

Multivariate Functions and Calculus: Review

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1 Continuous Function

- A univariate function f is *continuous* at x , if given any $\epsilon > 0$, $\exists \delta > 0$ such that $|y - x| < \delta$ implies that $|f(y) - f(x)| < \epsilon$
- Pictorially, the graph of f does not contain a ‘break’ at x
- A multivariate function F is *continuous* at x if, given any $\epsilon > 0$, $\exists \delta > 0$ such that if $\|y - x\| < \delta$ then $|F(y) - F(x)| < \epsilon$

2 Derivatives

- A univariate function f is said to be *differentiable* at x^* if following limit exists.

$$f'(x^*) = \lim_{h \rightarrow 0} \frac{f(x^* + h) - f(x^*)}{h}$$

Effectively, derivative at x^* is the rate of change of the function in the vicinity of x^* , which in limit is the *slope*.

- Example of a function which is continuous but non-differentiable function $|x|$.
- Further differentiating the derivatives yields higher order derivatives
- With multivariate function F , we talk about *partial derivatives* along each coordinate

$$\left. \frac{\partial F}{\partial x_i} \right|_{x^*} = \lim_{h \rightarrow 0} \frac{F(x_1^*, x_2^*, \dots, x_i^* + h, \dots, x_n^*) - F(x^*)}{h}$$

- Generally, F is said to be *differentiable* at x^* if it all n partial derivatives of F are continuous at x^*
- Gradient vector of F

$$\nabla F(x) = g(x) = \begin{pmatrix} \frac{\partial F}{\partial x_1} \\ \frac{\partial F}{\partial x_2} \\ \vdots \\ \frac{\partial F}{\partial x_n} \end{pmatrix}$$

- For a linear function F , gradient of F is a constant vector
- Similarly, we can define higher order partial derivatives

$$\frac{\partial}{\partial x_i} \left(\frac{\partial F}{\partial x_j} \right)$$

- Usually written as

$$\frac{\partial^2 F}{\partial x_i \partial x_j}, i \neq j; \quad \frac{\partial^2 F}{\partial x_i^2}, i = j;$$

3 Hessian

- *Hessian matrix* of $F(x)$ is used to represent second order derivatives.

$$\nabla^2 F(x) = G(x) = \begin{pmatrix} \frac{\partial^2 F}{\partial x_1^2} & \frac{\partial^2 F}{\partial x_2 \partial x_1} & \cdots & \frac{\partial^2 F}{\partial x_n \partial x_1} \\ \frac{\partial^2 F}{\partial x_1 \partial x_2} & \frac{\partial^2 F}{\partial x_2^2} & \cdots & \frac{\partial^2 F}{\partial x_n \partial x_2} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 F}{\partial x_1 \partial x_n} & \frac{\partial^2 F}{\partial x_2 \partial x_n} & \cdots & \frac{\partial^2 F}{\partial x_n^2} \end{pmatrix}$$

It can be shown that

$$\frac{\partial^2 F}{\partial x_i \partial x_j} = \frac{\partial^2 F}{\partial x_j \partial x_i}$$

Consequently, *Hessian* matrix is symmetric.

- If Hessian is constant, F is a *quadratic* function

$$F(x) = \frac{1}{2}x^T Gx + c^T x + \alpha$$

- Hessian matrix of the scalar function $F(x)$ is Jacobian matrix of vector function $g(x)$

Class of functions with continuous derivatives of order 1 through k is denoted by \mathcal{C}^k . Functions with high degrees of differentiability are referred as *smooth* function.

4 Order Notation

Let $f(h)$ be a univariate function of h . Then the function $f(h)$ is said to be of order h^p ($O(h^p)$) if there exists a finite number M , $M > 0$, independent of h , such that as $|h|$ approaches zero

$$|f(h)| \leq M|h|^p.$$

For sufficiently small $|h|$, the rate at which an $O(h^p)$ term will go to zero will increase as p increases. It is often desirable to find the largest value of p for which above inequality holds.

5 Taylor's Series

Theorem 1. If $f(x) \in \mathcal{C}^r$, then there exists a scalar θ ($0 \leq \theta \leq 1$), such that

$$f(x+h) = f(x) + hf'(x) + \frac{1}{2!}h^2 f''(x) + \cdots + \frac{1}{(r-1)!}h^{r-1} f^{(r-1)}(x) + \frac{1}{r!}h^r f^{(r)}(x+\theta h)$$

The above Taylor-series can also be written as

$$f(x+h) = f(x) + hf'(x) + \frac{1}{2}h^2 f''(x) + \cdots + \frac{1}{(r-1)!}h^{r-1} f^{(r-1)}(x) + O(h^r)$$

assuming that $|f(r)|$ is finite in the interval $[x, x+h]$.

- For $r < p$, $\mathcal{C}^r \subset \mathcal{C}^p$. Hence, the series can be truncated at lower order.
- If function is infinitely differentiable, then we have infinite Taylor's series

Taylor theorem also holds for sufficiently smooth multivariate functions. With point x , direction p and scalar h

$$F(x + hp) = F(x) + h\nabla(x)^T p + \frac{1}{2}h^2 p^T \nabla^2 F(x)p + \cdots \\ + \frac{1}{(r-1)!}h^{r-1}D^{r-1}F(x) + \frac{1}{r!}h^r D^r F(x + \theta hp)$$

where,

$$D^s F(x) = \sum_{i_1=1}^n \sum_{i_2=1}^n \cdots \sum_{i_s=1}^n \left\{ p_{i_1} p_{i_2} \cdots p_{i_s} \frac{\partial^s F(x)}{\partial x_{i_1} \partial x_{i_2} \cdots \partial x_{i_s}} \right\}$$

Restricting to first three terms

$$F(x + hp) = F(x) + h\nabla F(x)^T p + \frac{1}{2}h^2 p^T \nabla^2 F(x)p + O(h^3)$$

6 Review Exercises

1. (a) What is the order of the function?

(i) $f(x) = x + x^2$ i.e., is it $O(\epsilon^2)$ or $O(\epsilon)$?

(ii) $f(x) = 3x^{1.5} + 5x^2$, $x > 0$

where ϵ is a small number tending to zero.

- (b) What are the constants M in the above two cases.

2. Let $x \in \mathbb{R}^n$; show that $\nabla \|x\|_2 = \frac{1}{\|x\|_2} x$ for $x \neq 0$.

3. Let $F(x_1, x_2) = \frac{x_1 x_2}{x_1^2 + x_2^2}$ when $x \neq 0$. Show that $\nabla F(0)$ exist but, strictly, the function is not differentiable as it is not defined at $(0, 0)$.

4. Let $f(x) = x^3$; we are interested in (a) linear approximation of $f(x)$ around $x = 0$ and (b) its quadratic approximation. Find them and comment.

5. (a) Let $f(x) = 2 + x - x^3$; write the Taylor Series expression of $f(x)$ around $x = 0$ i.e.,

$$f(0 + \epsilon) = f(0) + \frac{\epsilon}{1!} f'(0) + \frac{\epsilon^2}{2!} f''(0) + \frac{\epsilon^3}{3!} f'''(\theta)$$

What would be the appropriate value of θ ?

- (b) Now consider that the Taylor Series is truncated at second derivative i.e.,

$$f(0 + \epsilon) = f(0) + \frac{\epsilon}{1!} f'(0) + \frac{\epsilon^2}{2!} f''(\theta)$$

What would be the value of θ you would choose? Make sure that θ is between 0 and ϵ .

References

- [1] Michael T. Heath, Scientific Computing: An Introductory Survey The McGraw-Hill Companies, 2nd edition, July 17, 2002.