

# IPCV - Assignment 1 (H1)

Team members (T1):

Anika Fuchs, 2580781  
Ankit Agrawal, 2581532  
Akshay Joshi, 2581346

## Problem 1 (Peak-Signal-to-Noise Ratio)

Let the image  $f = (f_{i,j})$  be a noisy version of  $g = (g_{i,j})$  degraded by additive noise  $n = (n_{i,j})$  with zero mean:

$$f_{i,j} = g_{i,j} + n_{i,j}$$

In the lecture the peak-signal-to-noise ratio (PSNR) is defined as a measure of quality.

(a) In which case do we have  $\text{PSNR}(f,g) = 0$ ?

$$\text{PSNR}(f,g) := 10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(f,g)} \right)$$

e.g. grey value range [0, 255]  $\Rightarrow \text{max grey value} = 255$

$$\text{MSE}(f,g) := \frac{1}{MN} \sum_{i=1}^M \sum_{j=1}^N (f_{i,j} - g_{i,j})^2$$

$$\text{PSNR}(f,g) = 0$$

$$10 \log_{10} (x) = 0$$

$$x = 1$$

$$\Rightarrow \frac{(\text{max grey value})^2}{\text{MSE}(f,g)} = 1$$

$$\Rightarrow \text{MSE}(f,g) = (\text{max grey value})^2$$

This is the case if  $(f_{i,j} - g_{i,j})^2$  averages to  $(\text{max grey value})^2$ ,

i.e.  $((g_{i,j} - n_{i,j}) - g_{i,j})^2$  averages to  $(\text{max grey value})^2$

$$\Rightarrow n_{i,j} \rightarrow \text{max grey value}$$

✓

(b) Let a filtered version  $u$  of  $f$  be given such that

$$\text{PSNR}(u,g) > \text{PSNR}(f,g) + 20 \text{ dB}$$

How has the MSE of  $f$  and  $g$  changed during the filtering

$$10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(u,g)} \right) = 10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(f,g)} \right) + 20 \text{ dB}$$

$$\Leftrightarrow 10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(u,g)} \right) - 10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(f,g)} \right) = 20 \text{ dB}$$

by the rule  $\log_b(x/y) = \log_b(x) - \log_b(y)$ :

$$10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(u,g)} \cdot \frac{\text{MSE}(f,g)}{(\text{max grey value})^2} \right) = 20 \text{ dB}$$

$$\Leftrightarrow 10 \log_{10} \left( \frac{\text{MSE}(f,g)}{\text{MSE}(u,g)} \right) = 20 \text{ dB} \quad | : 10$$

$$\Leftrightarrow \log_{10} \left( \frac{\text{MSE}(f,g)}{\text{MSE}(u,g)} \right) = 2 \text{ dB}$$

$$\Leftrightarrow \frac{\text{MSE}(f,g)}{\text{MSE}(u,g)} = 10^2 = 100$$

$$\Rightarrow \text{the new MSE after filtering } \text{MSE}(u,g) = \frac{\text{MSE before filtering}}{100} \quad \checkmark$$

(c) Assume that  $n \rightarrow 0$ . Calculate  $\text{PSNR}(f,g)$ . How can you interpret this

$$\text{PSNR}(f,g) = \lim_{n \rightarrow 0} 10 \log_{10} \left( \frac{(\text{max grey value})^2}{\text{MSE}(f,g)} \right)$$

$$\begin{aligned} \text{MSE}(f,g) &= \frac{1}{MN} \sum_{i=1}^M \sum_{j=1}^N |f_{i,j} - g_{i,j}| \\ &= \frac{1}{MN} \sum_{i=1}^M \sum_{j=1}^N \underbrace{((g_{i,j} - n_{i,j}) - g_{i,j})}_{=0} \\ &= 0 \quad (\checkmark) \end{aligned}$$

$$\Rightarrow \text{PSNR}(f,g) = \lim_{n \rightarrow 0} 10 \log_{10} \left( \frac{(\text{max grey value})^2}{0} \right)$$

UNDEFINED

f 

3/4

2) Problem 2 : Continuous Convolution

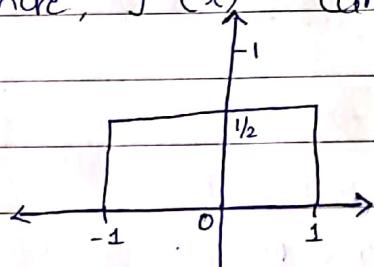
(a)  $f(z) = \begin{cases} \frac{1}{2} & (-1 \leq z \leq 1), \\ 0 & \text{else} \end{cases}$

Find:  
 •  $h_1 = f * f$   
 •  $h_2 = f * f * f$

→ WKT,

$$(g * f)x = \int_{-\infty}^{\infty} g(z) \cdot f(x-z) dz$$

where,  $f(x)$  can be visualized as follows



The signal is in the range  $(-1 \text{ to } 1)$

• Given,  $f(x-z) = \frac{1}{2}$  [Integral]

now, tracing the continuous signal in the range  
 i.e. • Between  $(-1 \leq x \leq 1)$   $f(x-z)$

$$(f * f)x = \int_{(x-1)}^{(x+1)} \frac{1}{2} \cdot g(z) dz$$

• Between  $(-2 \leq x \leq 0)$

$$(f * f)x = \underbrace{\int_{-1}^{-1} \frac{1}{2} f(z) dz}_{\text{outside the range so, 0}} + \underbrace{\int_{-1}^{x+1} \frac{1}{2} f(z) dz}_{\checkmark}$$

• Between  $(0 \leq x \leq 2)$

$$(f * f)x = \int_{x-1}^{-1} \frac{1}{2} f(z) dz + \underbrace{\int_{-1}^{x+1} \frac{1}{2} f(z) dz}_{\text{outside range so, eliminate}}$$

Finally,

$$\text{For } h_1 = (f * f)_x$$

> when  $-1 \leq x \leq 1$

$$\int_{x-1}^{x+1} \frac{1}{2} \cdot f(z) \cdot dz$$

> when  $-2 \leq x \leq 0$

$$\int_{-1}^{x+1} \frac{1}{2} \times \frac{1}{2} \underbrace{f(z)}_{=} dz = \frac{2+x}{4}$$

> when  $x < -2$

out of range, so  $\underline{\underline{0}}$

> when  $x \geq 2$

out of range, so  $\underline{\underline{0}}$

$$\therefore h_1 = \begin{cases} \frac{2+x}{4} & (-2 \leq x \leq 0) \\ 0 & (x \leq -2) \\ \frac{2-x}{4} & (0 \leq x \leq 2) \\ 0 & (x \geq 2) \end{cases}$$

Now,

$$h_2 = (f * f * f)_x$$

$$= \int_{x-1}^{x+1} \frac{1}{2} \cdot (f * f)_z \cdot \underbrace{f(z)}_{h_1} \cdot dz$$

(Associative property of convolution)

Again, tracing the signal across the range  
 Between  $(-1 \leq x \leq 1)$

$$(f * f * f)x = \int_{x-1}^0 \frac{2+z}{8} dz + \int_0^{x+1} \frac{2-z}{8} dz$$

$$= \int_{x-1}^0 \frac{1}{4} dz + \int_{x-1}^0 \frac{z}{8} dz + \int_0^{x+1} \frac{1}{4} dz - \int_0^{x+1} \frac{z}{8} dz$$

$$\text{Q. } \frac{1}{4} [z]_{x-1}^0 + \frac{1}{16} [z^2]_{x-1}^0 + \frac{1}{4} [z]_0^{x+1} - \frac{1}{16} [z^2]_0^{x+1}$$

$$= \frac{1}{16} [4(-x-1) - (x-1)^2 + 4(x+1) - (x+1)^2]$$

$$= \frac{1}{16} [6 - 2x^2]$$

✓

Between  $(1 \leq x \leq 3)$

$$(f * f * f)x = \int_{x-1}^2 \frac{2-z}{8} dz$$

$$= \int_{x-1}^2 \frac{1}{4} dz - \int_{x-1}^2 \frac{z}{8} dz$$

$$= \frac{1}{4} \cdot (z)_{x-1}^2 - \frac{1}{16} \cdot (z^2)_{x-1}^2$$

$$= \frac{x^2 - 6x + 9}{16}$$

✓

Between  $(-3 \leq x \leq -1)$

$$(f * f * f)x = \int_{-2}^{x+1} \frac{2+z}{8} dz = \int_{-2}^{x+1} \frac{1}{4} dz + \int_{-2}^{x+1} \frac{z}{8} dz$$

$$= \frac{1}{4} (z)_{-2}^{x+1} + \frac{1}{16} (z^2)_{-2}^{x+1}$$

$$= \frac{(x^2+9+6x)}{16}$$



• Between ( $x \geq 3$ )

$$(f * f * f)x = 0$$



• Between ( $x \leq -3$ )

$$(f * f * f)x = 0$$

$$\therefore h_2 = \begin{cases} 0 & (x \leq -3) \\ \frac{x^2+9+6x}{16} & (-3 \leq x \leq -1) \\ \frac{6-2x^2}{16} & (-1 \leq x \leq 1) \\ \frac{x^2+9-6x}{16} & (1 \leq x \leq 3) \\ 0 & (x \geq 3) \end{cases}$$



