

Lecture 9:

Mathematics of JPEG

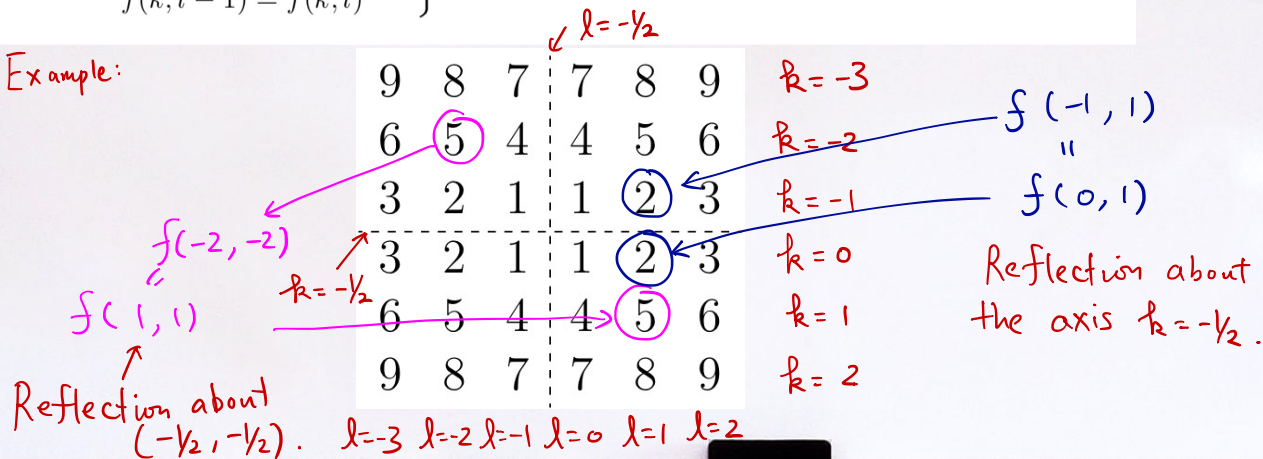
Consider a $N \times N$ image f . Extend f to a $2M \times 2N$ image \tilde{f} , whose indices are taken from $[-M, M - 1]$ and $[-N, N - 1]$.

Define $f(k, l)$ for $-M \leq k \leq M - 1$ and $-N \leq l \leq N - 1$ such that

$$f(-k - 1, -l - 1) = f(k, l) \quad \left. \vphantom{f(-k - 1, -l - 1)} \right\} \text{Reflection about } (-1/2, -1/2)$$

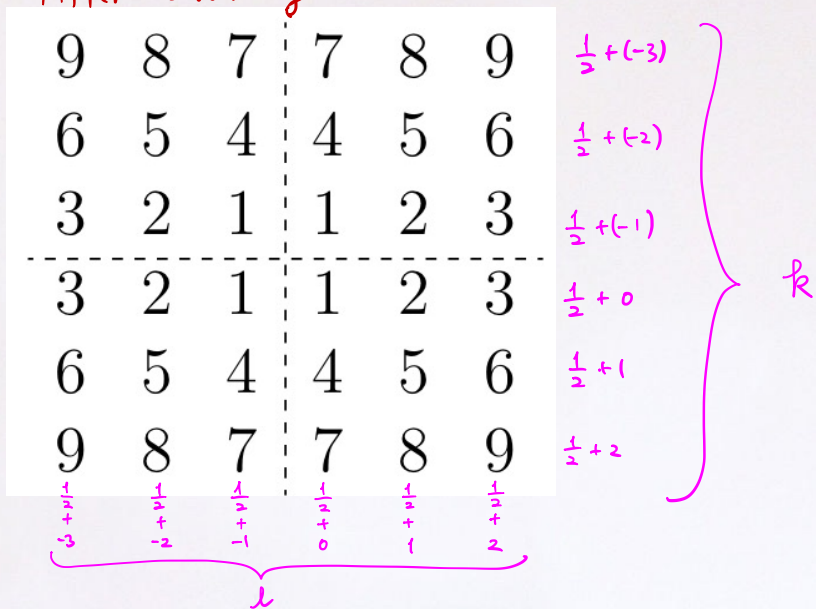
$$\left. \begin{aligned} f(-k - 1, l) &= f(k, l) \\ f(k, l - 1) &= f(k, l) \end{aligned} \right\} \text{Reflection about the axis } k = -1/2 \text{ and } l = -1/2$$

Example:



Make the extension as a reflection about $(0, 0)$, the axis $k=0$ and the axis $l=0$.
 Done by shifting the image by $(\frac{1}{2}, \frac{1}{2})$

After shifting



Now, we compute the DFT of (shifted) \tilde{f} :

$$\begin{aligned}
 F(m, n) &= \frac{1}{(2M)(2N)} \sum_{k=-M}^{M-1} \sum_{l=-N}^{N-1} f(k, l) e^{-j\frac{2\pi}{2M}m(k+\frac{1}{2})} e^{-j\frac{2\pi}{2N}n(l+\frac{1}{2})} \\
 &= \frac{1}{4MN} \sum_{k=-M}^{M-1} \sum_{l=-N}^{N-1} f(k, l) e^{-j(\frac{\pi}{M}m(k+\frac{1}{2})+\frac{\pi}{N}n(l+\frac{1}{2}))} \\
 &= \frac{1}{4MN} \left(\underbrace{\sum_{k=-M}^{-1} \sum_{l=-N}^{-1}}_{A_1} + \underbrace{\sum_{k=-M}^{-1} \sum_{l=0}^{N-1}}_{A_2} + \underbrace{\sum_{k=0}^{M-1} \sum_{l=-N}^{-1}}_{A_3} + \underbrace{\sum_{k=0}^{M-1} \sum_{l=0}^{N-1}}_{A_4} \right) \\
 &\quad f(k, l) e^{-j(\frac{\pi}{M}m(k+\frac{1}{2})+\frac{\pi}{N}n(l+\frac{1}{2}))}
 \end{aligned}$$

After some messy simplification, we can get:

$$A_1 + A_2 + A_3 + A_4 = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} f(k, l) \cos \left[\frac{m\pi}{M} \left(k + \frac{1}{2} \right) \right] \cos \left[\frac{n\pi}{N} \left(l + \frac{1}{2} \right) \right]$$

Definition: (Even symmetric discrete cosine transform [EDCT])

Let f be a $M \times N$ image, whose indices are taken as $0 \leq k \leq M - 1$ and $0 \leq l \leq N - 1$. The **even symmetric discrete cosine transform (EDCT)** of f is given by:

$$\hat{f}_{ec}(m, n) = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} f(k, l) \cos \left[\frac{m\pi}{M} \left(k + \frac{1}{2} \right) \right] \cos \left[\frac{n\pi}{N} \left(l + \frac{1}{2} \right) \right]$$

with $0 \leq m \leq M - 1, 0 \leq n \leq N - 1$

- Remark:
- Smart idea to get a decomposition consisting only of cosine function (by reflection and shifting!)
 - Can be formulated in matrix form
 - Again, it is a separable image transformation.

- The inverse of EDCT can be explicitly computed. More specifically, the **inverse EDCT** is defined as:

$$f(k, l) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} C(m)C(n) \hat{f}_{ec}(m, n) \cos \frac{\pi m(2k+1)}{2M} \cos \frac{\pi n(2l+1)}{2N} \quad (**)$$

where $C(0) = 1, C(m) = C(n) = 2$ for $m, n \neq 0$

Also involving cosine functions only!

- Formula (**) can be expressed as matrix multiplication:

$$f = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} \hat{f}_{ec}(m, n) \vec{T}_m \vec{T}_n^T$$

elementary images under EDCT!

where: $\vec{T}_m = \begin{pmatrix} T_m(0) \\ T_m(1) \\ \vdots \\ T_m(M-1) \end{pmatrix}, \vec{T}_n^T = \begin{pmatrix} T'_n(0) \\ T'_n(1) \\ \vdots \\ T'_n(N-1) \end{pmatrix}$ with $T_m(k) = C(m) \cos \frac{\pi m(2k+1)}{2M}$

and $T'_n(k) = C(n) \cos \frac{\pi n(2k+1)}{2N}$.

This is what JPEG does!!

Something similar can be developed:

Definition: (Odd symmetric discrete cosine transform [ODCT])

Let f be a $M \times N$ image, whose indices are taken as $0 \leq k \leq M - 1$ and $0 \leq l \leq N - 1$. The **odd symmetric discrete cosine transform (ODCT)** of f is given by:

$$\hat{f}_{oc}(m, n) = \frac{1}{(2M - 1)(2N - 1)} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} C(k)C(l)f(k, l) \cos \frac{2\pi mk}{2M - 1} \cos \frac{2\pi nl}{2N - 1}$$

where $C(0) = 1$ and $C(k) = C(l) = 2$ for $k, l \neq 0$, $0 \leq m \leq M - 1$, $0 \leq n \leq N - 1$.

The **inverse ODCT** is given by:

$$f(k, l) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} C(m)C(n)\hat{f}_{oc}(m, n) \cos \frac{2\pi mk}{2M - 1} \cos \frac{2\pi nl}{2N - 1}$$

where $C(0) = 1$, $C(m) = C(n) = 2$ if $m, n \neq 0$

Understanding convolution:

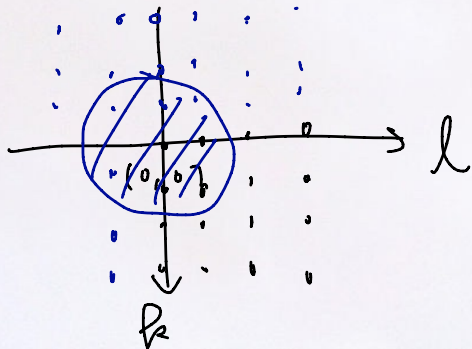
Recall: Discrete convolution:

$$V(n, m) = \sum_{n'=0}^{N-1} \sum_{m'=0}^{N-1} g(n-n', m-m') I(n', m')$$

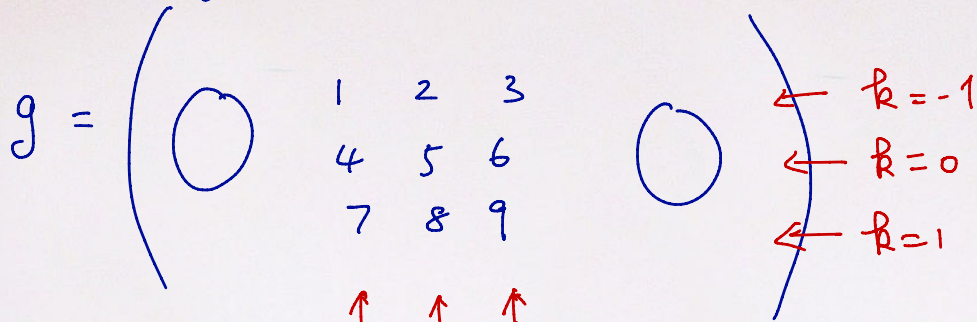
$g \times I(n, m)$

Linear combination of pixel values of I

In particular, if $g(k, l)$ is only non-zero around $(0, 0)$, then, $g \times I(n, m)$ is a linear combination of pixel value of I around (n, m) !!



Example: Suppose g looks like the following:



$$I * g(n, m) = \sum_{n'=0}^{N-1} \sum_{m'=0}^{N-1} g(n-n', m-m') I(n', m')$$

Linear combination of neighborhood pixel values

$$\begin{aligned} &= 1 \cdot I(n+1, m+1) + 2 \cdot I(n+1, m) + 3 \cdot I(n+1, m-1) \\ &+ 4 \cdot I(n, m+1) + 5 \cdot I(n, m) + 6 \cdot I(n, m-1) \\ &+ 7 \cdot I(n-1, m+1) + 8 \cdot I(n-1, m) + 9 \cdot I(n-1, m-1) \end{aligned}$$

Note:

(Spatial domain)

$I * g$

(Linear filtering:
Linear combination of
neighborhood pixel
values)

↓ DFT

(Frequency domain)

$MN \hat{I} \odot \hat{g}$
pixel-wise
multiplication

(Modifying the
Fourier coefficients
by multiplication)