

# TU: Linear Regression

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## Introduction

Fit Linear Regression

## Include path

```
addpath('../..../Library')
```

## Examples

### Example 1: Fit with Linear Regression

#### Data Acquisition

- Feature: One-Dimension,  $p=1$
- True value:  $y_{\text{true}} = 2X+3$

```
x = randn(100,1);  
y = x*2 + 3+randn(100,1);
```

#### Fit Linear Regression

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

```
mdl =  
선형 회귀 모델:  
y ~ 1 + x1
```

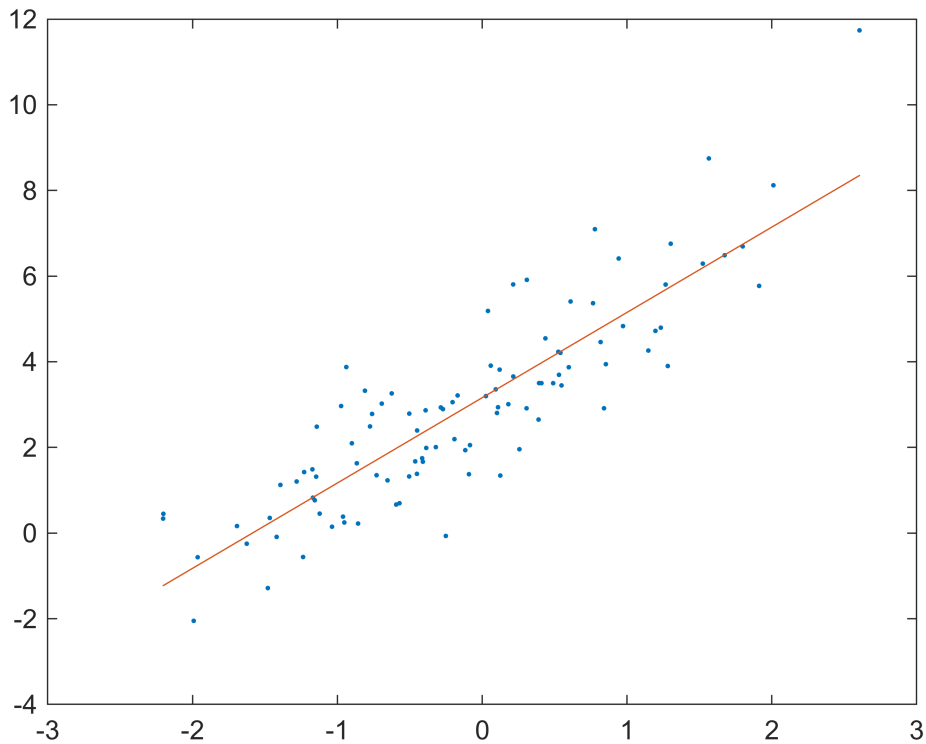
추정된 계수:

	Estimate	SE	tStat	pValue
(Intercept)	3.1607	0.11468	27.562	5.871e-48
x1	1.9916	0.11442	17.406	9.6192e-32

