In [17]:

```
import pandas as pd
import numpy as np
#********************
#
    Data Preprocessing
#
#************************

df = pd. read_excel('index_201101-201511_36. xlsx')
```

In [8]:

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 71842 entries, 0 to 71841
Data columns (total 8 columns):
数据时间
               71842 non-null datetime64[ns]
关键词
              71842 non-null object
数据类型
               71842 non-null object
             71842 non-null object
地区
指数
             71842 non-null float64
导入时间
               71842 non-null datetime64[ns]
               71842 non-null int64
数据周期
           71842 non-null int64
_AutoID_
dtypes: datetime64[ns](2), float64(1), int64(2), object(3)
memory usage: 4.4+ MB
```

```
#check stock data
def check_colinfo(df_index, col_name):
    "accept a dataframe of index collection with at least firm tickers and timestamp;
    print the anomalous column value information.
    #if there is null value in column
    if df index[col name].isnull().any():
        #get the array of the index of null value
        idx = np. where (df_index[col_name]. isnull())
        #get the tickers of firms with null value
        null_tic = list(df_index.ix[idx[0]]['关键词'].unique())
        if len(null tic) <=10:
            for i in null_tic:
                print(col_name, 'of', i, 'have/has NULL value, please CHECK!')
        else:
            print(len(null tic), 'indexes have NULL value, please CHECK!')
    #if there is negative value
    if (df_index[col_name]<0).any():</pre>
        neg_tic = list(df_index.ix[df_index[col_name]<0,'关键词'].unique())
        if len(neg_tic) <=10:
            for i in neg tic:
                print(col name, 'of', i, 'have/has negative value(s), please CHECK!')
        else:
            print(len(neg_tic), 'indexes have negative value(s), please CHECK!')
    #if there is repetitive record on the same day
    #for convenience we assume date as one seperate float column
    if df_index.duplicated(['关键词','数据时间']).any():
        dup_tic = list(df_index.ix[np.where(df_index.duplicated(['关键词','数据时间']))[0],'关键词']
        if len(dup\_tic) \leq 10:
            for i in dup_tic:
                print(col_name, 'of', i, 'have/has repetitive value(s), please CHECK!')
            print(len(dup tic), 'indexes have repetitive value(s), please CHECK!')
check_colinfo(df, '指数')
```

In [18]:

```
# since the way we collect data may mistake zero to some negative values.

df.ix[df['指数']<0,'指数'] = 0
# drop duplicates and choose the first value of duplicated items

df = df.drop_duplicates(subset=['数据时间', '关键词'], keep='first')

df.to_csv('index_201101-201511_36_cleaned.csv')
```

```
In [1]:
#************
#
#
    Data preprocessing
#
#>>> 1> Read file and Data preprocessing
# read excel file into dataframe
import pandas as pd
import numpy as np
df = pd. read_excel('final_index.xlsx')
#check data
print(df.info())
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 60 entries, 0 to 59
Data columns (total 28 columns):
年月
            60 non-null datetime64[ns]
发病数
            60 non-null int64
            60 non-null float64
HIV阳性
艾滋病论坛
              60 non-null float64
弓形体病
             60 non-null float64
淋病
            60 non-null float64
梅毒
            60 non-null float64
艾滋病初期症状图片
                 60 non-null float64
艾滋病检查
              60 non-null float64
                60 non-null float64
艾滋病检查费用
艾滋病能活多久
                60 non-null float64
              60 non-null float64
艾滋病症状
```

艾滋病治疗

带状疱疹

呼吸困难

食欲不振

肌肉痛 咳嗽

嗜睡

头痛 消瘦

胸痛

丘疹

None

血小板减少 阿昔洛韦

memory usage: 13.2 KB

抽搐

盗汗 恶性肿瘤 60 non-null float64

60 non-null float64

60 non-null float64 60 non-null float64

60 non-null float64

60 non-null float64 60 non-null float64

60 non-null float64

60 non-null float64

60 non-null float64 60 non-null float64

dtypes: datetime64[ns](1), float64(26), int64(1)

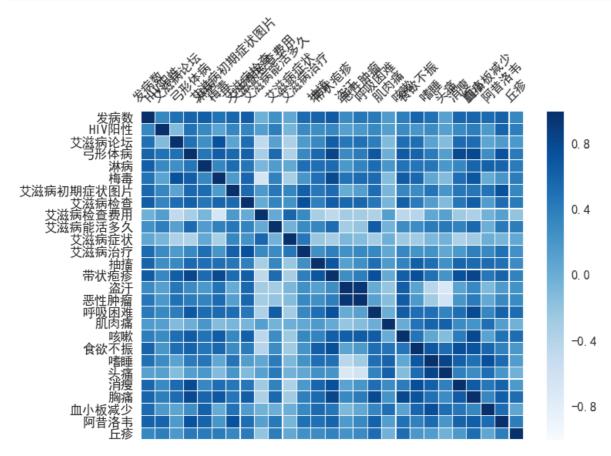
60 non-null float64

60 non-null float64 60 non-null float64

60 non-null float64

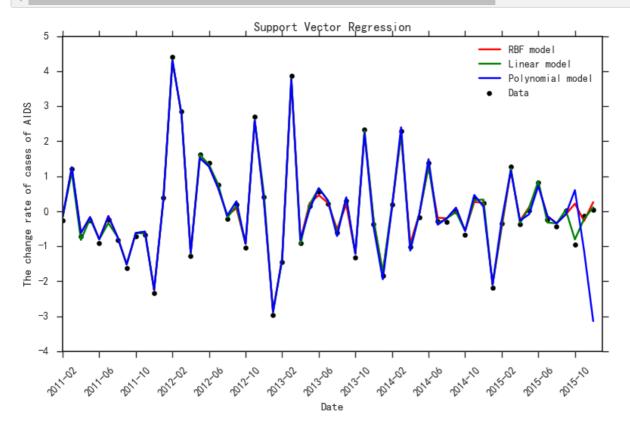
```
In [17]:
```

```
#>>> 0> # Function to visualize correlation in heatmap
def plot_corr(data):
    # set appropriate font and dpi
    sns. set (
           font='SimHei', # fix Chinese output
           font_scale=1.2,
           rc={'axes.unicode_minus':False} )# fix negative notation output)
   sns. set_style({"savefig.dpi": 100})
    # plot it out
   ax = sns.heatmap(data, cmap=plt.cm.Blues, linewidths=.1)
    # set the x-axis labels on the top
   ax. xaxis. tick top()
    # rotate the x-axis and y-axis labels
   plt. xticks (rotation=45)
   plt.yticks(rotation=0)
    # get figure (usually obtained via "fig, ax=plt. subplots()" with matplotlib)
    fig = ax.get_figure()
    # specify dimensions and save
# Get correlation
data = df. ix[:, 1:]. corr(method='pearson')
plot_corr(data)
plt. show()
```



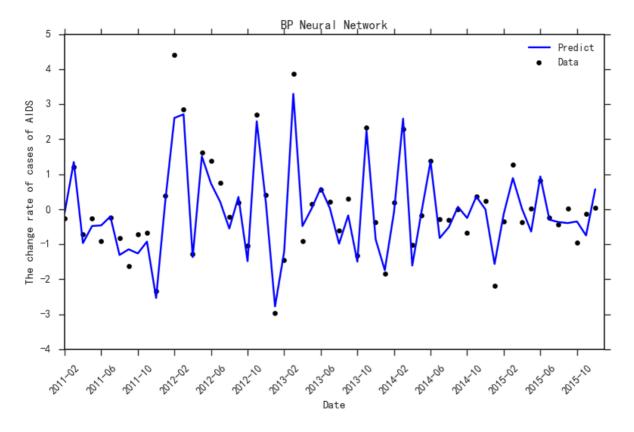
```
\#\rangle\rangle\rangle 0> Function for Splitting dataset into trainset and testset
def train_test_split(X, y, test_size=0.05):
   test_num = int(np.ceil(test_size * len(X)))
   X_train = X[:-test_num]
   X_test = X[-test_num:]
   y_train = y[:-test_num]
   y test = y[-test num:]
   return X_train, X_test, y_train, y_test
#>>> 3> Prepare datasets
from sklearn import preprocessing
# Get the percent change
column_names = df.columns
for name in column_names[1:]:
   df[name] = df[name].pct_change()
# Drop the first row without value
df. drop(0, axis=0, inplace=True)
# Standardize data
df_matrix = df.ix[:,1:].as_matrix()
df_matrix = preprocessing.scale(df_matrix.T).T
# Split dataset
X=df_matrix[:, 1:]
y=df_matrix[:,0]
X_train, X_test, y_train, y_test=train_test_split(X, y, 0.05)
# Convert format of date
date_tick = df.ix[:, 0].dt.strftime('%Y-%m')
```

