

# Lecture 11 : Discrete Cosine Transform

# Moving into the Frequency Domain

Frequency domains can be obtained through the transformation from one (**time** or **spatial**) domain to the other (**frequency**) via

- **Fourier Transform (FT)** (**see Lecture 3**) — **MPEG Audio**.
- **Discrete Cosine Transform (DCT)** (**new**) — Heart of **JPEG** and **MPEG Video**, **MPEG Audio**.

**Note:** We mention some image (and video) examples in this section with DCT (in particular) but also the FT is commonly applied to filter multimedia data.

External Link: [MIT OCW 8.03 Lecture 11 Fourier Analysis Video](#)

# Recap: Fourier Transform

The tool which converts a spatial (real space) description of audio/image data into one in terms of its frequency components is called the **Fourier transform**.

The new version is usually referred to as the **Fourier space description** of the data.

We then essentially process the data:

- *E.g.* for **filtering** basically this means attenuating or setting certain frequencies to zero

We then need to convert data back to real audio/imagery to use in our applications.

The corresponding **inverse** transformation which turns a Fourier space description back into a real space one is called the **inverse Fourier transform**.

# What do Frequencies Mean in an Image?

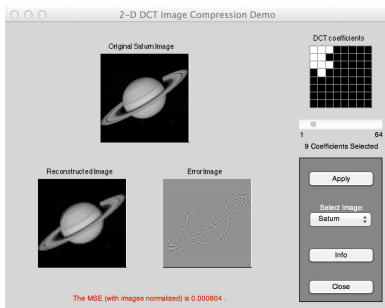
- Large values at **high** frequency components mean the data is changing rapidly on a short distance scale.  
*E.g* .: a page of small font text, brick wall, vegetation.
- Large **low** frequency components then the large scale features of the picture are more important.  
*E.g* . a single fairly simple object which occupies most of the image.

# The Road to Compression

How do we achieve compression?

- Low pass filter — ignore high frequency noise components
  - Only store lower frequency components
- High pass filter — spot gradual changes
  - If changes are too low/slow — eye does not respond so ignore?

# Low Pass Image Compression Example



MATLAB demo, [dctdemo.m](#), (uses DCT) to

- Load an image
- **Low pass filter** in frequency (DCT) space
- **Tune** compression via a single slider value  $n$  to select coefficients
- **Inverse DCT**, **subtract input and filtered image** to see **compression artefacts**.

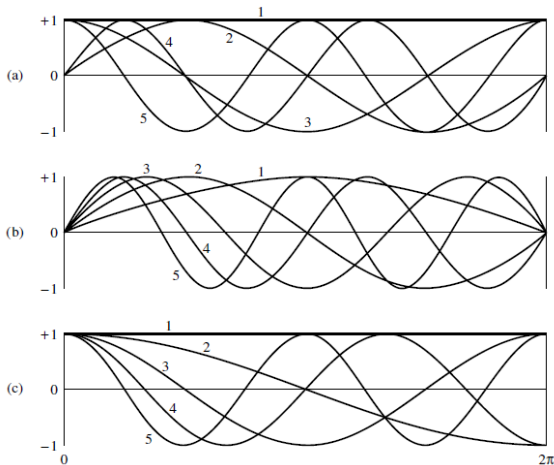
# The Discrete Cosine Transform (DCT)

## Relationship between DCT and FFT

**DCT (Discrete Cosine Transform)** is similar to the DFT since it decomposes a signal into a series of harmonic cosine functions. DCT is actually a **cut-down** version of the Fourier Transform or the Fast Fourier Transform (FFT):

- Only the **real** part of FFT (less data overheads).
- Computationally simpler than FFT.
- DCT— effective for multimedia compression (energy compaction).
- DCT **much** more commonly used (than FFT) in multimedia image/video compression — **more later**.
- Cheap MPEG Audio variant — **more later**.
- FT captures more frequency “fidelity” (*e.g.* phase).

# DCT vs FT



(a) Fourier transform, (b) Sine transform, (c) Cosine transform.



# DCT Example

Let's consider a DC signal that is a constant 100, *i.e*  $f(i) = 100$  for  $i = 0 \dots 7$  ([DCT1Deg.m](#)):

- So the domain is  $[0,7]$  for both  $i$  and  $u$ 
  - We therefore have  $N=8$  samples and will need to work 8 values for  $u=0 \dots 7$ .

We can now see how we work out:  $F(u)$

- As  $u$  varies we can work for each  $u$ , a component or a *basis*  $F(u)$ .
- Within each  $F(u)$ , we can work out the value for each  $F_i(u)$  to define a *basis function*
- Basis function can be pre-computed and simply looked up in DCT computation.

# 1D DCT

For  $N$  data items 1D DCT is defined by:

$$F(u) = \left(\frac{2}{N}\right)^{\frac{1}{2}} \sum_{i=0}^{N-1} \Lambda(u) \cos\left[\frac{\pi u}{2N}(2i+1)\right] f(i)$$

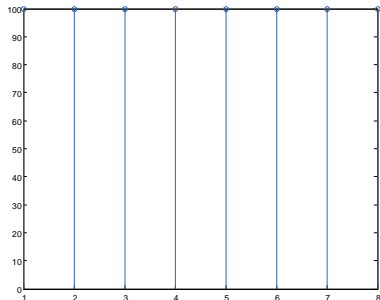
and the corresponding **inverse** 1D DCT transform is simply  $F^{-1}(u)$ , i.e.:

$$\begin{aligned} f(i) &= F^{-1}(u) \\ &= \left(\frac{2}{N}\right)^{\frac{1}{2}} \sum_{u=0}^{N-1} \Lambda(u) \cos\left[\frac{\pi u}{2N}(2i+1)\right] f(u) \end{aligned}$$

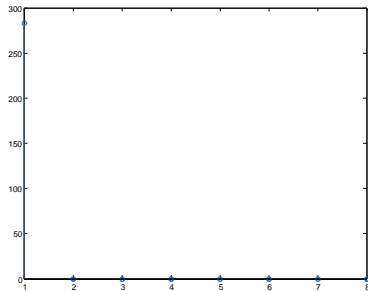
where

$$\Lambda(\xi) = \begin{cases} \frac{1}{\sqrt{2}} & \text{for } \xi=0 \\ 1 & \text{otherwise} \end{cases}$$

# Plots of $f(I)$ and $F(U)$



$f(i) = 100$  for  $i=0 \dots 7$



$F(u): F(0) \approx 283, F(1 \dots 7) = 0$

# DCT Example: $F(0)$

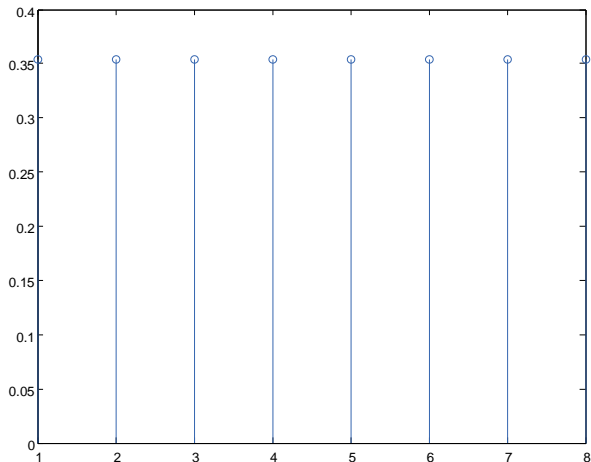
So for  $u = 0$ :

- Note:  $\Lambda(0) = \frac{1}{2\sqrt{2}}$  and  $\cos(0) = 1$
- So  $F(0)$  is computed as:

$$F(0) = \frac{1}{2\sqrt{2}} (1 \cdot 100 + 1 \cdot 100 + 1 \cdot 100 + 1 \cdot 100 + 1 \cdot 100 + 1 \cdot 100 + 1 \cdot 100 + 1 \cdot 100)$$
$$\approx 283$$

- Here the values  $F_i(0) = \frac{1}{2\sqrt{2}}$  ( $i = 0 \dots 7$ ).  
These are bases of  $F_i(0)$

# F(0) Basis Function Plot



F(0) basis function

# DCT Example: F(1 . . . 7)

So for  $u = 1$ :

Note:  $\Lambda(1) = 1$  and we have  $\cos$  to work out: so  $F(1)$  is computed as:

$$\begin{aligned} F(1) &= \frac{1}{2} \left( \cos \frac{\pi}{16} \cdot 100 + \cos \frac{3\pi}{16} \cdot 100 + \cos \frac{5\pi}{16} \cdot 100 + \cos \frac{7\pi}{16} \cdot 100 \right. \\ &\quad \left. + \cos \frac{9\pi}{16} \cdot 100 + \cos \frac{11\pi}{16} \cdot 100 + \cos \frac{13\pi}{16} \cdot 100 + \cos \frac{15\pi}{16} \cdot 100 \right) \\ &= 0 \end{aligned}$$

(since  $\cos(\frac{\pi}{16}) = -\cos(\frac{15\pi}{16})$ ,  $\cos(\frac{3\pi}{16}) = -\cos(\frac{13\pi}{16})$  etc.)

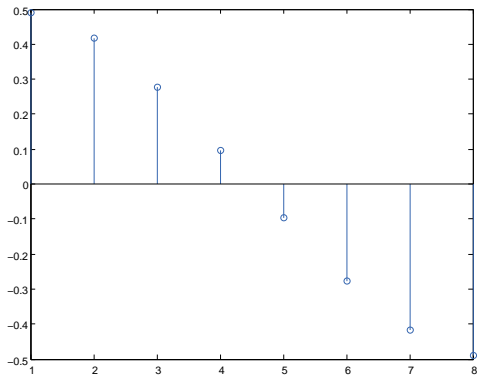
Here the values

$$F_i(1) = \left\{ \frac{1}{2} \cos\left(\frac{\pi}{16}\right), \frac{1}{2} \cos\left(\frac{3\pi}{16}\right), \frac{1}{2} \cos\left(\frac{5\pi}{16}\right), \dots, \frac{1}{2} \cos\left(\frac{11\pi}{16}\right), \frac{1}{2} \cos\left(\frac{13\pi}{16}\right), \frac{1}{2} \cos\left(\frac{15\pi}{16}\right) \right\}$$

form the **basis function**

$F(2 \dots 7)$  similarly = 0

# F(1) Basis Function Plot



F(1) basis function

## Note:

- Bases are *reflected about centre* and *negated* since

$$\cos\left(\frac{\pi}{16}\right) = -\cos\left(\frac{15\pi}{16}\right), \cos\left(\frac{3\pi}{16}\right) = -\cos\left(\frac{13\pi}{16}\right) \text{ etc.}$$

- **only** as our example function is a constant is F(1) zero.

# DCT Matlab Example

DCT1Deg.m explained:

```
i = 1:8 % dimension of vector  
f(i) = 100 % set function  
figure(1) % plotf  
stem(f);  
%compute DCT  
D = dct(f);  
figure(2) % plotD  
stem(D);
```

- Create our function f, and plot it.
- Use MATLAB 1D dct function to compute DCT of f and plot it.



# DCT Matlab Example

```
% Illustrate DCT bases compute DCT bases  
% with dctmtx
```

```
bases =dctmtx(8);  
% Plot bases:each row(j) of bases is the jth  
%DCT Basis Function
```

```
for j= 1:8  
figure %increment figure  
stem(bases(j,:)); %plot rows  
end
```

- MATLAB dctmtx function computes DCT basis functions.
- Each row  $j$  of bases is the basis function  $F(j)$ .
- Plot each row.

# DCT Matlab Example

```
% construct DCT from Basis Functions Simply  
% multiply f' (column vector) by bases
```

```
D1 =bases*f';
```

```
figure(1)    % plot D1  
stem(D1);
```

- Here we show how to compute the DCT from the basis functions.
- bases is an  $8 \times 8$  matrix, f an  $1 \times 8$  vector. Need column  $8 \times 1$  form to do matrix multiplication so transpose f.

# 2D DCT

For a 2D N by M image 2D DCT is defined:

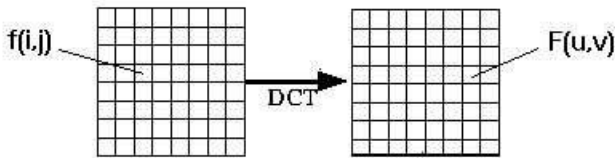
$$F(u, v) = \left(\frac{2}{N}\right)^{\frac{1}{2}} \cdot \left(\frac{2}{M}\right)^{\frac{1}{2}} \sum_{i=0}^{N-1} \sum_{j=0}^{M-1} \Lambda(u) \cdot \Lambda(v) \times \\ \cos \frac{\pi u}{2N} (2i+1) \cos \frac{\pi v}{2M} (2j+1) \cdot f(i, j)$$

and the corresponding **inverse** 2D DCT transform is simply  $F^{-1}(u, v)$ , i.e.:

$$f(i, j) = F^{-1}(u, v) \\ = \left(\frac{2}{N}\right)^{\frac{1}{2}} \cdot \left(\frac{2}{M}\right)^{\frac{1}{2}} \sum_{i=0}^{N-1} \sum_{j=0}^{M-1} \Lambda(u) \cdot \Lambda(v) \times \\ \cos \frac{\pi u}{2N} (2i+1) \cos \frac{\pi v}{2M} (2j+1) \cdot F(u, v)$$

# Applying The DCT

- Similar to the discrete Fourier transform:
  - It transforms a signal or image from the spatial domain to the frequency domain.
  - DCT can approximate lines well with fewer coefficients.



- Helps separate the image into parts (or spectral sub-bands) of differing importance (with respect to the image's visual quality).

# Performing DCT Computations

The basic operation of the DCT is as follows:

- The input image is  $N$  by  $M$ ;
- $f(i, j)$  is the intensity of the pixel in row  $i$  and column  $j$ .
- $F(u, v)$  is the DCT coefficient in row  $u_i$  and column  $v_j$  of the DCT matrix.
- For JPEG image (and MPEG video), the DCT input is usually an 8 by 8 (or 16 by 16) array of integers. This array contains each image window's respective colour band pixel levels.

# Compression with DCT

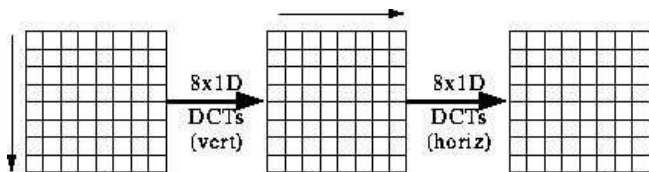
- For most images, much of the signal energy lies at low frequencies;
  - These appear in the upper left corner of the DCT.
- Compression is achieved since the lower right values represent higher frequencies, and are often small
  - Small enough to be neglected with little visible distortion.

# Separability

- One of the properties of the 2-D DCT is that it is separable meaning that it can be separated into a pair of 1-D DCTs.
- To obtain the 2-D DCT of a block a 1-D DCT is first performed on the rows of the block then a 1-D DCT is performed on the columns of the resulting block.
- The same applies to the IDCT.

# Separability

- Factoring reduces problem to a series of 1D DCTs (No need to apply 2D form directly):
  - As with 2D Fourier Transform.
  - Apply 1D DCT (vertically) to columns.
  - Apply 1D DCT (horizontally) to resultant vertical DCT.
  - Or alternatively horizontal to vertical.





# Computational Issues

- The equations are given by:

$$G(i, v) = \frac{1}{2} \sum_j \Lambda(v) \cdot \cos \frac{\pi v}{16} (2j+1) \cdot f(i, j)$$

$$F(u, v) = \frac{1}{2} \sum_i \Lambda(u) \cdot \cos \frac{\pi v}{16} (2i+1) \cdot G(i, v)$$

- Most software implementations use fixed point arithmetic. Some fast implementations approximate coefficients so all multiplies are shifts and adds.

## 2D DCT on an Image Block

- Image is partitioned into 8 x8 regions (See JPEG)—  
The DCT input is an 8 x 8 array of integers.
- So in  $N = M = 8$ , substitute these in DCT formula.
- An 8 point DCT is then:

$$F(u, v) = \frac{1}{4} \Lambda(u) \Lambda(v) \sum_{i=0}^7 \sum_{j=0}^7 \cos \frac{\pi u}{16} (2i+1) \times \\ \cos \frac{\pi v}{16} (2j+1) \cdot f(i, j)$$

where

$$\Lambda(\xi) = \begin{cases} \frac{1}{\sqrt{2}} & \text{for } \xi=0 \\ 1 & \text{otherwise} \end{cases}$$

- The output array of DCT coefficients contains integers; these can range from -1024 to 1023.

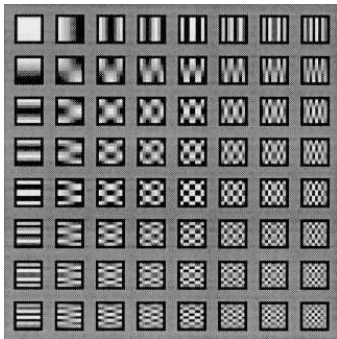
## 2D DCT Basis Functions

From the above formula, extending what we have seen with the 1D DCT we can derive basis functions for the 2D DCT:

- We have a basis for a 1D DCT (see bases = dctmtx(8) example above).
- We discussed above that we can compute a DCT by first doing a 1D DCT in one direction (e.g. horizontally) followed by a 1DCT on the intermediate DCT result.
- This is equivalent to performing matrix pre-multiplication by bases and matrix post-multiplication the transpose of bases.
  - take each row  $i$  in bases and you get 8 basis matrices for each  $j$ .
  - there are 8 rows so we get 64 basis matrices.

# Visualisation of DCT 2D Basis Functions

- Computationally easier to implement and more efficient to regard the DCT as a set of **basis functions**.
  - Given a known input array size (8 x 8) they can be precomputed and stored.
  - The values as simply calculated from DCT formula.



See MATLAB demo, [dctbasis.m](#), to see how to produce these bases.

# DCT Basis Functions

```
A = dctmtx(8);  
A = A';
```

```
Offset = 5;
```

```
basisim = ones(N*(N+offset))*0.5;
```

- Basically just set up a few things  $A := 1D$  DCT basis functions
- `basisim` will be used to create the plot of all 64 basis functions.

# DCT Basis Functions

```
B=zeros(N,N,N,N);  
for i=1:N  
    for j=1:N  
        B(:,:,i,j)=A(:,i)*A(:,j)';  
        %max(max(B(:,:,i,j)))-min(min(B(:,:,i,j)))  
    end;  
end;
```

- B = computation of 64 2D bases.
- Create a 4D array: first two dimensions store a 2D image for each i,j.
- 3rd and 4th dimension i and j store the 64 basis functions.

# DCT Basis Functions

```
for i=1:N
for j=1:N
minb = min(min(B(:, :, i, j)));
maxb = max(max(B(:, :, i, j)));
rangeb =maxb -minb;

if rangeb ==0
    minb =0;
    rangeb =maxb;
end;

imb = B(:, :, i, j) - minb;
imb =imb/rangeb;
```

# DCT Basis Functions

```
iindex1 =(i-1)*N + i*offset-1;  
iindex2 =iindex1 + N -1;
```

```
jindex1 =(j-1)*N + j*offset -1;  
jindex2 =jindex1 + N -1;
```

```
basisim(iindex1: iindex2, jindex1:jindex2) = imb;  
end;  
end;
```

- Basically just put up 64 2D bases in basisim as image data.



# DCT Basis Functions

```
figure(1)  
imshow(basisim)
```

```
figure(2)  
dispbasisim = imresize(basisim,4,'bilinear');  
imshow(dispbasisim);
```

- Plot normal size image and one 4 times up sampled.