ECE 398-MA

Introduction to Modern Communication with Python and SDR

Lab 8 – FSK

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1 Assignment 1

```
import matplotlib.pyplot as plt
import numpy as np
from scipy.signal import resample_poly, butter, filtfilt, find_peaks
def rrcosfilter(N, alpha, Tb, Fs):
Generates a root raised cosine (RRC) filter (FIR) impulse response.
Parameters
N: int
Length of the filter in samples.
alpha: float
Roll off factor (Valid values are [0, 1]).
Tb: float
Symbol period.
Fs: float
Sampling Rate.
Returns
h_rrc : 1-D ndarray of floats
Impulse response of the root raised cosine filter.
```

```
" " "
 T_delta = 1/float(Fs)
sample_num = np.arange(N)
 h_rrc = np.zeros(N, dtype=float)
 for x in sample_num:
 t = (x-N/2) * T_delta
 if t == 0.0:
 h_{rrc}[x] = 1.0 - alpha + (4*alpha/np.pi)
  elif alpha != 0 and t == Tb/(4*alpha):
 h_{rrc}[x] = (alpha/np.sqrt(2))*(((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha))))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha))))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi)* (np.sin(np.sin(np.pi)* (np.sin(np.pi)* (np.sin(np.sin(np.pi)* (np.sin(np.sin(np.pi)* (np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(np.sin(n
  elif alpha != 0 and t == -Tb/(4*alpha):
 h_{rrc}[x] = (alpha/np.sqrt(2))*(((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha))))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha))))) + ((1+2/np.pi)* (np.sin(np.pi/(4*alpha)))) + ((1+2/np.pi)* (np.sin(np.pi)* (np.sin(np.pi)*
 else:
 h_{rrc}[x] = (np.sin(np.pi*t*(1-alpha)/Tb) +
4*alpha*(t/Tb)*np.cos(np.pi*t*(1+alpha)/Tb))/ (np.pi*t*(1-(4*alpha*t/Tb)*(4+alpha)/Tb))/ (np.pi*t*(1-(4*alpha)/Tb))/ (np.pi*t*(1-(4*alpha)/T
return h rrc
# Create random binary data and BPSK symbols from the previous labs
num data symbols = 32
 sps = 16
Tb = 1
np.random.seed(0)
 bits = np.random.randint(0, 2, num_data_symbols) # 0 to 1
bpsk symbols = bits*2 - 1
######### YOUR CODE STARTS HERE ###########
barker = np. array ([1,1,1,1,1,-1,-1,1,1,-1,1,-1,1])
# Concatenate TWO sets of Barkers, guard interval, and data to create a fram
guard = np.zeros_like(barker)
frame = np.concatenate((barker, barker, guard, bpsk_symbols))
# Upsample and perform pulse-shaping here (the pulse is given)
up_sym = np.zeros(len(frame)*sps)
up_sym[::sps] = frame # Symbol-spaced upsampling
# Tx_ADC is the pulse-shaped signal
num_taps = 6*sps + 1
 pulse = rrcosfilter(N=num_taps, alpha=1.0, Tb=sps, Fs=1)
Tx_ADC = np.convolve(up_sym, pulse)
Tx_ADC = np.concatenate((np.zeros(sps*10), Tx_ADC)) # pad more zeros to help
```

########## YOUR CODE ENDS HERE ###########

```
## Simulate IQ modulator (Tx)
      M = 16 # upsample fs_adc for pass-band simulation
      xup = resample_poly(Tx_ADC, M, 1)
      fs adc = sps # sampling rate of ADC
      fs_rf = M * fs_adc # sampling rate for simulating carrier
      fc = (M*3/7) * fs_adc # carrier frequency
      t = 1/fs_rf_{np.arange}(len(xup)) # time vector at fs_rf
      # u(t): transmitted signal to the channel (passband)
      u = np.real(xup) * np.cos(2*np.pi*fc*t) - np.imag(xup) * np.sin(2*np.pi*fc*
      ## Simulate Channel
      ch att = 0.1 # channel attenuation
      h = np.zeros(M*sps)
      h[0] = ch att
      h = np.roll(h, np.random.randint(M*sps)) # random delay
      v = np.convolve(u, h)
      noise amplitude = 0.01
      noise = noise amplitude * np.random.randn(len(v))
# AWGN
      v = v + noise
      ## Simulate IQ demodulator (Rx)
      # Low-Pass Filter (LPF) @ fc
      Nfilt = 5
      cutoff = fc
      b, a = butter(Nfilt, Wn=cutoff, btype='low', fs=fs rf)
      t = 1/fs rf*np.arange(len(v))
      # Add CFO
      N = len(barker)
      cfo_limit = 1/(2*N*Tb)
      \# cfo_hz = cfo_limit *0.05
      \# cfo_hz = -cfo_limit *0.05
      cfo_hz = cfo_limit *1.05
      print('CFO_limit_(Hz):_+-_', cfo_limit)
      print('Applied_CFO_(Hz):_', cfo_hz)
```

```
yI = filtfilt(b, a, v*np.cos(2*np.pi*(fc + cfo hz)*t))
yQ = filtfilt(b, a, -v*np.sin(2*np.pi*(fc + cfo hz)*t))
Rx\_ADC = resample\_poly(yI + 1j*yQ, 1, M)
######### YOUR CODE STARTS HERE ##########
# Matched filtering and symbol timing recovery from the previous labs here
rx_matched = np.convolve(Rx_ADC, pulse)
# Compute energy for each alignment offset
energy = np.zeros(sps)
for k in range(sps):
# Compute energy for each offset by summing squared values of sampled segme
energy[k] = np.sum(np.abs(rx matched[k::sps])**2)
# Find the best offset
max ind = np.argmax(energy)
print("max ind:", max ind)
# Align samples using max ind
rx_aligned = rx_matched[max_ind::sps]
# Self-reference Frame synchronization: compute N-lagged auto-correlation (
N = len(barker)
corr = np.zeros(len(rx_aligned) - 2*N)
for n in range(len(corr)):
sum = 0
for k in range(N):
sum += np.conj(rx_aligned[n+k]) * rx_aligned[n+k+N]
corr[n] = np.abs(sum) / N
frame ind = np.argmax(corr)
# Plot the correlation and its peak
plt.figure
plt.plot(corr)
plt.plot(frame ind, corr[frame ind], 'x', label=f'frame ind={frame ind}')
plt.title('self-reference_correlation')
plt.grid(True)
