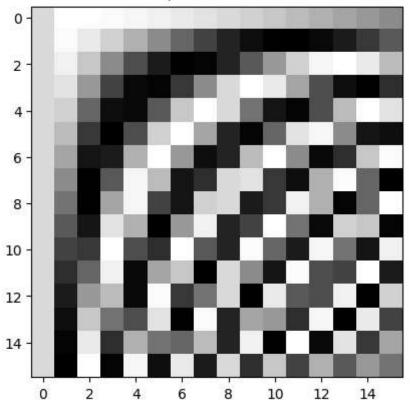
Here I will show how well DCT performs on spatial decorrelation using one segment of the signal. I also illustrate the effects of removing the lowest highest frequency coefficients.

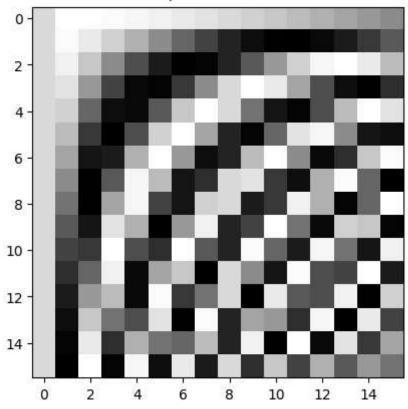
```
In [ ]: import numpy as np
        import matplotlib.pyplot as plt
        from scipy.fftpack import idct
        import warnings
        warnings.simplefilter("ignore")
        s = np.load('../test_data_sintef.npy')
In [ ]: # generate DCT basis
        pp = s.shape[1]
        D = np.zeros((pp, pp))
        # generate DCT basis using library
        for k in range(pp):
            a = np.zeros(pp)
            a[k] = 1
            D[:, k] = idct(a, norm='ortho') # use idct on the atom
        plt.figure()
        plt.imshow(D, cmap='gray')
        plt.title('DCT basis, atoms in the column')
        # generate DCT basis using equation
        DCT = np.zeros((pp, pp))
        c_0 = np.sqrt(1/pp)
        c_k = np.sqrt(2/pp)
        for k in range(pp):
            if (k == 0):
                DCT[:, k] = [c_0*np.cos(k*np.pi*(2*n+1)/(2*pp)) for n in range(pp)]
                DCT[:, k] = [c_k*np.cos(k*np.pi*(2*n+1)/(2*pp)) for n in range(pp)]
            DCT[:, k] = DCT[:, k] / np.linalg.norm(DCT[:, k], ord = 2)
        plt.figure()
        plt.imshow(DCT, cmap='gray')
        plt.title('DCT basis, atoms in the column')
```

Out[]: Text(0.5, 1.0, 'DCT basis, atoms in the column')

DCT basis, atoms in the column



DCT basis, atoms in the column

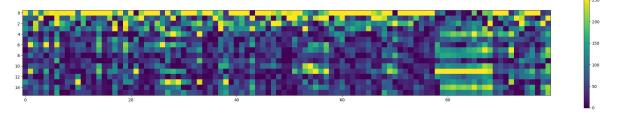


We see that the two DCT basis are identical. Next we compute the coefficients

```
c[:, i] = D.T@temp[i,:]
return c
```

```
In [ ]: temp = s.astype(int)
        # function to normalize each column of the dct coefficients
        def visualize_dct(c):
            for i in range(c.shape[0]):
                c[i,:] = 255* (c[i,:] - c[i,:].min()) / (c[i,:].max() - c[i,:].min())
            return c
        # perform DCT
        c = getDCT(temp, DCT)
        # reduce the number of columns for better visualization
        c_new = np.zeros((c.shape[0], M))
        step = c.shape[1] // M
        for i in range(M):
            c_{new}[:,i] = np.mean(c[:,i*step:i*step + step], axis=1)
        # visulaize the DCT coefficients
        plt.figure(figsize=[30,5])
        plt.imshow(visualize_dct(abs(c_new).T).T)
        plt.colorbar()
```

Out[]: <matplotlib.colorbar.Colorbar at 0x215ff65f490>



We see that most of the information is contained in the lowest frequencies, but that there still is some important information in higher frequencies too.

