
31606 Signals and Linear Systems in discrete time

Hands on 9
Group 1

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November 13, 2014

How finite should it be?

i) Like a hot knife through a Butterworth We are asked to design a filter meeting the following requirements:

- Passband from 0.2 to 0.3
- Stopbands from 0 to 0.1 and 0.4 to 1
- Ripples in passband no greater than 2dB
- Stopband attenuation min. 100dB

While the headline does imply that a butterworth filter could be used, this does not allow us to utilize that ripples in both the stop and passband are accepted.

We use the *Matlab* filter toolbox to design our filter:

```
% Construct an FDESIGN object and call its ELLIP method.  
h = fdesign.bandpass(Fstop1, Fpass1, Fpass2, Fstop2, Astop1, Apass, ...  
                  Astop2);  
Hd = design(h, 'ellip', 'MatchExactly', match);
```

This gives us a 12th order filter, compared to that a buttersworth filter would have to be of order 24 to meet the same specifications.

Pole-Zero plot of the filter can be seen in fig. 1, frequency response (and specification boundaries) can be found in fig. 2 and impulse response in fig. 3

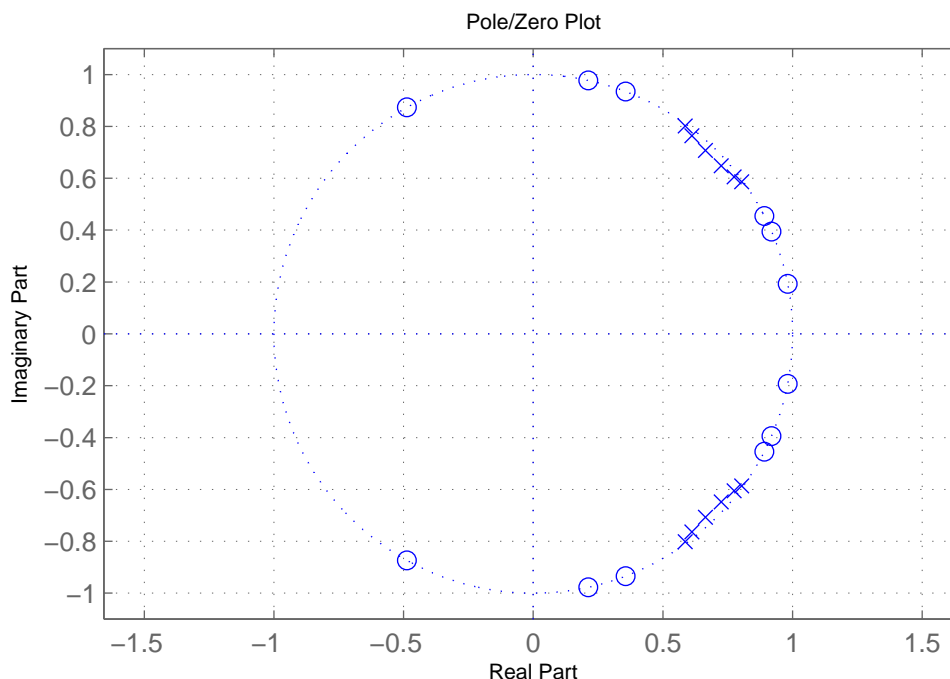


Figure 1: Pole-zero map of a 12th order elliptic bandpass filter with $-3dB$ cutoff frequencies at $0.3004\pi rad/sample$ and $1.997\pi rad/sample$

We find that point i in the impulse response where the maximum amplitude is below a value of 10% of the maximum amplitude of the whole signal so that:

$$x[0 : inf] * 10\% > x[i : inf]$$

This is done with the *Matlab* code:

```
%make list of max of remaining response:
maxrest = uncut_ir;
for i = 1:length(maxrest)
    maxrest(i) = max( abs( uncut_ir(i:end) ) );
    if maxrest(i) < maxval*sig
        break; %stop here, i is now the index, where all samples [i:inf] < max
    end
end
```

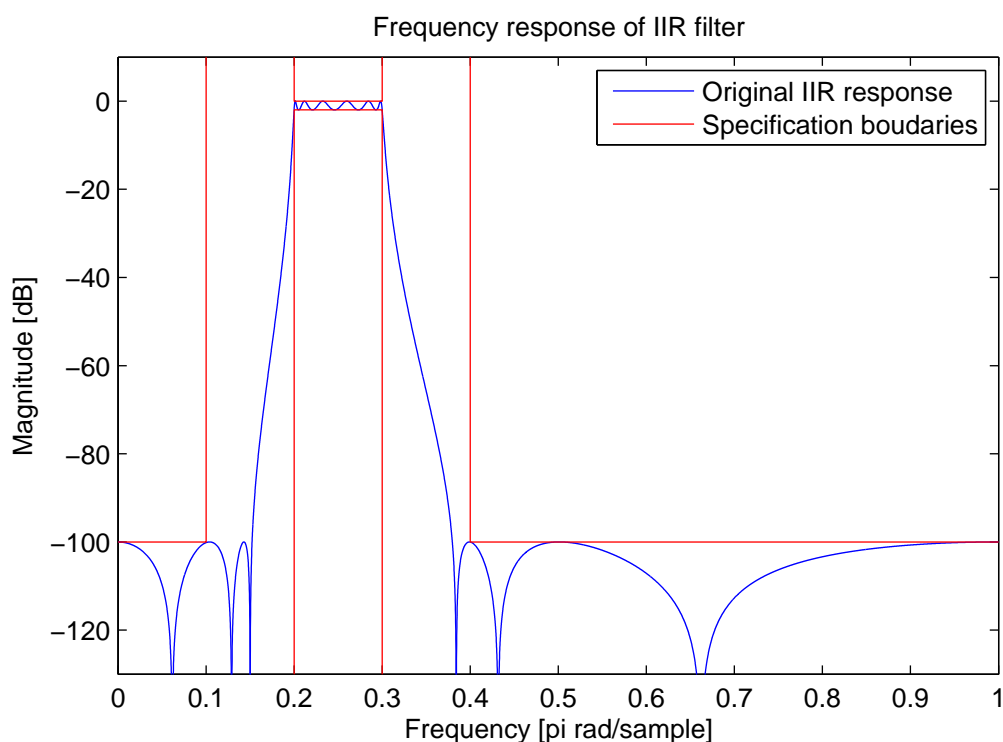


Figure 2: Frequency response of a 12th order elliptic bandpass filter with -3dB cutoff frequencies at $0.3004\pi\text{rad/sample}$ and $1.997\pi\text{rad/sample}$

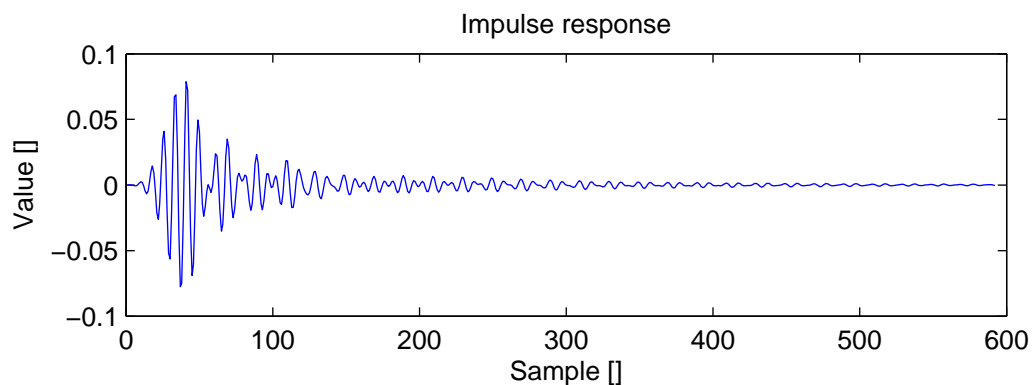


Figure 3: Impulse response of a 12th order elliptic bandpass filter with -3dB cutoff frequencies at $0.3004\pi\text{rad/sample}$ and $1.997\pi\text{rad/sample}$

This gives an “effective” length of 134 samples. (Had we chosen a buttersworth filter instead this would have been 146 samples).

We now use a rectangular window to cut the impulseresponse at 100%, 75%, 60%, 40% and 10% of the original. The frequency response of the filter with these impulseresponses as kernes are shown in fig. 4

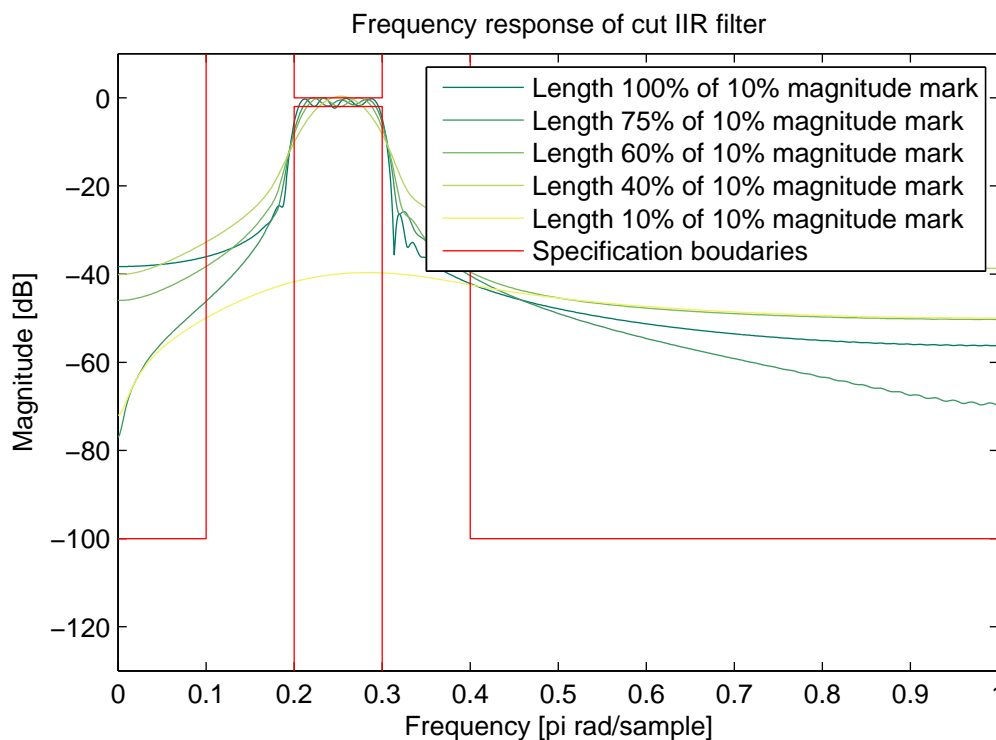


Figure 4: Frequency response of an FIR approximation of an IIR filter using a rectangular window, where the impulse response has been cut at different lengths. Notice how the ripple in the passband stays almost constant at 2dB , while the attenuation in the stopband is drastically reduced compared to the original IIR filter.

It is quite clear that neither of the filters meet the requirements. As the impulse length is shortened there, surprisingly, is not that big of a difference in the filter, until somewhere between 40% and 10% of the length.

Using a hann window, however, we can manipulate the filter responses to get much closer to the requirements - all without changing the length of the filter. See fig. 5. When applying the window it is important to notice, that the impulse response is already causal and $x[i] = 0$ for $i < 0$, so we only have to apply the half of the window on the right of the magnitude axis.

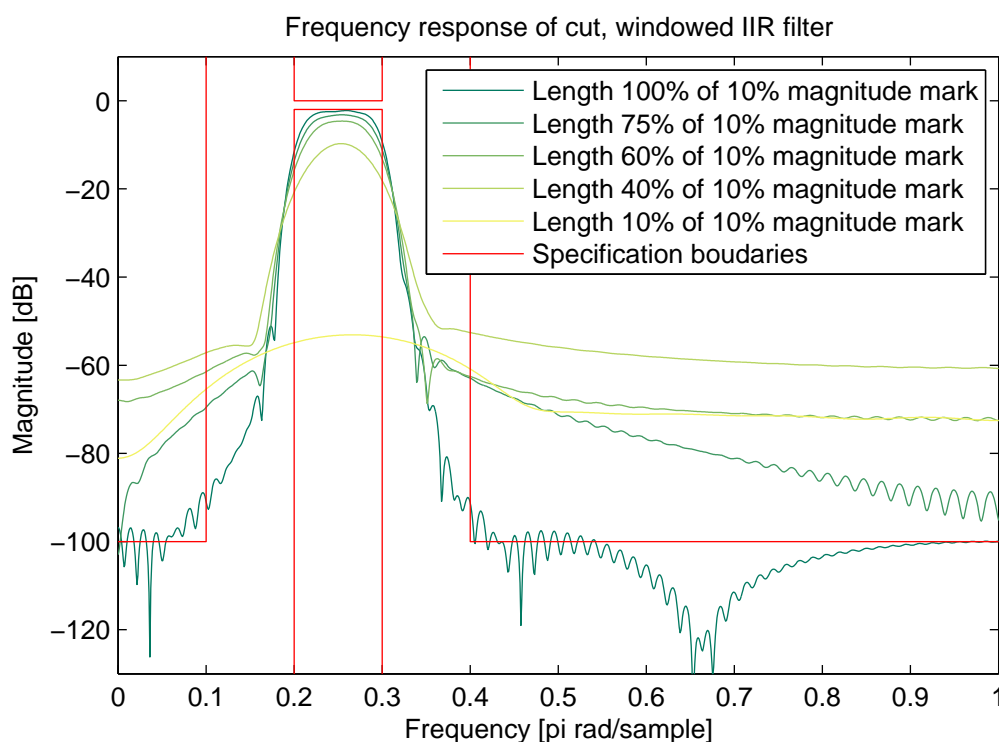


Figure 5: Frequency response of an FIR approximation of an IIR filter using a hann window, where the impulse response has been cut at different lengths. Notice how the passband ripple has been smeared out by convolution in the frequency domain, this has also made the corners of the filter rounder, resulting in the cutoff frequencies to move inside the desired passband. On the other hand stopband attenuation has been increased by a huge factor.

ii) 1.2 Shorter and shorter We are tasked with creating a filter for which the specification has been partly destroyed by a coffee addict. Luckily, the surviving information is sufficient to create the original information.

The filter should pass all frequencies between 0 and 5kHz with a gain on 0dB, this means the filter is a lowpass filter. As we can see from the sketched frequency response, 5kHz corresponds to one sixth of the sampling frequency since a normalised frequency of 1 corresponds to the nyquist frequency. The sampling frequency is 30kHz. Additionally the impulse response of the filter should be 601 samples long, meaning our filter is of the 600th order as it is a FIR filter where the length of the impulse response is $N + 1$ where N is the order of the filter.

The filter can be constructed in Matlab using the frequency domain equivalence method. This means we construct a perfect lowpass filter in the frequency domain, and sample it at the frequency bins given by our signal length. Our signal is 601 samples long, which at 30kHz yields a period of 0.02 seconds, meaning that bins in the frequency domain are separated by 50Hz. After sampling our perfect lowpass filter, we transform back to the time domain and delay by $\frac{N-1}{2}$ samples, obtaining the impulse response shown in figure 6 which is, not surprisingly, a time-shifted sinc function.

Since we want to please our boss, we decide to try minimizing the resources required to implement the filter. We do this by reducing the length of the impulse response and applying an appropriate window to reduce high-frequency components. We choose to use the hamming filter, applied to impulse responses of varying lengths shown in figure ?? . In the end we conclude that the filter with an impulse response length of 201 samples is actually better when windowed than the original impulse response in terms of rolloff without sacrificing any significant gain in the passband.

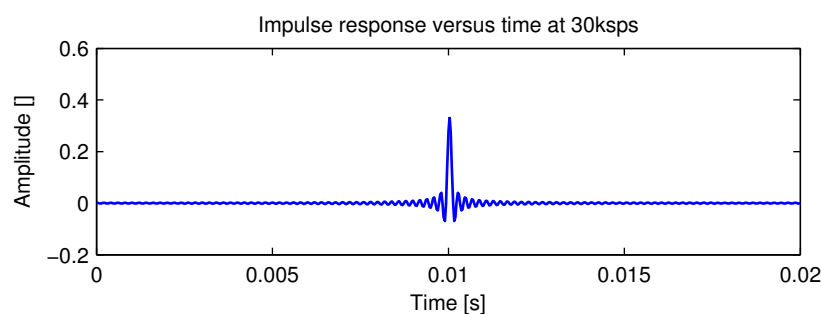


Figure 6: Impulse response of the 600th order FIR filter created by sampling a perfect low-pass filter in the frequency domain and transforming back to the time domain. It is a time-shifted sampled sinc function.

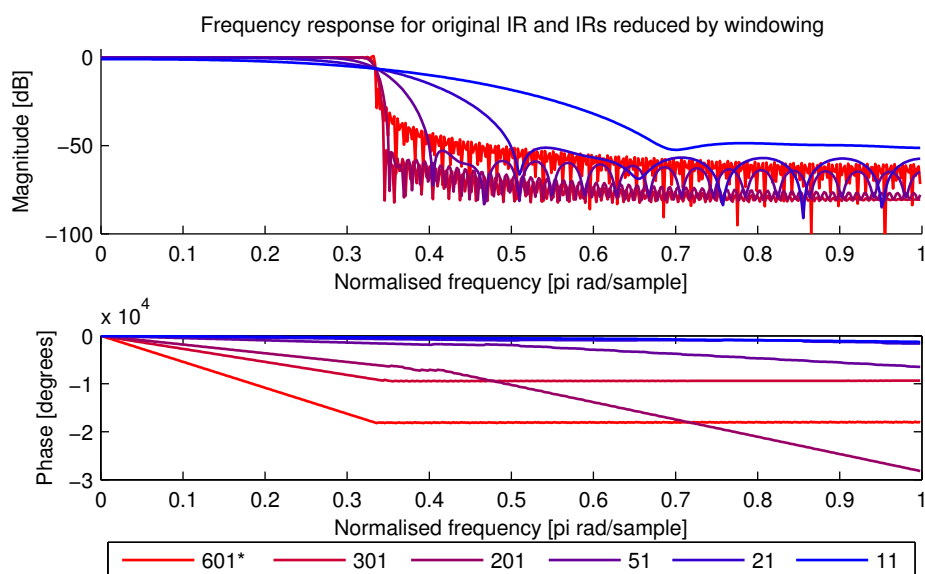


Figure 7: Frequency response of the *original impulse response as well as the impulse responses reduced in length by windowing with a hamming window. Legend shows the length of the impulse response in samples. It is clear that the windowed impulse responses have better maximum attenuation all the way to down to 11 samples, however the rolloff is worse for impulse responses shorter than 300 samples. Beyond 51 samples the filter behaviour is completely unsatisfactory for our purposes. 201 would be the best compromise between performance and cost. Note that the phase is linear until the cutoff frequency after which the sawtooth pattern is caused by the inability to represent small changes with floating point numbers at those magnitudes.

||| From bass/treble to an equalizer

The task is to implement a tunable FIR equalizer with five equally spaced (linear scale) bands. Using the frequency domain equivalence method the FIR filter coefficients can be determined by defining the magnitude response, transforming to the time-domain, windowing and translating to obtain a causal response. A function that constructs the filter coefficients was implemented where filter order and window type are parameters. The code is shown in Figure 8.

```
function [h] = equalizer5band(G1, G2, G3, G4, G5, n, W);
2 %INPUT: Band gains: G1, G2, G3, G4, G5. Approximate length: n. Window: W.
  %OUTPUT: Impulse response/FIR filter coefficients: h.
4 m = round(n/10);
  H = [G1*ones(1, m) , G2*ones(1, m) , G3*ones(1, m) , G4*ones(1, m) , ...
6       G5*ones(1, m) , G5*ones(1, m) , G4*ones(1, m) , G3*ones(1, m) , ...
       G2*ones(1, m) , G1*ones(1, m)];
8 h = ifftshift(fft(H, 'symmetric'));
  L = length(h);
10 if W == 0 %rectangular window (technically unnecessary code)
    h = h; %do nothing
12 elseif W == 1
    h = h.*hanning(L)';
14 elseif W == 2
    h = h.*hamming(L)';
16 end
end %eof
```

Figure 8: The equalizer5band function used to generate the FIR filter coefficients.

