lab4

April 14, 2024

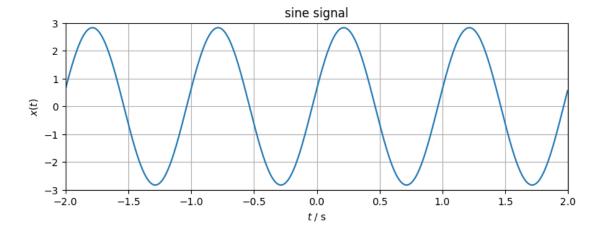
```
[1]: # most common used packages for DSP, have a look into other scipy submodules
     import numpy as np
     import matplotlib as mpl
     import matplotlib.pyplot as plt
     from scipy import signal
     def my_xcorr2(x, y, scaleopt='none'):
         N = len(x)
         M = len(y)
         kappa = np.arange(0, N+M-1) - (M-1)
         ccf = signal.correlate(x, y, mode='full', method='auto')
             if scaleopt == 'none' or scaleopt == 'raw':
                 ccf /= 1
             elif scaleopt == 'biased' or scaleopt == 'bias':
                 ccf /= N
             elif scaleopt == 'unbiased' or scaleopt == 'unbias':
                 ccf /= (N - np.abs(kappa))
             elif scaleopt == 'coeff' or scaleopt == 'normalized':
                 ccf /= np.sqrt(np.sum(x**2) * np.sum(y**2))
             else:
                 print('scaleopt unknown: we leave output unnormalized')
         return kappa, ccf
```

```
[2]: w = 2*np.pi*1 # f = 1 Hz
tend = 4 # theoretically the sine has infinite duration, here only 4s to plot
N = tend * 2**8 # 256 samples per period -> very sufficient oversampling
t = np.arange(N)/N * tend - tend//2 # here: [-2s...+2s)
A = np.sqrt(8) # this choice matches an ACF amplitude of 4
phi = np.pi/14 # arbitrary choice, note that ACF is not affected by phase
x = A*np.sin(w*t + phi) # create the sine signal

# estimate the sampling frequency from the time interval between two samples
fs = 1 / (t[1]-t[0])
print('fs = {0:f} Hz'.format(fs))
plt.figure(figsize=(9, 3))
```

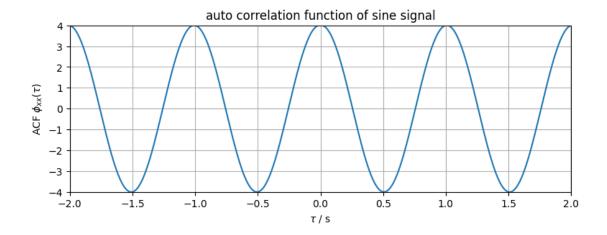
```
plt.plot(t, x)
plt.xlim(-2, 2)
plt.ylim(-3, 3)
plt.xlabel(r'$t$ / s')
plt.ylabel(r'$x(t)$')
plt.title('sine signal')
plt.grid(True)
```

fs = 256.000000 Hz



```
[3]: kappa, phixx = my_xcorr2(x, x, 'unbiased')

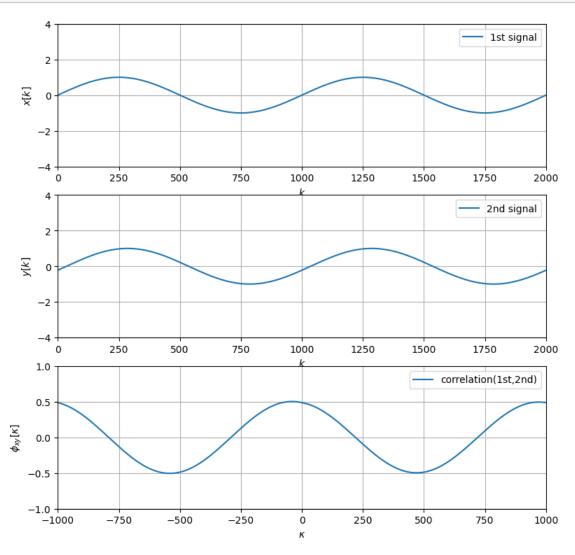
plt.figure(figsize=(9, 3))
plt.plot(kappa/fs, phixx)
plt.xlim(-2, 2)
plt.ylim(-4, 4)
plt.xlabel(r'$\tau$ / s')
plt.ylabel(r'ACF $\phi_{xx}(\tau)$')
plt.title('auto correlation function of sine signal')
plt.grid(True)
```



