Signal Processing Basics & Filtering

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https://gitlab.com/surligas/sdr-tutorial
https://gitlab.com/surligas/gr-tutorial

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Signals can be represented in the time domain

... or in the **frequency domain**



Time domain:

- X-axis represents the time
- Y-axis represents the amplitude (e.g Volts)

Frequency domain:

- X-axis represents the frequency
- Y-axis represents the power
- Commonly power is expressed in logarithmic scale $(10 \times log_{10}(x))$

Frequency domain:

 For the transition from time to frequency domain, the Discrete Fourier Transform (DFT) X_k is used

$$X_{k} = \sum_{n=0}^{N-1} x_{n} \cdot e^{-\frac{i2\pi}{N}kn}$$
$$= \sum_{n=0}^{N-1} x_{n} \cdot \left[\cos\left(\frac{2\pi}{N}kn\right) - i \cdot \sin\left(\frac{2\pi}{N}kn\right)\right]$$

- In simple words, the X_k tranformation, converts the original signal into N cosines, each one with a frequency $\frac{2\pi}{N}kn$ and amplitude x_n
- For real signals the imaginary part is 0

- Changing the signal in the time domain affects the frequency domain
- In most cases we use the frequency domain representation
- Different frequency components can be spotted very easily
- Signals can be spotted visually even with significant amount of noise

Use the **simple_example.grc** flowgraph located at the **examples** directory to play with time and frequency domain representation

Example 1: Finding the frequency components of a mixed signal



The two different sine waves can be spotted quite easily in the frequency domain comparing to the time domain

Example 2: Finding the frequency components of a noisy mixed signal



Even with the presence of significant noise, the signal components are distinguishable! (However this is not always the case)

Sampling & Quantization

Sampling is the process of converting an continuous-time signal into a discrete-time signal



Nyquist Theorem

A real signal x(t) band-limited to B Hz can be reconstructed without loss of information iff the sampling frequency is greater than $2 \times B$

- A band-limited signal *x*(*t*), is a signal that does not have spectral components above *B* Hz
- The minimum sampling frequency $R = 2 \times B$ is called **Nyquist** frequency

Sampling Theory

- In most cases there is a need to change the sampling rate of a signal
 - Hardware capabilities / restrictions
 - Processing maybe easier at specific sampling rates
 - Processing resources
- Decimation is the process of reducing the sample rate
- Interpolation increases the sampling rate
- Combining decimation and interpolation an arbitrary sampling rate can be achieved
- More about resampling on the next lecture!

Quantization converts the signal amplitude into digital form samples



An approximation of the Signal-to-Quantization Noise Ratio (SQNR) for an N-bit ADC/DAC is: $SQNR = 20 \log_{10}(2^N) \approx 6.02 \cdot N dB$

Filtering

- Filters are used to separate mixed signals
- $\cdot\,$ Extract a specific band of interest
- Avoid undesired effects (aliasing, imaging e.t.c)
- Restore distorted signals
- Pulse shaping

Filter types

Analog Filters:

- Implemented using electronic components (resistors, capacitors, coils, transistors)
- Used in analog front-ends
- Standard performance
- Difficult to change their properties
- Application specific



Filter types

Digital Filters:

- Implemented in software
- Variable performance depending the processing resources
- Change their properties relatively easily
- Can adapt to any application



- Low Pass Filters
 - Frequencies above a threshold do not pass the filter



- High Pass Filters
 - Frequencies below a threshold do not pass the filter



- Band Pass Filters
 - Frequencies of a specific band pass the filter



- **Cutoff frequency:** The frequency above (low pass filter) or below (high pass filter) which, the filter starts to attenuate the signal
- **Transition width:** Control how steep is the attenuation of the signal above (low pass filter) or below (high pass filter) the cutoff frequency

Filter parameters



• Large transition width:



- $\cdot\,$ Transition width specifies the number of taps of the filter
- $\cdot\,$ Steep filters have more taps than relaxed ones
- Designing a filter is always a balance between RF performance and CPU utilization
- Sampling rate plays a key role too

Example

- Assume a sampling rate of 1 MSPS
- $\cdot\,$ We are interested in a range of only 10 kHz
- The requirements demand a very steep filter and we have limited CPU resources (e.g RPi3)
- How???

Example

- Assume a sampling rate of 1 MSPS
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- How???
- Use a relaxed first stage relaxed filter
- Decimate the signal by a factor of e.g 50
- Apply a second stage steep filtering

- Filters can be used also for pulse shaping
- The goal is to minimize spurious signals and increase the spectral efficiency
- Spurious signals can be generated by extreme transitions of the source signal
- A simple square wave generated from a bit stream has an infinite bandwidth →non practical to transmit

Pulse Shaping Filtering

- A simple square wave generated from a bit stream has an infinite bandwidth →non practical to transmit
- A square wave is an infinite sum of sine waves





Pulse Shaping Filtering

- Commonly for shaping we use the Root Raised Cosine filter (RRC)
- Smooths the extreme transitions thus reducing the spurious

