COMPM012 - Coursework 1

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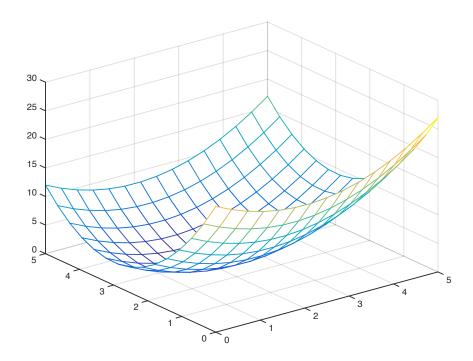
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1 Gradient Descent Visualisation

a) To produce the plot the following commands are used:

```
>> [X,Y] = \mathbf{meshgrid}(\mathbf{linspace}(0,5,15), \mathbf{linspace}(0,5,15));
>> \mathbf{mesh}(X,Y, fcarg(X,Y));
```

Which gives us the plot

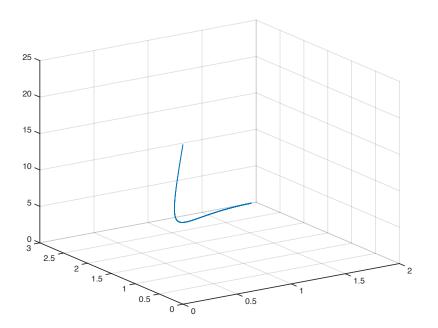


b) i) The modified graddesc function becomes:

```
function [soln, X, Y, Z] = graddesc(f, g, i,e, t)
         gi = feval(g,i);
         X = [];
         Y = [];
         Z = [];
         \mathbf{while} \, (\mathbf{norm}(\; \mathbf{gi} \,) \! > \! \mathbf{t} \,) \quad \, \% \  \, \textit{crude} \  \, \textit{termination} \  \, \textit{condition}
               i = i - e .* feval(g,i);
               gi = feval(g, i);
               X = [X i (1)];
               Y = [Y i (2)];
               Z = [Z fc(i)];
         end
         soln = i ;
   end
ii) We produce the plot with the commands
```

```
% Gradient descent
[result, X, Y, Z] = graddesc('fc', 'dfc', [0,0], 0.001, 0.1);
% Plot descent path
hold on
\mathbf{plot3}\left(\mathrm{X},\ \mathrm{Y},\ \mathrm{Z}\right)
grid on
hold off
```

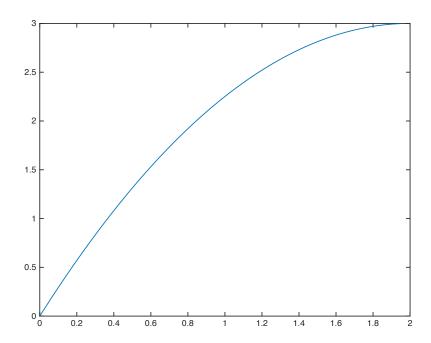
The obtained plot looks like



iii) We can plot the projection to the XY-plane by simply plotting the $X,\ Y$ values and ignore the Z values. We do this with the following commands

 $\begin{tabular}{ll} \% \ Plot \ projection \ to \ XY \ plane \\ \textbf{plot} (X, \ Y) \end{tabular}$

Which produces the following plot



2 Gradient Descent for Linear Regression

a) My implementation of gradient descent is

```
function [sol, guesses] = mygraddesc(A, b, guess, step, tol)
    mse = @(w) (A*w - b).' * (A*w - b);
    dmsedw = @(w) (-2*A.' * b) + (2*A.' * A * w);
    guesses = [guess; mse(guess)]; % Used for visualisation

while norm(dmsedw(guess)) > tol
    guess = guess - dmsedw(guess)*step;
    guesses = [guesses [guess; mse(guess)]];
end

sol = guess;
```

b) We can find the solution to the system using the gradient descent implementation by putting the coefficients for x_1 and x_2 in a matrix A and the right-hand-sides of the equations in a vector b. Passing A and

b as arguments to the gradient descent function along with an initial guess, step size, and tolerance gives us the solution to the system. In Matlab we do:

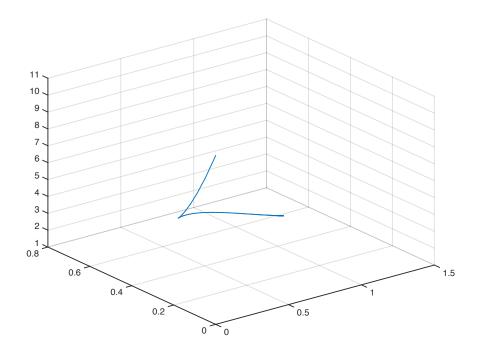
```
A = \begin{bmatrix} 1 & -1; & 1 & 1; & 1 & 2 \end{bmatrix}
b = \begin{bmatrix} 1; & 1; & 3 \end{bmatrix};
guess = \begin{bmatrix} 0; & 0 \end{bmatrix};
mygraddesc(A, b, guess, 0.01, 0.0001)
Which rerturns
ans = \begin{bmatrix} 1.2857 \\ 0.5714 \end{bmatrix}
```

So $x_1 = 1.2857$ and $x_2 = 0.5714$ is the least squares solution to the system.

c) My gradient descent implementation lets us retrieve the guesses for each iteration of the descent. We plot this as follows

```
[~, guesses] = mygraddesc(A, b, guess, 0.01, 0.001);
plot3(guesses(1,:), guesses(2,:), guesses(3,:));
hold on
grid on
hold off
```

Which produces the plot



3 Convergence of Gradient Descent in a Single Variable

- a)
- b)
- c)
- d)