

# Gradient Descent

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## Gradient

Gradient descent method is based on gradient

$$\nabla f = \frac{\partial f}{\partial x_1} \mathbf{e}_1 + \cdots + \frac{\partial f}{\partial x_n} \mathbf{e}_n$$

gradient always point to the ascent direction

## Gradient Descent

f is object function, and this is unconstrained

$$\min_x f$$

```
f = (@(X) (exp(X(1,:)-1) + exp(1-X(2,:)) + (X(1,:) - X(2,:)).^2));  
%f = (@(X) (sin(0.5*X(1,:)).^2 - 0.25 * X(2,:).^2 + 3) .* cos(2*X(1,:) + 1 - exp(X(2,:))) ) )
```

## Fail situation

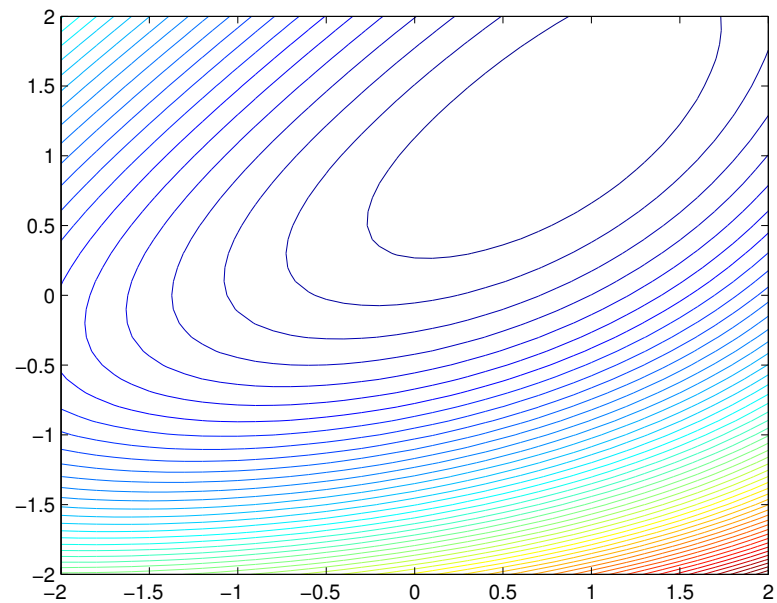
Rosenbrock function Gradient descent/ascent algorithm zig-zags, because the gradient is nearly orthogonal to the direction of the local minimum in these regions. It's hard to convergence

$$f(x, y) = (1 - x)^2 + 100 * (y - x^2)^2$$

```
%f = (@(X) (1-X(1,:)).^2 + 100 * (X(2,:) - X(1,:).^2).^2);
```

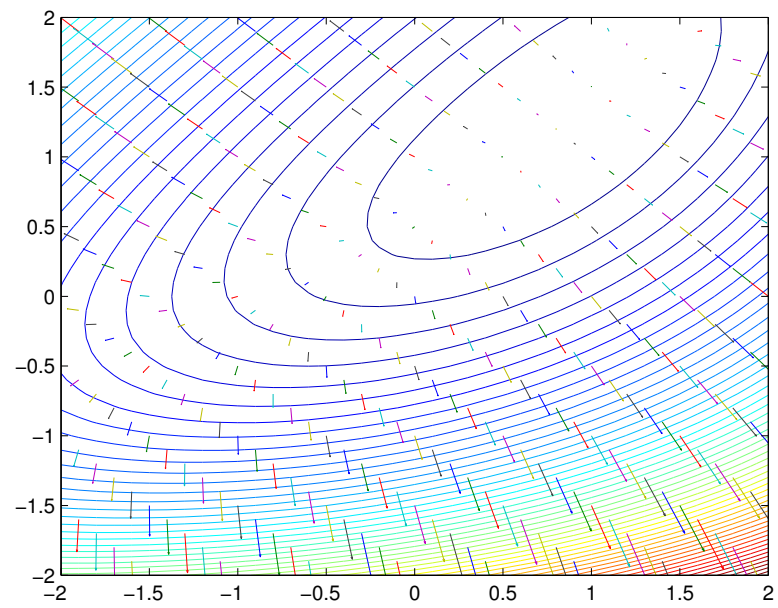
## Plot contour

```
[X, Y] = meshgrid(-2:0.1:2);  
XX = [reshape(X, 1, numel(X)); reshape(Y, 1, numel(Y))];  
%surf(X, Y, reshape(f(XX), length(X), length(X)))  
contour(X, Y, reshape(f(XX), length(X), length(X)), 50);  
  
hold on;
```



plot gradient of function

```
for i=1:5:length(XX)  
    tmp = XX(:,i);  
    g = gradient_of_function(f, tmp);  
    %plot([tmp(1),tmp(1)+g(1)*0.02],[tmp(1),tmp(2)+g(1)*0.02]);  
    quiver(tmp(1),tmp(2),g(1)*0.02,g(2)*0.02);  
end
```



calculation

```
x0 = [-1; -1];
[x, v, h] = gradient(f, x0)
```

```
% built-in method
[x_in, v_in] = fminunc(f, x0)
```

x =

```
0.7960
1.2038
```

v =

```
1.7974
```

h =

Columns 1 through 7

-1.0000	-1.0271	-0.4515	0.6432	0.8185	0.7755	0.7859
-1.0000	0.4778	0.2130	1.0809	1.1279	1.1801	1.2059

Columns 8 through 12

0.7925	0.7963	0.7956	0.7959	0.7960
1.2007	1.2024	1.2033	1.2039	1.2038

Warning: Gradient must be provided for trust-region algorithm;  
using line-search algorithm instead.

Local minimum found.

Optimization completed because the size of the gradient is less than  
the default value of the function tolerance.

x\_in =

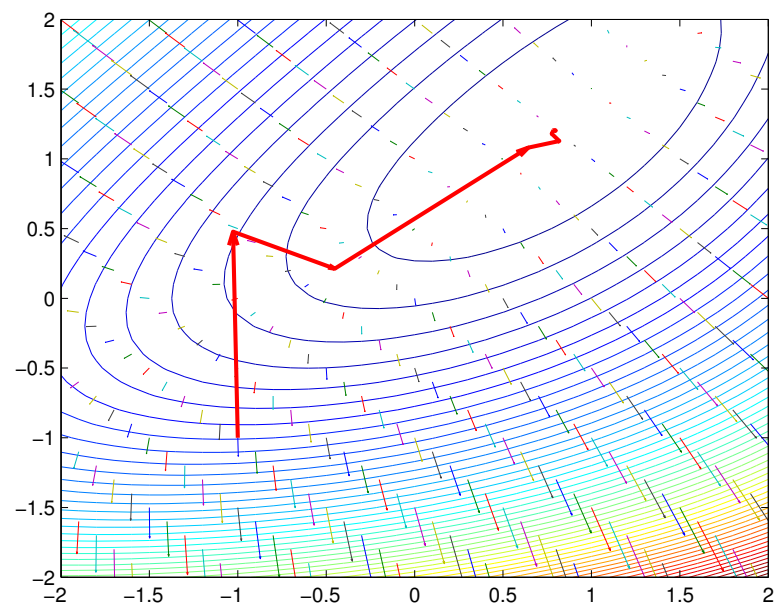
0.7961
1.2039

v\_in =

1.7974
--------

plot descent steps

```
for i=2:length(h)
    tmp1 = h(:,i-1);
    tmp2 = h(:,i);
    quiver(tmp1(1),tmp1(2),tmp2(1)-tmp1(1),tmp2(2)-tmp1(2), 0, 'r','LineWidth',2)
end
```



## Reference

1. <http://www.onmyphd.com/?p=gradient.descent>
2. Convex Optimization
3. <https://en.wikipedia.org/wiki/Gradient>
4. [https://en.wikipedia.org/wiki/Gradient\\_descent](https://en.wikipedia.org/wiki/Gradient_descent)
5. <http://stronglyconvex.com/blog/gradient-descent.html>