

Adaptive_Filters

February 28, 2020

```
In [1153]: import numpy as np
           import matplotlib.pyplot as plt

In [1154]: plt.style.use('seaborn')
           plt.rcParams['figure.figsize'] = 12,6
```

0.1 Define Constants

```
In [1404]: L = 21 # length of filter
           SeqLen = 1000 # length of test data
           SNR = 10

           lms_mu = 0.05

           nlms_mu = 1
           delta = 1e-3

           beta = .1
           reg = 1e-1
```

0.2 Generate Input Data

Run random data through a low-pass filter. LMS with same number of taps should be pretty good at guessing LPF response.

```
In [1405]: h = np.array([
           -0.02010411882885732,
           -0.05842798004352509,
           -0.061178403647821976,
           -0.010939393385338943,
           0.05125096443534972,
           0.033220867678947885,
           -0.05655276971833928,
           -0.08565500737264514,
           0.0633795996605449,
           0.310854403656636,
           0.4344309124179415,
           0.310854403656636,
```

```

0.0633795996605449,
-0.08565500737264514,
-0.05655276971833928,
0.033220867678947885,
0.05125096443534972,
-0.010939393385338943,
-0.061178403647821976,
-0.05842798004352509,
-0.02010411882885732,
])

```

```

In [1406]: x = np.random.normal(0, 1, SeqLen+L-1)
           y = np.zeros(SeqLen)
           for samp in range(SeqLen):
               y[samp] = x[samp:samp+L]@h + np.random.normal(0,1/SNR)

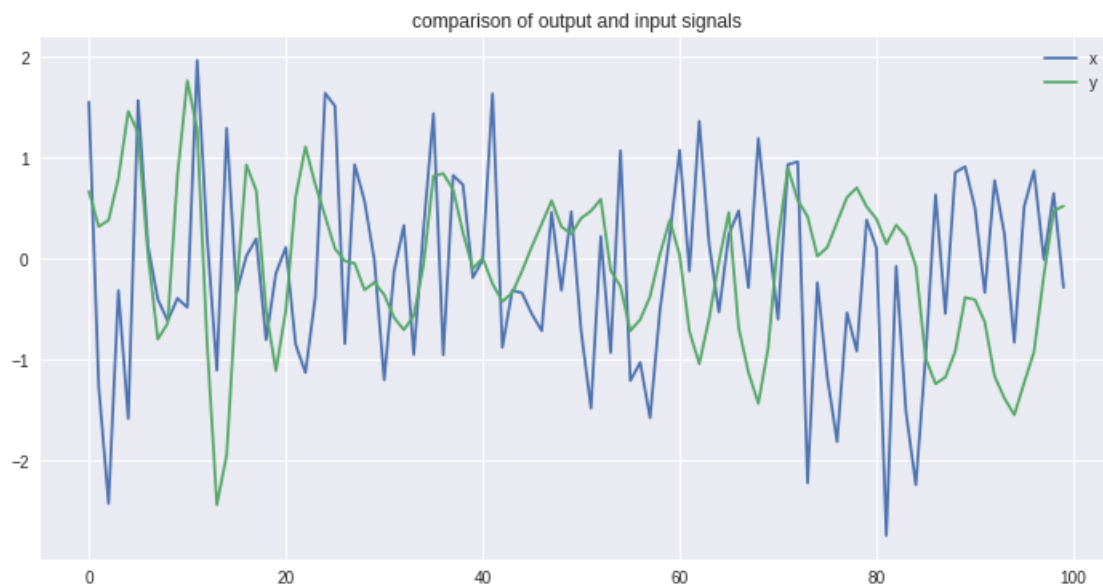
```

```

In [1418]: plt.plot(x[L:100+L], label="x")
           plt.plot(y[:100], label="y")
           plt.title("comparison of output and input signals")
           plt.legend()