```
#***********
#
#
     SVM
#
from sklearn.svm import SVR
from sklearn import cross_validation
from matplotlib.ticker import FuncFormatter, MaxNLocator
from sklearn. metrics import mean squared error
#>>> 1> Function for Using SVM to predict
def svr predict aids (X train, y train, dates, X, y, x):
    Visualize applying SVM
    # Build model
    svr_rbf = SVR(kernel= 'rbf', C= 1e3, gamma= 0.1)
    svr_rbf.fit(X_train, y_train)
    svr_lin = SVR(kernel= 'linear', C= 1e3)
    svr_lin.fit(X_train, y_train)
    svr_poly = SVR(kernel= 'poly', C= 1e3, degree= 2)
    svr_poly.fit(X_train, y_train)
   fig = plt. figure()
   ax = fig. add subplot(1, 1, 1)
    ax. set_xlim(0, len(date_tick))
    ax. scatter(np. reshape(list(range(len(date_tick))), (len(date_tick), 1)), y, color='black', label
    ax.plot(range(len(X)), svr_rbf.predict(X), color='red', label='RBF model') # plotting the line
    ax.plot(range(len(X)), svr_lin.predict(X), color='green', label='Linear model') # plotting the
    ax.plot(range(len(X)), svr_poly.predict(X), color='blue', label='Polynomial model') # plotting
    ax. set xlabel ('Date')
    ax.set_ylabel('The change rate of cases of AIDS')
    ax. set_title('Support Vector Regression')
    ax. xaxis. set_major_locator(MaxNLocator(nbins=15, integer=True))
    dates = list(date tick)
    dates. append (dates [-1])
    dates = np. array (dates)
    ax. set_xticklabels(dates. reshape(int(len(dates)/4), 4)[:, 0])
   plt.legend(loc='best')
    plt. xticks (rotation=45)
    plt. tight layout()
   plt. show()
    score_rbf = mean_squared_error(svr_rbf.predict(X_test), y_test)
    score lin = mean squared error(svr lin.predict(X test), y test)
    score_poly = mean_squared_error(svr_poly.predict(X_test), y_test)
    return score rbf, score lin, score poly
# SVM to predict
score_rbf, score_lin, score_poly = svr_predict_aids(X_train, y_train, date_tick, X, y, len(y)-1)
print ("MSE of rbf:", score rbf, '\n', "MSE of linear:", score lin, '\n', "MSE of poly:", score rbf)
```



MSE of rbf : 0.486988051034 MSE of linear : 0.0131702659883 MSE of poly : 0.486988051034

```
#
#
     Back propagation neural network
#
from keras. models import Sequential
from keras. layers. core import Dense, Dropout, Activation
from keras.wrappers.scikit learn import KerasRegressor
from sklearn.grid search import GridSearchCV
from keras. optimizers import SGD
model = Sequential()
# Create model
model.add(Dense(30, input dim=26, activation='relu'))
#mode1. add(Dropout(0.2))
model.add(Activation('relu'))
model.add(Dense(1, activation='linear'))
# Complie
optimizer = SGD(1r=0.01, momentum=0.9)
model.compile(loss='mean_squared_error', optimizer=optimizer)
# Fit
results = model.fit(X_train, y_train, epochs = 10, batch_size=10, verbose=0)
score_bp = model.evaluate(X_test, y_test)
print("MSE of testset IN BP: ", score bp)
print(y_test)
# Predict
y_pred = model.predict(X)
# plot prediction data
fig 2 = plt. figure()
ax = fig_2. add_subplot (111)
ax. set_xlim(0, len(date_tick))
ax.scatter(range(len(y_pred)), y, color='black', label='Data')
ax. plot (range (len (y pred)), y pred, 'b', label='Predict')
ax. set_xlabel('Date')
ax. set ylabel ('The change rate of cases of AIDS')
ax.set_title('BP Neural Network')
ax. xaxis. set_major_locator(MaxNLocator(nbins=15, integer=True))
dates = list(date tick)
dates. append (dates [-1])
dates = np. array (dates)
ax. set_xticklabels(dates. reshape(int(len(dates)/4), 4)[:, 0])
plt.legend(loc='best')
plt. xticks (rotation=45)
plt.tight layout()
plt.show()
```



