

# Lecture 20: Fourier Transform and Speech Recognition

University of Southern California

Linguistics 285

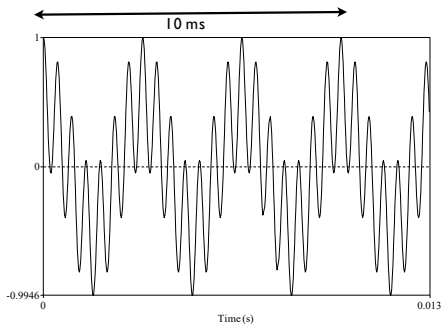
USC Linguistics

November 8, 2015



# Fourier Analysis (Fourier Transform)

- e How do we find the frequencies that compose a signal?
- e In a simple case, we can find them by careful observation of a waveform

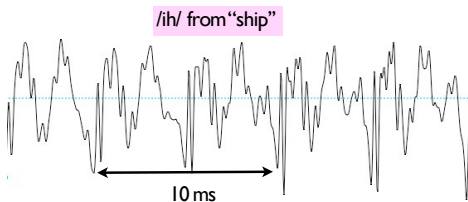
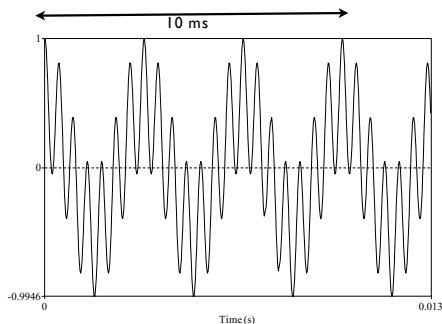


- e But not in a more complex case



# Fourier Analysis (Fourier Transform)

- How do we find the frequencies that compose a signal?
- Observation of waveform in simple, artificial case, but not in complex, real case



# Fourier Analysis (Fourier Transform)

- **Fourier Analysis** is the technique that does this.
- Produces something called the **Fourier Transform**, which contains the information in the spectrum.
- But how does it work?
- By using the inner product!
- Take the inner product of the signal (waveform) with pure tones of all possible frequencies.
- The size of the IP tells us how much that signal is similar to each tone.
- That quantitative similarity is the relative amplitude of the frequency in complex signal.

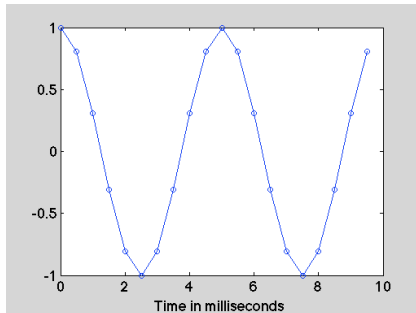


# Fourier Analysis (Fourier Transform)

- But how can we take the inner product of a signal with tones?
- Signals are just sequences of numbers, vectors!
- In this example, a new number (called a sample) every .5 ms

1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090  
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090

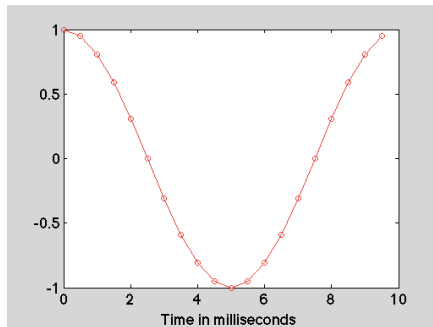
200 Hz tone



# Fourier Analysis (Fourier Transform)

1.0000  
0.9511  
0.8090  
0.5878  
0.3090  
0.0000  
-0.3090  
-0.5878  
-0.8090  
-0.9511  
-1.0000  
-0.9511  
-0.8090  
-0.5878  
-0.3090  
-0.0000  
0.3090  
0.5878  
0.8090  
0.9511

100 Hz tone

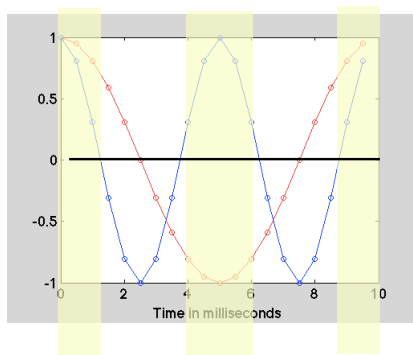


# Fourier Analysis via inner product

- Let's treat the 200 Hz tone as our signal whose spectrum we want to measure.
- Take inner product with 100 Hz, 200 Hz...500 Hz, etc.