```

Out[1418]: <matplotlib.legend.Legend at 0x7fea4f80fd90>



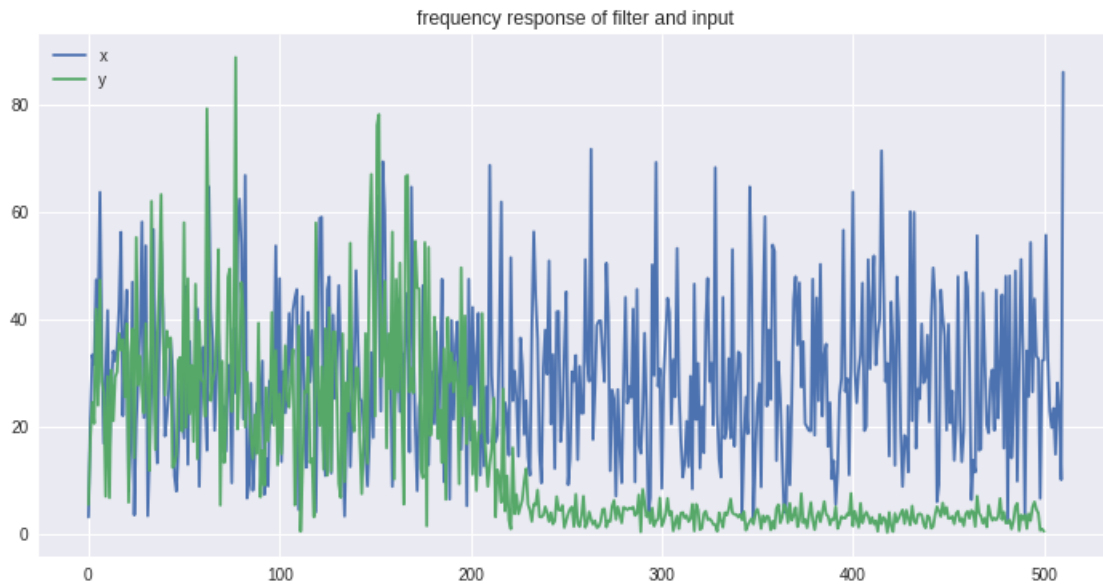
Confirm that low pass filter is working

```

In [1417]: plt.plot(abs(np.fft.rfft(x)), label="x")
           plt.plot(abs(np.fft.rfft(y)), label="y")
           plt.title("frequency response of filter and input")
           plt.legend()

```

Out[1417]: <matplotlib.legend.Legend at 0x7fea4f87d850>



1 Create and use filters

```
In [1409]: class LMS:
            """ Least mean squares filter
            Arguments:
                L : length of filter in samples
                mu: learning rate of filter
            """
            def __init__(self, L, mu):
                self.w = np.zeros(L)
                self.mu = mu

            def process(self, X, y):
                yhat = self.w.T*X
                e = y-yhat
                self.w += self.mu * e * X
                mse = e**2
                return mse, yhat

In [1410]: class NLMS:
            """ NLMS filter class
            Arguments:
                L      : length of filter in samples
                mu     : learning rate of filter
                delta  : factor to add for numerical stability
```

```

"""
def __init__(self, L, mu, delta):
    self.w = np.zeros(L)
    self.mu = mu
    self.delta = delta

def process(self, X, y):
    yhat = np.dot(self.w, X)
    e = y-yhat
    nu = self.mu / (self.delta + np.dot(X, X))
    self.w += nu * e * X
    mse = e**2
    return mse, yhat

```

In [1411]: `class RLS:`

```

""" Recursive least squares filter
Arguments:
    L      : length of filter in samples
    beta   : forgetting factor
    reg    : regularization factor
"""
def __init__(self, L, beta, reg):
    self.w = np.zeros(L)
    self.beta = beta
    self.reg = reg
    self.P = 1/reg*np.eye(L,L)

def process(self, X, y):
    yhat = np.dot(self.w.T, X)
    e = y-yhat

    num = np.dot(np.dot(np.dot(self.P,X),X.T),self.P)
    denom = self.beta + np.dot(np.dot(X,self.P),X.T)

    self.P = 1/self.beta * (self.P - num/denom)
    self.w += e*(self.P@X.T)
    mse = e**2
    return mse, yhat

