

## Maths6397

## HW3 Due on Tuesday 4/12/2016

### Part1 : DATA oil.gas.xlsx

Prices are listed for 1411 days Separate data into training set (first 1000 days) and a test set (last 411 days)

Use only the training data to perform 2 distinct standard linear regressions to predict separately 2 output variables “En” and “In” on day n

E = Conventional Gasoline Prices

I = Heating Oil Prices

using the same 4 Input variables Bn, Fn, Gn, Hn

B=Crude Oil Prices: West Texas ; F= Propane Prices: Mont Belvieu, Texas,

G= Henry Hub Natural Gas Spot Price; H= Crude Oil Prices: BrentEurope

Give the coefficients of the 4 regressions including the constant terms.

For each output variable Yn, call predYn the predicted value computed by the regression . Plot on the same graph the curves Yn and predYn to visualize the prediction errors

For each output variable Y compute separately the standard deviations STDtrain and STDtest of the prediction errors on the training set and on the test set

For each output variable Y compute mY = mean value of Y, and evaluate regression performance by STDstrain/mY and STDtest/mY

### Part 2 DATA metals.xlsx

Prices are listed for 838 days ‘Separate data into training set (first 600 days) and a test set (last 238 days)

Use only the training data to perform a **polynomial kernel non linear regression** to predict on day “n” the **future** output variable “J<sub>n+1</sub>” = price of aluminium on day “n+1”

The input variables will be the following 4 day “n” values

Dn= steel price ; En= nickel price; In= copper price; Jn= aluminium price

All variables D,E,I,J should first be normalized by the rescaling D/mD , E/mE, I/ml, J/mJ where mD, mE, ml, mJ are the respective mean values computed on the training set

Try only a polynomial kernel of degree r = 2.

For each day  $n$ , call  $\text{pred}J_{n+1}$  the predicted value of  $J_{n+1}$  computed by the nonlinear regression using the input values  $D_n, E_n, I_n, J_n$ .

Explain as explicitly as possible how  $\text{pred}J_{n+1}$  is computed in terms of  $D_n, E_n, I_n, J_n$ .

Plot on the same graph the curves  $J_{n+1}$  and  $\text{pred}J_{n+1}$  to visualize the prediction errors

Compute separately the standard deviations  $\text{STD}_{\text{train}}$  and  $\text{STD}_{\text{test}}$  of the prediction errors on the training set and on the test set

Evaluate regression performance by  $\text{STD}_{\text{train}}/\text{mJ}$  and  $\text{STD}_{\text{test}}/\text{mJ}$

### Part 3: Same data as Part2

Same problem as part 2 but implement a **non linear kernel ridge regression** (polynomial kernel of degree 2.) Optimize the performance by a proper choice of the coefficient  $\lambda$  involved in the ridge regression