plt.legend()
# plt.savefig('assignment1a.png')
# plt.savefig('assignment1d.png')
plt.savefig('assignment1g.png')
plt.show()
# Plot the received data and frame start
# plt.figure
```

```
# plt.plot(rx_aligned)
# plt.plot(frame_ind, rx_aligned[frame_ind], 'x', label=f'frame_ind={frame_
# plt.title('demodulated rx signal')
# plt.grid(True)
# plt.legend()
# plt.show()
# Separate preamble and data symbols
rx_preamble = rx_aligned[frame_ind:frame_ind+2*N]
rx_data = rx_aligned[frame_ind+3*N:frame_ind+3*N+num_data_symbols]
# Estimate CFO
# Calculate delta_phi
products = np.conj(rx_preamble[:N]) * rx_preamble[N:2*N]
sum_product = np.sum(products) / len(products)
delta phi = np.angle(sum product)
cfo_hat = delta_phi / (2*np.pi*N*Tb)
# Compare the estimation with the ground truth
print("cfo_hz:", cfo_hz)
# print("cfo_hat: ", cfo_hat)
print("cfo_hat:", cfo_hat)
# Correct CFO
t_cfo = np.arange(len(rx_aligned)) * Tb
rx_cfo_corrected = rx_aligned * np.exp(-1j*2*np.pi*cfo_hat*t_cfo)
rx_preamble_CFOcor = rx_cfo_corrected[frame_ind:frame_ind+2*N]
rx_data_CFOcor = rx_cfo_corrected[frame_ind+3*N:frame_ind+3*N+num_data_syml
# Create the IQ constellation
plt.figure
plt.subplot(1,2,1)
plt.plot(rx_preamble.real, rx_preamble.imag, '.')
plt.xlabel('I')
plt.ylabel('Q')
plt.title('Preamble_sym_Before_CFO_corr')
plt.grid(True)
plt.xlim([-1.2, 1.2])
plt.ylim([-1.2, 1.2])
```

```
ax = plt.gca()
ax.set aspect('equal', adjustable='box')
plt.subplot (1,2,2)
plt.plot(rx_data.real, rx_data.imag, '.')
plt.xlabel('I')
plt.ylabel('Q')
plt.title('Data_sym_Before_CFO_corr')
plt.grid(True)
plt.xlim([-1.2, 1.2])
plt.ylim([-1.2, 1.2])
ax = plt.gca()
ax.set_aspect('equal', adjustable='box')
# plt.savefig('assignment1b.png')
# plt.savefig('assignment1e.png')
plt.savefig('assignment1h.png')
plt.show()
# Create the IQ constellation
plt.figure
plt.subplot(1,2,1)
plt.plot(rx_preamble_CFOcor.real, rx_preamble_CFOcor.imag, '.')
plt.xlabel('I')
plt.ylabel('Q')
plt.title('Preamble_sym_After_CFO_corr')
plt.grid(True)
plt.xlim([-1.2, 1.2])
plt.ylim([-1.2, 1.2])
ax = plt.gca()
ax.set aspect('equal', adjustable='box')
plt.subplot(1,2,2)
plt.plot(rx data CFOcor.real, rx data CFOcor.imag, '.')
plt.xlabel('I')
plt.ylabel('Q')
plt.title('Data_sym_After_CFO_corr')
plt.grid(True)
plt.xlim([-1.2, 1.2])
plt.ylim([-1.2, 1.2])
ax = plt.gca()
ax.set_aspect('equal', adjustable='box')
# plt.savefig('assignment1c.png')
# plt.savefig('assignment1f.png')
plt.savefig('assignment1i.png')
plt.show()
```

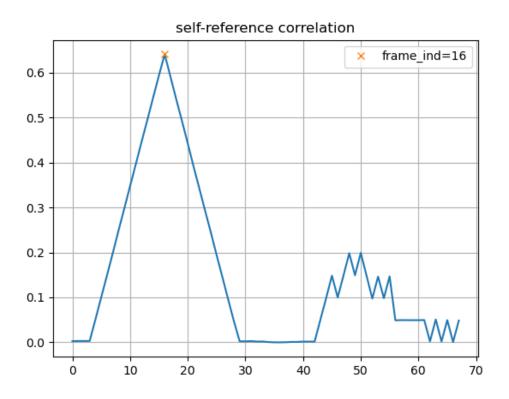


Figure 1: self-reference corr: cfo_hz = cfo_limit*0.05

SERIAL-OUT self-reference corr: cfo_hz = cfo_limit*0.05

CFO limit (Hz): +- 0.038461538461538464 Applied CFO (Hz): 0.0019230769230769232

max_ind: 8

cfo_hz: 0.0019230769230769232 cfo_hat: 0.001987514744180262

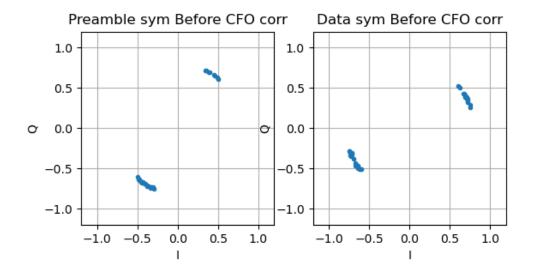


Figure 2: Preamble IQ Plot: cfo_hz = cfo_limit*0.05

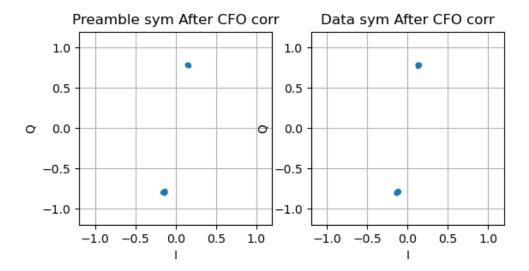


Figure 3: Data Symbols IQ Plot: cfo_hz = cfo_limit*0.05

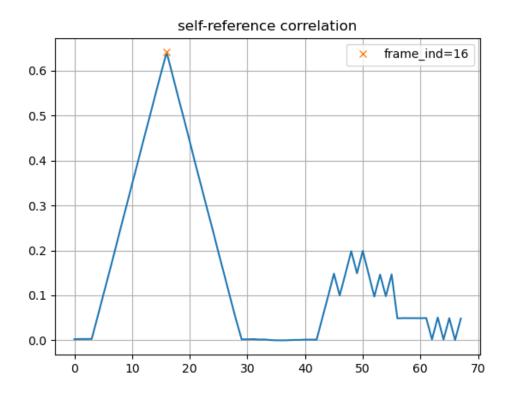


Figure 4: self-reference corr: cfo_hz = -cfo_limit*0.05

SERIAL-OUT self-reference corr: cfo_hz = -cfo_limit*0.05

CFO limit (Hz): +- 0.038461538461538464 Applied CFO (Hz): -0.0019230769230769232

max_ind: 8

cfo_hz: -0.0019230769230769232 cfo_hat: -0.001858604409919798

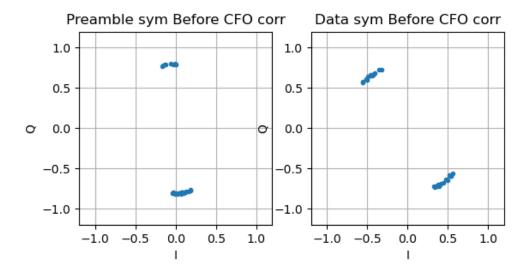


Figure 5: Preamble IQ Plot: cfo_hz = -cfo_limit*0.05

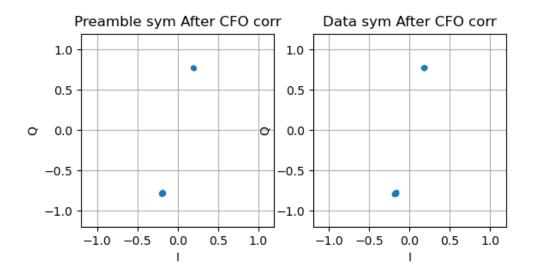


Figure 6: Data Symbols IQ Plot: -cfo_hz = cfo_limit*0.05

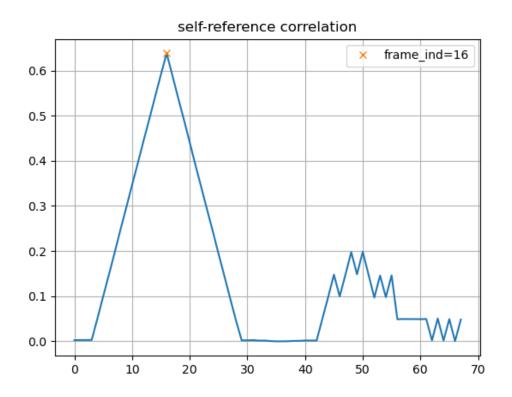


Figure 7: self-reference corr: cfo_hz = cfo_limit*1.05

SERIAL-OUT self-reference corr: cfo hz = cfo limit*1.05

CFO limit (Hz): +- 0.038461538461538464 Applied CFO (Hz): 0.04038461538461539

max_ind: 8

cfo_hz: 0.04038461538461539 cfo_hat: -0.036474784731455996

<clearly failed!!>this CFO correction failed because the channel error causes data
to perform a full rotation around the polar axis.

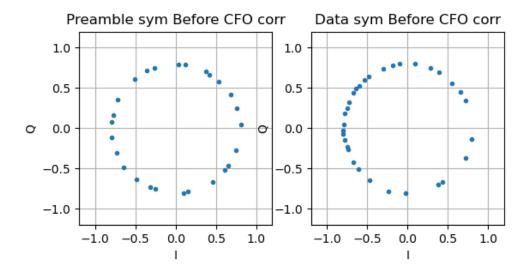


Figure 8: Preamble IQ Plot: cfo_hz = cfo_limit*1.05

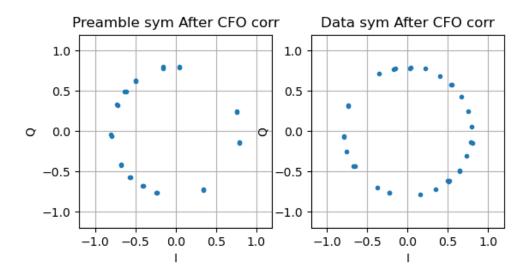


Figure 9: Data Symbols IQ Plot: cfo_hz = cfo_limit*1.05

```
plt.xlabel('I')
plt.ylabel('Q')
plt.title('Preamble_sym_After_CH_corr')
plt.grid(True)
plt.xlim([-1.2, 1.2])
plt.ylim([-1.2, 1.2])
ax = plt.gca()
ax.set_aspect('equal', adjustable='box')
plt.subplot(1,2,2)
plt.plot(rx_data_CHcor.real, rx_data_CHcor.imag, '.')
plt.xlabel('I')
plt.ylabel('Q')
plt.title('Data_sym_After_CH_corr')
plt.grid(True)
plt.xlim([-1.2, 1.2])
plt.ylim([-1.2, 1.2])
ax = plt.gca()
ax.set aspect('equal', adjustable='box')
# plt.savefig('assignment2a.png')
# plt.savefig('assignment2b.png')
plt.savefig('assignment2c.png')
plt.show()
# Recover the bits
rx bits = np.zeros(len(rx data CHcor), dtype=int)
for k in range(len(rx_data_CHcor)):
if rx_data_CHcor[k].real > 0: # BPSK
rx bits[k] = 1
else:
rx bits[k] = 0
# print("rx_data: ", rx_data)_
print("original_bits:_", bits)
print("rx bits:", rx bits)
# bit error
bit_err = np.sum( np.abs(bits - rx_bits) )
print("bit_error:_", bit_err)
print("bit_error_rate_(%):_", 100*bit_err/num_data_symbols)
```

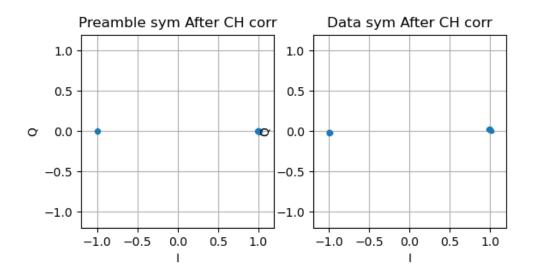


Figure 10: self-reference corr: cfo_hz = cfo_limit*0.05

SERIAL-OUT self-reference corr: cfo hz = cfo limit*0.05

CFO limit (Hz): +- 0.038461538461538464 Applied CFO (Hz): 0.0019230769230769232

max ind: 8

cfo_hz: 0.0019230769230769232 cfo_hat: -0.001987514744180262

h_hat: (-0.1453780186634645-0.7872008939992515j)

bit error: 0 bit error rate (

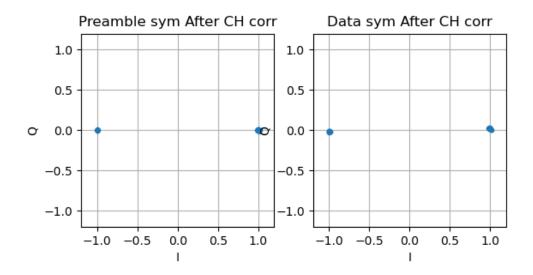


Figure 11: self-reference corr: cfo_hz = -cfo_limit*0.05

SERIAL-OUT self-reference corr: cfo hz = -cfo limit*0.05

CFO limit (Hz): +- 0.038461538461538464 Applied CFO (Hz): -0.0019230769230769232

max ind: 8

cfo_hz: -0.0019230769230769232 cfo_hat: 0.001858604409919798

bit error: 0

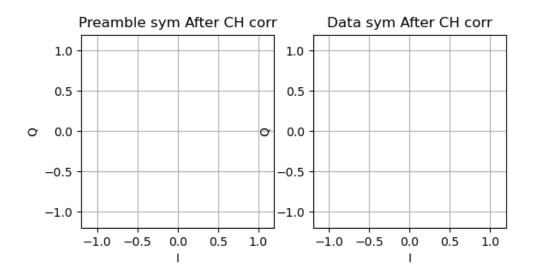


Figure 12: self-reference corr: cfo_hz = cfo_limit*1.05

SERIAL-OUT self-reference corr: $cfo_hz = cfo_limit*1.05$

CFO limit (Hz): +- 0.038461538461538464 Applied CFO (Hz): 0.04038461538461539

max ind: 8

cfo_hz: 0.04038461538461539 cfo_hat: 0.036474784731455996

h_hat: (0.00041507606418906146+0.0026299981006439386j)

bit error: 16 bit error rate (