```

In [1412]: `lms = LMS(L,lms_mu)`
`nlms = NLMS(L, nlms_mu, delta)`
`rls = RLS(L, beta, reg)`

In [1413]: `yhat_lms = np.zeros(y.shape)`
`yhat_nlms = np.zeros(y.shape)`
`yhat_rls = np.zeros(y.shape)`
`es_lms = np.zeros(y.shape)`

```

es_nlms = np.zeros(y.shape)
es_rls = np.zeros(y.shape)
for i in range(SeqLen):
    es_lms[i], yhat_lms[i] = lms.process(x[i:i+L], y[i])
    es_nlms[i], yhat_nlms[i] = nlms.process(x[i:i+L], y[i])
    es_rls[i], yhat_rls[i] = rls.process(x[i:i+L], y[i])

```

```

In [1414]: npoints = 300
plt.plot(yhat_lms[L:npoints], label=r"$\hat{y}$ LMS")
plt.plot(yhat_nlms[L:npoints], label=r"$\hat{y}$ NLMS")
plt.plot(yhat_rls[L:npoints], label=r"$\hat{y}$ RLS")
plt.plot(y[L:npoints], label=r"$y$")
plt.title("desired signal and estimate")
plt.legend()

```

Out[1414]: <matplotlib.legend.Legend at 0x7fea4f9e79d0>

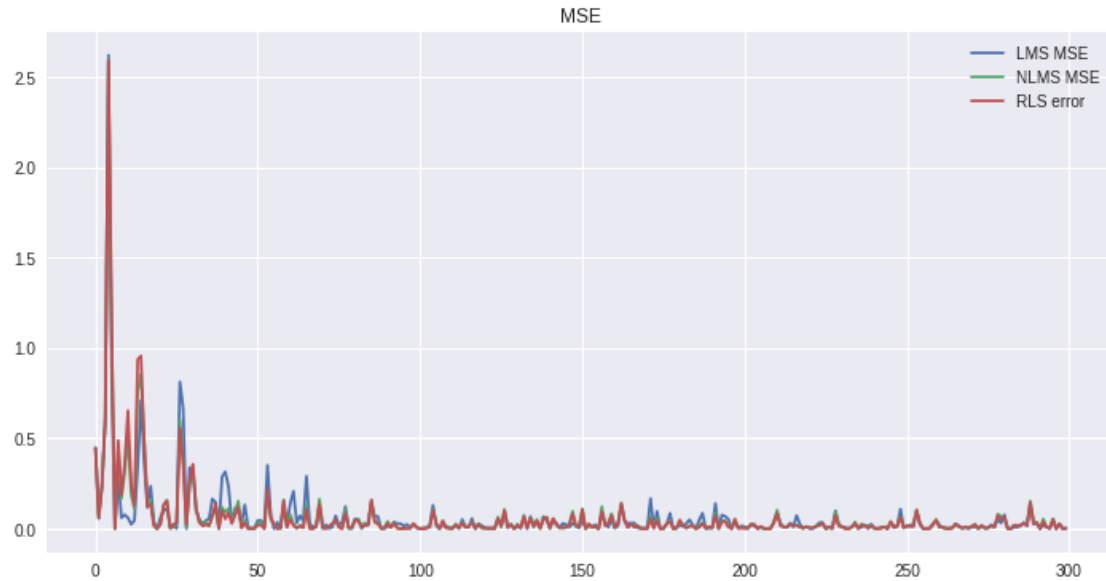


```

In [1415]: nerrors = 300
plt.plot(es_lms[:nerrors], label="LMS MSE")
plt.plot(es_nlms[:nerrors], label="NLMS MSE")
plt.plot(es_rls[:nerrors], label="RLS error")
plt.legend()
plt.title("MSE")

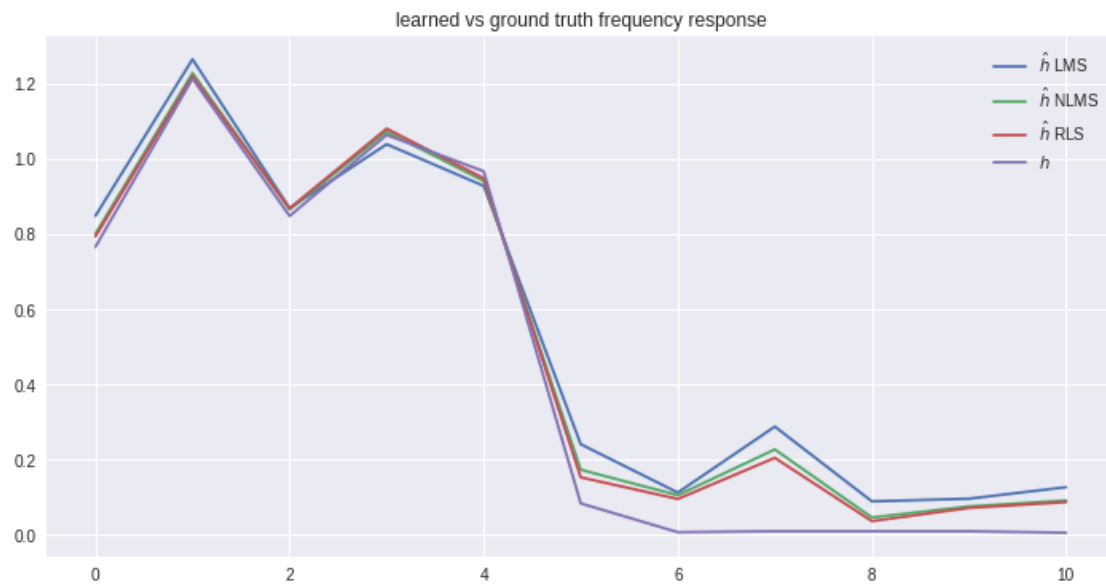
```

Out[1415]: Text(0.5, 1.0, 'MSE')



```
In [1416]: plt.plot(np.abs(np.fft.rfft(lms.w)), label=r"$\hat{h}$ LMS")
plt.plot(np.abs(np.fft.rfft(nlms.w)), label=r"$\hat{h}$ NLMS")
plt.plot(np.abs(np.fft.rfft(rls.w)), label=r"$\hat{h}$ RLS")
plt.plot(np.abs(np.fft.rfft(h)), label=r"$h$")
plt.title("learned vs ground truth frequency response")
plt.legend()
```

Out[1416]: <matplotlib.legend.Legend at 0x7fea4f8965d0>



2 Inspect Loss from C++ Implementation

This section inspects the .csv files exported by the c++ testbench

```
In [1367]: import pandas as pd
```

```
In [1398]: lmsf = "./hls/lms_classes/solution1/csim/build/LMS_MSEs.csv"
nlmsf = "./hls/nlms_classes/solution1/csim/build/NLMS_MSEs.csv"
lmsdf = pd.read_csv(lmsf)
nlmsdf = pd.read_csv(nlmsf)
```

```
In [1399]: df = nlmsdf
```

```
In [1400]: nt = 200
plt.plot(np.array(df.x[:nt]), label="x")
plt.plot(np.array(df.y[:nt]), label="y")
plt.legend()
```

```
Out[1400]: <matplotlib.legend.Legend at 0x7fea4fca3ad0>
```



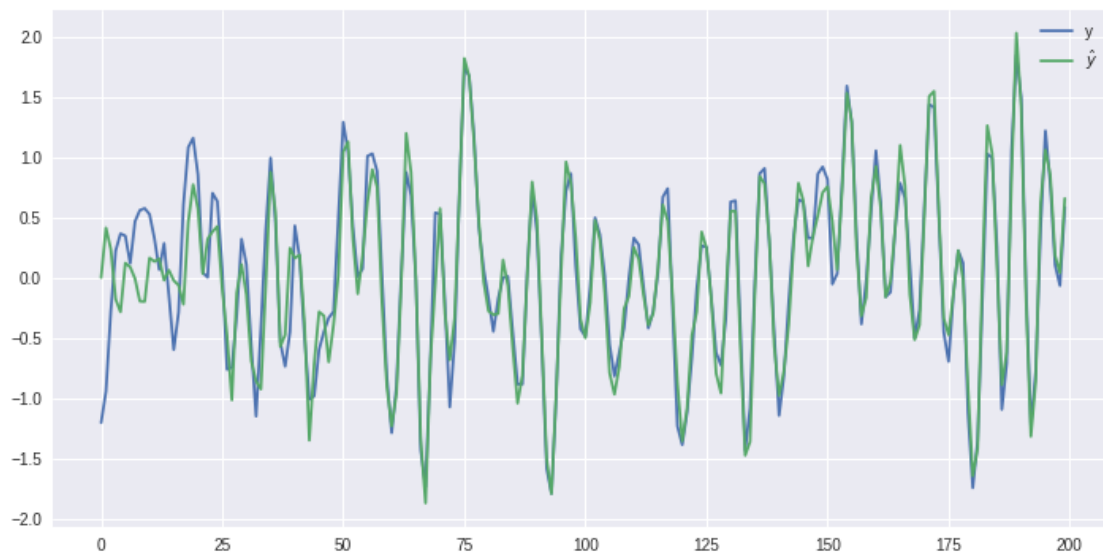
```
In [1401]: plt.plot(abs(np.fft.rfft(df.x)))
plt.plot(abs(np.fft.rfft(df.y)))
```

```
Out[1401]: [<matplotlib.lines.Line2D at 0x7fea4fc5a910>]
```



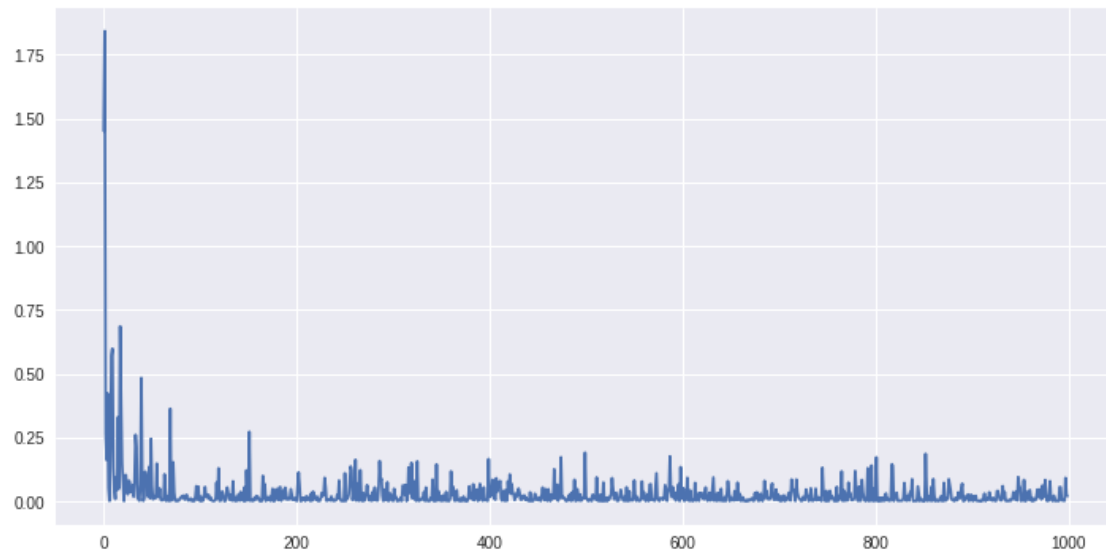
```
In [1402]: plt.plot(df.y[:nt], label='y')
           plt.plot(df.yhat[:nt], label=r'$\hat{y}$')
           plt.legend()
```

```
Out[1402]: <matplotlib.legend.Legend at 0x7fea4fbe9510>
```



```
In [1403]: plt.plot(df.MSE)
```

```
Out[1403]: [<matplotlib.lines.Line2D at 0x7fea4fb1cf90>]
```

In []: