

CS4243

Computer Vision & Pattern Recognition

AY 2023/24

Lab Session 1



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Arrangement

- Part 1 – Quick Recap from the Lecture (~20 min)
- Part 2 – Lab Tutorial (~30 min)
- Break (10 min)
- Part 3 – Lab Solution (~30 min)

Lab Materials

- GitHub Repo:
https://github.com/ldkong1205/cs4243_lab
- Slides
- Notebook & Solution
- Other Materials (image, media, etc.)

Canvas Chat

≡

CS4243 > Chat

Files
Syllabus
Outcomes
Rubrics
Quizzes
Modules
Collaborations
New Analytics
Studio
Zoom
Videos/Panopto
Student Feedback
Attendance
Course Readings
Chat
Course Roster
Microsoft Teams classes
Microsoft Teams meetings
Settings


Course chat

22 people online ▼
New message alerts

Today

KL

KONG LINGDONG 28/08/23 09:45
 Good morning!
 For those who are attending today's lab sessions, please download the following materials:
 Slides: https://github.com/ldkong1205/cs4243_lab/blob/main/slides/lab1_slides.pdf
 Notebook: https://github.com/ldkong1205/cs4243_lab/blob/main/notebook/lab1.ipynb

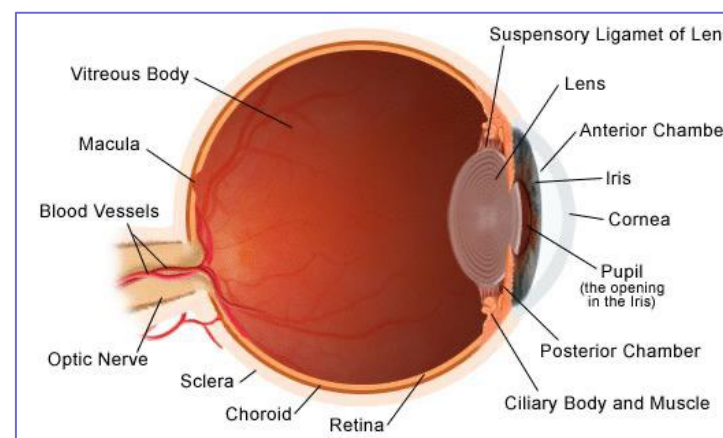

Send



Lesson 1

Introduction and Fundamentals

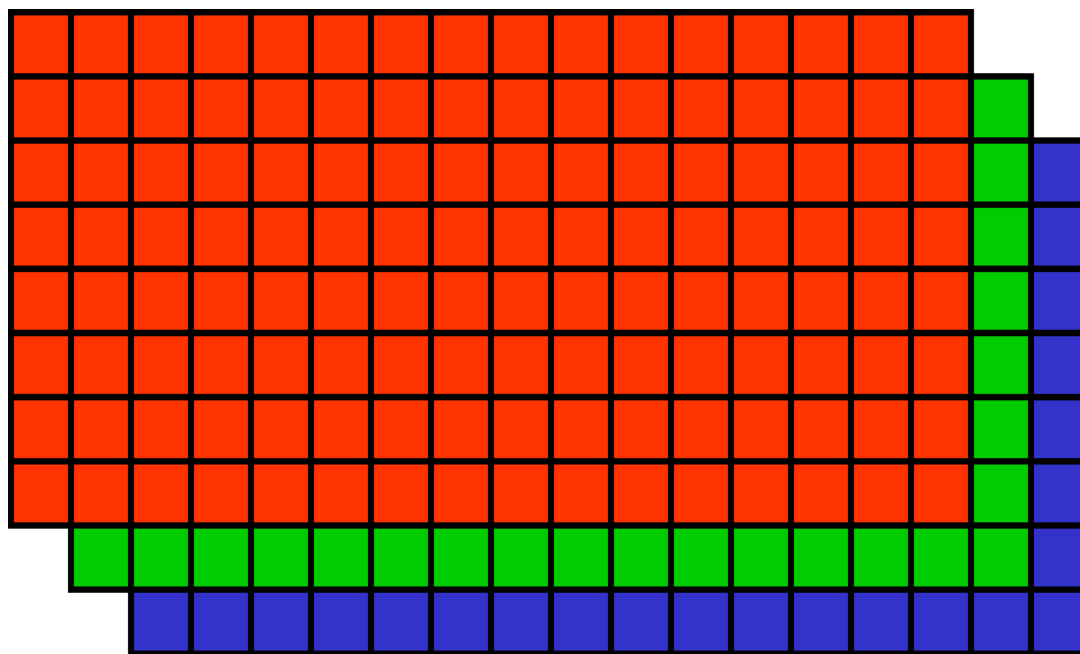
What is this?



