lab5

May 14, 2024

1 Quantization - Mateusz Kliś

1.1 Variant 5 - $\Omega c = \tan(t)$, t [-1, 1]

```
[10]: # imports
     import numpy as np
     import matplotlib as mpl
     import matplotlib.pyplot as plt
     from scipy import signal
     !pip install soundfile
     import soundfile as sf # requires 'pip install soundfile'
    Collecting soundfile
      Downloading soundfile-0.12.1-py2.py3-none-win_amd64.whl.metadata (14 kB)
    Requirement already satisfied: cffi>=1.0 in
    c:\users\klism\appdata\roaming\python\python312\site-packages (from soundfile)
    (1.16.0)
    Requirement already satisfied: pycparser in
    c:\users\klism\appdata\roaming\python\python312\site-packages (from
    cffi>=1.0->soundfile) (2.21)
    Downloading soundfile-0.12.1-py2.py3-none-win_amd64.whl (1.0 MB)
       ----- 0.0/1.0 MB ? eta -:--:-
       ----- 0.0/1.0 MB ? eta -:--:-
       ----- 0.2/1.0 MB 2.6 MB/s eta 0:00:01
       ----- 0.6/1.0 MB 5.5 MB/s eta 0:00:01
       ----- 1.0/1.0 MB 8.0 MB/s eta 0:00:00
    Installing collected packages: soundfile
    Successfully installed soundfile-0.12.1
[11]: # quantization function
     def my_quant(x, Q):
        r"""Saturated uniform midtread quantizer
        input:
        x input signal
        Q number of quantization steps
        output:
```

```
xq quantized signal
Note: for even Q in order to retain midtread characteristics,
we must omit one quantization step, either that for lowest or the highest
amplitudes. Typically the highest signal amplitudes are saturated to
the 'last' quantization step. Then, in the special case of log2(N)
being an integer the quantization can be represented with bits.
tmp = Q//2 \# integer div
quant_steps = (np.arange(Q) - tmp) / tmp # we don't use this
# forward quantization, round() and inverse quantization
xq = np.round(x*tmp) / tmp
# always saturate to -1
xq[xq < -1.] = -1.
# saturate to ((Q-1) - (Q \setminus 2)) / (Q \setminus 2), note that \ is integer div
tmp2 = ((Q-1) - tmp) / tmp # for odd N this always yields 1
xq[xq > tmp2] = tmp2
return xq
```

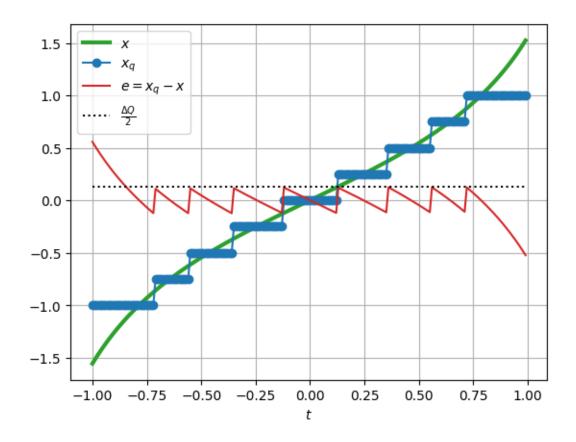
```
[12]: # cross corelation function
      def my_xcorr2(x, y, scaleopt='none'):
         r""" Cross Correlation function phixy[kappa] -> x[k+kappa] y
          input:
         x input signal shifted by +kappa
         y input signal
         scaleopt scaling of CCF estimator
          output:
         kappa sample index
              correlation result
          ccf
         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 N == M:
              if scaleopt == 'none' or scaleopt == 'raw':
              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
```

```
[13]: # midtread quantizer function
      def uniform_midtread_quantizer(x, deltaQ):
          r"""uniform_midtread_quantizer from the lecture:
          https://qithub.com/spatialaudio/diqital-siqnal-processing-lecture/blob/
       ⇒master/quantization/linear_uniform_quantization_error.ipynb
          commit: b00e23e
          note: we renamed the second input to deltaQ, since this is what the variable
          actually represents, i.e. the quantization step size
          input:
          x input signal to be quantized
          deltaQ
                  quantization step size
          output:
               quantized signal
          xq
          11 11 11
          # [-1...1) amplitude limiter
          x = np.copy(x)
          idx = np.where(x <= -1)
          x[idx] = -1
          idx = np.where(x > 1 - deltaQ)
          x[idx] = 1 - deltaQ
          # linear uniform quantization
          xq = deltaQ * np.floor(x/deltaQ + 1/2)
          return xq
```

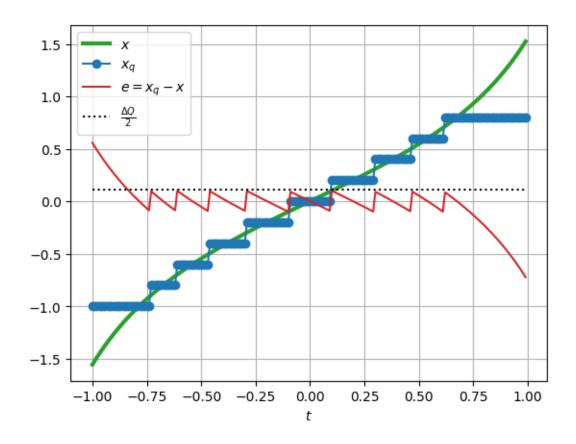
```
[14]: # signal quantization for Q = 9

Q = 9  # odd, number of quantization steps
t = np.arange(-1, 1, 0.01)
x = np.tan(t)
xq = my_quant(x, Q)
e = xq-x
# actually stem plots would be correct, for convenience we plot as line style
plt.plot(t, x, 'C2', lw=3, label=r'$x$')
plt.plot(t, xq, 'C0o-', label=r'$x_q$')
plt.plot(t, e, 'C3', label=r'$e=x_q-x$')
plt.plot(t, t*0+1/(Q-1), 'k:', label=r'$\frac{\Delta Q}{2}$')
plt.xlabel(r'$t$')
plt.legend()
plt.grid(True)
```



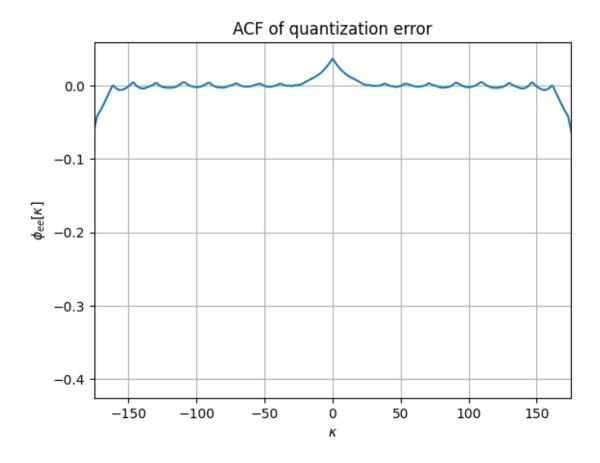
```
[15]: # signal quantization for Q = 10

Q = 10  # number of quantization steps
t = np.arange(-1, 1, 0.01)
x = np.tan(t)
xq = my_quant(x, Q)
e = xq-x
  # actually stem plots would be correct, for convenience we plot as line style
plt.plot(t, x, 'C2', lw=3, label=r'$x$')
plt.plot(t, xq, 'C00-', label=r'$x_q$')
plt.plot(t, e, 'C3', label=r'$e=x_q-x$')
plt.plot(t, t*0+1/(Q-1), 'k:', label=r'$\frac{\Delta Q}{2}$')
plt.xlabel(r'$t$')
plt.legend()
plt.grid(True)
```

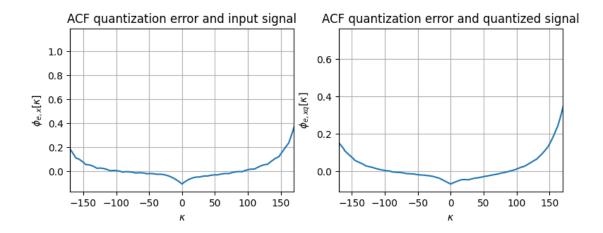


```
[16]: # Auto-correlation function

kappa, acf = my_xcorr2(e, e, 'unbiased')
plt.plot(kappa, acf)
plt.xlim(-175, +175)
plt.xlabel(r'$\kappa$')
plt.ylabel(r'$\kappa$')
plt.ylabel(r'$\phi_{ee}[\kappa]$')
plt.title('ACF of quantization error')
plt.grid(True)
```

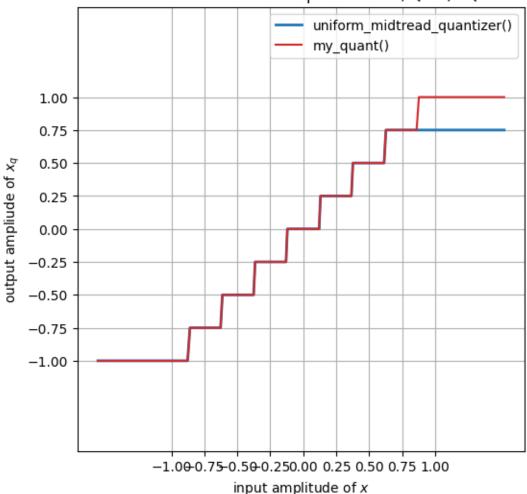


```
[17]: plt.figure(figsize=(9, 3))
      plt.subplot(1, 2, 1)
      kappa, acf = my_xcorr2(e, x, 'unbiased')
      plt.plot(kappa, acf)
      plt.xlim(-170, +170)
      plt.xlabel(r'$\kappa$')
      plt.ylabel(r'$\phi_{e,x}[\kappa]$')
      plt.title('ACF quantization error and input signal')
      plt.grid(True)
      plt.subplot(1, 2, 2)
      kappa, acf = my_xcorr2(e, xq, 'unbiased')
      plt.plot(kappa, acf)
      plt.xlim(-170, +170)
      plt.xlabel(r'$\kappa$')
      plt.ylabel(r'$\phi_{e,xq}[\kappa]$')
      plt.title('ACF quantization error and quantized signal')
      plt.grid(True)
```



```
[18]: \# midtread quantization for Q = 9
      Q = 9 # number of quantization steps, odd or even
      deltaQ = 1/(Q//2) # quantization step size, even/odd Q
      xq = my_quant(x, Q) # used in exercise
      xumq = uniform_midtread_quantizer(x, deltaQ) # as used in lecture
      plt.figure(figsize=(6, 6))
      plt.plot(x, xumq, 'CO', lw=2, label='uniform_midtread_quantizer()')
      plt.plot(x, xq, 'C3', label='my_quant()')
      plt.xticks(np.arange(-1, 1.25, 0.25))
      plt.yticks(np.arange(-1, 1.25, 0.25))
      plt.xlabel(r'input amplitude of $x$')
      plt.ylabel(r'output ampliude of $x_q$')
      plt.title(
          r'uniform saturated midtread quantization, Q={0:d}, $\Delta Q$={1:3.2f}'.
      →format(Q, deltaQ))
      plt.axis('equal')
      plt.legend()
      plt.grid(True)
```

uniform saturated midtread quantization, Q=9, ΔQ =0.25



```
def check_my_quant(Q):
    N = 5e2
    x = 2*np.arange(N)/N - 1
    xq = my_quant(x, Q)
    e = xq - x

plt.plot(x, x, color='C2', lw=3, label=r'$x[k]$')
    plt.plot(x, xq, color='C3', label=r'$x_q[k]$')
    plt.plot(x, e, color='C0', label=r'$e[k] = x_q[k] - x[k]$')
    plt.xticks(np.arange(-1, 1.25, 0.25))
    plt.yticks(np.arange(-1, 1.25, 0.25))
    plt.xlabel('input amplitude')
    plt.ylabel('output amplitude')
```

```
if np.mod(Q, 2) == 0:
    s = ' saturated '
else:
    s = ' '
plt.title(
        'uniform'+s+'midtread quantization with Q=%d steps, $\Delta Q$=%4.3e' %\_
$\(\text{Q}\), 1/(\text{Q}/\(2)\))
plt.axis('equal')
plt.legend(loc='upper left')
plt.grid(True)
```

```
<>:21: SyntaxWarning: invalid escape sequence '\D'
<>:21: SyntaxWarning: invalid escape sequence '\D'
C:\Users\klism\AppData\Local\Temp\ipykernel_21664\1586064171.py:21:
SyntaxWarning: invalid escape sequence '\D'
   'uniform'+s+'midtread quantization with Q=%d steps, $\Delta Q$=%4.3e' % (Q, 1/(Q//2)))
```

```
[20]: Q = 10 # number of quantization steps

deltaQ = 1 / (Q//2) # general rule

deltaQ = 2 / (Q-1) # for odd Q only

plt.figure(figsize=(5, 5))

check_my_quant(Q)
```

uniform saturated midtread quantization with Q=10 steps, ΔQ =2.000e-01