- There are times where the product is large where both vectors are positive, but each one is exactly canceled by a time when the vectors have opposite signs, so the product is negative. The sum (IP) is zero.

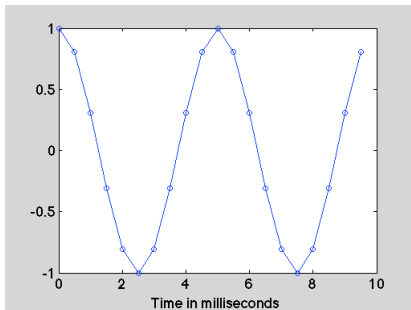


s200	s100	s200.*s100
1.0000	1.0000	1.0000
0.8090	0.9511	0.7694
0.3090	0.8090	0.2500
-0.3090	0.5878	-0.1816
-0.8090	0.3090	-0.2500
-1.0000	0.0000	-0.0000
-0.8090	-0.3090	0.2500
-0.3090	-0.5878	0.1816
0.3090	-0.8090	-0.2500
0.8090	-0.9511	-0.7694
1.0000	-1.0000	-1.0000
0.8090	-0.9511	-0.7694
0.3090	-0.8090	-0.2500
-0.3090	-0.5878	0.1816
-0.8090	-0.3090	0.2500
-1.0000	-0.0000	0.0000
-0.8090	0.3090	-0.2500
-0.3090	0.5878	-0.1816
0.3090	0.8090	0.2500
0.8090	0.9511	0.7694

Sum = 0







s200

1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090  
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090  
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090

s200

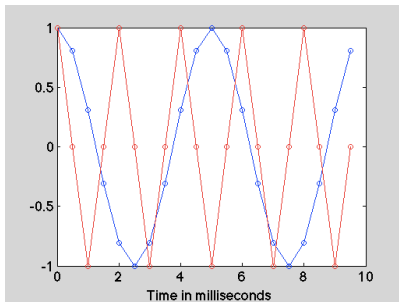
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090  
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090  
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090

s200.\*s200

1.0000  
0.6545  
0.0955  
0.0955  
0.6545  
1.0000  
0.6545  
0.0955  
0.0955  
0.6545  
1.0000  
0.6545  
0.0955  
0.0955  
0.6545  
1.0000  
0.6545  
0.0955  
0.0955  
0.6545  
1.0000  
0.6545  
0.0955  
0.0955  
0.6545  
1.0000  
0.6545  
0.0955  
0.0955  
0.6545

Sum = 10





s200

1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090  
1.0000  
0.8090  
0.3090  
-0.3090  
-0.8090  
-1.0000  
-0.8090  
-0.3090  
0.3090  
0.8090

s500

1.0000  
0.0000  
-1.0000  
-0.0000  
1.0000  
0.0000  
-1.0000  
-0.0000  
1.0000  
0.0000  
-1.0000  
-0.0000  
1.0000  
-0.0000  
-1.0000  
-0.0000  
1.0000  
-0.0000  
-1.0000  
-0.0000

s200.\*s500

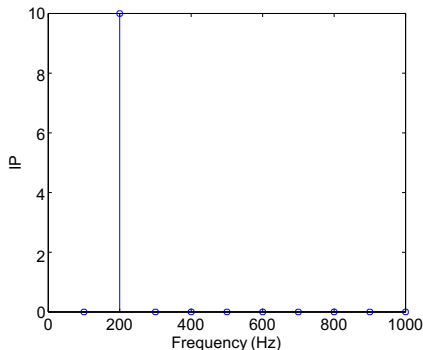
1.0000  
0.0000  
-0.3090  
0.0000  
-0.8090  
-0.0000  
0.8090  
0.0000  
0.3090  
0.0000  
-1.0000  
-0.0000  
0.3090  
0.0000  
0.8090  
0.0000  
-0.8090  
0.0000  
-0.3090  
-0.0000

Sum = 0



# Fourier Analysis via inner product

- Other frequencies will also give 0 as the IP, so this our resulting spectrum:



- This will also work with more complex signals!



