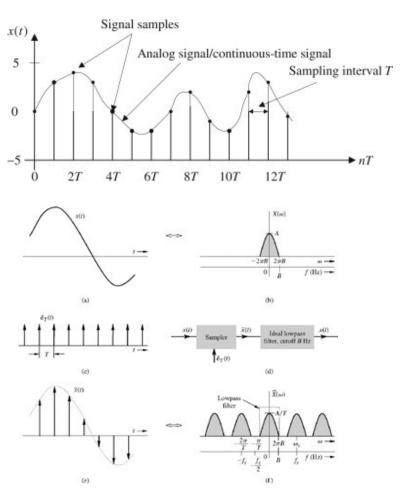
Introduction to Speech Signal

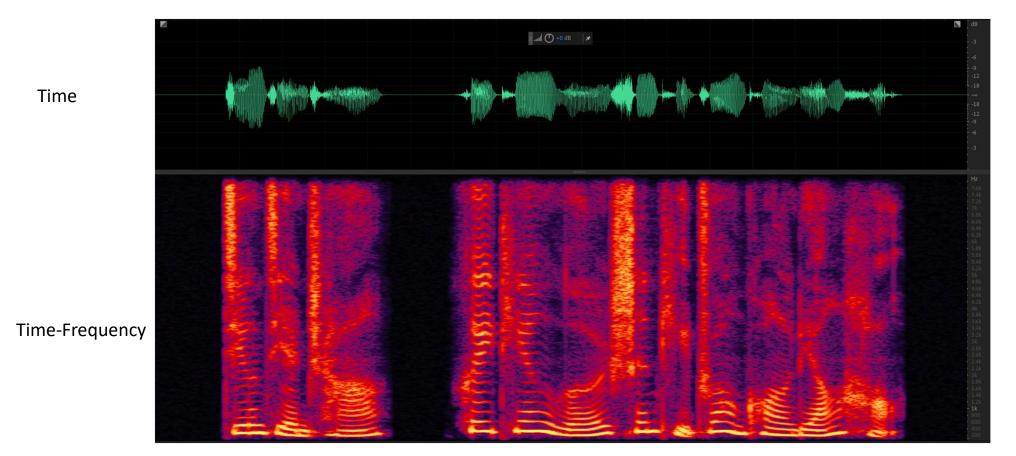
Meng Yu 08/29/2022

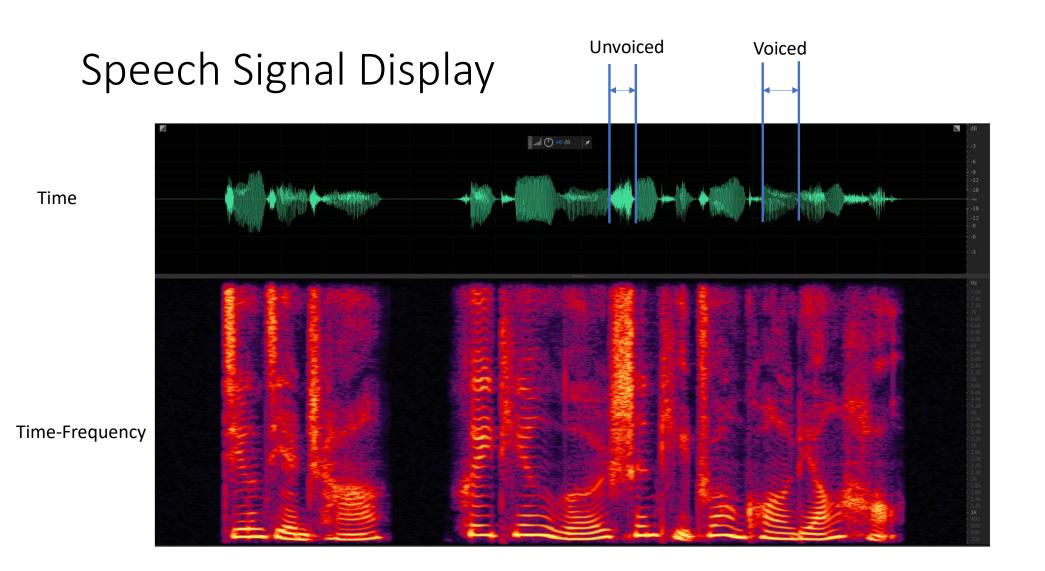
Speech Signal Representation

- Represent continuous signal into discrete form
- Max measurable frequency is half sampling rate (Shannon sampling theorem)
- Quantization Representing real value of each amplitude as integer 8-bit (-128 to 127) or 16-bit (-32768 to 32767)



