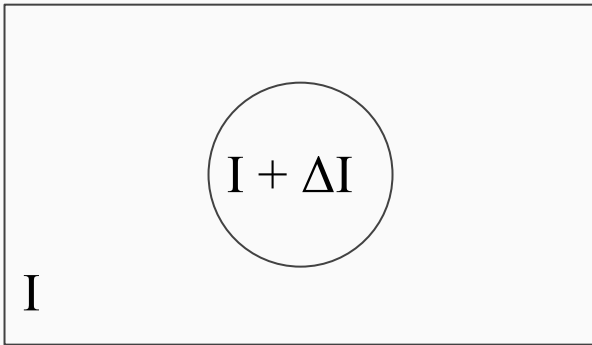


Human Visual System (HVS)

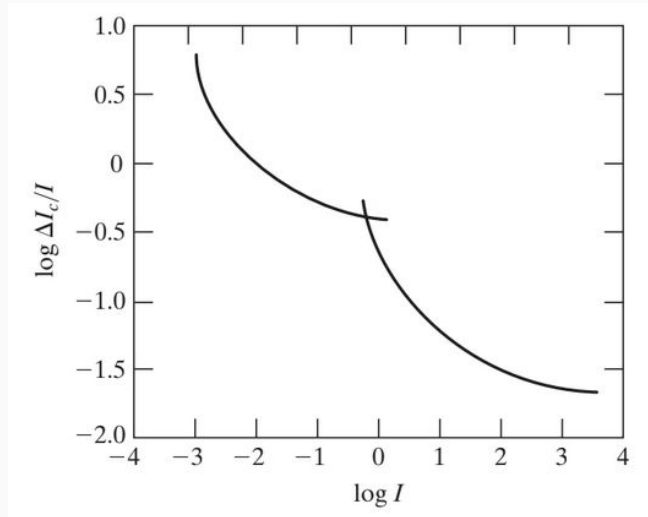
Brightness Discrimination



- I = background intensity
- ΔI = change in intensity required for “just noticeable difference”
- Weber ratio = $\Delta I/I$
- A smaller Weber ratio \rightarrow only a small intensity change is distinguishable (good brightness discrimination)

Human Visual System (HVS)

Brightness Discrimination



- Weber ratio as a function of intensity
- The power of brightness discrimination increases with the background intensity level
- Need “contrast stretching” for poorly illuminated images (Week 2)

Human Visual System (HVS)

Brightness Discrimination

| | | |
|---|----|---|
| 0 | 0 | 0 |
| 0 | 10 | 0 |
| 0 | 0 | 0 |

$I = 0$

$$\Delta I = 10$$

| | | |
|-----|-----|-----|
| 255 | 255 | 255 |
| 255 | 245 | 255 |
| 255 | 255 | 255 |

$I = 255$

Human Visual System (HVS)

Brightness Discrimination

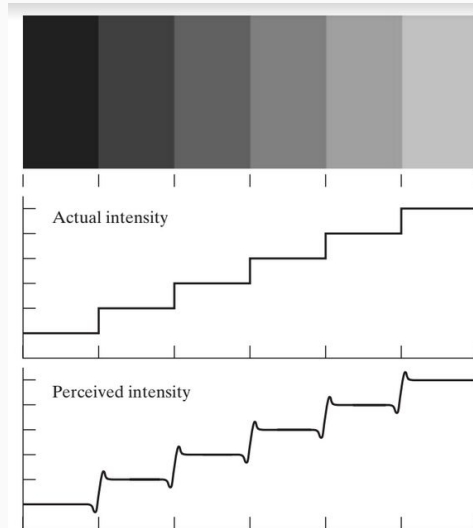


Human Visual System (HVS)

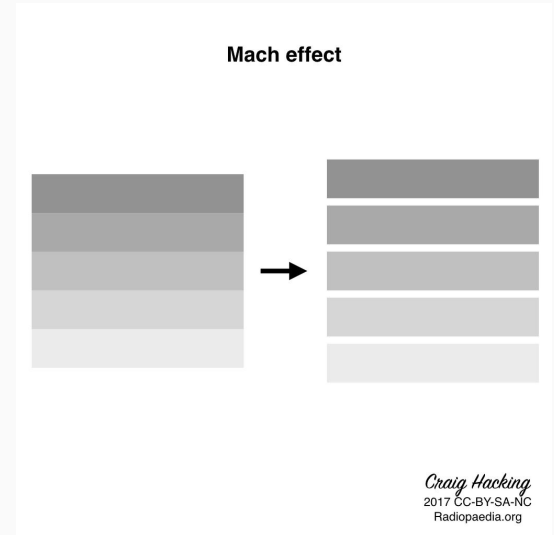
Brightness Discrimination (Mach bands)



By The original uploader was Aliwiki at French Wikipedia. - Transferred from fr.wikipedia to Commons by Korrigan using CommonsHelper., FAL, <https://commons.wikimedia.org/w/index.php?curid=4770182>