관측값 개수: 100, 오차 자유도: 98  
RMS 오차: 1.14

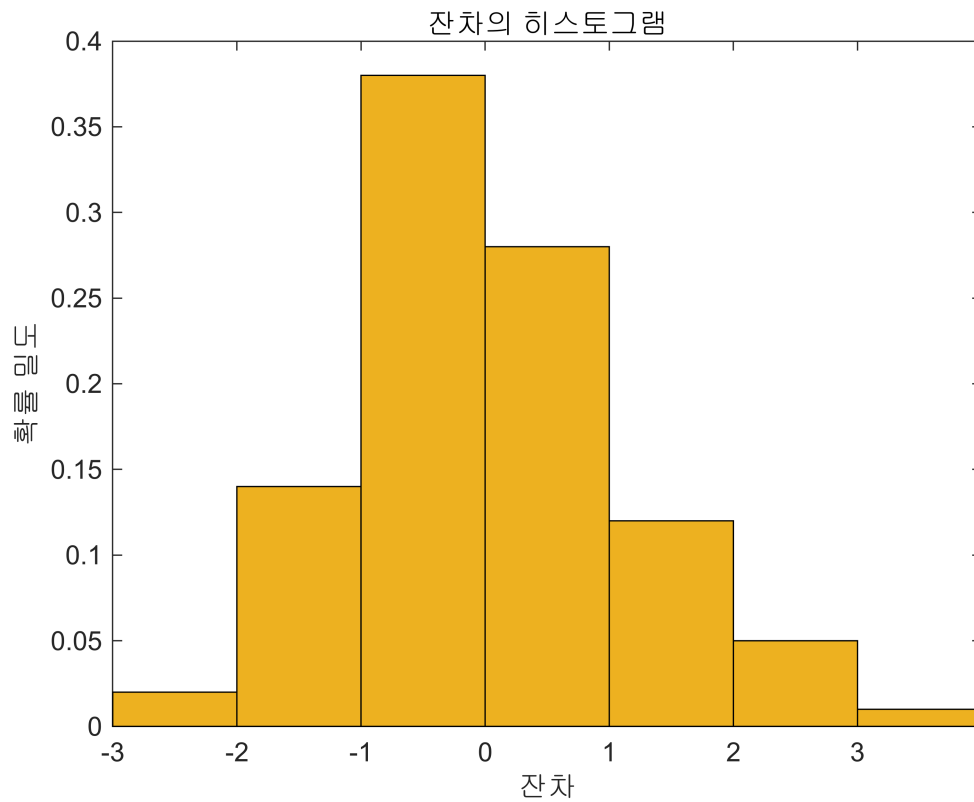
결정계수: 0.756, 수정된 결정계수: 0.753  
상수 모델에 대한 F-통계량: 303, p-값 = 9.62e-32

```
% Plot prediction  
plot(x,y,'.',x,mdl.Fitted, '-')
```



## Analyze Linear Regression

```
% Plot residual histogram  
plotResiduals(mdl)
```



```
% From remove outlier from histogram
outl = find(abs mdl.Residuals.Raw) > 2);
mdl.Residuals.Raw(outl)
```

```
ans = 8×1
    2.2185
    3.3878
    2.4698
   -2.7272
    2.3842
    2.1376
    2.5840
   -2.0667
```

```
% Fit with outlier removed
mdl2 = fitlm(x,y,'Exclude',outl)
```

```
mdl2 =
선형 회귀 모델:
y ~ 1 + x1
```

추정된 계수:

	Estimate	SE	tStat	pValue
(Intercept)	3.0119	0.094688	31.809	9.7326e-51
x1	1.8087	0.096003	18.84	5.4685e-33

관측값 개수: 92, 오차 자유도: 90  
 RMS 오차: 0.89  
 결정계수: 0.798, 수정된 결정계수: 0.795  
 상수 모델에 대한 F-통계량: 355, p-값 = 5.47e-33

## Predict with Test data

```
% Ypredict and confidence interval
```

```
Xnew=2;
```

```
[Ynew,YnewI]=predict(md12,Xnew)
```