Speech Signal Display

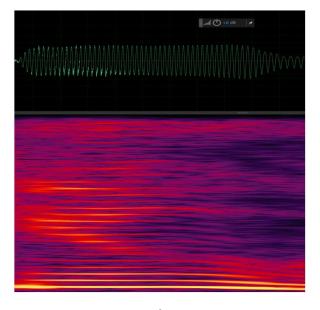




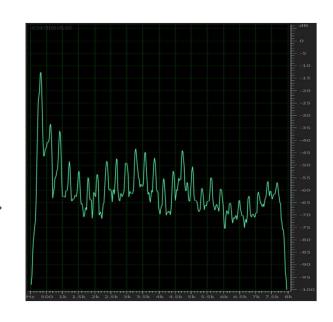
Voiced

Sinusoid in time domain

Harmonics in frequency domain



Time and Time-Frequency domain

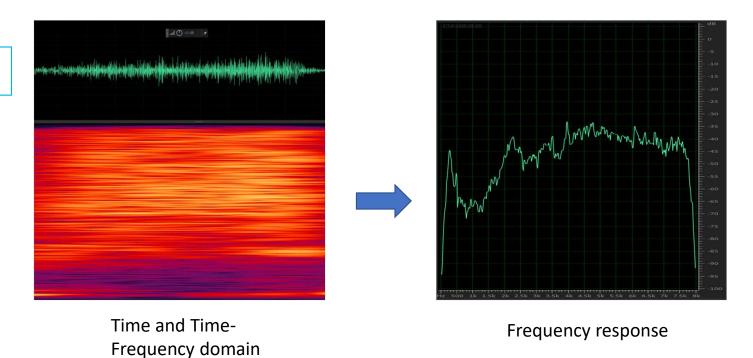


Frequency response

浊音声带紧绷,气流来了后,张弛振动,周期性的开启和闭合,形成准周期性的脉冲状空气流(周期为基音周期)。声带越短、厚度越薄、张力越大,则音调越高,即浊音的基音频率越高。男性基音频率50-250Hz,女性基音频率100-500Hz

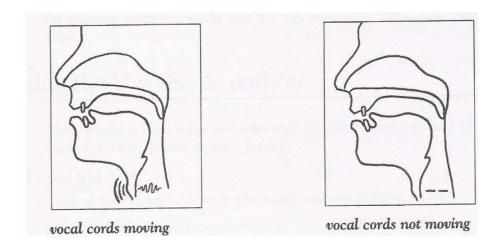
Unvoiced

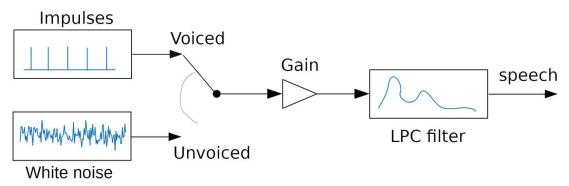
Like a white noise



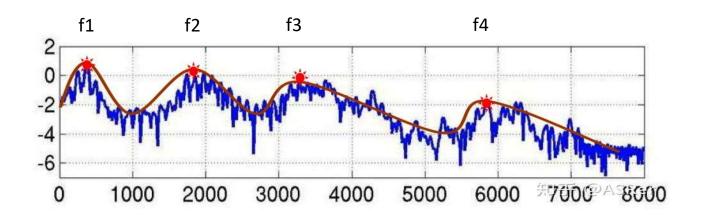
清音声带完全舒展,气流来了后,两种情况:第一,声道某个部位收缩为狭窄通道,声流高速冲过,形成摩擦音或者清音;第二,声道某部位完全闭合(如闭嘴),则形成爆破音。

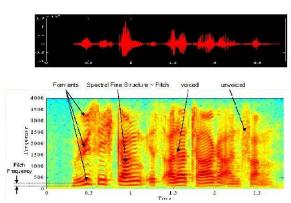
Speech Production





Formant (共振峰)