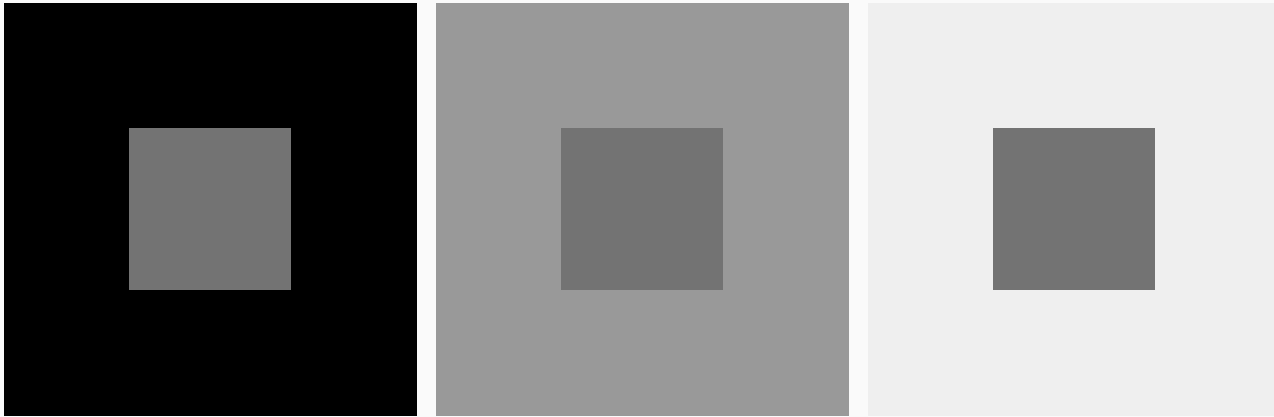
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Craig Hacking
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Radiopaedia.org

Human Visual System (HVS)

Optical Illusion Due to Perceived Brightness



Identify the middle-box with the highest intensity

Human Visual System (HVS)

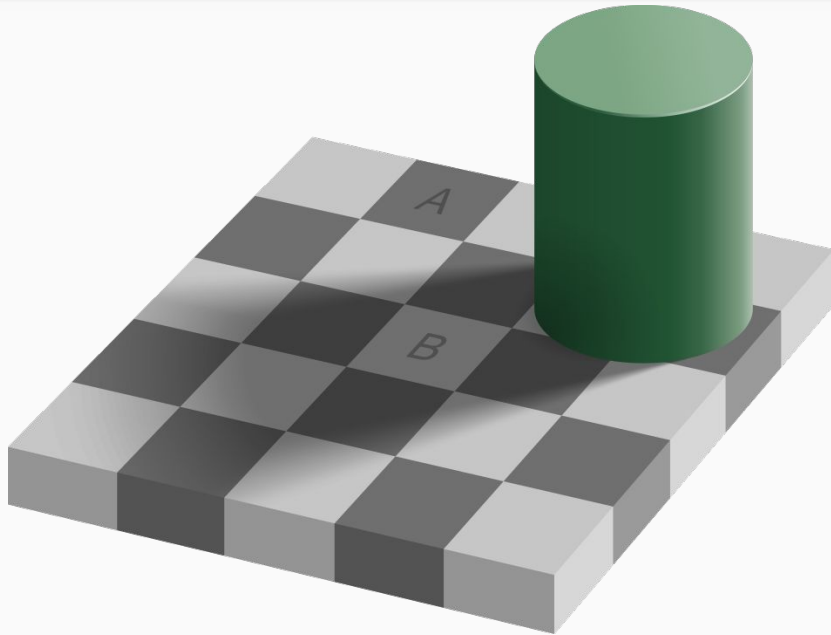
Optical Illusion Due to Perceived Brightness



They have the same intensity!

Human Visual System (HVS)

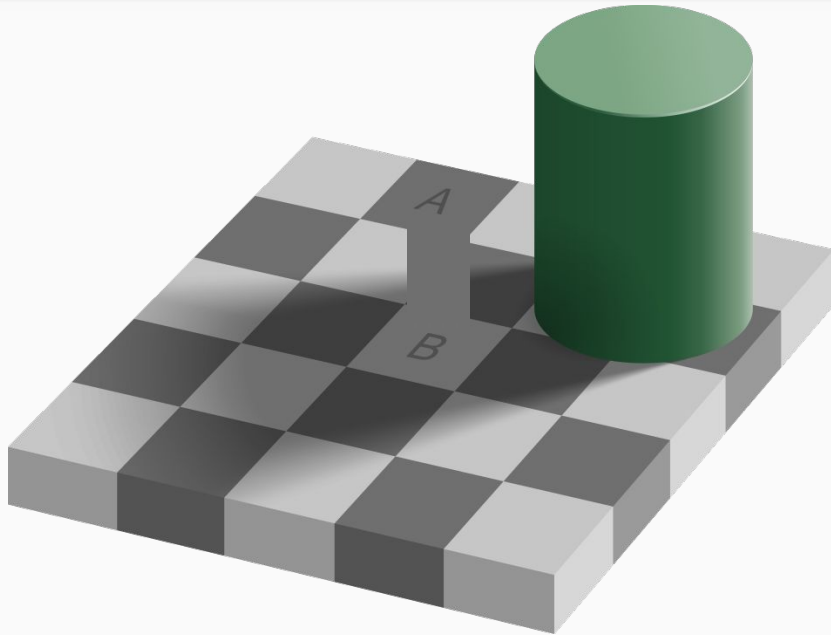
Optical Illusion Due to Experience



Which one has more intensity,
A or B?

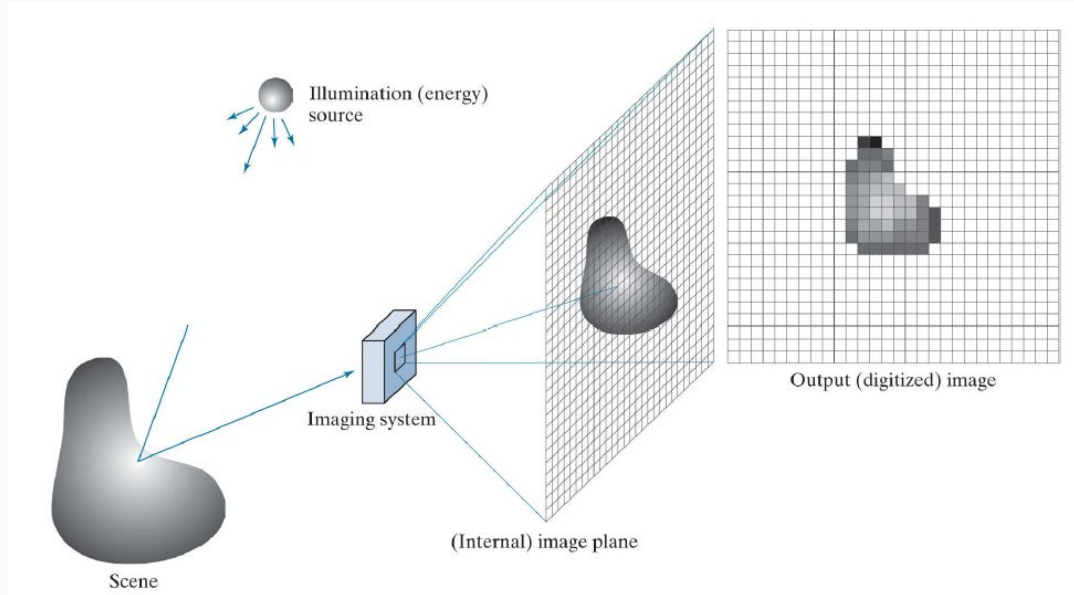
Human Visual System (HVS)

Optical Illusion Due to Experience



They have the exact same intensity!

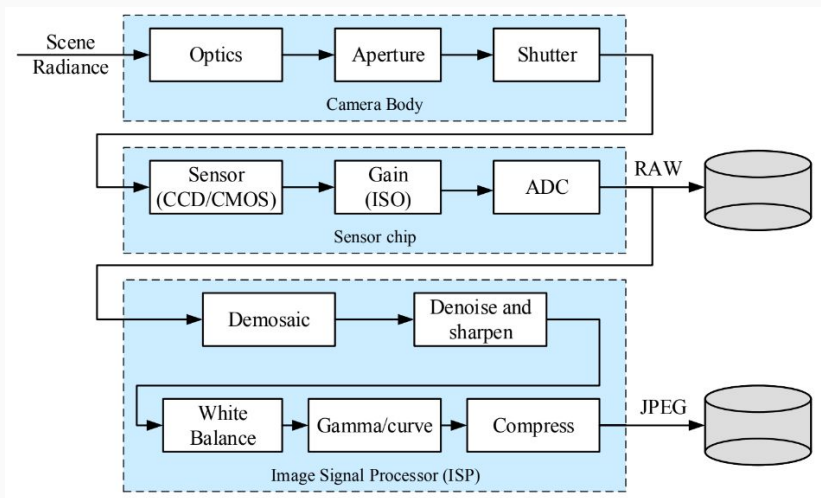
Digital Image Acquisition



Example pipeline of digital image acquisition using a camera with CCD sensor array

- Image Formation
- Sensing
- Digitization
- Representation

Image Sensing Pipeline



Example pipeline of digital image acquisition using a camera with CCD sensor array

- Image Formation
- Sensing
- Digitization
- Representation

Digital Image Acquisition

Image Formation Model

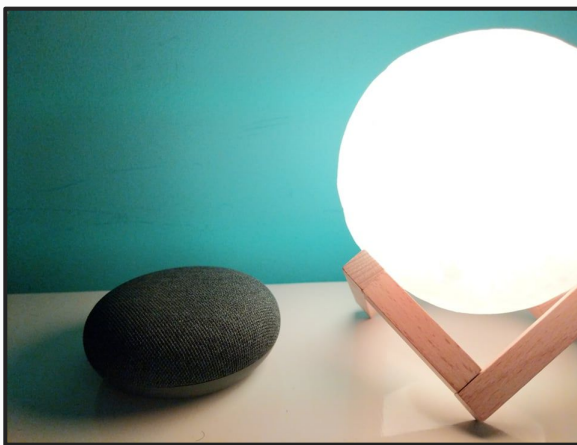
- The captured intensity $0 \leq f(x, y) < \infty$
- $f(x, y) = i(x, y) r(x, y)$
- Illumination, $0 \leq i(x, y) < \infty$
- Reflectance, $0 \leq r(x, y) \leq 1$
- $L_{\min} \leq f(x, y) \leq L_{\max}$. The ratio L_{\max} / L_{\min} is called the dynamic range.
- The interval $[L_{\min}, L_{\max}]$ is called gray scale. Typical indoor values, $L_{\min} \approx 10$, $L_{\max} \approx 1000$

Digital Image Acquisition

Image Formation Model



Clipped at L_{\min}



Clipped at L_{\max} (saturated)



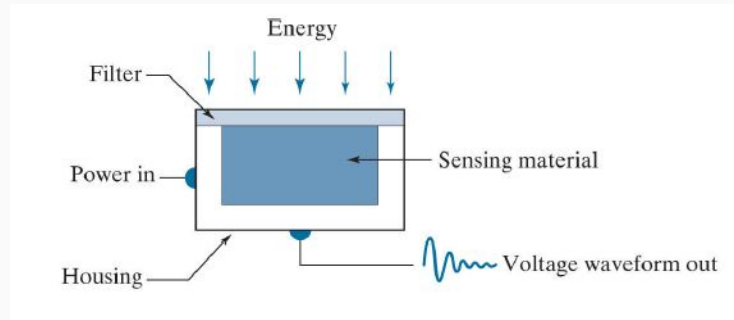
High Dynamic Range



Low Dynamic Range

Digital Image Acquisition

Image Sensing - Single Sensing Element

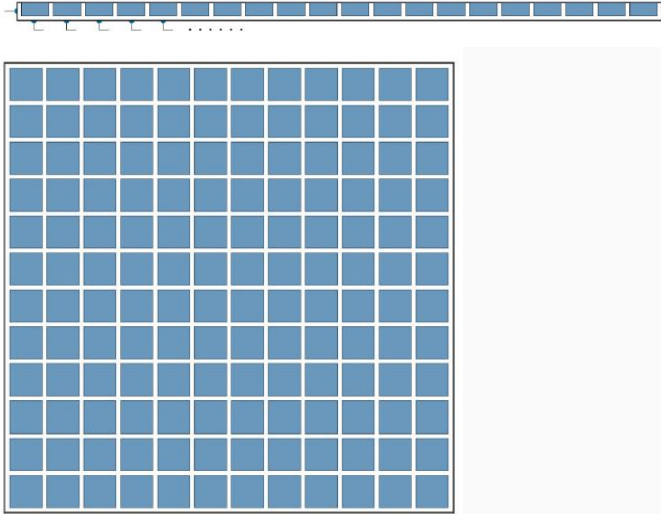


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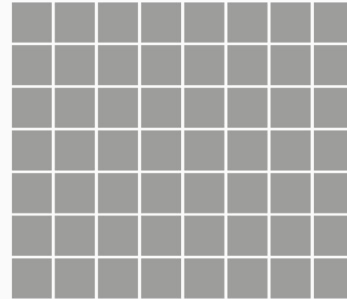
- A typical CCD camera has 4000×4000 sensors in array
- Electrical response to each sensor \propto Integral of light intensity on surface over time, and thus reducing noise
- Electrical signal is stored in digitized form as a digital image

Digital Image Acquisition

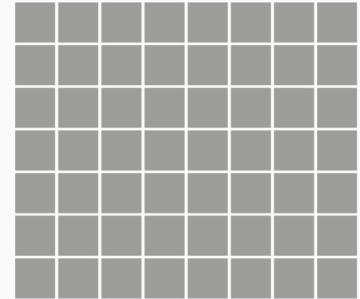
Image Sensing - Sensor Array



Rolling shutter



Global shutter



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<https://www.baumer.com/us/en/service-support/function-principle/function-principle-and-applications-of-rolling-shutter-cmos-cameras/a/CMOS-rolling-shutter-cameras>

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Line and grid Array of sensors

Exposure method - Rolling shutter vs global shutter

Digital Image Acquisition

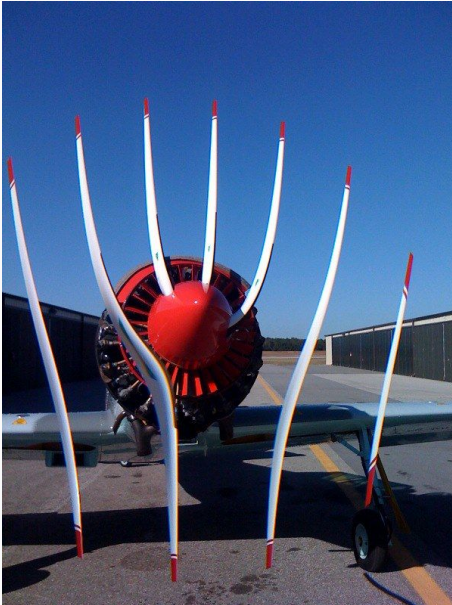
Image Sensing - Rolling Shutter Effect



- Number of blades? (Assuming the scan direction is from left to right)
- Direction of rotation?
- Speed of rotation? (RPM)

Digital Image Acquisition

Image Sensing - Rolling Shutter Effect



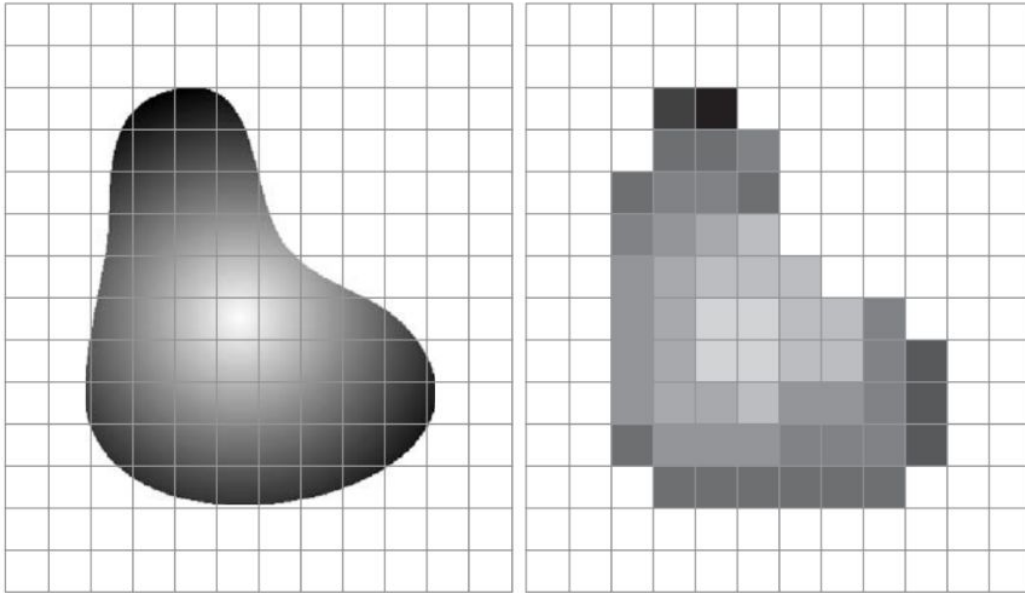
Ragsdale, S. 2009. "Airplane Prop + CMOS Rolling Shutter = WTF". Online Image. <<https://www.flickr.com/photos/sorenragdale/3192314056/in/photostream/>>



Cole, J. 2014. "Rolling Shutters". Online Image. <<https://jasmcole.com/2014/10/12/rolling-shutters/>>

Digital Image Acquisition

Digitization

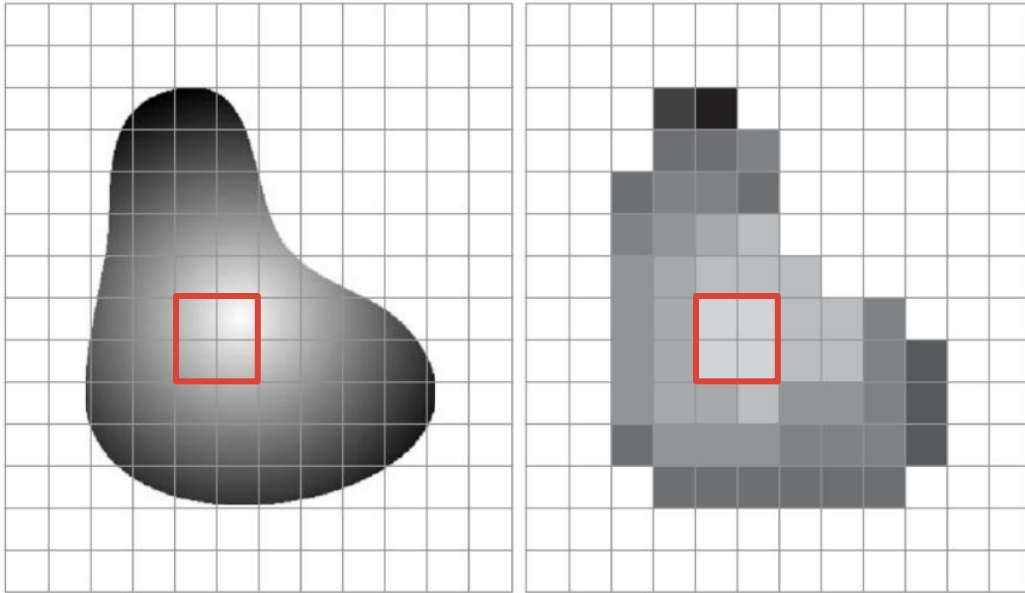


Two steps required:

- Sampling
- Quantization

Digital Image Acquisition

Digitization

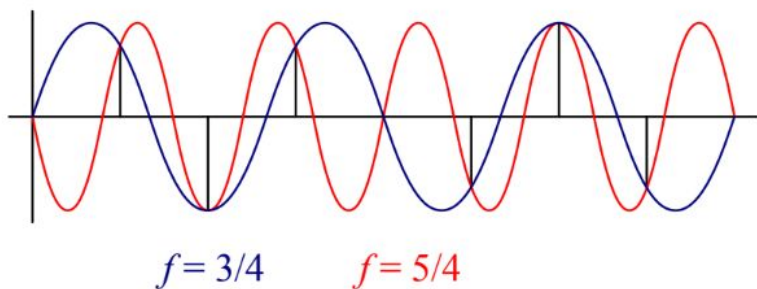


Two steps required:

- Sampling
- Quantization

Digital Image Acquisition

Sampling and Aliasing in 1D



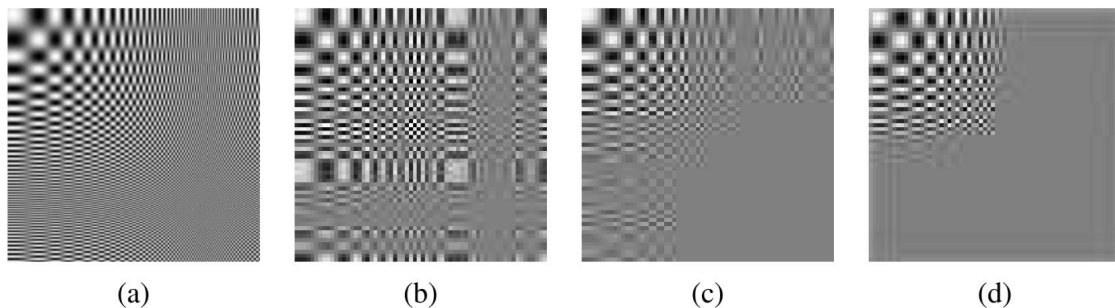
Aliasing of a one-dimensional signal:

- The blue sine wave at $f = 3/4$ and the red sine wave at $f = 5/4$ have the same digital samples, when sampled at $f = 2$
- These two signals are said to be “aliased”
- We are now no longer able to reconstruct the original signal

Shannon's Sampling Theorem: the minimum sampling rate required to reconstruct a signal from its instantaneous samples must be at least twice the highest frequency: $f_s \geq 2 \cdot f_{\max}$

Digital Image Acquisition

Sampling and Aliasing in 2D



Aliasing of a two-dimensional signal:

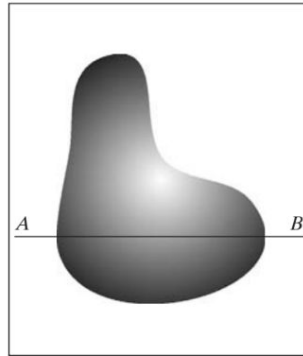
- (a) original full-resolution image
- (b) downsampled $4 \times$ with a 25% fill factor box filter
- (c) downsampled $4 \times$ with a 100% fill factor box filter
- (d) downsampled $4 \times$ with a high-quality 9-tap filter