```
#>>> 0> # Function to view results
def show_best_score(grid_result):
    print best choice for keras regression
   mse_arr = np.array([])
    for params, mean_score, scores in grid_result.grid_scores_:
        print("%f (%f) with: %r" % (scores.mean(), scores.std(), params))
        mse_arr = np. concatenate((mse_arr, np. array([scores. mean(), params])))
   mse_arr = mse_arr.reshape((int(len(mse_arr)/2),2)) # reshape to 2 dimensions for calculating mse
    print("Best: %f using %s" % (mse_arr[mse_arr[:,0]==mse_arr[:,0].min()][0][0], mse_arr[mse_arr[:,0].min()][0][0]
#>>> 1> # Function to create model for tuning activation
def create_model_1(activation = 'sigmoid'):
    # create model
    model = Sequential()
    model.add(Dense(30, input_dim=26, activation='relu'))
    model. add (Activation ('relu'))
   model. add(Dense(1, activation=activation))
    # Compile model
    model.compile(loss='mean squared error', optimizer='SGD')
    return model
# Find optimal activation
model = KerasRegressor(build_fn=create_model_1, epochs = 10, batch_size=10, verbose=0)
# define the grid search parameters
activation = ['softmax', 'softplus', 'softsign', 'relu', 'tanh', 'sigmoid', 'hard_sigmoid', 'linear'
param grid = dict(activation=activation)
grid = GridSearchCV(estimator=model, param_grid=param_grid)
grid_result = grid.fit(X_train, y_train)
show best score (grid result)
2.847612 (0.711022) with: {'activation': 'softmax'}
2.131128 (0.797848) with: {'activation': 'softplus'}
1.991538 (0.976736) with: {'activation': 'softsign'}
2.027911 (0.789575) with: {'activation': 'relu'}
1.915652 (0.760310) with: {'activation': 'tanh'}
2.021431 (0.806095) with: {'activation': 'sigmoid'}
2.\,\,082010\ (0.\,816970)\ \text{with: \{'activation': 'hard\_sigmoid'\}}
2.131582 (0.930664) with: {'activation': 'linear'}
Best: 1.915652 using {'activation': 'tanh'}
```

```
#>>> 2> # Function to create model for tuning optimal batchsize and training epochs
def create_model_2(batch_size = 30, input_dim = 26):
    # create model
   model = Sequential()
    model. add(Dense(batch_size, input_dim=input_dim, activation='relu'))
    model.add(Activation('relu'))
    model.add(Dense(1, activation='linear'))
    # Compile model
    model.compile(loss='mean_squared_error', optimizer='SGD')
    return model
model = KerasRegressor(build fn=create model 2, verbose=0)
batch size = [10, 20, 30]
epochs = [10, 50]
param_grid = dict(batch_size=batch_size, nb_epoch=epochs)
grid = GridSearchCV(estimator=model, param_grid=param_grid)
grid_result = grid.fit(X_train, y_train)
show best score(grid result)
2.131330 (0.813639) with: {'nb_epoch': 10, 'batch_size': 10}
2.046298 (1.029935) with: {'nb_epoch': 50, 'batch_size': 10}
2.580343 (1.162664) with: {'nb_epoch': 10, 'batch_size': 20}
2.222788 (0.926217) with: {'nb_epoch': 50, 'batch_size': 20}
1.812820 (0.855541) with: {'nb_epoch': 10, 'batch_size': 30}
1.993234 (0.840102) with: {'nb_epoch': 50, 'batch_size': 30}
Best: 1.812820 using {'nb_epoch': 10, 'batch_size': 30}
```

```
#>>> 3> # Function to create model for tuning optimizer
def create_model_3(optimizer = 'SGD'):
    # create model
   model = Sequential()
    model.add(Dense(30, input_dim=26, activation='relu'))
    model.add(Activation('relu'))
    model.add(Dense(1, activation='linear'))
    # Compile model
    model.compile(loss='mean squared error', optimizer=optimizer)
    return model
# Find optimaizer
model = KerasRegressor(build fn=create model 3, epochs = 10, batch size=10, verbose=0)
# define the grid search parameters
optimizer = ['SGD', 'RMSprop', 'Adagrad', 'Adadelta', 'Adam', 'Adamax', 'Nadam']
param_grid = dict(optimizer=optimizer)
grid = GridSearchCV(estimator=model, param_grid=param_grid)
grid_result = grid.fit(X_train, y_train)
show best score(grid result)
1.876631 (0.959333) with: {'optimizer': 'SGD'}
2.059060 (1.042237) with: {'optimizer': 'RMSprop'}
1.884204 (0.733410) with: {'optimizer': 'Adagrad'}
1.824416 (0.792038) with: {'optimizer': 'Adadelta'}
1.846100 (0.716938) with: {'optimizer': 'Adam'}
2.383727 (0.977617) with: {'optimizer': 'Adamax'}
1.947258 (0.973459) with: {'optimizer': 'Nadam'}
Best: 1.824416 using {'optimizer': 'Adadelta'}
```

