

BI476: Biostatistics - Case Studies
Assignment 1: Math Fundamentals

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1 Calculus

1.1 For $\mathbf{y} = \mathbf{x}^T \mathbf{A} \mathbf{x}$; $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, $\mathbf{A} \in \mathbb{R}^{n \times n}$

(1) Prove that $\frac{\partial \mathbf{y}}{\partial \mathbf{x}} = \mathbf{A} \mathbf{x}$

(2) Compute $\frac{\partial \mathbf{y}}{\partial \mathbf{x}^T}$

1.2 For $\mathbf{z} = \mathbf{x}^T \mathbf{A} \mathbf{y}$, where $\mathbf{x} \in \mathbb{R}^m$, $\mathbf{y} \in \mathbb{R}^n$, $\mathbf{A} \in \mathbb{R}^{m \times n}$

(1) Compute $\frac{\partial \mathbf{z}}{\partial \mathbf{x}}$

(2) Compute $\frac{\partial \mathbf{z}}{\partial \mathbf{y}}$

1.3 Linear regression

A linear regression problem $\mathbf{y} = \mathbf{X} \beta$ can be solved either using least squares or maximum likelihood. Can you write down the two objective functions and reach the normal equations.

2 Linear algebra

2.1 Trace

Here is definition of the trace for a square matrix \mathbf{X} :

$$\text{tr}(\mathbf{X}) = |\mathbf{X}| = \sum_{i=1}^n x_{ii}$$

Let $\mathbf{A} \in \mathbb{R}^{n \times n}$ and $\mathbf{B} \in \mathbb{R}^{n \times n}$ are two square matrices, and $c \in \mathbb{R}$. Prove that:

(1) $\text{tr}(\mathbf{A} \pm \mathbf{B}) = \text{tr}(\mathbf{A}) + \text{tr}(\mathbf{B})$

(2) $\text{tr}(\mathbf{A}^T) = \text{tr}(\mathbf{A})$

- (3) $tr(cA) = ctr(A)$
- (4) $tr(AB) = tr(BA)$
- (5) $tr(AA^T) = tr(A^T A)$

2.2 Determinant

We know that the determinant of a square matrix $A \in \mathbb{R}^{n \times n}$ is defined by

$$\|A\| = \sum_{i=1}^n (-1)^{i+j} a_{ij} \|M_{ij}\|, M_{ij} = A_{-i,-j}$$

Prove that

- (1) If A is diagonal or triangular, then $\|A\| = \prod_{i=1}^n a_{ii}$

2.3 Spectral Decomposition

The spectral decomposition of a square matrix $A \in \mathbb{R}^{n \times n}$ can be defined as

$$\mathbf{A} = \mathbf{\Gamma} \mathbf{\Lambda} \mathbf{\Gamma}^T$$

where $\mathbf{\Gamma}$ are orthonormal matrix such that $\mathbf{\Gamma} \mathbf{\Gamma}^T = I$ and $\mathbf{\Lambda}$ is the diagonal matrix of the eigenvalues of \mathbf{A} . Write down the form of \mathbf{A}^m .

2.4 Eigen-decomposition

- (1) What are positive-definite (PD) matrix and positive semi-definite (PS) matrices?
- (2) Can you use eigen-decomposition to determine if a square matrix is positive definite or not?

2.5 Other Matrix Decomposition Techniques

- (1) What is QR-decomposition and Cholesky-decomposition?
- (2) Write some comments on the applications of the above matrix decomposition techniques.
- (3) `base::qr` and `base::chol` can be used to conduct the two decompositions. Give an example to illustrate the usage of the two decompositions?

3 Mathematical Optimization

Find the minima of the function $f(x) = -\sin(x) - (.25x - 2)^3$. You only write down the idea, but not the final result. If you can, you should write some R code to find the minima.

4 Probability

4.1 Sufficient statistic

- (1) What is sufficient statistic?
- (2) How to prove that a statistic is a sufficient statistic for a parameter?

4.2 Negative binomial distribution

- (1) What is negative-binomial distribution?
- (2) Can you write down the expectation and variance of the distribution?
- (3) Both Poisson and negative binomial distributions can be used to fit the counts.
What are the differences between these two distributions?

4.3 Beta-Binomial model

This is a beta-binomial, can you interpret the process?

```
y <- 13
N <- 35

a <- 1
b <- 1

theta <- seq(0, 1, len = 100)
likelihood <- theta ^ y * (1-theta)^(N-y) #dbinom(y, N
, theta)
prior <- theta ^ (a-1) * (1-theta)^(b-1) #dbeta(theta,
a, b)
posterior1 <- prior * likelihood
posterior2 <- dbeta(theta, y + a, N - y + b)

plot(theta, likelihood, ty='l')
abline(v = y/N, col="red")

par(mfrow=c(1,3))
plot(theta, prior, ty='l')
plot(theta, posterior1, ty='l')
plot(theta, posterior2, ty='l', col="blue")
```