

lab1

March 12, 2024

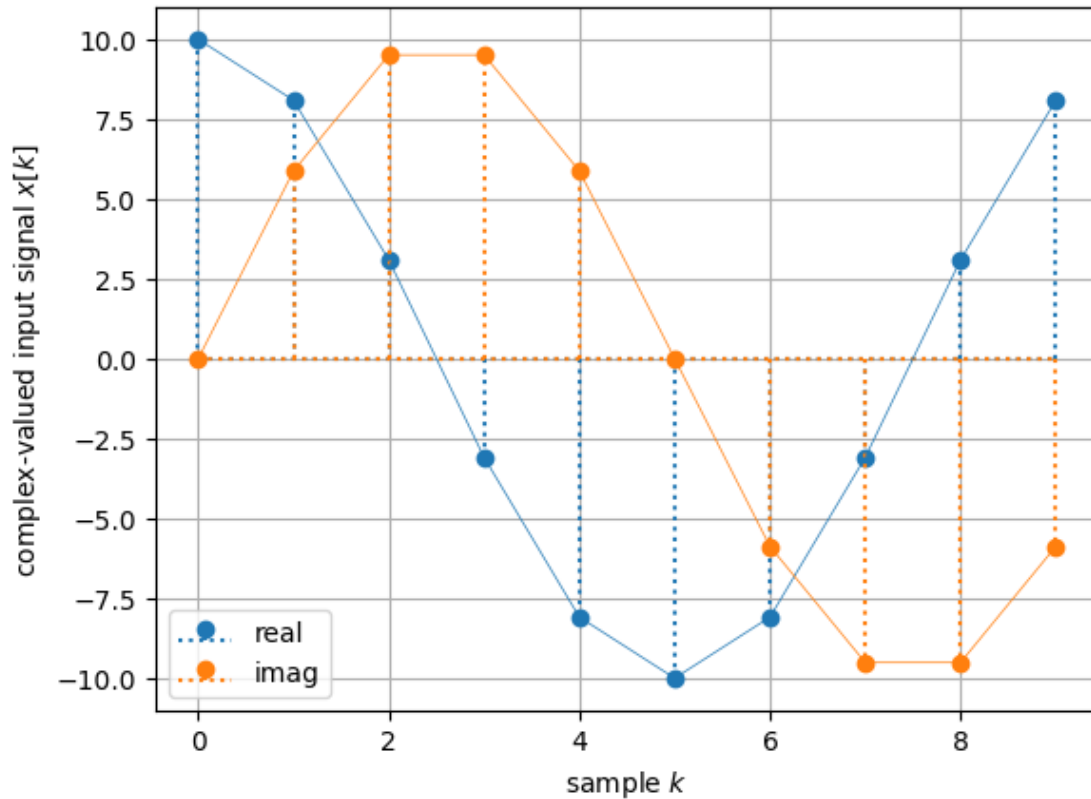
```
[1]: import numpy as np
import matplotlib.pyplot as plt
from numpy.linalg import inv
from numpy.fft import fft, ifft
#from scipy.fft import fft, ifft

[2]: N = 10 # signal block length
k = np.arange(N) # all required sample/time indices
A = 10 # signal amplitude

tmpmu = 2-1/2 # DFT eigenfrequency worst case
tmpmu = 1 # DFT eigenfrequency best case

x = A * np.exp(tmpmu * +1j*2*np.pi/N * k)

# plot
plt.stem(k, np.real(x), markerfmt='C0o',
         baselfmt='C0:', linefmt='C0:', label='real')
plt.stem(k, np.imag(x), markerfmt='C1o',
         baselfmt='C1:', linefmt='C1:', label='imag')
plt.plot(k, np.real(x), 'C0-', lw=0.5)
plt.plot(k, np.imag(x), 'C1-', lw=0.5)
plt.xlabel(r'sample $k$')
plt.ylabel(r'complex-valued input signal $x[k]$')
plt.legend()
plt.grid(True)
```



```
[3]: # DFT with for-loop:
X_ = np.zeros((N, 1), dtype=complex) # alloc RAM, init with zeros