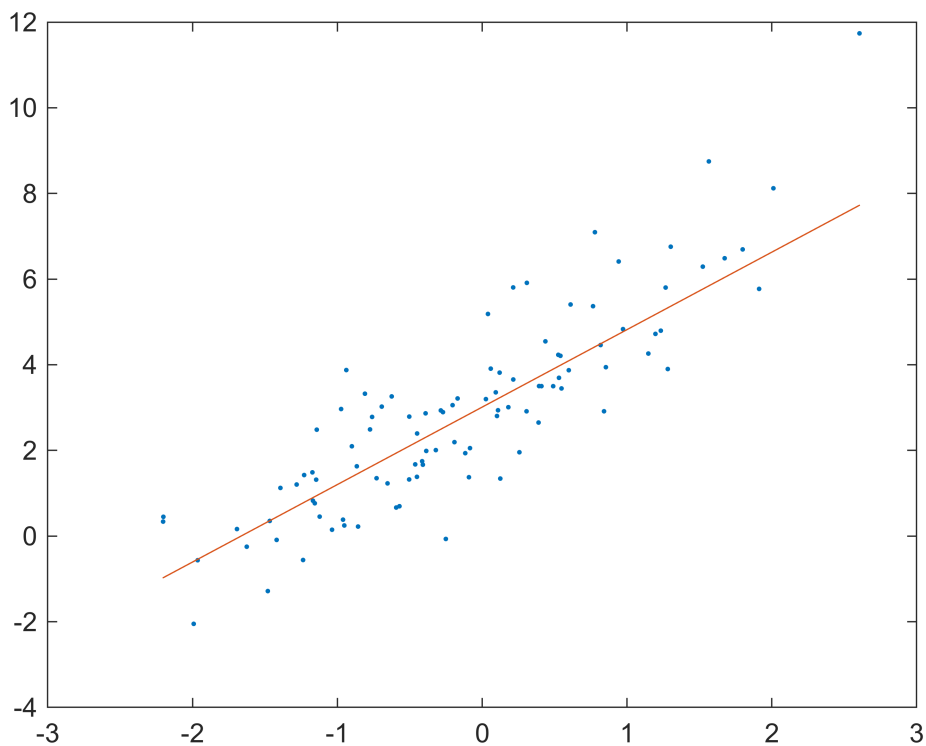
```
Ynew = 6.6294
```

```
YnewI = 1x2
```

```
6.1717 7.0870
```

```
% Plot prediction
```

```
plot(x,y,'.',x,md12.Fitted, '-')
```



## Example 2: Predict Car MPG

Find linear relationship of MPG(연비) with Weight & Displacement

Then, Predict MPG for a car with Weight=3000 , Displacement=300

## Data Acquisition

- Dependent Variable: MPG
- Independent Variables: Weight, Displacement

```
clear
load carsmall

tbl = table(MPG,Weight, Displacement); % table type
```

## Fit Linear Regression

```
mdl = fitlm(tbl,'MPG~Weight+Displacement')
```

mdl =  
선형 회귀 모델:  
MPG ~ 1 + Weight + Displacement

추정된 계수:

	Estimate	SE	tStat	pValue
(Intercept)	46.925	2.0858	22.497	6.0509e-39
Weight	-0.0068422	0.0011337	-6.0353	3.3838e-08
Displacement	-0.014593	0.0082695	-1.7647	0.080968

관측값 개수: 94, 오차 자유도: 91  
RMS 오차: 4.09  
결정계수: 0.747, 수정된 결정계수: 0.741  
상수 모델에 대한 F-통계량: 134, p-값 = 7.22e-28

```
mdl.CoefficientNames
```

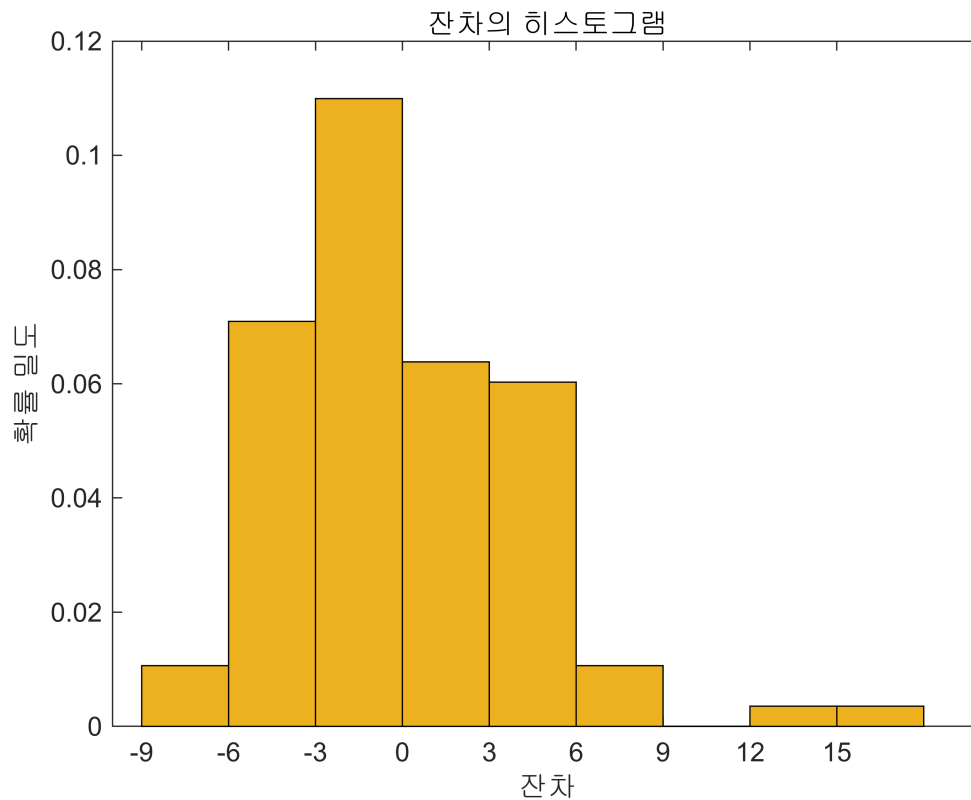
```
ans = 1x3 cell
'(Intercept)' 'Weight' 'Displacement'
```

```
mdl.Coefficients.Estimate
```

```
ans = 3x1
46.9247
-0.0068
-0.0146
```

