

sEMG De-noising and Muscle Performance Evaluation

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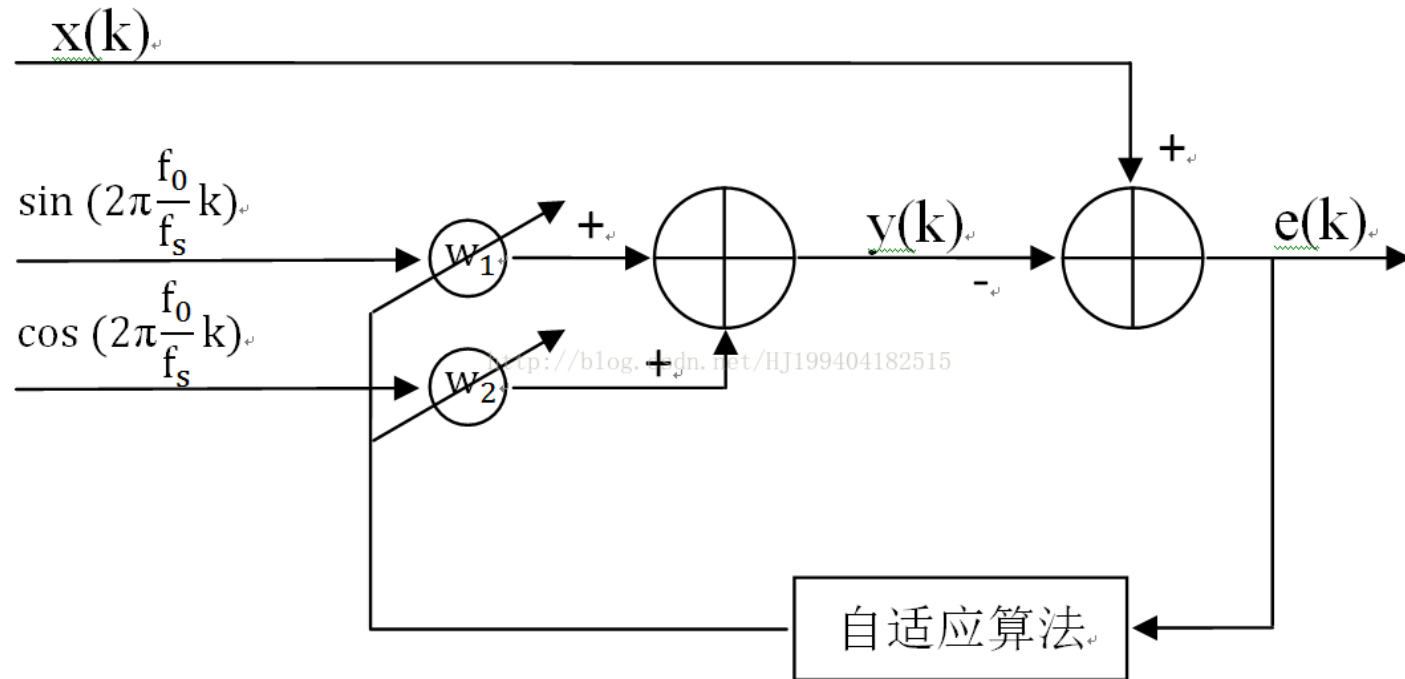
Adaptive filter

➤ Advantages

- adjustable parameters
- adaption to changing input signal environment and output signal characteristics
- better performance in sEMG denoising



Theory



$x(k)$: original signal with noises

f_s : sampling frequency

$y(k)$: estimated noise

adaptive algorithm: Least Mean Square(LMS); Recursive Least Mean Square(RMS)

f_0 : frequency of noise

w_1 w_2 : coefficients

$e(k)$: error function



LMS vs. RLS

Filter	EMG1	EMG2	EMG3	EMG4
LMS algorithm	6.32	6.98	8.12	8.53
RLS algorithm	8.77	7.41	9.27	9.14
Wavelet db4	7.66	7.35	8.45	8.13
0.34-1 HZ bandpass filter	5.44	4.63	4.88	5.75

SNR values (in db) of denoised uterine EMG signals



RLS algorithm

RLS 算法: ↵

$$x(k) = \begin{bmatrix} \sin(2\pi \frac{f_0}{f_s} k) \\ \cos(2\pi \frac{f_0}{f_s} k) \end{bmatrix} \quad \text{↵}$$

f_0 为噪声信号频率, f_s 为采样频率, $x(k)$ 为信号矢量 ↵

$$K(k) = \frac{P(k)x(k)}{\lambda + x^T(k)P(k)x(k)} \quad \text{↵}$$

$K(k)$ 为增益矢量维度为 2×1 ; λ 为遗忘因子, 一般取值为 0.95-0.9995; $P(0) = \delta^{-1}I$, 其中 δ 为一个小正数 (程序中取为 0.01) 保证矩阵 P 非奇异, I 为 2×2 单位矩阵, 故 P 为 2×2 对角矩阵 ↵

$$f(k) \text{ 为滤波器系数矢量维度为 } 2 \times 1, \quad f(0) = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \text{↵}$$

$$y(k) = x^T(k)f(k) \quad \text{↵}$$

$$e(k) = d(k) - y(k) \quad \text{↵}$$

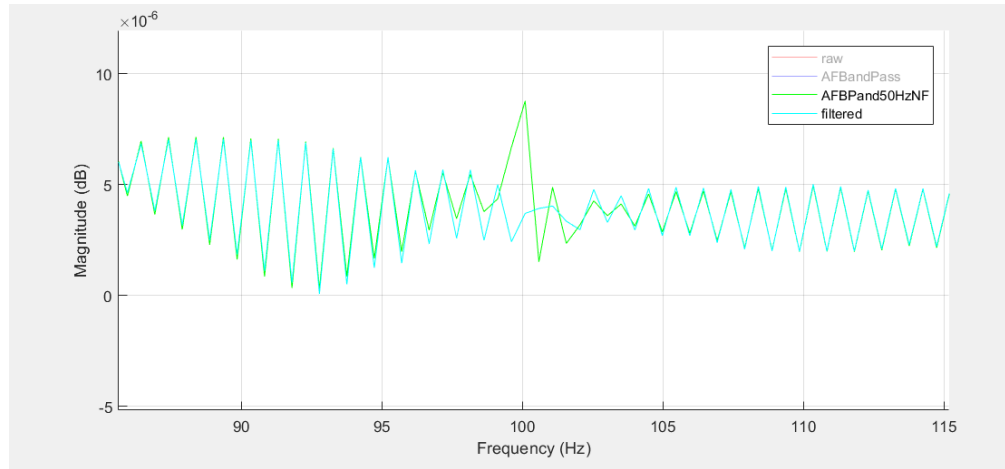
$d(k)$ 为原始信号, $y(k)$ 为计算出来需要滤除的信号, $e(k)$ 为滤波后的信号 ↵

$$f(k+1) = f(k) + K(k)e(k) \quad \text{↵}$$

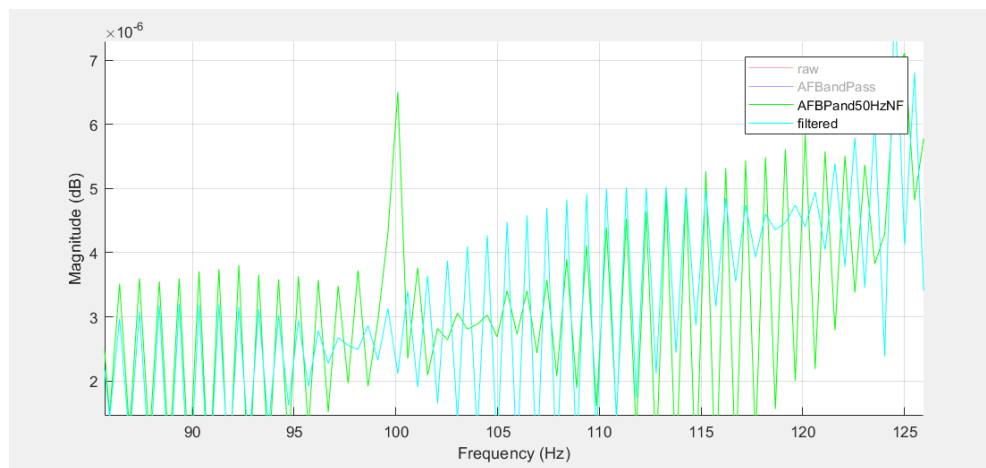
$$P(k+1) = \lambda^{-1}P(k) - \lambda^{-1}K(k)x^T(k)P(k) \quad \text{↵}$$



Conclusion



100Hz notch filter with fixed parameters



100Hz adaptive notch filter



Literature Review

➤ Muscle force estimation

- Neural Muscle Force Estimator(NMFE)
- Fuzzy Muscle Force Estimator(FMFE)

➤ Procedure

- first layer: neural network
- second layer: neural network theory; fuzzy logic algorithms

乌萨马.根据肌电图(EMG)信号评估人手臂肌肉力和肌肉疲劳的人工智方法[D].
武汉:华中科技大学博士学位论文.2013.

Thanks