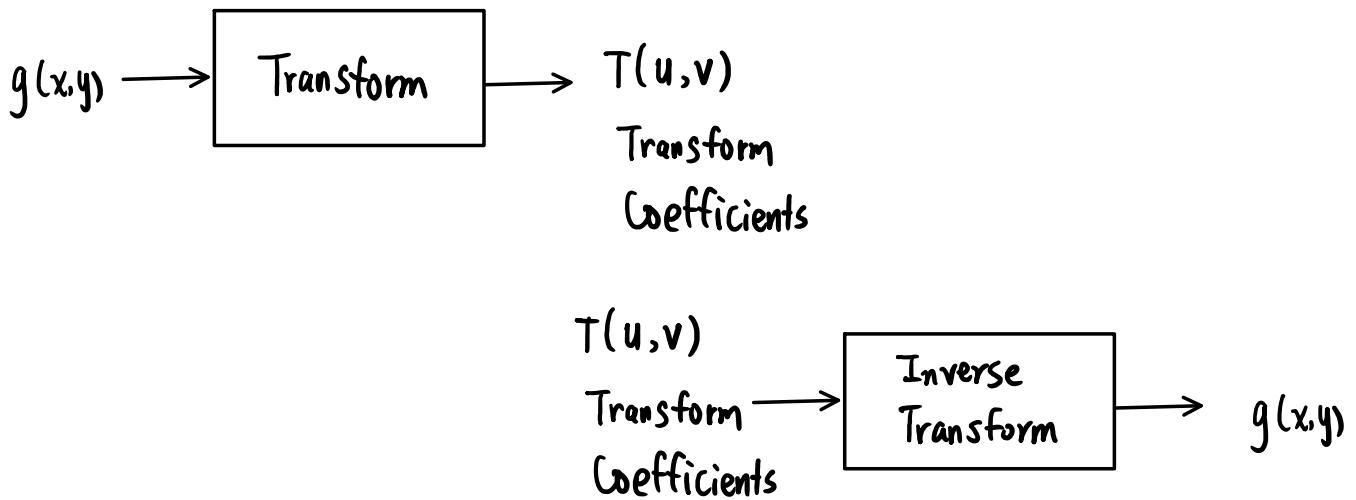
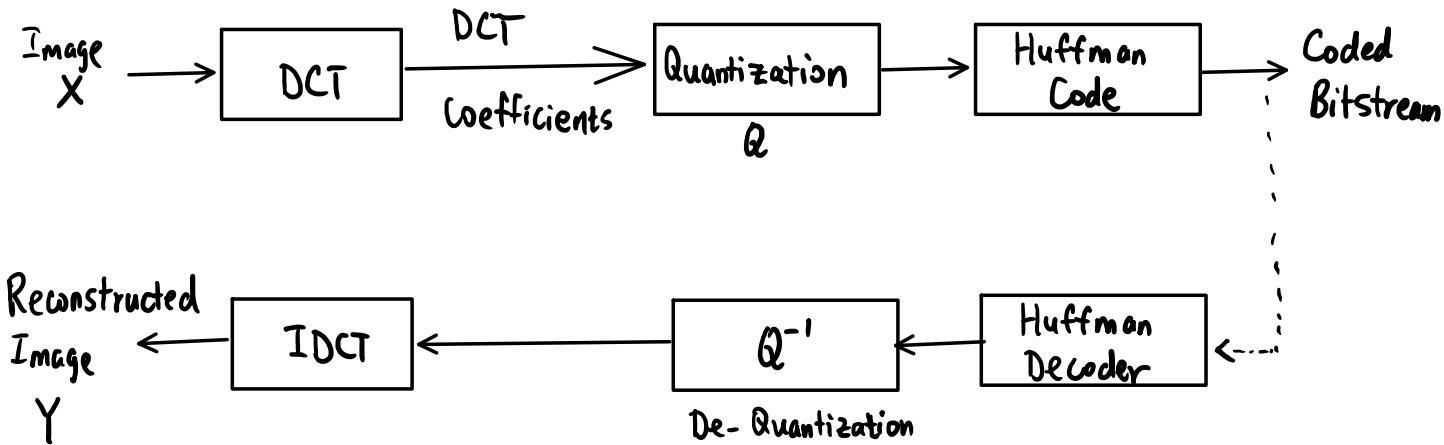


