

System Identification Project



A brief description

We are given:

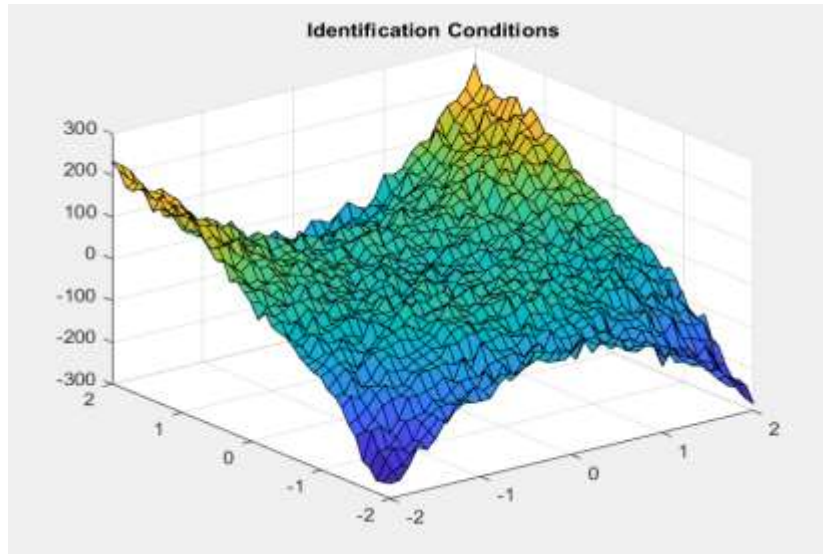
A set of grid coordinates X for the inputs, where X is a cell array of two vectors, each vector $X\{1\}(i), X\{2\}$ containing n grid points for input dimension dim .

A set of corresponding outputs Y , a matrix of size $n \times n$, where $Y(i,j)$ is equal to the value of f at point $(X\{1\}(i), X\{2\}(j))$.

Using the identification set we calculate Φ matrix (the regressors matrix) and further using that we can generate the values of θ vector.

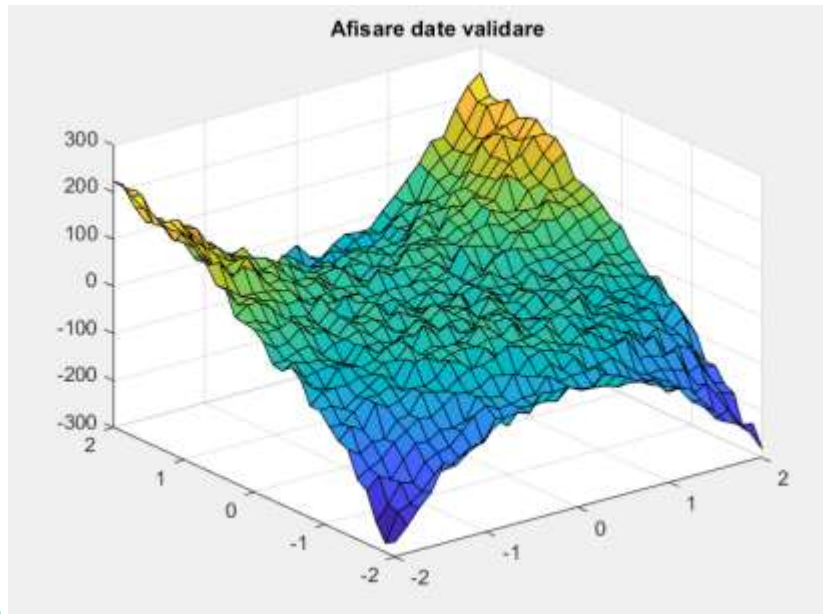
After generating the θ vector we calculate the Φ matrix for de validation set using it to create our approximation (\hat{y}).