A Color Image

- An $M \times N \times 3$ matrix
- Each pixel is an 8-bit number between 0 and 255
- RED, GREEN, BLUE
- $M \times N$
 = resolution
 = number of pixels in
 columns / rows

Caution: OpenCV uses BGR format



Operations

Anything Can be applied on your
Digital Image or Video

Logical
Operations

Transforms

Statistical
Operations

Geometrical
Operations

Mathematical
Operations

Morphology,
Coding,

Fourier,
Walsh, PCA

Histograms,
Correlation,
Max/Min

Affine
Transforms

Filtering, **

Histogram

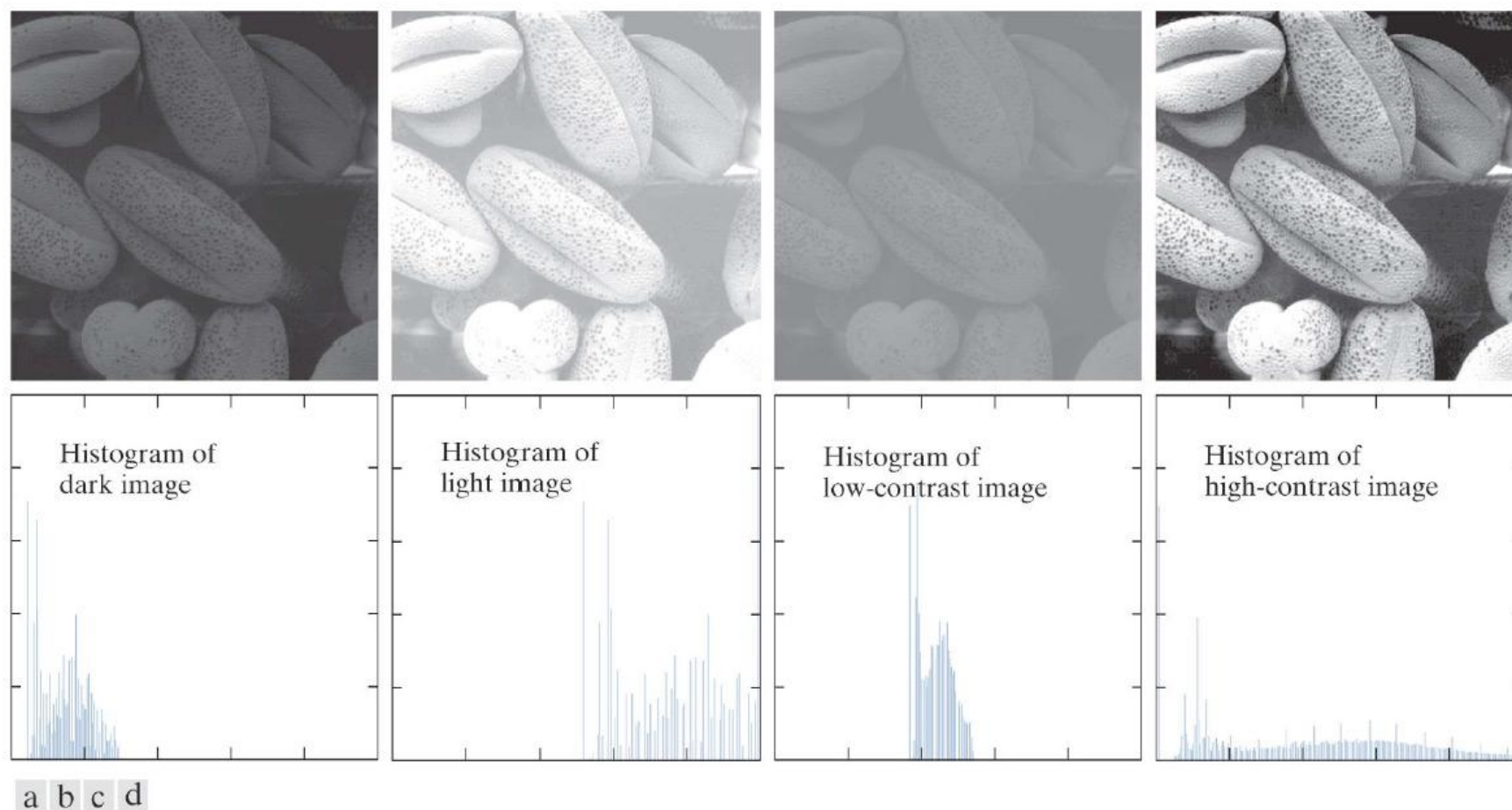


FIGURE 3.16

Four image types and their corresponding histograms. (a) dark; (b) light; (c) low contrast; (d) high contrast. The horizontal axis of the histograms are values of r_k and the vertical axis are values of $p(r_k)$.



