

## 1 Basic Linear Algebra

- $\mathbf{A}^{-T} \triangleq (\mathbf{A}^T)^{-1} = (\mathbf{A}^{-1})^T$
- $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$
- $(\mathbf{AB})^{-1} = \mathbf{B}^{-1} \mathbf{A}^{-1}$ , iff  $\mathbf{A}$  and  $\mathbf{B}$  are invertible
- Frobenius norm  $\|\mathbf{A}\|_F = \sqrt{\text{Tr}(\mathbf{A}\mathbf{A}^T)}$
- **trace:**
  - $\text{Tr}(\mathbf{A}^T) = \text{Tr}(\mathbf{A})$
  - $\text{Tr}(\mathbf{A} + \mathbf{B}) = \text{Tr}(\mathbf{A}) + \text{Tr}(\mathbf{B})$
  - $\text{Tr}(\mathbf{AB}^T) = \text{Tr}(\mathbf{A}^T \mathbf{B}) = \sum_{i,j} (\mathbf{A} \circ \mathbf{B})_{(i,j)}$
  - $\text{Tr}(\mathbf{ABC}) = \text{Tr}(\mathbf{CAB}) = \text{Tr}(\mathbf{BCA})$
- **determinant:**
  - $\det \mathbf{A}^T = \det \mathbf{A}$
  - $\det \mathbf{A}^{-1} = (\det \mathbf{A})^{-1}$
  - $\det(\mathbf{AB}) = \det(\mathbf{A})\det(\mathbf{B})$ , for square matrices of equal size.
  - If  $\mathbf{A}$  is a triangular matrix (lower triangular or upper triangular),

$$\det \mathbf{A} = \prod_i A_{(i,i)}$$

## 2 Cholesky decomposition

The Cholesky decomposition of a positive-definite matrix  $\mathbf{A}$  is a decomposition of the form:

$$\mathbf{\Sigma} = \mathbf{L}\mathbf{L}^T$$

where  $\mathbf{L}$  is a lower triangular matrix

- $\mathbf{\Sigma}^{-1} = (\mathbf{L}\mathbf{L}^T)^{-1} = \mathbf{L}^{-T} \mathbf{L}^{-1}$
- $\mathbf{L} \setminus \mathbf{x} \triangleq \mathbf{L}^{-1} \mathbf{x}$
- $\mathbf{\Sigma}^{-1} \mathbf{x} = \mathbf{L}^T \setminus \mathbf{L} \setminus \mathbf{x}$
- $\mathbf{x}^T \mathbf{\Sigma}^{-1} \mathbf{x} = \|\mathbf{L} \setminus \mathbf{x}\|_2^2$

- $\text{Tr}(\Sigma_b^{-1} \Sigma_a) = \|\mathbf{L}_b \setminus \mathbf{L}_a\|_F^2$   
Proof:

$$\begin{aligned} \text{Tr}(\Sigma_b^{-1} \Sigma_a) &= \text{Tr}\left(\left(\mathbf{L}_b \mathbf{L}_b^T\right)^{-1} \mathbf{L}_a \mathbf{L}_a^T\right) \\ &= \text{Tr}\left(\mathbf{L}_b^{-T} \mathbf{L}_b^{-1} \mathbf{L}_a \mathbf{L}_a^T\right) = \text{Tr}\left(\mathbf{L}_a^T \mathbf{L}_b^{-T} \mathbf{L}_b^{-1} \mathbf{L}_a\right) \\ &= \text{Tr}\left(\left(\mathbf{L}_b^{-1} \mathbf{L}_a\right)^T \left(\mathbf{L}_b^{-1} \mathbf{L}_a\right)\right) = \|\mathbf{L}_b \setminus \mathbf{L}_a\|_F^2 \end{aligned}$$

- $\log(|\Sigma|) = 2 \sum_i \log(L_{(i,i)}) = 2\text{Tr}(\log(\mathbf{L}))$

### 3 Inverse

$$(\mathbf{I} + \mathbf{P})^{-1} = \mathbf{I} - (\mathbf{I} + \mathbf{P})^{-1} \mathbf{P} \quad (1)$$

$$(\mathbf{I} + \mathbf{PQ})^{-1} \mathbf{P} = \mathbf{P} (\mathbf{I} + \mathbf{QP})^{-1} \quad (2)$$

#### 3.1 Matrix inversion lemma (Sherman-Morrison-Woodbury)

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$$(\mathbf{A} + \mathbf{BCD})^{-1} = \mathbf{A}^{-1} - \mathbf{A}^{-1} \mathbf{B} (\mathbf{C}^{-1} + \mathbf{DA}^{-1} \mathbf{B})^{-1} \mathbf{DA}^{-1} \quad (3)$$

- $(\mathbf{A} + \mathbf{BCD})^{-1} = \mathbf{A}^{-1} - \mathbf{A}^{-1} \mathbf{B} (\mathbf{I} + \mathbf{CDA}^{-1} \mathbf{B})^{-1} \mathbf{CDA}^{-1}$
- $\left(\mathbf{A} + \mathbf{XBX}^T\right)^{-1} = \mathbf{A}^{-1} - \mathbf{A}^{-1} \mathbf{X} \left(\mathbf{B}^{-1} + \mathbf{X}^T \mathbf{A}^{-1} \mathbf{B}\right)^{-1} \mathbf{X}^T \mathbf{A}^{-1}$
- $(\mathbf{A} + \mathbf{BCD})^{-1} \mathbf{BC} = \mathbf{A}^{-1} \mathbf{B} (\mathbf{C}^{-1} + \mathbf{DA}^{-1} \mathbf{B})^{-1}$ , using 3 and 1 2
- $(\mathbf{A} + \mathbf{B})^{-1} = \mathbf{A}^{-1} (\mathbf{A}^{-1} + \mathbf{B}^{-1})^{-1} \mathbf{B}^{-1}$
- $(\mathbf{A}^{-1} + \mathbf{B}^{-1})^{-1} = \mathbf{A} (\mathbf{A}^{-1} + \mathbf{B}^{-1}) \mathbf{B}$

where  $\mathbf{A}$  and  $\mathbf{B}$  are square and invertible matrices.

### 4 square

$$\mathbf{x}^T \mathbf{M} \mathbf{x} - 2\mathbf{b}^T \mathbf{x} = (\mathbf{x} - \mathbf{M}^{-1} \mathbf{b})^T \mathbf{M} (\mathbf{x} - \mathbf{M}^{-1} \mathbf{b}) - \mathbf{b}^T \mathbf{M}^{-1} \mathbf{b} \quad (4)$$

### A Notation

Notation	Description
$\underline{\mathbf{X}} \in \mathbb{R}^{I_1 \times I_2 \times \dots \times I_N}$	Tensor of order $N$
$x, \mathbf{x}, \mathbf{X}$	Scalar, vector and matrix. Non-bold letters do not strictly represent scalars, in many cases their meaning should be extracted from context
$x_{i_1, i_2, \dots, i_N}, \underline{\mathbf{X}}_{(i_1, i_2, \dots, i_N)}$	$(i_1, i_2, \dots, i_N)$ th entry of $\underline{\mathbf{X}}$
$\mathbf{x}_{:, i}, \mathbf{X}_{(:, i)}$	$i$ th column of the matrix $\mathbf{X}$ . Colons are used for indexing an entire dimension.
$\mathbf{x}_{:, i_2, \dots, i_N}, \underline{\mathbf{X}}_{(:, i_2, \dots, i_N)}$	Mode-1 fiber of $\underline{\mathbf{X}}$
$\mathbf{X}_{:, :, \dots, i_N}, \underline{\mathbf{X}}_{(:, :, \dots, i_N)}$	Frontal slice of $\underline{\mathbf{X}}$
$\underline{\mathbf{X}}_{(+, :, \dots, :)}$	Partial sum-reduction over first dimension
$\underline{\mathbf{X}}_{(+)}$	Sum-reduction over all elements in the tensor
$\underline{\mathbf{1}}$	Tensor whose elements are equal to one. Their order and size is usually extracted from context
$\underline{\mathbf{A}} \odot \underline{\mathbf{B}}$	Element-wise product between tensors $\underline{\mathbf{A}}$ and $\underline{\mathbf{B}}$
$\underline{\mathbf{C}}_{(i, j)} = \underline{\mathbf{A}}_{(i, k)} \underline{\mathbf{B}}_{(k, j)}$	Tensor contraction using Einstein notation. In this case is just the matrix multiplication between $\underline{\mathbf{A}}$ and $\underline{\mathbf{B}}$

Table 1: Notation for vectors, matrices and tensors