Digital Signal Processing

Lecture 8 – Discrete Cosine Transform

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Approach to DCT

- Very useful tool.
 - To compress data in MP3, JPEG and MPEG.
 - DCT is similar to DFT, which has been used for spectral analysis.
- Steps to learn DCT.
 - Synthesis: given a set of freq. components and their amplitudes, how can we construct a wave.
 - Rewrite the synthesis problem using Numpy arrays.
 - Analysis: given a signal and a set of frequencies, how can we find the amplitude of each freq. component?
 - Find more efficient analysis algorithm using some linear algebra principles.

Synthesis I

Suppose I gave you a list of amplitudes and a list of frequencies, and ask you to construct a signal that is the sum of these freq. components.

amps is a list of amplitudes, fs is the list of frequencies, and ts is the sequence of times where the signal should be evaluated.

components is a list of CosSignal objects, one for each amplitude-frequency pair. SumSignal represents the sum of these frequency components.

Synthesis 2

Usage of synthesize I function

```
amps = np.array([0.6, 0.25, 0.1, 0.05])
fs = [100, 200, 300, 400]
framerate = 11025

ts = np.linspace(0, 1, framerate)
ys = synthesize1(amps, fs, ts)
wave = thinkdsp.Wave(ys, framerate)
```

This example makes a signal that contains a fundamental frequency at 100 Hz and three harmonics (100 Hz is a sharp G2). It renders the signal for one second at 11,025 frames per second and puts the results into a Wave object.

Synthesis with array 1

```
def synthesize2(amps, fs, ts):
    args = np.outer(ts, fs)
    M = np.cos(PI2 * args)
    ys = np.dot(M, amps)
    return ys
```

- np.outer computes the outer product of ts and fs. The result is an array with one row for each element of ts and one column for each element of fs. Each element in the array is the product of a frequency and a time, ft.
- We multiply args by 2π and apply cos, so each element of the result is cos(2πft). Since the ts run down the columns, each column contains a cosine signal at a particular frequency, evaluated at a sequence of times.
- np.dot multiplies each row of M by amps, element-wise, and then adds up the products. In terms of linear algebra, we are multiplying a matrix, M, by a vector, amps. In terms of signals, we are computing the weighted sum of frequency components.

Synthesis with array 2

Each row of M corresponds to a time from 0.0 to 1.0 seconds

Each column of M corresponds to a freq. from 100 to 400 Hz.

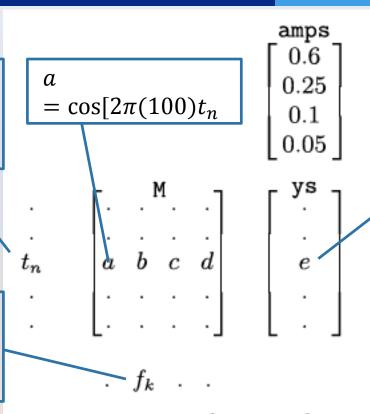


Figure 6.1: Synthesis with arrays.

$$M = \cos(2\pi t \otimes f)$$
$$y = Ma$$

$$e$$
= $0.6a + 0.25b$
+ $0.1c + 0.05d$

So, each element of ys is the sum of four frequency components, evaluated at a point in time, and multiplied by the corresponding amplitudes.



- Suppose I gave you a wave and tell you that it is the sum of cosines with a given set of frequencies.
- How would you find the amplitude for each freq. component? (Given ys, ts and fs, recover amps)
 - The first step : compute $M = \cos(2\pi t \otimes f)$.
 - Find a so that y = Ma.

```
def analyze1(ys, fs, ts):
    args = np.outer(ts, fs)
    M = np.cos(PI2 * args)
    amps = np.linalg.solve(M, ys)
    return amps
```

Solving a linear system

In general, we can only solve a system of linear equations if the matrix is square (the number of equations (rows) = the number of unknowns (columns)

In this example, we have only 4 frequencies, but we have evaluated the signal at 11,025 times. So we have many more equations than unknowns.

```
n = len(fs)
amps2 = analyze1(ys[:n], fs, ts[:n])_
```

Because we know that ys were generated by adding only 4 freq. components

Slow

Orthogonal matrices 1

- \blacksquare To solve the equation y = Ma,
 - $M^{-1}y = M^{-1}Ma = a$
- If we can compute M^{-1} efficiently, we can find a with a simple matrix multiplication.
 - This algorithms has n^2 complexity, witch is better than n^3 .
- Inverting a matrix is slow, but some special cases are faster.
 - If M is orthogonal, the M^{-1} is just the M^T , which implies $M^TM = I$.

Orthogonal matrices 2

- Synthesize2()
 - Since M has 11,025 rows, let's work with a smaller example.

time_unit drops out of the computation.

Since there are 4 freq. components, we chose 4 samples in time That way, M is square.

time unit = 0.001 is a arbitrary choice.

Since N samples per time_unit, the framerate is N/time unit and the Nyquist freq. is framerate/2, which is 2000Hz. So fs is a vector of equally spaced frequencies between 0 and 2000Hz.

```
def test1():
   amps = np.array([0.6, 0.25, 0.1, 0.05])
   N = 4.0
   time_unit = 0.001
   ts = np.arange(N) / N * time_unit
   max_freq = N / time_unit / 2
   fs = np.arange(N) / N * max_freq
  ys = synthesize2(amps, fs, ts)
```

$$M = \begin{bmatrix} 1. & 1. & 1. & 1. \\ 1. & 0.707 & 0. & -0.707 \\ 1. & 0. & -1. & 0. \\ 1. & -0.707 & 0. & 0.707 \end{bmatrix} 0.707 = \frac{\sqrt{2}}{2} = \cos\frac{\pi}{4}$$

$$M^{T} = M$$

$$M^T M = \begin{bmatrix} 4. & 1. & 0. & 1. \\ 1. & 2 & 1. & 0. \\ 0. & 1. & 2. & 1. \\ 1 & 0 & 1 & 2 \end{bmatrix} M \text{ is not orthogonal.}$$

DCT-IV I

- If we choose ts and fs carefully, we can make M orthogonal.
 - There are several ways to do it, so there are several versions of the Discrete Cosine Transform (DCT).
 - One simple option is to shift ts and fs by a half unit. (DCT-IV)

Two changes:

I. time_unit is canceled out.

2. 0.5 is added to ts and fs

$$M = \begin{bmatrix} 0.981 & 0.831 & 0.556 & 0.195 \\ 0.831 & -0.195 & -0.981 & -0.556 \\ 0.556 & -0.981 & 0.195 & 0.831 \\ 0.195 & -0.556 & 0.831 & -0.981 \end{bmatrix}$$

$$M^T M = \begin{bmatrix} 2. & 0. & 0. & 1. \\ 0. & 2 & 0. & 0. \\ 0. & 0. & 2. & 0. \\ 0. & 0. & 0. & 2. \end{bmatrix} = 2I$$

2I is almostorthogonal.
$$\frac{M^TM}{2} = I$$

DCT-IV 2

M is symmetric and almost orthogonal.

$$M \frac{M^T}{2} = M \frac{M}{2} = I$$
, so $M^{-1} = \frac{M}{2}$.

Now we can write a more efficient version of analyse.

Instead of using np.linalg.solve, we just multiply by M/2

def analyze2(ys, fs, ts):
 args = np.outer(ts, fs)
 M = np.cos(PI2 * args)
 amps = np.dot(M, ys) / 2
 return amps

ys is the wave array. No need of ts and fs

```
def dct_iv(ys):
    N = len(ys)
    ts = (0.5 + np.arange(N)) / N
    fs = (0.5 + np.arange(N)) / 2
    args = np.outer(ts, fs)
    M = np.cos(PI2 * args)
    amps = np.dot(M, ys) / 2
    return amps
```

Inverse DCT

■ Notice that analyze2() and synthesize2() are almost identical.

```
def synthesize2(amps, fs, ts):
    args = np.outer(ts, fs)
    M = np.cos(PI2 * args)
    ys = np.dot(M, amps)
    return ys
def analyze2(ys, fs, ts):
    args = np.outer(ts, fs)
    M = np.cos(PI2 * args)
    amps = np.dot(M, ys) / 2
    return amps
```

We can use the above insight to compute the inverse DCT

```
def inverse_dct_iv(amps):
    return dct_iv(amps) * 2
```

The biggest difference is about 1e-16.

```
amps = [0.6, 0.25, 0.1, 0.05]

ys = inverse_dct_iv(amps)

amps2 = dct_iv(ys)

max(abs(amps - amps2))
```

The DCT class I

To make a Dct object, you can invoke make_dct on a Wave.

```
signal = thinkdsp.TriangleSignal(freq=400)
wave = signal.make_wave(duration=1.0, framerate=10000)
dct = wave.make_dct()
dct.plot()
```

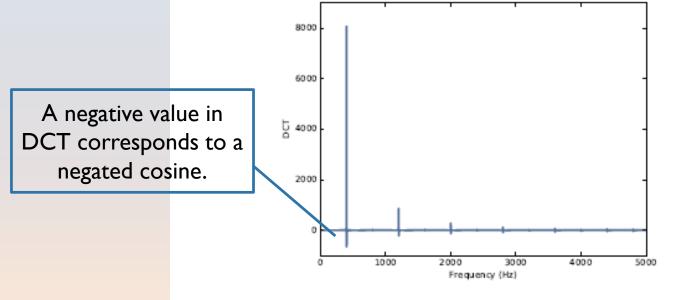


Figure 6.2: DCT of a triangle signal at 400 Hz, sampled at 10 kHz.

The DCT class 2

DCT-II in scipy.fftpack

```
import scipy.fftpack

# class Wave:
    def make_dct(self):
        N = len(self.ys)
        hs = scipy.fftpack.dct(self.ys, type=2)
        fs = (0.5 + np.arange(N)) / 2
        return Dct(hs, fs, self.framerate)
```

No difference in result. Check with this!

```
wave2 = dct.make_wave()
max(abs(wave.ys-wave2.ys))
```