

```

1  function J = computeCostMulti(X, y, theta)
2  %COMPUTECOSTMULTI Compute cost for linear regression with multiple
  • variables
3  % J = COMPUTECOSTMULTI(X, y, theta) computes the cost of using theta
  • as the
4  % parameter for linear regression to fit the data points in X and y
5
6  % Initialize some useful values
7  m = length(y); % number of training examples
8
9  % You need to return the following variables correctly
10 J = 0;
11
12 % ===== YOUR CODE HERE =====
13 % Instructions: Compute the cost of a particular choice of theta
14 %               You should set J to the cost.
15
16 h=X*theta; % funcion de hipotesis
17 e=h.-y; % error de cada prediccion
18 e_cuadrado=e.^2; % cuadrado de los errores
19 suma_e_cuadrado=sum(e_cuadrado); % suma de los cuadrados
20
21 J=1/(2*m)*suma_e_cuadrado % Cost Function
22
23 %
  • =====
24

```

```

1  function [theta, J_history] = gradientDescentMulti(X, y, theta, alpha,
•  num_iters)
2  %GRADIENDESCENTMULTI Performs gradient descent to learn theta
3  %   theta = GRADIENDESCENTMULTI(x, y, theta, alpha, num_iters) updates
•  theta by
4  %   taking num_iters gradient steps with learning rate alpha
5
6  % Initialize some useful values
7  m = length(y); % number of training examples
8  J_history = zeros(num_iters, 1);
9
10 for iteration = 1:num_iters
11     % Perform a single gradient step on the parameter vector theta.
12
13     % we minimize the value of J(theta) by changing the values of the
14     % vector theta NOT changing X or y
15
16     % alpha = learning rate as a single number
17
18     % hypothesis = mx1 column vector
19     % X = mxn matrix
20     % theta = nx1 column vector
21     hypothesis = X * theta;
22
23     % errors = mx1 column vector
24     % y = mx1 column vector
25     errors = hypothesis .- y;
26
27     newDecrement = (alpha * (1/m) * errors' * X);
28
29     theta = theta - newDecrement';
30
31     % Save the cost J in every iteration
32     J_history(iteration) = computeCostMulti(X, y, theta);
33
34 end
35
36 end

```

```
1  function [theta] = normalEqn(X, y)
2  %NORMALEQN Computes the closed-form solution to linear regression
3  %  NORMALEQN(X,y) computes the closed-form solution to linear
4  %  regression using the normal equations.
5
6  theta = zeros(size(X, 2), 1);
7
8  % ===== YOUR CODE HERE =====
9  % Instructions: Complete the code to compute the closed form solution
10 %                to linear regression and put the result in theta.
11 %
12
13 % ----- Sample Solution -----
14
15 theta = pinv(X' * X) * X' * y;
16
17
18 % -----
19
20
21 % =====
22
23 end
24
```