

Hidden Markov Model (HMM) is a statistical Markov model in which the system being modeled is assumed to be a Markov process with unobserved (i.e. hidden) states. Observed data is our market features, hidden states are our market behavior.

It is simple enough. But also its rich enough.

We can use it as a baseline model.

<https://www.youtube.com/watch?v=TPRoLreU9IA>

```
import quandl
import numpy as np
import pandas as pd
import statsmodels.api as sm
from scipy import stats
from matplotlib import cm, pyplot as plt
from hmmlearn.hmm import GaussianHMM
import scipy
import datetime
import json
import seaborn as sns
import joblib
import quandl
from dateutil.relativedelta import relativedelta

import yfinance as yf
import plotly.express as px
from statistics import mode

import numpy as np
from sklearn.decomposition import SparsePCA, PCA
from sklearn.preprocessing import StandardScaler, MinMaxScaler

import warnings
warnings.filterwarnings("ignore")
```

load data

```
vix_df = pd.read_pickle('/data/workspace_files/data_vix.pickle')
rf_df = pd.read_pickle('/data/workspace_files/data_rf.pickle')
fut_return = pd.read_pickle('/data/workspace_files/data_fut_return.pickle')
```

PCA

```
def pca_algo(data_df):  
    x92 = StandardScaler().fit_transform(data_df) #standardize  
    pca_transformer = PCA()  
    pca_transformer.fit(x92)  
    px.line(pd.DataFrame(pca_transformer.explained_variance_ratio_).cumsum())  
  
    pca_model = PCA(n_components=20).fit(x92)  
    pca_data = pca_model.transform(x92)  
    pca_data = pd.DataFrame(pca_data, index=data_df.index)  
    return pca_data
```

SPARSE PCA

```
def spca_algo(data_df):  
    x92 = StandardScaler().fit_transform(data_df) #standardize  
    spca_transformer = SparsePCA(n_components=12, alpha=1., random_state=0)  
    spca_transformer.fit(x92)  
    spca_data = spca_transformer.transform(x92)  
    spca_data = pd.DataFrame(spca_data, index=data_df.index)  
    return spca_data
```

HMM Model

```

def plot_hidden_states(model, data, X, close_series, fut_ret_series):
    plt.figure(figsize=(15, 15))
    fig, axs = plt.subplots(model.n_components, 3, figsize = (15, 15))
    colours = cm.prism(np.linspace(0, 1, model.n_components))
    hidden_states = model.predict(X)

    print(data.shape, X.shape, close_series.shape, fut_ret_series.shape)

    ret_list = []
    for i, (ax, colour) in enumerate(zip(axs, colours)):
        mask = hidden_states == i
        ax[0].plot(data.index, close_series, c = 'grey')
        ax[0].plot(data.index[mask], close_series[mask], '.', c = colour)
        ax[0].set_title("{0}th hidden state".format(i))
        ax[0].grid(True)

        ax[1].hist(fut_ret_series[mask], bins = 30)
        ax[1].set_xlim([-0.1, 0.1])
        ax[1].set_title("future return distribution at {0}th hidden state".format(i))
        ax[1].grid(True)

        cum_return = (fut_ret_series[mask] + 1).cumprod()
        ax[2].plot(cum_return, c = colour)
        ax[2].set_title("cumulative future return at {0}th hidden state".format(i))
        ax[2].grid(True)
        ret_list.append(cum_return.iloc[-1].squeeze())
    plt.tight_layout()
    return ret_list

```

```

# Brute force modelling
def hmm_model(data, state, max_iter = 10000):
    best_model = GaussianHMM(n_components = state,
                             random_state = 100,
                             covariance_type = "full", n_iter = max_iter).fit(data)

    return best_model

```

```

def get_data(mtype, data_df):
    if mtype == "spca":
        spca_data = spca_algo(data_df)
        return spca_data.copy()
    elif mtype == "pca":
        pca_data = pca_algo(data_df)
        return pca_data.copy()
    elif type(mtype)==list:
        return data_df[mtype].copy()

def train_hmm_model(train_data, data_df, close_data, fut_return, filen):
    train_ind = int(np.where(train_data.index >= pd.to_datetime('2018-01-01 00:00:00'))[0][0])
    train_set = train_data.iloc[:train_ind]
    test_set = train_data.iloc[train_ind:]
    test_set_orig = data_df.loc[test_set.index]
    print("train set size is ", train_set.shape[0], ", test set size is ", test_set.shape[0])

    model = hmm_model(train_set, state=3, max_iter=1000000)
    print("Best model with {0} states ".format(str(model.n_components)))

    rets = plot_hidden_states(model,
                              data_df.iloc[:train_ind],
                              train_set,
                              close_data[:train_ind],
                              fut_return.iloc[:train_ind])

    test_hidden_states = model.predict(test_set)

    bear_id = rets.index(min(rets))
    bull_id = rets.index(max(rets))
    neutral_id = 3 - bear_id - bull_id
    print("bull", bull_id, " bear", bear_id, " neutral", neutral_id)

    states = {"bear": bear_id,
              "bull": bull_id,
              "static": neutral_id}

    signal_map = {v: k for k, v in states.items()}
    test_set_res = train_data.iloc[train_ind:]
    test_orig_data = data_df.loc[train_data.index][train_ind:]

    test_orig_data['signal'] = test_hidden_states
    test_set_res['signal'] = test_hidden_states
    actions = np.select([
        (test_orig_data[['signal']] == states['bull']),
        (test_orig_data[['signal']] == states['bear']),
        (test_orig_data[['signal']] == states['static'])],
        [1, -1, 0])
    test_set_res['weight'] = actions

    train_h = pd.DataFrame(model.predict(train_set), columns=['signal'], index=train_set.index)
    train_out = pd.DataFrame(np.select([
        (train_h[['signal']] == states['bull']),
        (train_h[['signal']] == states['bear']),
        (train_h[['signal']] == states['static'])],
        [1, -1, 0])

