Newton-Raphson Method for Convex Optimization

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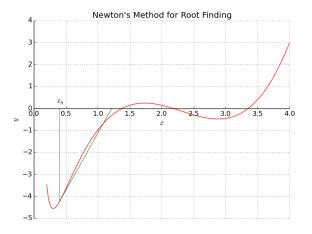
Given a differentiable function $f : \mathbb{R} \to \mathbb{R}$ we want to find the instances when f(x) = 0 (not generally solvable in closed form).

Newton's Method:

- 1. Take a starting position x_0
- 2. Find where the tangent line $y = f(x_n) + (x_{n+1} x_n)f'(x_n)$ is 0 and iterate:

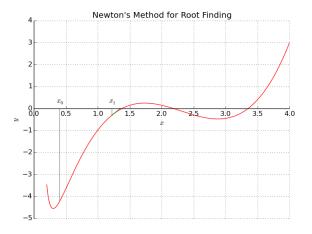
$$y = f(x_n) + (x_{n+1} - x_n)f'(x_n) = 0$$
$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

Initial
$$x_0 = 0.4$$
, $f(x) = (x - 1)(x - 3)^2 + e^{\frac{1}{3x}} - \cos(\frac{x}{2}) - 1.5$



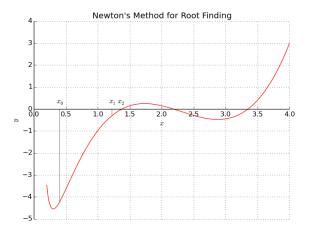
$$x_0 = 0.4 \quad \Delta = 0.9805$$

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$$x_1 = 1.2167$$
 $\Delta = 0.1638$

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$$x_2 = 1.3487$$
 $\Delta = 0.0319$



Optimizing Function Using Newton's Method

Instead of finding where f(x) = 0, how can we find where f'(x) = 0?

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Instead of:

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

Use derivatives:

$$x_{n+1} = x_n - \frac{f'(x_n)}{f''(x_n)}$$

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Def: The gradient of $f: \mathbb{R}^n \mapsto \mathbb{R}$ is

$$\nabla f = \begin{bmatrix} \frac{\partial}{\partial x_1} f & \frac{\partial}{\partial x_2} f & \dots & \frac{\partial}{\partial x_n} f \end{bmatrix}$$

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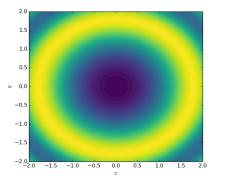
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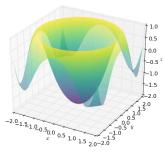
$$\nabla f = \begin{bmatrix} \frac{\partial}{\partial x_1} f & \frac{\partial}{\partial x_2} f & \dots & \frac{\partial}{\partial x_n} f \end{bmatrix}$$

Def: The Hessian of $f: \mathbb{R}^n \to \mathbb{R}$ is

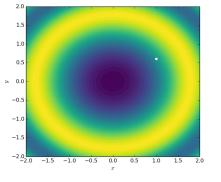
$$H_{f} = \begin{bmatrix} \frac{\partial^{2} f}{\partial x_{1} \partial x_{1}} & \frac{\partial^{2} f}{\partial x_{1} \partial x_{2}} & \cdots & \frac{\partial^{2} f}{\partial x_{1} \partial x_{n}} \\ \frac{\partial^{2} f}{\partial x_{2} \partial x_{1}} & \ddots & \frac{\partial^{2} f}{\partial x_{2} \partial x_{n}} \\ \vdots & & \ddots & \vdots \\ \frac{\partial^{2} f}{\partial x_{n} \partial x_{1}} & \frac{\partial^{2} f}{\partial x_{n} \partial x_{2}} & \cdots & \frac{\partial^{2} f}{\partial x_{n} \partial x_{n}} \end{bmatrix}$$

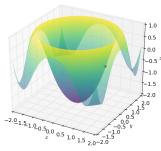
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, $f(x) = -\cos(x^2 + y^2) - e^{-(x^2 + y^2)}$





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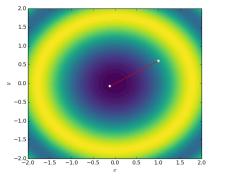


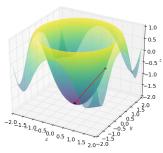


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$$\Delta = 1.1661$$

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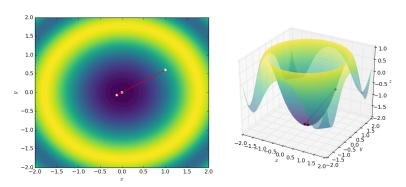




$$x_1 = \begin{bmatrix} -0.1167 & -0.0700 \end{bmatrix}$$

$$\Delta = 0.1361$$

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, $f(x) = -\cos(x^2 + y^2) - e^{-(x^2 + y^2)}$



$$x_2 = \begin{bmatrix} -7.8 \times 10^{-5} & -4.7 \times 10^{-5} \end{bmatrix}$$
 $\Delta = 9.15 \times 10^{-5}$

Converges in 4 iterations.