## Example recognizer: Match unknown (test) vowels from one speaker to template vowels from another speaker

1. Select a short waveform chunk from each template and test vowel.
2. Save **waveform** chunks to file and import into a matrix in ~~Matlab~~ python
3. Derive the **spectrum** of each vowel by using the Fourier Transform: take the inner product of the vowel waveform with the waveforms of a series of tones with different frequencies.
4. Compare the **similarity** of each test vowel to each template vowel by taking the inner product of the spectrum of the test vowel with each template vowel:
  - ) **Recognition:** The largest IP value obtained for a given test vowel is the vowel that our recognizer selects as the one spoken.



# Perform Fourier Transform for each vowel

- Take the inner product of each vowel waveform (vector) with the waveform of tones from 100 Hz to 3000 Hz, in steps of 100 Hz.
- The resulting vector of inner products for each of the 30 frequencies is the spectrum.
- The tone waveforms are stored in a matrix called **freqs**.
- Each column is a different frequency. The rows contain the successive time samples.

```
>>size (freqs)  
ans =  
2000 30
```

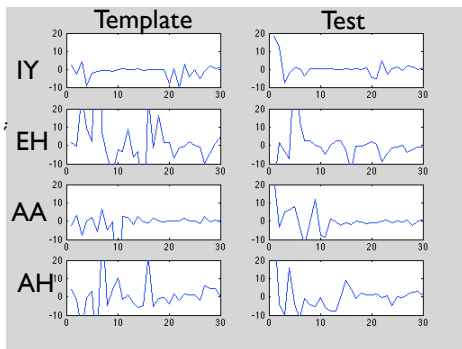


# Perform Fourier Transform for each vowel

- Plot the resulting spectra

```
%plot_spec
```

```
for i = 1:4  
    subplot (4,2,2*i-1), plot(spec_temp(:,i));  
    subplot (4,2,2*i), ylim([-10 20]);  
    subplot (4,2,2*i), plot(spec_test(:,i));  
    subplot (4,2,2*i), ylim([-10 20]);  
end
```



# Compare each unknown vowels to all the templates

- Go thru each test vowel in a loop.
- For each test vowel, go through each template vowel in a loop and take the IP.
- Save the IP in a matrix in which the rows correspond to the test vowels and the columns the template vowels.

```
%do_test
for j = 1:4
    for i=1:4
        IP(j,i) = sum((spec_test(:,j)-mean(spec_test(:,j)))*(spec_temp(:,i)-mean(spec_temp(:,i)))));
    end
end
```

```
>>IP
```

```
IP =
```

```
1.0e+03 *
    0.0023   -0.2601    0.0546    0.0495
    0.0855    1.1562    0.1832   -0.5468
   -0.0204   -0.4350    0.2976   -0.3989
   -0.0648   -1.5917    0.0845    0.7764
```



# Results

- How do we interpret the IP matrix?
- Each row has the results for one test vowel, comparing it to all template vowels, across the columns

		Template			
		IY	EH	AA	AH
Test	IY	0.0023	-0.2601	0.0546	0.0495
	EH	0.0855	1.1562	0.1832	-0.5468
	AA	-0.0204	-0.4350	0.2976	-0.3989
	AH	-0.0648	-1.5917	0.0845	0.7764

- The largest IP for each test vowel is the vowel category our system has recognized for that vowel
- If our recognizer worked, the diagonal cells of the matrix should be larger than the off-diagonal cells.
- 3 vowels were correctly recognized. One was not (/IY/).





# Important functions in numpy

- Numpy is a math library in python that does a lot of things
- linspace: Return evenly spaced numbers over a specified interval
- sin: Computes sinus of number



# Important functions in scipy

- Numpy is a science library in python that does a lot of things
- io.wavfile: Reads wav files – returns the sample rate and the data
- fftpack.fft: Computes fft of data vector (fourier transform)

