

```
In [1]: import os
import glob

import numpy as np

import matplotlib.pyplot as plt
import IPython.display as ipd
import IPython
import matplotlib.colors as colors

from scipy import signal
from copy import deepcopy

import utils
```

```
In [2]: dataurl = 'http://www.openslr.org/resources/1/waves_yesno.tar.gz'
```

```
In [3]: utils.download_and_extract_data(dataurl)
```

```
In [4]: path_to_sound_data = './waves_yesno/'
```

```
In [5]: ### STFT hyperparameters
```

```
nperseg = 512
window_size = 5
```

```
In [6]: noise_std_array = [1e-1]; ## all noise levels to consider. feel free to pl
```

```
In [7]: ### for plotting
```

```
ini = 34500-1
end = 37500+1

ini2 = 34875
end2 = 35025
```

## Creating Dataset

```
In [8]: train_dataset, train_fs_array, train_max_array, val_dataset, val_fs_array,
```

train\_dataset and val\_dataset contains time domain signals normalized so that maximum amplitude is 1

```
In [9]: max( x.max() for x in train_dataset )
```

```
Out[9]: 1.0
```

The actual maximum amplitude of signal is saved in train\_max\_array and val\_max\_array. We will use this maximum value to un-normalize the signals before plotting or listening to it.

```
actual_signal[k] = train_dataset[k] * train_max_array[k]
```

The sampling rate of each signal is given in `train_fs_array` and `val_fs_array`. In our case, all of them have same sampling rate. We will use sampling rate for `stft` function

## Get Noise Function

```
In [10]: def get_noise(data, noise_std = 0.1):
          noise = np.random.randn(*data.shape);
          noise = noise * noise_std;
          return noise
```

## STFT Hard and Block Thresholding

```
In [11]: def plot_stft(stft_array, save_str=''):

          """ utility function to plot stft """
          f, t, Zxx = stft_array;
          Zxx_abs = np.abs(Zxx)
          Zxx_abs[Zxx_abs<1e-5] = 1e-5
          plt.figure()
          plt.pcolormesh(t, f,Zxx_abs, norm=colors.LogNorm(vmin=Zxx_abs.min(), vm
          tick_size = 18
          label_size = 18
          cb = plt.colorbar()
          plt.tick_params(labelsize=tick_size)
          cb.ax.tick_params(labelsize=tick_size)
          plt.ylabel('Frequency (Hz)',fontsize=label_size)
          plt.xlabel('Time (s)',fontsize=label_size)
          plt.savefig('plots/stft_denoising_'+save_str+'_stft.pdf',bbox_inches="t
          plt.show()
```

```

In [12]: def get_block_L2_norm(mat, window_size):

    """ mat: an nxn matrix
        window_size: postivie integer (assume odd)
        return: nxn matrix where the (i,j)th entry is the L2 norm of a wind
            neighbourhood centered at (i, j)

        to obtain an nxn output, assume that edges are zero padded.

        hint: implement using a convolution

        sample output for get_block_L2_norm(np.ones([5, 5]), 3):

        [[2.          , 2.44948974, 2.44948974, 2.44948974, 2.          ],
         [2.44948974, 3.          , 3.          , 3.          , 2.44948974],
         [2.44948974, 3.          , 3.          , 3.          , 2.44948974],
         [2.44948974, 3.          , 3.          , 3.          , 2.44948974],
         [2.          , 2.44948974, 2.44948974, 2.44948974, 2.          ]]
    """
    pad = (window_size-1)//2
    mat = np.sqrt(signal.convolve2d(mat*mat,np.ones((window_size,window_siz

    return mat

```

## Implement Hard and Block Thresholding

```

In [13]: def stft_denoising(source, noise_std, fs, nperseg, thresh,
                        window_size, block_thresh = None, plot_res = True, ind=0)

    """
    implements hard and block thresholding.

    thresh - threshold for hard thresholding. Thresholding is implemented p
    block_thresh - threshold for block thresholding. Thresholding is implem

    """

    noisy = source + get_noise(source, noise_std);

    if block_thresh is None:
        block_thresh = thresh;

    source_stft = signal.stft(source, fs = fs, nperseg=nperseg);
    noisy_stft = signal.stft(noisy, fs = fs, nperseg=nperseg);
    ## FILL YOUR CODE
    coeffs = noisy_stft[2]
    hardmask = np.greater(np.abs(coeffs),thresh)
    block12s= get_block_L2_norm(coeffs, window_size)
    blockmask= np.greater(block12s,block_thresh)

    denoised_stft = ( noisy_stft[0], noisy_stft[1], hardmask*coeffs)#HARD T
    block_denoised_stft = ( noisy_stft[0], noisy_stft[1], blockmask*coeffs

    if plot_res:
        plot_stft(source_stft,save_str='_clean_'+str(ind));
        plot_stft(noisy_stft,save_str='_noisy_'+str(ind));
        plot_stft(denoised_stft,save_str='_denoised_'+str(ind));
        plot_stft(block_denoised_stft,save_str='_block_denoised_'+str(ind))

    _, source_istft = signal.istft(source_stft[2], fs=fs, nperseg=nperseg)
    _, noisy_istft = signal.istft(noisy_stft[2], fs=fs, nperseg=nperseg);
    _, denoised_istft = signal.istft(denoised_stft[2], fs=fs, nperseg=npe
    _, block_denoised_istft = signal.istft(block_denoised_stft[2], fs=fs,

    return np.real(source_istft), np.real(noisy_istft), np.real(denoised_is

```

## Choosing Threshold Based on Train Error

We will compute the error for different values of threshold and pick the threshold based on the lowest error on the training set.

We don't really do any training here. The denoising method only has one hyperparameter - the threshold. We're picking it based on the training set.

The next cells should run off hand if you have filled `get_block_L2_norm()` and `stft_denoising()`

```
In [14]: np.sqrt(1e-2)
```

```
Out[14]: 0.1
```

```
In [15]: # threshold_array = [1e-10, 1e-8, 1e-6, 1e-4, 1e-2, 1e0, 1e2]  
threshold_array = np.logspace(-10, 1, num = 10)
```

```

In [16]: error_dict = {}
lock_error_dict = {}
est_threshold_dict = {}
lock_best_threshold_dict = {}
or noise_std in noise_std_array:

    error_dict[noise_std] = np.zeros_like(threshold_array);
    block_error_dict[noise_std] = np.zeros_like(threshold_array);

    for thresh_i, thresh in enumerate(threshold_array):
        total_error = 0.0;
        block_total_error = 0.0;
        total_length = 0.0;
        for i, x in enumerate(train_dataset):
            rec_source, rec_noisy, rec_denoised, rec_denoised_block = stft_de
            total_error += np.linalg.norm( rec_denoised - rec_source)**2;
            block_total_error += np.linalg.norm( rec_denoised_block - rec_sou
            total_length += len(rec_source);

        error_dict[noise_std][thresh_i] = (total_error/total_length);
        block_error_dict[noise_std][thresh_i] = (block_total_error/total_leng

    print('noise std: ', noise_std);
    fig, axes = plt.subplots(1, 2, sharex=True, sharey=True, figsize = (8, 4)

    axes[0].loglog(threshold_array, error_dict[noise_std])
    axes[0].set_xlabel('threshold')
    axes[0].set_ylabel('error')

    best_threshold_dict[noise_std] = threshold_array[np.argmin(error_dict[no

    axes[0].set_title('single coeff best thresh: '+str( round(best_threshold

    axes[1].loglog(threshold_array, block_error_dict[noise_std])
    axes[1].set_xlabel('threshold')
    axes[1].set_ylabel('error')

    block_best_threshold_dict[noise_std] = threshold_array[np.argmin(block_e

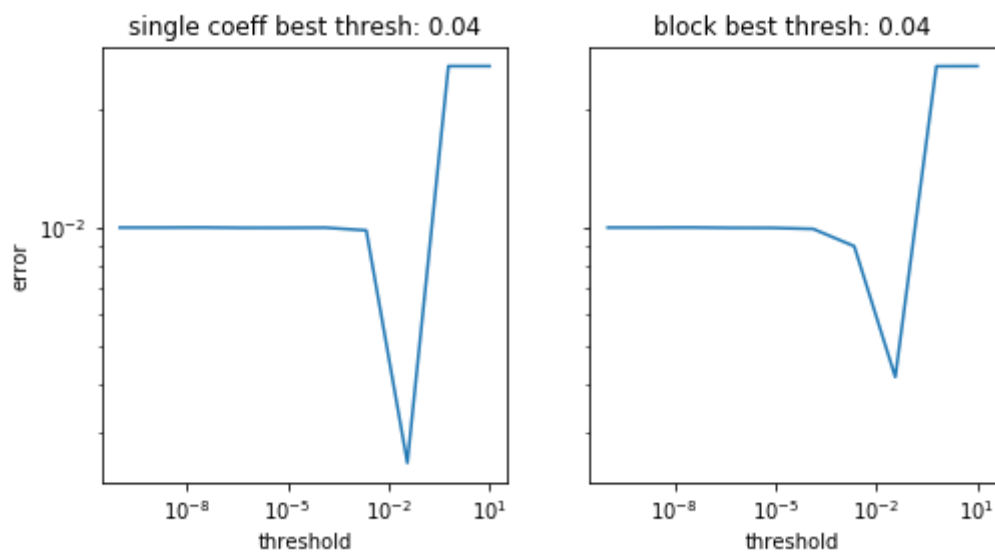
    axes[1].set_title('block best thresh: '+str( round(block_best_threshold_c

    plt.show()

    print('='*50+'\n')