```
#>>> 4> # Function to create model for tuning learning rate and momentum
def create model 4(learn rate=0.01, momentum=0.9):
    # create model
    model = Sequential()
    model. add(Dense(30, input_dim=26, activation='relu'))
    model.add(Activation('relu'))
    model. add (Dense (1, activation='linear'))
    # Compile model
    optimizer = SGD(1r=learn_rate, momentum=momentum)
    model.compile(loss='mean_squared_error', optimizer=optimizer)
    return model
# Find optimal learning rate and momentum
model = KerasRegressor(build_fn=create_model_4, epochs = 10, batch_size=10, verbose=0)
# define the grid search parameters
learn_rate = [0.001, 0.01, 0.1, 0.2, 0.3]
momentum = [0.0, 0.2, 0.4, 0.6, 0.8, 0.9]
param grid = dict(learn rate=learn rate, momentum=momentum)
grid = GridSearchCV(estimator=model, param_grid=param_grid)
grid result = grid. fit (X train, y train)
show_best_score(grid_result)
2.407559 (0.934983) with: {'momentum': 0.0, 'learn_rate': 0.001}
2.108527 (0.787898) with: {'momentum': 0.2, 'learn_rate': 0.001}
2.366757 (0.852796) with: {'momentum': 0.4, 'learn_rate': 0.001}
2.075800 (0.708892) with: {'momentum': 0.6, 'learn rate': 0.001}
2.064825 (0.953951) with: {'momentum': 0.8, 'learn_rate': 0.001}
1.889727 (0.806629) with: {'momentum': 0.9, 'learn rate': 0.001}
2.394459 (1.200493) with: {'momentum': 0.0, 'learn_rate': 0.01}
2.028620 (0.982316) with: {'momentum': 0.2, 'learn rate': 0.01}
1.848259 (0.886741) with: {'momentum': 0.4, 'learn rate': 0.01}
2.250743 (0.679329) with: {'momentum': 0.6, 'learn_rate': 0.01}
1.986153 (0.722411) with: {'momentum': 0.8, 'learn rate': 0.01}
2.412243 (0.328795) with: {'momentum': 0.9, 'learn_rate': 0.01}
2.288509 (0.680591) with: {'momentum': 0.0, 'learn_rate': 0.1}
2.121893 (0.390046) with: {'momentum': 0.2, 'learn_rate': 0.1}
2.502005 (0.290244) with: {'momentum': 0.4, 'learn rate': 0.1}
2.429795 (0.835475) with: {'momentum': 0.6, 'learn rate': 0.1}
nan (nan) with: {'momentum': 0.8, 'learn_rate': 0.1}
nan (nan) with: {'momentum': 0.9, 'learn rate': 0.1}
   (nan) with: {'momentum': 0.0, 'learn rate': 0.2}
    (nan) with: {'momentum': 0.2, 'learn rate': 0.2}
nan (nan) with: {'momentum': 0.4, 'learn_rate': 0.2}
   (nan) with: {'momentum': 0.6, 'learn rate': 0.2}
   (nan) with: {'momentum': 0.8, 'learn_rate': 0.2}
    (nan) with: {'momentum': 0.9, 'learn rate': 0.2}
   (nan) with: {'momentum': 0.0, 'learn_rate': 0.3}
    (nan) with: {'momentum': 0.2, 'learn rate': 0.3}
    (nan) with: {'momentum': 0.4, 'learn rate': 0.3}
nan (nan) with: {'momentum': 0.6, 'learn_rate': 0.3}
nan (nan) with: {'momentum': 0.8, 'learn_rate': 0.3}
nan (nan) with: {'momentum': 0.9, 'learn rate': 0.3}
```

In []:

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In [ ]:
```