

Module: Computer Vision - Session 2

Computer Vision

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**Online Training & Certification Course on Artificial Intelligence
& Machine Learning
Defence Institute of Advanced Technology (DU), Pune.**



Computer Vision

Dr Sunita Dhavale
Image Transformation



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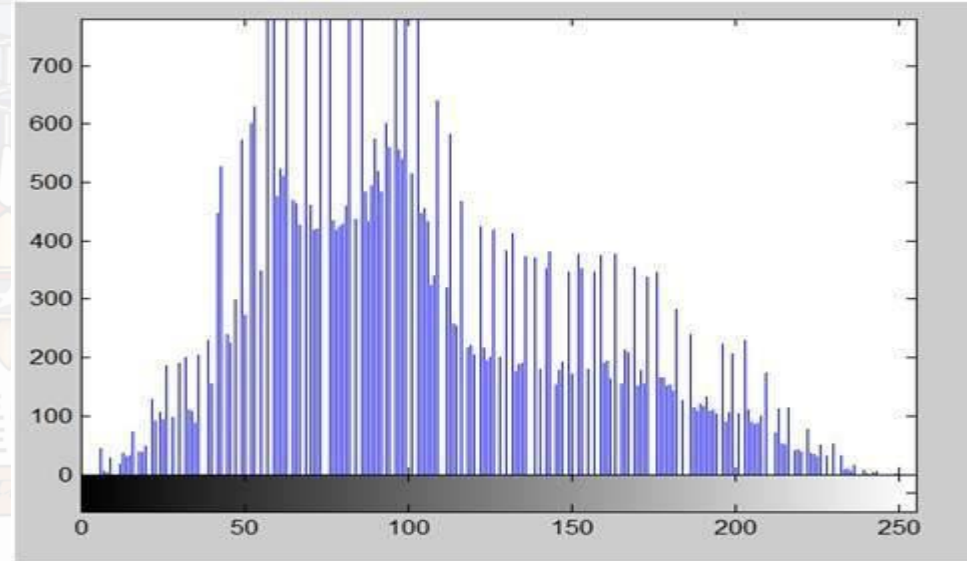
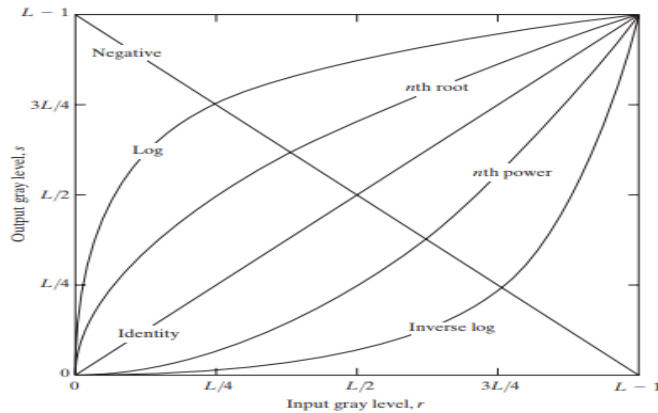
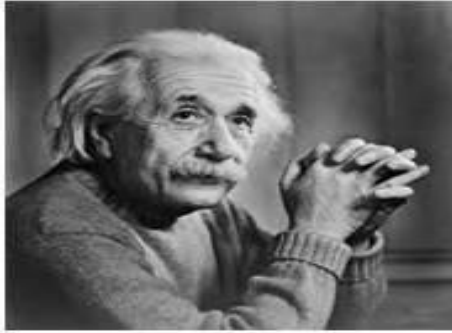


Outline of Presentation

- Image Transformation
- Contrast enhancement
- Histogram Equalization
- Convolution
- Filters
- Noise
- DFT



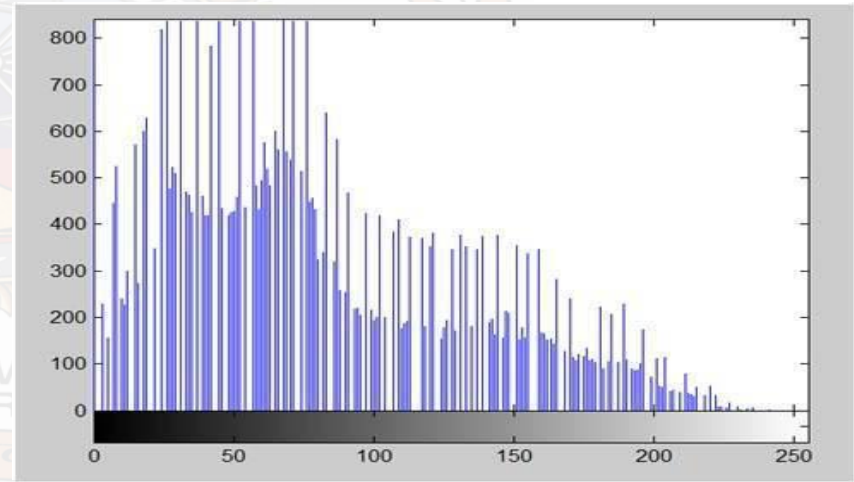
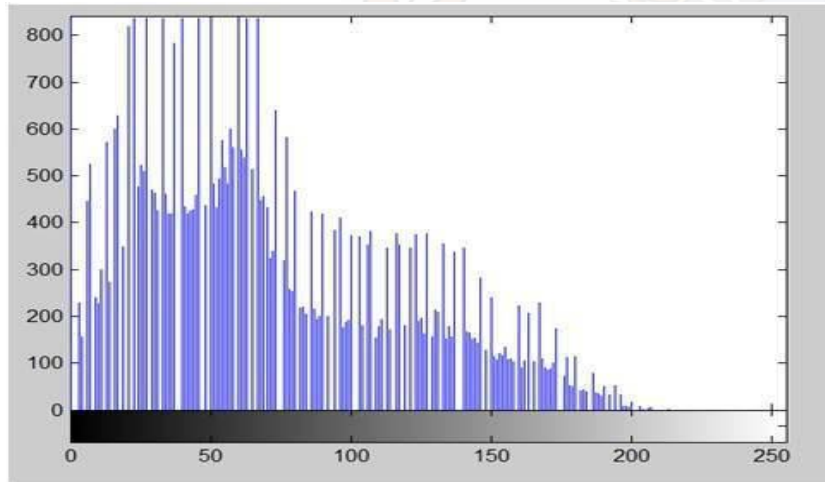
Image Transformation





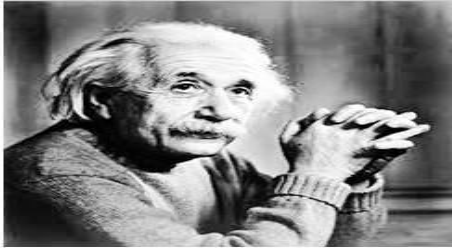
Contrast Enhancement

$$g(x,y) = \frac{f(x,y)-f_{\min}}{f_{\max}-f_{\min}} * 2^{bpp}$$

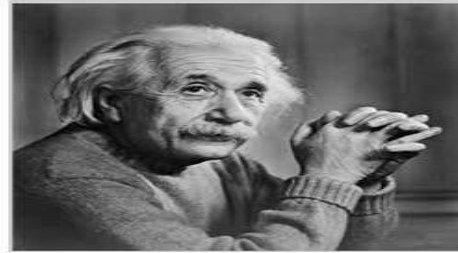


Histogram Equalization

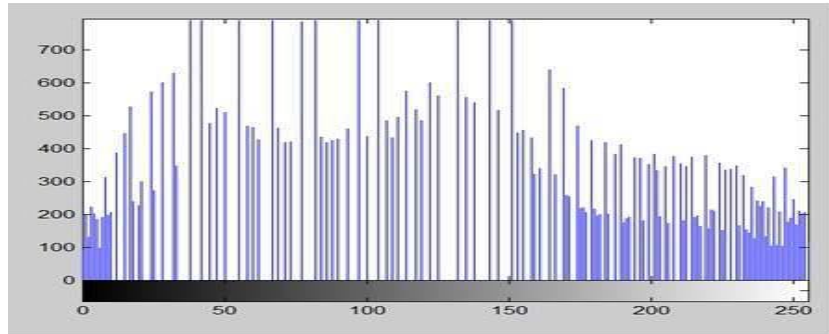
New Image



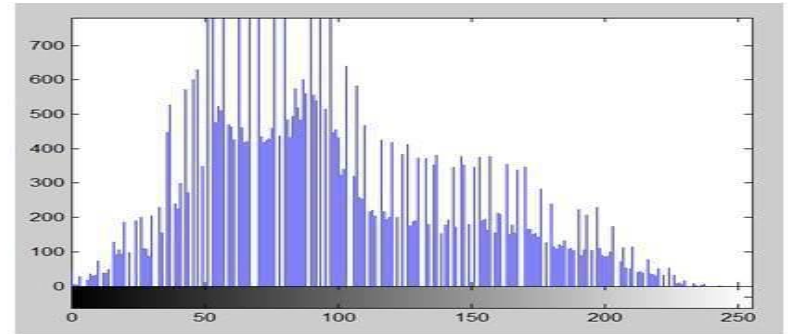
Old image



New Histogram



Old Histogram



Convolution/Spatial/Time domain Filtering

- $g(x,y) = h(x,y) * f(x,y) = f(x,y) * h(x,y)$
- convolution operator (*)





Example of convolution

1	2	3	3	2	1	9	8	7	2	4	6
4	5	6	6	5	4	6	5	4	8	10	12
7	8	9	9	8	7	3	2	1	14	16	18
Original mask			Flipping the mask horizontally			Flipping the mask vertically			Original Image		

9		8		7	
6	2	5	4	4	6
3	8	2	10	1	12
	14		16		18



Types of Filters

- linear/spatial domain filters or frequency domain filters
- Box/Mean/average filter,
- Weighted average filter
- Gaussian filter.



Noise

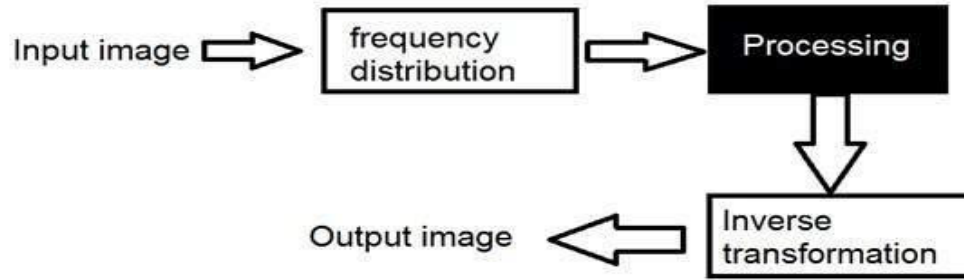
- Speckle/random noise
- Transient frequency spike
- Jitter and judder line noise
- Motion blur
- Gaussian noise
- Impulse noise
- Salt and pepper noise
- common measure of noise is the signal to noise ratio
- additive noise model
- multiplicative noise model

$$S/Nratio = \frac{\sum_{(i,j)} f^2(i,j)}{\sum_{(i,j)} v^2(i,j)}$$

$$f(i,j) = g(i,j) + v(i,j)$$

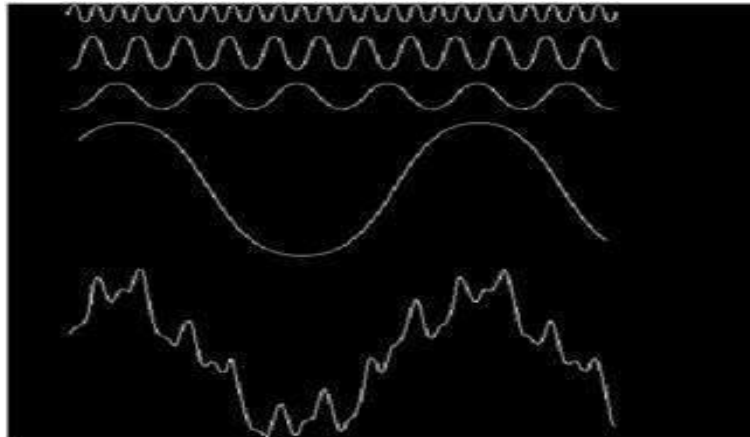
$$f(i,j) = g(i,j) + g(i,j).v(i,j)$$

Frequency domain analysis



$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi(ux+vy)} dx dy$$

$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{j2\pi(ux+vy)} du dv$$



Discrete Fourier transform



Ideal high pass filter, Ideal low pass filter, Gaussian high pass filter, Gaussian low pass filter

$$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$

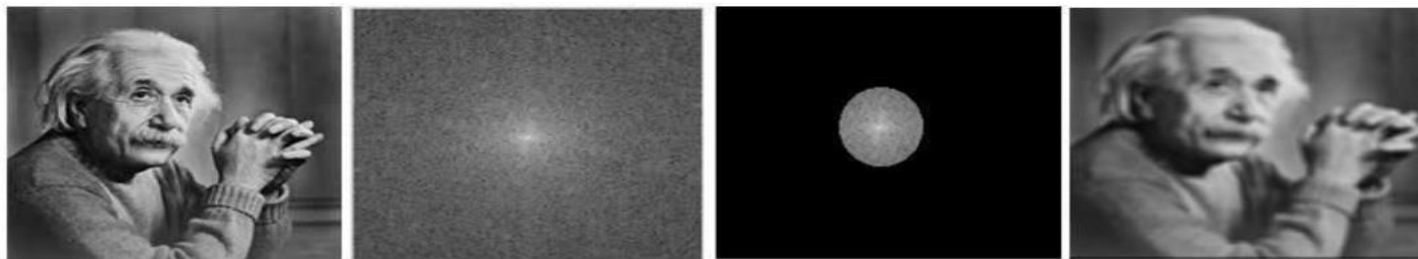
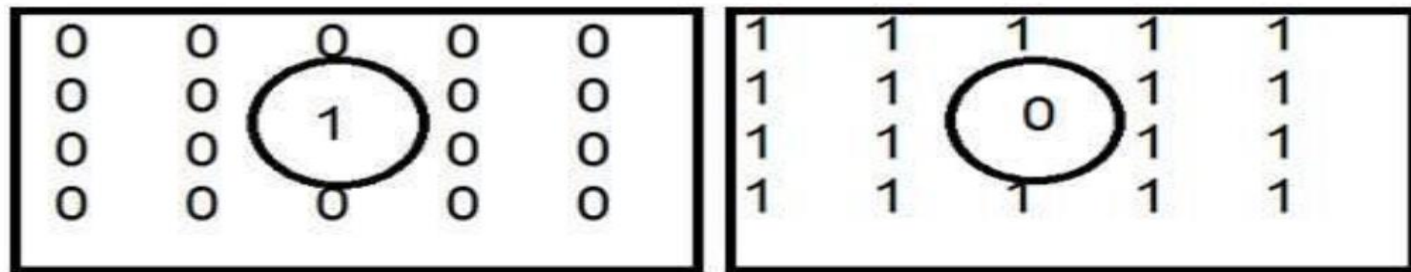
$$f(x, y) = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi(ux/M + vy/N)}$$

$$G(u, v) = F(u, v) \times H(u, v)$$



Filtering an image in frequency domain

Ideal low pass and Ideal High pass filters



Sample image, Image in frequency domain, Applying LPF filter over this image, Resultant Image

- Histogram Equalization
- Gaussian Filter
- DFT





Reference Material

- 1. E. R. Davies, "Computer & Machine Vision", Fourth Edition, Academic Press, 2012.
- 2. R. Szeliski, "Computer Vision: Algorithms and Applications", Springer 2011.
- 3. Simon J. D. Prince, "Computer Vision: Models, Learning, and Inference", Cambridge University Press, 2012.
- 4. Mark Nixon and Alberto S. Aquado, "Feature Extraction & Image Processing for Computer Vision", Third Edition, Academic Press, 2012.
- 5. Sunita Dhavale, "Advanced Image-Based Spam Detection and Filtering Techniques", Book Published by CyberTech: An Imprint of MKP Technologies, Hershey, PA, USA IGI Global, March 2017, ISBN13: 9781683180135|ISBN10: 1683180135|EISBN13: 9781683180142|DOI: 10.4018/978-1-68318-013-5.
- Gonzalez and R. Woods Digital Image Processing, Addison-Wesley Publishing Company, 1992, p 442.



<<Epilogue>>

- We will meet in next scheduled lecture.
- Implement algorithms using python.
- Feel free to ask your questions.
- Email: sunitadhavale@diat.ac.in



Thank You!

