

Here I will show how well DCT performs on temporal 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

s = np.load('../test_data_sintef.npy')
```

## DCT temporal coding

### setup

```
In [ ]: # generate DCT basis
pp = 500

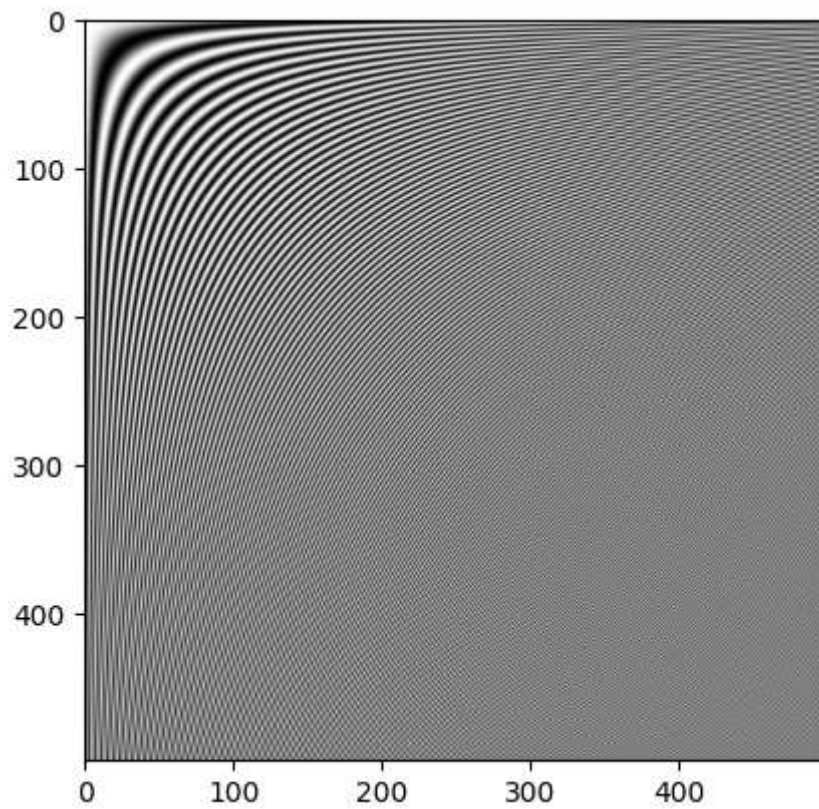
D = np.zeros((pp, pp))

# generate DCT basis
for k in range(pp):
    a = np.zeros(pp)
    a[k] = 1
    D[:, k] = idct(a, norm='ortho') # use idct on the atom

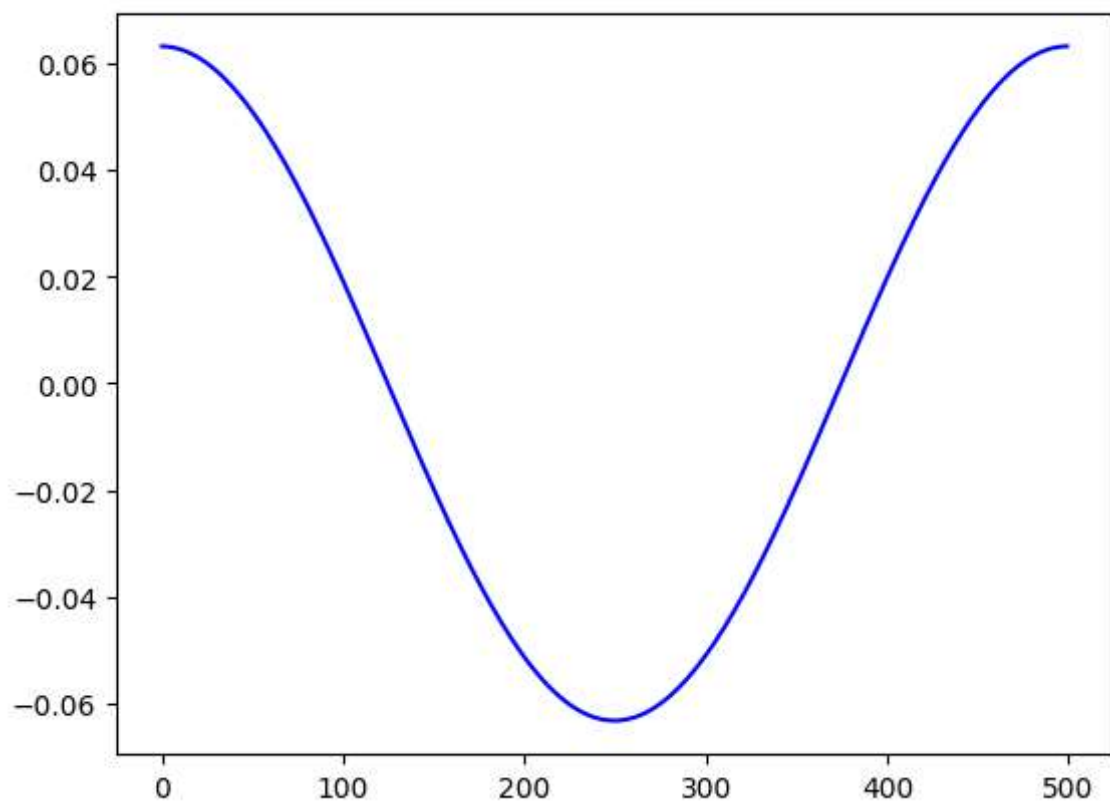
# plot DCT matrix
plt.figure()
plt.imshow(D, cmap='gray')

