

Colorizing Mars

Documented proposed solution or strategy

This section documents the strategy for coloring high-resolution black-and-white Mars imagery using low-resolution color data. The approach combines high-performance geospatial libraries (**Rasterio**) with computer vision libraries (**OpenCV**, **scikit-image**) to handle alignment, color separation, and filtering. The core idea is to use the $L^*a^*b^*$ color space to separate luminance (detail) from chrominance (color) and apply **histogram matching** for color consistency.

Approach Details:

- **Read Input Files:**
 - Read the high-resolution B&W image (JPEG or GeoTIFF) using `cv2.imread` for simple images or `rasterio` for GeoTIFFs.
 - Read the low-resolution color GeoTIFF using `rasterio`.
- **Align Data:**
 - The low-resolution image data, read as a NumPy array.
 - Convert the low-resolution image data from `(Channel, Height, Width)` to `(Height, Width, Channel)` to match the high-resolution image.
 - Resize the low-res image to the high-res dimensions using `cv2.resize` with interpolation (`INTER_CUBIC` or `INTER_LINEAR`).
 - This step allows both images to be processed pixel-to-pixel.
- **Convert to $L^*a^*b^*$:**
 - Convert the resized low-res RGB data to the $L^*a^*b^*$ color space using `cv2.cvtColor(..., cv2.COLOR_RGB2LAB)`.
 - LAB separates **brightness (L^*)** from **color channels (a^* , b^*)**, making it easier to manipulate color without affecting details.
 - $L^* \rightarrow$ Luminance (*lightness / brightness / detail*)
 - $a^* \rightarrow$ Chrominance channel (*green \rightarrow red*)
 - $b^* \rightarrow$ Chrominance channel (*blue \rightarrow yellow*)
- **Histogram Matching:**
 - Use `skimage.exposure.match_histograms` on the luminance channel.
 - This updates the pixel intensities of the high-res image to match the statistical distribution of the low-res luminance channel.
 - Ensures brightness in the high-res image follows the low-res reference.

Syntax: `skimage.exposure.match_histograms(image, reference, *, channel_axis=None, multichannel=False)`

- **Refine Chroma Channels:**

- Smooth or sharpen the a^* and b^* channels using edge-preserving filters:
 - `cv2.bilateralFilter`: Smooths colors based on proximity and color similarity.
 - `cv2.ximgproc.guidedFilter`: Uses the high-res luminance channel as a guide to preserve sharp edges in the color data.

- **Reconstruct**

- Combine the matched L^* channel with the filtered a^* and b^* channels into a final $L^*a^*b^*$ array.
- Convert this back to RGB (`cv2.cvtColor(..., cv2.COLOR_Lab2RGB)`).

- **Georeference & Save:**

- Use `rasterio` to save the output image as a GeoTIFF, by using the low-res GeoTIFF metadata, bounds, and CRS to save the high-resolution colored image as a GeoTIFF using Rasterio.
- This preserves spatial referencing for GIS applications or Cesium.

Concluding Note:

To my knowledge, in previous projects I have used OpenCV and CNNs (Neural Networks) for image resolution enhancement. For these Mars images, I experimented with OpenCV and Rasterio. The challenge was that one image was JPEG and the other a GeoTIFF. I initially used OpenCV combined with Rasterio to maintain georeferencing, and through research and experimentation, I learned how to use OpenCV Lab* color space to separate brightness and color, histogram matching for reference-based adjustments, and bilateral and guided filters for edge-preserving color refinement. Since OpenCV is implemented in C++, this entire workflow can also be performed in C++ for maximum performance.