# alg\_OMP\_WND

## August 22, 2023

```
[]: import matplotlib.pyplot as plt
     # Improve the quality of figures
     plt.rcParams['figure.dpi'] = 300
     # Times New Roman font
     plt.rcParams["font.family"] = "Times New Roman"
     import numpy as np
     import networkx as nx
[ ]: N_RF = 500
     SNR = 5 \# dB
    Read Generated Channel Data
[]: | # read data from \generate_channel_data\generate_channel_data.mat
     import scipy.io as sio
     generate_channel_data = sio.loadmat('./data/generate_channel_data.mat')
     cluster_para = generate_channel_data['meta_data']['cluster_para']
     # xi = generate_channel_data['meta_data']['xi']
     f_c = generate_channel_data['meta_data']['f_c']
```

```
import scipy.io as sio
generate_channel_data = sio.loadmat('./data/generate_channel_data.mat')
cluster_para = generate_channel_data['meta_data']['cluster_para']
# xi = generate_channel_data['meta_data']['xi']
f_c = generate_channel_data['meta_data']['f_c']
N_x = int(generate_channel_data['meta_data']['N_x'])
N_y = int(generate_channel_data['meta_data']['N_y'])
lambda_c = generate_channel_data['meta_data']['lambda_c']
delta = generate_channel_data['meta_data']['delta']
N = int(generate_channel_data['meta_data']['L_x']
L_x = generate_channel_data['meta_data']['L_x']
L_y = generate_channel_data['meta_data']['L_y']

H_channel = np.mat(generate_channel_data['channel']['H_channel'][0][0])
vec_H_a = np.mat(generate_channel_data['channel']['vec_H_a'][0][0])
variance = np.mat(generate_channel_data['channel']['variance'][0][0])
Psi = np.mat(generate_channel_data['channel']['Psi'][0][0])
```

```
for l_x in range(-l_x_abs_max, l_x_abs_max + 1):
    # for start with -l_y_abs_max to l_y_abs_max
    for l_y in range(-l_y_abs_max, l_y_abs_max + 1):
        k_x = 2 * np.pi * l_x / L_x
        k_y = 2 * np.pi * l_y / L_y
        if k_x ** 2 + k_y ** 2 < (2 * np.pi / lambda_c) ** 2:
            xi.append((l_x, l_y))</pre>
```

 $l_x_abs_max = 32$  $l_y_abs_max = 32$ 

#### Visualize Channel

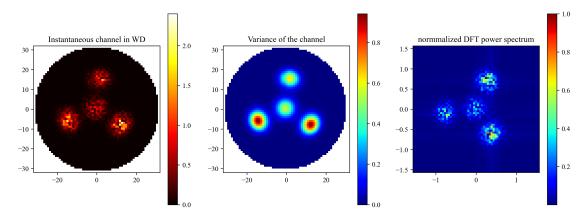
```
[]: # visualize the channel
     def visualize_wavenumber_domain_channel(vec_H_a):
         mat_H_a = np.mat(np.zeros((2*l_x_abs_max+1, 2*l_y_abs_max+1), ___)
      →dtype=complex))
         # all set to np.nan in mat_H_a
         mat_H_a[:,:] = np.nan
         for 1, l_pair in enumerate(xi):
             l_x = l_pair[0]
             l_y = l_pair[1]
             mat_H_a[l_x + l_x_abs_max, l_y + l_y_abs_max] = vec_H_a[l]
         # plt.figure(figsize=(10, 10))
         plt.imshow(np.abs(mat_H_a), extent=[-l_x_abs_max, l_x_abs_max,_
      →-l_y_abs_max, l_y_abs_max], cmap='hot')
         plt.colorbar()
         plt.title('Instantaneous channel in WD')
     def visualize_wavenumber_variance(variance):
         # plt.fiqure(fiqsize=(8, 6))
         plt.imshow(variance, cmap='jet', extent=[-l_x_abs_max, l_x_abs_max,_
      →-l_y_abs_max, l_y_abs_max])
         plt.colorbar()
         plt.title('Variance of the channel')
     def DFT_power_spectrum(H_channel):
         # DFT power spectrum of the channel
         DFT_power = np.abs(np.fft.fft2(np.reshape(H_channel, (N_x, N_y))))
         # element shift according to the DFT power spectrum
         DFT_power = np.fft.fftshift(DFT_power)
         # normalize the DFT power spectrum
         DFT_power = DFT_power / np.max(DFT_power)
         # visualize the DFT power spectrum
```

```
# plt.figure(figsize=(10, 10))
    plt.imshow(DFT_power, cmap='jet', extent=[-np.pi/2, np.pi/2, -np.pi/2, np.
    pi/2])
    plt.colorbar()
    plt.title('normmalized DFT power spectrum')

plt.figure(figsize=(14, 5))
# sub out of 3 in one row
plt.subplot(131)
visualize_wavenumber_domain_channel(vec_H_a)

plt.subplot(132)
visualize_wavenumber_variance(variance)

plt.subplot(133)
DFT_power_spectrum(H_channel)
```



### Observation Model

$$y = C h_a + n$$

```
[]: # Measurement matrxi for compressed sensing, {N}_{RF} \times N
measurement_matrix = np.random.randn(N_RF, N) * 1 / np.sqrt(N)

y = np.dot(
    measurement_matrix,
    np.dot(
        Psi, vec_H_a
    )
)
Phi = np.dot(measurement_matrix, Psi)
```

### Self-made OMP alg

```
[]: from tool_box.OMP import OMP
# add noise
noise = np.random.randn(*y.shape) + 1j * np.random.randn(*y.shape)
noise = noise / np.linalg.norm(noise) * np.linalg.norm(y) / 10 ** (SNR / 20)
y = y + noise

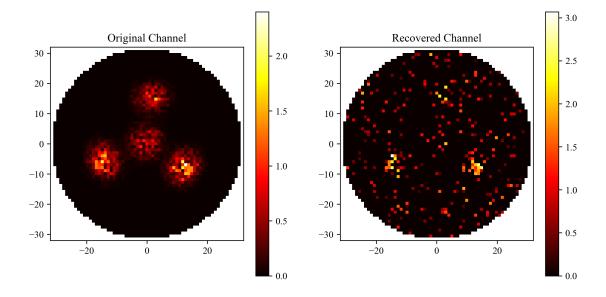
myOMP = OMP(
    A = np.array(Phi),
    y = np.array(y),
    vec_H_a = vec_H_a,
    sparseness = 0.1,
    norm_err_threshold = 1e-1)
vec_H_a_recovered, vec_H_a_nonzero_idx, vec_H_a_err = myOMP.run()
```

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```
[]: plt.figure(figsize=(10, 5))
# sub 1
plt.subplot(1, 2, 1)
visualize_wavenumber_domain_channel(vec_H_a)
# title
plt.title("Original Channel")

# sub 2
plt.subplot(1, 2, 2)
visualize_wavenumber_domain_channel(vec_H_a_recovered)
# title
plt.title("Recovered Channel")
```

### []: Text(0.5, 1.0, 'Recovered Channel')

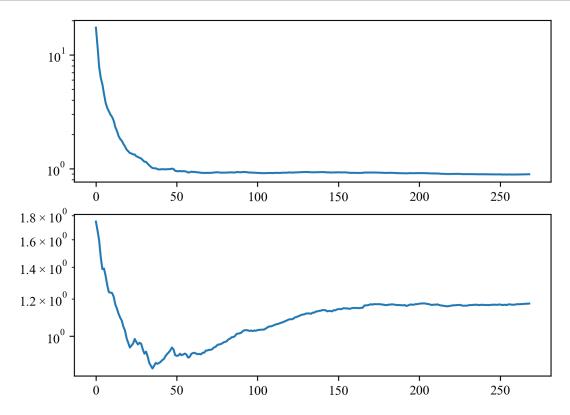


## []:

```
[]: NMSE_list_v2 = myOMP.NMSE_list_v2
NMSE_list = myOMP.NMSE_list
plt.figure()
plt.subplot(211)
plt.plot(NMSE_list)
plt.yscale('log')

plt.subplot(212)
plt.plot(NMSE_list_v2)
plt.yscale('log')

# set log scale
```



## []: