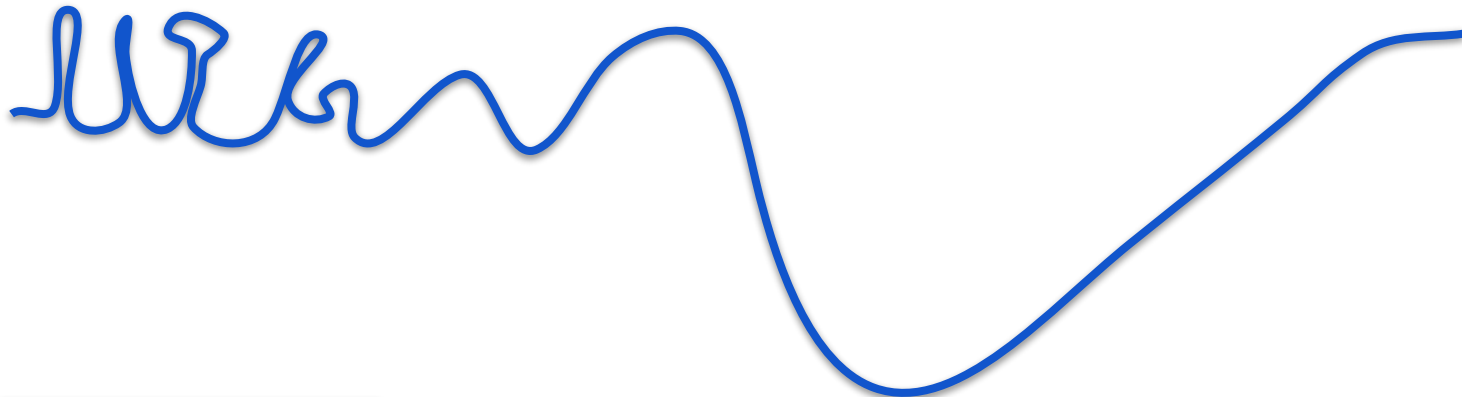


Computing with Signals



DA 623

Jan - May 2024

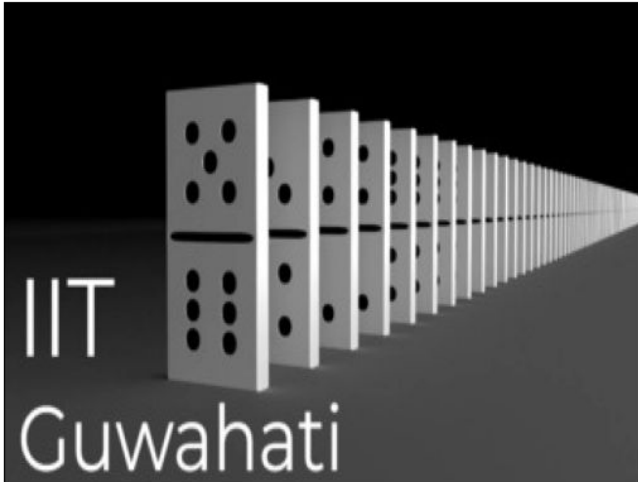
IIT Guwahati

Instructors: Neeraj Sharma

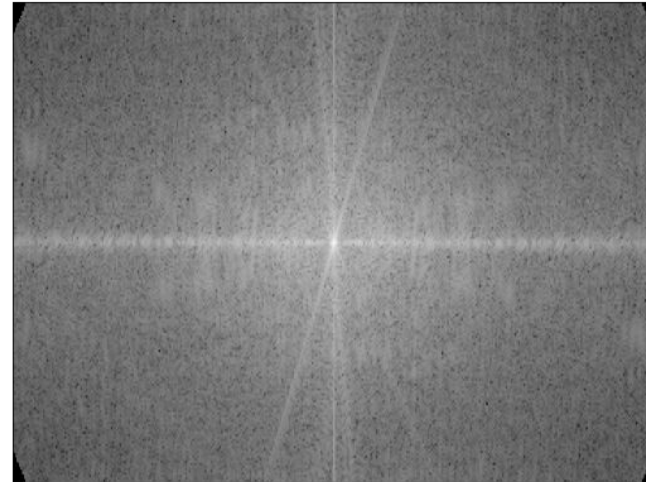
Lecture-18

Using Convolution to Design Linear Image Filters

Lowpass filtered image
(the cutoff frequency is gradually increasing)



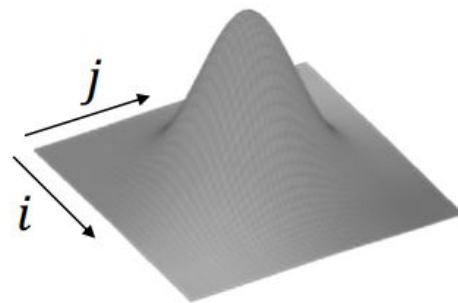
2-D Fourier Spectrum
(in dB, post circular lowpass filtering)



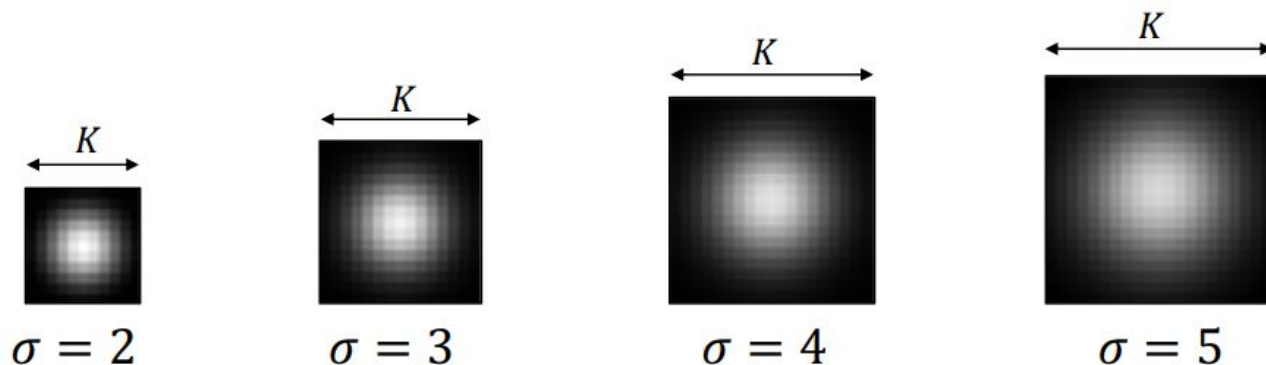
Gaussian Kernel

$$n_{\sigma}[i, j] = \frac{1}{2\pi\sigma^2} e^{-\frac{1}{2}\left(\frac{i^2+j^2}{\sigma^2}\right)}$$

1



σ^2 : Variance



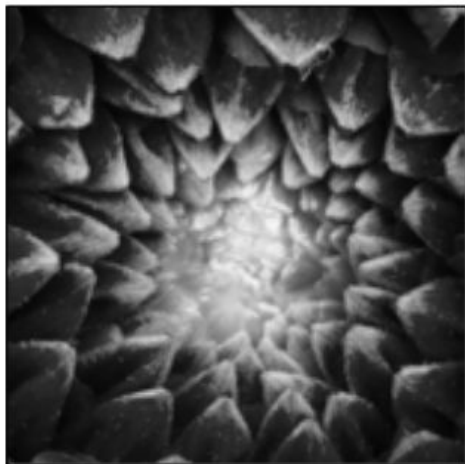
Rule of thumb: Set kernel size $K \approx 2\pi\sigma$

Reference:

<https://fpcv.cs.columbia.edu/Monographs>

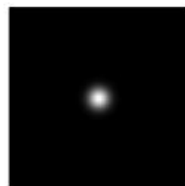
Gaussian Smoothing

Input



$f(x, y)$

*



$\sigma = 4$

=

Output



$g(x, y)$

Larger the kernel (or σ), more the blurring

Reference:
<https://fpcv.cs.columbia.edu/Monographs>

Gaussian Smoothing

Input



$f(x, y)$

*

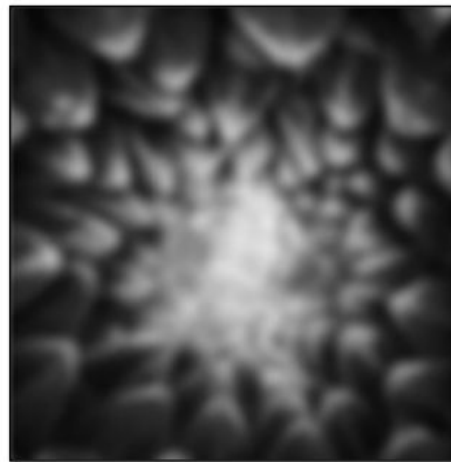


$\sigma = 16$

$n_{16}(x, y)$

=

Output



$g(x, y)$

Larger the kernel (or σ), more the blurring

Gaussian Smoothing is Separable

$$g[i,j] = \frac{1}{2\pi\sigma^2} \sum_{m=1}^K \sum_{n=1}^K e^{-\frac{1}{2}\left(\frac{m^2+n^2}{\sigma^2}\right)} f[i-m, j-n]$$

$$g[i,j] = \frac{1}{2\pi\sigma^2} \sum_{m=1}^K e^{-\frac{1}{2}\left(\frac{m^2}{\sigma^2}\right)} \cdot \sum_{n=1}^K e^{-\frac{1}{2}\left(\frac{n^2}{\sigma^2}\right)} f[i-m, j-n]$$

Using One 2D Gaussian Filter \equiv Using Two 1D Gaussian Filters

The diagram illustrates the separability of Gaussian smoothing. It shows that a 2D Gaussian filter applied to an image f is equivalent to applying two 1D Gaussian filters sequentially. The first 1D filter is vertical with height K , and the second is horizontal with width K .

Gaussian Smoothing is Separable

Using One 2D Gaussian Filter \equiv Using Two 1D Gaussian Filters

$$f * \begin{array}{c} \text{2D Gaussian Filter} \\ \leftarrow K \rightarrow \end{array} = f * \begin{array}{c} \text{1D Gaussian Filter (vertical)} \\ \updownarrow K \end{array} * \begin{array}{c} \text{1D Gaussian Filter (horizontal)} \\ \leftarrow K \rightarrow \end{array}$$

Which one is faster? Why?

K^2 Multiplications

$K^2 - 1$ Additions

$2K$ Multiplications

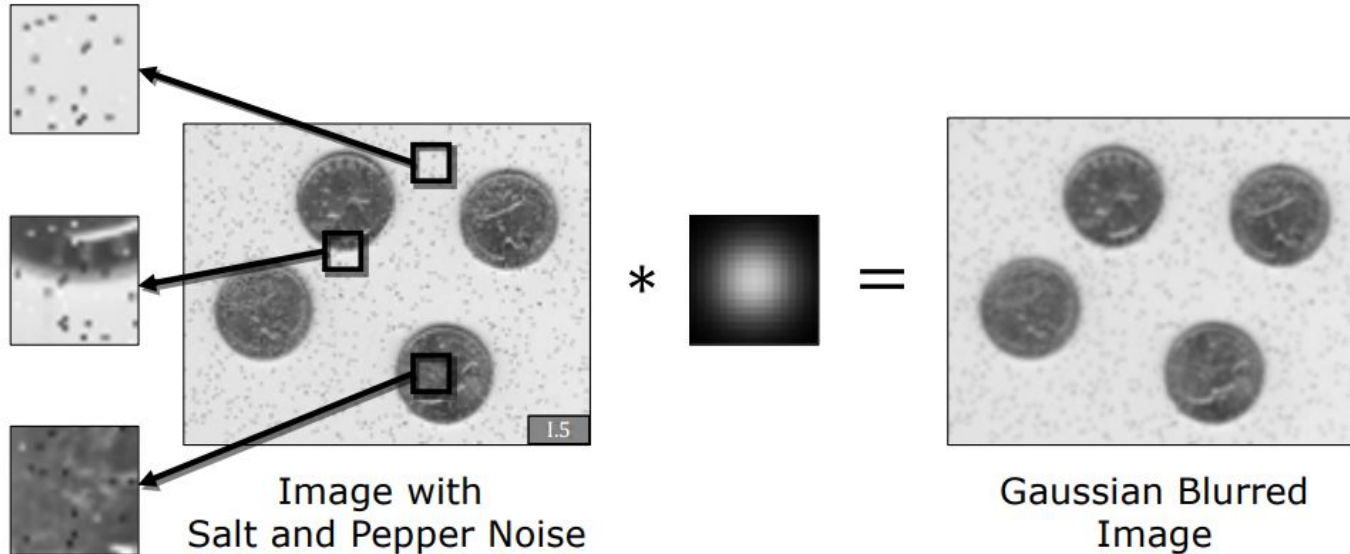
$2(K - 1)$ Additions

Reference:

<https://fpcv.cs.columbia.edu/Monographs>

Non-Linear Image Filters

Smoothing to Remove Image Noise



Problem with Smoothing:

- Does not remove outliers (Noise)
- Smooths edges (Blur)

Median Filtering

1. Sort the K^2 values in window centered at the pixel
2. Assign the Middle Value (Median) to pixel

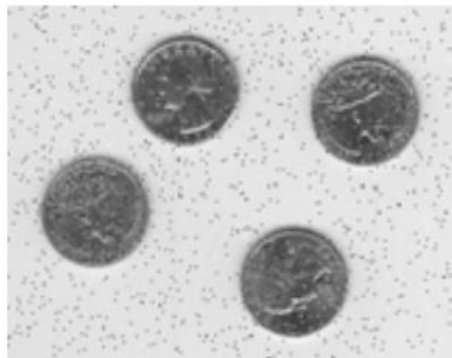
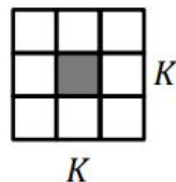


Image with
Salt and Pepper Noise



Median Filtered
Image ($K = 3$)

Non-linear Operation
(Cannot be implemented using convolution)

Median Filtering

Not Effective when Image Noise is not a Simple Salt and Pepper Noise.

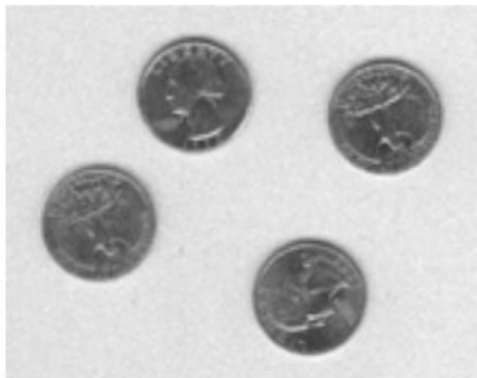
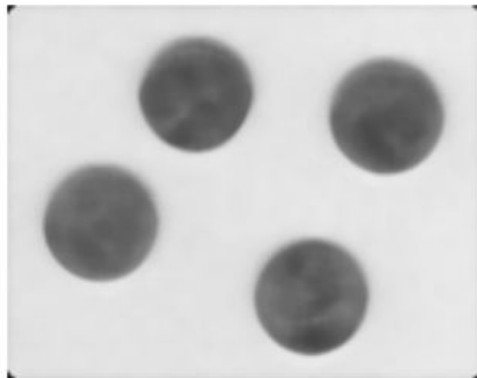


Image with Noise



Median Filtered
Image ($K = 11$)

Larger K causes blurring of image detail

Template Matching

Template Matching



Template

How do we locate the template in the image?

Minimize:

$$E[i, j] = \sum_m \sum_n (f[m, n] - t[m - i, n - j])^2$$

Template Matching



Template

How do we locate the template in the image?

Minimize:

$$E[i, j] = \sum_m \sum_n (f[m, n] - t[m - i, n - j])^2$$

$$E[i, j] = \sum_m \sum_n (f^2[m, n] + t^2[m - i, n - j] - \frac{2f[m, n]t[m - i, n - j]}{\text{Maximize}})$$

1

Maximize

Template Matching



Template

How do we locate the template in the image?

Maximize:

$$R_{tf}[i,j] = \sum_m \sum_n f[m,n]t[m-i,n-j] = t \otimes f$$

(Cross-Correlation)

Convolution vs. Correlation

Convolution:

$$g[i, j] = \sum_m \sum_n f[m, n] \underline{t[i - m, j - n]} = t * f$$

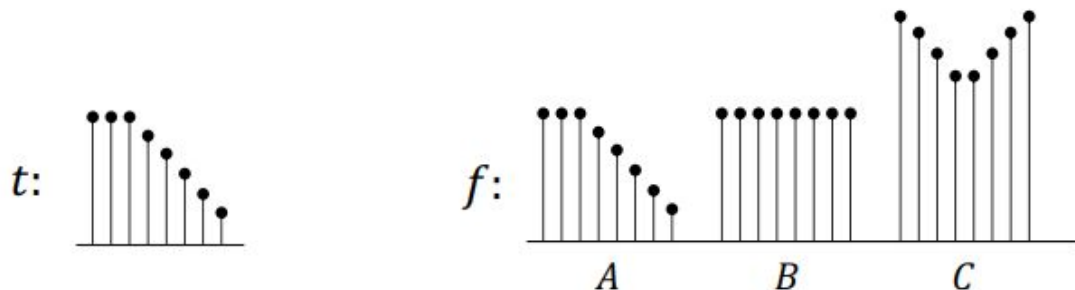
Correlation:

$$R_{tf}[i, j] = \sum_m \sum_n f[m, n] \underline{t[m - i, n - j]} = t \otimes f$$

No Flipping in Correlation

Problem with Cross-Correlation

$$R_{tf}[i,j] = \sum_m \sum_n f[m,n]t[m-i,n-j] = t \otimes f$$



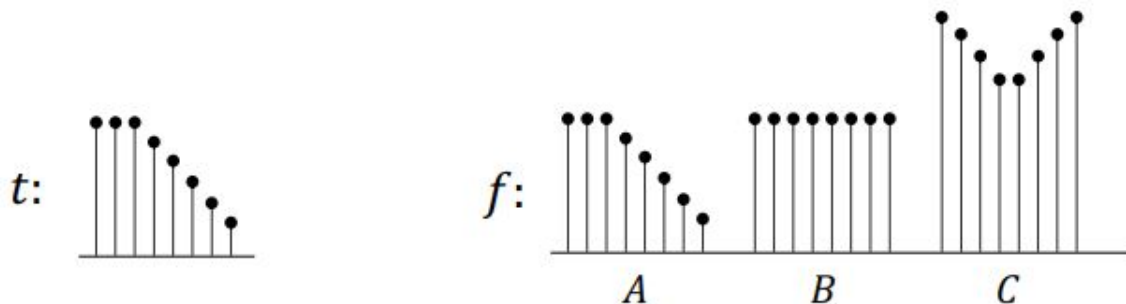
$$R_{tf}(C) > R_{tf}(B) > R_{tf}(A)$$

We need $R_{tf}(A)$ to be the maximum!

Normalized Cross-Correlation

Account for energy differences

$$N_{tf}[i,j] = \frac{\sum_m \sum_n f[m,n]t[m-i,n-j]}{\sqrt{\sum_m \sum_n f^2[m,n]} \sqrt{\sum_m \sum_n t^2[m-i,n-j]}}$$



$$N_{tf}(A) > N_{tf}(B) > N_{tf}(C)$$

Normalized Cross-Correlation

Account for energy differences

$$N_{tf}[i,j] = \frac{\sum_m \sum_n f[m,n]t[m-i,n-j]}{\sqrt{\sum_m \sum_n f^2[m,n]} \sqrt{\sum_m \sum_n t^2[m-i,n-j]}}$$