```

noise std: 0.1



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## Comparison Between STFT Hard and Block Threshold

```
In [17]: val_idx = 2

source = val_dataset[val_idx]
fs = val_fs_array[val_idx]
max_val = val_max_array[val_idx]
```

```
In [18]: fig_size=(20,6)
```

```

In [20]: for ind_fig, noise_std in enumerate(noise_std_array):

    print('Noise Std: ', noise_std)

    noise_sample = get_noise(source, noise_std = noise_std);

    istft_source, istft_noisy, istft_denoised, istft_denoised_block = stft_c
                                                thresh = best
                                                window_size =
                                                block_thresh

    istft_denoised_block *= max_val
    istft_denoised *= max_val;
    istft_noisy *= max_val;
    istft_source *= max_val;

    print('STFT Denoised: ')
    IPython.display.display(ipd.Audio(istft_denoised, rate=fs))

    print('Block STFT Denoised: ')
    IPython.display.display(ipd.Audio(istft_denoised_block, rate=fs))

    label_size = 18
    font_size = 18

    t_indices = np.arange(len(istft_source))/fs

    plt.figure(figsize = fig_size)
    plt.plot(np.real(istft_source))
    plt.xlabel('Time (s)',fontsize=font_size)
    plt.tick_params(labelsize=label_size)

    plt.figure(figsize = fig_size)
    plt.plot(t_indices[ini:end],np.real(istft_denoised[ini:end]))
    plt.xlabel('Time (s)',fontsize=font_size)
    plt.tick_params(labelsize=label_size)
    plt.savefig('plots/stft_stft_denoised_' + str(ind_fig) + '.pdf',bbox_inc

    plt.figure(figsize = fig_size)
    plt.plot(t_indices[ini:end],np.real(istft_denoised_block[ini:end]))
    plt.xlabel('Time (s)',fontsize=font_size)
    plt.tick_params(labelsize=label_size)
    plt.savefig('plots/stft_block_denoised_' + str(ind_fig) + '.pdf',bbox_in

    plt.figure(figsize = fig_size)
    plt.plot(t_indices[ini2:end2],np.real(istft_source[ini2:end2]), '--o',mar
    plt.plot(t_indices[ini2:end2],np.real(istft_denoised[ini2:end2]), 'x',col
    plt.plot(t_indices[ini2:end2],np.real(istft_noisy[ini2:end2]), '.',color=
    plt.xlabel('Time (s)',fontsize=font_size)
    plt.tick_params(labelsize=label_size)
    plt.legend(fontsize=font_size)
    plt.savefig('plots/stft_stft_denoised_' + str(ind_fig) + '_zoom.pdf',bbc

    plt.figure(figsize = fig_size)

