

5- pixels and filters

image types:

- ↳ binary = pixels are either black (0) or white (1)
- ↳ grayscale = pixels are shade of gray black [0, 255] white
- ↳ color = have multiple channel colors RGB, LAB, HSV, 3d tensor [0, 255] in RGB
- image histograms = measure the frequency of brightness within the image
- ↳ how many times a particular pixel value appear in an image
- images as functions = $f: [a,b] \times [c,d] \rightarrow [0, 255] \rightarrow \text{grayscale}$
 $g: [x,y] = \begin{bmatrix} r(x,y) \\ g(x,y) \\ b(x,y) \end{bmatrix} \rightarrow \text{color}$

resolution = dots per inch (dpi)

filters/linear systems

- ↳ used to extract useful information or to adjust the visual properties of the image
- ↳ they are example of systems $f[m,n] \rightarrow g[m,n]$

example filters

- moving average = sets the value of a pixel to be the average of its neighbors
- ↳ smooth out the sharper edges $g[m,n] = \frac{1}{9} \sum_{i=-2}^2 \sum_{j=-2}^2 f[m-i, n-j]$
- ↳ blurred effect
- image segmentation = sets the value of a pixel either too high or too low value depending on the threshold
- ↳ $g[m,n] = \begin{cases} 255 & f[m,n] > 120 \\ 0 & \text{otherwise} \end{cases}$ divides pixel into binary classification of bright and dark regions

linear system = system that satisfies the property of superposition \rightarrow no powers of $x(t)$
 both $x(t)$, $t \cdot x(t) \rightarrow$ linear

properties of systems

- homogeneity = $S[\alpha f(x)] = \alpha S[f(x)]$
- superposition = $S[f(x) + f(y)] = S[f(x)] + S[f(y)]$
- stability = if system's output for all possible inputs is bounded
- invertibility = if distinct inputs $x(t)$ not t leads to distinct outputs $\frac{t \cdot x(t)}{x(t)} = y(t)$ \times
- causality = if the output depends on past or present values of an input
- ↳ if $h(t) = 0$ for all $n < n_0 \rightarrow$ causal $y(t) = x(t) \times y(t) = x(t)$ future \times
- time-invariance = if shifting input in time causes a identical shift in the output as well
- $y(t) = h(x(t)) \rightarrow y(t+t_0) = h(x(t+t_0))$ no t outside $x(t)$ $2 \times x(t) \times$
- memory = if output depend on past or future values of an input
- ↳ memoryless if output depends only the present value

ex: $y[n] = n \cdot x[n]$ linear = \checkmark no powers of $x[n]$

shift/ time invariant = \times n outside the $x[n]$

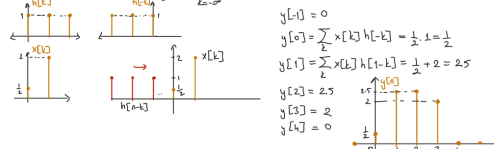
causal = memoryless \rightarrow depends only the present so causal \checkmark

stable = \times $x[n] \rightarrow \infty \rightarrow y[n] \rightarrow \infty$ not bounded

convolution = uses information from neighboring pixels to filter the target pixel

- $f[n,m] * g[n,m]$ = function being multiplied by a shifted impulse response
- $f[n] = \begin{cases} 1 & n=0 \\ 0 & \text{else} \end{cases} \rightarrow x[n] = \sum_k x[k] f[n-k] \rightarrow$ any signal
- $y[n] = \sum_k x[k] h[n-k] \rightarrow$ convolution of the signal x with impulse response h
- $y[n,m] = \sum_k \sum_j x[k,j] h[n-k, m-j]$ (2d)

1d convolution example = $y[n] = \sum_{k=-\infty}^{\infty} x[k] h[n-k] = x[n] * h[n]$



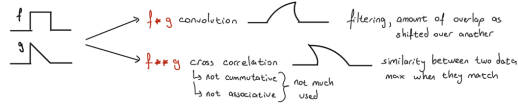
2d convolution example = $\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} * \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 16 & 25 & 34 \\ 25 & 36 & 45 \\ 34 & 45 & 54 \end{bmatrix}$

some filters: $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ no change, $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ shifted 1 pixel right, $\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ blur/average, $\begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ sharper

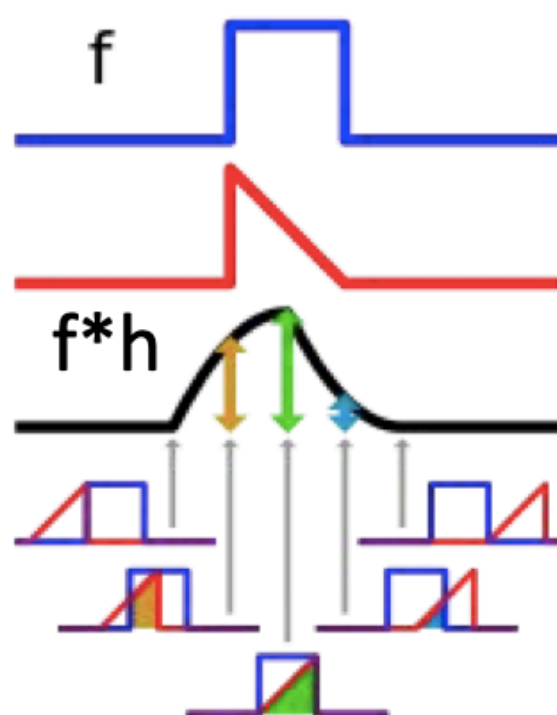
shape after convolution = $(M1 \times M2) * (N1 \times N2) = (M1+N1-1) \times (M2+N2-1)$
 image filter \rightarrow if no padding is added

cross correlation = used to find known features by using a kernel that contains target features

- $r[k,L] = \sum_n f[n+k, L] g[n, L] = f[k,L] * g[k,L]$



Convolution



Cross-correlation

