hw2

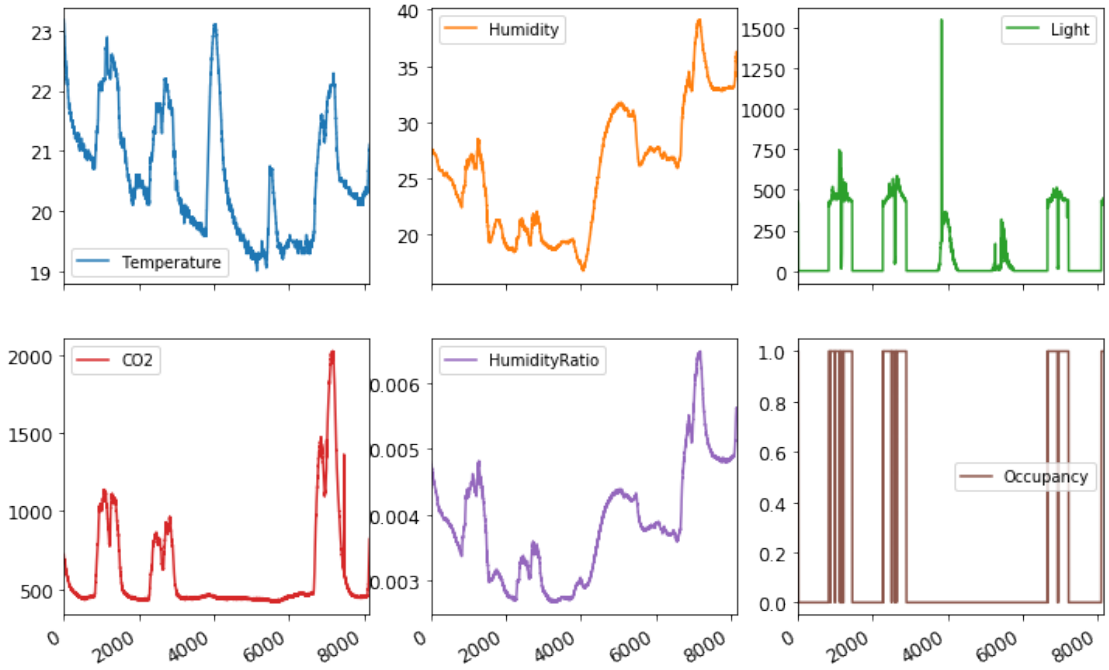
September 17, 2017

In [1]:
```
#%%writefile read_data.py
#%load read_data.py
import pandas as pd
import os
def load_data(filename):
    csv_path = os.path.join(".", filename)
    return pd.read_csv(csv_path)

room_data = load_data('train_data.txt')
room_data.head()
```

Out[1]:

	date	Temperature	Humidity	Light	CO2	HumidityRatio	Occupancy
1	2015-02-04 17:51:00	23.18	27.2720	426.0	721.25	0.004793	1
2	2015-02-04 17:51:59	23.15	27.2675	429.5	714.00	0.004783	1
3	2015-02-04 17:53:00	23.15	27.2450	426.0	713.50	0.004779	1
4	2015-02-04 17:54:00	23.15	27.2000	426.0	708.25	0.004772	1
5	2015-02-04 17:55:00	23.10	27.2000	426.0	704.50	0.004757	1

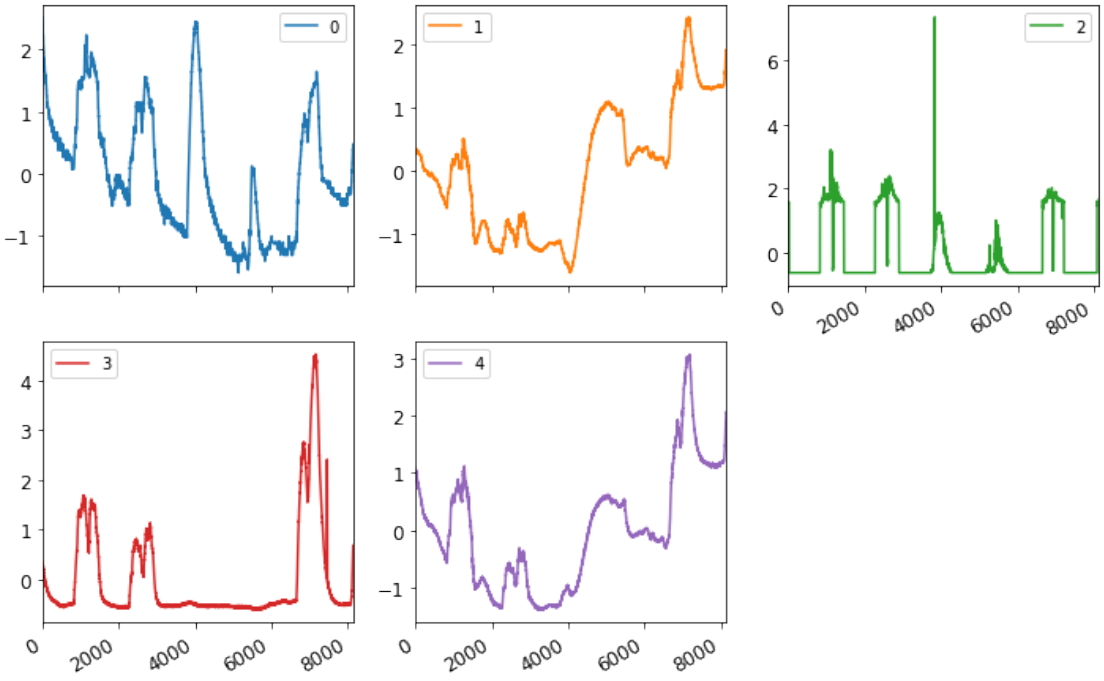
In [2]:
```
#%%writefile plot_data.py
#%load plot_data.py
%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt
plt.rcParams['axes.labelsize'] = 14
plt.rcParams['xtick.labelsize'] = 12
plt.rcParams['ytick.labelsize'] = 12
room_data.plot(subplots=True, layout=(2,3), figsize=(12, 8));
```



In [3]:
```
#%%writefile preprocess_data.py
# %load preprocess_data.py

from sklearn import preprocessing
from sklearn.base import BaseEstimator, TransformerMixin
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler
import numpy as np
import re
room_data.describe()

#attributes = [attr for attr in list(data) if not re.search(attr, r'date|Occupancy')]
# Create a class to select numerical or categorical columns
# since Scikit-Learn doesn't handle DataFrames yet
class DataFrameSelector(BaseEstimator, TransformerMixin):
    def __init__(self, attribute_names):
        self.attribute_names = attribute_names
    def fit(self, X, y=None):
        return self
    def transform(self, X):
        return X[self.attribute_names].values

# list of attributes for the DataFrameSelector (pandas to numpy)
room_attrib = [attr for attr in list(room_data) if not re.search(attr, r'date|Occupancy
pipeline = Pipeline([
```

```
        ('selector', DataFrameSelector(room_attrib)),
        ('std_scaler', StandardScaler()),
    ])

    room_prepared = pipeline.fit_transform(room_data)
    print(np.shape(room_prepared))
    print(room_prepared[0])
    df = pd.DataFrame(room_prepared)
    df.plot(subplots=True, layout=(2,3), figsize=(12, 8));
```

```
(8143, 5)
[ 2.51847007  0.27852622  1.57376283  0.36494808  1.09175697]
```



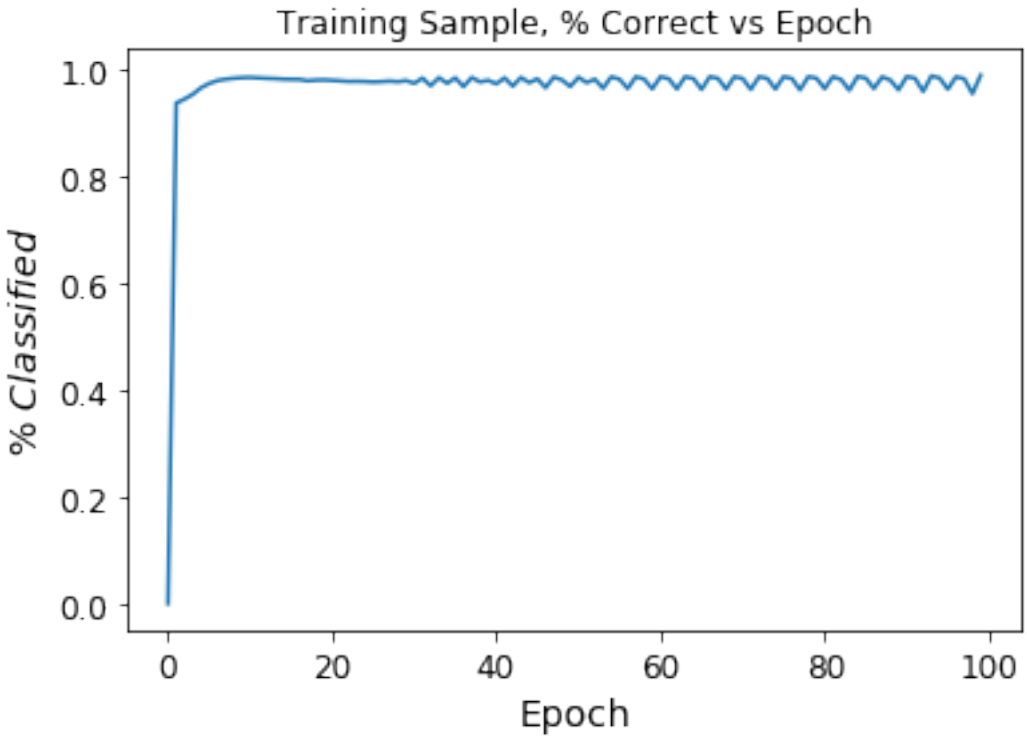
In [4]:
```
#%%writefile train.py
# %load train.py
import os as os
import numpy as np
import scipy.io
import scipy.optimize as optimization

N_EPOCH = 100

n_mis = []
def dwp(w,d,X):
```

```
        # using numpy einstein summation to vectorize the computation
        # calculating w.x.d -- w.x.d < 0 => incorrect classification
        C = d*np.einsum('i,ji->j', w, X) # shape is (100,)
        # use enumerate for the equivalent to each_with_index (Ruby)
        xd = [ d[i]*X[i] for i,c in enumerate(C) if c <= 0]
        # the error is the number of misclassified teachers
        E = len(xd)/float(len(d))
        # sum xd element-wise
        # dw is x.t summed over misclassified teachers
        dw =  np.einsum('ij->j',xd) if(len(xd) > 0) else 0
        return dw, E

    def train():
        # data
        room_with_bias = np.full((len(room_prepared),6),1.0)
        room_with_bias[:,1:] = room_prepared
        # shape is (nrow, 6)
        X = room_with_bias
        #print(np.shape(X))
        # use domain -1,1 for teachers so we can use x.w.d<0 for the classification test
        # shpae is (nrow,1)
        t = [z if z == 1 else -1 for z in room_data['Occupancy']]
        #print(np.shape(t))
        # starting vector of weights
        w = np.zeros((np.shape(room_with_bias)[1]))
        perf = []
        # sweep through the data N_EPOCH times
        for _ in range(N_EPOCH):
            dw, E = dwp(w,t,X)
            w += dw
            perf.append(1-E)
        return w, perf

    # invoke main
    w_trained, performance = train()
```

In [5]:
```
    #%%writefile plot_train_perf.py
    # To plot pretty figures
    %matplotlib inline
    import matplotlib
    import matplotlib.pyplot as plt
    plt.rcParams['axes.labelsize'] = 14
    plt.rcParams['xtick.labelsize'] = 12
    plt.rcParams['ytick.labelsize'] = 12

    plt.ylabel(r'$\%\ Classified$')
    plt.xlabel('Epoch')
    plt.title('Training Sample, % Correct vs Epoch')
```

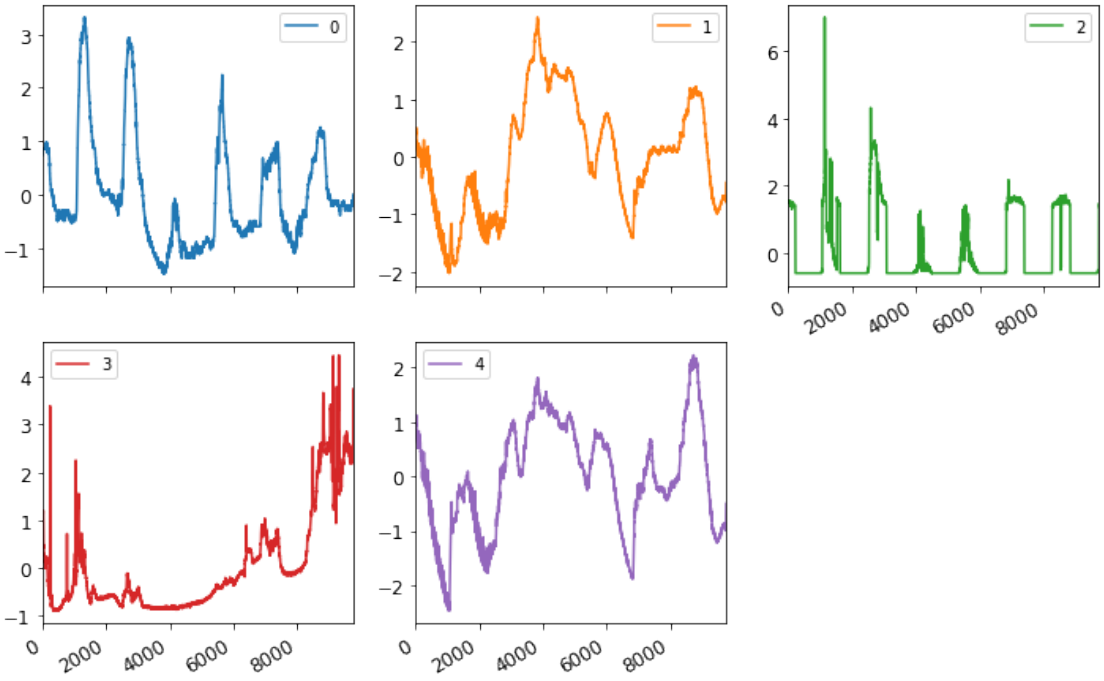
```
plt.plot(performance)
```

Out[5]: [<matplotlib.lines.Line2D at 0x115ca8390>]



In [6]:
```
#%%writefile test.py
def test(w):
    test_data = load_data('test_data.txt')
    # same transform pipeline on the test data as the training data
    test_prepared = pipeline.fit_transform(test_data)
    df = pd.DataFrame(test_prepared)
    df.plot(subplots=True, layout=(2,3), figsize=(12, 8));
    # add a bias to the prepared test data
    test_with_bias = np.full((len(test_prepared),6),1.0)
    test_with_bias[:,1:] = test_prepared
    # shape is (nrow, 6)
    X = test_with_bias
    t = [z if z == 1 else -1 for z in test_data['Occupancy']]
    # all we care about here is E
    _, E = dwp(w, t, X)
    print(1-E)

test(w_trained)
```

```
0.9520098441345365
```


0
1
2
3
4
0
2000
4000
6000
8000