## Lecture 24

### Transform Coding



### JPEG Lossy Image Compression Using DCT



Block Transform:  
Input:  $n \times n$  block  
Output:  $n \times n$  block

$$\left\{ \begin{array}{l} T(u,v) = \sum_{x=0}^{n-1} \sum_{y=0}^{n-1} g(x,y) r(x,y, u, v) \\ g(x,y) = \sum_{u=0}^{n-1} \sum_{v=0}^{n-1} T(u,v) s(x,y, u, v) \end{array} \right.$$

```

x =
16 2 3 13
5 11 10 8
9 7 6 12
4 14 15 1
>> y = dct2(x)
y =
34.0000 0 0 0
0 0 13.5140 0
0 3.3785 0 2.9302
0 0 11.7206 0

```

### ✓ Frobenius Norm

The Frobenius norm of an  $m$ -by- $n$  matrix  $X$  (with  $m, n \geq 2$ ) is defined by

$$\|X\|_F = \sqrt{\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2} = \sqrt{\text{trace}(X^\dagger X)} .$$

```
>> (norm(x,'fro'))^2
```

```
ans =
1496
```

```
>> (norm(y,'fro'))^2
```

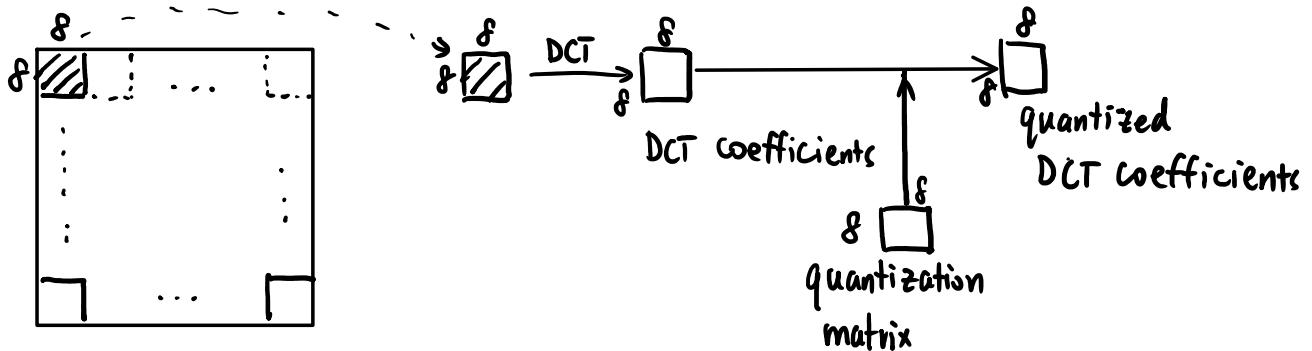
```
ans =
1.4960e+03 (energy is preserved)
```

```

>> x_bar = idct2(y)
x_bar =
16.0000 2.0000 3.0000 13.0000
5.0000 11.0000 10.0000 8.0000
9.0000 7.0000 6.0000 12.0000
4.0000 14.0000 15.0000 1.0000

```

Block-based lossy compression using DCT and Quantization



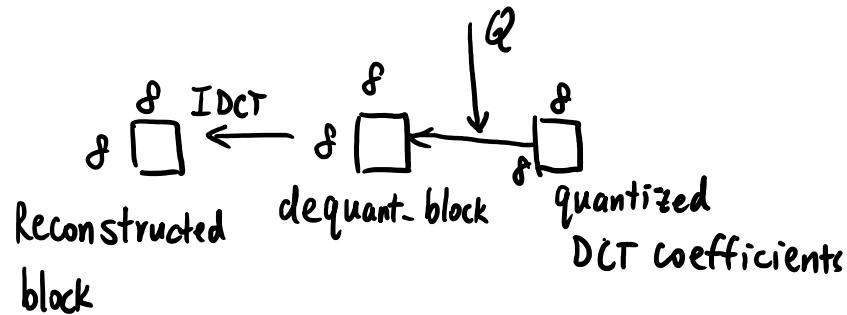
$$\begin{bmatrix} 16 & 11 & 10 & 16 & 24 & 40 & 51 & 61 \\ 12 & 12 & 14 & 19 & 26 & 58 & 60 & 55 \\ 14 & 13 & 16 & 24 & 40 & 57 & 69 & 56 \\ 14 & 17 & 22 & 29 & 51 & 87 & 80 & 62 \\ 18 & 22 & 37 & 56 & 68 & 109 & 103 & 77 \\ 24 & 35 & 55 & 64 & 81 & 104 & 113 & 92 \\ 49 & 64 & 78 & 87 & 103 & 121 & 120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

[https://en.wikipedia.org/wiki/Quantization\\_\(image\\_processing\)](https://en.wikipedia.org/wiki/Quantization_(image_processing))

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

quant\_coeff (i, j) = round (dct\_coeff (i, j) / Quant\_Matrix (i,j));

dequant\_block(i, j) = quant\_coeff (i, j) x Quant\_Matrix (i, j);



```
>> I = imread('coins.png');
>> imshow(I)
>> J = dct2(I);
>> whos J
  Name      Size            Bytes  Class    Attributes
  J        246x300        590400  double

>> whos I
```

Name	Size	Bytes	Class	Attributes
I	246x300	73800	uint8	

```
>> max(J(:))  
ans =  
( 2.7975e+04 )
```

```
>> min(J(:))  
ans =  
-4.2375e+03
```

```
>> sum(I(:))/(sqrt(246*300))  
ans =  
( 2.7975e+04 )
```

*Using DCT 2  
definition*

```
>> J(1:1)  
ans =  
( 2.7975e+04 )
```

*'DC' component*

```
>> (max(J(:)))^2/(norm(double(I),'fro')^2)  
ans =  
0.7725
```

77.25% of total energy contained  
by the 'DC' DCT Coefficient.

## ✓ Discrete Cosine Transform

The discrete cosine transform (DCT) is closely related to the discrete Fourier transform. It is a separable linear transformation; that is, the two-dimensional transform is equivalent to a one-dimensional DCT performed along a single dimension followed by a one-dimensional DCT in the other dimension. The definition of the two-dimensional DCT for an input image A and output image B is

$$B_{pq} = \alpha_p \alpha_q \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} A_{mn} \cos \frac{\pi(2m+1)p}{2M} \cos \frac{\pi(2n+1)q}{2N}, \quad 0 \leq p \leq M-1, \quad 0 \leq q \leq N-1$$

where

If  $p=0, q=0$ , then

$$\alpha_p = \begin{cases} \frac{1}{\sqrt{M}}, & p=0 \\ \sqrt{\frac{2}{M}}, & 1 \leq p \leq M-1 \end{cases}$$

$$B_{00} = \alpha_0 \alpha_0 \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} A_{mn}$$

where

$$\alpha_0 = \frac{1}{\sqrt{M}}$$

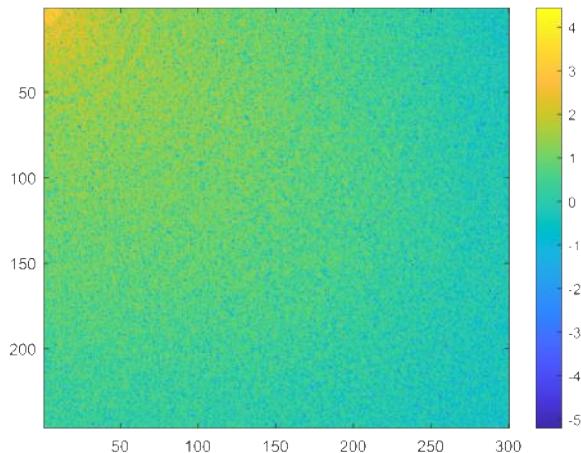
and

$$\alpha_q = \begin{cases} \frac{1}{\sqrt{N}}, & q=0 \\ \sqrt{\frac{2}{N}}, & 1 \leq q \leq N-1 \end{cases}$$

$$\alpha_0 = \frac{1}{\sqrt{N}}$$

$M$  and  $N$  are the row and column size of A, respectively.

```
>> figure; imagesc(log10(abs(J)));
>> colorbar;
```



```
>> K = idct2(J);
>> whos K
Name      Size            Bytes  Class
Attributes
K        246x300          590400  double
```

```
>> isequal(uint8(K), I)
ans =
logical
1
```