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COM3502/4502/6502 SPEECH PROCESSING

Lecture 14

Waveform Processing

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Block Processing

- After capture through a microphone and digitisation through the analogue-to-digital converter, an incoming speech signal becomes a sequence of quantised samples
- Digital signal processing is typically performed on a fixed-length sequence of samples called '**blocks**' or '**frames**'
 - e.g. in Pure Data the default '**block size**' is 64 samples (i.e. 1.45 msec frame at the default 44.1 kHz sampling rate)
- Because of the quasi-stationary nature of speech, the frame size is a compromise of ...
 - having *sufficient* data in a frame to make the required measurements
 - having *small enough* amount of data that the stationarity assumption is fulfilled
- It is also necessary to ensure that there are a sufficient number of frames to capture the *non-stationary* properties

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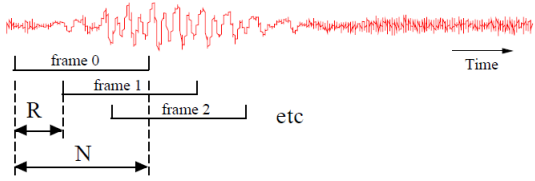
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Block Processing

- To accommodate all these constraints, it is usual to use *overlapping* frames in speech processing
 - 'frame size' (N): number of samples per frame
 - 'frame shift' (R): number of samples between the start of successive frames
- Frame size is often expressed in time ...
 - NT seconds (where T is the sample period)
- Frame shift is often expressed as the 'frame rate' ...
 - $f_r = 1/RT$ frames per second (fps)



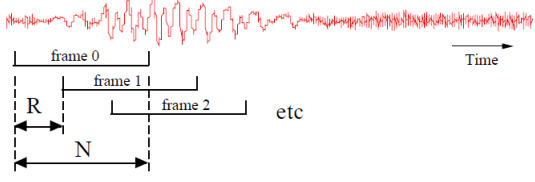
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Block Processing

- In speech, it is usual to have ...
 - frame length (NT) \approx 30 msecs
 - frame rate (f_r) \approx 100 fps
- For example ...
 - sample rate (f_s) = 10 kHz (10,000 samples/sec)
 - sample period (T) = $1/f_s = 100$ μ secs/sample
 - frame size (N) = $NT/T = 0.03/0.0001 = 300$ samples
 - frame shift (R) = $1/f_r T = 1/(100 * 0.0001) = 100$ samples
 - frame overlap ($N-R$) = $300-100 = 200$ samples (66%)



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Block Processing in Pure Data

pd~

Signal Block

A₁ A₂ A₃ A₄

31.4 15.9 26.5 35.8

Wire

B₁ B₂ B₃ B₄

97.9 42.3 84.6 26.4

Object Box

Inlet

+~

A₁+B₁ A₂+B₂ A₃+B₃ A₄+B₄

129.3 58.2 111.1 62.2

Taken from: Farnell, A. (2008). *Designing Sound*. London: Applied Scientific Press Limited.

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Block Processing in Pure Data

pd~

- The default settings in Pure Data are ...
 - sampling rate = 44.1 kHz (*defined by the sound card*)
 - block size = 64 samples (*1.45 ms*)
 - no overlap
- These can be overridden using the `[block~]` object (*but not in the same patch as the `[adc~]` or `[dac~]` objects*)
- `[block~]` takes the following parameters ...
 - block size (*in samples, power of 2*)
 - overlap (*power of 2*)
 - up/down sampling ratio (*relative to parent window*)
- Only one `[block~]` object is allowed in a window
- A reasonable setting for speech is `[block~ 1024 2 1]`
 - 23 msec frame size
 - 86 frames per second
 - 50% frame overlap

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Block Processing in Pure Data

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Sample-by-Sample Processing in Pd

- Sometimes it is necessary to process data one sample at a time
- Pd provides an object for this: **[fexpr~]**
- [fexpr~] takes the following arguments ...
 - \$i#: integer input variable on inlet #
 - \$f#: float input variable on inlet #
- Expressions in [fexpr~] are constructed using ...
 - \$x#[n]: the sample from inlet # indexed by n
 - \$y[n]: the output value indexed by n
(*\$x# is shorthand for the current input*)
(*\$y is shorthand for the previous output: \$y[-1]*)
- E.g. [fexpr~ \$x1+\$y] is a simple accumulator

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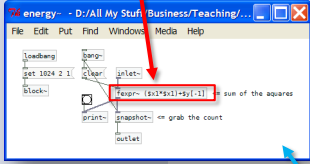
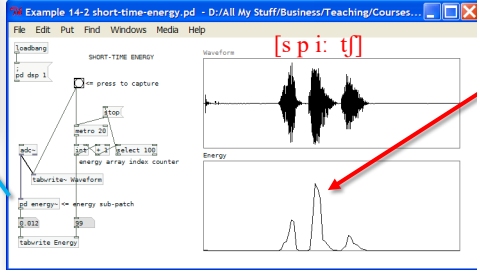
Short-Time Energy


'Short-time energy' = sum of the squares of the samples in one frame

Sample by sample computation

$$E = \sum_{i=0}^{N-1} s_i^2$$

Energy is large in voiced speech



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Short-Time Energy


energy - D:/All My Stuff/Business/Teaching/...

File Edit Put Find Windows Media Help

```

loadbang
set 1024 2 1
block~

bang~
clear
inlet~
fexpr~ (sx1*sx1)+sy[-1] <= sum of the aquares
print~
snapshot~ <= grab the count
outlet
    
```



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Zero-Crossing Rate

'ZCR' = number of times the zero axis is crossed in one frame

zerocross~ - D:/All My Stuff/Business/Teachin...

```

loadbang
set 1024 2 1
block~

inlet~
x~ 1e+006 => multiply by a very large number
clip~ -1 1 => clip to create a square wave

bang~
fexpr~ -$x1*$x1[-1] => multiply adjacent samples
                    (each '1' indicates a zero-crossing)
clear~
clip~ 0 1 => remove all the '-1's
fexpr~ $x1+$y[-1] => count the remaining '1's
print~
snapshot~ => grab the count
outlet~
                    
```

Example 14-3 zero-crossings.pd~ - D:/All My Stuff/Business/Teaching/Courses/2...

ZCR is large in unvoiced speech

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Zero-Crossing Rate

zerocross~ - D:/All My Stuff/Business/Teachin...

```

File Edit Put Find Windows Media Help

loadbang
set 1024 2 1
block~

inlet~
x~ 1e+006 =<= multiply by a very large number
clip~ -1 1 =<= clip to create a square wave

bang~
fexpr~ -$x1*$x1[-1] =<= multiply adjacent samples
                    (each '1' indicates a zero-crossing)
clear~
clip~ 0 1 =<= remove all the '-1's
fexpr~ $x1+$y[-1] =<= count the remaining '1's
print~
snapshot~ =<= grab the count
outlet~
                    
```

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Speech/Non-Speech Detection

- In speech processing it is often useful to be able to detect when someone is speaking
- Accurate speech 'end-point detection' is very difficult
- A simple 'speech/non-speech detector' can be constructed using short-time energy *and* zero-crossing rate
 - energy is high in voiced speech
 - ZCR is high in unvoiced speech



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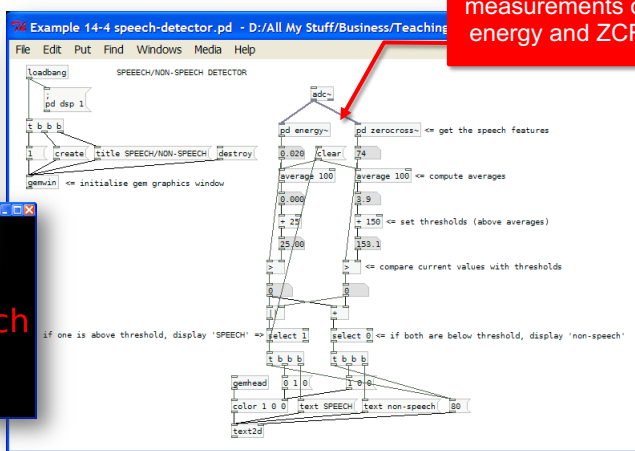
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Speech/Non-Speech Detection

Based on measurements of energy and ZCR





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Autocorrelation Function

The '**autocorrelation function**' computes the correlation of a signal with itself (as a function of time)

$$r_k = \sum_{i=0}^{N-k-1} s_i \cdot s_{i+k}$$

The diagram illustrates the computation of the autocorrelation function. It shows a signal s_i and a delayed signal s_{i+k} with a time delay k . The diagram highlights the process of multiplying corresponding samples and summing the results to produce r_k . A red box indicates that r_0 represents energy. The x-axis for the correlation function is labeled k , ranging from 0 to $N-1$.

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Autocorrelation Function

- The autocorrelation function (**ACF**) emphasises periodicity
- ACF is the basis for many spectrum analysis methods
- Short-time ACF (STACF) is the basis for many '**pitch detectors**' (fundamental frequency estimators)
- ACF is fairly expensive to compute (because there is an inner loop running for every data sample)
- STACF is often combined with ZCR to construct a '**voiced/unvoiced detector**'

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Autocorrelation Function

[æ]

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Autocorrelation Function

[s]

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Correlation

The 'correlation' between two discrete-time signals s and t over an N point interval is ...

$$q = \sum_{i=0}^{N-1} s_i \cdot t_i$$


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Cosine Correlation

For two sinusoids (where N and T are chosen so the summation is over an *integer* number of cycles for both signals) ...