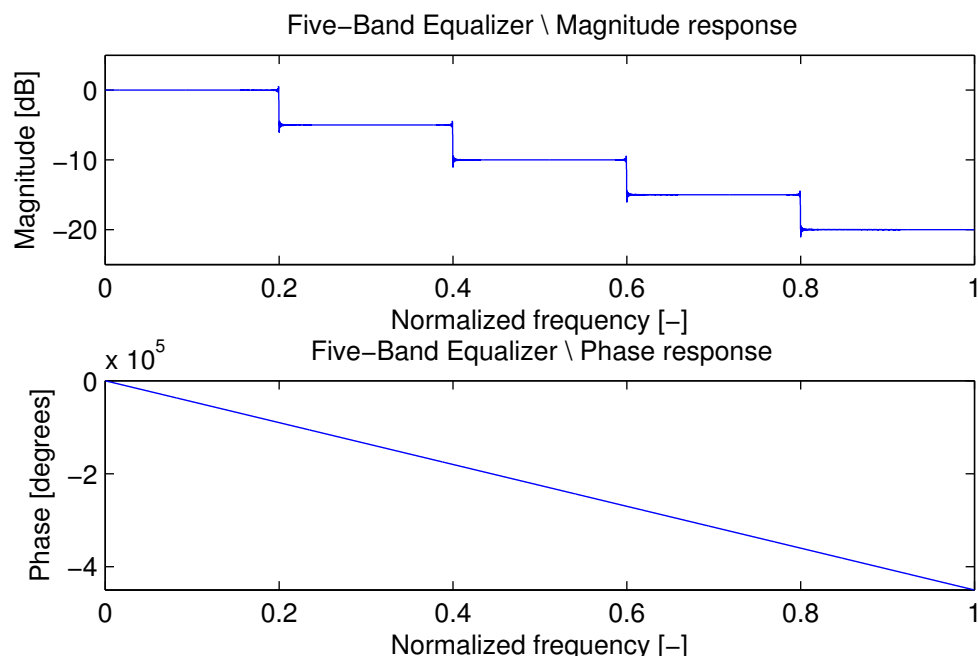


Figure 9: Magnitude and phase response of the transfer function of a five-band FIR equalizer with band gains decreasing in steps of 5 dB from the lowest to the highest normalized frequency. Ringing at the edges is caused by the use of a rectangular window in this case. See Figure 8. The filter order in this case is 4999 (IR length 5000).

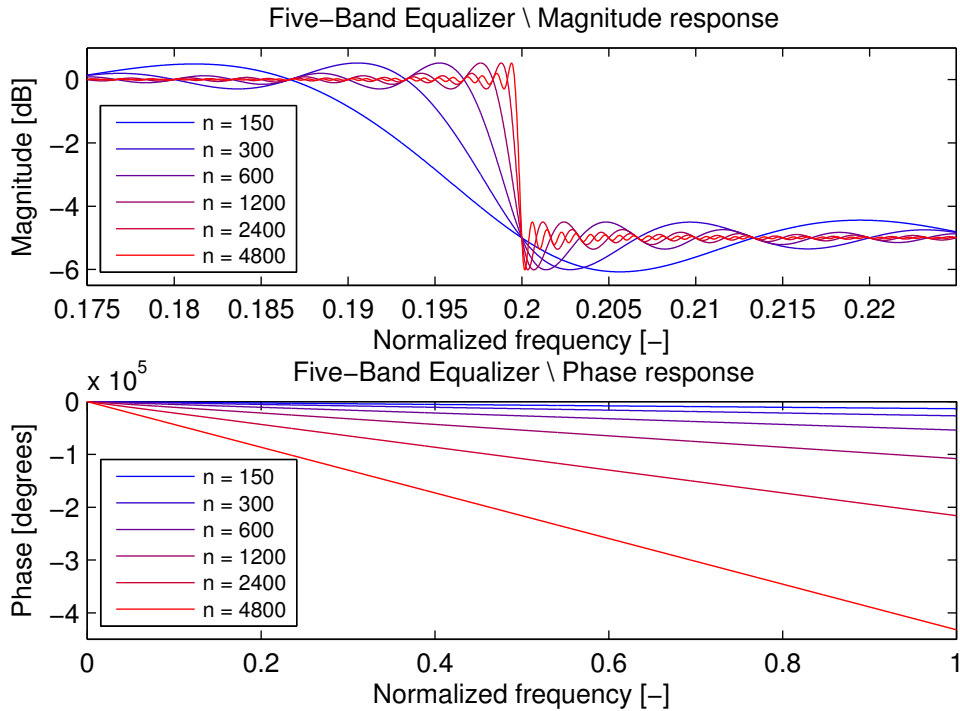


Figure 10: Equalizer frequency response (rectangular window function) for several IR lengths n . Ringing caused by the discontinuous rectangular window is clear and increases with the order. The phase response is linear, since the filter type is FIR and had zero-phase before translation.

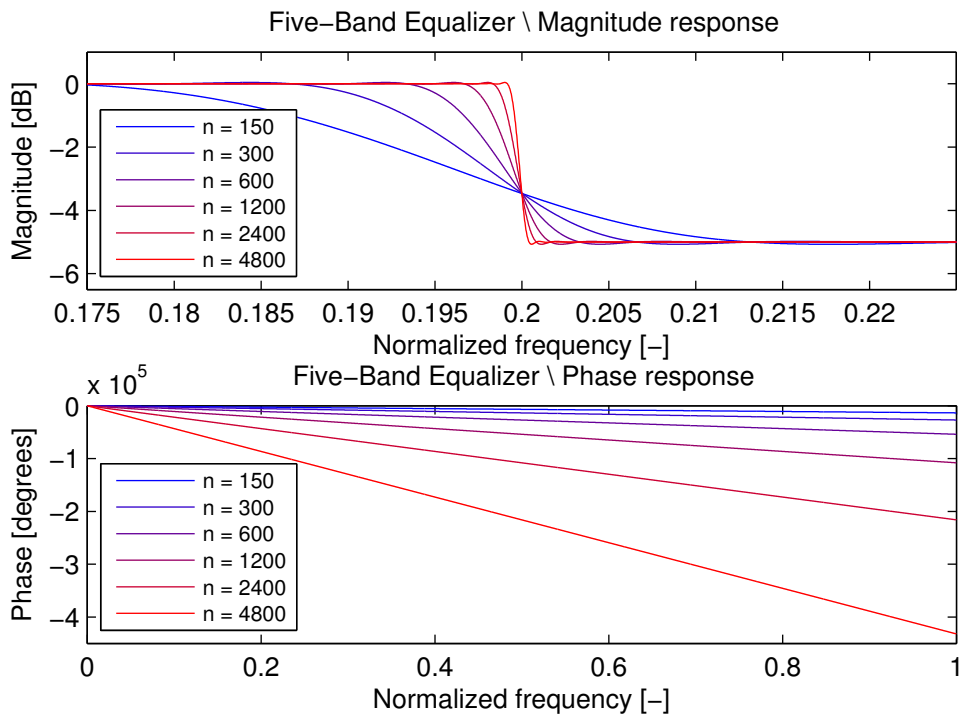


Figure 11: Equalizer frequency response (using Hann window function) for several IR lengths n . Compare with Figure 10 to see the clear improvement in smoothness of the magnitude response. Note also that the slope of the phase response is proportional to the IR length (also Figure 10). This is also as expected since a time translation by τ corresponds to a phase shift proportional to τ .

Higher order results in a greater phase slope and therefore more initial delay (besides the cost), but the approximation is much better. The relations between order, phase and approximation are illustrated in Figure 10 (rectangular window function) and Figure 11 (Hann window function). A filter order of 499 (corresponding to IR length $n = 500$) may be chosen for practical purposes (e.g. equalizing music), since an order of this size gives a somewhat reasonable approximation without too many samples (less processing time, less memory usage and less delay as benefits). The magnitude response in Figure 11 show that this choice is middle ground.

Zero-padding the IR increases samples in the time domain and hence in the frequency domain. The spectral resolution increases with the number of samples. By comparing the resulting plots this effect was observed in the magnitude response (but not in the phase response however). Choosing a smoother window gives a less pronounced effect (since there is less ringing).

The result of processing white noise using the equalizer settings in Figure 9 is shown in Figure 12. As expected, a stepwise descent corresponding to the magnitude response can be clearly seen.

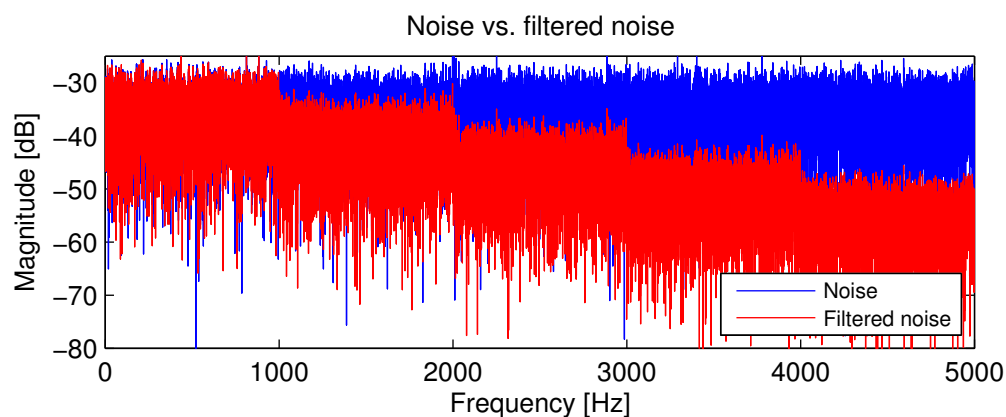


Figure 12: White noise signal filtered using the frequency response from Figure 9 (order 4999). The expected descent by 5 dB for each step is clearly visible in the output spectrum.

The result of processing `piano.wav` using the equalizer settings in Figure 9 is shown in Figure 13. A stepwise descent is no longer clear, but attenuation appears to increase steadily with frequency.

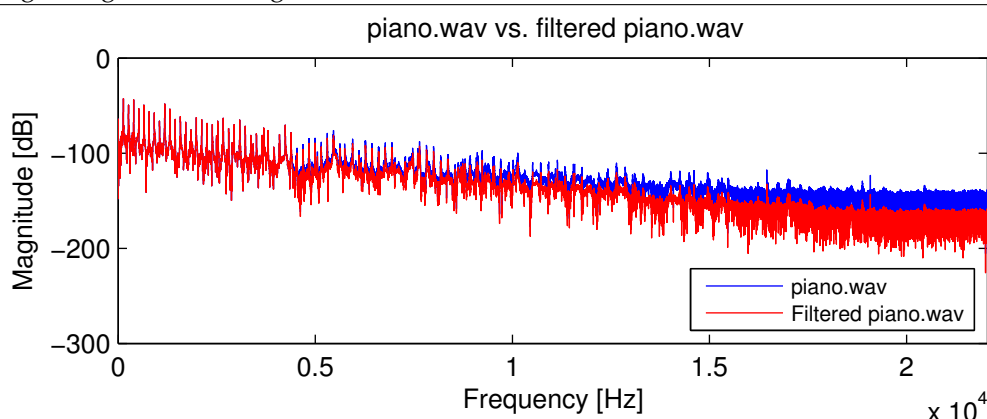


Figure 13: Sound `piano.wav` filtered using the frequency response from Figure 9 (order 4999). Attenuation increases steadily in accordance with the magnitude response of the applied filter.

||| The grandparents and DSP

i) Granma rolling with Maadlab My bored grandmother fires up *Matlab* and generates a linear sweep from 50Hz to 5kHz with a sampling frequency of 5kHz.

i,a What I expect to hear Assuming *Matlab* and my soundcard likes eachother and can up-sample and filter away any frequencies above half the sampling frequency, I expect to hear a cosine sweep from 50Hz to 2500Hz and then (with opposite phase) back again from 2500Hz to 0Hz. This sweep has been illustrated in fig. 14.

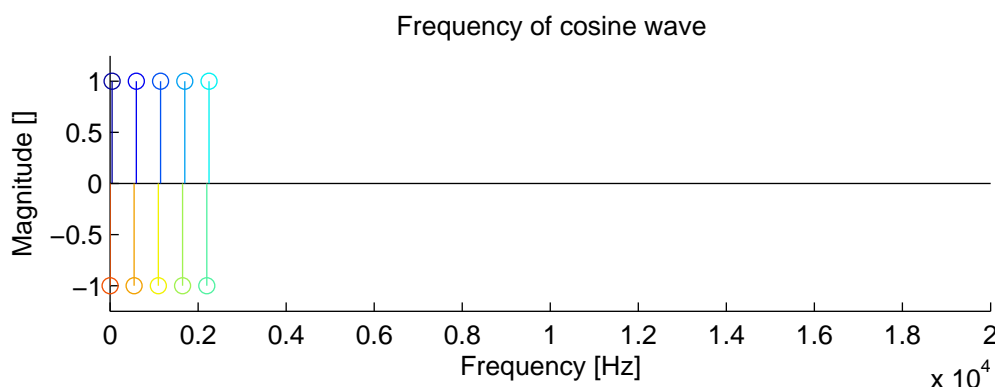


Figure 14: Linear sweep from 50Hz to 5kHz with a sampling frequency of 5kHz an brickwall filter at 2.5kHz. Each color represents the frequency response at a different time, starting with blue and mooving to red.

Now I know that is probably not the case (by listening this is quite easy to check). Neither *Matlab* or my soundcard has a brickwall lowpass filter at 2500Hz, so what I hear is actually multipe sweeps moving up and down. The audible range of the first 25% of the sweep is illustrated in fig 15. Or with a zero-order hold *DAC* (*digital to analog converter*) : 16.

i,a What my parrents expected was of course to actually hear a sweep from 50Hz to 5kHz, because they do not know about the nyquist theorem and frequency folding. If I was to explain this to them, I'd probably tell them it's like that effect when you photograph a checkered shirt and get those wavy patterns instead.

i,a Explanation The reason for this frequency folding is sampling in the time domain, which leads to preodicity in the frequency domain. This leads artefacts if care is not taken to ensure

that all frequencies stay inside their “period” in the frequency spectrum, but it also leads to artefacts if one is not careful when outputting the signal through ex. a DAC.

ii) Grandpa returns from the dead The filter is an order 3 running average. Taking the z-transform of the impulse response gives the frequency response

$$H[z] = \frac{1}{3}(1 + z^{-1} + z^{-2})$$

and on the unit circle $z = e^{it}$ we have

$$H[e^{it}] = \frac{1}{3}(1 + e^{-it} + e^{-2it}) = \frac{1}{3}e^{-it}(2\cos(t) + 1)$$

Therefore the magnitude and phase response of this filter is $|H[e^{it}]| = \frac{1}{3}(2\cos(t) + 1)$ and $\text{Arg}(H[e^{it}]) = e^{-it}$ (linear phase). Provided that the impulse response is allowed to be non-causal, then

$$H[z] = \frac{1}{3}(z + 1 + z^{-1})$$

and on the unit circle $z = e^{it}$ we have

$$H[e^{it}] = \frac{1}{3}(1 + e^{-it} + e^{it}) = \frac{1}{3}(2\cos(t) + 1)$$

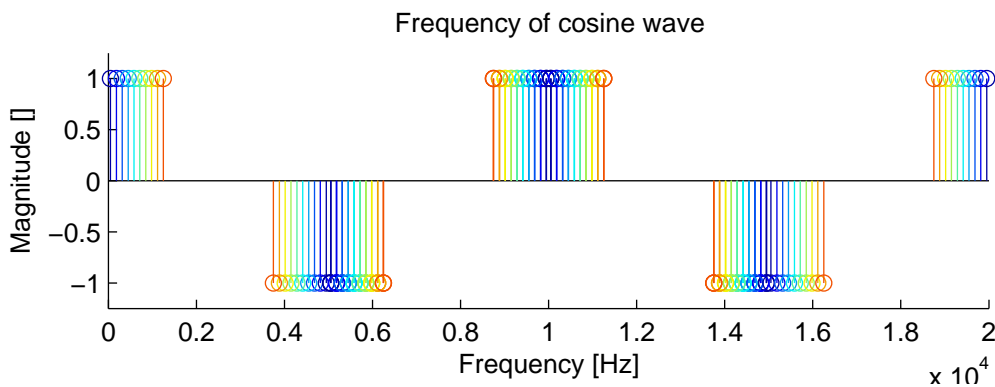


Figure 15: Linear sweep from 50Hz to 25% · 5kHz with a sampling frequency of 5kHz. Each color represents the frequency response at a different time, starting with blue and moving to red.

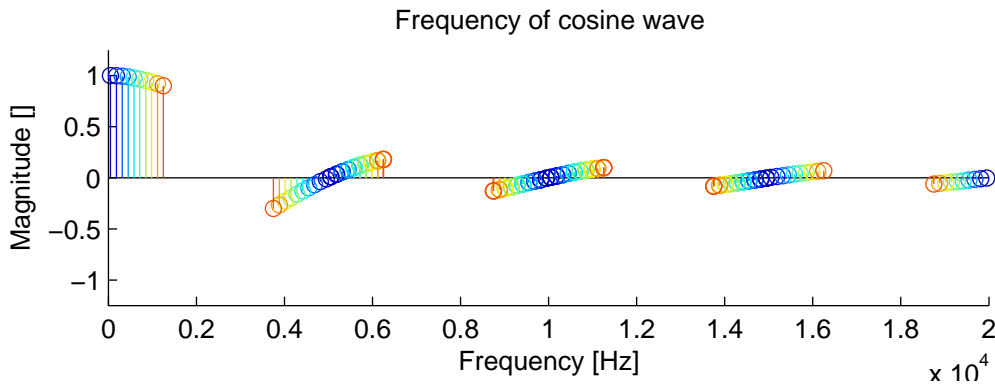


Figure 16: Linear sweep from 50Hz to 25% · 5kHz with a sampling frequency of 5kHz run through a zero-order hold DAC. Each color represents the frequency response at a different time, starting with blue and moving to red.

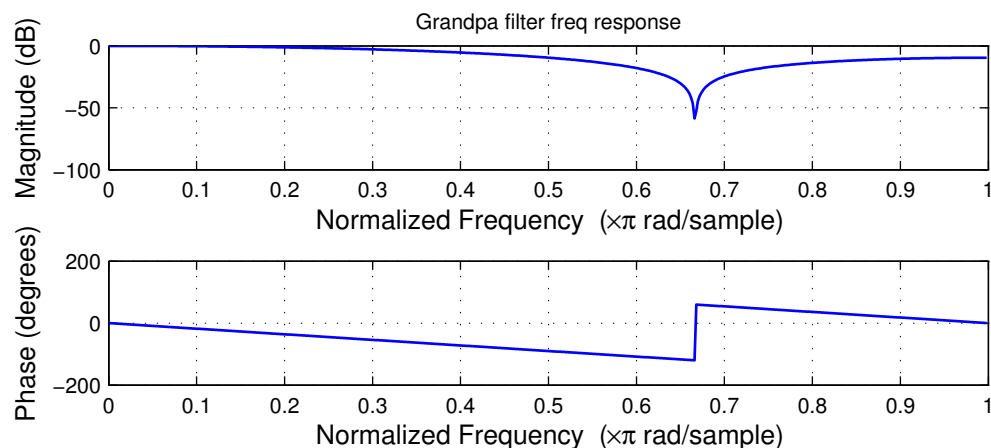


Figure 17: Grandpa's smoothing filter. The suppressed frequency is seen at $2/3$ because the plot shows normalised frequency over π . It is also clear that the phase is linear apart from unwrapping artefacts.

Therefore the magnitude and phase response of this filter is $|H[e^{it}]| = \frac{1}{3}(2 \cos(t) + 1)$ and $\text{Arg}(H[e^{it}]) = 0$ (zero phase). The normalized frequencies which are completely suppressed corresponds to the zeroes $|H[e^{it}]| = \frac{1}{3}(2 \cos(t) + 1) = 0$ for $t \in [0, \pi]$ giving one normalized frequency $t = \frac{2}{3}\pi$. The actual suppressed frequency is then $\frac{2}{3} \frac{f_s}{2} = \frac{1}{3}f_s$ where f_s is the sampling frequency. If $T_s = 1$ s, then $f_s = \frac{1}{T_s} = 1$ s⁻¹ so the suppressed frequency is 1/3 Hz. Using normalized frequency allows for the frequency response to be expressed without knowing the sampling frequency.