## Problem1

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#### 1 Introduction

# 2 Packages

# 3 Cleaning and Acquiring Data

```
In [10]: ### reading data and cleaning (less needed here though)
     ##### Actual GDP
     gdp_data = pd.read_csv('./GDPC1.csv', index_col=0, parse_dates=[0])
     gdp_data.columns = ['Actual GDP']
     gdp_data = gdp_data.reset_index()
     gdp_data['DATE'] = pd.to_datetime(gdp_data['DATE'])
     gdp_data = gdp_data.set_index('DATE')
     ##### Oil Production (Finished_Motor_Gasoline)
     ###### using the data on oil production from 1991 to 2018
     oil_production = pd.read_csv('./Weekly_US_Product_Supplied_of_Finished_Motor_Gasoline.c
                        skiprows=[0,1,2, 3])
     oil_production.columns=['Date', 'Productions']
```

```
oil_production.loc[:, 'Date'] = pd.to_datetime(oil_production['Date'])
oil_production_cleaned = oil_production.set_index('Date')
### resampling Quarterly
oil_prod_quarterly = oil_production_cleaned.resample('Q').mean()
##### 3-MONTH LIBOR (some cleaning...)
**************************************
libor3M = pd.read_csv('./USD3MTD156N.csv')
libor3M['DATE'] = pd.to_datetime(libor3M['DATE'])
libor3M =libor3M.set_index('DATE')
libor3M['USD3MTD156N'] = pd.to_numeric(libor3M['USD3MTD156N'].values[:],
     errors='coerce')
null_col = libor3M.columns[libor3M.isnull().any()] # obvious !
num_nulls = len(libor3M[libor3M.isnull().any(axis=1)][null_col]) ## ne 0
libor3M = libor3M.dropna() # num_nulls = 0
### This is a three month libor but in case the daycount is different...
libor3M = libor3M.resample('Q').mean()
##### DURABLE GOODS ORDERS
durable = pd.read csv('./DGORDER.csv')
durable = durable.dropna(how='any')
durable['DATE'] = pd.to datetime(durable['DATE'])
durable =durable.set_index('DATE')
### resampling
durable = durable.resample('Q').mean()
##### US Population
population = pd.read_csv('./POPTHM.csv')
population = population.dropna(how='any')
population['DATE'] = pd.to_datetime(population['DATE'])
population =population.set_index('DATE')
#### resampling
population = population.resample('Q').mean()
```

#### 4 Illustration

```
### there might be better ways ...
tO_oil = datetime.datetime.strptime(str(oil_prod_quarterly.index[0]), "%Y-%m-%d %H:%N
t0_gdp = datetime.datetime.strptime(str(gdp_data.index[0]), "%Y-%m-%d %H:%M:%S")
tO_libor = datetime.datetime.strptime(str(libor3M.index[0]), "%Y-%m-%d %H:%M:%S")
tO_pop = datetime.datetime.strptime(str(population.index[0]), "%Y-%m-%d %H:%M:%S")
tO_durb = datetime.datetime.strptime(str(durable.index[0]), "%Y-%m-%d %H:%M:%S")
tf_oil = datetime.datetime.strptime(str(oil_prod_quarterly.index[-1]), "%Y-%m-%d %H:%
tf_gdp = datetime.datetime.strptime(str(gdp_data.index[-1]), "%Y-%m-%d %H:%M:%M:%S")
tf_libor = datetime.datetime.strptime(str(libor3M.index[-1]), "%Y-%m-%d %H:%M:%S")
tf_pop = datetime.datetime.strptime(str(population.index[-1]), "%Y-%m-%d %H:%M:%S")
tf_durb = datetime.datetime.strptime(str(durable.index[-1]), "%Y-%m-%d %H:%M:%S")
start_date = max([t0_oil, t0_gdp, t0_libor, t0_pop, t0_durb])
end_date = min([tf_oil, tf_gdp, tf_libor, tf_pop, tf_durb])
##### filter and adjust the date
libor3M = libor3M.loc[start_date:end_date]
libor3M = libor3M.reset_index()
libor3M['DATE'] = libor3M['DATE'].apply(lambda x: x + datetime.timedelta(days=1))
libor3M = libor3M.set_index('DATE')
oil_prod_quarterly = oil_prod_quarterly.loc[start_date:end_date]
oil_prod_quarterly = oil_prod_quarterly.reset_index()
oil_prod_quarterly['Date'] = oil_prod_quarterly['Date'].apply(lambda x: x + datetime.ti
oil_prod_quarterly = oil_prod_quarterly.set_index('Date')
population = population.loc[start_date:end_date]
population = population.reset_index()
population['DATE'] = population['DATE'].apply(lambda x: x + datetime.timedelta(days=1))
population = population.set_index('DATE')
durable = durable.loc[start_date:end_date]
```

durable['DATE'] = durable['DATE'].apply(lambda x: x + datetime.timedelta(days=1))

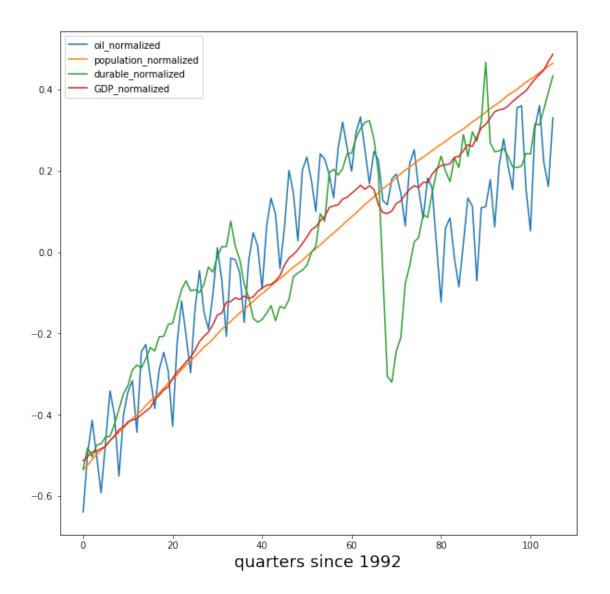
durable = durable.reset\_index()

durable = durable.set\_index('DATE')

## 5 Analysis

### 5.1 R2-squared

```
In [13]: #### visualization:
        final_dt = pd.concat([oil_prod_quarterly,
                             libor3M,
                             population,
                             durable,
                             gdp_data], axis=1).dropna()
        final_dt.columns=['oil',
                          'libor',
                          'population',
                         'durable',
                         'GDP',
        ## normalize for illustration
        final_dt_norm = (final_dt - final_dt.mean())/(final_dt.max() - final_dt.min())
        ### seems like libor is not a good fit unless we consider some non-linear relation
        ### between Libor and GDP ... so let's drop it for now.
        fig, ax = plt.subplots(figsize=[10,10])
        for col in range(final_dt_norm.shape[1]):
            if final_dt_norm.columns[col]!= 'libor':
                ax.plot(final_dt_norm.values[:, col], label=str(final_dt_norm.columns[col])+'_n
        ax.set_xlabel('quarters since 1992', **{'fontsize':18})
        plt.legend()
        plt.show()
        ###### let's now the correlation variace around the fit
        ##### with the gdp or not.
        ###### From the plot they are not perfactly in shape ... but anyway...
        y_data = final_dt.iloc[:, -1]
                                         ## qdp the target
        for i in range(4): ### since we have chosen four indicators
            X_data = final_dt.iloc[:, [i]] ## the indexs we chose
            lin_mod = linear_model.LinearRegression()
            _ = lin_mod.fit(X_data, y_data)
            print ' ========== ' ====== '
            print 'The indecator chosen is : %s ' %(final_dt.columns[i])
            print('R-Squared of Linear Regression Model:', lin_mod.score(X_data, y_data))
            print ' ========= \n\n'
```



### 5.2 Others