$$s[nT] = A \cdot \cos(\omega_s nT)$$

$$t[nT] = \cos(\omega_t nT)$$

$$q = s[0] \cdot t[0] + s[T] \cdot t[T] + s[2T] \cdot t[2T] + \dots + s[(N-1)T] \cdot t[(N-1)T]$$

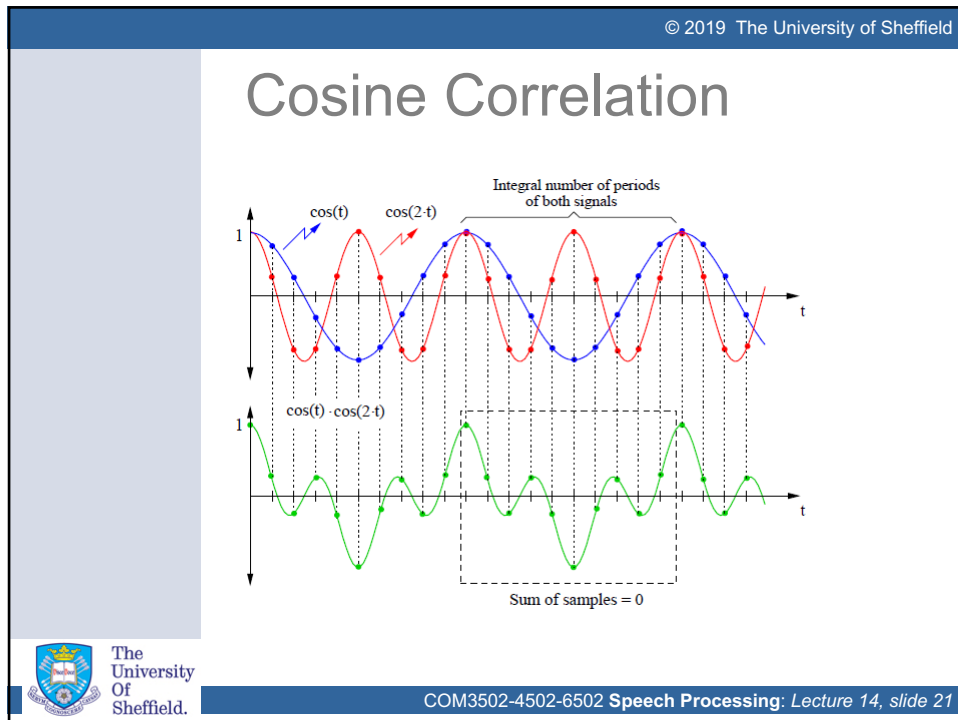
It can be shown that ...

$$q = \begin{cases} \alpha \cdot A & \text{if } \omega_s = \omega_t \\ 0 & \text{otherwise} \end{cases}$$


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Cosine Correlation

- The correlation between a test signal $t[NT]$ and a target signal $s[NT]$ is proportional to the amplitude A of the target signal when ...

$$\cos(\omega_s NT) = \cos(\omega_t NT)$$
- So, given that Fourier analysis shows that any signal can be decomposed into sinusoidal waves, cosine correlation can be used as a method to find (*extract*) the cosine components of an arbitrary signal
- This is only possible if the correlation is computed over an integer number of p cycles (*in the test signal*)

$$\omega_t = \frac{2\pi p}{NT} \quad p = 0, 1, \dots, N-1$$

$$f_t = \frac{p}{NT}$$

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Cosine Correlation

So, the spectrum computed by cosine correlation is ...

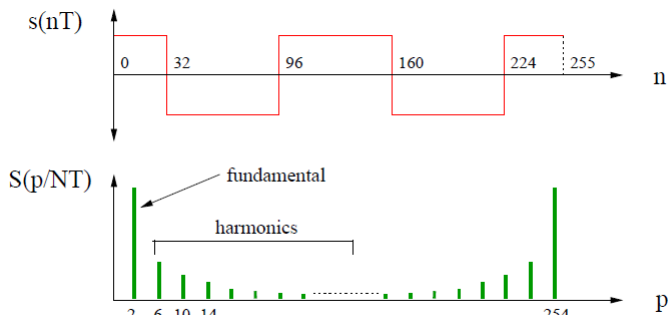
$$S_p = \sum_{n=0}^{N-1} s_n \cdot \cos\left(\frac{2\pi np}{N}\right) \quad p = 0 \dots N-1$$

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Cosine Correlation



$S_p = \sum_{n=0}^{N-1} s_n \cdot \cos\left(\frac{2\pi np}{N}\right) \quad p = 0 \dots N-1$

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Cosine Correlation

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This lecture has covered ...

- Processing blocks/frames
- Processing samples
- Short-time energy
- Zero crossing rate
- Speech/non-speech detection
- Autocorrelation
- Cosine correlation


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Any Questions ?



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Next time ...

The Fourier Transform

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