

Summary of Matrix Calculus

15.1 Definitions, Notation and Preliminaries

- Derivative of scalar-valued function f with input $\mathbf{x} = (x_1, \dots, x_m)$
 - Interior point: $\{\mathbf{x} \in \mathcal{R}^{m \times 1} : \|\mathbf{x} - \mathbf{c}\| < r\}$ for some pos. const r
Applies to matrices too:
 $\{\mathbf{X} \in \mathcal{R}^{m \times n} : \|\mathbf{X} - \mathbf{C}\| < r\}$
 - **Def:** j^{th} (first) partial derivative (scalar-value function with vector input $\mathbf{x} = (x_1, \dots, x_m)^T$)
 $D_j f(\mathbf{c})$ denotes the j th part. deriv. of f at \mathbf{c} :
 $D_j f(\mathbf{c}) = \lim_{t \rightarrow 0} \frac{f(\mathbf{c} + t\mathbf{u}_j) - f(\mathbf{c})}{t}$ where \mathbf{u}_j is the j th row of identity mx.
Alternative notation: $\frac{\partial f(\mathbf{x})}{\partial x_j}$
Alternative notation: At times, it may be more convenient to reshape vector x as matrix \mathbf{X} and denote its partial derivative wrt element x_{ij} as $\frac{\partial f(\mathbf{X})}{\partial x_{ij}}$. The way we treat this derivative depends on whether the elements of \mathbf{X} are dependent (*e.g.* symmetric matrix) or independent (Sec. 15.1.f)
 - **Def:** Vector of partial derivatives $\mathbf{D}f(\mathbf{c})$ denotes vector of all part. derivs of f at \mathbf{c} :
 $\mathbf{D}f(\mathbf{c}) = (D_1 f(\mathbf{c}), D_2 f(\mathbf{c}), \dots, D_m f(\mathbf{c}))'$
Alternative notation: $\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}'}$
 - **Def:** $(\mathbf{D}f)'$ is called *gradient vector*
Alternative notation: $\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}}$
 - **Def:** *continuously differentiable*:
Function f with domain $S \in \mathcal{R}^{m \times 1}$ is continuously differentiable at the interior pt $\mathbf{c} \in S$ if $D_1 f(\mathbf{x}), \dots, D_m f(\mathbf{x})$ exist and are continuous at every pt in some neighbourhood of \mathbf{c} .
In this case the following holds: $\lim_{\mathbf{x} \rightarrow \mathbf{c}} \frac{f(\mathbf{x}) - [f(\mathbf{c}) + \mathbf{D}f(\mathbf{c})(\mathbf{x} - \mathbf{c})]}{\|\mathbf{x} - \mathbf{c}\|} = 0$
 - **Def:** ij^{th} (second) partial derivative $D_{ij}^2 f(\mathbf{x})$
Alternative notation: $\frac{\partial^2 f(\mathbf{x})}{\partial x_i \partial x_j}$
 - **Def:** *Hessian Matrix* $\mathbf{H}f$
An $m \times m$ matrix whose ij th element is $D_{ij}^2 f(\mathbf{x})$
- Derivative of vector-valued fn $\mathbf{f} = (f_1, \dots, f_p)'$ where each f_i takes input $\mathbf{x} = (x_1, \dots, x_m)'$.
 - $D_j f_s(\mathbf{c})$: j th partial derivative of f_s
 - $D_j \mathbf{f}(\mathbf{c})$ is $p \times 1$ vector $D_j \mathbf{f}(\mathbf{c}) = [D_j f_1(\mathbf{c}), \dots, D_j f_p(\mathbf{c})]'$
Alternative notation $\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}'}$ (row vector, $1 \times p$ – see Sec 15.1.c #287)
Alternative notation $\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}}$ (column vector, $p \times 1$)
 - $\mathbf{D}f$ is $p \times m$ matrix: $\mathbf{D}f(\mathbf{c}) = [D_1 \mathbf{f}(\mathbf{c}), \dots, D_p \mathbf{f}(\mathbf{c})]$
 $\mathbf{D}f$ is called *Jacobian* of \mathbf{f} and it's the matrix whose sj th element is $D_j f_s$.
 $(\mathbf{D}f)'$ is called the *gradient (matrix)* of \mathbf{f} .
Alternative notation to Jacobian: $\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}'}$ and it's the matrix whose sj th element is $\frac{\partial f_s(\mathbf{x})}{\partial x_j}$
- Derivative of matrix of functions $F = f_{st}$ where \mathbf{F} is $p \times q$
 - It's preferable to keep the j th partial derivatives of \mathbf{F} as a separate $p \times q$ matrix, denoted as:
 $\frac{\partial \mathbf{F}(\mathbf{x})}{\partial x_j}$ or $D_j \mathbf{F}(\mathbf{x})$

15.2 Differentiation of Scalar-valued Functions

- Lem 15.2.1 – If $f(\mathbf{x})$ does not vary wrt x_j at \mathbf{c} then $D_j f(\mathbf{c}) = 0$
- Lem 15.2.2 – Let l, h, r be functions defined as: $l = af + bg$, $h = fg$ and $r = f/g$. The rules of derivative for single-variable calculus function apply for the j th partial derivative of l, h, g .

15.3 Differentiation of Linear and Quadratic Forms

- Let $\mathbf{a} = (a_1, \dots, a_m)$ be constant (or fn of \mathbf{x} that is invariant wrt x_j), and \mathbf{A} be an $m \times m$ constant matrix (or matrix of functions invariant wrt x_j). Then:

- $\frac{\partial \mathbf{a}'\mathbf{x}}{\partial x_j} = a_j$ (see #294)
- $\frac{\partial \mathbf{a}'\mathbf{x}}{\partial \mathbf{x}} = \mathbf{a}$ or $\frac{\partial \mathbf{a}'\mathbf{x}}{\partial \mathbf{x}'} = \mathbf{a}'$
- $\frac{\partial \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial x_j} = \sum_{i=1}^m a_{ij}x_i + \sum_{k=1}^m a_{jk}x_k$ (see #295)
- $\frac{\partial \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial \mathbf{x}} = (\mathbf{A} + \mathbf{A}')\mathbf{x}$ — if \mathbf{A} symmetric then $\frac{\partial \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial \mathbf{x}} = 2\mathbf{A}\mathbf{x}$
- $\frac{\partial^2 \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial x_s \partial x_j} = a_{sj} + a_{js}$ (see #295)
- $\frac{\partial^2 \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial \mathbf{x}^2} = (\mathbf{A} + \mathbf{A}')$ — if \mathbf{A} symmetric then $\frac{\partial^2 \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial \mathbf{x}^2} = 2\mathbf{A}$