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## relation between positive function and its gradient when its Hessian matrix is bounded

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Let  $f : R^n \rightarrow R$  a positive function, twice differentiable everywhere. Furthermore, let  $\|\mathbf{H}_f(\mathbf{x})\|_2 \leq M, M > 0 \forall \mathbf{x} \in R^n$ , where  $\mathbf{H}_f(\mathbf{x})$  is the Hessian matrix of  $f(\mathbf{x})$ . Then, for any  $\mathbf{x} \in R^n$ ,

$$\|\nabla f(\mathbf{x})\|_2 \leq \sqrt{2Mf(\mathbf{x})}$$

*Proof.* Let  $\mathbf{x}, \mathbf{x}_0 \in R^n$  be arbitrary points. By positivity of  $f(\mathbf{x})$ , writing Taylor expansion of  $f(\mathbf{x})$  with Lagrange error formula around  $\mathbf{x}_0$ , a point  $\mathbf{c} \in R^n$  exists such that:

$$\begin{aligned} 0 &\leq f(\mathbf{x}) \\ &= f(\mathbf{x}_0) + \nabla f(\mathbf{x}_0) \cdot (\mathbf{x} - \mathbf{x}_0) + \frac{1}{2}(\mathbf{x} - \mathbf{x}_0)^T \cdot \mathbf{H}_f(\mathbf{c}) \cdot (\mathbf{x} - \mathbf{x}_0) \\ &= \left| f(\mathbf{x}_0) + \nabla f(\mathbf{x}_0) \cdot (\mathbf{x} - \mathbf{x}_0) + \frac{1}{2}(\mathbf{x} - \mathbf{x}_0)^T \cdot \mathbf{H}_f(\mathbf{c}) \cdot (\mathbf{x} - \mathbf{x}_0) \right| \\ &\leq f(\mathbf{x}_0) + \|\nabla f(\mathbf{x}_0)\|_2 \|\mathbf{x} - \mathbf{x}_0\|_2 + \frac{1}{2} |(\mathbf{x} - \mathbf{x}_0)^T \cdot \mathbf{H}_f(\mathbf{c}) \cdot (\mathbf{x} - \mathbf{x}_0)| \\ &\leq f(\mathbf{x}_0) + \|\nabla f(\mathbf{x}_0)\|_2 \|\mathbf{x} - \mathbf{x}_0\|_2 + \frac{1}{2} \|\mathbf{H}_f(\mathbf{c})\|_2 \|\mathbf{x} - \mathbf{x}_0\|_2^2 \quad (\text{by Cauchy-Schwartz inequality}) \\ &\leq f(\mathbf{x}_0) + \|\nabla f(\mathbf{x}_0)\|_2 \|\mathbf{x} - \mathbf{x}_0\|_2 + \frac{1}{2} M \|\mathbf{x} - \mathbf{x}_0\|_2^2 \end{aligned}$$

The rightest side is a second degree polynomial in variable  $\|\mathbf{x} - \mathbf{x}_0\|_2$ ; for it to be positive for any choice of  $\|\mathbf{x} - \mathbf{x}_0\|_2$  (that is, for any choice of  $\mathbf{x}$ ), the discriminant

$$\|\nabla f(\mathbf{x}_0)\|_2^2 - 4 \cdot \frac{1}{2} M f(\mathbf{x}_0)$$

must be negative, whence the thesis.  $\square$

Note: The condition on the boundedness of the Hessian matrix is actually needed. In fact, in the Lagrange form remainder, the constant  $\mathbf{c}$  depends upon the point  $\mathbf{x}$ . Thus, if we couldn't rely on the condition  $\|\mathbf{H}_f(\mathbf{x})\|_2 \leq M$ , we could only state  $f(\mathbf{x}_0) + \|\nabla f(\mathbf{x}_0)\|_2 \|\mathbf{x} - \mathbf{x}_0\|_2 + \frac{1}{2} \|\mathbf{H}_f(\mathbf{c}(\mathbf{x}))\|_2 \|\mathbf{x} - \mathbf{x}_0\|_2^2 \geq 0$  which, not being a second degree polynomial, wouldn't imply any particular further condition. Moreover, in the case  $n = 1$ , the lemma assumes the simpler form: Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  a positive function, twice differentiable everywhere. Furthermore, let  $f''(x) \leq M, M > 0 \forall x \in \mathbb{R}$ . Then, for any  $x \in \mathbb{R}$ ,  $|f'(x)| \leq \sqrt{2Mf(x)}$ .