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# COM3502-4502-6502 SPEECH PROCESSING

## Lecture 17 Linear Filters

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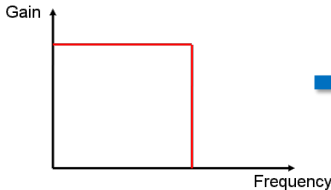
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## Idealised vs. Practical Filters

- Filters are often described in idealised terms, e.g. as **'brickwall'** filters
- In reality, practical filters approximate the idealised forms

*Idealised 'Brickwall' Filter*

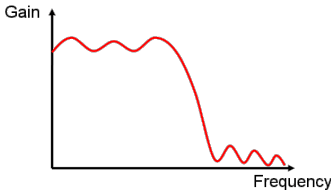


Gain

Frequency

➔

*Practical Filter*



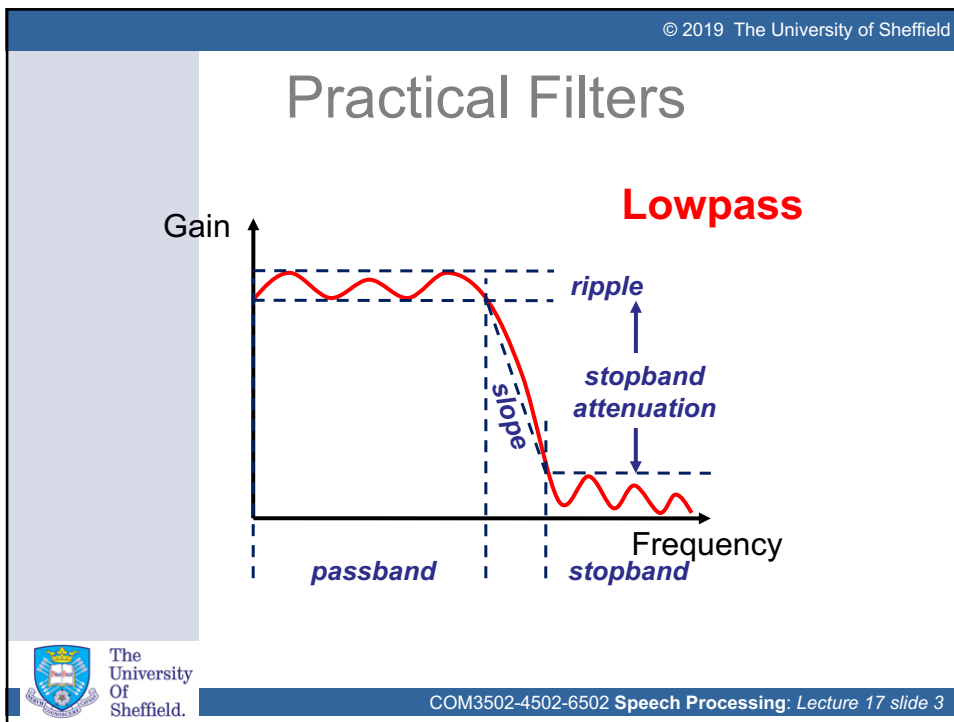
Gain

Frequency

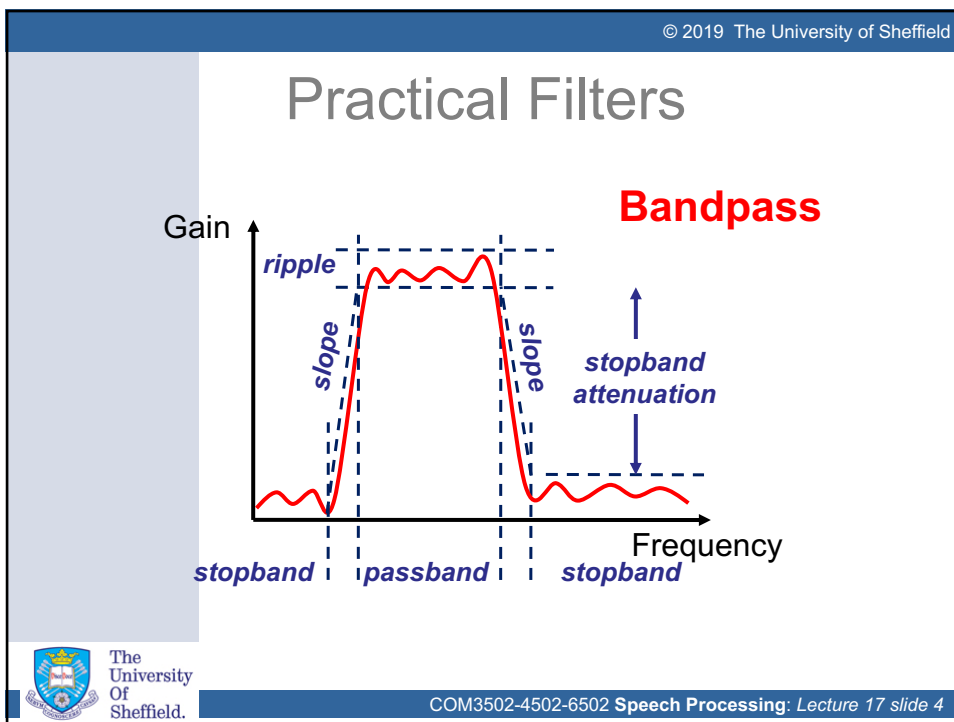
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## Filter Design

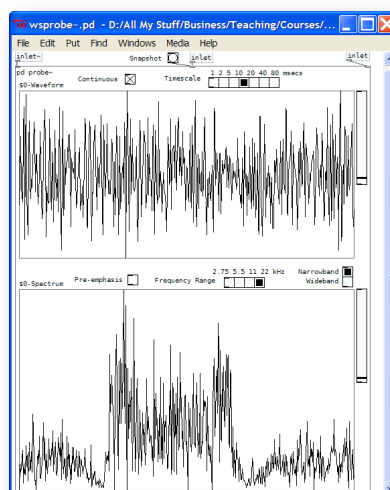
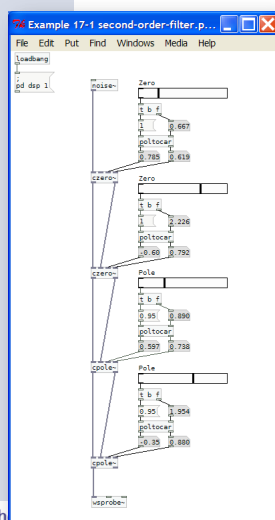
- The behaviour of a filter is defined by the number and the location of its **'poles'** and **'zeros'** (as we saw in *Lecture #16*)
- The complexity of a filter is often characterised by its **'order'**
- In a digital filter, order is defined as the number of *past* (input or output) values that are involved in the calculation
- For example, a filter that takes two past values is termed a **'second-order filter'**
- Filter order can thus be related to the number of poles and zeros
- Hence, the degree to which a practical filter approximates the idealised form is a function of its order (i.e. the number of poles and zeros)



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## 2-Pole 2-Zero 1<sup>st</sup>-Order Filtering



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## General Difference Equation

- The general difference equation may include any number of past inputs and outputs ...

$$y[nT] = a_1 y[(n-1)T] + a_2 y[(n-2)T] + \dots + a_p y[(n-p)T] + b_0 x[nT] + b_1 x[(n-1)T] + b_2 x[(n-2)T] + \dots + b_q x[(n-q)T]$$

- Taking Z transforms ...

$$Y(Z) = Y(z) \sum_{i=1}^p a_i z^{-i} + X(z) \sum_{i=0}^q b_i z^{-i}$$


$$H(z) = \frac{\sum_{i=0}^q b_i z^{-i}}{1 - \sum_{i=1}^p a_i z^{-i}}$$

$\sum_{i=0}^q b_i z^{-i}$ 

← zeros

$1 - \sum_{i=1}^p a_i z^{-i}$ 

← poles



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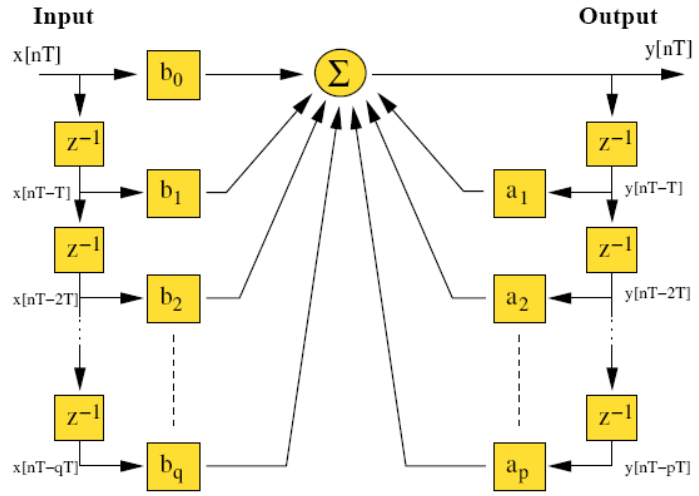
## General Digital Filter


**Input**

$x[nT]$

**Output**

$y[nT]$





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## General Digital Filter

- A very wide range of transfer functions (*and associated frequency responses*) may be obtained by appropriate choice of the filter parameters
- $p$  = number of poles,  $q$  = number of zeros
- For the case  $p = 0$ , the resulting non-recursive filter is said to be a 'finite impulse response - FIR' filter
- If  $p > 0$ , the resulting recursive filter is said to be an 'infinite impulse response - IIR' filter


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## 'FIR' Filters

- Finite impulse response (*FIR*) filters express each output sample as a weighted sum of the last  $N$  inputs (*where  $N$  is the order of the filter*)
- Advantages
  - inherently **stable** since they don't use feedback (*i.e. only zeros*)
  - the coefficients are usually symmetrical, hence the phase response is **linear** and signals of all frequencies are delayed equally
  - overflow is straightforward to avoid
  - generally easier to design than IIR filters
- Disadvantages
  - may require significantly more processing and memory resources than the equivalent IIR filter
  - often require a much higher filter order than IIR filters
  - delay can be much greater than for an equivalent IIR filter


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## ‘IIR’ Filters

- Infinite impulse response (*IIR*) filters are the digital counterpart to analog filters
  - ... because they contain an internal state, and the output and the next internal state are determined by the previous inputs and outputs
- Advantages
  - normally require less computing resources than an FIR filter of similar performance
- Disadvantages
  - due to the feedback, high-order IIR filters may have problems with instability, arithmetic overflow and limit cycles
  - careful design is required to avoid such pitfalls
  - the time delay through such a filter is frequency-dependent (*since the phase shift is inherently a non-linear function of frequency*)

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## ‘IIR’ Filters

- IIR filters all approximate the ideal *brickwall* filter ...
  - Butterworth
  - Chebyshev (*types I and II*)
  - elliptic
  - Bessel
- High-order IIR filters can easily become unstable
- This is much less of a problem with first and second-order filters
- 2<sup>nd</sup>-order IIR filters are often called ‘**biquads**’
- Higher-order filters are typically implemented as a cascade of ‘**biquad sections**’

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## ‘Biquad’ Filters

- A **‘biquadratic’** (*biquad*) filter is a 2<sup>nd</sup>-order recursive linear IIR filter, containing *two* poles and *two* zeros
- The name refers to the fact that its Z domain transfer function is the ratio of two *quadratic* functions ... 
$$H(z) = \frac{b_0 + b_1z^{-1} + b_2z^{-2}}{1 + a_1z^{-1} + a_2z^{-2}}$$
- This is derived from the following difference equation ...

$$y[kT] = b_0x[kT] + b_1x[(k-1)T] + b_2x[(k-2)T] - a_1y[(k-1)T] - a_2y[(k-2)T]$$

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## ‘Biquad’ Filters

**Input**  $x[nT]$  **Output**  $y[nT]$


$$y[kT] = b_0x[kT] + b_1x[(k-1)T] + b_2x[(k-2)T] - a_1y[(k-1)T] - a_2y[(k-2)T]$$

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## The Biquad Filter in Pd



- Pd provides a raw biquadratic filter object **[biquad~]**
- The five coefficients ( $b_0, b_1, b_2, a_1, a_2$ ) can be sent as a single message
- For a bandpass filter, the coefficients can be calculated using the **[bandpass]** object
- Pd provides the following objects for calculating filter coefficients for [biquad~] ...
  - [lowpass] [highpass]
  - [bandpass] [notch]
  - [lowshelf] [highshelf] [hlshelf]
  - [equalizer]



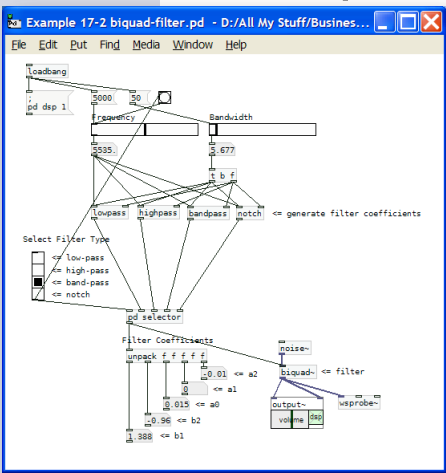
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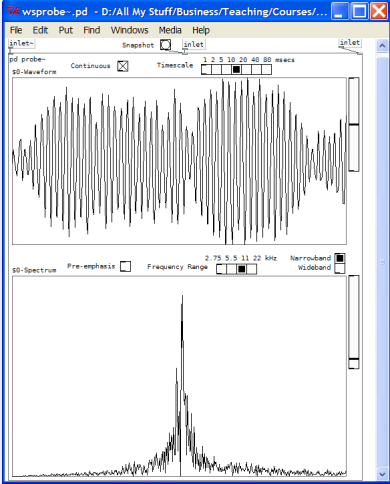
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## ‘Biquad’ Filters







