

# Discrete cosine transform

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# Motivation

- Fourier transform
- sparse image representation, compression, JPEG
- integral transform

$$c(k) = \int f(t) g_k(t) dt$$

$\{g_k\}$  ... kernel, basis

$$f(t) = \int c(k) g_k^{-1}(t) dk$$

$\{g_k^{-1}\}$  ... inverse kernel

# 1D DCT

- basis of the complex space  $\mathbb{C}^N$  (signal length of  $N$ )

$$\left\{ g_k[n] = \lambda_k \sqrt{\frac{2}{N}} \cos \left[ \frac{k\pi}{N} \left( n + \frac{1}{2} \right) \right] \right\}_{0 \leq k < N}$$

$$\lambda_k = \begin{cases} 1/\sqrt{2} & : k = 0 \\ 1 & : k \neq 0 \end{cases}$$

- forward transform  $f \in \mathbb{C}^N$

$$c[k] = \sum_{n=0}^{N-1} f[n] g_k[n]$$

- inverse transform

$$f[n] = \sum_{k=0}^{N-1} c[k] g_k[n]$$

- basis of  $\mathbb{C}^{N \times N}$  space (image block of  $N \times N$  pixels)

$$\left\{ g_{k,j}[n, m] = \lambda_k \lambda_j \frac{2}{N} \cos \left[ \frac{k\pi}{N} \left( n + \frac{1}{2} \right) \right] \cos \left[ \frac{j\pi}{N} \left( m + \frac{1}{2} \right) \right] \right\}_{0 \leq k, j < N}$$

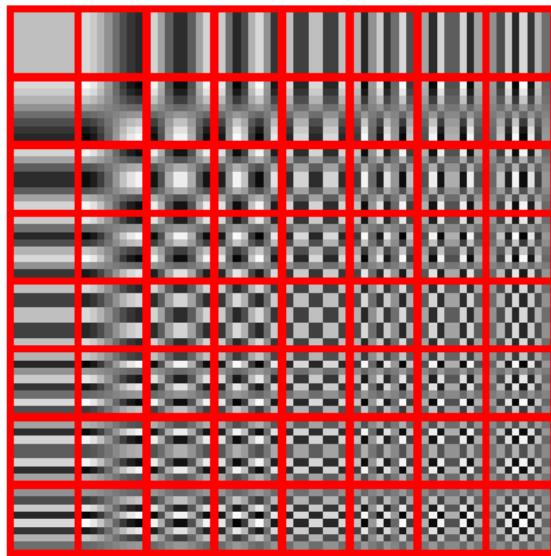
- forward transform

$$c[k, j] = \sum_{n=0}^{N-1} \sum_{m=0}^{N-1} f[n, m] g_{k,j}[n, m]$$

- inverse transform

$$f[n, m] = \sum_{k=0}^{N-1} \sum_{j=0}^{N-1} c[k, j] g_{k,j}[n, m]$$

# 2D DCT basis



## Separable 2D transform

- transform on  $\mathbb{C}^{N \times N}$  space has  $O(N^4)$  complexity
- identical separable transform has  $O(N^3)$  complexity
- kernel of the transform can be separated

$$g_{k,j}[n, m] = g_k[n] g_j[m]$$

(simplified notation)

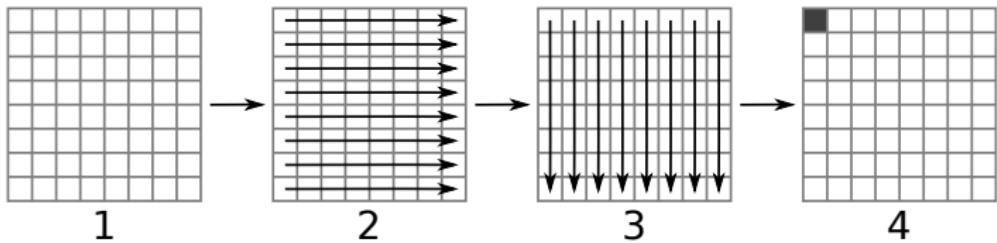
$$\{g_j[m]\} = \{g_k[n]\}^T$$

# Separable 2D transform

- computation in two steps – the rows, columns

$$\begin{aligned} c[k, j] &= \sum_{n=0}^{N-1} \sum_{m=0}^{N-1} f[n, m] g_{k,j}[n, m] \\ &= \sum_{n=0}^{N-1} \sum_{m=0}^{N-1} f[n, m] g_k[n] g_j[m] \\ &= \sum_{n=0}^{N-1} g_k[n] \sum_{m=0}^{N-1} f[n, m] g_j[m] = \sum_{n=0}^{N-1} g_k[n] c_j[n] \\ &= \sum_{m=0}^{N-1} g_j[m] \sum_{n=0}^{N-1} f[n, m] g_k[n] = \sum_{m=0}^{N-1} g_j[m] c_k[m] \end{aligned}$$

# Separable 2D transform



# Coefficient quantization

coarser quantization  $\Rightarrow$  larger loss of information  $\Rightarrow$  better compression

$$Q[k, j] = \begin{bmatrix} 16 & 11 & 10 & 16 & 24 & 40 & 51 & 61 \\ 12 & 12 & 14 & 19 & 26 & 58 & 60 & 55 \\ 14 & 13 & 16 & 24 & 40 & 57 & 69 & 56 \\ 14 & 17 & 22 & 29 & 51 & 87 & 80 & 62 \\ 18 & 22 & 37 & 56 & 68 & 109 & 103 & 77 \\ 24 & 35 & 55 & 64 & 81 & 104 & 113 & 92 \\ 49 & 64 & 78 & 87 & 103 & 121 & 120 & 101 \\ 72 & 92 & 95 & 98 & 112 & 100 & 103 & 99 \end{bmatrix}$$

$$b[k, j] = \text{round}\left(\frac{c[k, j]}{Q[k, j]}\right) \quad \tilde{c}[k, j] = b[k, j] Q[k, j]$$

- documentation – <http://opencv.willowgarage.com/>

```
Mat img = Mat_<double>(
imread("Lenna.png", CV_LOAD_IMAGE_GRAYSCALE) );

img.at<double>(y, x)

Mat dst = src.t();

Rect rect (x, y, N, N);
Mat block(img, rect);

imshow("image", Mat_<uchar>(img) );
```

# Tasks

- ① play with two DCT implementation
- ② play with coefficient quantization