

Project Assignment
System
Identification
2023-2024

Part 1. Fitting an unknown function

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Problem Statement

- The primary objective of the project is to create an appropriate approximation for an unknown function, using the data set provided by the teacher. In pursuit of this goal, I constructed a polynomial approximator for the function given by the data (index 9) and established a model fitting using linear regression. A lot of advanced models use linear regression as a base. It offers a simple model, characterized by its ease of understanding.

Generation of Regressors and tuning process

In order to find the polynomial of a certain degree m , I created a function: `generatePoly`, which is my key feature, that automates the process. Let's consider the regressed variable Y , defined as:

$$Y = \begin{bmatrix} y[1,1] \\ y[1,2] \\ \dots \\ y[L1,L2] \end{bmatrix}$$

,where $L1$ represents the length of $x1$ (input parameter given in the data set) and $L2$ represents the length of $x2$ (input parameter given in the data set).

The regressor ϕ will be determined using the function I mentioned before, starting from the formula:

$$\phi = [\phi(1, 1), \phi(1, 2), \dots, \phi(L1, L2)]^T$$

,where the elements are the coefficients determined by my function.

The parameter vector θ :

$$\theta = \begin{bmatrix} \theta_1 \\ \theta_2 \\ \dots \end{bmatrix}$$

The determination of θ using the formula:

$$Y = \phi^* \theta$$

I extract θ using left matrix division:

$$\theta = \phi \backslash Y$$

Having the regressor ϕ and θ , I can find the approximation:

$$\hat{Y} = \phi \cdot \theta$$

After I calculated \hat{Y} , I can calculate the MSE as a function of m with the formula:

$$\frac{1}{N} \sum_{i=1}^N (y_i - \hat{y}_i)^2$$

, where N represents the number of data points, y_i is the output parameter for the identification data and the other one is the approximation computed for the identification data.

The **evolution of the errors** is presented below (Figure 1). With blue is represented the evolution of MSE for the identification data set and with red for validation data set .