Notice how the higher frequencies are aliased into visible frequencies with the lower quality filters, while the 9-tap filter completely removes these higher frequencies. (Anti-aliasing filter)

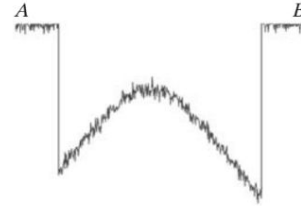
Digital Image Acquisition

Digitization

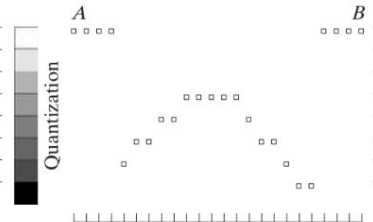
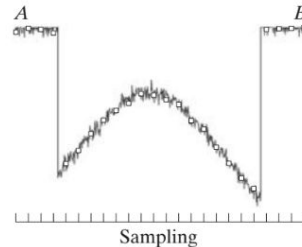
1) Image ➡



➡ 2) Scan line



3) Sampling ➡



➡ 4) Quantization

Digital Image Acquisition

Effect of Sampling

400 x 400



200 x 200



100 x 100



50 x 50



25 x 25



Digital Image Acquisition

Effect of Quantization



| | |
|------------|------------|
| 256 levels | 128 levels |
| 64 levels | 32 levels |

Digital Image Acquisition

Effect of Quantization

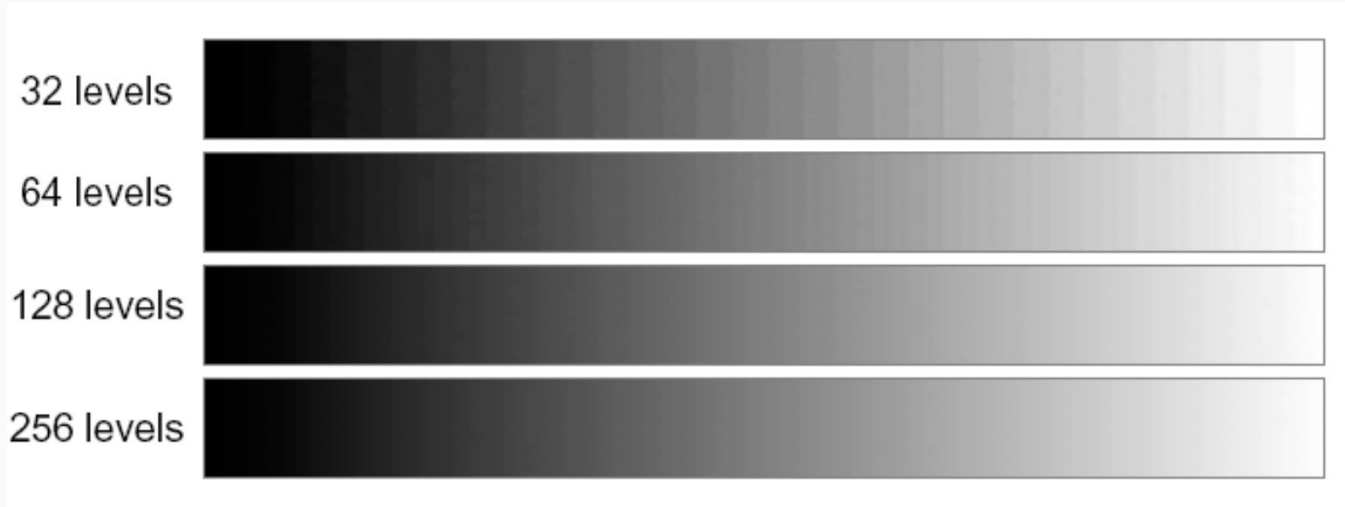


| | |
|-----------|-------------------|
| 16 levels | 8 levels |
| 4 levels | 2 levels (binary) |

False contouring

Digital Image Acquisition

Effect of Quantization



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Digital Image Acquisition

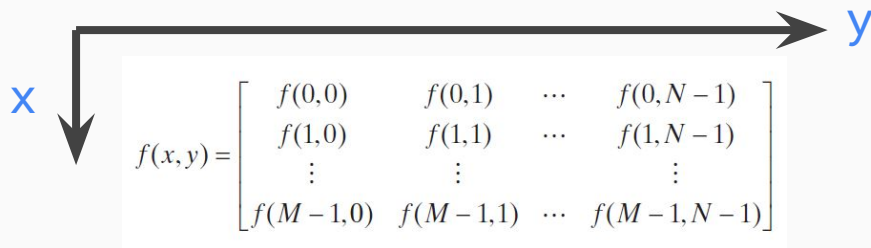
Representation

- $L_{\min} \leq f(x, y) \leq L_{\max}$
- $[L_{\min}, L_{\max}]$ often mapped to $[0, L-1]$ (for digital image)
- $[0, L-1] \Rightarrow L$ quantization levels
- 0 is called black level, $L-1$ is called white level
- We choose $L = 2^k$, where k is the number of bits required
- 256 levels = 2^8 , hence $k = 8$, called 8-bit image (most common)