Understanding the project

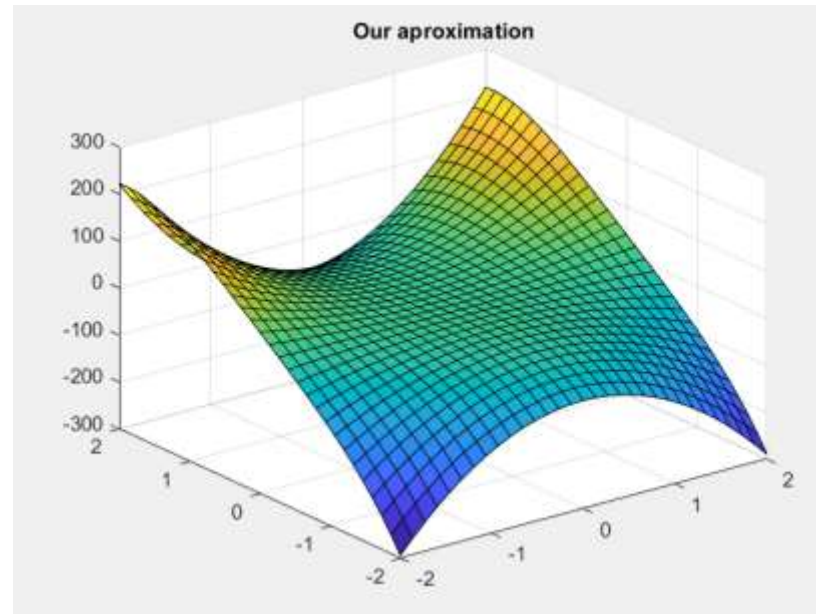


Step 1

θ

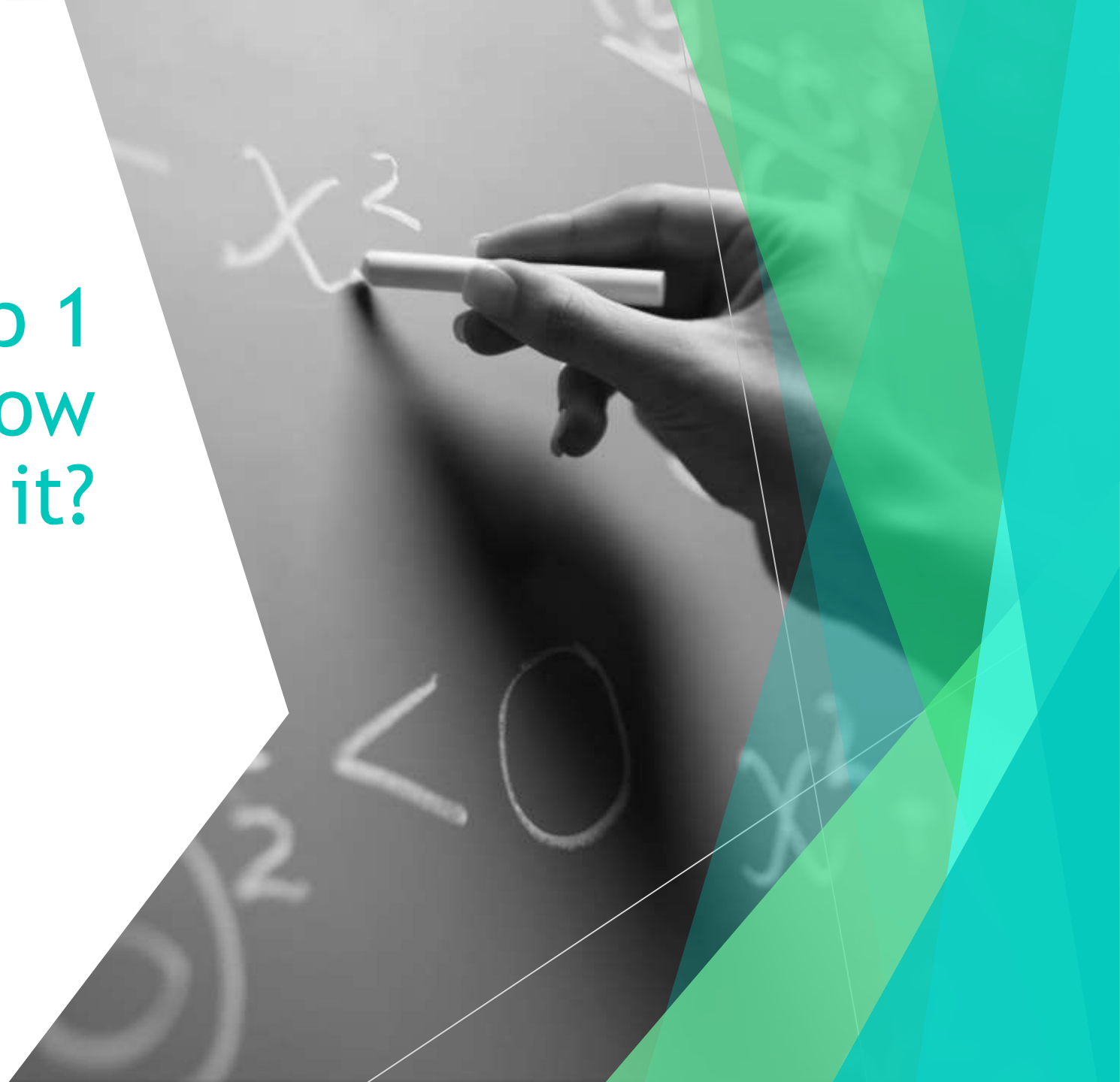


Step 2



Step 1

Who is θ and how do we calculate it?



To determine how many columns we have in the Φ matrix we observe:

$n = 1 \Rightarrow 3$ terms

$n = 2 \Rightarrow 6$ terms

$n = 3 \Rightarrow 10$ terms

$n = 4 \Rightarrow 15$ terms

We can see that for the n -th degree we use the formula:

$$\text{No. of Terms on each line} = \frac{(n+1)(n+2)}{2}$$

The matrix of regressors for a specific degree has this formula for each line

$$\underline{x^2} + x\underline{y} + \underline{y^2} + \underline{x} + \underline{y} + \underline{1} \quad n=2$$

$$\underline{x^3} + x^2\underline{y} + x\underline{y^2} + \underline{y^3} + \underline{x^2} + x\underline{y} + \underline{y^2} + x + y + \underline{1} \quad n=3$$

$$\underline{x^4} + x^3\underline{y} + x^3 + x^2\underline{y^2} + x^2\underline{y} + x^2 + x\underline{y^3} + x\underline{y^2} + x\underline{y} + x + y^4 + y^3 + y^2 + y + \underline{1} \quad n=4$$

We observe that the sum is symmetric

For this example we consider grade = 4

$$\begin{aligned} & x^{n-0}y^0 + x^{n-1}y^1 + x^{n-2}y^2 + x^{n-3}y^3 + x^{n-4}y^4 + \text{pt } n = \text{grad} \\ & + x^{n-0}y^0 + x^{n-1}y^1 + x^{n-2}y^2 + x^{n-3}y^3 + \text{pt } n = \text{grad} - 1 \\ & + x^{n-0}y^0 + x^{n-1}y^1 + x^{n-2}y^2 + \text{pt } n = \text{grad} - 2 \\ & + x^{n-0}y^0 + x^{n-1}y^1 + \text{pt } n = \text{grad} - 3 \\ & + x^0y^0 \quad n = \text{grad} - 4 \end{aligned}$$

What the system of equations looks like:

$$\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(N) \end{bmatrix} = \begin{bmatrix} p_1(1) & p_2(1) & \dots & p_n(1) \\ p_1(2) & p_2(2) & \dots & p_n(2) \\ \vdots & \vdots & & \vdots \\ p_1(N) & p_2(N) & \dots & p_n(N) \end{bmatrix} \cdot \begin{bmatrix} \theta_1 \\ \theta_2 \\ \vdots \\ \theta_n \end{bmatrix}$$

Where

$$n = \frac{(\text{grad} + 1)(\text{grad} + 2)}{2}$$

N = length of x_1 multiplied with the length of x_2

Step 2

Creating our aproximation



My biggest issue
with the project.
(symbolic variables)



$$\text{expand}((1+a+b)^{\text{degree}})$$

degree = 2

$$\Rightarrow a^2 + 2ab + 2a + b^2 + 2b + 1$$

use coeffs $\rightarrow [1 \quad 2 \quad 2 \quad 1 \quad 2 \quad 1]$

$$\text{Phi} \rightarrow \left[\frac{a^2}{1} \quad \frac{2ab}{2} \quad \frac{2a}{2} \quad \frac{b^2}{1} \quad \frac{2b}{2} \quad 1 \right]$$