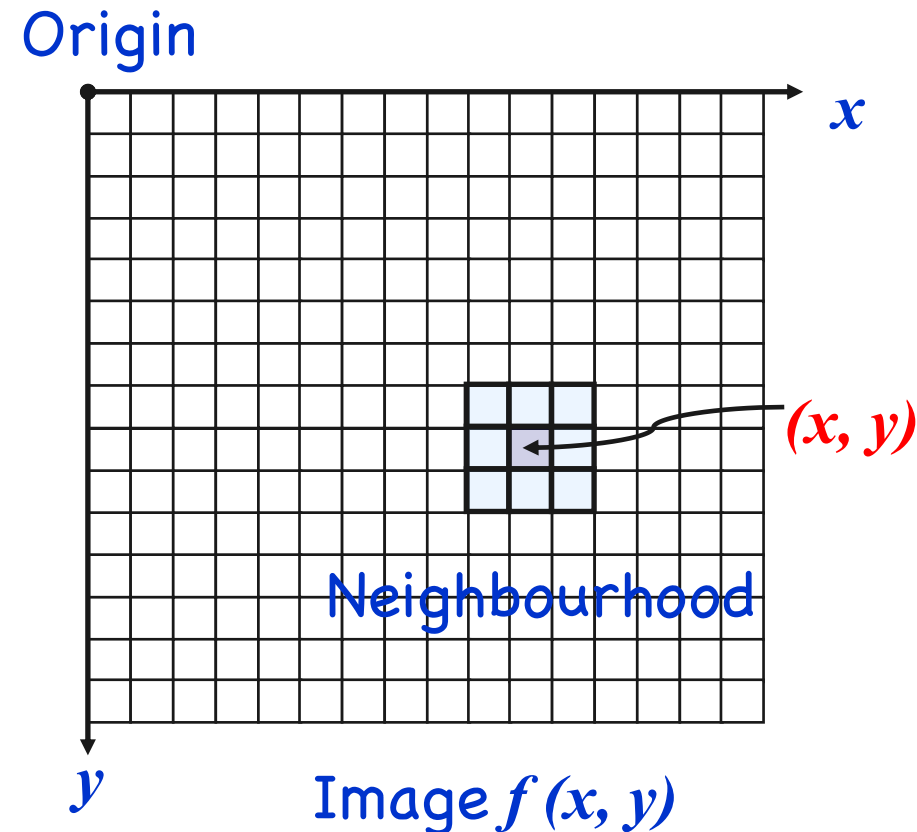
Lesson 2

Local Operations, Filtering, and Convolution

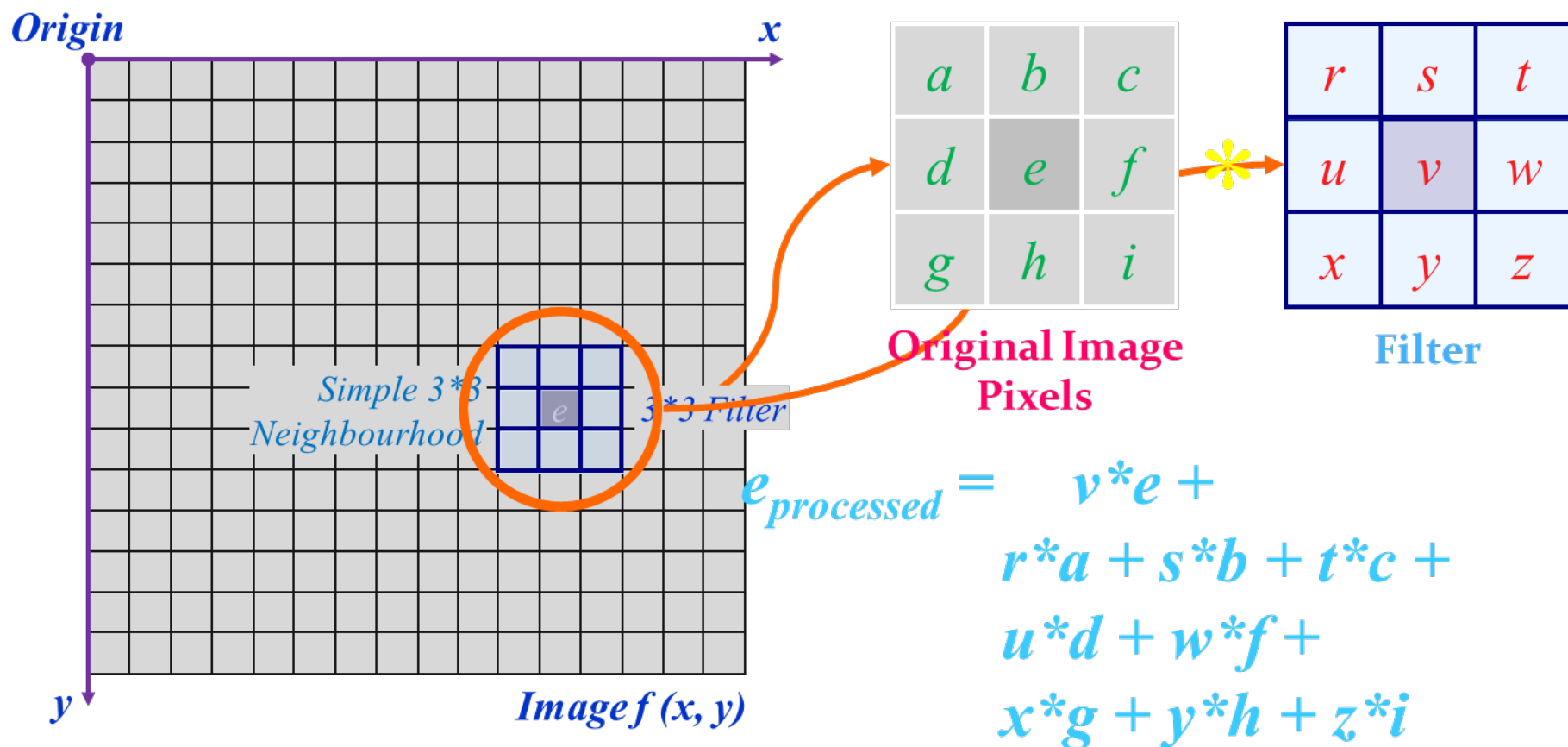
- Where we see how we can filter an image
- Every system in this big universe is a filter

Local Operations

- An image in the spatial domain
- A local operator uses the pixel and its neighbours to compute the new value of that pixel.
- E.g., replace each pixel $f(x,y)$, with the average of that and its 8 neighbours.
- $g(x,y) = T[f(x \pm \Delta x, y \pm \Delta y)]$

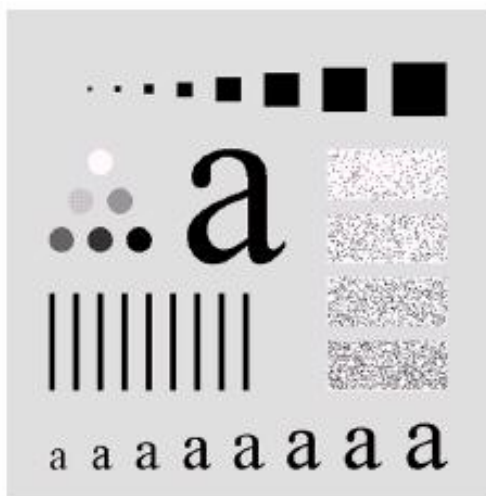


Local Operations

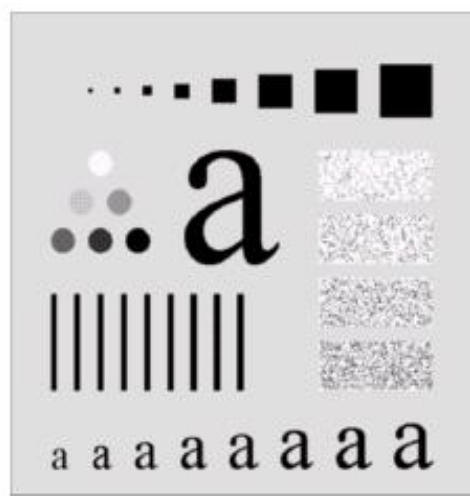


The **filter** would slide over the image, shift or stride is 1 column and then 1 row

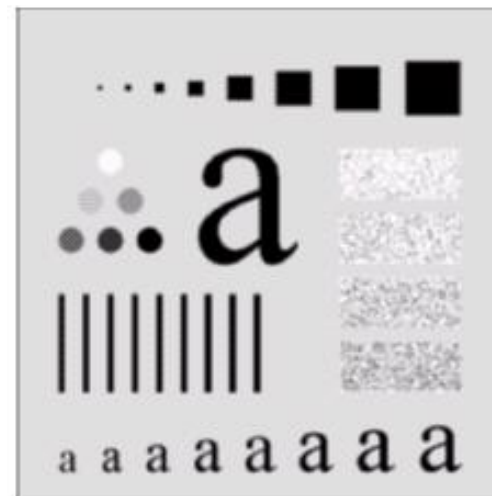
Low-Pass Filters



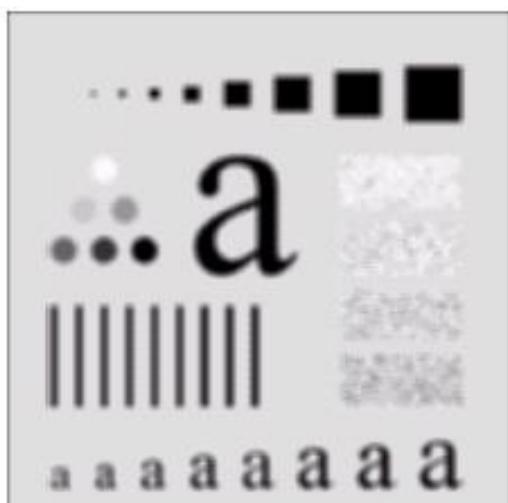
Original image



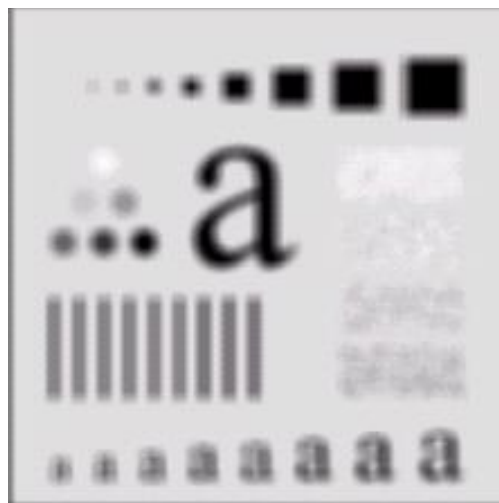
Filtered with 3x3



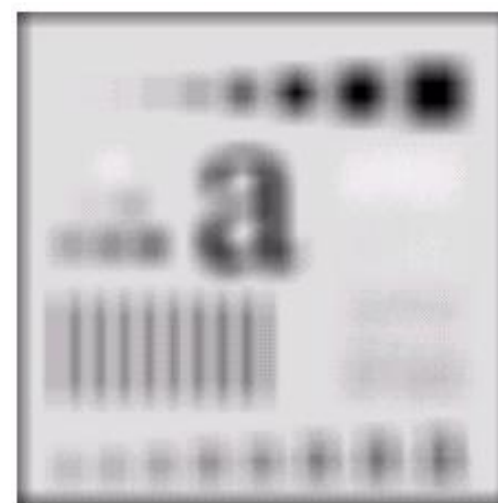
Filtered with 5x5



Filtered with 9x9

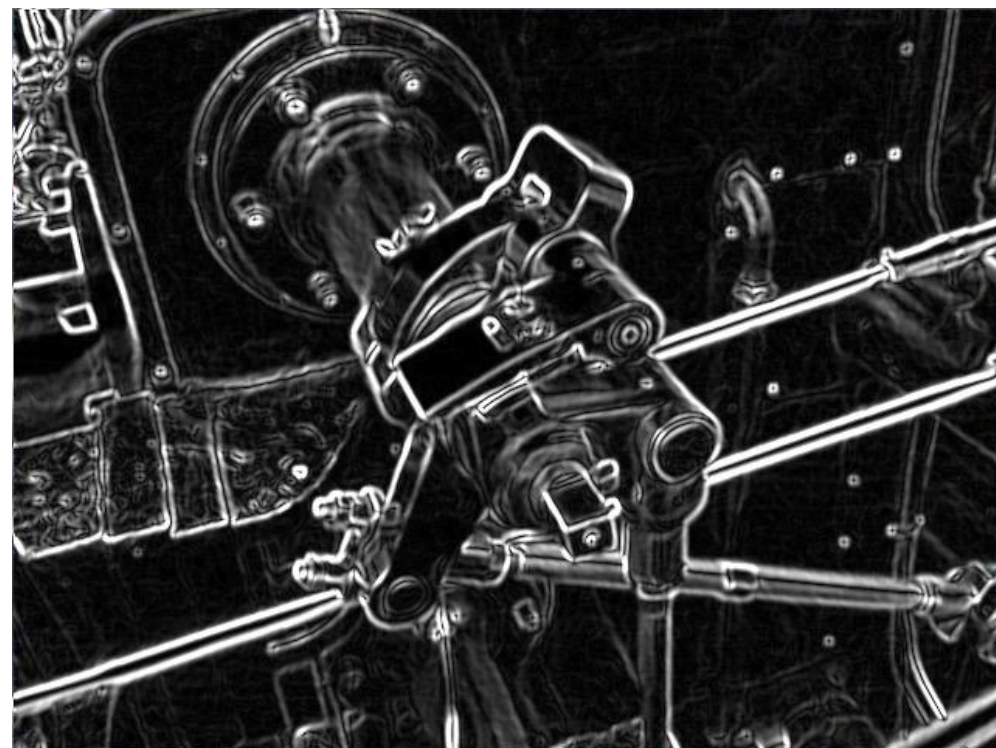
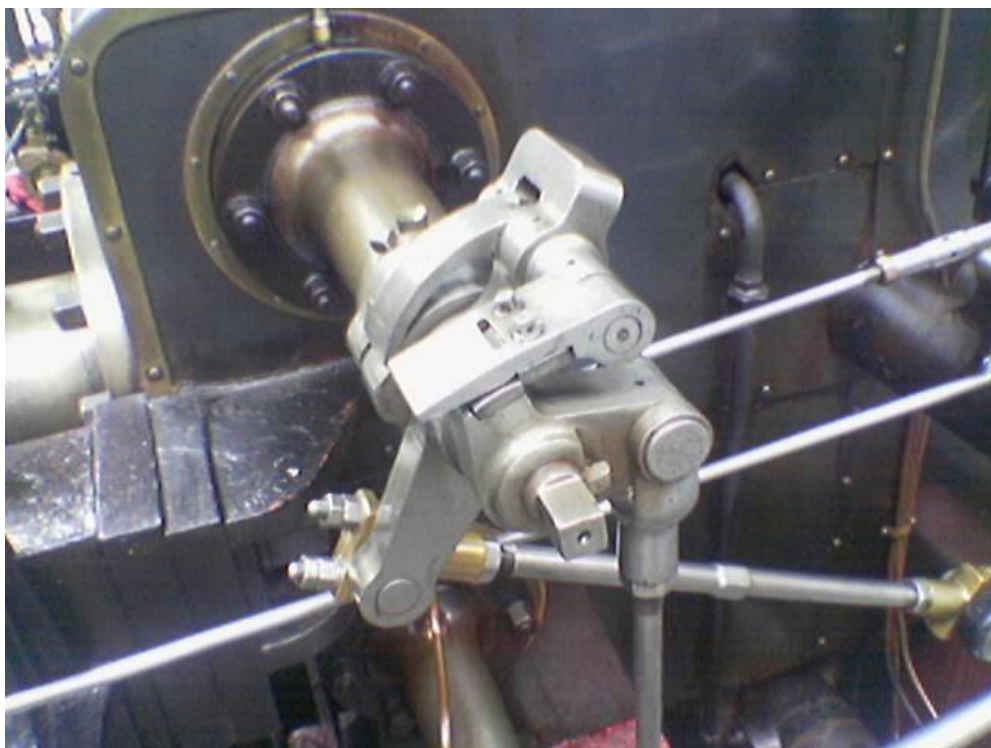


Filtered with 15x15



Filtered with 35x35

High-Pass Filters

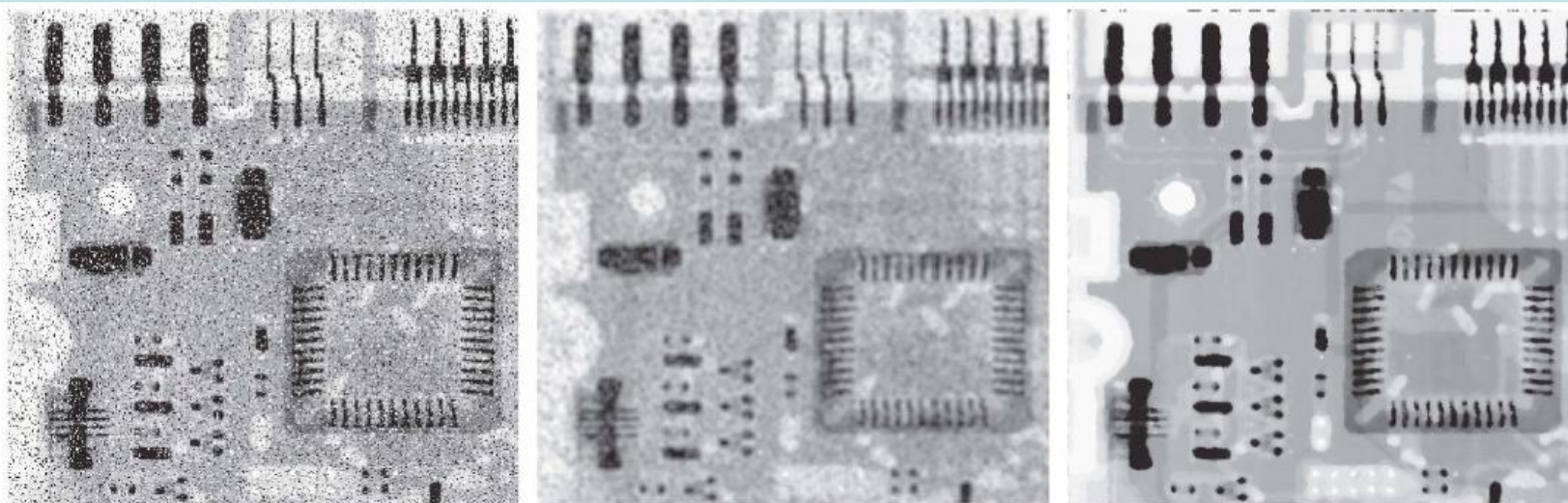


Discrete 1st derivative or **gradient** of an image.

An image and result of **edge detection** using the Sobel algorithm.

Median Filters

- In a small $n \times n$ patch/neighborhood
- Sort the pixels and find the median, M
- Replace the central pixel of your patch with M
- Slide one pixel to the next neighborhood

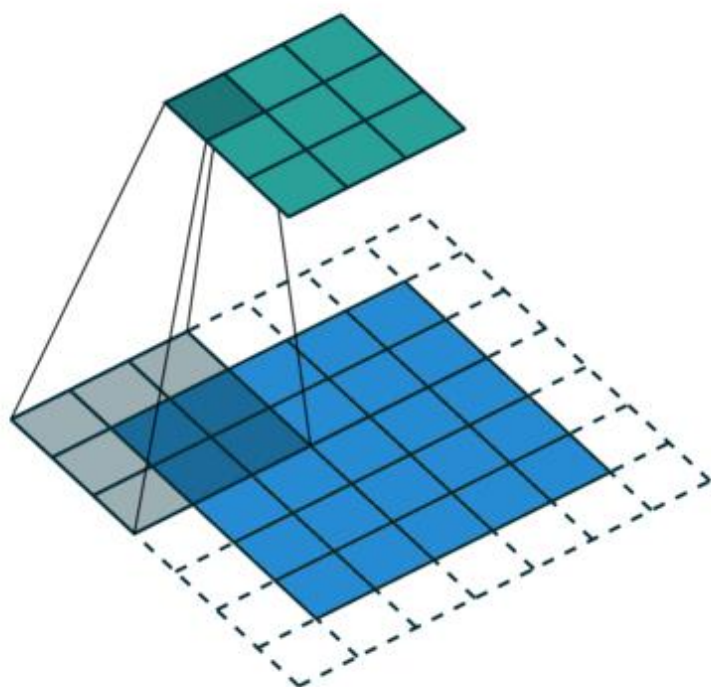


a b c

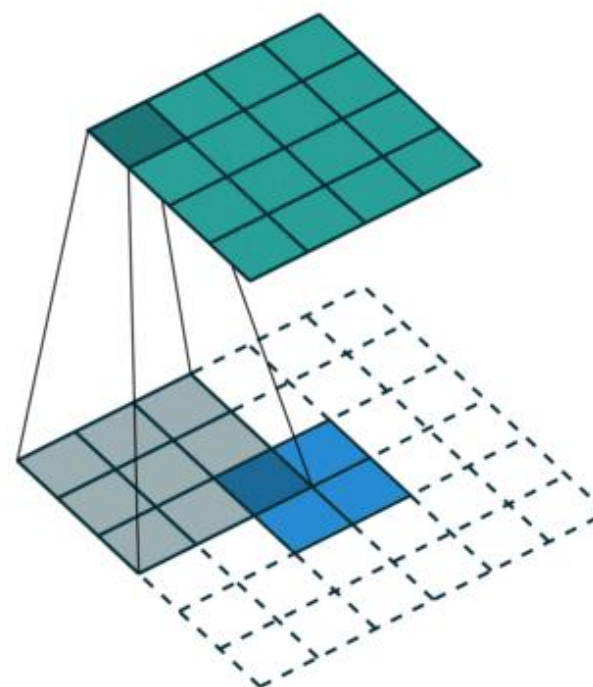
FIGURE 3.49

(a) X-ray image of a circuit board, corrupted by salt-and-pepper noise. (b) Noise reduction using a 19×19 Gaussian lowpass filter kernel with $\sigma = 3$. (c) Noise reduction using a 7×7 median filter. (Original image courtesy of Mr. Joseph E. Pascente, Lixi, Inc.)

Convolution, 2D

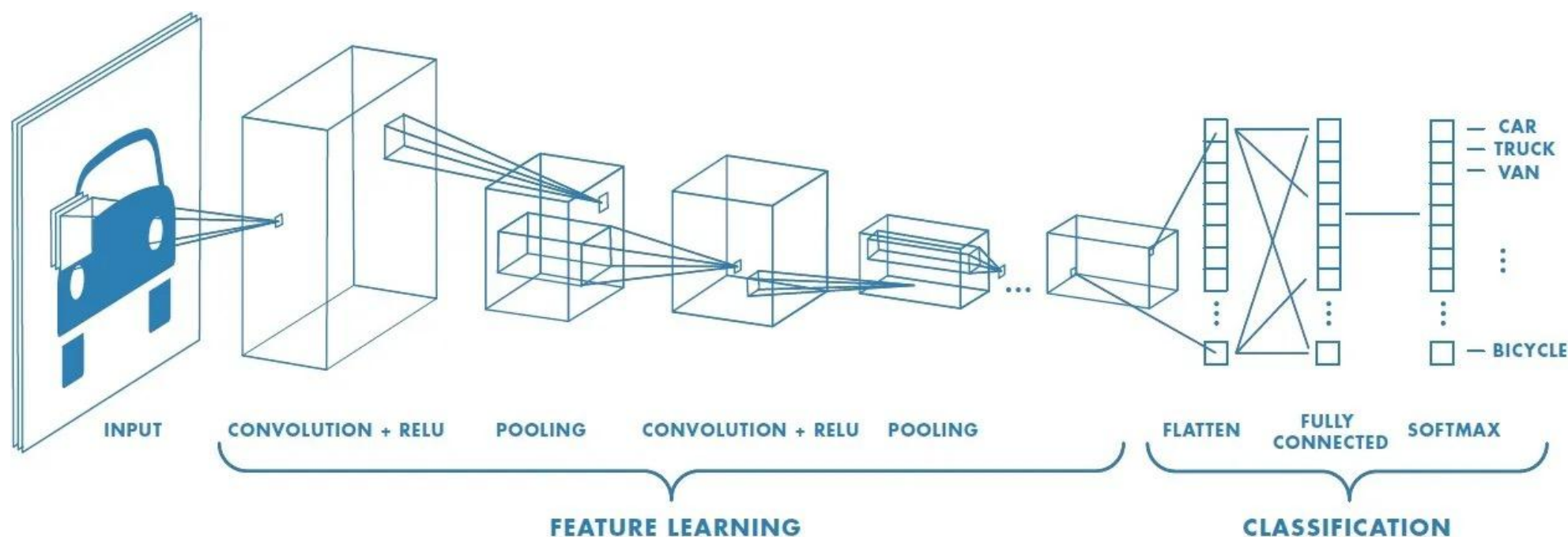


convolution with stride = 1



convolution with stride = 2

Convolutional Neural Nets (CNNs)



Lab Session 1

Background & Global Operations



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