```



```

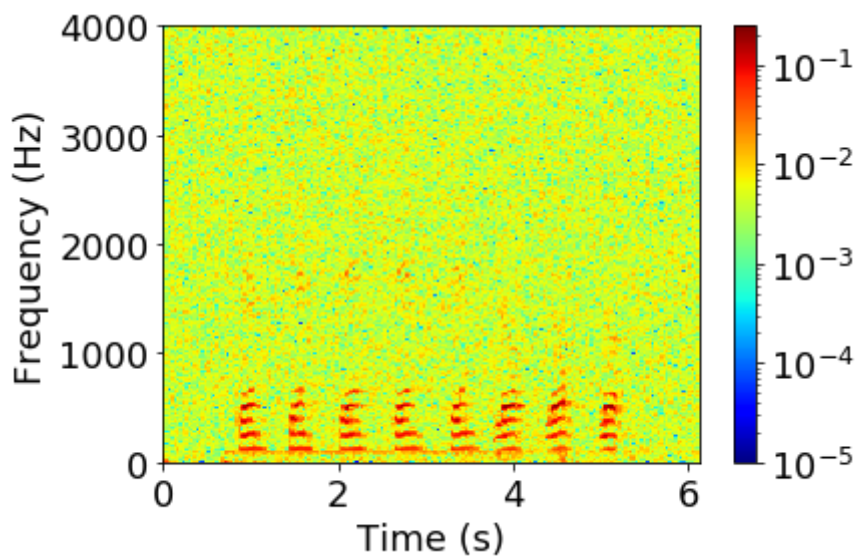
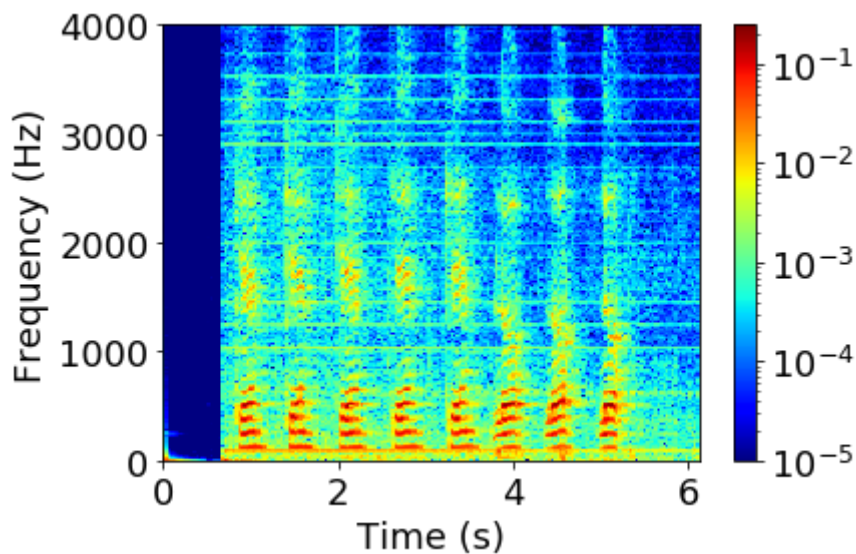
plt.plot(t_indices[ini2:end2],np.real(istft_source[ini2:end2]),'--o',mar
plt.plot(t_indices[ini2:end2],np.real(istft_denoised_block[ini2:end2]),'
plt.plot(t_indices[ini2:end2],np.real(istft_noisy[ini2:end2]),'.',color=
plt.xlabel('Time (s)',fontsize=font_size)
plt.tick_params(labelsize=label_size)
plt.legend(fontsize=font_size)
plt.savefig('plots/stft_block_denoised_' + str(ind_fig) + '_zoom.pdf',bk

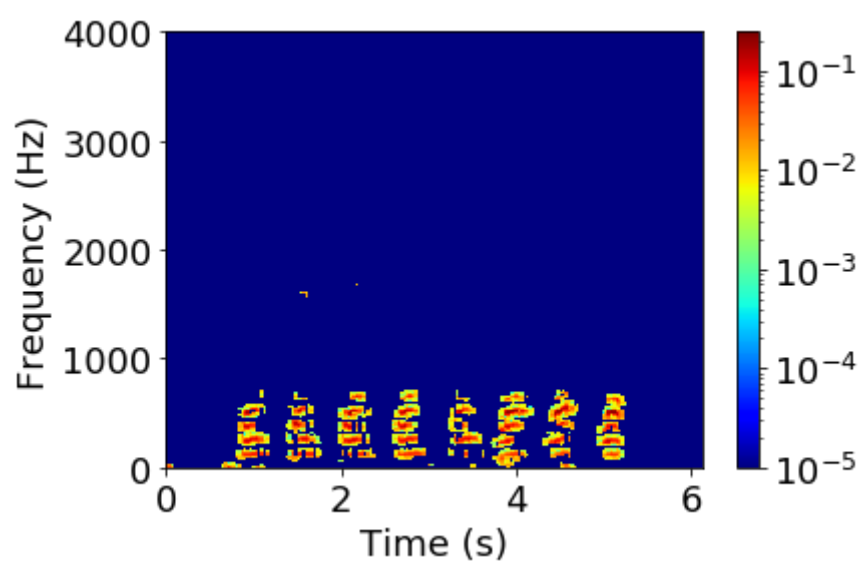
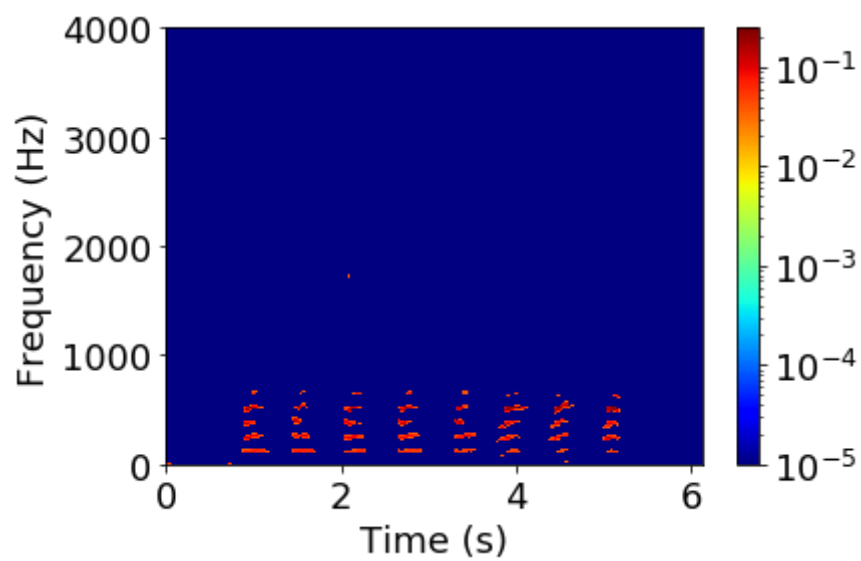
plt.show()

print('='*50 + '\n')

```

Noise Std: 0.1



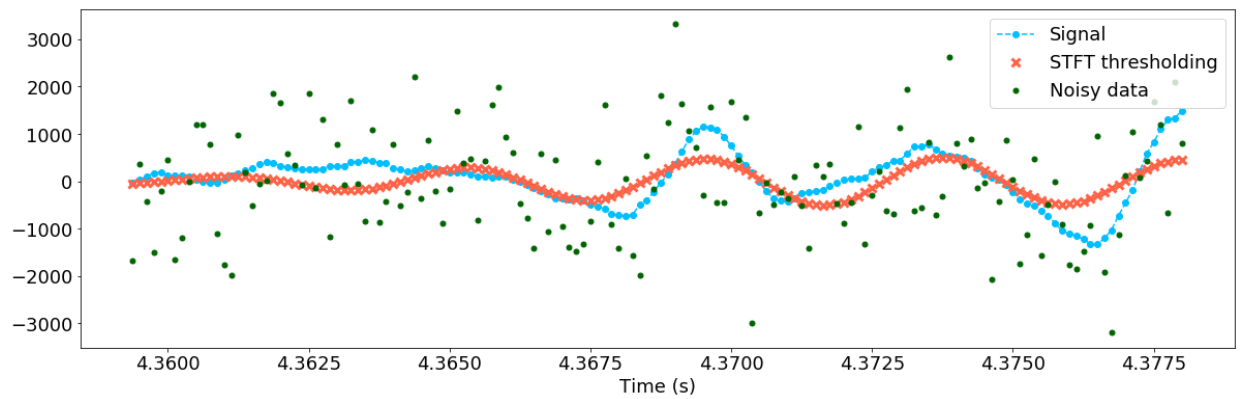
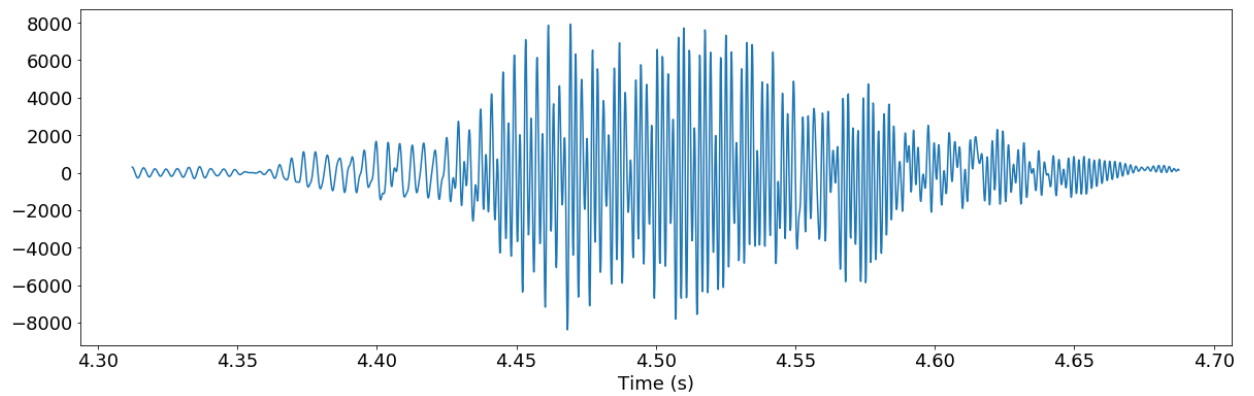
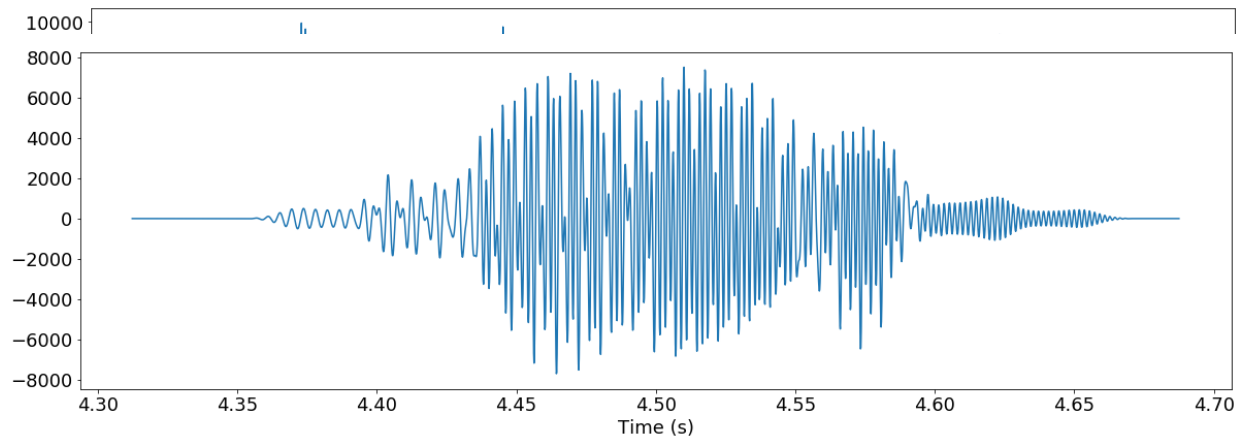


STFT Denoised:

0:06 / 0:06

Block STFT Denoised:

0:02



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In [ ]: