REPORT

Subject: Digital Signal Processing

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Laboratory #3	Bartosz Bieniek
Date: 23.11.2024	IT Science
Topic: Projektowanie i analiza filtrów cyfrowych:	II degree, 1
implementacja filtrów FIR i IIR w Matlabu/Pythonie.	semester, gr.A
Filtracja adaptacyjna: zastosowanie algorytmów filtracji	
adaptacyjnej w redukcji szumów.	
Second variant (2)	

1. Task:

Variant 2:

- Design an FIR filter with the following coefficients and implement it in Python to reduce noise in a noisy sinusoidal signal.

FIR Filter Coefficients: b = {1, 1, 2}

- Design an IIR filter with the following coefficients and implement it in Python to reduce noise in the same noisy sinusoidal signal.

IIR Filter Coefficients: $b = \{0.5, 0.2\}, a = \{1, -0.8\}$

- Implement an adaptive LMS filter in Python with a step size μ = 0.05 and filter length

M = 4 to reduce noise in the same noisy sinusoidal signal.

2. Code description **Github Repository**

```
import numpy as np
import matplotlib.pyplot as plt
def fir_filter(x, b):
   return np.convolve(x, b, mode='same')
def iir_filter(x, b, a):
    y = np.zeros_like(x)
    for n in range(len(x)):
        y[n] = b[0] * x[n]
elif n == 1:
          y[n] = b[0] * x[n] + b[1] * x[n-1]
          y[n] = b[0] * x[n] + b[1] * x[n-1] - a[1] * y[n-1]
def lms_filter(x, d, mu, M):
   N = len(x)
   y = np.zeros(N)
   e = np.zeros(N)
   w = np.zeros(M)
    for n in range(M, N):
       x_n = x[n-M:n][::-1]
       y[n] = np.dot(w, x_n)
       e[n] = d[n] - y[n]
w = w + mu * e[n] * x_n
```

TD. 1. Definition of filters

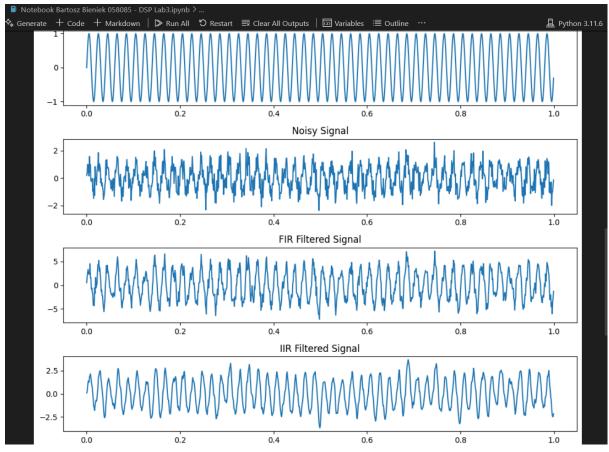
Functions shown at technical drawing number one are implementing basic digital signal processing filters in Python. The FIR filter (fir_filter) applies a finite impulse response filter to the input signal using convolution with given coefficients. The IIR filter (iir_filter) processes the input signal with recursive coefficients, where each output depends on previous outputs. The LMS adaptive filter (Ims_filter) adjusts its filter weights based on the approximation between the desired signal and the current output, optimizing for noise reduction in real-time.

```
fs = 1000
                                                                                                                           t = np.arange(0, 1, 1/fs)
f_signal = 50
signal = np.sin(2 * np.pi * f_signal * t)
noise = np.random.normal(0, 0.5, len(t))
noisy_signal = signal + noise
b_iir = np.array([0.5, 0.2])
a_iir = np.array([1, -0.8])
fir_output = fir_filter(noisy_signal, b_fir)
iir_output = iir_filter(noisy_signal, b_iir, a_iir)
lms_output, lms_error = lms_filter(noisy_signal, signal, mu, M)
plt.figure(figsize=(10, 8))
plt.subplot(4, 1, 1)
plt.plot(t, signal)
plt.title('Original Signal')
plt.subplot(4, 1, 2)
plt.plot(t, noisy_signal)
plt.title('Noisy Signal')
plt.subplot(4, 1, 3)
plt.plot(t, fir_output)
plt.title('FIR Filtered Signal')
plt.subplot(4, 1, 4)
plt.plot(t, iir_output)
plt.title('IIR Filtered Signal')
plt.tight_layout()
plt.show()
```

TD. 2. Applying data from task and plotting.

The code displayed above assigns the data provided with task and generating a noisy sinusoidal signal, applying the previously defined filters, and visualizing the results.

The code creates a clean sinusoidal signal and then adds a noise, mimicking a real-world noisy signal. Then, the FIR filter, IIR filter, and LMS adaptive filter are applied to see the result of how does this signal works. Each filter is designed to reduce noise, so filtered signals supposed to be comparable to the original and noisy signals in the plots.



T.D. 3. The result of nosing signal and filtering.

3. Conclusions

The FIR and IIR filters significantly reduce the noise, but very noisy signal drastically changes comparing to the original.

The FIR filter, though simple, performs very well in smoothing the signal. Meanwhile the IIR filter (with its recursive nature) achieves a more efficient way to get similar results with fewer coefficients.