for mu_ in range(N): # do for all DFT frequency indices
    for k_ in range(N): # do for all sample indices
        X_[mu_] += x[k_] * np.exp(-1j*2*np.pi/N*k_*mu_)

[4]: # IDFT with for-loop:
x_ = np.zeros((N, 1), dtype=complex) # alloc RAM, init with zeros
for k_ in range(N):
    for mu_ in range(N):
        x_[k_] += X_[mu_] * np.exp(+1j*2*np.pi/N*k_*mu_)
x_ *= 1/N # normalization in the IDFT stage

[5]: # k = np.arange(N) # all required sample/time indices, already defined above
# all required DFT frequency indices, actually same entries like in k
mu = np.arange(N)

# set up matrices
K = np.outer(k, mu) # get all possible entries k*mu in meaningful arrangement
W = np.exp(+1j * 2*np.pi/N * K) # analysis matrix for DFT
```

```

[6]: # visualize the content of the Fourier matrix
# we've already set up (use other N if desired):
# N = 10
# k = np.arange(N)
# mu = np.arange(N)
# W = np.exp(+1j*2*np.pi/N*np.outer(k, mu)) # set up Fourier matrix

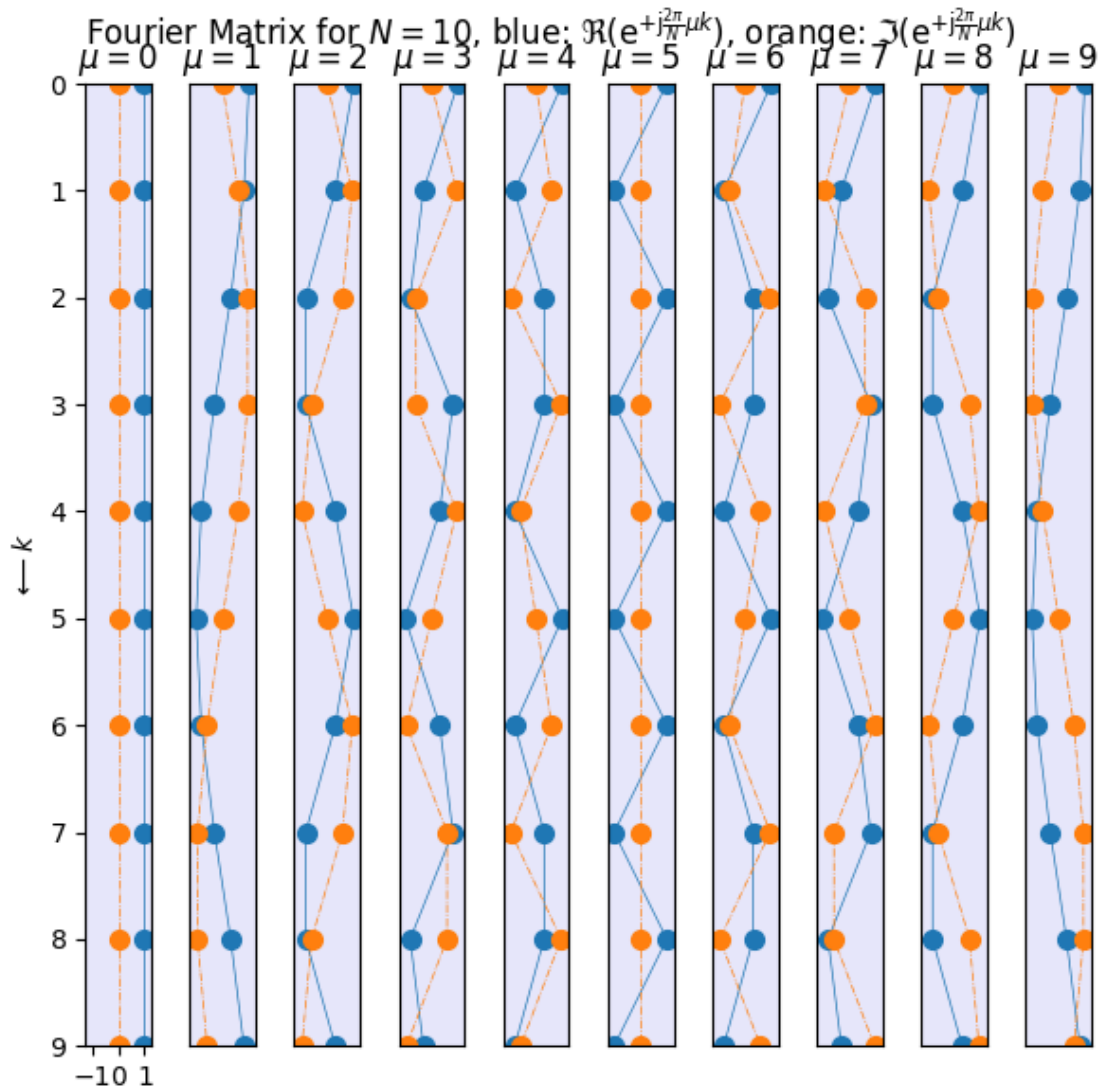