We then test on just the tetrode channels:

```
In []: temp = s[:,:12].astype(int)

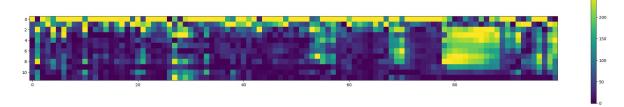
# generate DCT basis
pp = temp.shape[1]

# compute new DCT basis that is 12x12

DCT = np.zeros((pp, pp))
c_0 = np.sqrt(1/pp)
c_k = np.sqrt(2/pp)
for k in range(pp):
    if (k == 0):
        DCT[:, k] = [c_0*np.cos(k*np.pi*(2*n+1)/(2*pp)) for n in range(pp)]
    else:
        DCT[:, k] = [c_k*np.cos(k*np.pi*(2*n+1)/(2*pp)) for n in range(pp)]
    DCT[:, k] = DCT[:, k] / np.linalg.norm(DCT[:, k], ord = 2)

# perform DCT
```

Out[]: <matplotlib.colorbar.Colorbar at 0x215ff6b5c10>



We see that the result is more sparse, as expected

Threshold coefficients

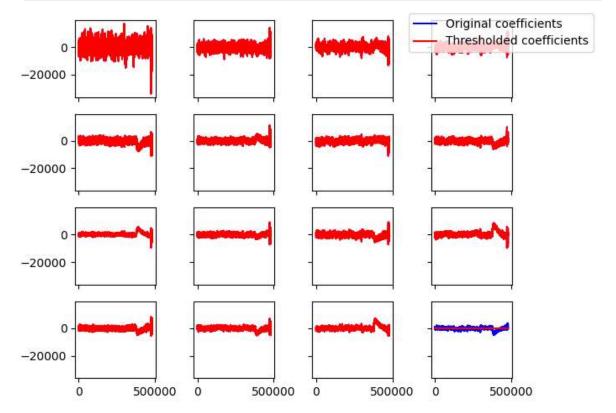
Threshold to see how much loss we get from removing the highest frequency coefficients

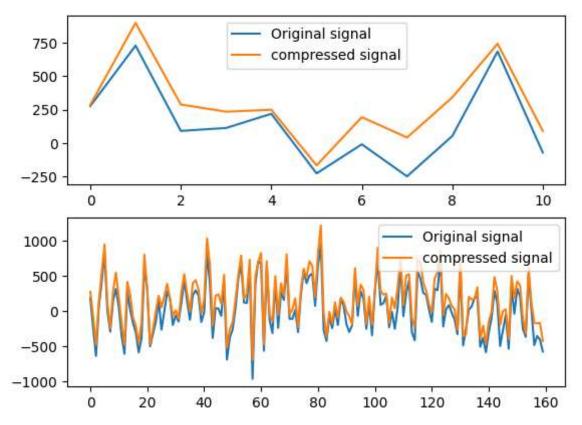
```
In [ ]: # threshold coefficients
        n_remove = 1
        temp = s.astype(int)
        # threshold coefficients
        c = getDCT(temp, D)
        # remove the n_remove highest frequencies
        c_{th} = c[:-n_{remove}, :]
        c_th_full = np.pad(c_th, ((0,n_remove), (0,0)), 'constant', constant_values=(0))
        # Visualize the coefficients
        fig, axs = plt.subplots(4, 4, sharey=True, sharex=True)
        for i in range(temp.shape[1]):
            axs.flatten()[i].plot(c[i, :], 'b')
            axs.flatten()[i].plot(c_th_full[i,:], 'r')
        fig.legend(['Original coefficients', 'Thresholded coefficients'], loc='upper rig
        plt.tight_layout()
        plt.show()
        # reconstuct signal
        temp_hat = (D@c_th_full).T
        # plot error
        fig2, axs2 = plt.subplots(2,1)
        axs2[0].plot(temp[140:151,5])
        axs2[0].plot(temp_hat[140:151,5])
```

```
axs2[0].legend(['Original signal', 'compressed signal'])
axs2[1].plot(temp[40:200,5])
axs2[1].plot(temp_hat[40:200,5])
axs2[1].legend(['Original signal', 'compressed signal'])
plt.show()

SNDR = 0
for i in range(temp.shape[1]):
    P_d = np.linalg.norm(temp[:,i])
    P_error = np.linalg.norm(temp[:,i] - temp_hat[:,i])

SNDR += 1/(temp.shape[1]) * 20*np.log10(P_d/P_error)
print(f'Mean SNDR across channels: {SNDR:.2f} dB')
```





Mean SNDR across channels: 16.90 dB