## Analyze Linear Regression

```
plotResiduals(mdl)
```



```
% Remove outlier
out1 = find((mdl.Residuals.Raw) > 9);

% Fit with outlier removed
mdl2 = fitlm(tbl, 'MPG~Weight+Displacement', 'Exclude', out1)
```

```
mdl2 =
선형 회귀 모델:
MPG ~ 1 + Weight + Displacement
```

추정된 계수:

	Estimate	SE	tStat	pValue
(Intercept)	45.548	1.8056	25.226	2.5559e-42
Weight	-0.0062503	0.00097845	-6.3879	7.4767e-09
Displacement	-0.018035	0.0071414	-2.5255	0.013324

관측값 개수: 92, 오차 자유도: 89  
 RMS 오차: 3.51  
 결정계수: 0.796, 수정된 결정계수: 0.792  
 상수 모델에 대한 F-통계량: 174, p-값 = 1.78e-31

```
mdl2.Coefficients.Estimate
```

```
ans = 3x1
45.5477
-0.0063
-0.0180
```

## Predict

Detailed look at the interactions

```
% Ypredict and confidence interval
```

```
Xnew=[3000 300];
```

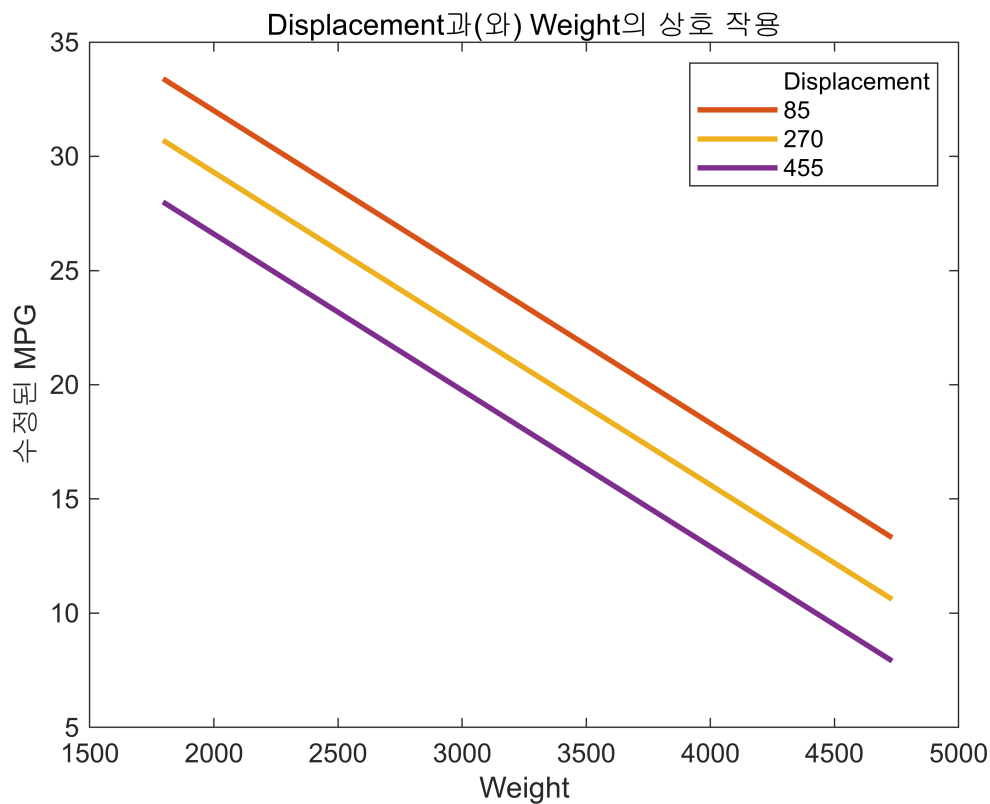
```
[Ynew,YnewI]=predict(md12,Xnew)
```

```
Ynew = 21.3863
```

```
YnewI = 1×2
```

```
19.8515 22.9211
```

```
plotInteraction(md1,'Displacement','Weight','predictions')
```



## Exercise

### Exercise 1 : Gradient Descent

Linear Regression Using Gradient Descent

Hypothesis.

$$h_{\theta}(x) = \theta_1 x + \theta_0$$

To find the parameters, repeat until convergence

$$\theta_k = \theta_k - \alpha \frac{\partial}{\partial \theta_k} J(\theta_1, \theta_0)$$

where, cost(error) function

$$J(\theta_1, \theta_0) = \frac{1}{2n} E = \frac{1}{2n} \sum_{i=1}^n (y_i - h(x_i))^2$$

and

$$\frac{\partial J}{\partial \theta_1} = -\frac{1}{n} \sum_{i=1}^n x_i (y_i - (\theta_1 x_i + \theta_0))$$

$$\frac{\partial J}{\partial \theta_0} = -\frac{1}{n} \sum_{i=1}^n (y_i - (\theta_1 x_i + \theta_0))$$

## Data Acquisition

Feature: One-Dimension,  $p=1$

True:  $y = 2X+3$

```
N=100;  
X = randn(N,1);  
Y = X*2 + 3+randn(N,1);
```

## Fit Linear Regression: Gradient Descent

```
% dJ/dx  
lamda=0.1; % learning rate  
  
% Initialization for t0, t1  
t0=0.5;  
t1=0.5;  
loss=1;  
itrN=1000;  
k=1;
```



```

t1_prev=0;
t0_prev=0;

%% ADD your code here
while(loss>0.0001 && k<itrN)

    % Define function h(x)
    h=t1*X+t0;

    % Define gradient w.r.t theta_1 and theta_0
    dJt1 = -sum(X.*(Y-(t1*X+t0))) / N;
    dJt0 = -sum(Y-(t1*X+t0)) / N;

    % Update theta1, theta0
    t1=t1-lamda*(dJt1 + dJt0);
    t0=t0-lamda*(dJt1 + dJt0);

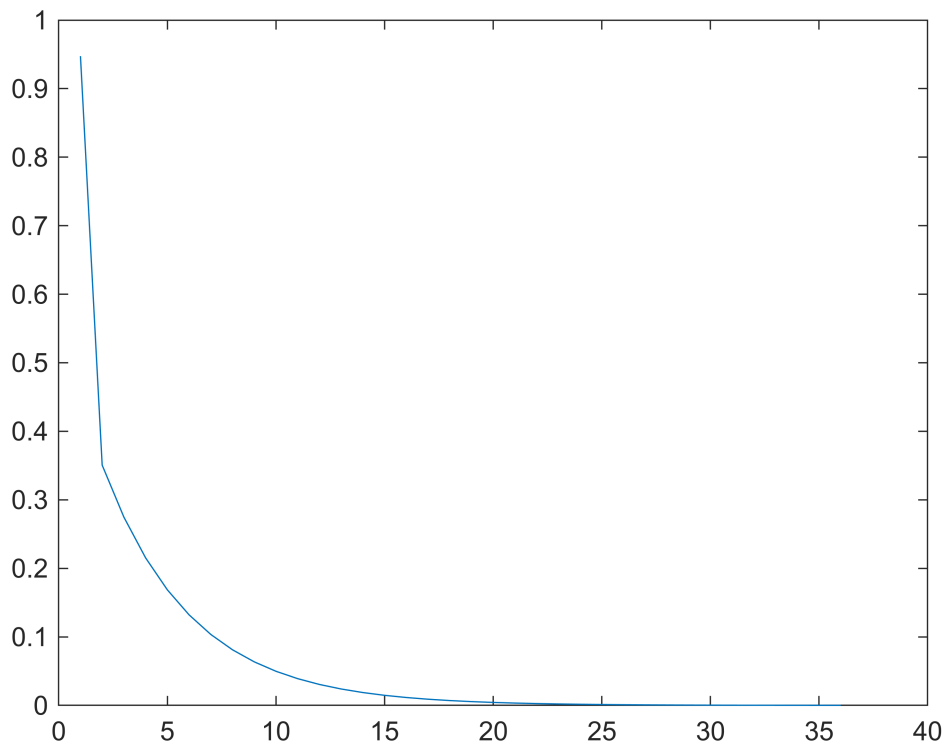
    % Cost Function v1
    loss=sum((Y-h).^2)/(2*N);

    % Cost Function v2
    loss= (norm(t1-t1_prev)+norm(t0-t0_prev))/2;
    t1_prev=t1;
    t0_prev=t0;

    loss_hist(k) = loss;
    k=k+1;
end

% Plot loss vs iteration
figure
plot(loss_hist)

```



```
% Predicted hypothesis y(x)
ypred=t1_prev*X + t0_prev; %%% ADD your code here

disp('optimal paraterms are')
```

optimal paraterms are

t1

t1 = 2.5670

t0

t0 = 2.5670

```
% MATLAB fit linear using fitlm(x,y)
mdl = fitlm(X, Y) %%% ADD your code here
```

mdl =  
선형 회귀 모델:  
 $y \sim 1 + x1$

추정된 계수:

	Estimate	SE	tStat	pValue
(Intercept)	3.0111	0.10578	28.466	3.5158e-49

x1                      2.0901            0.11019            18.968            1.4282e-34

관측값 개수: 100, 오차 자유도: 98

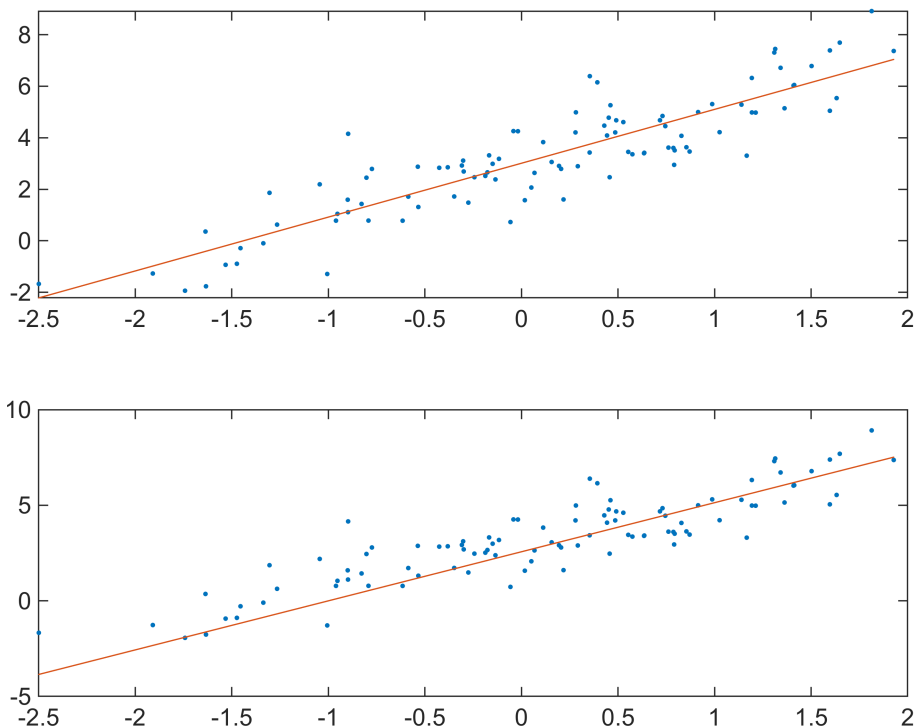
RMS 오차: 1.05

결정계수: 0.786, 수정된 결정계수: 0.784

상수 모델에 대한 F-통계량: 360, p-값 = 1.43e-34

**% Plot and compare predictions**

```
figure
subplot(2,1,1)
plot(X,Y,'.',X,mdl.Fitted, '-')
subplot(2,1,2)
plot(X,Y,'.',X,ypred, '-')
```