```

```

        ),
        index=train_set.index)
train_out.to_pickle('/data/workspace_files/output/hmm_%s_train.pickle'%filen)

out = pd.DataFrame(test_set_res['weight'], index=test_set.index)
out.to_pickle('/data/workspace_files/output/hmm_%s_test.pickle'%filen)

plot_hidden_states(model,
                    data_df.iloc[train_ind:],
                    test_set,
                    close_data.iloc[train_ind:],
                    fut_return.iloc[train_ind:])

return rets, model, test_set_res, test_set_orig

```

Baseline

```

data_df92 = pd.read_pickle('/data/workspace_files/data_df92.pickle')
close_data92 = data_df92[['Close']]
data_df92 = data_df92[['last_return']]
fut_return92 = fut_return.loc[data_df92.index]

rets, model, test_set_res, test_set_orig = train_hmm_model(data_df92,
                                                            data_df92,
                                                            close_data92,
                                                            fut_return92,
                                                            "baseline")

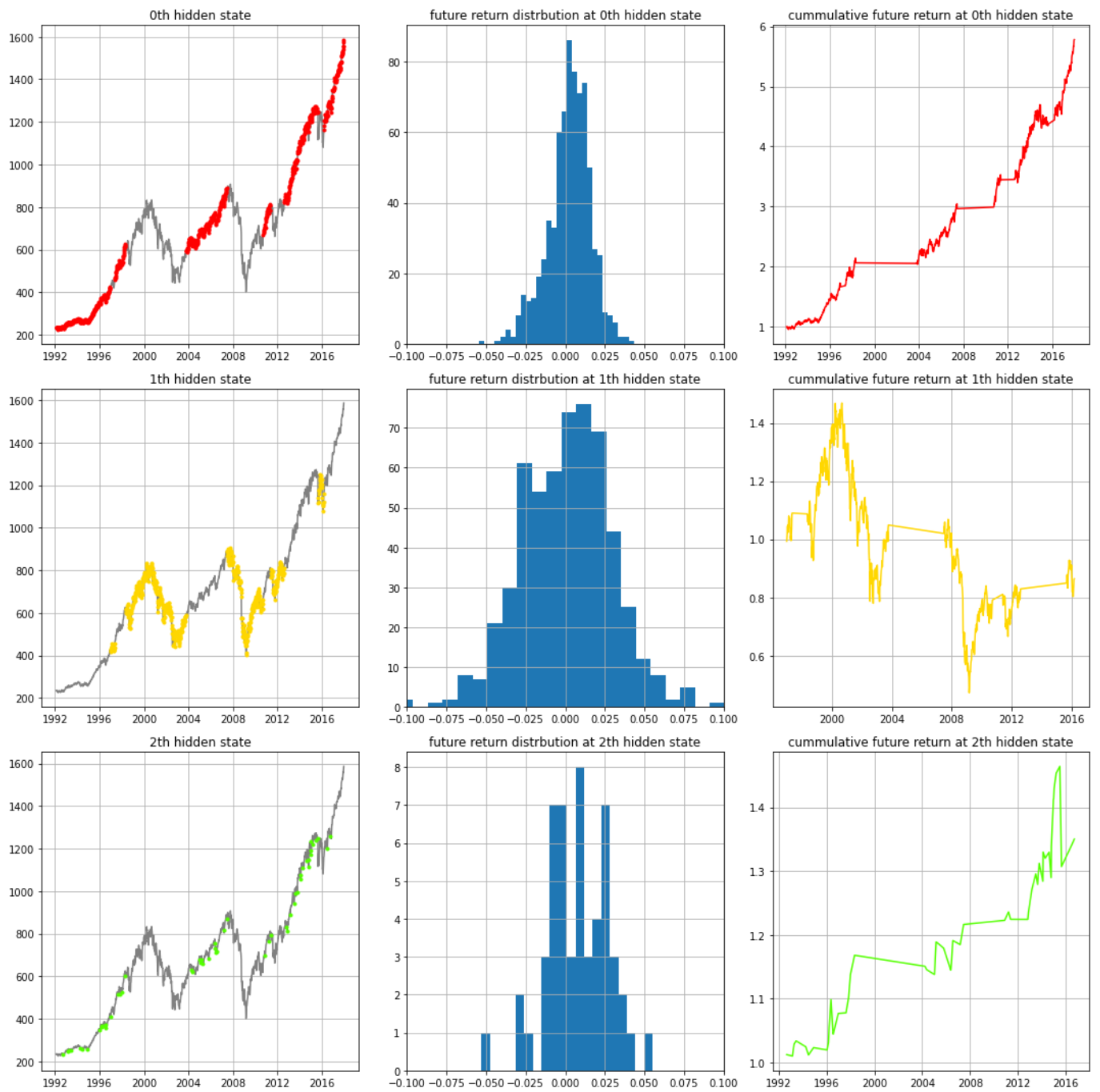
```

```

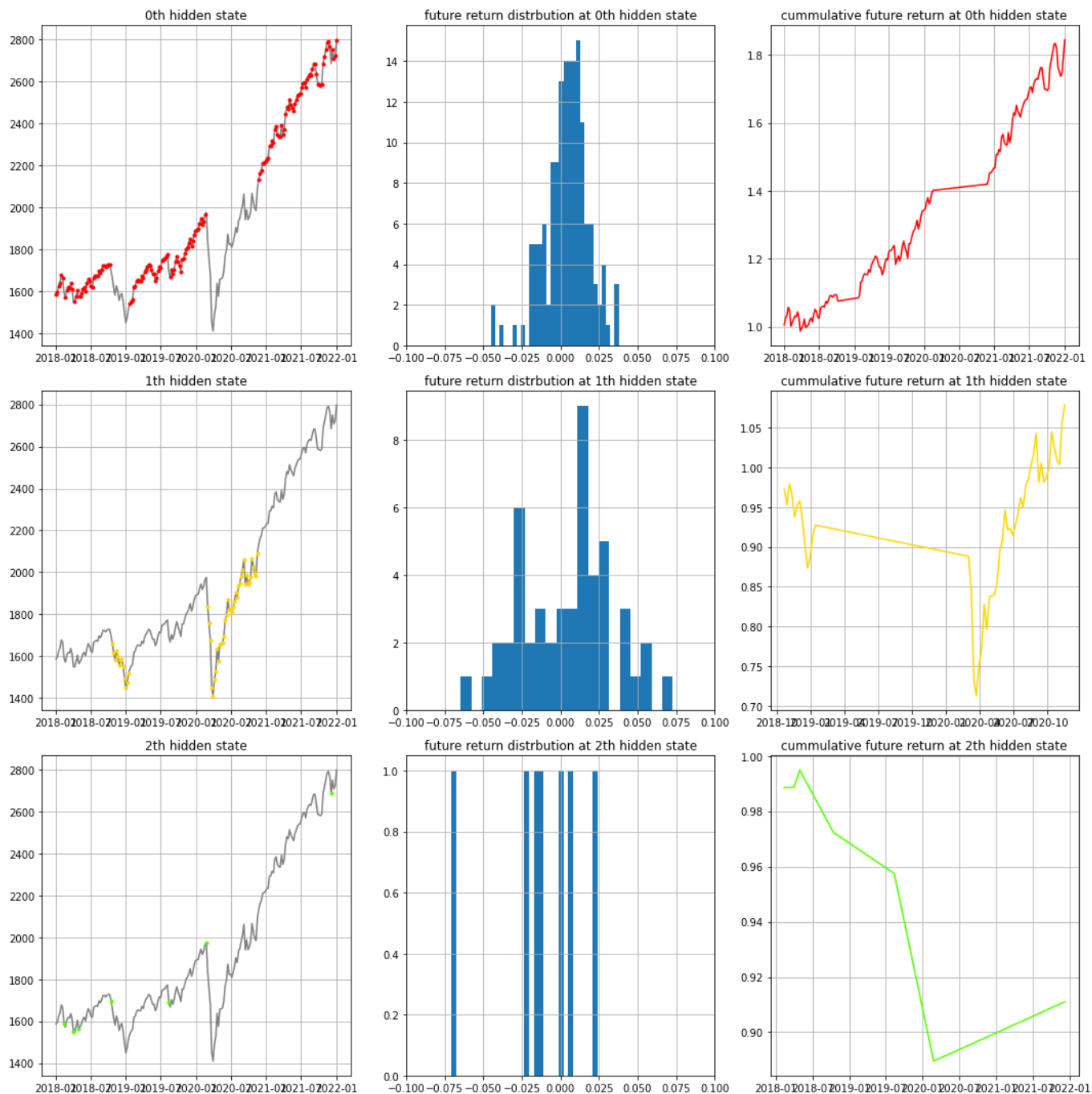
train set size is 1351 , test set size is 210
Best model with 3 states
(1351, 1) (1351, 1) (1351, 1) (1351, 1)
bull 0 bear 1 neutral 2
(210, 1) (210, 1) (210, 1) (210, 1)

```

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<Figure size 1080x1080 with 0 Axes>



Trading Strategy

[illegible]