Speech signal processing using MATLAB Basics and applications

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

Outline

Digital signals
Sampling
Time domain
Frequency domain
Filters

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 i 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 $N = \lfloor Lf_S \rfloor$

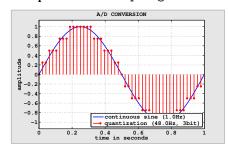
- >> N = floor(L * fS); % number of samples
 >> ti = (0:N-1) / fS; % quantized time values
- \triangleright bits per sample $n_{\rm S}$, quantization of amplitude

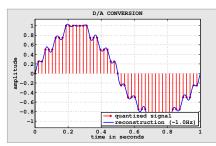
$$x(t) \to x_i = \frac{\lfloor 2^{n_s - 1} x(t_i) \rfloor}{2^{n_s - 1}}$$

>> 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 $RMS_{\rm dB} = 20 \log_{10}(RMS)$

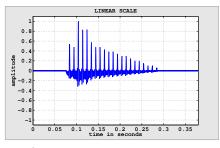
```
>> PdB = 10 * log10( P ); % power-like
>> RMSdB = 20 * log10( RMS ); % magnitude-like
```

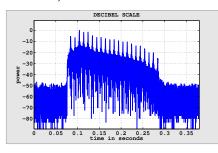
zero-crossings rate

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

Time domain

example: matlab/decibel.m (matlab/sound.wav)





exercise:

- compare linear and logarithmic scales
- ► explain **negative decibel values** (e. g. −3 dB power, −6 dB magnitude)
- specify the power of silence in decibels

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

 \blacktriangleright 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

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

>> 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 $f_{k>1}$

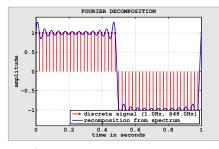
$$x(t) = \sqrt{P_1} + \sqrt{2} \sum_{k>1} \sqrt{P_k} \sin(2\pi f_k t)$$

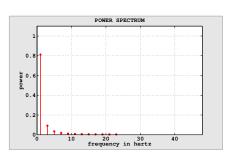
Frequency domain

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

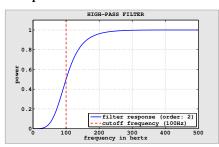


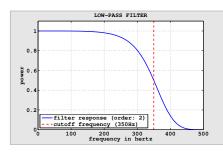


- exercise:
 - examine spectra of different wave forms (sines, square, sawtooth, ...)
 - examine spectral frequency range
 - verify loss of phase information in (real valued) power spectra

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

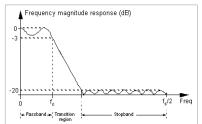


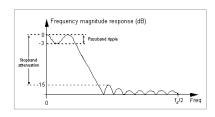


- filters are represented by **filter coefficients** b_i (feedforward) and a_i (feedback)
- ▶ high **filter order** *m* increases computational complexity but thereby quality

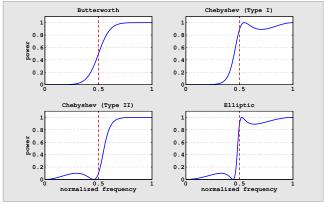
$$y_i = \underbrace{\frac{1}{a_1} \left(\sum_{j=0}^m b_{j+1} x_{i-j} - \sum_{j=1}^m a_{j+1} y_{i-j} \right)}_{\text{FIR}} \text{ with } i \in \{1, \dots, N\}$$

- ► 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)





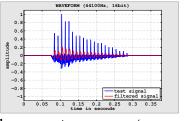
example: matlab/filters2.m

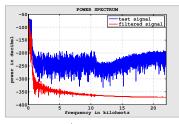


- ▶ many **filter families** with different characteristics
- normalized frequency

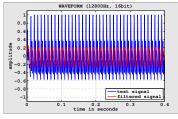
$$\tilde{f}_k = \frac{f_k}{f_{N_V}} = \frac{2f_k}{f_S} \in [0, 1] \text{ with } k \in \{1, ..., N\}$$

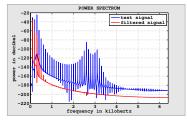
example: matlab/spectrum.m (matlab/sound.wav)





example: matlab/spectrum.m (matlab/ivowel.wav)





- exercise:
 - observe the occurrence of filter delay

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
```

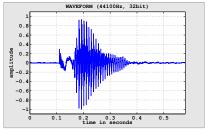
• or in zero-phase version (without filter delay)

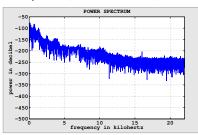
```
>> y = filtfilt( b, a, x ); % zero-phase filtering
```

Acoustic signals/Short-time analysis

Short-time analysis

- spectral and temporal analysis is essential for speech acoustics
- problem:
 - power spectrum has no temporal information anymore (matlab/tam.wav)

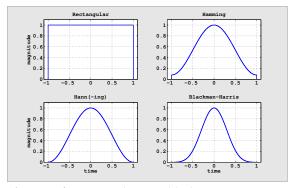




- ▶ solution:
 - choose short overlapping segments (windows) at different time points
 - ▶ length of the segments (window size) is crucial
 - overlap and window function control spectral leakage
 - ► aligning Fourier transforms of these (altered) segments leads to **spectrograms**

Short-time analysis

example: matlab/windows.m



• optimal overlapping for minimal spectral leakage

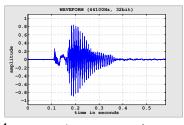
Rectangular: any value
Hamming: 50%
Hann(-ing): 50%
Blackman-Harris: 66.1%

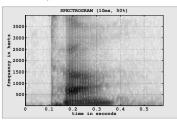
▶ other commonly used window functions: Welch, Kaiser, Gaussian, ...

Acoustic signals/Spectrograms

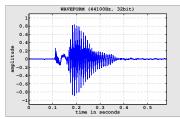
Spectrograms

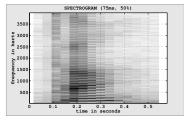
example: matlab/specgram.m (matlab/tam.wav)





example: matlab/specgram.m (matlab/tam.wav)





- exercise:
 - ► impact of window size → broad-band vs. narrow-band spectrogram

Spectrograms

- broad-band spectrograms have good temporal but poor spectral resolution
- narrow-band spectrograms have poor temporal but good spectral resolution

spectrogram: broad-band narrow-band window size: < 20 ms > 20 ms harmonics

set up windowing

```
>> wsize = 10; % window size in milliseconds
>> woverlap = 66; % window overlap in percent
>> wfunc = @blackmanharris; % window function
```

compute the spectrogram

```
>> [Xk, fk, ti] = spectrogram( xi, ... % signal
    wfunc( ceil( wsize/1000 * fS ) ), ... % window function values
    ceil( woverlap/100 * wsize/1000 * fS ), ... % window overlap samples
    4096, fS ); % fourier transform
```

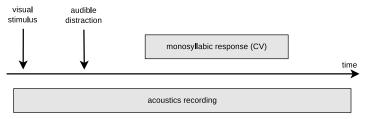
▶ plot the spectrogram

```
>> colormap( flipud( colormap( 'gray' ) ) ); % set color coding
>> imagesc( ti, fk, Pk .^ 0.1 ); % plot spectral powers
```

Acoustic signals/Activity detection

Activity detection

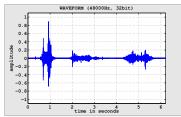
- experimental data often contain a lot of noise and little of information
- for automatic processing restriction to important parts is essential
- consider the following experiment:

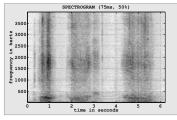


- with features of interest:
 - responded syllable (out of a specific set → classification task)
 - ► voice onset time (→ landmarks detection)
 - ► formants onsets (frequency and time → formants tracking/detection)
- ▶ all of these require (human) activity detection as an initial processing pass

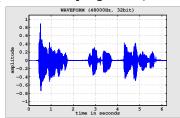
Activity detection

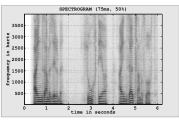
- ▶ in literature usually called **voice activity detection** (VAD)
- exploiting spectral differences in human speech and ambient sound/noise
- example: matlab/specgram.m (matlab/chair.wav)





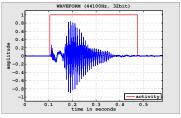
example: matlab/specgram.m (matlab/haiku.wav)

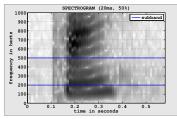




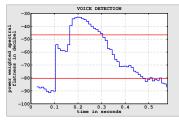
Activity detection

- example: matlab/activity.m (matlab/tam.wav)
- ▶ applying an equal loudness filter and limiting to frequency band 0 . . . 1000 Hz





adaptive thresholds for power-weighted spectral flatness and subband power





combining activity states based on thresholds gives overall voice activity

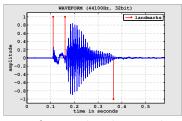
Activity detection/References

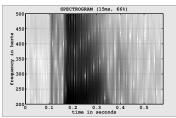
- ▶ D. Robinson. Equal loudness filter. 2001.
- ▶ D. Burileanu, L. Pascalin, C. Burileanu, M. Puchiu. An adaptive and fast speech detection algorithm. Springer, 2000.
- ▶ M. Prcin, L. Müller. Heuristic and statistical methods for speech/non-speech detector design. Springer, 2002.
- ► Y. Ma, A. Nishihara. Efficient voice activity detection algorithm using long-term spectral flatness measure. Springer, 2013.

Acoustic signals/Landmarks detection

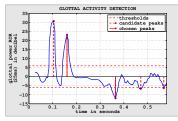
Landmarks detection

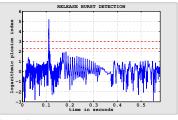
- example: matlab/landmarks.m (matlab/tam.wav)
- ▶ applying an **equal loudness filter** and limiting to **frequency band** 200 . . . 500 Hz





▶ power rate-of-rise and sophisticated peak detection, thresholding and pairing





computation of plosion indices to find release bursts

Landmarks detection/References

- ▶ D. Robinson. *Equal loudness filter.* 2001.
- ▶ P. Mermelstein. Automatic segmentation of speech into syllabic units. ASA, 1975.
- ▶ S. Liu. Landmark detection for distinctive feature-based speech recognition. MIT, 1995.
- ► T. Ananthapadmanabha, A. Prathosh, A. Ramakrishnan. Detection of the closure-burst transitions of stops and affricates in continuous speech using the plosion index. ASA, 2014

Acoustic signals/Formants detection

Formants detection