fig, ax = plt.subplots(1, N)
fig.set_size_inches(6, 6)
fig.suptitle(
    r'Fourier Matrix for $N=%d$, blue: $\mathrm{Re}(\mathrm{e}^{+ \mathrm{j} \frac{2\pi}{N} \mu k})$, orange: $\mathrm{Im}(\mathrm{e}^{+ \mathrm{j} \frac{2\pi}{N} \mu k})$' % N)

for tmp in range(N):
    ax[tmp].set_facecolor('lavender')
    ax[tmp].plot(W[:, tmp].real, k, 'C0o-', ms=7, lw=0.5)
    ax[tmp].plot(W[:, tmp].imag, k, 'C1o-', ms=7, lw=0.5)
    ax[tmp].set_ylim(N-1, 0)
    ax[tmp].set_xlim(-5/4, +5/4)
    if tmp == 0:
        ax[tmp].set_yticks(np.arange(0, N))
        ax[tmp].set_xticks(np.arange(-1, 1+1, 1))
        ax[tmp].set_ylabel(r'$\longrightarrow k$')
    else:
        ax[tmp].set_yticks([], minor=False)
        ax[tmp].set_xticks([], minor=False)
    ax[tmp].set_title(r'$\mu=%d$' % tmp)
fig.tight_layout()
fig.subplots_adjust(top=0.91)

fig.savefig('fourier_matrix.png', dpi=300)

# TBD: row version for analysis

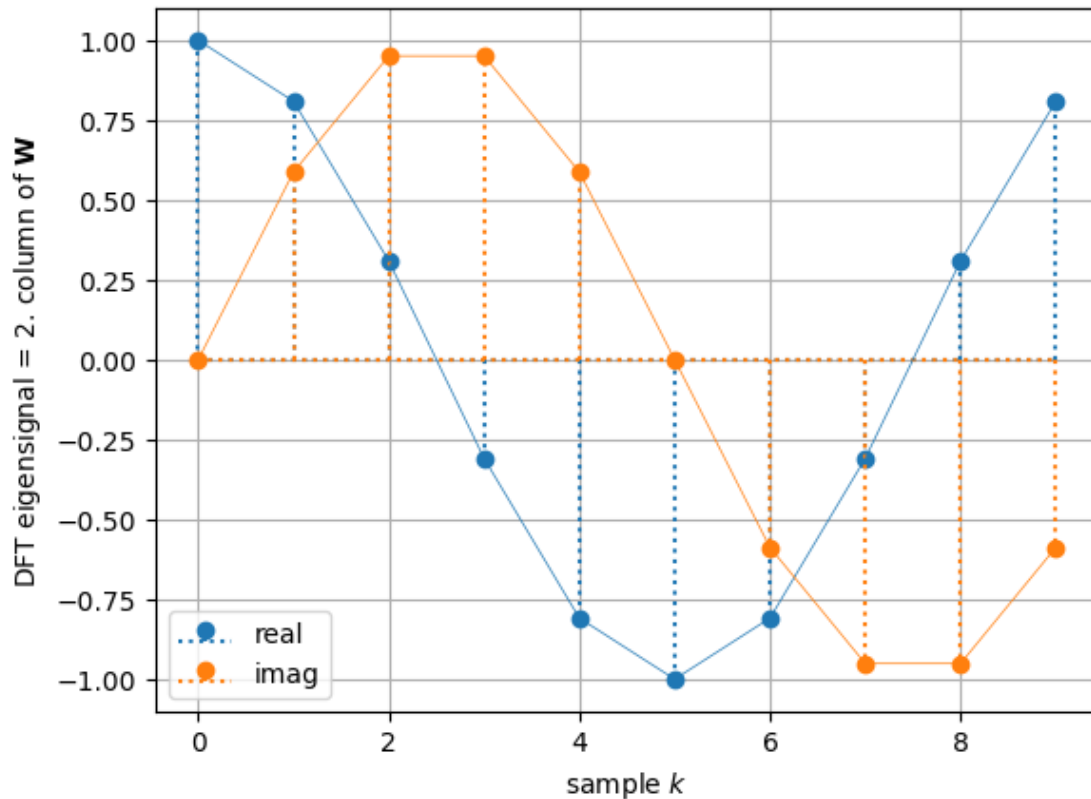
```



```
[7]: tmpmu = 1 # column index

plt.stem(k, np.real(W[:, tmpmu]), label='real',
         markerfmt='C0o', basefmt='C0:', linefmt='C0:')
plt.stem(k, np.imag(W[:, tmpmu]), label='imag',
         markerfmt='C1o', basefmt='C1:', linefmt='C1:')
# note that connecting the samples by lines is actually wrong, we
# use it anyway for more visual convenience
plt.plot(k, np.real(W[:, tmpmu]), 'C0-', lw=0.5)
plt.plot(k, np.imag(W[:, tmpmu]), 'C1-', lw=0.5)
plt.xlabel(r'sample $k$')
plt.ylabel(r'DFT eigensignal = '+str(tmpmu+1)+' column of $\mathbf{W}$')
plt.legend()
```

```
plt.grid(True)
```



```
[8]: np.dot(np.conj(W[:, 0]), W[:, 0]) # same eigensignal, same eigenfrequency
      # np.vdot(W[:,0],W[:,0]) # this is the suitable numpy function
```

```
[8]: (10+0j)
```

```
[9]: np.dot(np.conj(W[:, 0]), W[:, 1]) # different eigensignals
      # np.vdot(W[:,0],W[:,1]) # this is the suitable numpy function
      # result should be zero, with numerical precision close to zero:
```

```
[9]: (-3.3306690738754696e-16+0j)
```

```
[10]: if N == 10:
        X_test = np.array([6, 2, 4, 3, 4, 5, 0, 0, 0, 0])
        # x_test = 1/N*W@X_test # >= Python3.5
        x_test = 1/N * np.matmul(W, X_test)

        plt.stem(k, np.real(x_test), label='real',
                  markerfmt='C0o', basefmt='C0:', linefmt='C0:')
        plt.stem(k, np.imag(x_test), label='imag',
```

```

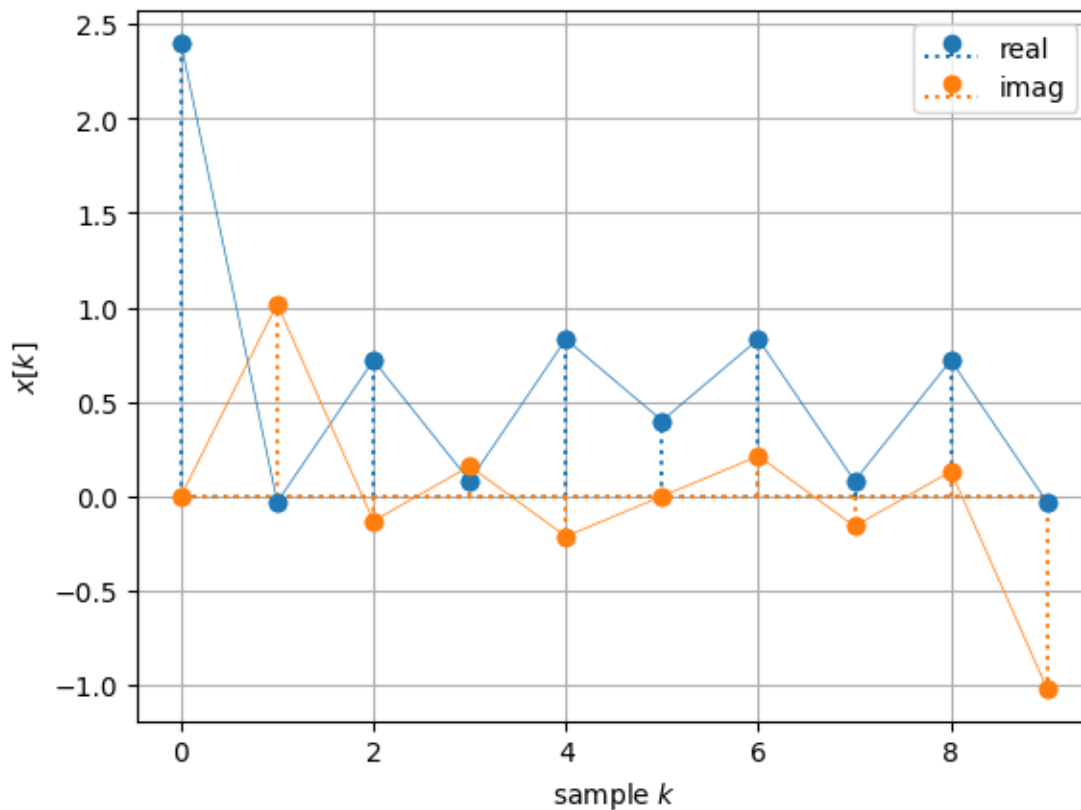
        markerfmt='C1o', basefmt='C1:', linefmt='C1:')
# note that connecting the samples by lines is actually wrong, we
# use it anyway for more visual convenience
plt.plot(k, np.real(x_test), 'C0o-', lw=0.5)
plt.plot(k, np.imag(x_test), 'C1o-', lw=0.5)
plt.xlabel(r'sample $k$')
plt.ylabel(r'$x[k]$')
plt.legend()
plt.grid(True)

# check if results are identical with numpy ifft package
print(np.allclose(fft(X_test), x_test))
print('DC is 1 as expected: ', np.mean(x_test))

```

True

DC is 1 as expected: (0.6+6.66133814775094e-17j)



```

[11]: if N == 10:
        x_test2 = X_test[0] * W[:, 0] + X_test[1] * W[:, 1] + X_test[2] * W[:, 2]

```

```
[12]: if N == 10:
      x_test2 *= 1/N
      print(np.allclose(x_test, x_test2))  # check with result before
```

False

```
[13]: if N == 10:
      # X_test2 = np.conj(W)@x_test  # >= Python3.5
      X_test2 = np.matmul(np.conj(W), x_test)  # DFT, i.e. analysis
      print(np.allclose(X_test, X_test2))  # check with result before
```

True

```
[14]: if N == 10:
      print(np.allclose(fft(x_test), X_test))
```

True

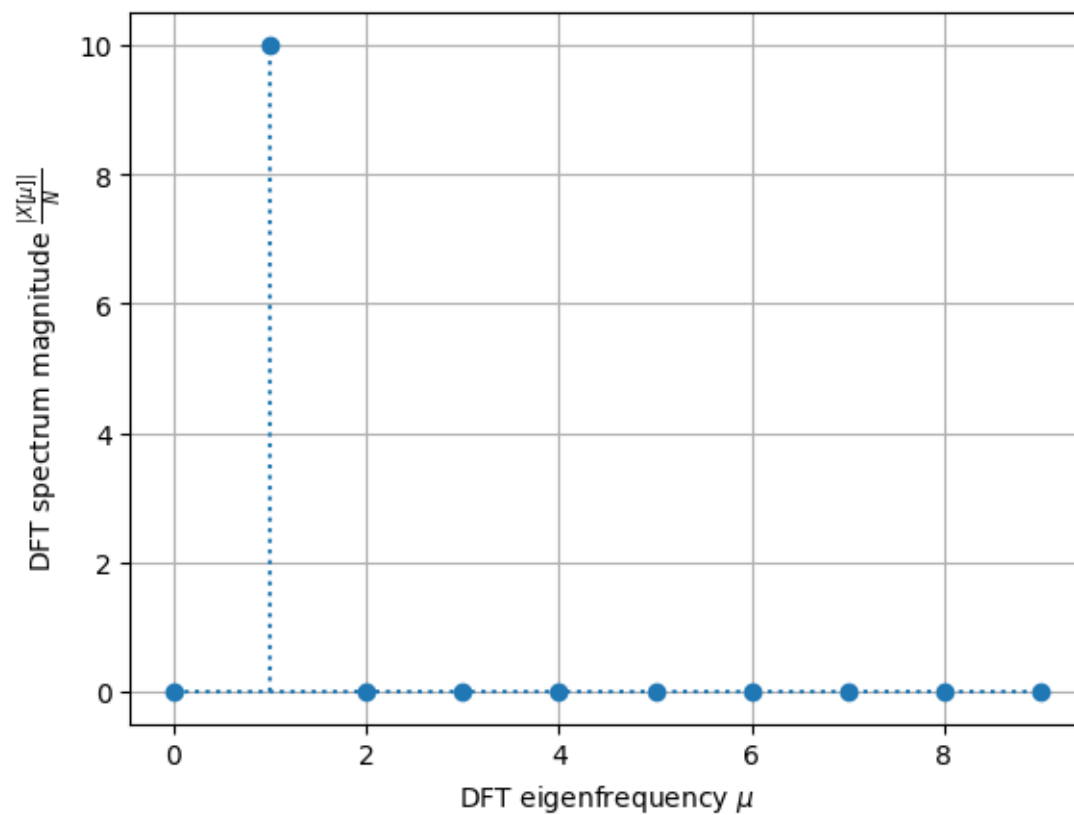
```
[15]: if N == 10:
      print(np.conj(W[:, 0])@x_test)
      print(np.conj(W[:, 1])@x_test)
      print(np.conj(W[:, 2])@x_test)
```

```
(6+6.661338147750939e-16j)
(1.9999999999999998+4.440892098500626e-16j)
(3.9999999999999996+0j)
```

```
[16]: X = fft(x)
      # print(np.allclose(np.conj(W)@x, X))  # >=Python 3.5
      print(np.allclose(np.matmul(np.conj(W), x), X))
```

True

```
[17]: plt.stem(mu, np.abs(X)/N, markerfmt='Co', basefmt='C:', linefmt='C:')
      # plt.plot(mu, np.abs(X)/N, 'C', lw=1)  # this is here a misleading plot and
      # hence not used
      plt.xlabel(r'DFT eigenfrequency $\mu$')
      plt.ylabel(r'DFT spectrum magnitude $\frac{|X[\mu]|}{N}$')
      plt.grid(True)
```



[]: