

Machine Learning

Practical-1

Non-linear regression (NLR)

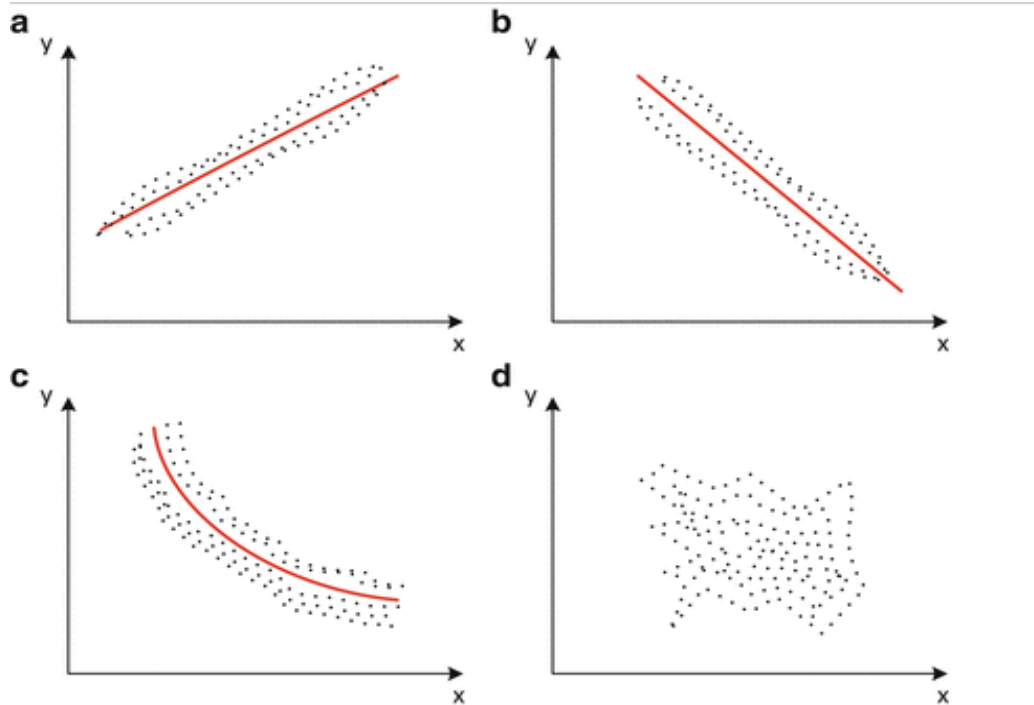
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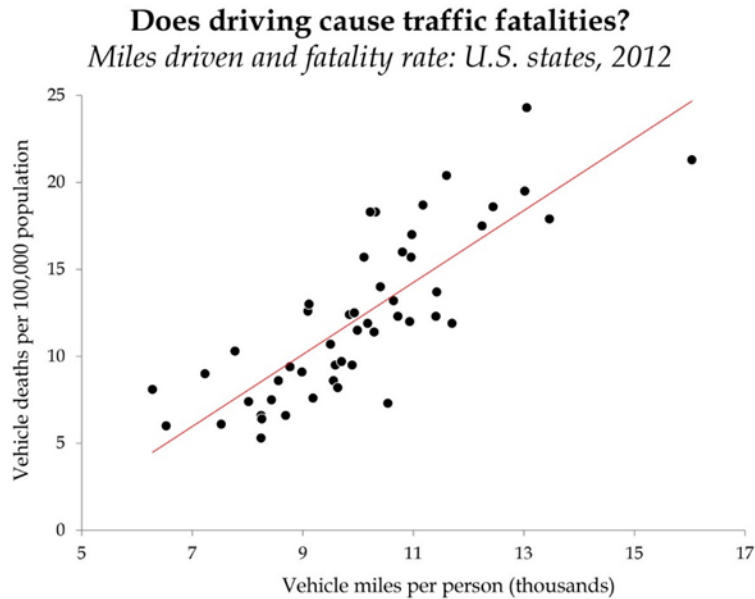
Regression analysis

Regression models describe the correlation between input and output variables.

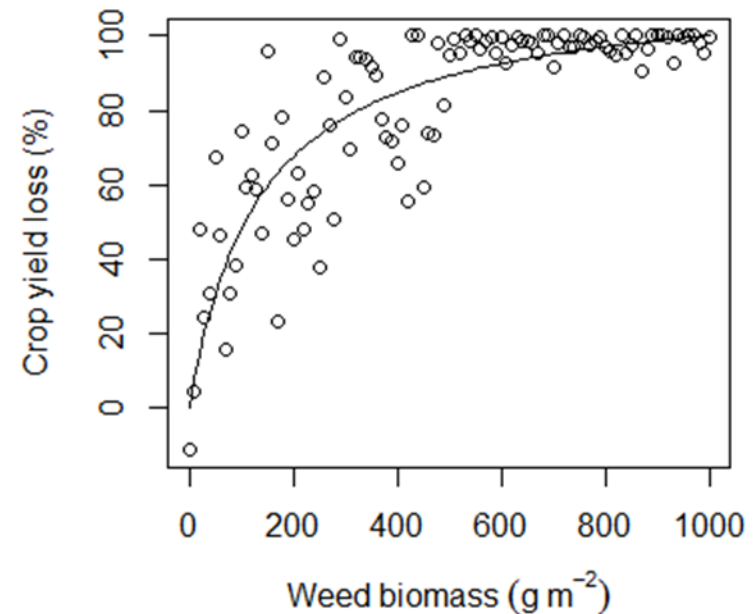


Different types of correlation between Y and X . **(a)** Positive linear correlation **(b)** negative linear correlation **(c)** nonlinear correlation **(d)** no correlation

Example of linear and non-linear regression model



Linear Regression Model



Nonlinear Regression model

Linear regression model:

Linear regression is used to model a linear relationship between a continuous dependent variables $Y(output)$ and independent variables $X(input)$.

- **fitlm:** Creates linear regression model.

Syntax:

```
mdl = fitlm(x,y);  
mdl = fitlm(x,y,modelspec);
```

Description:

x: Predictor variables(Input)

y: Response variable(output)

modelspec: Model specification

- **Predict:** Predict response of linear regression model

Syntax:

```
ypred = predict(mdl,Xnew);
```

Example of Linear regression function: fitlm

```
clear;clc;  
rng(1);  
load hald % built-in data set in Matlab  
mdl = fitlm(ingredients,heat,'linear') ;  
heatpred = predict(mdl,ingredients);  
table( heat( : ),heatpred( : ), 'VariableNames',...  
       {' Heat',' Predicted_heat'}) %Show the results  
of output data and predicted output.
```

Example of Linear regression function: fitlm (cont.)

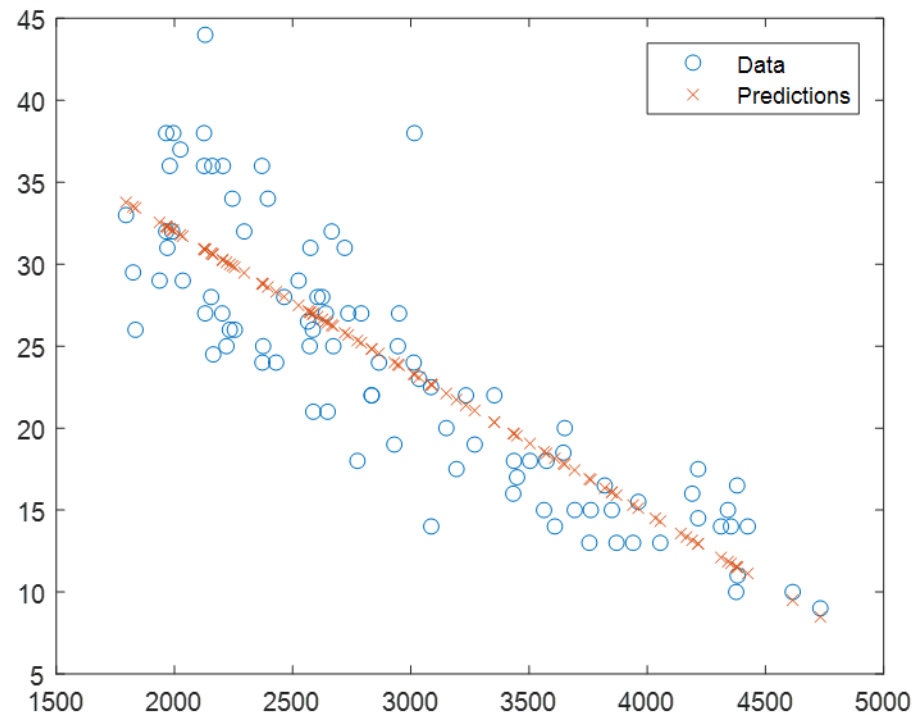
13×2 table

Heat	Predicted_heat
78.5	78.495
74.3	72.789
104.3	105.97
87.6	89.327
95.9	95.649
109.2	105.27
102.7	104.15
72.5	75.675
93.1	91.722
115.9	115.62
83.8	81.809
113.3	112.33
109.4	111.69

Example of Linear regression function (2): fitlm (cont.)

```
clear;clc;
rng(1);
Filename='regression1.xlsx';
Sheetread='Sheet1';
Input1='A1:A100';
output1='B1:B100';
Input=xlsread(Filename,Sheetread,Input1); %Read
Microsoft Excel
Target=xlsread(Filename,Sheetread,output1 );
x=Input;
t=Target;
t= fillmissing(t,'spline'); %fill in the missing output
data
```

```
mdl = fitlm(X,y,'linear');  
ypred = predict(mdl,X);  
plot(X,y,'o',X,ypred,'x')  
legend('Data','Predictions')
```



Nonlinear regression models/functions

Nonlinear regression model is a nonlinear correlation of a continuous dependent variables $Y(output)$ and independent variables $X(input)$.

- **Examples of Nonlinear functions**

ASN(X)	Arc sine of X
ATN (X)	Arc tangent of X
COS(X)	Cosine of X
EXP(X)	Exponential of X
INT(X)	Integer part of X
LN(X)	Log base e of X
LOG(X)	Log base 10 of X
SQR(X)	Square root of X
TAN(X)	Tangent of X

Examples of nonlinear models in Matlab

1. $Y = b(1) + b(2) * x(:,1).^b(3) + b(4) * x(:,2).^b(5)$
2. $Y = x(:,1) .* \exp(x(:,2))$
3. $Y = (b(1) * x^2 - x(:,3)/b(5))./(1 + b(2) * x(:,1) + b(3) * x(:,2) + b(4) * x(:,3));$
4. $Y = b(1) + b(2) * x(:,1) + b(3) * x(:,2) + b(4) * x(:,3)$
5. $Y = b(1) + b(2) * x(:,1) + b(3) * x(:,2) + b(4) * x(:,3) + b(5) * (x(:,1).^2) + b(6) * x(:,3) + b(7) * ((x(:,1) .* x(:,2) .* x(:,3))));$

Matlab function for non-linear polynomial models: polyfit

- **polyfit:** A Matlab solver for non-linear polynomial models.

Syntax:

```
p = polyfit(x,y,n)
```

Describe:

Returns the coefficients for a polynomial $p(x)$ of degree n that is a best fit for the data in y . x is the input, y is the output.

- **polyval**

Syntax:

```
ypredicted = polyval(p,x)
```

Describe:

To predict y based on x .

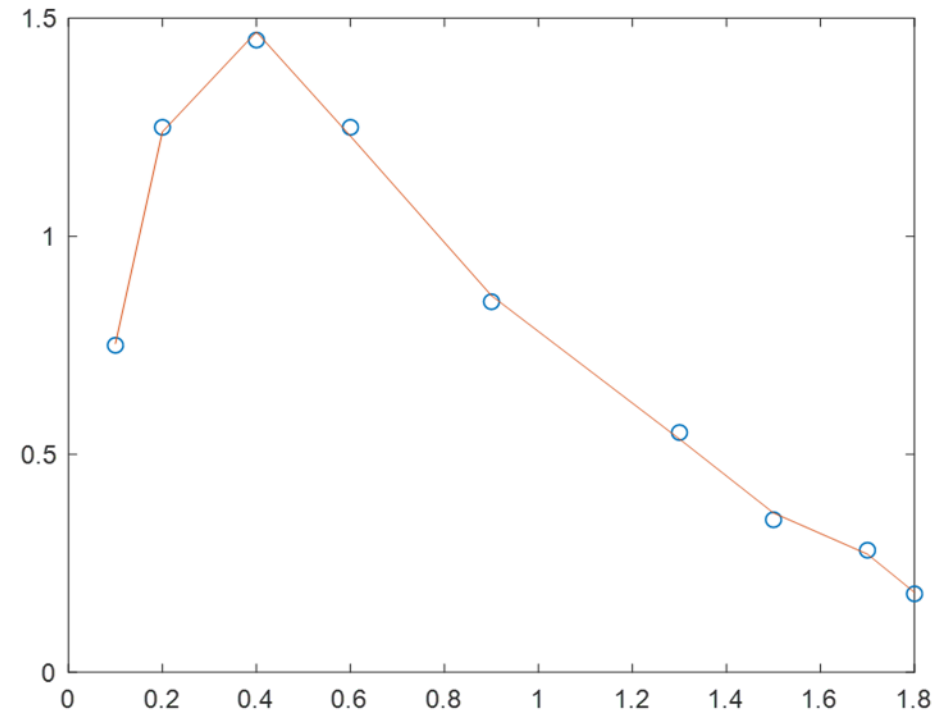
Example of 'polyfit' Matlab function 1:

```
clear ;clc;
x=[0.1 0.2 0.4 0.6 0.9 1.3 1.5 1.7 1.8]
y=[0.75 1.25 1.45 1.25 0.85 0.55 0.35 0.28 0.18]
p1=polyfit(x,y,7);% returns the coefficients for a polynomial
p(x) of degree 7
y2=polyval(p1,x);
plot(x,y,'o',x,y2,'-');
table( y( : ),y2( : ), 'VariableNames',...
    {' Original_y',' Predicted_y'}) %Show the results of data
of the output and predicted output.
```

Example of 'polyfit' Matlab function (cont.) 2:

9×2 [table](#)

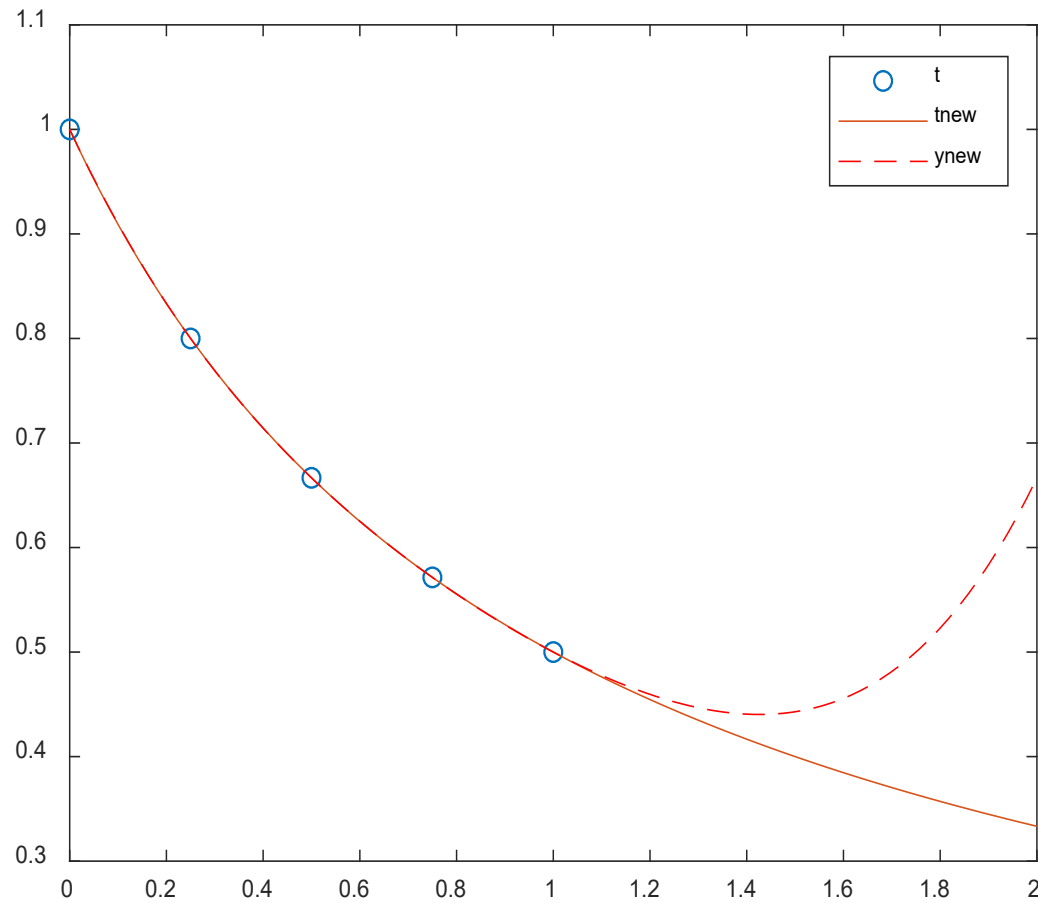
Original_y	Predicted_y
0.75	0.75368
1.25	1.2395
1.45	1.4686
1.25	1.2298
0.85	0.86389
0.55	0.53485
0.35	0.36554
0.28	0.27082
0.18	0.18327



Example of 'polyfit' Matlab function 2 (cont.):

```
clear;clc;
x = linspace(0,1,5); % input
t = 1./(1+x); % output
p = polyfit(x,t,4); %gives the coefficient
xnew = linspace(0,2); % new x (generate linearly)
tnew = 1./(1+xnew); % new y
ynew = polyval(p,xnew); % polynomial evaluation
figure
plot(x,t,'o')
hold on
plot(xnew,tnew)
plot(xnew,ynew,'r--')
legend('t','tnew','ynew')
```

Example of 'polyfit' Matlab function 2 (cont.):



Example of 'polyfit' Matlab function 3:

Data =

0	0.562
0	0.58
0	0.549
25	0.572
25	0.6
25	0.572
57	0.744
57	0.749
57	0.742
90	0.776
90	0.789
90	0.824
115	0.959
115	0.993
115	0.949
136	1.248
136	1.282
136	1.304
165	1.492
165	1.468
165	1.49
186	1.854
186	1.804

Example of 'polyfit' Matlab function 3 (cont.):

```
clear;clc;

figure(1);

plot(data(:,1),data(:,2),'ks');grid on;% plotting
with black squares

x=data(:,1);

y=data(:,2);

m1=polyfit(x,y,1); %build a first order polynomial
model ,m1 are the coefficients

p1=polyval(m1,x);

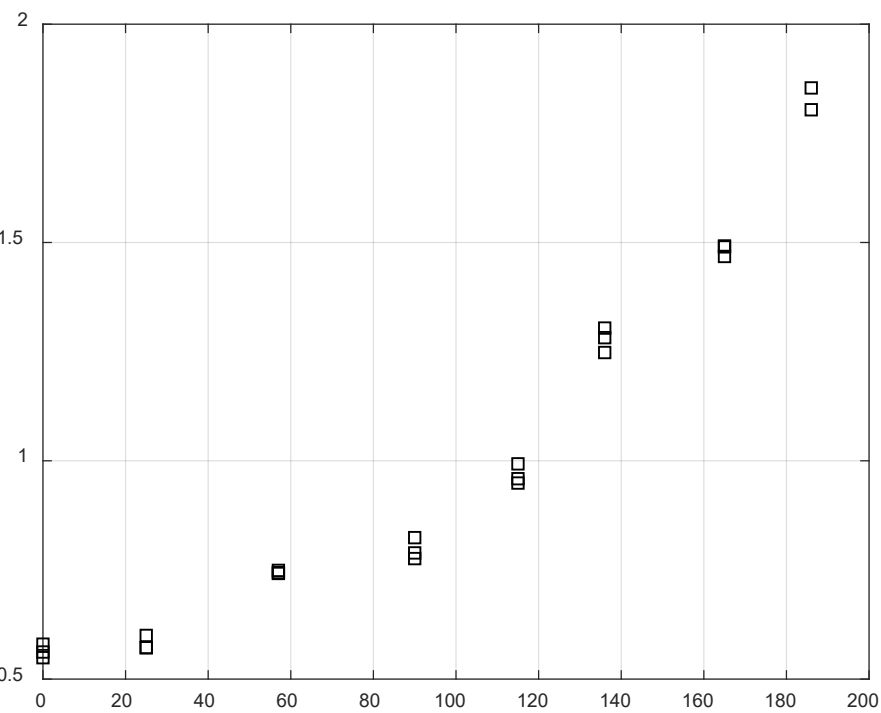
m2=polyfit(x,y,2);%build a second order polynomial
model ,m2 are the coefficients

p2=polyval(m2,x);

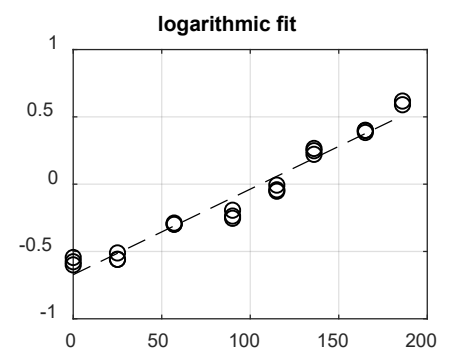
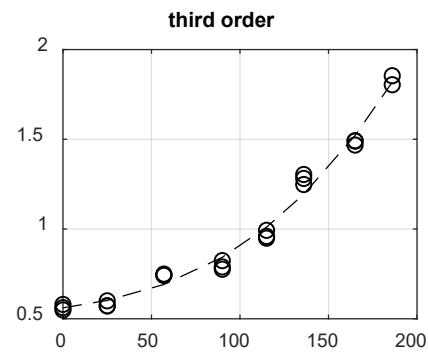
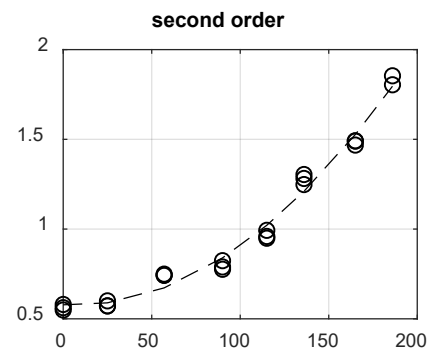
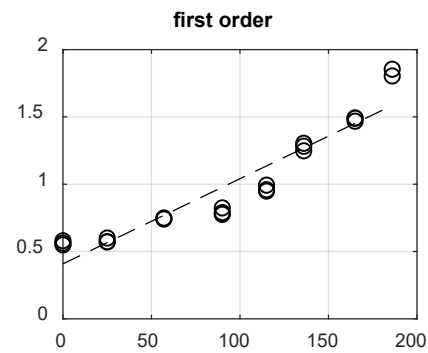
m3=polyfit(x,y,3);%build a second order polynomial
model ,m3 are the coefficients
```

Example of 'polyfit' Matlab function 3 (cont.):

```
p3=polyval(m3,x);  
m4=polyfit(x,log(y),1);%build a logarithmic model of order 1  
%m4 are the coefficients  
p4=polyval(m4,x);  
figure(2);  
subplot(2,2,1),plot(x,y,'ko'),hold on,plot(x,p1,'k--');grid  
on,title('first order');  
subplot(2,2,2),plot(x,y,'ko'),hold on,plot(x,p2,'k--');grid  
on,title('second order');  
subplot(2,2,3),plot(x,y,'ko'),hold on,plot(x,p3,'k--');grid  
on,title('third order');  
subplot(2,2,4),plot(x,log(y),'ko'),hold on,plot(x,p4,'k--  
';grid on,title('logarithmic fit');
```



Original data



Original data and polynomial fit

Nonlinear functions solver: lsqcurvefit

- **lsqcurvefit**: A Matlab solver for non-linear models.

Syntax:

```
m = lsqcurvefit(modelfun, beta0,x,t)
```

Description:

Solve non-linear curve-fitting (data-fitting) problems in least-squares sense.

fun : Function you want to fit.

beta0 — Coefficients

x: Input data for model

t : Out put for model

- **Prediction**

Syntax:

```
Ypredicted=modelfun(m,x);
```

Example of Non-linear regression function (1) : lsqcurvefit

```
clear;

clc;

rng(1);

x = linspace(0,3);

t = exp(-1.3*x) + 0.05*randn(size(x));

fun = @(b,x)b(1)*exp(b(2)*x); % Defining the Model
function

b0 = [1/2,-2]; %Coefficient initiation

mdl = lsqcurvefit(fun,b0,x,t); % obtaining the
coefficients.

ypredicted=fun(mdl,x);

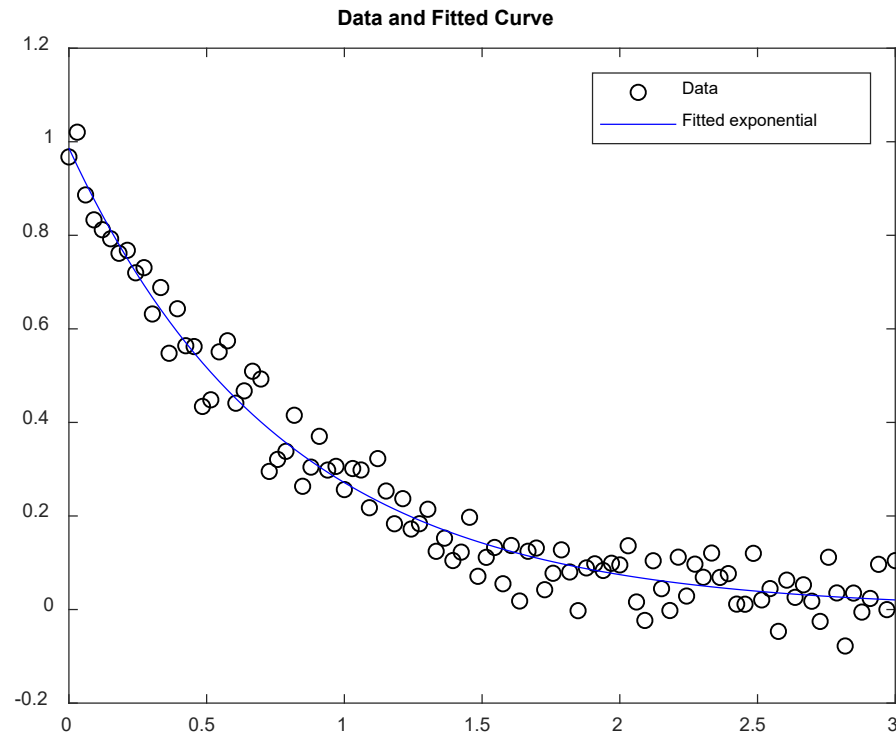
plot(x,t,'ko',x,ypredicted,'b-')

legend('Data','Fitted exponential')

title('Data and Fitted Curve')
```