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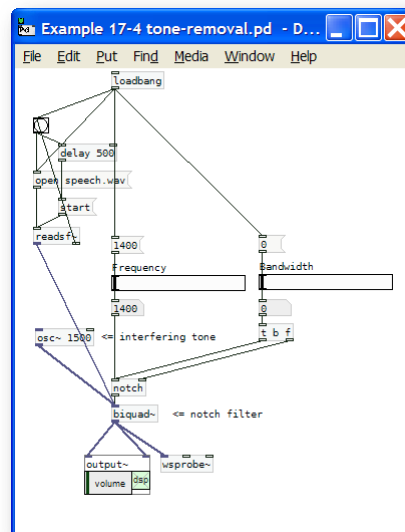


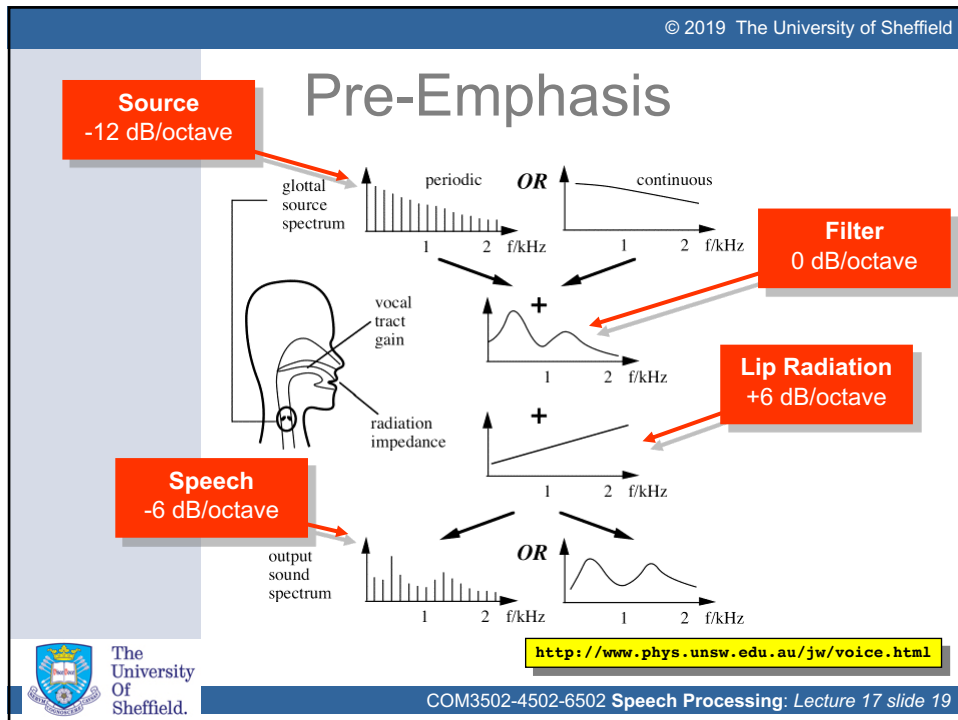
## Filters in Speech Processing

- General filtering
  - notch filtering  
(to remove interference and noise)
  - pre-emphasis  
(to equalise the frequency response)
- Modelling speech production
  - the source-filter model (*Lecture 3*)
- Modelling the auditory system
  - filter-bank analysis (*Lecture 4*)



## Tone Removal by Notch Filtering





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## Pre-Emphasis

- Averaged over time, speech has an overall spectral slope of  $\sim -6$  dB/octave
- This means that there can be a large amplitude difference between the lowest ( $\sim 50$  Hz) and highest ( $\sim 8$  kHz) frequency components
- E.g. over 8 octaves there will be a  $(6 \times 8)$  48 dB difference (which is comparable to the dynamic range from the quietest to the loudest speech sounds)
- This can make processing difficult (especially for functions such as spectral peak picking, i.e. formant tracking)
- It is therefore usual to 'pre-emphasise' a speech signal by giving it a +6 dB/octave lift
- In Pd this can be achieved by placing a 'real zero' near the origin, e.g. `[rzero~ 1]`

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## 'Gammatone' Filter

$$g[t] = ak^{n-1}e^{-2\pi bt} \cos(2\pi\omega t + \phi)$$

Slaney, M. (1993). *An efficient implementation of the Patterson-Holdsworth auditory filter bank: Apple Computer.*


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## The Gammatone in Pure Data



- Daniel Pressnitzer and Dan Gnaniasia have implemented a gammatone filterbank in Pd: `[gammabank_dm~]`
- It is part of an 'Audition Library' which contains other useful functions, such as ...
  - gammatone re-synthesis
  - multiplexers
  - de-multiplexers
  - outer and middle ear filters
- The code is based on:
  - Hohmann, V. (2002). Frequency analysis and synthesis using a Gammatone filterbank. *Acta Acustica*, 88, 433-442.
- The Audition toolkit is *not* available on the University machines, but is downloadable (*for Windows*) from: <http://lumiere.ens.fr/Audition/tools/realtime/>

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# Any Questions ?



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# Next time ...

## Linear Prediction

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