

EE3414

Multimedia Communication Systems - I

DCT and Transform Coding

Yao Wang

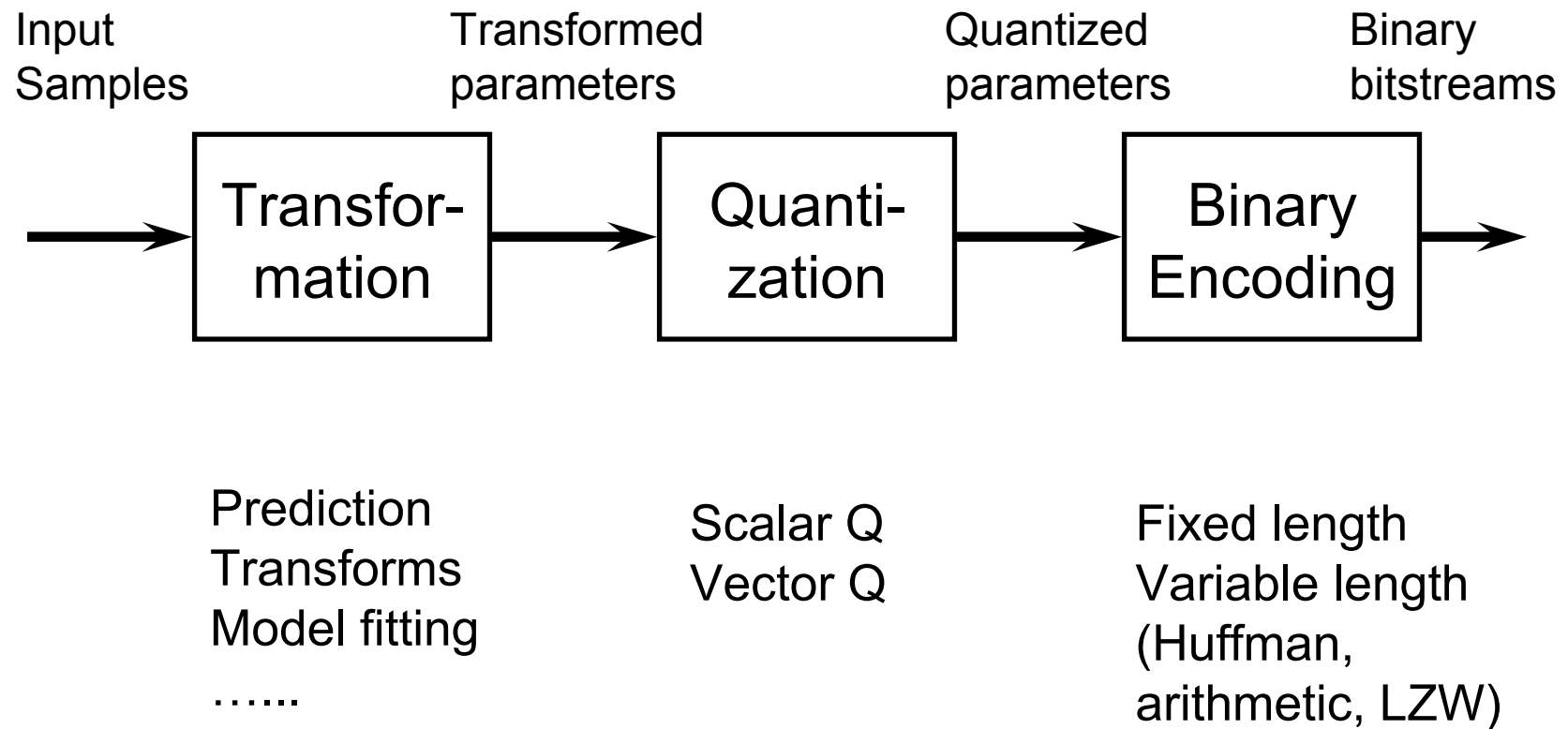
Polytechnic University, Brooklyn, NY11201

<http://eeweb.poly.edu>

Outline

- Transform coding
 - General principle
- DCT
 - Definition, basis images
 - Energy distribution
 - Approximation with different number of basis images
 - Quantization of DCT coefficients

A Typical Compression System

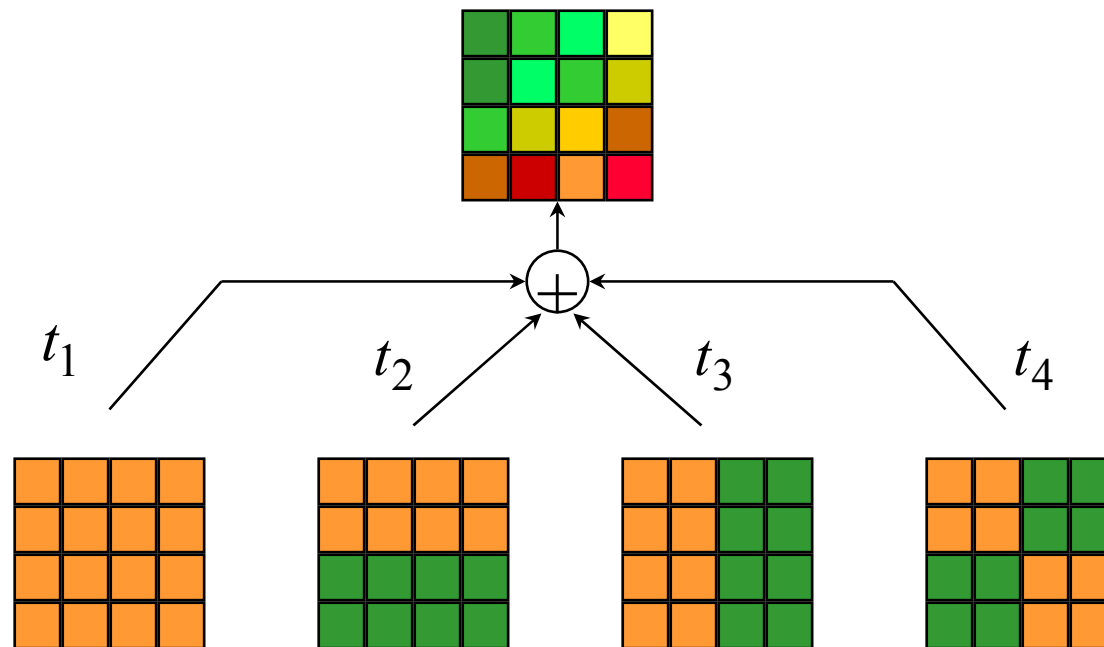


Motivation for “Transformation”