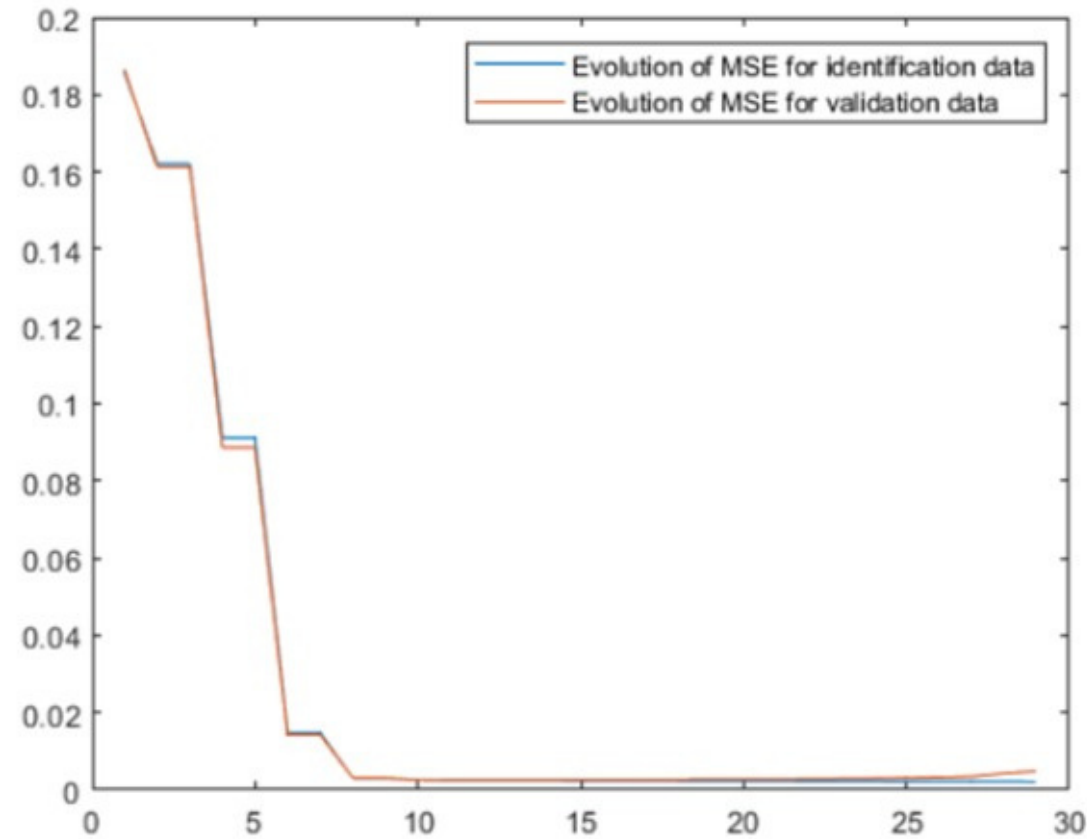


Figure 1



Tuning Results

- In Figure 2 I have 3 subplots containing the identification data representation, having the best approximator data representation for $m = 30$ and $MSE = 0.0021451$ and the approximator fitting on the identification data.