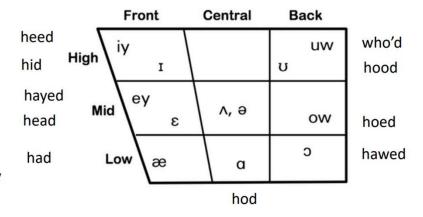


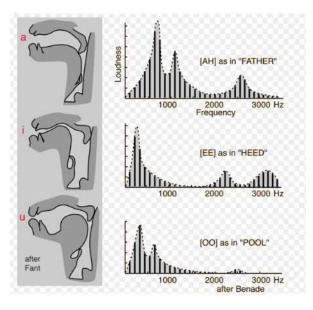
通常有3-5个共振峰 开口度越大,f1越高; 舌位越靠前,f2越高; 不圆唇元音的f3比圆唇元音高。

Vowel

- Tongue height:
 - Low: e.g., /a/
 - Mid: e.g., /e/
 - High: e.g., /i/
- Tongue advancement:
 - Front : e.g., /i/
 - Central: e.g., /ə/
 - Back : e.g., /u/
- Lip rounding:
 - Unrounded: e.g., /I, ε, e, ə/
 - Rounded: e.g., /u, o, ɔ/
- Tense/lax:
 - Tense: e.g., /i, e, u, o, ɔ, a/
 - Lax: e.g., /I, ε, æ, ə/

Simple & Glided Vowels

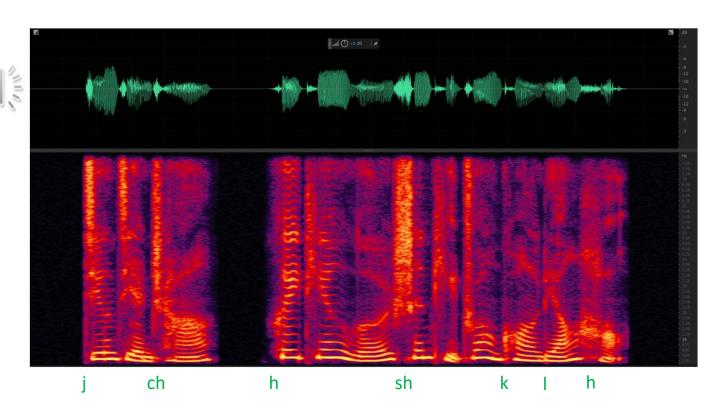




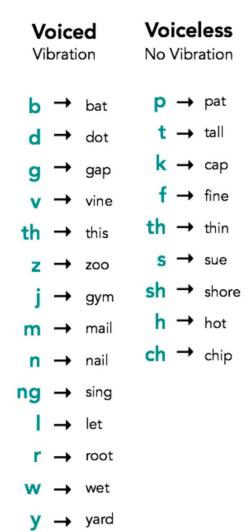
Consonant

- 辅音也叫"子音"。发音时,气流从肺中呼出,经过声门、咽腔、口腔或鼻腔时,受到各个器官不同程度的阻碍,不能畅通。由于阻碍部位(发音部位)和阻碍的方法及除去阻碍的方式(发音方法)不同,造成不同的辅音
- - 寒音 Stops: /p, t, k, b, d, g/
- − 擦音 Fricatives: /f, s, v, z/
- - 塞擦音 Affricates: /ts, dz/
- - 近音/边音 Approximants/Liquids: /l, r, w, j/
- - 鼻音 Nasals: /m, n, ng/

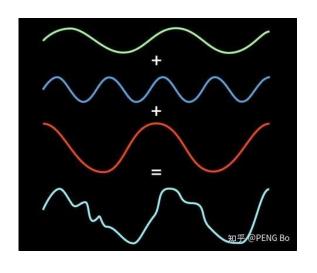
Consonant contains voiced sound

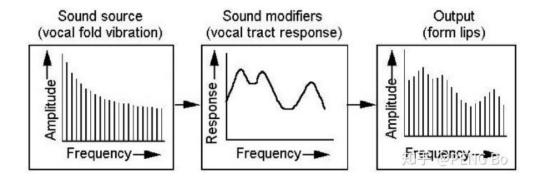


浊音会在清音的基础上有周期性的精细结构



Speech Signal

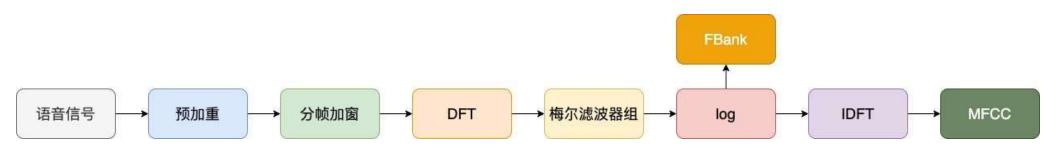




Green Function

- 对线性算子 L, 在点源 δ 作用下的输出(或响应)就是格林函数G, 即: LG=δ
- 声波波动问题,线性算子为 $L = \frac{\partial^2}{\partial t^2} c^2 \nabla^2$
- 若已知格林函数与源分布(包括时间上与空间上),则可通过格林函数与源的卷积求得在此源作用下系统的输出(或响应)
- Lφ=Q,其中 L 是线性算子, Q 为源分布, φ 为待求输出。利用卷积的性质, 可得: φ=φ*δ=φ*(LG)=(Lφ)*G=Q*G.

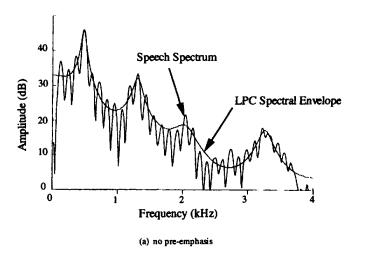
Feature Extraction

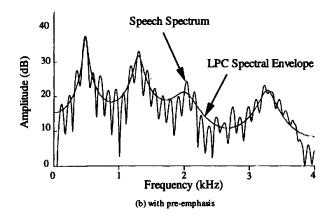


Pre-emphasis

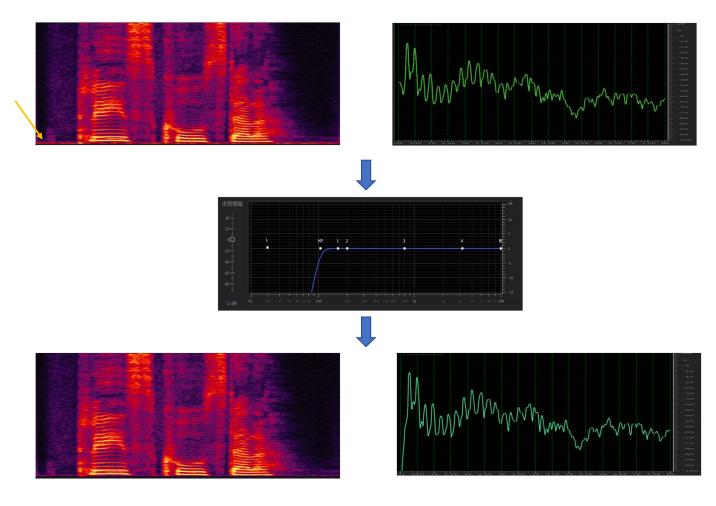
- Boost energy in the high frequencies
- Spectrum for voiced segments has more energy at low frequencies than high frequencies, called spectral tilt, caused by glottal pulse

$$y[n] = x[n] - \alpha x[n-1], \qquad 0.9 \le \alpha \le 1.0$$





High Pass Filter



Frame Segment

Why divide speech signal into successive overlapping frames?

Speech is not a stationary signal

We want information about a small enough region that the spectral information is a useful cue.

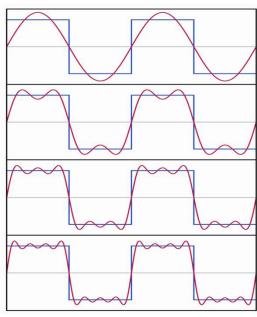
Frames

Frame size: typically, 10 -25 ms

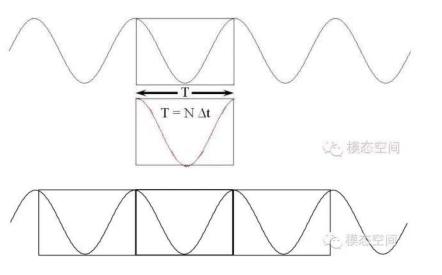
Frame shift: the length of time between successive frames, typically, 5 -10 ms

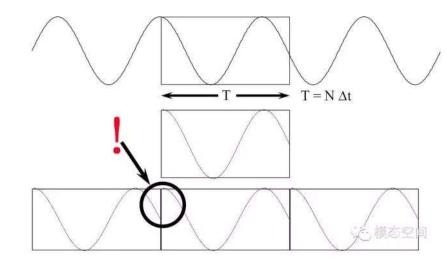
Fourier Transform

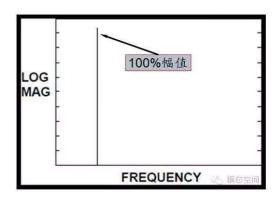
- 周期信号,表示为可数个正弦波的叠加
- 对于非周期信号,我们不能简单地将它展开为可数个正弦波的叠加,但是可以利用傅里叶变换展开为不可数的正弦波的叠加



Fourier Transform on Truncated Signal







假设原始信号的频率为f Hz,则周期为1/f s。因为截取的时间长度T为信号周期的整数倍(假设为k倍),即

T=k/*f*

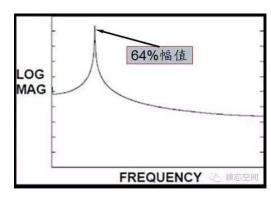
而频率分辨率为1/7,即

 $\Delta f = 1/T = f/k$

因而,信号的频率成分

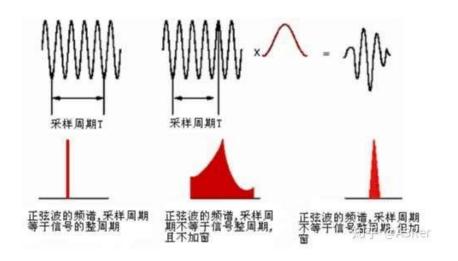
f =k*∆*f*

即信号的频率成分为频率分辨率**Δf**的整数倍,也就是说频 谱图中有一条谱线与信号的频率成分相同,这也就是所谓 的信号"**压谱线**"



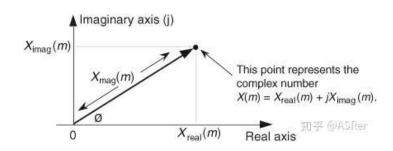
Window for Fourier Transform

- 避免频谱泄露
- 频谱泄露就是分析结果中,出现了本来没有的频率分量。比如说,50Hz的纯正弦波,本来只有一种频率分量,分析结果却包含了与50Hz频率相近的其它频率分量。



Discrete Fourier Transform

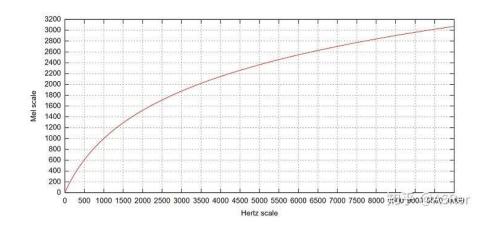
- Input:
 - Windowed signal x[n]...x[m]
- Output:
 - > For each of N discrete frequency bands
 - ➤ A complex number X[k] representing magnitude and phase of that frequency component in the original signal Discrete Fourier Transform (DFT)
- ullet DFT: $X(m) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi n m/N}$

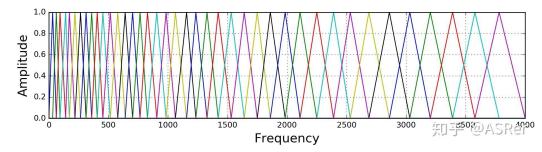


Mel Scale

- Mel-scale Human hearing is not equally sensitive to all frequency bands
- Less sensitive at higher frequencies, roughly > 1000 Hz
- Human perception of frequency is non-linear
- 在 Mel 频域内,人的感知能力为线性关系,如果两段语音的 Mel 频率差两倍,则人在感知上也差两倍
- Mel 频率与频率 Hz 转换的公式如下:

$$m=2595log_{10}(1+rac{f}{700})$$





40dim Mel Filter Bank

Reference

- https://zhuanlan.zhihu.com/p/147386972
- https://www.semanticscholar.org/paper/Noise-Cancellation-for-CELP-Voice-Encoders-in-an-Heide/f03c7d2d677f6ee13bec09e435b9cc694885a8ca
- https://slidetodoc.com/lsa-352-speech-recognition-and-synthesis-dan-jurafsky-5/
- https://www.sciencedirect.com/topics/engineering/original-signal-spectrum
- https://mp.weixin.qq.com/s/aLSmlrgQF7FBxh YXXfq6w
- https://zhuanlan.zhihu.com/p/40329331
- https://blog.csdn.net/weixin 42846157/article/details/104486434
- https://jmvalin.ca/demo/lpcnet/
- https://courses.engr.illinois.edu/ece417/fa2017/ece417fa2017lecture8.pdf
- http://campusweb.howardcc.edu/ehicks/YE618/Mastering%20American%20Pronunciation/Voiced Voiceless Sounds/Voiced Voiceless Sounds print.html
- https://myenglishfaves.blogspot.com/2017/05/voiceless-and-voiced-consonants-chart.ht
- https://www.zhihu.com/question/24190826