

Experimental Class

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Image upsampling

The original color image is represented as a $64 \times 64 \times 3$ matrix of intensities I . The image is provided by “lena64color.tiff”. The downsampled color image is represented as a $32 \times 32 \times 3$ matrix of intensities D . The image is provided by “lena32color.tiff”. Your job is to upsample the image D to be a 64×64 color image, by guessing the missing pixels. Generally, for a upsampling application, the original high resolution image is not available. If the upsampled image will be represented by U , we can try to guarantee that U satisfies the condition: $U_{ij} = D_{mk}$, $m = i/2+1$, $k = j/2+1$. So:

1. Please design the convex optimization model to derive the upsampled image. You can think about minimizing ℓ_2 , ℓ_1 and ℓ_∞ norm and compare the effect of each of them;
2. Please try to introduce some regularization items to your model for some smooth regions to guarantee the smoothness in the variation of intensities, such as ℓ_2 regularization:

$$\sum_{i=2}^{512} \sum_{j=2}^{512} \left((U_{ij} - U_{i-1,j})^2 + (U_{ij} - U_{i,j-1})^2 \right),$$

and ℓ_1 regularization:

$$\sum_{i=2}^{512} \sum_{j=2}^{512} \left(|U_{ij} - U_{i-1,j}| + |U_{ij} - U_{i,j-1}| \right).$$

3. Image I cannot be used in the model. I
4. Compare the upsampled image U with I using PSNR (Average PSNR of RGB). Let's see who can win among you.
5. Please provide:
 - (\mathcal{A}) The m file (including the code of calculating the PSNR performance) and the upsampled images;
 - (\mathcal{A}) The report describes your model, your implementation, results and performance comparison in detail.