## Exercise 2 : Linear Regression with dim=2

Find the linear regression. Remove Outlier and predict a test value  $X_{test}=[2;1]$ . You can use MATLAB `fitlm()`

- Feature: 2-Dimensions,  $p=2$
- True value:  $y = 2 \cdot X_1 + 4 \cdot X_2 + 3$

### Data Acquisition

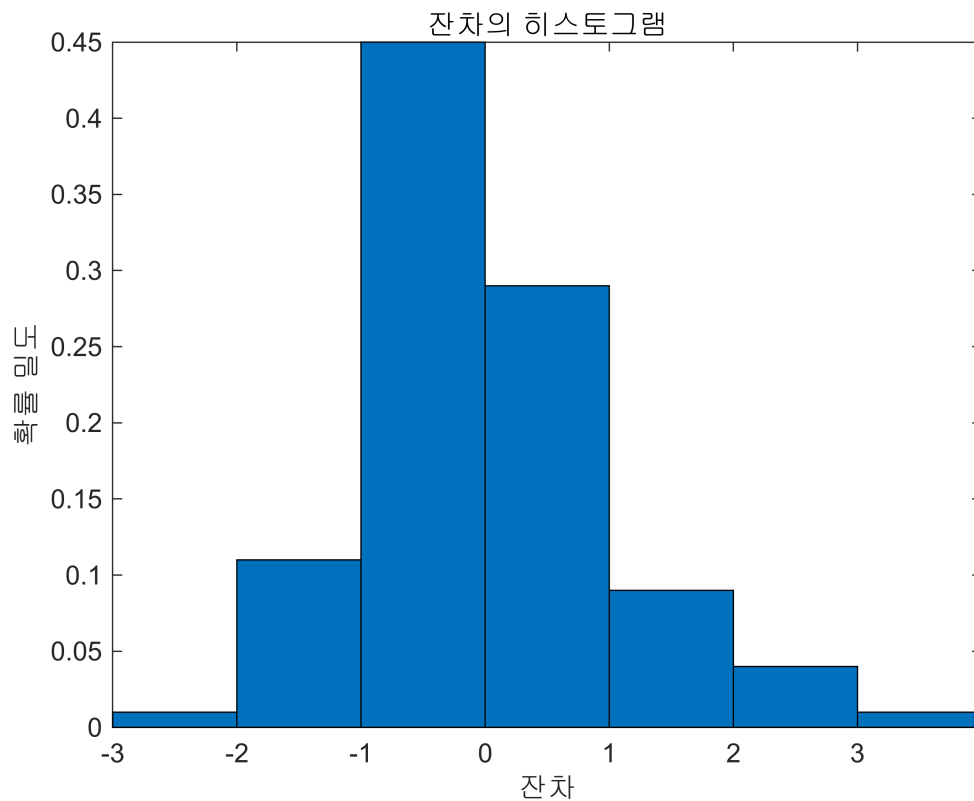
```
X = randn(100,2);
y = X*[2;4] + 3+ randn(100,1);
```

### Fit Linear Regression

```
mdl=fitlm(X,y);
% Your code goes here
```

## Analyze

```
% Remove outlier
% -plot residual histogram
figure
plotResiduals(mdl)
```



```
% - remove outlier from histogram analysis
out1 = find(mdl.Residuals.Raw > 2);
mdl.Residuals.Raw(out1)
```

```
ans = 5x1
2.4975
2.2773
2.4835
3.1499
2.9978
```

```
% Fit linear regression (fitlm)
mdl2 = fitlm(X, y, 'Exclude', out1);
```

```
% Your code goes here
```