Example of Non-linear regression function (1) : lsqcurvefit (cont.)

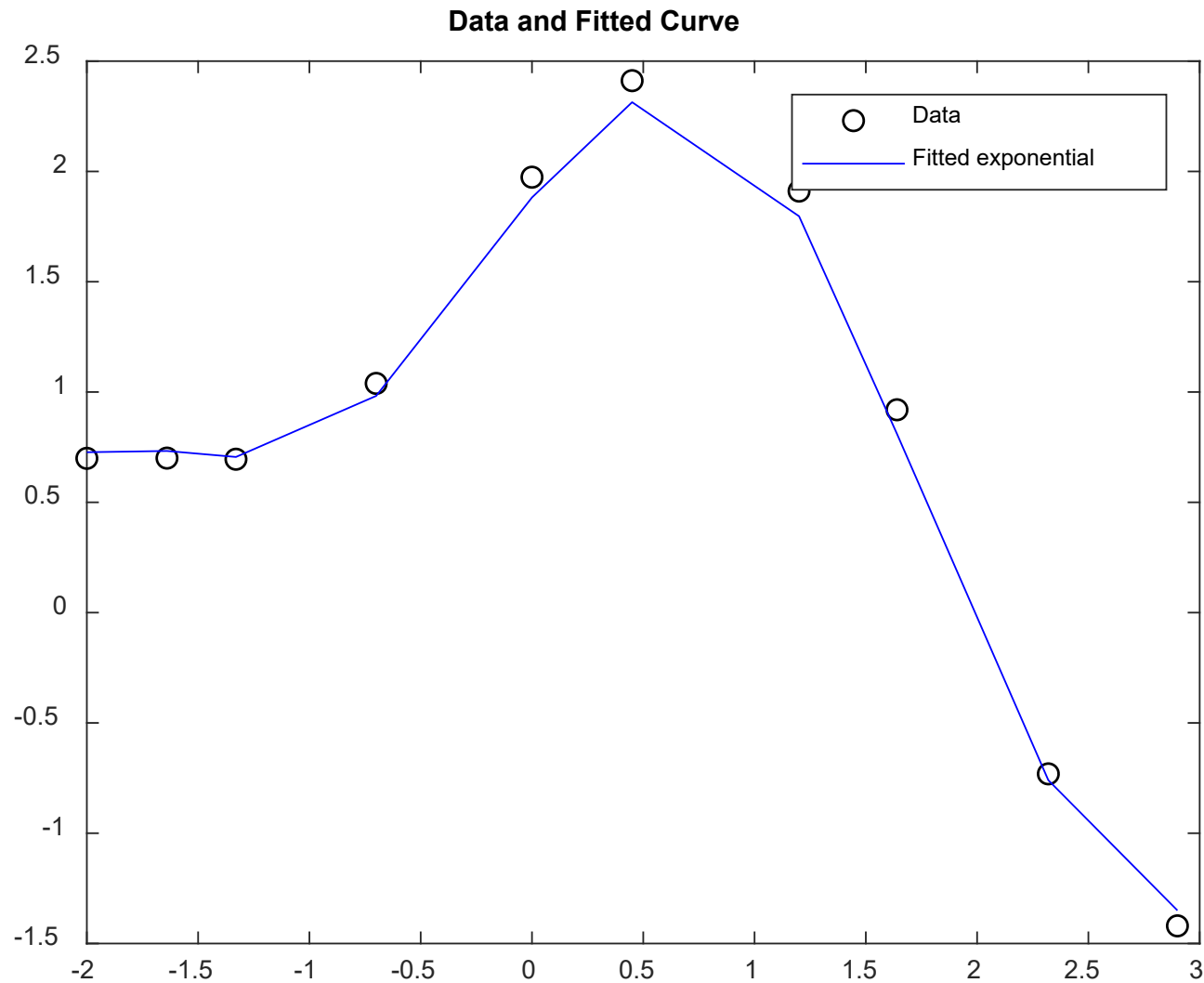
mdl		
1x2 double		
	1	2
1	0.9863	-1.2909
2		
3		
4		
5		
6		



Example of Non-linear regression function (2) : lsqcurvefit

```
clear ;  
clc;  
rng(1) ;  
x = [-2,-1.64,-1.33,-0.7,0,0.45,1.2,1.64,2.32,2.9] ;  
t =  
[0.699369,0.700462,0.695354,1.03905,1.97389,2.41143,  
1.91091,0.919576,-0.730975,-1.42001] ;  
fun = @(b,x) b(1)*cos(b(2)*x)+b(2)*sin(b(1)*x) ;  
b = [1,0.2] ;  
mdl = lsqcurvefit(fun,b,x,t) ; % mdl will be the  
coefficients  
ypredicted=fun(mdl,x) ; %This equals to yexpected  
=mdl(:,1)*cos(mdl(:,2)*x)+mdl(:,2)*sin(mdl(:,1)*x)  
plot(x,t,'ko',x,ypredicted,'b-')
```

Example of Non-linear regression function (2) : lsqcurvefit (cont.)



Example of Non-linear regression function (2) : lsqcurvefit (cont.)

```
legend('Data','Fitted exponential')  
title('Data and Fitted Curve')  
table( t(:),ypredicted(:), 'VariableNames',...  
       {' Original_y',' Predicted_y'})
```

10×2 table

Original_y	Predicted_y
0.69937	0.72706
0.70046	0.73275
0.69535	0.7056
1.0391	0.98242
1.9739	1.8818
2.4114	2.3138
1.9109	1.7969
0.91958	0.81025
-0.73098	-0.75969
-1.42	-1.35

Non-linear regression Matlab function (fitnlm):

- **fitnlm:** A non-linear regression Matlab solver for non-linear models.

Syntax:

```
mdl = fitnlm(x,y,modelfun,beta0)
```

x: Predictor variables(Input)

y: Response variable(output)

Modelfun: Functional form of the model (model you want to fit)

Beta0: Coefficients

- **Predict:** Predict response of the regression model

Syntax:

```
ypred = predict(mdl,Xnew);
```

Example of Nonlinear regression function 1: fitnlm

```
rng(1);  
Filename='regression2.xlsx';  
Sheetread='Sheet1';  
Input1='A1:B406';  
output1='C1:C406';  
Input=xlsread(Filename,Sheetread,Input1); %Read  
Microsoft Excel  
Target=xlsread(Filename,Sheetread,output1 );  
x=Input;  
t=Target;  
modelfun = @(b,x) (b(1) + b(2)*x(:,1).^b(3) + ...  
b(4)*x(:,2).^b(5)); % nonlinear model
```

Example of Nonlinear regression function 1: fitnlm (cont.)

```
beta0 = [-50 500 -1 500 -1]; % coefficients  
initiation  
  
mdl = fitnlm(x,t,modelfun,beta0);  
  
y_expected = predict(mdl,x);  
  
table( t (10:20 ), y_expected( 10:20 ),  
    'VariableNames', ...  
        { ' TrueLabel', ' PredictedLabel' }) %Show the  
results of 1st  to 10th data of the output and  
predicted output.  
  
MSE_training= (mean((t - y_expected).^2));  
  
RMSE_training = sqrt(mean((t - y_expected).^2));
```

ans =

11×2 table

TrueLabel	PredictedLabel
15	14.533
16.027	20.696
16.702	14.267
16.982	15.054
16.821	13.87
16.175	14.962
15	16.225
14	16.373
15.241	18.223
15	16.149
14	16.956

mdl.Coefficients	
	1 Estimate
1 b1	-31.0524
2 b2	149.8782
3 b3	-0.5187
4 b4	921.5878
5 b5	-0.3937

MSE_training =16.1174

MSE_testing =4.0147

Example of Nonlinear regression function 2: fitnlm

```
clear;clc;

rng(1);

Filename='regression3.xlsx';

Sheetread='Sheet1';

Input1='A1:C13';

output1='D1:D13';

Input=xlsread(Filename,Sheetread,Input1); %Read Microsoft
Excel

Target=xlsread(Filename,Sheetread,output1 );

x=Input;

t=Target;

beta = [1 1 1 1 1]; % coefficient initiation

fun = @(b,x) ((b(1)*x(:,2) -
x(:,3)/b(5))./(1+b(2)*x(:,1)+b(3)*x(:,2)+b(4)*x(:,3))));

mdl = fitnlm(x,t,fun,beta)
```

Example of Nonlinear regression function 2: fitnlm (cont.)

```
y_expected = predict(mdl,x);  
table( t( 5:10 ),y_expected( 5:10 ), 'VariableNames',...  
      {' Actual_Y',' PredictedY'}) %Show the results of 5th  
to 10th data of the output and predicted output.  
MSE_training= (mean((t - y_expected).^2));  
RMSE_training = sqrt(mean((t - y_expected).^2));
```

mdl.Coefficients		
	1	
	Estimate	
1 b1	1.2526	
2 b2	0.0628	
3 b3	0.0400	
4 b4	0.1124	
5 b5	1.1914	

MSE_training =0.0230

MSE_testing =0.1516

Example of Nonlinear regression function 2: fitnlm (cont.)

Actual_Y	PredictedY
8.55	8.4179
3.79	3.9542
4.82	4.9109
0.02	-0.010952
2.75	2.6358
14.39	14.34
2.54	2.5662
4.35	4.0385
13	13.029
8.5	8.3904
0.05	-0.021563
11.32	11.47
3.13	3.4326

Example of Nonlinear regression function 3: fitnlm

```
Clear;clc;

rng(1);

Filename='regression4.xlsx';

Sheetread='Sheet1';

Input1='A1:C72';

output1='D1:D72';

Input=xlsread(Filename,Sheetread,Input1); %Read Microsoft
Excel

Target=xlsread(Filename,Sheetread,output1 );

x=Input;

t=Target;

[xn,sxn] = mapminmax(x');% Standardize x

[tn,stn]= mapminmax(t');% Standardize t

Sheetread1='Sheet2';

Input2='A1:C3';
```

Example of Nonlinear regression function 3: fitnlm (cont.)

```
Target2 = 'D1:D3';  
Inputnew=xlsread(Filename,Sheetread1,Input2);  
Targetnew=xlsread(Filename,Sheetread1,Target2 );  
xnew=Inputnew;  
tnew=Targetnew;  
  
xnewn = mapminmax('apply',xnew',sxn); % The same Process  
setting of  
  
%standardization for x should also be applied for xnew.  
  
%xnewn is the  
  
%standardized xnew  
  
xn=xn';% standardized x  
  
tn=tn';% standardized t  
  
xnewn=xnewn';%standardized xnew  
  
beta = [1 1 1 1 1 1]; % coefficient initiation
```

Example of Nonlinear regression function 3: fitnlm (cont.)

```
fun=@(b,xn)b(1)+b(2)*xn(:,1)+b(3)*xn(:,2)+b(4)*xn(:,3)+b(5)  
)*(xn(:,1).^2)+b(6)*((xn(:,1).*xn(:,2).*xn(:,3)));%
```

nonlinear model with standardized x

```
mdl = fitnlm(xn,tn,fun,beta);% find coefficients(beta) of  
model(fun) using normalized x and t
```

```
yfitn = predict(mdl,xn);% make prediction based on  
normalized x
```

```
yfit = mapminmax('reverse', yfitn,stn); % To reverse the  
prediction to original state using the same process  
setting of t
```

```
table( t( 10:20 ),yfit( 10:20 ), 'VariableNames',...
```

```
    {' TrueLabel',' PredictedLabel'}) %Show the results of  
5th  to 10th data of the output and predicted output.
```

```
MSE_training=sum((yfit-t).^2)/numel(t); % Calculate MSE  
for data
```

```
RMSE_training=sqrt(sum((yfit-t).^2)/numel(t)); % Calculate  
RMSE for data
```

Example of Nonlinear regression function 3: fitnlm (cont.)

```
ynewn=predict(mdl,xnewn);% make prediction based on  
normalized new data  
  
ynew = mapminmax('reverse', ynewn,stn); % To reverse the  
normalized ynew and use the processing setting of t  
  
table(tnew(:),ynew(:),'VariableNames',{'ObservedValue_Newd  
ata',' PredictedValue_newdata'}) % show data in output and  
predicted output  
  
% MSE_testing1=mse(tnew,ynew);  
  
MSE_testing=sum((tnew-ynew).^2)/numel(tnew); % Calculate  
MSE for new data  
  
RMSE_testing=sqrt(sum((tnew-ynew).^2)/numel(tnew)); %  
Calculate RMSE for new data  
  
Errorpercentage=((ynew-tnew)./tnew)*100; % Calculate error  
percentage for tnew and ynew
```

```
ans =
```

```
11×2 table
```

TrueLabel	PredictedLabel
518	513.76
512	512.98
509	512.7
513	512.4
508	515.75
513	512.32
507	514.13
512	511.24
517	515.37
514	514.31
514	512.81

MSE_training =11.0469

MSE_testing =3.98557

ans =

3x2 [table](#)

ObservedValue_Newdata	PredictedValue_newdata
495	497.12
498	499.45
498	500.31

RMSE_training =3.3237

RMSE_testing =1.9964

Errorpercentage			xn
3x1 double			
	1	2	
1	0.4290		
2	0.2911		
3	0.4643		
4			

Examples of non-linear model functions (Using combination of Power , Exponential, polynomial functions) :

1. `fun=@(b,x) b(1)+b(2)*x(:,1)+b(3)*x(:,2)+b(4)*x(:,3)+b(5)*(x(:,1).^2)+b(6)*((x(:,1).*x(:,2).*x(:,3))) ;`
2. `fun=@(b,x) b(1)+b(2)*x(:,1)+b(3)*x(:,2)+b(4)*x(:,3)+b(5)*(x(:,1).^2) ;`
3. `fun=@(b,x) b(1)+b(2)*x(:,1)+b(3)*x(:,2)+b(4)*x(:,3)+b(5)*(x(:,1).^2)+b(6)*x(:,2).^2 ;`
4. `fun=@(b,x) b(1)+b(2)*x(:,1)+b(3)*x(:,2)+b(4)*x(:,3)+b(5)*(x(:,1).^2)+b(6)*exp(x(:,1)) ;`
5. `fun=@(b,x) b(1)+b(2)*x(:,1)+b(3)*x(:,2)+b(4)*x(:,3)+b(5)*(x(:,1).^2)+b(6)*x(:,3) +b(7)*((x(:,1).*x(:,2).*x(:,3))) ;`
6. `fun=@(b,xn) b(1)+b(2)*xn(:,1)+b(3)*xn(:,2)+b(4)*xn(:,3)+b(5)*(xn(:,1).^2)+b(6)*((xn(:,1).*xn(:,2).*xn(:,3)))+b(7).*exp(b(8)*xn(:,3))`

References for regression functions and methods:

- <https://au.mathworks.com>
- S. Araghinejad, Data-Driven Modeling: Using MATLAB® in Environmental Engineering