ML & LinAlg Math Cheat Sheet

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1 Notation

Vectors are column vectors denoted by lower-case bolded variables, such that

$$oldsymbol{x} = egin{bmatrix} x_1 \ dots \ x_N \end{bmatrix}.$$

A row vector is denoted $\mathbf{x}^{\top} = [x_1 \dots x_N]$. A matrix is indicated by a bolded upper-case variable, such that an $N \times M$ matrix is

$$m{A} = \{a_{ij}\} = [m{a}_1 \cdots m{a}_M] = \left[egin{array}{c} m{a}_1^{ op} \ dots \ m{a}_N^{ op} \end{array}
ight] = \left[egin{array}{c} a_{1,1}^{ op} & \cdots & a_{1,M} \ dots & \ddots & dots \ a_N^{ op} & \cdots & a_{N,M} \end{array}
ight].$$

2 Derivative

2.a Vector Gradient

$$\nabla_{\boldsymbol{x}} \boldsymbol{y} = \left[\frac{\partial \boldsymbol{y}}{\partial x_1}, \dots, \frac{\partial \boldsymbol{y}}{\partial x_N}\right] \tag{1}$$

3 Determinant Operator

3.a Random Properties

For scalar c and $N \times N$ identity matrix I,

$$\det(cI) = c^N.$$

4 Trace Operator

4.a Derivatives

4.a.i
$$\nabla_{\boldsymbol{x}}\operatorname{tr}(\boldsymbol{x}\boldsymbol{x}^{\top}\boldsymbol{A}) = 2\boldsymbol{x}^{\top}\boldsymbol{A}$$

For square $N \times N$ matrix \boldsymbol{A} ,

$$\nabla_{\boldsymbol{x}}\operatorname{tr}(\boldsymbol{x}\boldsymbol{x}^{\top}\boldsymbol{A}) = \frac{d}{d\boldsymbol{x}}\operatorname{tr}(\boldsymbol{x}\boldsymbol{x}^{\top}\boldsymbol{A}) = \frac{d}{d\boldsymbol{x}}\sum_{i}^{N}\sum_{k}^{N}x_{i}x_{k}a_{ik}.$$

Recall eq. (1), and consider for any $j \in \{1, \dots, N\}$:

$$\frac{\partial}{\partial x_j} \sum_{i=1}^{N} \sum_{k=1}^{N} x_i x_k a_{ik} = x_1 a_{j,1} + x_2 a_{j,2} + \dots + \frac{\partial}{\partial x_j} \sum_{k=1}^{N} x_j x_k a_{jk} + \dots + x_N a_{j,N}$$

$$= \sum_{i=1}^{N} x_i a_{ji} - x_j a_{jj} + \sum_{k=1}^{N} x_k a_{jk} - x_j a_{jj} + \frac{\partial}{\partial x_j} x_j x_j a_{jj}$$

$$= 2 \sum_{i=1}^{N} x_i a_{ji} - 2 x_j a_{jj} + 2 x_j a_{jj}$$

$$= 2 \mathbf{x}^{\top} \mathbf{a}_j.$$

This equally applies for any j in $1 \dots N$, and so for the full gradient:

$$\nabla_{\boldsymbol{x}}\operatorname{tr}(\boldsymbol{x}\boldsymbol{x}^{\top}\boldsymbol{A}) = \frac{d}{d\boldsymbol{x}}\sum_{i}^{N}\sum_{k}^{N}x_{i}x_{k}a_{ik} = [2\boldsymbol{x}^{\top}\boldsymbol{a}_{1}\cdots 2\boldsymbol{x}^{\top}\boldsymbol{a}_{N}] = 2\boldsymbol{x}^{\top}\boldsymbol{A}.$$
 (2)

4.b Relation to Determinant