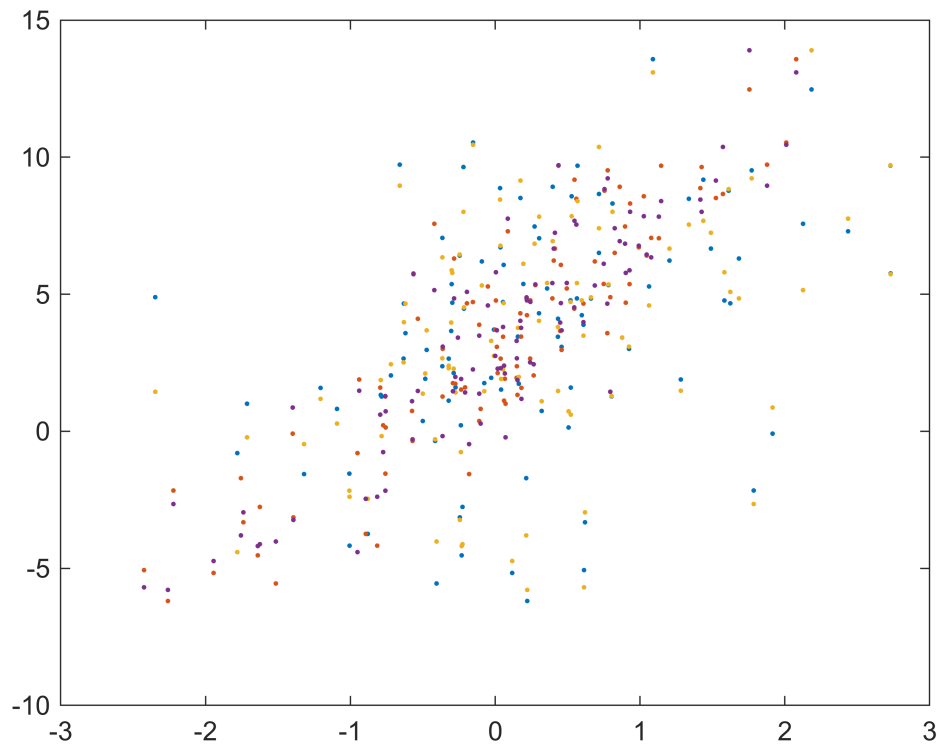
## Predict

Predict for a Test value

```
Xnew=[2, 1];  
Ynew=predict mdl2,Xnew)
```

```
Ynew = 10.5491
```

```
plot(X,y,'.', X,mdl2.Fitted,'.')
```



## Exercise 3: Linear Regression with dim=4

Find linear relationship of

- $MPG \sim 1 + \text{Acceleration} + \text{Displacement} + \text{Horsepower} + \text{Weight}$

Choose a test data and predict

## Data Acquisition

```
clear  
load carsmall
```

```
tbl = table(MPG,Acceleration,Displacement,Horsepower,Weight);
```

## Fit Linear Regression

```
mdl = fitlm(tbl, 'MPG~Acceleration+Displacement+Horsepower+Weight')
```

mdl =

선형 회귀 모델:

MPG ~ 1 + Acceleration + Displacement + Horsepower + Weight

추정된 계수:

	Estimate	SE	tStat	pValue
(Intercept)	48.117	3.9008	12.335	6.9194e-21
Acceleration	-0.060312	0.21167	-0.28493	0.77636
Displacement	-0.0066826	0.011594	-0.57638	0.56583
Horsepower	-0.037547	0.026139	-1.4364	0.15442
Weight	-0.006084	0.0013823	-4.4014	3.01e-05

관측값 개수: 93, 오차 자유도: 88

RMS 오차: 4.11

결정계수: 0.753, 수정된 결정계수: 0.742

상수 모델에 대한 F-통계량: 67.1, p-값 = 6.49e-26

## Analyze

```
% Remove outlier
% -plot residual histogram
% - remove outlier from histogram analysis
outl = find(abs(mdl.Residuals.Raw) > 9);

% Fit linear regression (fitlm)
mdl2 = fitlm(tbl, 'MPG~Acceleration+Displacement+Horsepower+Weight', 'Exclude', outl);
```

## Predict

Predict for a Test value

```
Xnew = [10 300 150 3500];
Ynew= predict(mdl2, Xnew);
plotInteraction(mdl, 'Acceleration', 'Displacement', 'predictions')
```