(b) We have already evaluated functions  $h_1$  &  $h_2$  in (a), let us summarize at the discrete points rather than summarizing in the range.

$$h_1 = (f * f)x$$

$$\text{For } x = -2 \text{ (or) } x \leq -2$$

$$(f * f)x = 0$$

$$\text{For } x = -1 \text{ (or) } -2 \leq x \leq 0$$



$$(f * f)x = \frac{x+2}{4} = \frac{-1+2}{4} = \frac{1}{4}$$



For  $x = 0$ ,

$$(f * f)x = \frac{2-x}{4} = \frac{1}{2}$$



For  $x = 1$ ,

$$(f * f)x = \frac{1}{4}$$



For  $x = 2$

$$(f * f)x = 0$$



$$* h_2 = (f * f * f)x$$

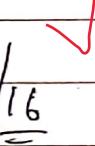
For  $x = -3$

$$(f * f * f)x = 10$$



For  $x = -2$

$$(f * f * f)(-2) = \frac{x^2 + 9 + 6x}{16} = \frac{(-2)^2 + 9 + 6(-2)}{16} = \frac{1}{16}$$



For  $x = -1$ ,

$$\Rightarrow \frac{6 - 2x^2}{16} = \frac{6 - 2(-1)^2}{16} = \frac{1}{4}$$



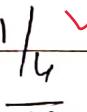
For  $x = 0$ ,

$$\Rightarrow \frac{6 - 2(0)^2}{16} = \frac{6}{16}$$



For  $x = 1$ ,

$$\Rightarrow \frac{(x)^2 + 9 - 6(x)}{16} = \frac{1}{4}$$



For  $x = 2$ ,

$$\Rightarrow \frac{(2)^2 + 9 - 6(2)}{16} = \frac{1}{16}$$



For  $x = 3$ ,

$$\Rightarrow \frac{0}{16}$$



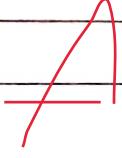
③ Comparison of results b/w continuous convolution & Discrete convolution from P1

→ Differences:

1. In the functions of discrete convolution, we don't see  $\underline{\text{time}}(t)$ , which is a continuous parameter 
2. In Discrete convolution, the parameter  $t$  is replaced by  $n$  (sample)
3. When we compare  $h_2$  of both discrete & continuous convolution, we notice that in a continuous signal, the amplitude is 0 at  $-3$ , peaks at 0 & then vanishes again at  $+3$ . Whereas, in discrete, it is 0 for all values below 0 & then amplitude gradually increases  $\nearrow$  positive values.

Similarity:

1. Continuous convolution would appear discrete if we sample the signal / replace time  $\underline{t}(t)$  by sample  $(n)$ .
4. In discrete, the amplitude is consistently in the form  $\frac{x}{F}$  [here:  $x$  is any integer]  
whereas, in continuous, it is always of the form  $\frac{1}{F} f$

 4/6

3)

### Properties of Convolution [Continuous]

- Commutativity Property  
 $f * g = g * f$

Proof :- w.k.t.

$$f(t) * g(t) = \int_{-\infty}^{\infty} f(\tau) \cdot g(t - \tau) d\tau$$

$(f * g)(t)$

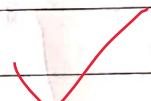
Integral form of continuous signal

Let,  $z = t - \tau$

$$\Rightarrow - \int_{\infty}^{\infty} f(t - z) \cdot g(z) dz$$

$$\Rightarrow \int_{-\infty}^{\infty} g(z) \cdot f(t - z) dz$$

$$= g(t) * f(t)$$



- Associativity Property [Ref: Univ of Tennessee]  
 $(f * g) * h = f * (g * h)$

Proof: w.k.t for a continuous convolution

$$(f * g)(t) = \int_0^t f(z) \cdot g(t - z) dz$$

Now with 3 functions  $f, g & h$

Thus,

$$((f * g) * h)(t) = \int_0^t (f * g)(z) \cdot h(t - z) dz$$

Integrating this part b/w ( $y=0 & z$ )

$$= \int_{z=0}^t \left[ \int_{y=0}^z f(y) g(z-y) dy \right] \cdot h(t-z) dz$$

combining both the integrals

$$= \int_0^t \int_0^z f(y) \cdot g(z-y) \cdot h(t-z) \cdot dy \cdot dz$$

→ Rearranging, the integrals to t.

$$= \int_{y=0}^t \int_{z=y}^t f(y) \cdot g(z-y) \cdot h(t-z) \cdot dz \cdot dy$$

$$y=0 \quad z=y$$

$$= \int_{y=0}^t \int_{z=0}^{t-y} f(y) \cdot g(z) \cdot h(t-z-y) \cdot dz \cdot dy$$

$$= \int_{y=0}^t f(y) \cdot \left[ \int_{z=0}^{t-y} g(z) \cdot h(t-y-z) \cdot dz \right] \cdot dy$$

$$= \int_{y=0}^t f(y) (g * h)(t-y) \cdot dy$$

$$= (\underline{f * (g * h)})(t) \quad (\checkmark)$$

### Distributive Property

$$(f+g) * h = f * h + g * h \quad &$$

$$f * (g+h) = f * g + f * h$$

Proof :-

$$R.H.S \Rightarrow f * g + f * h$$

Applying the conditions of convolution

$$\Rightarrow f(x) * g(x) + f(x) * h(x) \quad (=) \quad (f * g)(x)$$

$$\text{Let, } y_1(x) = f(x) * g(x) \quad &$$

$$y_2(x) = f(x) * h(x)$$

$\therefore$  Adding both  $y_1(x)$  &  $y_2(x)$  together  
i.e.  $y(x) = y_1(x) + y_2(x)$   
 $= f(x) * g(x) + f(x) * h(x)$   
 $= \sum_{z=-\infty}^{\infty} f(z) \cdot g(x-z) + \sum_{z=-\infty}^{\infty} f(z) \cdot h(x-z)$

$\rightarrow$  Taking  $f(z)$  common

$$= \sum_{z=-\infty}^{\infty} f(z) [g(x-z) + h(x-z)]$$
 $= f(z) * (g(x) + h(x))$ 
 $= f * (\underline{g+h}) \quad (\text{L.H.S.})$



$\text{f}$

- Differentiability Property [Ref: Univ. of Tennessee]  
if  $f \in C^0(\mathbb{R})$  &  $g \in C^\infty(\mathbb{R})$   
then,  $(f * g) \in C^\infty(\mathbb{R})$

Proof:

Let  $(f * g)(t)$  be the convolution of  $f(t)$  with  $g(t)$

so,

$$(f * g)(t) = \int_{-\infty}^{\infty} f(\tau) \cdot g(t - \tau) \cdot d\tau$$

Differentiating b.s w.r.t  $t$

$$(f * g)'(t) = \int_{-\infty}^{\infty} f(\tau) \cdot g'(t - \tau) \cdot d\tau$$

$$(f * g)'(t) = \underline{f(t)} \cdot \underline{g'(t)}$$

$\text{f}$

[Applying commutative property]

$$(f * g)'(t) = \underline{g(t)} \cdot \underline{f'(t)} * g(t) \quad | (f * g) \in C^\infty$$

$\text{f}$

- Linearity Property  
 $(\alpha f + \beta g) * h = \alpha (f * h) + \beta (g * h)$

Proof:

Let,  $\star(x) = \alpha f(x) + \beta g(x)$

NOW, substitute  $\star(x)$  in L.H.S

$$h(\star(x)) = h(\alpha f(x) + \beta g(x)) ? \quad (h * f) \neq h(f)$$

Applying Associativity,

$$\begin{aligned} h(\alpha f(x) + \beta g(x)) &= \alpha h(f(x)) + \beta h(g(x)) \\ &= \alpha (f * h) + \beta (g * h) \end{aligned}$$

f

• Shift Invariance

$$(T_b f) * g = T_b(f * g)$$

& translations  $T_b$  with  $(T_b f)(x) = f(x-b)$

Proof:- W.K.T.

$$(T_b f) * g = \int_{-\infty}^{\infty} T_b f(t) \cdot g(x-t) dt$$

Let,  $t = t-b$  over  $f(t)$  as,  $x = x-b$   
 $\Rightarrow \int_{-\infty}^{\infty} f(t-b) \cdot g(x-t) dt$  so,  $t = t-b$

Transform to Laplace Domain & perform integration

Before That, let's assume:  $s = t-b$   
 $\Rightarrow t = s+b$

$$\Rightarrow \int_{-\infty}^{\infty} f(s) \cdot g(x - [s+b]) ds$$

$$= \int_{-\infty}^{\infty} f(s) \cdot g[(x-b) - s] ds$$

~~f~~  $(f * g)(x-b)$

$$\Rightarrow (f * g)(s) \quad \rightarrow \text{Transform back from domain.}$$

$$\Rightarrow \boxed{T_b(f * g)(x)} \quad (\checkmark)$$

2/4