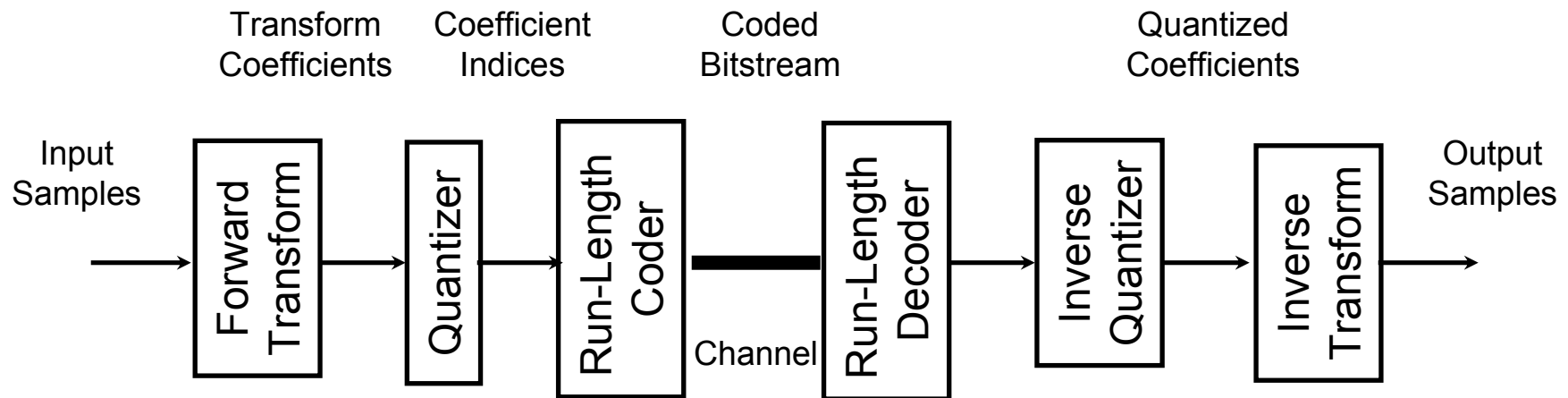
- Motivation for transformation:
 - To yield a more efficient representation of the original samples.
 - The transformed parameters should require fewer bits to code.
- Types of transformation used:
 - For speech coding: prediction
 - Code predictor and prediction error samples
 - For audio coding: subband decomposition
 - Code subband samples
 - For image coding: DCT and wavelet transforms
 - Code DCT/wavelet coefficients

Transform Coding

- Represent an image as the linear combination of some basis images and specify the linear coefficients.



A Typical Transform Coder



Transform Basis Design

- Optimality Criteria:
 - *Energy compaction*: a few basis images are sufficient to represent a typical image.
 - *Decorrelation*: coefficients for separate basis images are uncorrelated.
- **Karhunen Loeve Transform (KLT)** is the Optimal transform for a given covariance matrix of the underlying signal.
- **Discrete Cosine Transform (DCT)** is close to KLT for images that can be modeled by a first order Markov process (*i.e.*, a pixel only depends on its previous pixel).

1D Unitary Transform

Consider the N – point signal $s(n)$ as an N - dimensional vector

$$\mathbf{s} = \begin{bmatrix} s_0 \\ s_1 \\ \vdots \\ s_{N-1} \end{bmatrix}$$

The inverse transform says that \mathbf{s} can be represented as the sum of N basis vectors

$$\mathbf{s} = t_0 \mathbf{u}_0 + t_1 \mathbf{u}_1 + \dots + t_{N-1} \mathbf{u}_{N-1}$$

where \mathbf{u}_k corresponds to the k - th transform kernel :

$$\mathbf{u}_k = \begin{bmatrix} u_{k,0} \\ u_{k,1} \\ \vdots \\ u_{k,N-1} \end{bmatrix}$$

The forward transform says that the expansion coefficient t_k can be determined by the inner product of \mathbf{s} with \mathbf{u}_k :

$$t_k = (\mathbf{u}_k, \mathbf{s}) = \sum_{n=0}^{N-1} u_{k;n}^* s_n$$

1D Discrete Cosine Transform

- Can be considered “real” version of DFT
 - Basis vectors contain only co-sinusoidal patterns

DFT

$$u_{k,n} = \frac{1}{\sqrt{N}} \exp\left(j \frac{2\pi k}{N} n\right) = \frac{1}{\sqrt{N}} \left(\cos\left(\frac{2\pi k}{N} n\right) + j \sin\left(\frac{2\pi k}{N} n\right) \right)$$

$$\mathbf{u}_k = \frac{1}{\sqrt{N}} \begin{bmatrix} \cos\left(\frac{2\pi k}{N} 0\right) \\ \cos\left(\frac{2\pi k}{N} 1\right) \\ \dots \\ \cos\left(\frac{2\pi k}{N} (N-1)\right) \end{bmatrix} + j \frac{1}{\sqrt{N}} \begin{bmatrix} \sin\left(\frac{2\pi k}{N} 0\right) \\ \sin\left(\frac{2\pi k}{N} 1\right) \\ \dots \\ \sin\left(\frac{2\pi k}{N} (N-1)\right) \end{bmatrix}$$

DCT

$$u_{k,n} = \alpha(k) \cos\left(\frac{\pi k}{2N} (2n+1)\right)$$

$$\alpha(0) = \sqrt{\frac{1}{N}}, \alpha(k) = \sqrt{\frac{2}{N}}, k = 1, 2, \dots, N-1$$

$$\mathbf{u}_k = \alpha(k) \begin{bmatrix} \cos\left(\frac{\pi k}{2N} 1\right) \\ \cos\left(\frac{\pi k}{2N} 3\right) \\ \dots \\ \cos\left(\frac{\pi k}{2N} (2N+1)\right) \end{bmatrix}$$

Example: 4-point DCT

Using $u_{k,n} = \alpha(k) \cos\left(\frac{k\pi}{2*4}(2n+1)\right)$, $\alpha(0) = \sqrt{\frac{1}{4}} = \frac{1}{2}$, $\alpha(k) = \sqrt{\frac{2}{4}} = \sqrt{\frac{1}{2}}$, $k \neq 0$,

$$\text{1D DCT basis are: } \mathbf{u}_0 = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}; \mathbf{u}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} \cos\left(\frac{\pi}{8}\right) \\ \cos\left(\frac{3\pi}{8}\right) \\ \cos\left(\frac{5\pi}{8}\right) \\ \cos\left(\frac{7\pi}{8}\right) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 0.9239 \\ 0.3827 \\ -0.3827 \\ -0.9239 \end{bmatrix}; \mathbf{u}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} \cos\left(\frac{\pi}{4}\right) \\ \cos\left(\frac{3\pi}{4}\right) \\ \cos\left(\frac{5\pi}{4}\right) \\ \cos\left(\frac{7\pi}{4}\right) \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}; \mathbf{u}_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} \cos\left(\frac{3\pi}{8}\right) \\ \cos\left(\frac{9\pi}{8}\right) \\ \cos\left(\frac{15\pi}{8}\right) \\ \cos\left(\frac{21\pi}{8}\right) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 0.3827 \\ -0.9239 \\ 0.9239 \\ -0.3827 \end{bmatrix}$$

For $\mathbf{s} = \begin{bmatrix} 2 \\ 4 \\ 5 \\ 3 \end{bmatrix}$, determine the transform coefficients t_k . Also determine the reconstructed vector from all coefficients and two largest coefficients.

Go through on the board

2D Separable Transform

Consider the $M \times N$ – point image \mathbf{S} as a $M \times N$ - dimensional array (matrix)

$$\mathbf{S} = \begin{bmatrix} S_{0,0} & S_{0,1} & \dots & S_{0,N-1} \\ S_{1,0} & S_{1,1} & \dots & S_{1,N-1} \\ \dots & \dots & \dots & \dots \\ S_{M-1,0} & S_{M-1,1} & \dots & S_{M-1,N-1} \end{bmatrix}$$

The inverse transform says that \mathbf{s} can be represented as the sum of $M \times N$ basis images

$$\mathbf{S} = T_{0,0} \mathbf{U}_{0,0} + T_{0,1} \mathbf{U}_{0,1} + \dots + T_{M-1,N-1} \mathbf{U}_{M-1,N-1}$$

where $\mathbf{U}_{k,l}$ corresponds to the (k,l) - th transform kernel :

$$\mathbf{U}_{k,l} = \mathbf{u}_k (\mathbf{u}_l)^T = \begin{bmatrix} u_{k,0} \\ u_{k,1} \\ \dots \\ u_{k,N-1} \end{bmatrix} \begin{bmatrix} u_{l,0}^* & u_{l,1}^* & \dots & u_{l,N-1}^* \end{bmatrix} = \begin{bmatrix} u_{k,0} u_{l,0}^* & u_{k,0} u_{l,1}^* & \dots & u_{k,0} u_{l,N-1}^* \\ u_{k,1} u_{l,0}^* & u_{k,1} u_{l,1}^* & \dots & u_{k,1} u_{l,N-1}^* \\ \dots & \dots & \dots & \dots \\ u_{k,N-1} u_{l,0}^* & u_{k,N-1} u_{l,1}^* & \dots & u_{k,N-1} u_{l,N-1}^* \end{bmatrix}$$

The forward transform says that the expansion coefficient S_k can be determined by the inner product of \mathbf{S} and $\mathbf{U}_{k,l}$:

$$T_{k,l} = (\mathbf{U}_{k,l}, \mathbf{S}) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} U_{k,l;m,n}^* S_{m,n}$$

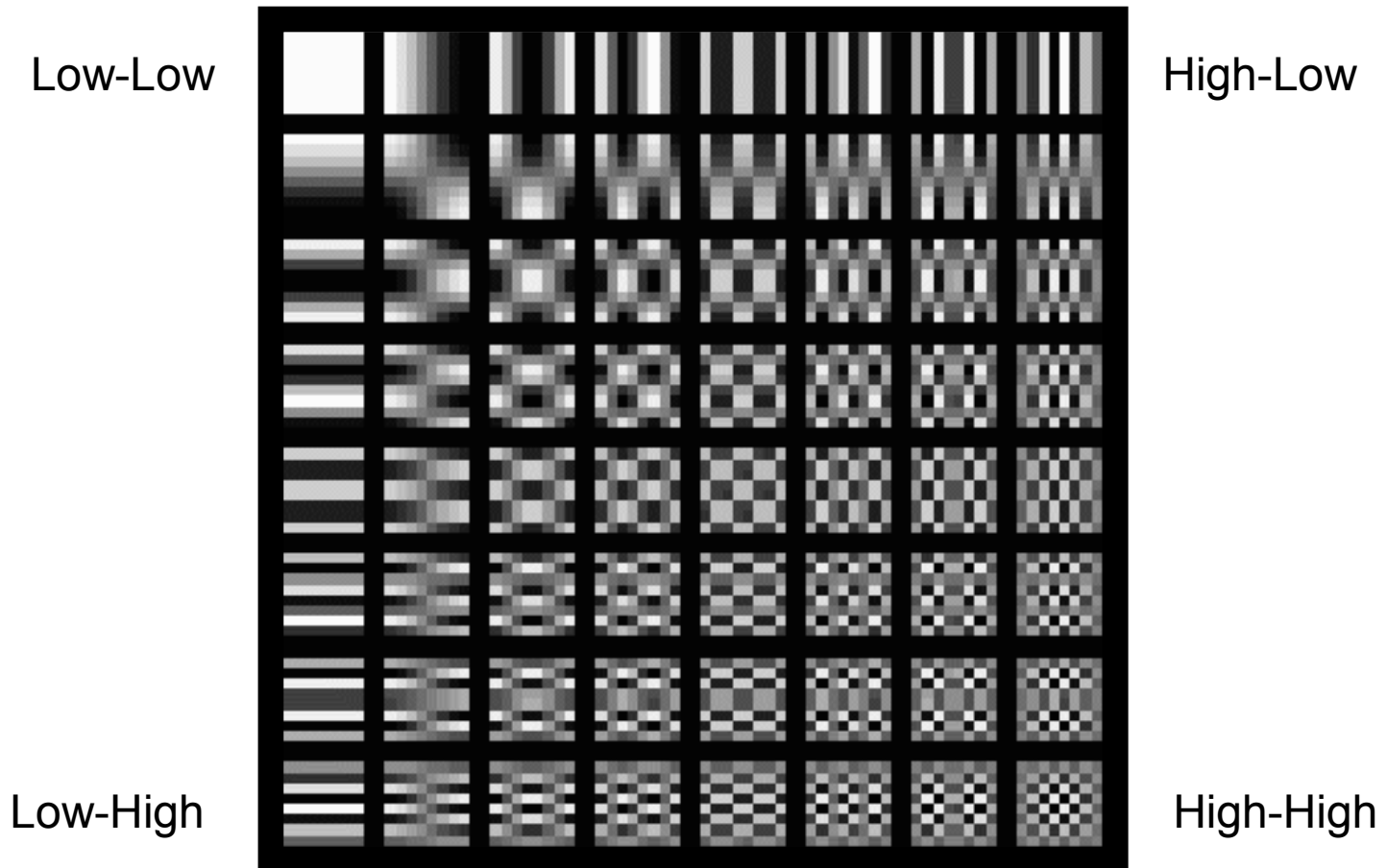
2D Discrete Cosine Transform

- Basis image = outer product of 1D DCT basis vector

$$\mathbf{u}_{k;N} = \alpha(k) \begin{bmatrix} \cos\left(\frac{\pi k}{2N} 1\right) \\ \cos\left(\frac{\pi k}{2N} 3\right) \\ \dots \\ \cos\left(\frac{\pi k}{2N} (2N+1)\right) \end{bmatrix}, \quad \alpha(0) = \sqrt{\frac{1}{N}}, \alpha(k) = \sqrt{\frac{2}{N}}, k = 1, 2, \dots, N-1$$

$$\begin{aligned} \mathbf{U}_{k,l;M,N} &= \mathbf{u}_{k;M} (\mathbf{u}_{l;N})^T \\ &= \alpha(k)\alpha(l) \begin{bmatrix} \cos\left(\frac{k\pi}{2M} 1\right) \cos\left(\frac{l\pi}{2N} 1\right) & \cos\left(\frac{k\pi}{2M} 1\right) \cos\left(\frac{l\pi}{2N} 3\right) & \dots & \cos\left(\frac{k\pi}{2M} 1\right) \cos\left(\frac{l\pi}{2N} (2N+1)\right) \\ \cos\left(\frac{k\pi}{2M} 3\right) \cos\left(\frac{l\pi}{2N} 1\right) & \cos\left(\frac{k\pi}{2M} 3\right) \cos\left(\frac{l\pi}{2N} 3\right) & \dots & \cos\left(\frac{k\pi}{2M} 3\right) \cos\left(\frac{l\pi}{2N} (2N+1)\right) \\ \dots & \dots & \dots & \dots \\ \cos\left(\frac{k\pi}{2M} (2M+1)\right) \cos\left(\frac{l\pi}{2N} 1\right) & \cos\left(\frac{k\pi}{2M} (2M+1)\right) \cos\left(\frac{l\pi}{2N} 3\right) & \dots & \cos\left(\frac{k\pi}{2M} (2M+1)\right) \cos\left(\frac{l\pi}{2N} (2N+1)\right) \end{bmatrix} \end{aligned}$$

Basis Images of 8x8 DCT



Example: 4x4 DCT

Using $u_{k,n} = \alpha(k) \cos\left(\frac{k\pi}{2*4}(2n+1)\right)$, $\alpha(0) = \sqrt{\frac{1}{4}} = \frac{1}{2}$, $\alpha(k) = \sqrt{\frac{2}{4}} = \sqrt{\frac{1}{2}}$, $k \neq 0$,

$$\text{1D DCT basis are: } \mathbf{u}_0 = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}; \mathbf{u}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} \cos\left(\frac{\pi}{8}\right) \\ \cos\left(\frac{3\pi}{8}\right) \\ \cos\left(\frac{5\pi}{8}\right) \\ \cos\left(\frac{7\pi}{8}\right) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 0.9239 \\ 0.3827 \\ -0.3827 \\ -0.9239 \end{bmatrix}; \mathbf{u}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} \cos\left(\frac{\pi}{4}\right) \\ \cos\left(\frac{3\pi}{4}\right) \\ \cos\left(\frac{5\pi}{4}\right) \\ \cos\left(\frac{7\pi}{4}\right) \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}; \mathbf{u}_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} \cos\left(\frac{3\pi}{8}\right) \\ \cos\left(\frac{9\pi}{8}\right) \\ \cos\left(\frac{15\pi}{8}\right) \\ \cos\left(\frac{21\pi}{8}\right) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 0.3827 \\ -0.9239 \\ 0.9239 \\ -0.3827 \end{bmatrix}$$

using $\mathbf{U}_{k,l} = \mathbf{u}_k(\mathbf{u}_l)^T$ yields:

$$\mathbf{U}_{0,0} = \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}, \mathbf{U}_{0,2} = \frac{1}{4} \begin{bmatrix} 1 & -1 & -1 & 1 \\ 1 & -1 & -1 & 1 \\ 1 & -1 & -1 & 1 \\ 1 & -1 & -1 & 1 \end{bmatrix}, \mathbf{U}_{2,0} = \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & -1 & -1 & -1 \\ -1 & -1 & -1 & -1 \\ 1 & 1 & 1 & 1 \end{bmatrix}, \mathbf{U}_{2,2} = \frac{1}{4} \begin{bmatrix} 1 & -1 & -1 & 1 \\ -1 & 1 & 1 & -1 \\ -1 & 1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}, \dots$$

For $\mathbf{S} = \begin{bmatrix} 1 & 2 & 2 & 0 \\ 0 & 1 & 3 & 1 \\ 0 & 1 & 2 & 1 \\ 1 & 2 & 2 & -1 \end{bmatrix}$, compute $T_{k,l}$

Using $T_{k,l} = (\mathbf{U}_{k,l}, \mathbf{S})$ yields, e.g,

$$T_{0,0} = (\mathbf{U}_{0,0}, \mathbf{S}) = \frac{1}{4} \left(\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 2 & 2 & 0 \\ 0 & 1 & 3 & 1 \\ 0 & 1 & 2 & 1 \\ 1 & 2 & 2 & -1 \end{bmatrix} \right) = \frac{1}{4} (1+2+2+0+0+1+3+1+0+1+2+1+1+2+2-1) = \frac{18}{4} = 4.5$$

$$T_{2,2} = (\mathbf{U}_{3,2}, \mathbf{S}) = \frac{1}{4} \left(\begin{bmatrix} 1 & -1 & -1 & 1 \\ -1 & 1 & 1 & -1 \\ -1 & 1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 2 & 2 & 0 \\ 0 & 1 & 3 & 1 \\ 0 & 1 & 2 & 1 \\ 1 & 2 & 2 & -1 \end{bmatrix} \right) = \frac{1}{4} (1-2-2+1+3-1+1+2-1+1-2-2-1) = -\frac{2}{4} = -0.5$$

Completing calculation using Matlab yields (using "dct2(S)") :

$$\mathbf{T} = \begin{bmatrix} 4.5 & -0.0793 & -3 & 1.1152 \\ 0.4619 & -0.5 & 0.1913 & 0 \\ 0 & 2.0391 & -0.5 & -0.3034 \\ -0.1913 & 0 & 0.4619 & -0.5 \end{bmatrix}$$

Reconstruction from top 2x2 coefficients only yields :

$$\mathbf{S} = \begin{bmatrix} 1.09 & 1.54 & 1.38 & 0.86 \\ 1.44 & 1.9 & 1.63 & 0.95 \\ 1.11 & 1.38 & 1.63 & 0.95 \\ 0.61 & 0.68 & 0.31 & 0.05 \end{bmatrix}$$

The reconstructed image varies slower than the original, because we cut off some high frequency coefficients

DCT on a Real Image Block

```
>>imblock = lena256(128:135,128:135)
```

```
imblock=
```

```
182 196 199 201 203 201 199 173
175 180 176 142 148 152 148 120
148 118 123 115 114 107 108 107
115 110 110 112 105 109 101 100
104 106 106 102 104 95 98 105
99 115 131 104 118 86 87 133
112 154 154 107 140 97 88 151
145 158 178 123 132 140 138 133
```

```
>>dctblock =dct2(imblock)
```

```
dctblock=1.0e+003*
```

```
1.0550 0.0517 0.0012 -0.0246 -0.0120 -0.0258 0.0120 0.0232
0.1136 0.0070 -0.0139 0.0432 -0.0061 0.0356 -0.0134 -0.0130
0.1956 0.0101 -0.0087 -0.0029 -0.0290 -0.0079 0.0009 0.0096
0.0359 -0.0243 -0.0156 -0.0208 0.0116 -0.0191 -0.0085 0.0005
0.0407 -0.0206 -0.0137 0.0171 -0.0143 0.0224 -0.0049 -0.0114
0.0072 -0.0136 -0.0076 -0.0119 0.0183 -0.0163 -0.0014 -0.0035
-0.0015 -0.0133 -0.0009 0.0013 0.0104 0.0161 0.0044 0.0011
-0.0068 -0.0028 0.0041 0.0011 0.0106 -0.0027 -0.0032 0.0016
```

Low frequency coefficients (top left corner) are much larger than the rest!

Reconstructed Block

Original block

```
182 196 199 201 203 201 199 173
175 180 176 142 148 152 148 120
148 118 123 115 114 107 108 107
115 110 110 112 105 109 101 100
104 106 106 102 104 95 98 105
99 115 131 104 118 86 87 133
112 154 154 107 140 97 88 151
145 158 178 123 132 140 138 133
```

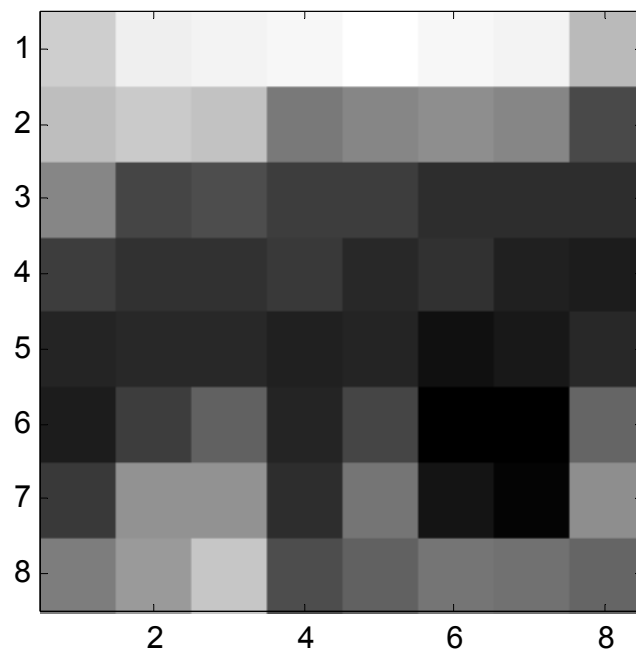
Reconstructed using top 4x4 coefficients

```
190 192 195 197 196 189 179 172
175 169 163 163 164 160 150 141
147 136 124 120 122 121 114 106
115 110 103 98 96 94 93 92
96 104 109 104 93 89 96 104
102 117 130 123 106 98 109 124
126 139 148 138 119 111 121 136
148 155 156 144 125 118 126 138
```

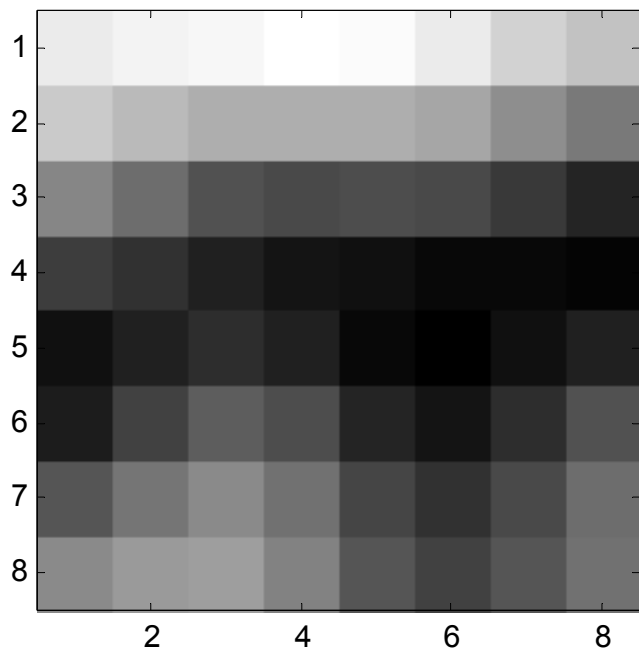
Reconstructed using top 2x2 coefficients only

```
162 161 158 154 149 146 143 141
159 157 154 151 147 143 140 138
153 151 149 145 141 137 135 133
145 144 141 138 134 131 128 126
137 135 133 130 126 123 121 119
129 128 125 122 119 116 114 113
123 122 119 117 114 111 109 108
119 118 116 114 111 108 106 105
```

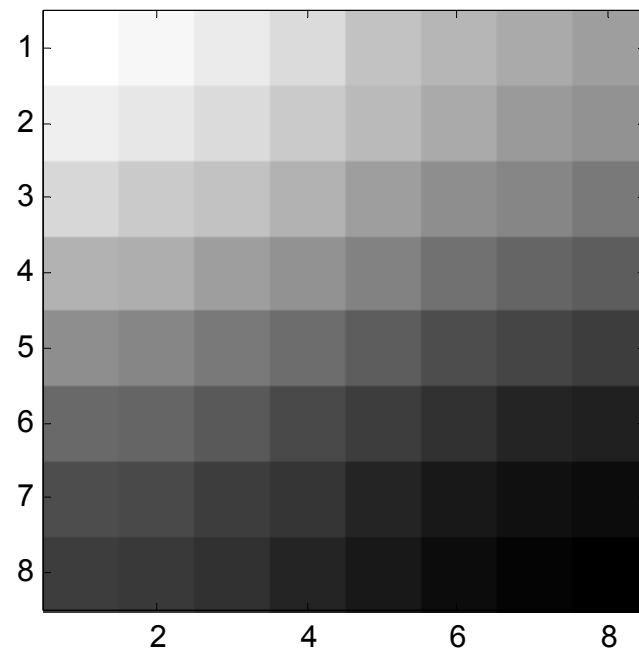
original



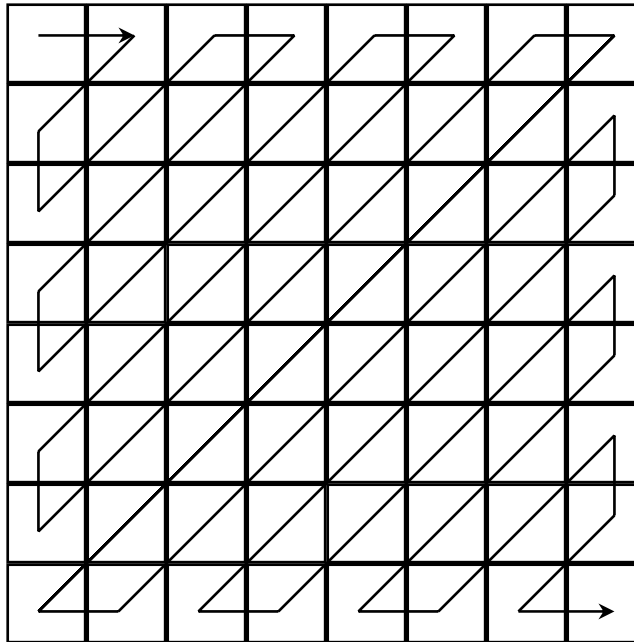
From 4x4 coef



from 2x2 coef



Zig-Zag Ordering of DCT Coefficients



Zig-Zag ordering: converting a 2D matrix into a 1D array, so that the frequency (horizontal+vertical) increases in this order, and the coefficient variance (average of magnitude square) decreases in this order.

DCT Coefficient Variance Distribution

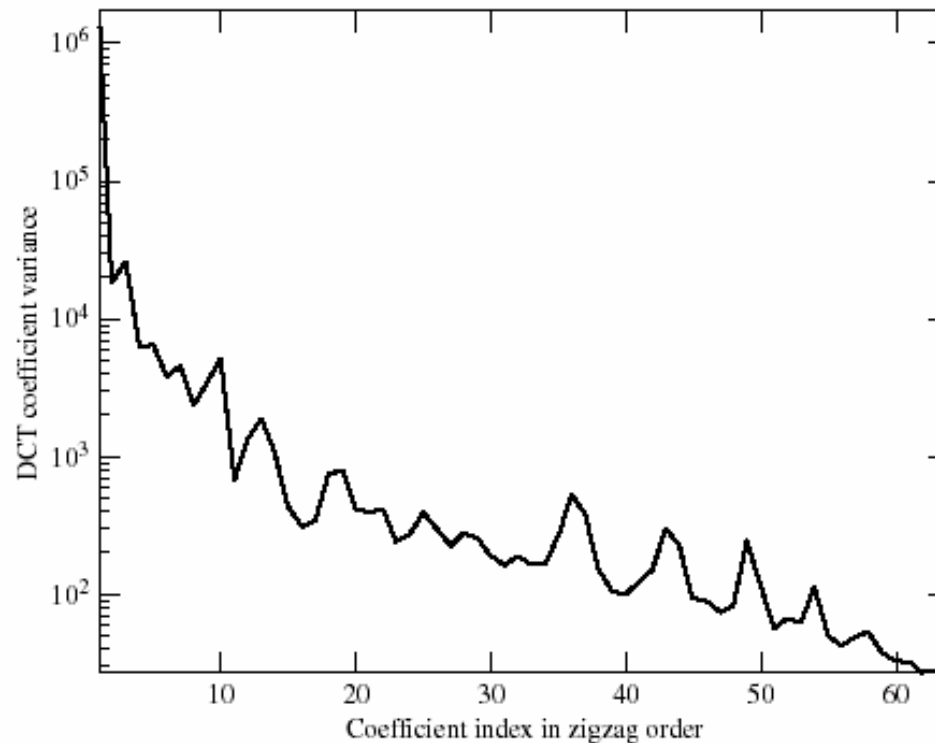


Figure 9.4 Energy distribution of the 8×8 DCT coefficients of the test image "flower."

Low frequency coefficients have much higher variances.

Variance for 1 coefficient is the mean of the coefficient square over all 8×8 blocks

Approximation by DCT Basis

Original



With 16/64
Coefficients



With 8/64
Coefficients



With 4/64
Coefficients



Matlab demo: dctdemo

VCdemo

- Can show DCT images as collection or as blocks
- Can perform quantization, show quantized image and bit rate

Quantization of DCT Coefficients

- Use uniform quantizer on each coefficient
- Different coefficient is quantized with different step-size (Q):
 - Human eye is more sensitive to low frequency components
 - Low frequency coefficients with a smaller Q
 - High frequency coefficients with a larger Q
 - Specified in a normalization matrix
 - Normalization matrix can then be scaled by a scale factor (QP)

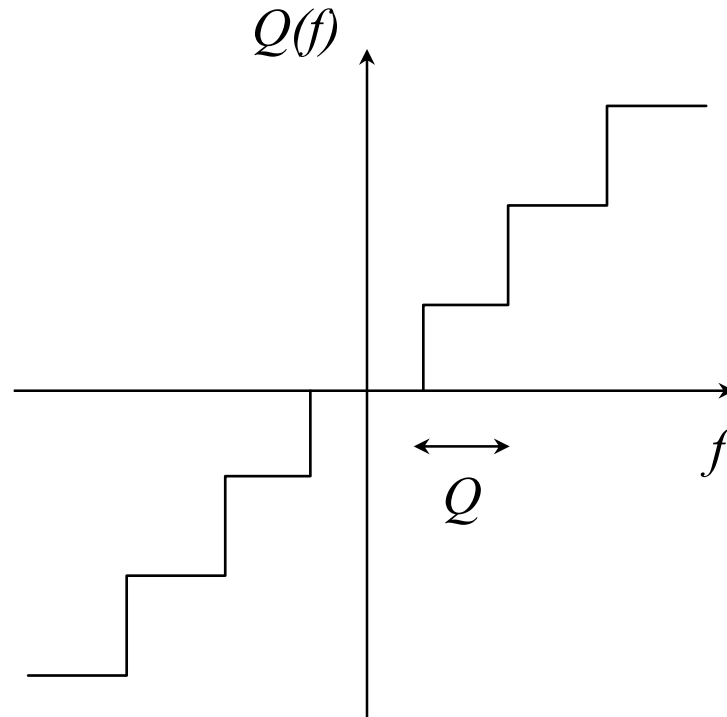
Default Normalization Matrix in JPEG

For Luminance
component

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

Actual step size for C(i,j): $Q(i,j) = QP * M(i,j)$

Uniform Quantization



$$Q_{\text{index}}(f) = \text{floor}((f + Q/2)/Q)$$
$$Q(f) = Q_{\text{index}}(f) * Q$$

Example: Quantized Indices

```
>>dctblock =dct2(imblock)
```

```
dctblock=1.0e+003*
```

```
1.0550 0.0517 0.0012 -0.0246 -0.0120 -0.0258 0.0120 0.0232
0.1136 0.0070 -0.0139 0.0432 -0.0061 0.0356 -0.0134 -0.0130
0.1956 0.0101 -0.0087 -0.0029 -0.0290 -0.0079 0.0009 0.0096
0.0359 -0.0243 -0.0156 -0.0208 0.0116 -0.0191 -0.0085 0.0005
0.0407 -0.0206 -0.0137 0.0171 -0.0143 0.0224 -0.0049 -0.0114
0.0072 -0.0136 -0.0076 -0.0119 0.0183 -0.0163 -0.0014 -0.0035
-0.0015 -0.0133 -0.0009 0.0013 0.0104 0.0161 0.0044 0.0011
-0.0068 -0.0028 0.0041 0.0011 0.0106 -0.0027 -0.0032 0.0016
```

```
>>QP=1;
```

```
>>QM=Qmatrix*QP;
```

```
>>qdct=floor((dctblock+QM/2)./(QM))
```

```
qdct =
```

```
66 5 0 -2 0 -1 0 0
9 1 -1 2 0 1 0 0
14 1 -1 0 -1 0 0 0
3 -1 -1 -1 0 0 0 0
2 -1 0 0 0 0 0 0
0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0
```

Only 19 coefficients are retained out of 64

Example: Quantized Coefficients

```
%dequantized DCT block
```

```
>> iqdct=qdct.*QM
```

```
iqdct=
```

```
1056    55     0   -32     0   -40    0    0
  108    12   -14    38     0    58    0    0
  196    13   -16     0   -40     0    0    0
   42   -17   -22   -29     0     0    0    0
   36   -22     0     0     0     0    0    0
    0     0     0     0     0     0    0    0
    0     0     0     0     0     0    0    0
    0     0     0     0     0     0    0    0
```

```
Original DCT block
```

```
dctblock=1.0e+003*
```

```
1.0550  0.0517  0.0012 -0.0246 -0.0120 -0.0258  0.0120  0.0232
0.1136  0.0070 -0.0139  0.0432 -0.0061  0.0356 -0.0134 -0.0130
0.1956  0.0101 -0.0087 -0.0029 -0.0290 -0.0079  0.0009  0.0096
0.0359 -0.0243 -0.0156 -0.0208  0.0116 -0.0191 -0.0085  0.0005
0.0407 -0.0206 -0.0137  0.0171 -0.0143  0.0224 -0.0049 -0.0114
0.0072 -0.0136 -0.0076 -0.0119  0.0183 -0.0163 -0.0014 -0.0035
-0.0015 -0.0133 -0.0009  0.0013  0.0104  0.0161  0.0044  0.0011
-0.0068 -0.0028  0.0041  0.0011  0.0106 -0.0027 -0.0032  0.0016
```

Example: Reconstructed Image

%reconstructed image block

```
>> qimblock=round(idct2(iqdet))
```

qimblock=

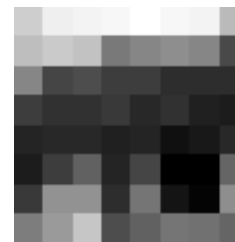
```
186 196 213 207 189 196 195 166
 73 166 167 161 150 152 147 126
149 130 117 116 115 109 104 101
120 109 97 102 108 93 91 113
 97 111 107 108 112 89 87 128
 95 131 127 114 117 91 84 129
112 160 146 118 129 112 98 136
131 182 158 122 144 139 121 151
```



Original image block

imblock=

```
182 196 199 201 203 201 199 173
 75 180 176 142 148 152 148 120
148 118 123 115 114 107 108 107
115 110 110 112 105 109 101 100
104 106 106 102 104 95 98 105
 99 115 131 104 118 86 87 133
112 154 154 107 140 97 88 151
145 158 178 123 132 140 138 133
```



Matlab Implementation

```
% This function does the following processing on each block: DCT, quantize DCT,  
    inverse DCT  
% This function is called by dct_exp  
% Yao Wang 4/10/2003  
  
function qimblock=blkdct_quant(imblock,QP)  
Qmatrix=[16,11,10,16,26,40,51,61;  
    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];  
  
dctblock=dct2(imblock);  
QM=Qmatrix*QP;  
qdct=floor((dctblock+QM/2)./QM);  
iqdct=qdct.*QM;  
qimblock=round(idct2(iqdct));
```

Matlab Implementation

```
%For each image block, perform DCT, quantization and inverse DCT
%User can specify the QP for quantization
%Yao Wang, 4/10/2003
%This program calls function "blkdct_quant"

function dct_exp(fname,QP)
img=imread(fname);
qimg=blkproc(img,[8 8],'blkdct_quant',QP);
figure;
imagesc(qimg),axis image, truesize, axis off; colormap(gray);
title('DCT Domain Quantized Image');
```

Note: the “blkproc” function in matlab performs “blkdct_quant” on each 8x8 block of the image. Using this function is much faster than using a loop that going through each block.

lena256_gray.tif



QP=1



QP=0.5



QP=2



What Should You Know

- How to perform 2D DCT: forward and inverse transform
 - Manual calculation for small sizes, using inner product notation
 - Using Matlab: `dct2`, `idct2`
- Why DCT is good for image coding
 - Real transform, easier than DFT
 - Most high frequency coefficients are nearly zero and can be ignored
 - Different coefficients can be quantized with different accuracy based on human sensitivity
- How to quantize DCT coefficients
 - Varying stepsizes for different DCT coefficients based on visual sensitivity to different frequencies
 - A quantization matrix specifies the default quantization stepsize for each coefficient
 - The matrix can be scaled using a user chosen parameter (QP) to obtain different trade-offs between quality and size

References

Gonzalez and Woods, *Digital image processing*, 2nd edition, Prentice Hall, 2002. Sec.8.5.2.

Vcdemo site:

<http://www-ict.its.tudelft.nl/~inald/vcdemo/>

This site also contain pdf slides of lectures on transform coding.