Image Denoising

Homework n°2 - Exercises

Exercise 5.1

We denote by n the dimension of the initial image u_0 , and k the dimension of the low-scale image u_1 .

Let us explain how u_1 is obtained from u_0 .

- First, we consider the DCT transform of u_0 , denoted by DCT_n^{iso} (u_0) .
- Then, we pad to remove high frequencies, and we obtain $ZP_k(DCT_n^{iso}(u_0))$.
- Because we changed the size (from n to k), we have to scale the image by a factor $\frac{k}{n}$ $\sqrt{\frac{k \times k}{n \times n}} = \sqrt{\frac{\text{NUMPIX}(layer)}{\text{NUMPIX}(input)}}, \text{ that is we consider } \sqrt{\frac{\text{NUMPIX}(layer)}{\text{NUMPIX}(input)}} \text{ZP}_k(\text{DCT}_n^{iso}(u_0)).$

This is what is done in Algorithm 5, in the function ExtractScale, lines 14 and 16.

• Finally, we convert back the above to an image, which means we get (by linearity of $IDCT_k^{iso}$) $\sqrt{\frac{NUMPIX(layer)}{NUMPIX(input)}}IDCT_k^{iso} (ZP_k(DCT_n^{iso}(u_0))).$

So we have
$$u_1 = \sqrt{\frac{\text{NUMPIX}(layer)}{\text{NUMPIX}(input)}} \text{IDCT}_k^{iso} \left(\text{ZP}_k(\text{DCT}_n^{iso}(u_0)) \right).$$

Since $u_0 = \tilde{u}_0 + n_0$ with $n_0 \sim \mathcal{N}(0, \sigma_0^2)$, we have $u_1 = \tilde{u}_1 + n_1$ with $n_1 \sim \mathcal{N}(0, \sigma_1^2)$, where:

$$\sigma_{1}^{2} = \operatorname{Var}(n_{1})$$

$$= \operatorname{Var}(u_{1})$$

$$= \operatorname{Var}\left(\sqrt{\frac{\operatorname{NUMPIX}(layer)}{\operatorname{NUMPIX}(input)}}\operatorname{IDCT}_{k}^{iso}\left(\operatorname{ZP}_{k}(\operatorname{DCT}_{n}^{iso}(u_{0}))\right)\right)$$

$$= \left(\sqrt{\frac{\operatorname{NUMPIX}(layer)}{\operatorname{NUMPIX}(input)}}\right)^{2} \operatorname{Var}\left(\operatorname{IDCT}_{k}^{iso}\left(\operatorname{ZP}_{k}(\operatorname{DCT}_{n}^{iso}(u_{0}))\right)\right)$$

$$= \frac{\operatorname{NUMPIX}(layer)}{\operatorname{NUMPIX}(input)}\sigma_{0}^{2},$$

where, for the last equality, we used the fact that DCT and IDCT preserve the variance (because they are isometries).

We therefore have $\sigma_1 = \sigma_0 \sqrt{\frac{\text{NUMPIX}(layer)}{\text{NUMPIX}(input)}}$ i.e., the noise standard deviation at the lower