision.

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Inpainting is like a jigsaw puzzle without pieces



We complete jigsaw puzzles by trying to match patches that fit well with the surrounding data.

We perform inpainting in a similar way, but teach a computer to <u>generate</u> the missing jigsaw pieces, such that: 1. The filled patches look structurally similar to rest of image 2. The filled patches connect well to the rest of the image 3. The filled images looks 'plausible'



Inpainting with a neural network





Example application on satellite data

Image masked by cloud cover



Inpainted image



Ground truth



The inpainted images appear realistic, and correctly *connect* plumes that were disjoint by cloud cover

Note: at this stage, we do <u>not</u> recommend to use inpainted data for emissions estimations, just as a way to connect disjoint plume patches.

Wind field

(addition of wind field reduces average L₁ error by 55%)



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Neurite detection



Varying background strengths (due to stitched microscope images), varying contrast, in presence of noise



[Meijering et al. (2004)]



Plume detection





Plume detection



Empa



Plume detection





Discussion & Conclusion

- The methods thus far were only applied or trained on SMARTCARB data, but as of yet untested on real satellite images
- Meijering method not well-suited for broad point-sources (such as city of Berlin), and detection of the late tail of plumes is only of limited value as emissions are more dispersed at that stage
- Many further uncertainties remain to not just detect but also quantify plume emissions
- BM3D shows a great potential for improving image quality by considering joint XCO₂ and NO₂ images
- Image inpainting can connect disjoint plume pixels due to cloud cover remarkably well. NB: we recommend to use this for small cloud clover fractions only, as the data quality will be bad for large cloud cover fractions.
- Meijering method can detect narrow plume features for long distances



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