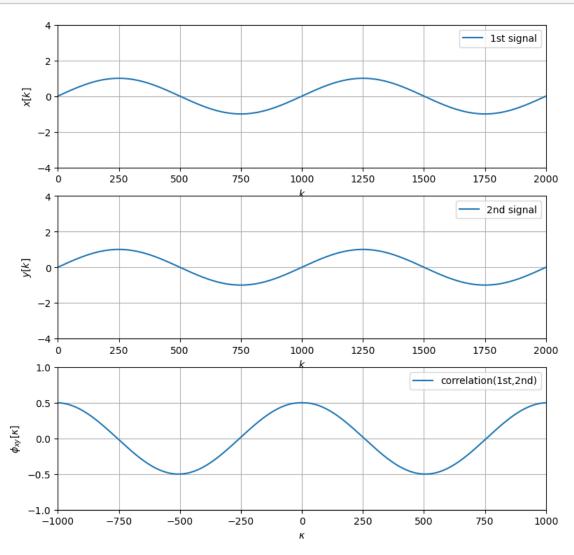
```
[4]: def my_ccf_plot(x, y, scaleopt):
         kappa, ccf = my_xcorr2(x, y, scaleopt)
         plt.figure(figsize=(9, 9))
         plt.subplot(3, 1, 1)
         plt.plot(x, label='1st signal')
         plt.xlim(0, 2000)
         plt.ylim(-4, 4)
         plt.xlabel(r'$k$')
         plt.ylabel(r'$x[k]$')
         plt.legend()
         plt.grid(True)
         plt.subplot(3, 1, 2)
         plt.plot(y, label='2nd signal')
         plt.xlim(0, 2000)
         plt.ylim(-4, 4)
         plt.xlabel(r'$k$')
         plt.ylabel(r'$y[k]$')
         plt.legend()
         plt.grid(True)
         plt.subplot(3, 1, 3)
         plt.plot(kappa, ccf, label='correlation(1st,2nd)')
         plt.xlim(-1000, 1000)
         plt.ylim(-1, 1)
         plt.xlabel(r'$\kappa$')
         plt.ylabel(r'$\phi_{xy}[\kappa]$')
         plt.legend()
         plt.grid(True)
```

```
[5]: N = 5000
k = np.arange(N)
Omega1 = 5 * 2*np.pi/N
Omega2 = 20 * 2*np.pi/N
```

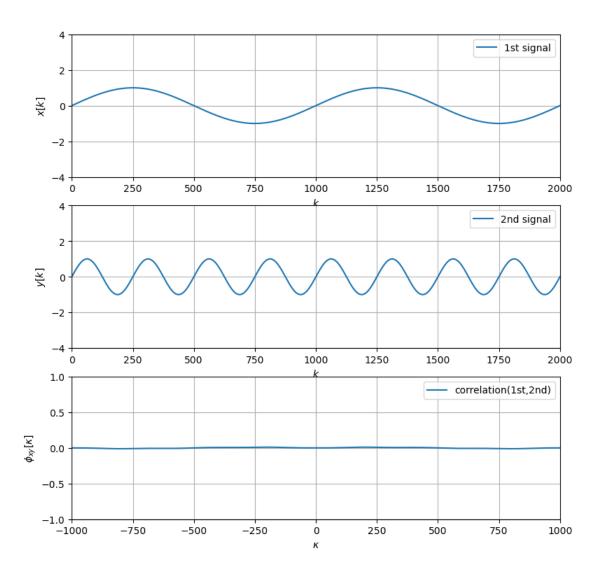
```
[6]: # a)
x = np.sin(Omega1*k)
# b)
y = np.sin(Omega1*k - np.pi/14)
# c) and d)
my_ccf_plot(x, y, scaleopt='unbiased')
```



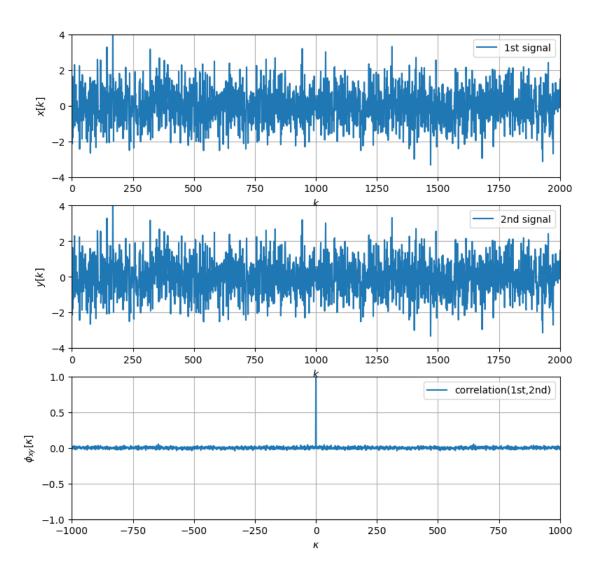
```
y = x
my_ccf_plot(x, y, scaleopt='unbiased')
```



```
[8]: # g)
x = np.sin(Omega1*k)
y = np.sin(Omega2*k)
my_ccf_plot(x, y, scaleopt='unbiased')
```



```
[9]: # h)
    np.random.seed(2) # arbitrary choice
    x = np.random.randn(N)
    y = x
    my_ccf_plot(x, y, scaleopt='biased')
```



```
[10]: x = (+1., +2., -3.)
kappa, acf = my_xcorr2(x, x, 'biased') # use own function defined above

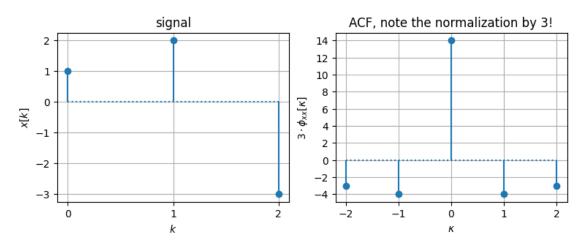
for i in acf:
    print('{:f}'.format(i), end=', ')

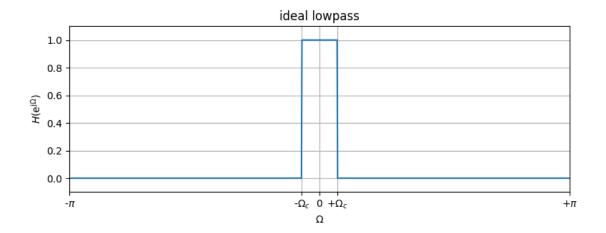
plt.figure(figsize=(9, 3))
plt.subplot(1, 2, 1)
plt.stem(x, basefmt='CO:')
plt.xticks(np.arange(0, 3))
plt.yticks(np.arange(-3, 3, 1))
plt.ylabel(r'$k$')
plt.ylabel(r'$x[k]$')
plt.title('signal')
```

```
plt.grid(True)

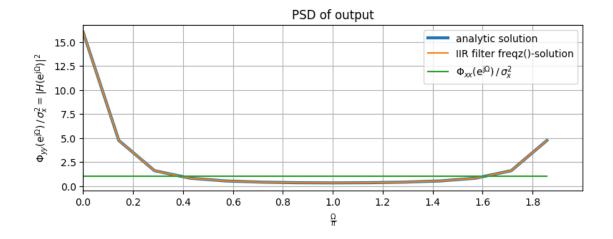
plt.subplot(1, 2, 2)
# note that we plot 3*ACF ! this gives simple integer results in the plot
plt.stem(kappa, acf*3, basefmt='CO:')
plt.xticks(np.arange(-2, 3))
plt.yticks(np.arange(-4, 16, 2))
plt.xlabel(r'$\kappa$')
plt.ylabel(r'$3 \cdot \phi_{xx}[\kappa]$')
plt.title('ACF, note the normalization by 3!')
plt.grid(True)
```

-1.000000, -1.333333, 4.666667, -1.333333, -1.000000,

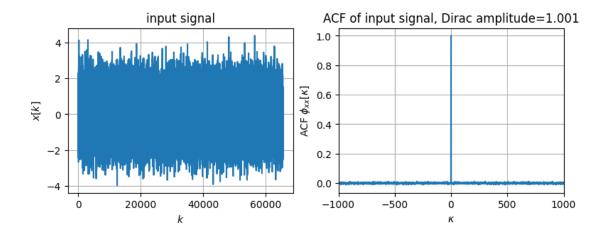




```
[12]: N = 14
      Omega = np.arange(N) * 2*np.pi/N
      H2 = 2 / (25/8 - 3*np.cos(Omega)) # analytic
      Omega, H_{IIR} = signal.freqz(b=(1), a=(1, -3/4), worN=Omega)
                                                                     # numeric
      plt.figure(figsize=(9, 3))
      plt.plot(Omega/np.pi, H2, lw=3, label='analytic solution')
      plt.plot(Omega/np.pi, np.abs(H_IIR)**2, label='IIR filter freqz()-solution')
      plt.plot(Omega/np.pi, Omega*0+1,
               label=r'$\Phi_{xx}(\mathrm{e}^{\mathrm{j}\Omega})\,/\,\sigma_x^2$')
      plt.xlabel(r'$\frac{\Omega}{\pi}$')
      plt.ylabel(
          r'$\Phi_{yy}(\mathrm{e}^{\mathrm{j}\Omega})\,/\,\sigma_x^2 =__
       \hookrightarrow |H(\mathrm{e}^{\mathrm{j}\Omega})|^2$')
      plt.title('PSD of output')
      plt.xlim(0, 2)
      plt.xticks(np.arange(0, 20, 2)/10)
      plt.legend()
      plt.grid(True)
```



```
[13]: np.random.seed(2) # arbitrary choice
      Nx = 2**16
      k = np.arange(Nx)
      x = np.random.randn(Nx)
      kappa, phixx = my_xcorr2(x, x, 'biased') # we use biased here, i.e. 1/N_{L}
       \neg normalization
      idx = np.where(kappa==0)[0][0]
      plt.figure(figsize=(9, 3))
      plt.subplot(1, 2, 1)
      plt.plot(k, x)
      plt.xlabel('$k$')
      plt.ylabel('$x[k]$')
      plt.title('input signal')
      plt.grid(True)
      plt.subplot(1, 2, 2)
      plt.plot(kappa, phixx)
      plt.xlim(-1000, +1000)
      plt.xlabel('$\kappa$')
      plt.ylabel('ACF $\phi_{xx}[\kappa]$')
      plt.title('ACF of input signal, Dirac amplitude=%4.3f' % phixx[idx])
      plt.grid(True)
```

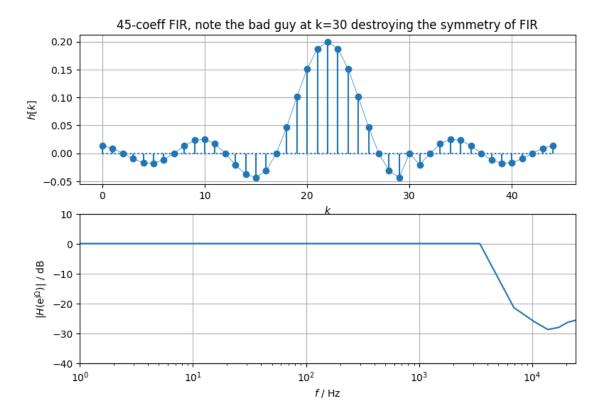


```
[14]: fs = 48000 # sampling frequency in Hz
      fc = 4800 # cut frequency in Hz
      number_fir_coeff = 45 # FIR taps
      h = signal.firls(numtaps=number_fir_coeff, # example for demo
                       bands=(0, fc, fc+1, fs//2),
                       desired=(1, 1, 0, 0),
                       fs=fs)
      Nh = h.size
      k = np.arange(Nh)
      # make the IR unsymmetric by arbitray choice for demonstration purpose
      idx = 30
      h[idx] = 0 # then FIR is not longer linear-phase, see the spike in the plot
      print('h[0]={0:4.3f}, DC={1:4.3f} dB'.format(h[0], 20*np.log10(np.sum(h))))
      N = 14
      Omega = np.arange(0, N) * 2*np.pi/N
      _, H = signal.freqz(b=h, a=1, worN=Omega)
      plt.figure(figsize=(9, 6))
      plt.subplot(2, 1, 1)
      plt.stem(k, h, basefmt='C0:')
      plt.plot(k, h, 'CO-', lw=0.5)
      plt.xlabel(r'$k$')
      plt.ylabel(r'$h[k]$')
      plt.title(str(Nh)+'-coeff FIR, note the bad guy at k=%d destroying the symmetry_

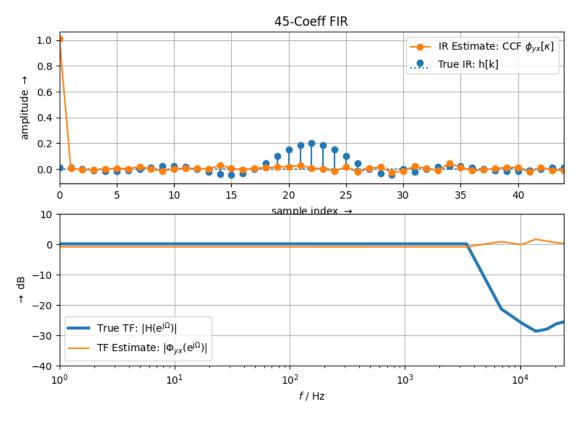
of FIR' % idx)
      plt.grid(True)
      plt.subplot(2, 1, 2)
      plt.semilogx(Omega / (2*np.pi) * fs, 20*np.log10(np.abs(H)))
      plt.xlabel(r'$f$ / Hz')
```

```
plt.ylabel(r'$|H(\mathrm{e}^{\mathbb{j}\0mega})|$ / dB')
plt.xlim(1, fs//2)
plt.ylim(-40, 10)
plt.grid(True)
```

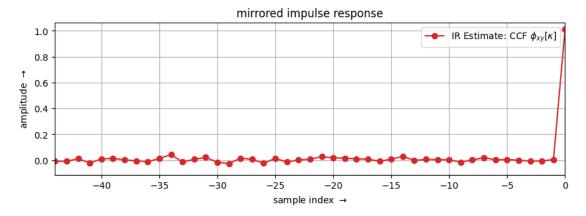
h[0]=0.014, DC=0.298 dB

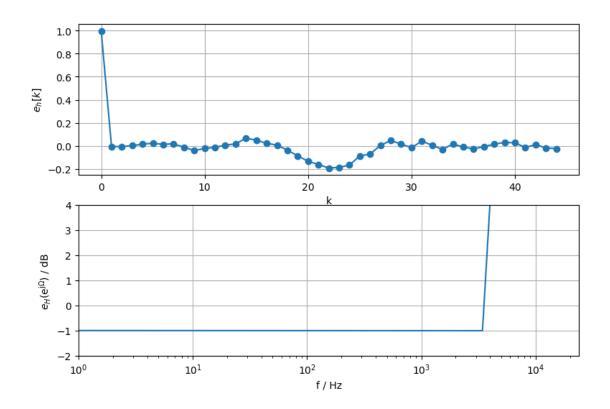


```
plt.xlabel(r'sample index $\rightarrow$')
plt.ylabel(r'amplitude $\rightarrow$')
plt.title(str(Nh)+'-Coeff FIR')
plt.legend()
plt.grid(True)
plt.subplot(2, 1, 2)
plt.semilogx(Omega/2/np.pi*fs, 20*np.log10(np.abs(H)), lw=3,
             label=r'True TF: $|\mathrm{H}(\mathrm{e}^{\mathrm{j}\Omega})|$')
plt.semilogx(Omega/2/np.pi*fs, 20*np.log10(np.abs(Phiyx)),
             label='TF Estimate: $|\Phi_{yx}(\mathrm{e}^{\mathbb{j}}0mega})|$')
plt.xlabel(r'$f$ / Hz')
plt.ylabel(r'$\rightarrow$ dB')
plt.xlim(1, fs//2)
plt.ylim(-40, 10)
plt.legend()
plt.grid(True)
```



```
plt.xlim(-(Nh-1), 0)
plt.xlabel(r'sample index $\rightarrow$')
plt.ylabel(r'amplitude $\rightarrow$')
plt.title('mirrored impulse response')
plt.legend()
plt.grid(True)
```





[]: