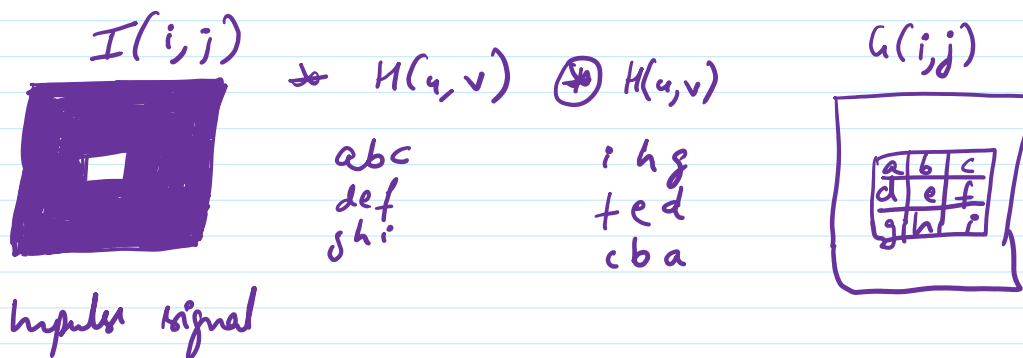


Convolution :-

$$G(i, j) = \sum_{u=-k}^k \sum_{v=-k}^k H(u, v) I(i-u, j-v)$$

$$G = H * I$$

Equivalent to flip the filter in both the directions (bottom to top, right to left) and apply cross correlation.



Linear Shift Invariant Operators :-

Both correlation and convolution are LSI.

① Linearity (or superposition principle)

$$I \circ (h_0 + h_1) = I \circ h_0 + I \circ h_1$$

② Shift-invariance

Shifting / translating a signal commutes with applying the operator.

related

when applying the operator.

$$g(i, j) = h(i+k, j+l) \leftarrow \begin{matrix} \text{Translated} \\ g \end{matrix}$$

a new filter



$$(f \circ g)(i, j) = (f \circ h)(i+k, j+l)$$

effect of operator is same everywhere as long as the image has same characteristics.

at at top / bottom.
same output

Convolution properties :-

① ^{Commutative}
 $a * b = b * a$

Conceptually no difference between filter and signal.

② Associativity

$$a * (b * c) = (a * b) * c$$

→ we often apply filters one after the other $((a * b_1) * b_2) * b_3$

$$\rightarrow \Rightarrow a * (b_1 * b_2 * b_3)$$

③ Distributive over addition :-

$$a * (b + c) = (a * b) + (a * c)$$

→ We can combine the response of a signal over two or more filters by combining the filters.

④ Scalar factor out

$$k a * b = a * kb = k(a * b)$$

⑤ Identity

Unit impulse $e = [---, 0, 0, 1, 0, 0, ---]$

$$a * e = a$$

⑥ Separability

Convolution operator requires k^2 operations per pixel, where k is the width (and height) of a convolution kernel.

How to reduce?

Let $k = 3 \Rightarrow 3 \times 3$ 9 operations

$N \times N$ image \Rightarrow cost $N^2 \times 9$

\Rightarrow 1D horizontal convolution followed by
 a 1D vertical convolution requires
 $2k$ operations. \Rightarrow separability
 $N^2 \times 6$

$$k = v h^T$$

v, h are 1D kernels and k is the
 2D kernel

Example 8-

$$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix} \quad v = h = \frac{1}{4} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

$$\frac{1}{8} \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix} \quad v = \frac{1}{4} \begin{bmatrix} -1 \\ -2 \\ -1 \end{bmatrix} \quad h = \frac{1}{2} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

How to tell if a given kernel k is
 separable?

look at SVD, and if only one
 singular value is non-zero, then
 it is separable

$$k = U \Sigma V^T$$