

# Linear Algebra Review

## BIOS 526

### Definitions

1. Inner product:  $\mathbf{a}'\mathbf{b} = \sum_i a_i b_i$ .
2. Euclidean norm for vectors (length):  $\|\mathbf{a}\|_2 = \sqrt{\mathbf{a}'\mathbf{a}}$ .
3. Vectors  $\mathbf{a}$  and  $\mathbf{b}$  are orthogonal if  $\mathbf{a}'\mathbf{b} = 0$ .
4. Linear independence: vectors  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$  are linearly independent if  $\sum_i c_i \mathbf{a}_i \neq 0$  unless  $c_i = 0$  for all  $i$ .
5. Column space of matrix  $\mathbf{A}$  = the linear space spanned by the columns of  $\mathbf{A}$ .
6. Rank of matrix  $\mathbf{A}$  = minimum( $n, k$ ) where  $n$  is the number of linearly independent columns of  $\mathbf{A}$  and  $k$  is the number of linearly independent rows of  $\mathbf{A}$ .
7. Trace:  $tr(\mathbf{A}) = \sum_i \mathbf{A}_{ii}$
8. Eigenvalue: If  $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$  where  $\mathbf{x} \neq 0$ , then  $\lambda$  is an eigenvalue of  $\mathbf{A}$  and  $\mathbf{x}$  is a corresponding eigenvector.
9. A matrix  $\mathbf{P}$  is idempotent if  $\mathbf{P}^2 = \mathbf{P}$ . A symmetric idempotent matrix is called a projection matrix.

### Properties of Matrix Operations

1. If  $\mathbf{A}$  and  $\mathbf{B}$  are invertible, then  $(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}$ .
2.  $(\mathbf{AB})' = \mathbf{B}'\mathbf{A}'$
3.  $(\mathbf{A} + \mathbf{B})' = \mathbf{A}' + \mathbf{B}'$ .

### Properties of Covariance Matrix

Let  $\mathbf{X}$  be a random vector in  $\mathbb{R}^p$ .

1. Symmetric:  $\text{cov}(\mathbf{X}) = [\text{cov}(\mathbf{X})]'$ .

2.  $\text{cov}(\mathbf{X})$  is positive semidefinite.
3.  $\text{cov}(\mathbf{X} + \mathbf{a}) = \text{cov}(\mathbf{X})$  if  $\mathbf{a}$  is a constant vector.
4.  $\text{cov}(\mathbf{A}\mathbf{X}) = \mathbf{A}\text{cov}(\mathbf{X})\mathbf{A}'$  if  $\mathbf{A}$  is a constant matrix.
5.  $\text{cov}(\mathbf{X}) = E[\mathbf{X}\mathbf{X}'] - E[\mathbf{X}](E[\mathbf{X}])'$ .
6.  $\text{cov}(\mathbf{A}\mathbf{X}, \mathbf{B}\mathbf{Y}) = \mathbf{A}\text{cov}(\mathbf{X}, \mathbf{Y})\mathbf{B}'$  if  $\mathbf{A}$  and  $\mathbf{B}$  are constant matrices.

**Properties of Multivariate Normal Distribution:**

1. If  $\mathbf{Y} \sim N_n(\boldsymbol{\mu}, \boldsymbol{\Sigma})$  and  $\mathbf{C}_{p \times n}$  is a constant matrix, then  $\mathbf{C}\mathbf{Y} \sim N_p(\mathbf{C}\boldsymbol{\mu}, \mathbf{C}\boldsymbol{\Sigma}\mathbf{C}')$ .
2. Partition  $\mathbf{Y} = [\mathbf{Y}_1, \mathbf{Y}_2]$ . The conditional distribution of  $\mathbf{Y}_1$  given  $\mathbf{Y}_2$  is
$$\mathbf{Y}_1 | \mathbf{Y}_2 = \mathbf{y}_2 \sim N_n(\boldsymbol{\mu}_1 + \boldsymbol{\Sigma}_{12}\boldsymbol{\Sigma}_{22}^{-1}(\mathbf{y}_2 - \boldsymbol{\mu}_2), \boldsymbol{\Sigma}_{11} - \boldsymbol{\Sigma}_{12}\boldsymbol{\Sigma}_{22}^{-1}\boldsymbol{\Sigma}_{21}).$$
3. If  $\mathbf{Y} \sim N_n(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ , then  $(\mathbf{Y} - \boldsymbol{\mu})'\boldsymbol{\Sigma}^{-1}(\mathbf{Y} - \boldsymbol{\mu}) \sim \chi_n^2$ .