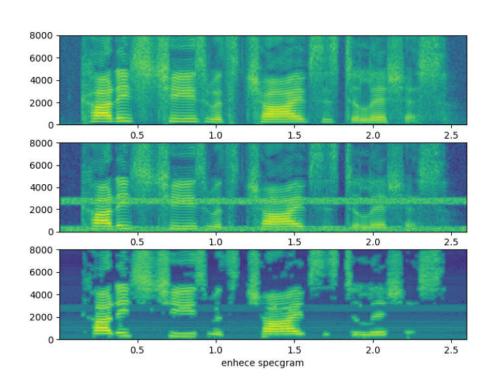


# 语音增强-谱减法

#### **Speech Enhancement-Spectral Subtraction**



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## 语音增强 (去噪)

• 消除语音中的噪声,增加语音听感与可懂度

#### 谱减法的 基本原理

$$y(n) = x(n) + e(n)$$

$$Y(\omega) = X(\omega) + E(\omega)$$

$$|\hat{X}(\omega)| = |Y(\omega)| - |E(\omega)|$$

$$\hat{X}(\omega) = |\hat{X}(\omega)| e^{j\varphi_{Y}(\omega)}$$

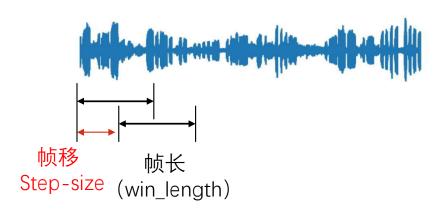
$$|\hat{X}(\omega)| = |\hat{X}(\omega)| e^{j\varphi_{Y}(\omega)}$$

$$|\hat{X}(\omega)| = |\hat{X}(\omega)| = |\hat$$



### 实现过程

•1将输入语音分帧进行短时傅里叶变换(STFT)



- 2 取前N帧进行噪声估计(获取  $E(\omega)$ )
- 3 谱减
- 4 ISTFT 还原获取干净语音*x*(n)

#### 人工智能学院

#### 智能语音处理



```
import librosa
from librosa.core.spectrum import amplitude to db
import numpy as np
import soundfile as sf
import matplotlib.pyplot as plt
if name == " main ":
    clean wav file = "sf1 cln.wav"
    clean,fs = librosa.load(clean wav file,sr=None)
    print(fs)
    noisy wav file = "sf1 n0L.wav"
    noisy, fs = librosa.load(noisy wav file, sr=None)
    # 计算 nosiy 信号的频谱
    S noisy = librosa.stft(noisy,n fft=256, hop length=128, win length=256) # D x T
    D,T = np.shape(S noisy)
    Mag noisy= np.abs (S noisy)
    Phase nosiy = np.angle (S noisy)
    Power nosiy = Mag noisy**2
    print(fs)
    # 估计噪声信号的能量
    # 由于噪声信号未知 这里假设 含噪 (noisy) 信号的前30帧为噪声
    Mag nosie = np.mean(np.abs(S noisy[:,:30]),axis=1,keepdims=True)
    Power nosie = Mag nosie**2
    Power nosie = np.tile(Power nosie,[1,T])
```

干净



噪声





```
# 能量减
                                                                  5000 -
Power enhenc = Power nosiy-Power nosie
# 保证能量大于0
Power enhenc[Power enhenc<0]=0
Mag enhenc = np.sqrt (Power enhenc)
                                                                                0.5
                                                                                          1.0
                                                                                                     1.5
                                                                                                               2.0
# 幅度减
# Mag enhenc = np.sqrt(Power nosiy) - np.sqrt(Power nosie)
                                                                  5000 -
# Mag enhenc[Mag enhenc<0]=0
# 对信号进行恢复
S enhec = Mag enhenc*np.exp(1j*Phase nosiy)
enhenc = librosa.istft(S enhec, hop length=128, win length=256)
                                                                                0.5
                                                                                           1.0
                                                                                                     1.5
                                                                                                               2.0
                                                                                                                         2.5
sf.write("enhce.wav",enhenc,fs)
print(fs)
                                                                  5000
# 绘制谱图
plt.subplot(3,1,1)
plt.specgram(clean,NFFT=256,Fs=fs)
plt.xlabel("clean specgram")
                                                                                0.5
                                                                                          1.0
                                                                                                     1.5
                                                                                                               2.0
                                                                                                                         2.5
plt.subplot(3,1,2)
                                                                                                                                音乐噪声
                                                                                           enhece specgram
plt.specgram(noisy,NFFT=256,Fs=fs)
                                                   增强语音
plt.xlabel("noisy specgram")
plt.subplot(3,1,3)
plt.specgram(enhenc,NFFT=256,Fs=fs)
plt.xlabel("enhece specgram")
plt.show()
plt.imshow(librosa.amplitude to db(Mag enhenc, ref=np.max), origin='lower')
plt.show()
                                                                                           100
                                                                                                  150
                                                                                                         200
                                                                                                                      300
                                                                                                               250
    2021/7/29
                                                           谱减法
```

较小 0.001~0.0001



### 过减法

$$P_{x}(\omega) = \left(P_{Y}(\omega)^{\gamma} - \alpha P_{Y}(\omega)^{\gamma}\right)^{\frac{1}{\gamma}}$$

$$P_{x}(\omega) = \begin{cases} \beta P_{e}(\omega) & P_{X}(\omega) < \beta P_{e}(\omega) \\ P_{x}(\omega) & \text{!} \text{!!} \text{!!} \end{cases}$$

$$|X(\omega)| = \sqrt{P_x(\omega)}$$



```
## 方法2 超减
# 引入参数
alpha = 4
gamma = 1
Power enhenc = np.power(Power nosiy,gamma) - alpha*np.power(Power nosie,gamma)
Power enhenc = np.power (Power enhenc, 1/gamma)
# 对于过小的值用 beta* Power nosie 替代
beta = 0.0001
                                                                      5000 -
mask = (Power enhenc>=beta*Power nosie) -0
print(mask.shape)
Power_enhenc = mask*Power_enhenc + beta*(1-mask)*beta*Power_nosie
                                                                                   0.5
                                                                                             1.0
                                                                                                       1.5
                                                                                                                 2.0
Mag enhenc = np.sqrt(Power enhenc)
                                                                     5000
                                                                                   0.5
                                                                                             1.0
                                                                                                       1.5
                                                                                                                 2.0
                                                                                                                           2.5
                                                                     5000
                                                                        0
                                              谱线的不连续
                                                                                   0.5
                                                                                             1.0
                                                                                                       1.5
                                                                                                                 2.0
                                                                                                                           2.5
```



引入平滑机制

最大噪声残差

$$\max(\omega) = \arg\max \sum_{t=0}^{T_{noise}} E_t(\omega) - E(\omega)$$
估计的噪声 (E\_t(w)的均值)

$$\left| \mathbf{X}(\omega) \right| = \begin{cases} \arg\min \sum_{t-k}^{t+k} \left| \mathbf{X}_{t}(\omega) \right| & \left| \mathbf{X}(\omega) \right| < \max(\omega) \\ \left| \mathbf{X}(\omega) \right| & \text{其他} \end{cases}$$

对于过小的部分用相邻帧的最小值取代





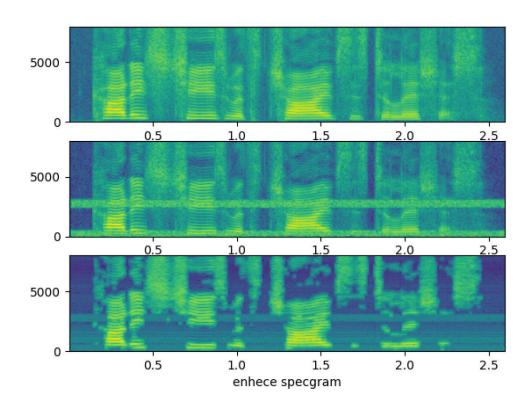
```
Mag_enhenc_new = np.copy(Mag_enhenc)
# 计算最大噪声残差
maxnr = np.max(np.abs(S_noisy[:,:31])-Mag_nosie,axis =1)

k = 1
for t in range(k,T-k):
    index = np.where(Mag_enhenc[:,t]<maxnr)[0]
    temp = np.min(Mag_enhenc[:,t-k:t+k+1],axis=1)
    Mag_enhenc_new[index,t] = temp[index]

# 对信号进行恢复
S_enhec = Mag_enhenc_new*np.exp(1j*Phase_nosiy)
enhenc = librosa.istft(S_enhec, hop_length=128, win_length=256)
sf.write("enhce_3.wav",enhenc,fs)
print(fs)

Print(fs)
```









2021/7/29