Speech signal processing using MATLAB Basics and applications

Stephan Kuberski (kuberski@uni-potsdam.de)

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Slides and MATLAB scripts at https://github.com/murtex/spl

Outline

Digital signals
Sampling
Time domain
Frequency domain
Filters
Noise

Acoustic signals
Short-time analysis
Spectrograms
Activity detection
Landmarks detection
Formants detection

Digital signals/Sampling

Sampling

► **continuous signal** (normalized magnitude, length *L* in seconds)

$$x(t) \in [-1, 1]$$
 with $t \in [0, L]$

```
>> x = Q(t) \sin(2*pi*f * t); % continuous sine with frequency f
```

 \triangleright sampling rate f_S , quantization of time

$$t \to t_i = \frac{i-1}{f_S}$$
 with $i \in \{1, ..., N\}$ and $N = \lfloor Lf_S \rfloor$

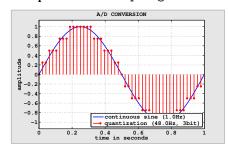
- >> N = floor(L * fS); % number of samples
 >> ti = (0:N-1) / fS; % quantized time values
- \triangleright bits per sample $n_{\rm S}$, quantization of amplitude

$$x(t) \to x_i = \frac{\lfloor 2^{n_s - 1} x(t_i) \rfloor}{2^{n_s - 1}}$$

>> xi = round(2^(nS-1) * x(ti)) / 2^(nS-1); % quantized amplitudes

Sampling

example: matlab/sampling.m





- exercise:
 - verify from reconstruction that Nyquist frequency holds

$$f_{\rm Ny} = \frac{f_{\rm S}}{2}$$

 compare commonly used sampling standards (telephony, Audio-CD, professional audio equipment, ...) Digital signals/Time domain

Time domain

total energy, average power and root mean square

$$E = \sum_{i=1}^{N} x_i^2$$
, $P = \frac{1}{N} \sum_{i=1}^{N} x_i^2$ and $RMS = \sqrt{\frac{1}{N} \sum_{i=1}^{N} x_i^2}$

```
>> E = sum( xi .* xi ); % total energy
>> P = mean( xi .* xi ); % average power
>> RMS = sqrt( mean( xi .* xi ) ); % root mean square
```

▶ decibel full scale, different for power- and magnitude-like quantities, e. g.

$$P_{\rm dB} = 10 \log_{10}(P)$$
 and $RMS_{\rm dB} = 20 \log_{10}(RMS)$

```
>> PdB = 10 * log10( P ); % power-like
>> RMSdB = 20 * log10( RMS ); % magnitude-like
```

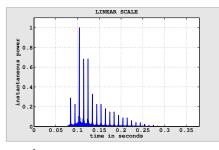
zero-crossings rate

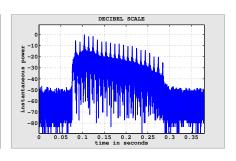
```
>> fZ = sum( abs( diff( xi >= 0 ) ) ) / N * fS;
```

Time domain

example (playback): matlab/decibel.m

$$P_i = x_i^2$$
 and $P_{i,dB} = 10 \log_{10}(x_i^2)$





- exercise:
 - ► TODO!

Digital signals/Frequency domain

Frequency domain

▶ discrete Fourier transform, time domain → frequency domain

$$X_k = \sum_{i=1}^N x_i e^{-2\pi i \frac{(i-1)(k-1)}{N}} \in \mathbb{C} \quad \text{with} \quad k \in \{1, \dots, N\}$$

>> Xk = fft(xi) / N; % complex Fourier coefficients

▶ *k* is a **frequency index** (as *i* was a time index)

$$k \to f_k = \frac{k-1}{N} f_{\rm S}$$

>> fk = (0:N-1) / N * fS; % frequency values

▶ frequencies beyond Nyquist frequency are negative frequencies

$$f_k \to f_k = \begin{cases} f_k - f_{\rm S} & \text{if } f_k > f_{\rm Ny} \\ f_k & \text{otherwise} \end{cases}$$

>> fk(fk > fNy) = fk(fk > fNy) - fS; % imply negative frequencies

Frequency domain

power spectral density (also known as power spectrum)

$$P_k = |X_k|^2 \in \mathbb{R} \quad \Leftarrow \quad \sum_{k=1}^N P_k = P$$

- >> Pk = abs(Xk) .^ 2; % power spectral density
- ▶ real valued signals $(x_i \in \mathbb{R})$ imply a special symmetry

$$X_{f_k} = X_{-f_k}^* \quad \Rightarrow \quad P_{f_k} = P_{-f_k}$$

restrict to one-sided spectrum

```
>> Pk(fk < 0) = []; % remove negative frequency components
>> Xk(fk < 0) = [];
>> fk(fk < 0) = [];
>> Pk(2:end) = 2 * Pk(2:end); % rescale to match total power
>> Xk(2:end) = sqrt( 2 ) * Xk(2:end);
```

 $ightharpoonup P_1$ is DC offset, $P_{k>1}$ are contributions of sines with frequencies $f_{k>1}$

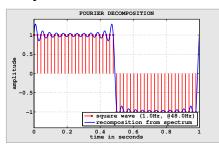
$$x(t) = \sqrt{P_1} + \sqrt{2} \sum_{k>1} \sqrt{P_k} \sin(2\pi f_k t)$$

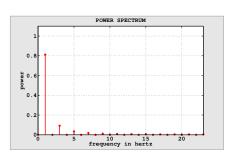
Frequency domain

complex valued but without loss of phase information

$$x(t) = X_1 + \sqrt{2} \sum_{k>1} X_k e^{2\pi i f_k t}$$

example: matlab/fdomain.m

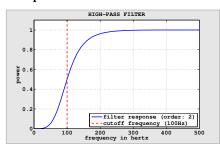


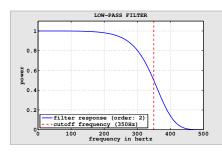


- exercise:
 - is there any loss of information by changing to real values?
 - ► TODO!

Digital signals/Filters

- general filter types:
 - ► low-pass: passes low frequencies (cuts high ones)
 - high-pass: passes high frequencies (cuts low ones)
 - **band-pass**: passes a range of frequencies (combination of low- and high-pass)
 - **band-stop** (notch): cuts a range of frequencies (opposite of band-pass)
- ▶ **cutoff frequency** at which output power is (generally) reduced by -3 dB
- example: matlab/filters.m

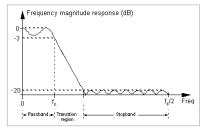


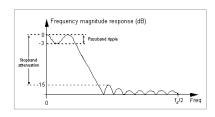


- filters are represented by **filter coefficients** b_i (feedforward) and a_i (feedback)
- ▶ high **filter order** *m* increases computational complexity but thereby quality

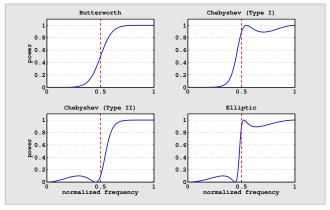
$$y_i = \underbrace{\frac{1}{a_1} \left(\sum_{j=0}^m b_{j+1} x_{i-j} - \sum_{j=1}^m a_{j+1} y_{i-j} \right)}_{\text{FIR}} \quad \text{with} \quad i \in \{1, \dots, N\}$$

- ► FIR filters (finite impulse response) are slow to compute but stable
- ► IIR filters (infinite impulse response) are fast to compute but might be unstable
- ▶ some often used additional terms (images from http://dspguru.com)





► many filter families with different characteristics (matlab/filters2.m), e.g.



normalized frequency

$$\tilde{f}_k = \frac{f_k}{f_{\text{Nv}}} = \frac{2f_k}{f_{\text{S}}} \in [0, 1] \text{ with } k \in \{1, \dots, N\}$$

► Butterworth filter (high-pass, second-order, 100 Hz cutoff)

```
>> m = 2; % filter order
>> cutoff = 100; % cutoff frequency
>> [b, a] = butter( m, cutoff / (fS/2), 'high' );
```

► Chebyshev filter (high-pass, 1 dB ripple, 40 dB attenuation, 100 Hz cutoff)

```
>> cutoff = 100; % cutoff frequency
>> stopband = 90; % stopband frequency
>> ripple = 1; % passband ripple
>> attenuation = 40; % stopband attenuation
>> m = cheb2ord( cutoff / (fS/2), stopband / (fS/2), ripple, attenuation );
>> [b, a] = cheby2( m, attenuation, stopband / (fS/2) );
```

apply any filter

```
>> y = filter( b, a, x ); % filter signal x using coefficients a, b
```

• or in zero-phase version (without filter delay)

```
>> y = filtfilt( b, a, x ); % zero-phase filtering
```

- exercise:
 - ► Can you image what filter delay means?
 - ► TODO: home audio/hifi, bass and treble

Digital signals/Noise

Noise

► spectral flatness measure

$$SFM = \dots$$

• spectral entropy using probability densities p_k and skipping constant DC

$$H = -\sum_{k>1} p_k \log(p_k) \quad \text{with} \quad p_k = \frac{P_k}{\sum_{k>1} P_k}$$

```
>> pk = Pk(2:end) / sum( Pk(2:end) ); % probability densities >> H = -sum( log2( pk .^ pk ) ); % entropy in bits
```

- additive noise
- ▶ signal-to-noise ratio
- spectral subtraction

Acoustic signals/Short-time analysis

Short-time analysis

Acoustic signals/Spectrograms

Spectrograms

Acoustic signals/Activity detection

Activity detection

Acoustic signals/Landmarks detection

Landmarks detection

Acoustic signals/Formants detection

Formants detection