

Homework 2:Exercise 5.1:

let n : dimension of initial image

k : dimension of image at lower scale

u_0 : initial image

u_1 : lower scale image

DCT_n^{iso} : isometric DCT at scale n

DCT_k^{iso} : isometric DCT at scale k

ZP_k : the padding operator keep the top left $k \times k$ coef.

To go from u_0 to u_1 , we need to:

i) transform u_0 to the frequency space ie

$$u' = DCT_n^{iso}(u_0)$$

ii) pad the resulting image, ie keep the lowest frequency.

$$u'' = ZP_k(u') = ZP_k(DCT_n^{iso}(u_0))$$

iii) we then adjust the scale of the image.

To do so, we multiply the current image by $\frac{k}{n} = \sqrt{\frac{k \times k}{n \times n}} = \sqrt{\frac{N_{\text{output}}(\text{layer})}{N_{\text{output}}(\text{input})}}$

We have:

$$u''' = \sqrt{\frac{k \times k}{n \times n}} u''$$

iv) The last step is to transform the current image from frequency space to real image space ie

$$u_1 = DCT_k^{-1, iso}(u''')$$

$$u_1 = \sqrt{\frac{k \times k}{n \times n}} IDCT_k^{iso}(ZP_k(DCT_n^{iso}(u_0)))$$

The final image u_1 contains a white gaussian uniform noise ie

$$u_1 = \tilde{u}_1 + n_1, \quad n_1 \sim \mathcal{U}(0, \sigma_1^2)$$

and

$$\begin{aligned} \text{Var}(u_i) = \sigma_i^2 &= \frac{b \times b}{n \times n} \text{Var} \left(\text{IDCT}_k^{\text{iso}} \left(\text{ZPa} \left(\text{OCT}_n^{\text{iso}}(u_0) \right) \right) \right) \\ &= \frac{b \times b}{n \times n} \sigma_0^2 \end{aligned}$$

we

$$\sigma(u_i) = \sqrt{\frac{\text{Numpix}(\text{Oyer})}{\text{Numpix}(\text{Input})}} \sigma_0$$

The standard deviation has been reduced by a factor $\sqrt{\frac{\text{Numpix}(\text{Oyer})}{\text{Numpix}(\text{Input})}}$.

Exercise 6.1:

Let $(n(i))_{i \in \mathbb{Z}}$ a sequence of white noise with σ^2 variance.
 $a(j)$, $j \in [-m; m]$ a filter of size $2m+1$
 $*$ the convolution product.

Let's compute the variance of $a * n(i)$

For a pixel $i \in \mathbb{Z}$,

$$a * n(i) = \sum_j n(i-j) a(j)$$

$$\text{we } \text{Var}[a * n(i)] = \sum_j a^2(j) \text{Var}[n(i-j)]$$

because noise are independent

$$\text{we } \boxed{\text{Var}[a * n(i)] = \sigma^2 \sum_j a^2(j)}$$

We suppose $\sum_j a(j) = 1$. We also have $a(j) \geq 0 \quad \forall j \in [-m; m]$

So, for any $j \in [-m; m]$

$$a^2(j) \leq a(j) \quad \text{we } \sum_j a^2(j) \leq 1$$

$$\text{we have } \boxed{\text{Var}[a * n(i)] \leq \text{Var}[n(i)]}$$

The convolution helps at reducing the noise by a fixed factor.

Now let's look at the optimal filter weights. We want

$$a = \arg \min_a \sum_{i=-m}^m a^2(i) \\ \sum a(i) = 1$$

This is a continuous coercive convex cost function with linear inequality constraint ie the problem is convex.

let's apply KKT conditions ie

For the solution \hat{a} , there exists $\mu \in \mathbb{R}$ such that $\forall i \in [-m, m]$

$$2\hat{a}(i) + \mu = 0 \quad \text{and} \quad \sum_{i=-m}^m \hat{a}(i) = 1$$

ie $\hat{a}(i) = -\frac{\mu}{2} \quad \text{and} \quad \sum_i \hat{a}(i) = 1$

$$\text{ie } \hat{a}(i) = -\frac{\mu}{2} \quad \text{and} \quad -(2m+1)\frac{\mu}{2} = 1$$

$$\text{ie } \boxed{\hat{a}(i) = \frac{1}{2m+1}, \quad \forall i \in [-m, m]}$$