Conclusion

The task I had for the project was simple, yet complicated. Creating the approximation for a 3D system using linear regression appeared to be an interesting topic that made me push my mind to find the best algorithm I could and make the code work as I wanted

My conclusion regarding this project is that despite the hard beginning I had in understanding the task, with the help of my project teacher Tudor I managed to make small steps in the right direction and after a lot of work that I put into this project I an algorithm that calculates a correct approximation using linear regression that is fast and optimized.

My goal was to write a code that can be understood by everyone and I hope that I did a good job presenting my idea to you.

% Algoritm pentru substitutia valorilor simbolice una cate

una

syms a b;

PHI = [];

k= expand((1+a+b).^degree)

child = children(k);

L = length(child);

for index=1:L

regresor(index) = child{index};

coe(index) = coeffs(child{index});

row(index) = regresor(index)/coe(index);

end

for i = 1:N

Y = [Y, y(i,:)];

ROW = subs(row,a,x1(i));

for j = 1:N

ROW2 = subs(ROW,b,x2(j));

PHI = [PHI; ROW2];

end

end

PHI = double(PHI);

theta = PHI \ Y';

```
%% INCEPUT DE PROGRAM  
tic;  
clear; clc;  
%% INCARCAM SI CITIM DATELE
```

```
load("proj_fit_19.mat")
```

```
%Date Identificare
```

```
x1 = id.X{1};
```

```
x2 = id.X{2};
```

```
y = id.Y;
```

```
%Date Validare
```

```
valx1 = val.X{1};
```

```
valx2 = val.X{2};
```

```
valy = val.Y;
```

```
Yval = []; % Reshape matricea valy pentru a fi o singura linie
```

```
YhatFinal = []; % Matricia Yhat care este pentru degree ul cel mai optim
```

```
Y = []; % Acelasi lucru ca la Yval
```

```
MSE = []; % Vectorul MSE pentru Erori la Validare
```

```
MSEid = []; % Vectorul MSE pentru Erori la Identificare
```

```
n=30; % Aici alegem Degree ul
```

%% AFLAREA MATRICII PHI, YHAT, THETA, MSE

for k=1:n % for-ul care face cate o matrice PHI pentru fiecare degree

Y = reshape(y, [], 1); % dam reshape la y intr un vector coloana sa il pot afla pe theta

Yval = reshape(valy, [], 1); % dam reshape la valy ca sa putem afla yhat (avem nevoie de matricie linie)

valreshape = ((k+1)*(k+2))/2; % un fel de suma lui Gauss, ca sa stim cum trebuie sa i dam reshape lui PHI in functie de degree

PHIval = []; % declaram o matrice PHI de validare la fiecare iteratie a lui k

PHI = []; % declaram o matrice PHI la fiecare iteratie a lui k

% DATELE DE IDENTIFICARE

for i=1:length(x1) %for-ul care creaza linile matricii PHI. selecteaza primul element din x1 care se inmulteste cu toate valorile din x2, etc. Contorul lui x1

row = []; % cream un rand nou la fiecare linie noua

for j=1:length(x2) % for-ul care creaza fiecare element de pe o linie. Contorul lui x2

m = k; % m este gradul ca sa stim de unde coboara contorul

while(m >= 0) % cat timp m este mai mare decat 0

contor = 0; % avem un contor care ia 0 si va creste pana la m

while(contor <= m) % cat timp contorul este mai mic decat m

row = [row; x1(i)^(contor)*x2(j)^(m-contor)]; % contruim fiecrae element astfel: a^0*b^m | a^1*b^{m-1} | a^2*b^{m-2} | etc... (in coloana)

contor = contor + 1; % crestem contorul

end

m = m - 1; % scadem m-ul

end

end

PHI = [PHI; row]; % concatenam fiecare rand in matricea PHI (va fi un vector coloana)

end

PHI = reshape(PHI, valreshape, []); % dam reshape la matricea PHI in functie de ce degree aste folosind VALRESHAPE ce l am calculat mai sus

PHI = PHI'; % transpunem matricea PHI ca sa o putem impartii cu matricea Y

theta = PHI \ Y; % aflam THETA

Yhatid = PHI * theta; % aflam YHAT pentru datele de identificare

N = length(x1); % N este lungimea vectorului x din datele de identificare

for c=1:N % facem suma de erori cu un for de la 1 la N

mse = (Y(c)-Yhatid(c)).^2; % calculam MSE ul cu formula de $Y - Yhat$ de Identificare

end

MSEid(k)=1/N * mse; % folosim formula finala pentru MSE care este suma inmultita cu 1/N

if k == 1

Minid = MSEid(1);

elseif MSEid(k)<Minid

Minid = MSEid(k);

YhatIDFinal = Yhatid;

end

% DATELE DE VALIDARE

% pentru datele de validare facem acelasi lucru ce am facut si pentru

% datele de identificare :)

for i=1:length(valx1)

rowVal = [];

for j=1:length(valx2)

m = k;

while(m >= 0)

contor = 0;

while(contor <= m)

rowVal = [rowVal; valx1(i)^(contor)*valx2(j)^(m-contor)];

contor = contor + 1;

end

m = m - 1;

end

end

PHlval = [PHlval; rowVal];

end

PHlval = reshape(PHlval, valreshape, []);

PHlval = PHlval';

yHat = PHlval * theta;

N = length(valx1); % N este lungimea vectorului x din datele de validare

for c=1:N

mse = (Yval(c)-yHat(c)).^2;

end

MSE(k)=1/N * mse;

%%%

%%%

% alegem primul minim ca fiind MSE(1) iar dupa in else if verificam %

% daca gasim un MSE mai bun decat cel gasit anterior iar daca este mai %

% bun, YhhatFinal o sa devina Yhat de degree-ul unde am gasit MSE-ul %

% cel mai mic %

%%%

%%%

if k == 1

Min = MSE(1);

elseif MSE(k)<Min

Min = MSE(k);

YhatFinal = yHat;

end

end


```
%% GRAFICELE DE MSE-uri
```

```
% plotez grafu de erori
```

```
b = 1:n;
```

```
figure;
```

```
subplot(221)
```

```
plot(b, MSE);
```

```
title(["MSE graphic (from 1 to " num2str(n)]);
```

```
subplot(223)
```

```
plot(b, MSEid);
```

```
title(["MSEid graphic (from 1 to " num2str(n)]);
```

```
subplot(222)
```

```
plot(b, MSE);
```

```
axis([1 30 -5 40])
```

```
title(["Zoomed MSE graphic (from 1 to " num2str(n)]);
```

```
subplot(224)
```

```
plot(b, MSEid);
```

```
axis([1 30 -5 40])
```

```
title(["Zoomed MSEid graphic (from 1 to " num2str(n)]);
```

```
%% GRAFICUL DATELOR DE IDENTIFICARE
%conditiile initiale (id X si Y )
figure;
surf(x1, x2, y);
title("Identification set");
%% GRAFICUL DATELOR DE VALIDARE
%conditiile de verificare (val X1 si X2 si Yhat)
YHAT = reshape(YhatFinal, 31, 31);
figure
surf(valx1,valx2,val.Y);
title("Validation set");
%% APROXIMAREA VALIDARE
figure
surf(valx1, valx2, YHAT);
title("Our aproximation for the validation set");
%% APROXIMAREA IDENTIFICARE
figure
YHATID = reshape(YhatIDFinal, 41, 41);
surf(x1, x2, YHATID);
title("Our aproximation for the identification set");
%% VALIDAREA SI APROXIMAREA PE ACELASI GRAFIC
YHAT = reshape(YhatFinal, 31, 31);
figure
surf(valx1,valx2,val.Y);
title("Afisare date validare");
hold on;
surf(valx1, valx2, YHAT);
title("The validation set and the aproximation on the same graph");

toc;
```