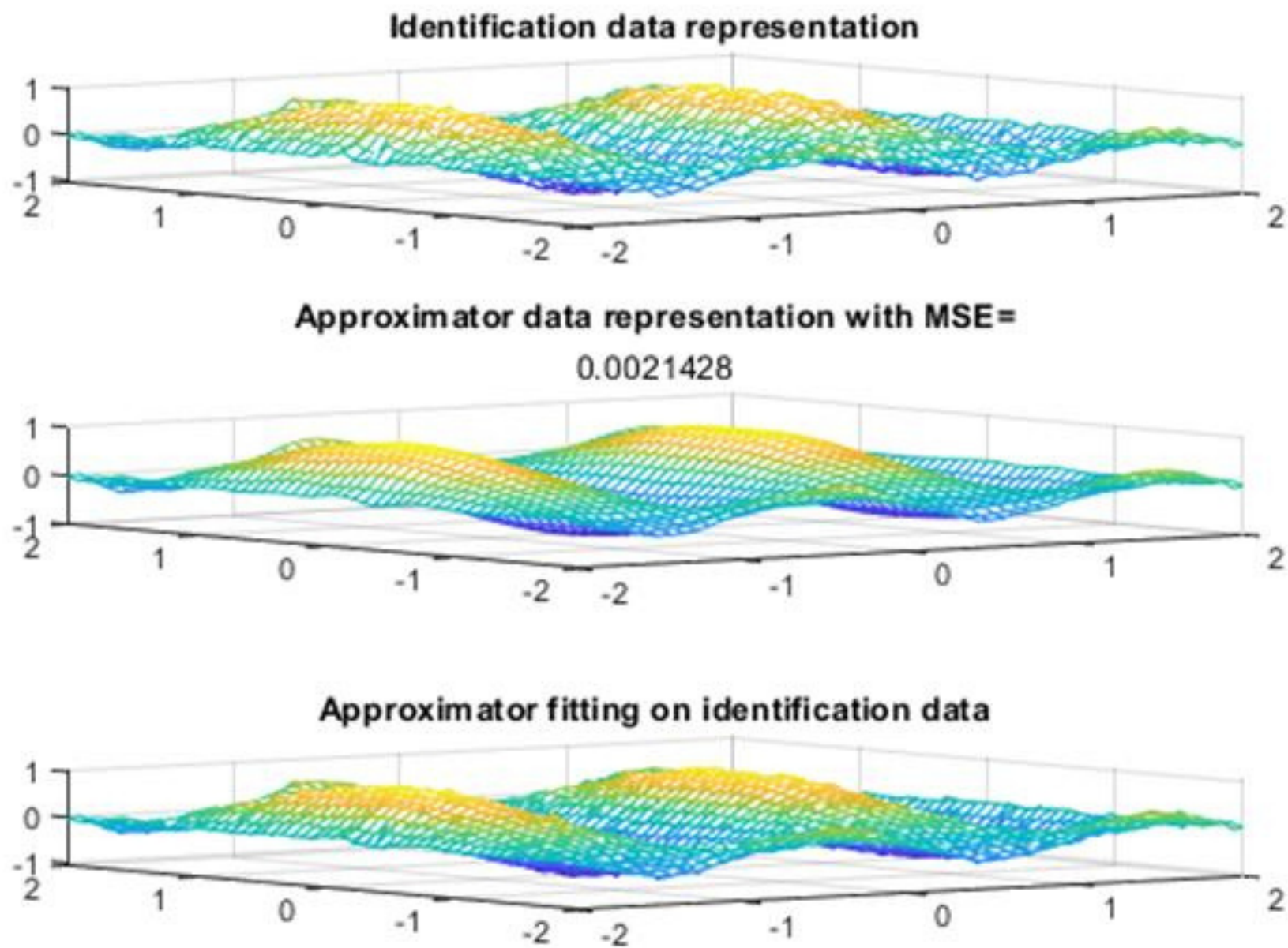


Figure 2

In Figure 3 I have 3 subplots containing the Validation data representation, the best approximator data representation for $m = 11$ and $MSE = 0.0025808$ and the approximator fitting on the validation data.

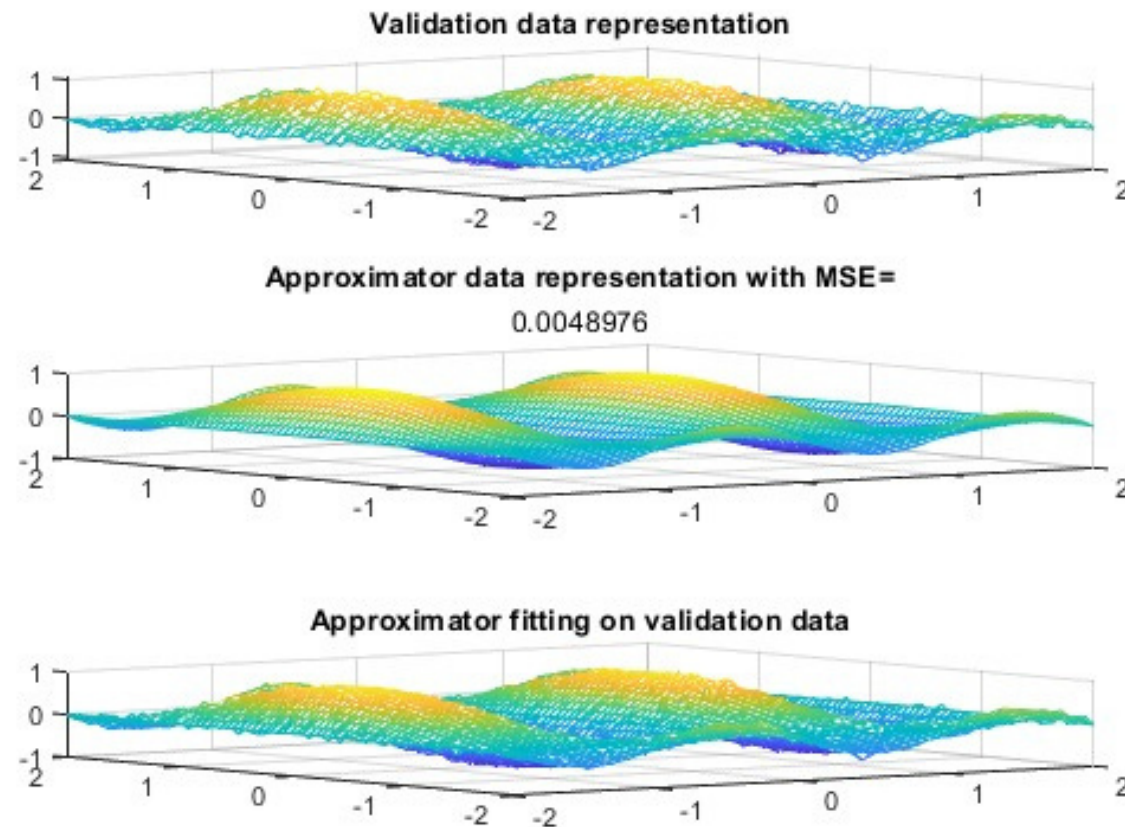


Figure 3

Code

```
load("proj_fit_09.mat")
% We declare our variables
x1=id.X{1,1};
x2=id.X{2,1};
y=id.Y;

x1v=val.X{1,1};
x2v=val.X{2,1};
yval=val.Y;

MSEmin=200;
MSEidMin=300;
mseID_vector=[];
mse_vector=[];
yHatID_min=[];
yHat_min=[];
```

```

% We start a for loop to determine the most convenient degree of the
% polynomial
for m=2:30
    PHI = [];% We need to calculate the regressor for id data
    line_nr= 1;

    for i=1:length(x1)
        for j=1:length(x2)
            PHI_line = generatePoly(i,j,m,x1,x2);% We created a function to
generate the polynomial
            for index=1:length(PHI_line)
                PHI(line_nr,index) = PHI_line(index);
            end
            line_nr = line_nr + 1;
        end
    end

% Calculation of theta
y_resh=reshape(y,length(x1)*length(x2),1);
THETA=PHI\y_resh;

yhatID=PHI*THETA;% Calculation of the aproximation of Y for identification

```

```

y_r = reshape(y,length(y)*length(y), 1);
MSEid=1/length(yhatID)*sum((y_r-yhatID).^2);% Calculation of the MSE of the
identification data
mseID_vector(m-1)=MSEid;

% Finding the smallest MSE that will give the most convenient degree for
% the identification data
    if MSEidMin>MSEid
        MSEidMin=MSEid;
        yHatID_min=yhatID;
        degreeID_min=m;
    end

% We calculate the regressor for the validation data
PHIval = [];
linenr_val = 1;
    for i=1:length(x1v)
        for j=1:length(x2v)
            PHI_line = generatePoly(i,j,m,x1v,x2v);
            for index=1:length(PHI_line)
                PHIval(linenr_val,index) = PHI_line(index);
            end
            linenr_val = linenr_val + 1;
        end
    end

yhat=PHIval*THETA;% Calculation of the aproximation of Y for validation

```

```
yval_r = reshape(yval, length(yval)*length(yval), 1);  
MSE=1/length(yhat)*sum((yval_r-yhat).^2);% Calculation of the MSE of the  
validation data  
mse_vector(m-1)=MSE;
```

```
% Finding the smallest MSE that will give the most convenient degree for  
% the validation data
```

```
    if MSEmin>MSE  
        MSEmin=MSE;  
        yHat_min=yhat;  
        degree_min=m;  
    end
```

```
end
```

```
subplot(311)  
mesh(x1v,x2v,yval);  
title('Validation data representation')  
subplot(312)  
yHat_r=reshape(yHat_min,71,71);  
mesh(x1v,x2v,yHat_r);  
title('Approximator data representation with MSE=',MSEmin)
```



```
subplot(313)
mesh(x1v,x2v,yval);
hold on;
newZLim=[-1,1];
zlim(newZLim);
mesh(x1v,x2v,yHat_r);
title('Approximator fitting on validation data')
hold off;
figure;
subplot(311)
mesh(x1,x2,y);
title('Identification data representation')
subplot(312)
yHatID_r=reshape(yHatID_min,51,51);
mesh(x1,x2,yHatID_r);
title('Approximator data representation with MSE=',MSEidMin)
subplot(313)
mesh(x1,x2,y);
hold on
mesh(x1,x2,yHatID_r);
hold off;
title('Approximator fitting on identification data')
figure;
plot(mseID_vector)
hold on;
plot(mse_vector)
hold off;
legend('Evolution of MSE for identification data','Evolution of MSE for validation data')
```

Function:

```
function poly = generatePoly(i,j,m,x1,x2)
    poly = [];
    for x1_power = 0:m
        for x2_power = 0:m
            if (x1_power + x2_power) <= m
                poly = [poly, x1(i)^x1_power * x2(j)^x2_power];
            end
        end
    end
end
```