# plot one element from the DCT matrix
k = 2
plt.figure()
plt.plot(D[:, k], 'b')
plt.title(f'element: {k+1} from the DCT basis')
```

```
Out[ ]: Text(0.5, 1.0, 'element: 3 from the DCT basis')
```



element: 3 from the DCT basis



```
In [ ]: def getDCT(temp):
    nSteps = int(s.shape[0]/pp)
    c_mat = np.zeros((pp, nSteps, s.shape[1]))

    for j in range(s.shape[1]):
        for i in range(nSteps):
            c_mat[:,i,j] = D.T@temp[pp*i:pp*i + pp, j]
```

```

    return c_mat

def getiDCT(temp):
    nSteps = int(s.shape[0]/pp)
    d_hat = np.zeros_like(s)

    for j in range(s.shape[1]):
        for i in range(nSteps):
            d_hat[pp*i:pp*i + pp, j] = D@temp[:,i,j]

    return d_hat

```

## perform

```

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_mat = getDCT(temp).astype(int)

# flatten coefficient matrix
c = c_mat.reshape(c_mat.shape[0]*c_mat.shape[1], c_mat.shape[2]).T

# reduce the number of columns and rows for better visualization
M = 60
N = 50
c_new = np.zeros((N,M))
step_col = c_mat.shape[1] // M
step_row = c_mat.shape[0] // (N)
for j in range(N):
    for i in range(M):
        c_new[j,i] = np.mean(c_mat[j*step_row :j*step_row + step_row, i*step_col

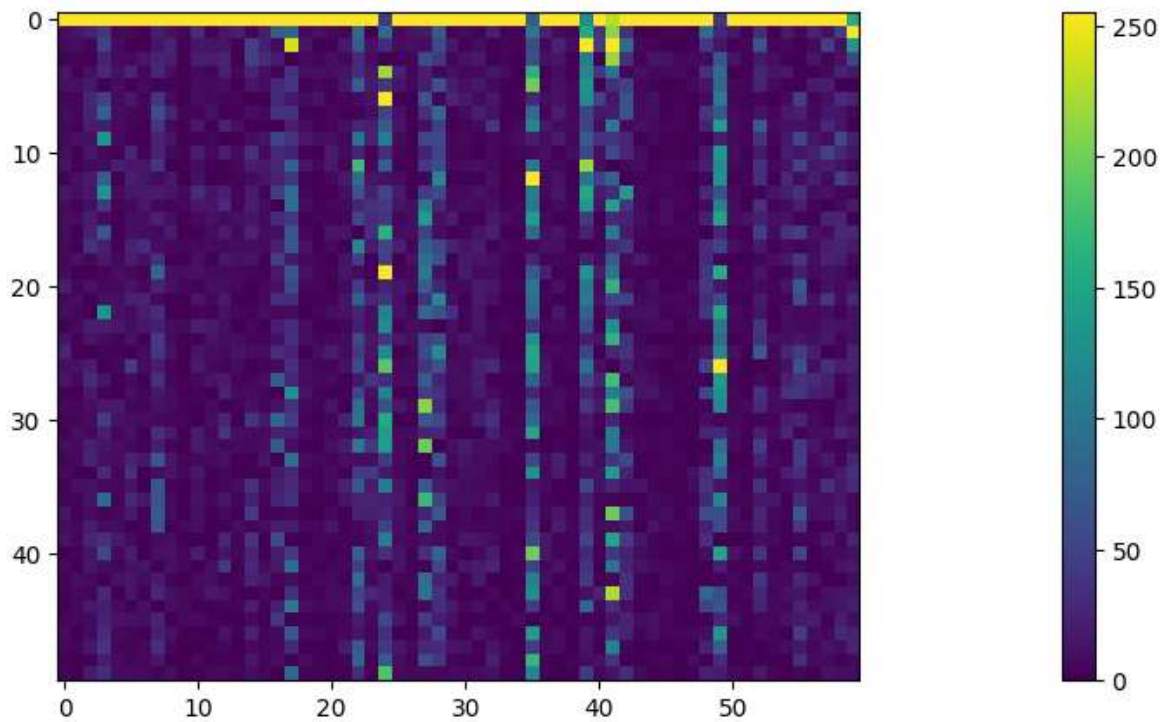
# visulaize the DCT coefficients
plt.figure(figsize=[30,5])
plt.imshow(visualize_dct(abs(c_new).T).T)
plt.colorbar()

```

```

Out[ ]: <matplotlib.colorbar.Colorbar at 0x1ea68782750>

```



this shows the coefficients of the first channel, we see that the DC offset has the biggest contribution.

## Threshold coefficients

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

```
In [ ]: # threshold coefficients
n_remove = 100
c_mat = getDCT(temp).astype(int)

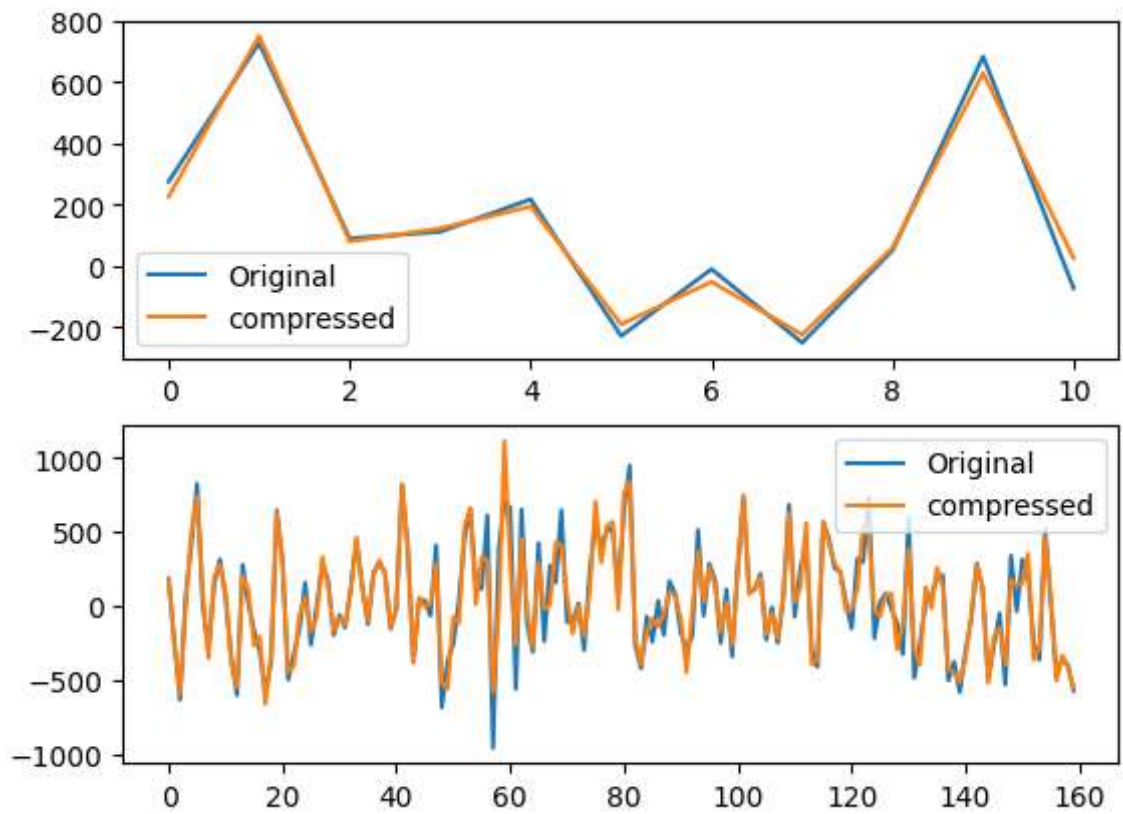
c_mat_th = c_mat[:-n_remove,:,:] # remove n_remove highest freqs

temp_hat = getIDCT(np.pad(c_mat_th, ((0,n_remove), (0,0), (0,0)), 'constant', cc

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

# compute SNDR
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: 17.94 dB