Assignment 1 (Part-A)

Machine Learning (SE-807)



Submitted By:-

Shaheryar Alam Khan

Reg no:-

364823

Submitted To:-

Dr. Qasim Umar Khan

Date: - 21, March, 2022

Q: Take a 5-featured Multi-Linear Regression Problem having at least 1000 training examples, plot the cost function vs. number of iterations, after that show the surface 3D plot of Cost function having variable θ_0 and θ_1 , at the end also show the contour plot of the cost function.

I have taken a problem of climate **Temperature** prediction using other features in relation to it like, **Humidity**, **Wind direction**, **Atmospheric pressure**, **Windspeed** and **Visibility**.

First of all I downloaded the dataset from Kaggle having link as follows: -

https://www.kaggle.com/datasets/zakriarehman/weather-data-for-linear-regression

Then I selected my feature for prediction as Temperature and selected other features as inputs.

After arranging in a useable format in excel as follows:-

					3					,	
	L1 ▼ (f _x										
1	Α	В	С	D	Е	F	G	Н	I	J	K
1	Humidity	,	Wind_Bearing_degrees	,	Pressure_millibars	,	Wind_Speed_kmh	,	Visibility_km	,	Temperature_c
2	0.92	,	130	,	1021.6	,	11.27	,	8.05	,	-0.55555556
3	0.73	,	330	,	1017	,	20.93	,	16.1	,	21.11111111
4	0.97	,	193	,	1013.99	,	5.9731	,	14.9086	,	16.6
5	0.82	,	300	,	1031.59	,	3.22	,	16.1	,	1.6
6	0.6	,	116	,	1020.88	,	10.8836	,	9.982	,	2.194444444
7	0.32	,	190	,	1015.33	,	21.4613	,	10.3523	,	27.53888889

Then I converted this excel data file to text form for usage in my MATLAB Program as shown on next page: -

Shahze	eb_Awan_d	data_wea	ther - Notepad					×
File Edit	Format	View	Help					
0.92	,	130	,	1021.6 ,	11.27	8.05	-0.55555556	
0.73	,	330	,	1017 ,	20.93 ,	16.1 ,	21.11111111	
0.97	,	193	,	1013.99 ,	5.9731 ,	14.9086 ,	16.6	
0.82	,	300	,	1031.59 ,	3.22	16.1 ,	1.6	
0.6	,	116	,	1020.88 ,	10.8836 ,	9.982	2.194444444	
0.32	,	190	,	1015.33 ,	21.4613 ,	10.3523 ,	27.53888889	
0.84	,	170	,	1009.04,	7.9695	11.1251 ,	19.97777778	
0.86	,	30	,	1009.6	14.49	15.134 ,	11.11111111	
0.73	,	351	,	1018.39	14.007	15.8263 ,	8.405555556	
0.81	,	320	,	1003.89	6.44	7.8568	1.7	
0.88	,	141	,	1021.28 ,	14.007	6.0214	-2.22222222	
0.6	,	204	,	1019.52	1.4168	15.8263 ,	21.9	
0.87	,	1	,	1015.92 ,	11.0285	14.9086 ,	17.10555556	
0.73	,	297	,	1013.06 ,	4.0733	9.7566	17.77222222	
0.39	,	35	,	1025.59	7.6636	9.982	24.95	
0.92	,	310	,	1024.3	3.22	3.4615	-2.711111111	
0.78	,	180	,	1018.76	4.83	9.982	18.88888889	
മാ	•	200	•	1000	רסמכ מר	1E 0163	A 207777770	

Main code:-

```
%% Load Data
data = load('Shahzeb Awan data weather.txt');
X = data(:, 1:5);% input features
y = data(:, 6); % output column is number 6
m = length(y); % length of dataset
%% ========= Part 1: Feature Normalization ============
% Scale features and set them to zero mean
fprintf('Normalizing Features ...\n');
[X mu sigma] = featureNormalize(X); % This function is explained in this
%report on later pages in detail
% Add intercept term to X
X = [ones(m, 1) X];
%% ======= Part 2: Gradient Descent ========
fprintf('Running gradient descent ...\n');
\mbox{\%} Choosing some alpha value
alpha = 0.01;
num iters = 1000;
% Init Theta and Running Gradient Descent
theta = [1;44;9;2;5;1]; % nice then zeros (comparitively better)
%gradient decent function is also explained later on
[theta, J1] = gradientDescentMulti2(X, y, theta, alpha, num iters);
```

```
% Plot the convergence graph
figure;
plot(1:numel(J1), J1, 'b');
xlabel('Number of iterations');
ylabel('Cost J');
% Display gradient descent's result
fprintf('Theta computed from gradient descent: \n');
fprintf(' %f \n', theta);
fprintf('\n');
%if Humidity is 0.88 , Wind bearing degrees 141, pressure in mb is 1021.28,
wind speed in km/h is 14.007
% and visibility in km is 6.0214 then Temperature should be round about :-2.22
Temperature=[1, (0.88-mu(1))/sigma(1), (141-mu(2))/sigma(2), (1021.28-mu(1))/sigma(2), (1021.28-mu(1))/sigma(1), (141-mu(2))/sigma(2), (1021.28-mu(1))/sigma(1), (141-mu(2))/sigma(2), (1021.28-mu(1))/sigma(2), (1021.28-mu(1))
mu(3))/sigma(3),(14.007-mu(4))/sigma(4),(6.0214-mu(5))/sigma(5)]*theta;
fprintf(['Predicted Temperature ' ...
                  '(using gradient descent):\n %f\n'], Temperature);
%% ========== Part 3: Normal Equations ===========
%% Analytical solution through ordinary least squares
fprintf('Solving with normal equations...\n');
%% Load Data
data = csvread('Shahzeb Awan data weather.txt');
X = data(:, 1:5);
y = data(:, 6);
m = length(y);
% Add intercept term to X
X = [ones(m, 1) X];
% Calculate the parameters from the normal equation
theta = normalEqn(X, y); %this is also explained separately in following pages
% Display normal equation's result
fprintf('Theta computed from the normal equations: \n');
fprintf(' %f \n', theta);
fprintf('\n');
%if Humidity is 0.88 , Wind bearing degrees 141, pressure in mb is 1021.28,
wind speed in km/h is 14.007
% and visibility in km is 6.0214 then Temperature should be round about :-2.22
Temperature = [1,0.88], 141, 1021.28, 14.007, 6.0214]*theta;
```

```
fprintf(['Predicted Temperature ' ...
         '(using normal equations): \n %f\n'], Temperature);
%% ======== Part 4: Visualizing J(theta 0, theta 1) =========
fprintf('Visualizing J(theta 0, theta 1) ...\n')
% Grid over which we will calculate J
theta0 vals = linspace(-60, 100, 100);
theta1 vals = linspace(-60, 60, 100);
% initialize J vals to a matrix of 0's
J vals = zeros(length(theta0 vals), length(theta1 vals));
% Fill out J vals
for i = 1:length(theta0 vals)
    for j = 1:length(theta1 vals)
     t = [theta0 vals(i); theta1_vals(j);0.002519;-0.003176;-
0.179258; 0.361726];
     J vals(i,j) = computeCostMulti(X, y, t);
end
% Because of the way meshgrids work in the surf command, we need to
% transpose J vals before calling surf, or else the axes will be flipped
J vals = J vals';
% Surface plot
figure;
surf(theta0 vals, theta1 vals, J vals)
xlabel('\theta 0'); ylabel('\theta 1');
% Contour plot
figure;
% Plot J vals as 15 contours spaced logarithmically between 0.01 and 100
contour(theta0 vals, theta1 vals, J vals, logspace(-2, 3, 20))
xlabel('\theta 0'); ylabel('\theta 1');
hold on;
plot(theta(1), theta(2), 'rx', 'MarkerSize', 10, 'LineWidth', 2);
%my minimum cost Function Result at my Selected thetas from gradient decent:-
J=computeCostMulti(X, y, theta)
```

Gradient-decent function:-

```
function [theta, J history] = gradientDescentMulti2(X, y, theta, alpha,
num iters)
% Initialize some useful values
m = length(y); % number of training examples
J history = zeros(num iters, 1);
for iter = 1:num iters
% Perform a single gradient step on the parameter vector theta.
gradient=zeros(6,1);
  for i=1:m,
    for j=1:6,
    gradient(j,1) = gradient(j,1) + (theta'*X(i,:)'-y(i))*X(i,j);
    end
  end
theta=theta-alpha/m*gradient;
    % Save the cost J in every iteration
    J history(iter) = computeCostMulti(X, y, theta);
end
```

Feature normalization:-

```
function [X_norm, mu, sigma] = featureNormalize(X)
%    the mean value of each feature is 0 and the standard deviation
%    is 1.
X_norm = X;
mu = zeros(1, 5);
sigma = zeros(1, 5);

mu=mean(X);
sigma=std(X);
X_norm=(X-mu)./sigma;
end
```

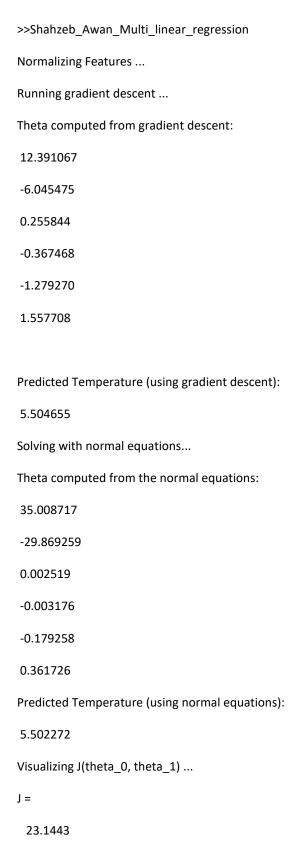
Closed Form Solution using normal Eqn:

```
function [theta] = normalEqn(X, y)
%Computes the closed-form solution to linear regression
theta = zeros(5, 1);
theta=pinv(X'*X)*X'*y;
end
```

My Cost function:-

```
function J = computeCostMulti(X, y, theta)
m = length(y); % number of training examples
J = 0;
J=1/(2*m)*(X*theta-y)'*(X*theta-y);
end
```

Command Window Result:-



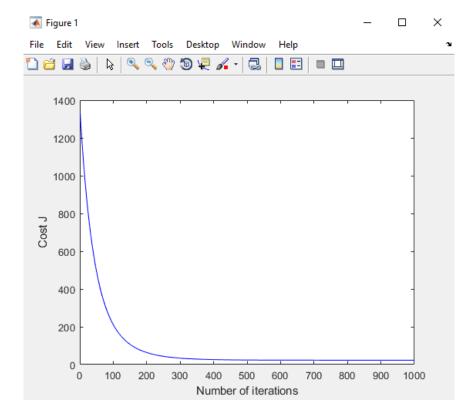


Figure 1 Cost function vs number of iterations

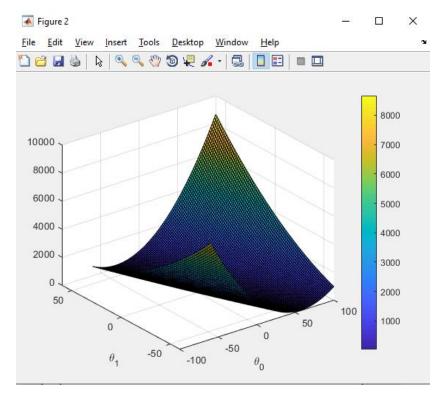


Figure 2 Cost Function vs variable Theta 0 and Theta 1

.

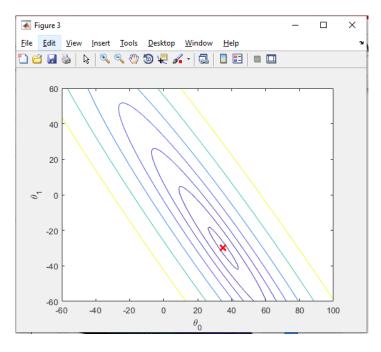


Figure 3 Contour plot of Cost function for theta 0 and theta 1

Test dataset:-

Humidity	Wind Angle (deg)	Pressure in (mb)	Wind speed in (km/h)	Visibility in (km)	Temperature in (C), to be predicted
0.39	35	1025.59	7.6636	9.982	24.95
0.83	190	1014.18	1.7549	16.1	7.761
0.7	13	1012.96	12.5097	9.982	17.81

Theta computed from gradient descent:

[Theta 0=12.391067] [Theta 1=-6.045475] [Theta 2=0.255844]

[Theta 3=-0.367468] [Theta 4=-1.279270] [Theta 5=1.557708]