Digital Image Acquisition

Representation

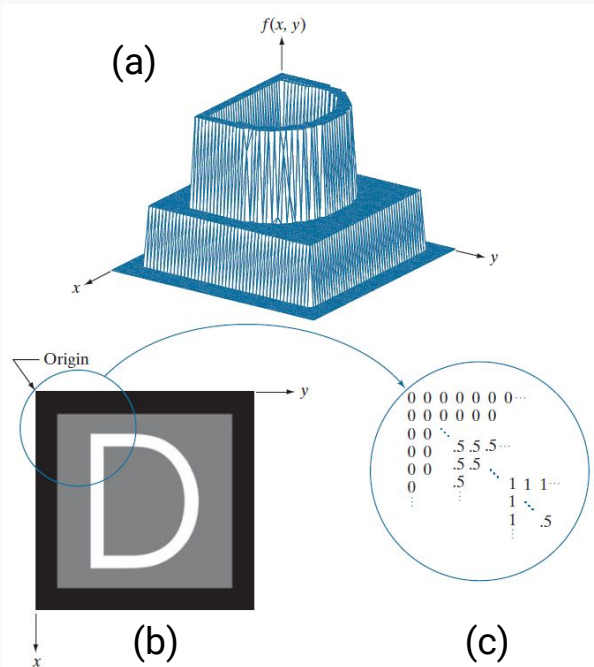
- For an image with M rows and N columns
 - $x = 0, 1, 2, \dots, (M - 1)$ where M is the **height** of the image (also called H)
 - $y = 0, 1, 2, \dots, (N - 1)$ where N is the **width** of the image (also called W)
- Numerical array form $[f(x, y)]$
- (i, j) th pixel value $[f(i, j)]$ is the image intensity at point (i, j)
- In Python - `img[x, y]`, `img.shape = (H, W)`


$$f(x,y) = \begin{bmatrix} f(0,0) & f(0,1) & \cdots & f(0,N-1) \\ f(1,0) & f(1,1) & \cdots & f(1,N-1) \\ \vdots & \vdots & & \vdots \\ f(M-1,0) & f(M-1,1) & \cdots & f(M-1,N-1) \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} a_{0,0} & a_{0,1} & \cdots & a_{0,N-1} \\ a_{1,0} & a_{1,1} & \cdots & a_{1,N-1} \\ \vdots & \vdots & & \vdots \\ a_{M-1,0} & a_{M-1,1} & \cdots & a_{M-1,N-1} \end{bmatrix}$$

Digital Image Acquisition

Representation



- (a) Image plotted as a surface
- (b) Image displayed as a visual intensity array
- (c) Image shown as a 2-D numerical array. (The numbers 0, .5, and 1 represent black, gray, and white, respectively.)

Digital Image Acquisition

Coordinate Convention

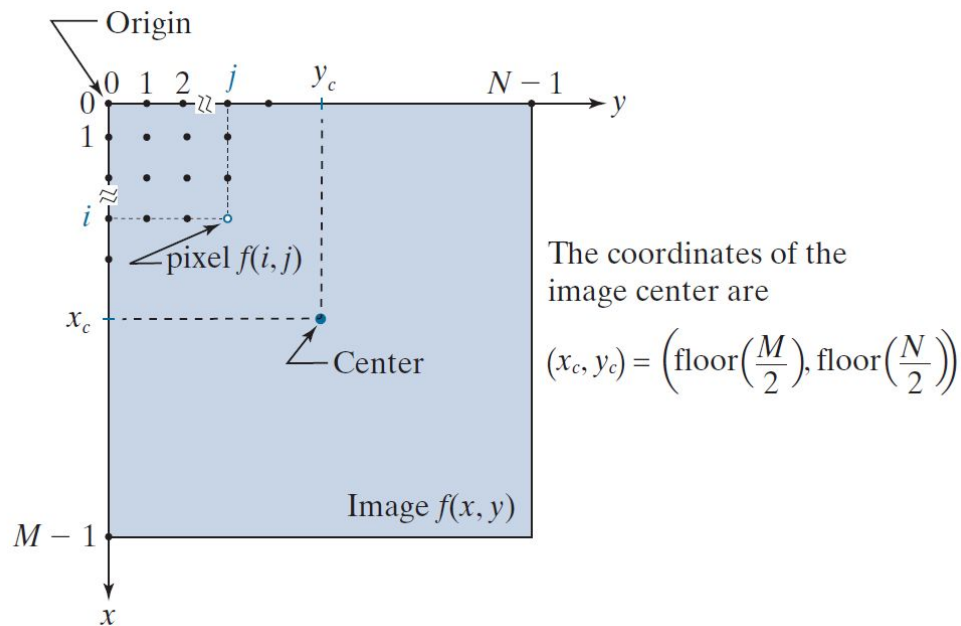


Image Resampling

Downsampling

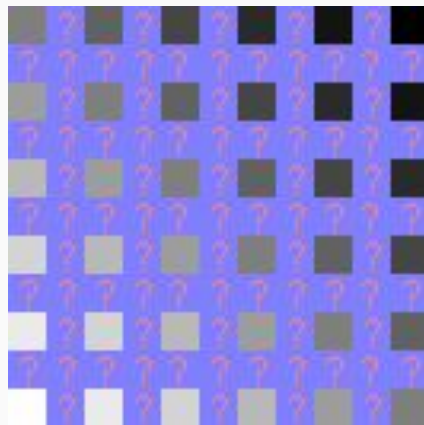
- k times downsample \rightarrow take the k'th sample
- In Python - `inp[y, x, n]`, `inp.shape = (H, W, N)`
- `out[y, x, n]`, `out.shape = (H//k, W//k, N)`
- `for y in range(0, H, k):`
 `for x in range(0, W, k):`
 `out[y//k, x//k, :] = inp[y, x, :]`
- Or, thanks to numpy slicing, `out = inp[::k, ::k,]`

Image Resampling

Upsampling and Interpolation



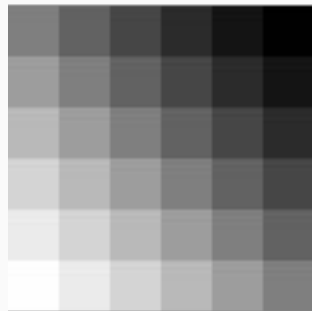
Original



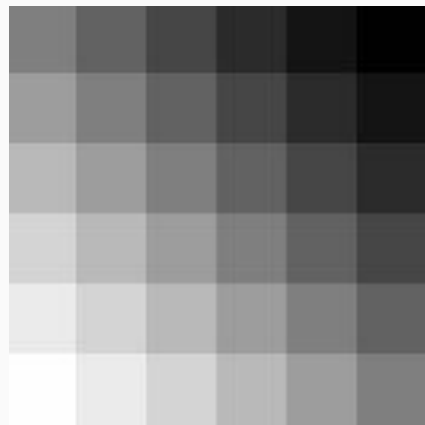
Upscaled - Before
interpolation

Image Resampling

Upsampling and Interpolation



Original



Upscaled - Nearest neighborhood

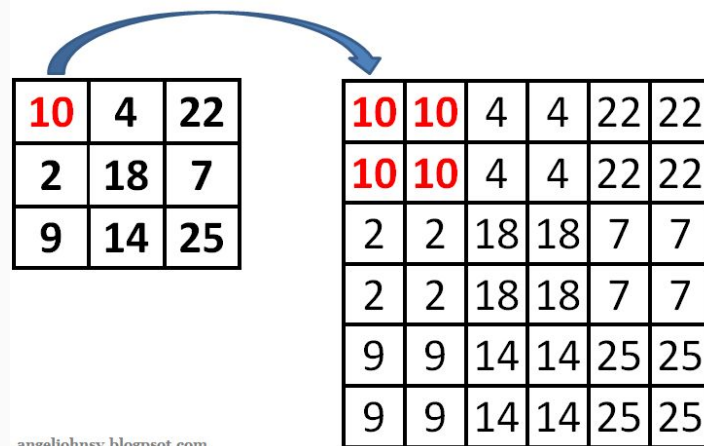
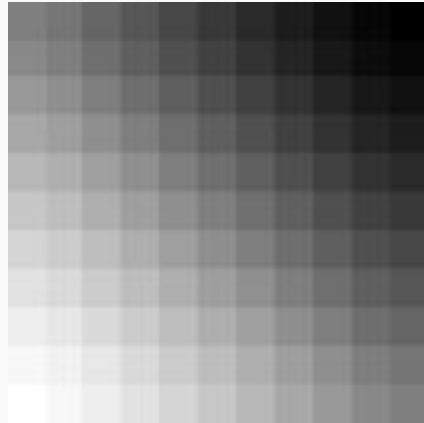


Image Resampling

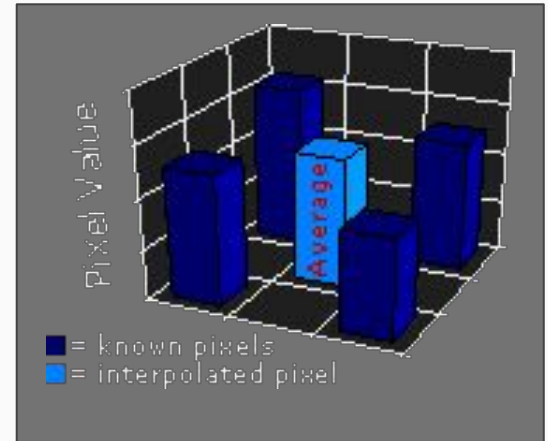
Upsampling and Interpolation



Original



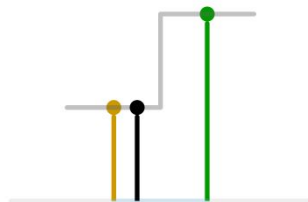
Upscaled - Bi-linear
interpolation



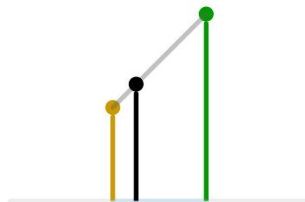
Bi-linear interpolation

Image Resampling

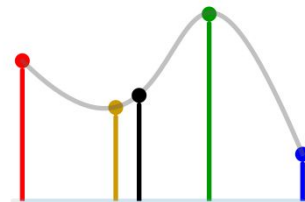
Upsampling and Interpolation



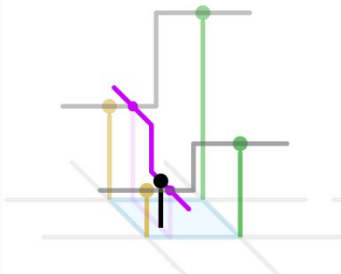
1D nearest-neighbour



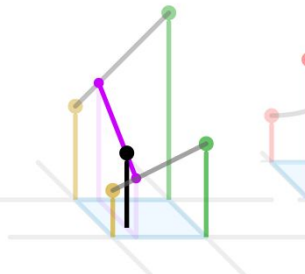
Linear



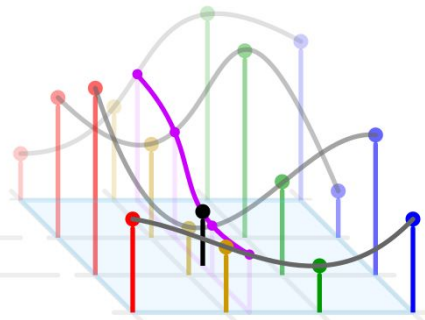
Cubic



2D nearest-neighbour



Bilinear



Bicubic

Questions?