

# CS229

## Problem Set 0

### Linear Algebra and Multivariable Calculus

Kangwei Ling @kevinkwl

February 2, 2017

#### 1. Gradient and Hessians

*Solution.*

- (a) Note that  $x^T Ax = \sum_{1 \leq i, j \leq n} A_{ij} x_i x_j$ , and since  $A$  is a symmetric matrix,  $x^T Ax = \sum_{i=1}^n A_{ii} x_i^2 + \sum_{i=1}^{n-1} \sum_{j=i+1}^n 2A_{ij} x_i x_j$ . Thus,

$$\nabla f(x) = \begin{bmatrix} \vdots \\ 1/2(2A_{ii}x_i + \sum_{1 \leq j \leq n, j \neq i} 2A_{ij}x_j) + b_i \\ \vdots \end{bmatrix} = \begin{bmatrix} \vdots \\ \sum_{1 \leq j \leq n} A_{ij}x_j + b_i \\ \vdots \end{bmatrix} = Ax + b$$

- (b) Use the **Chain Rule**.

$$\nabla f(x) = \begin{bmatrix} g'(h(x)) \cdot \frac{\partial}{\partial x_1} h(x) \\ \vdots \end{bmatrix}$$

- (c) from the answer in (a), it's easy to find out that,

$$\nabla^2 f(x) = A$$

- (d)

$$\begin{aligned} \nabla f(x) &= g'(a^T x) a \\ \nabla^2 f(x) &= g''(a^T x) a a^T \end{aligned}$$

□

#### 2. Positive definite matrices

- (a) Firstly,  $A^T = (zz^T)^T = zz^T = A$ . Then for all  $x \in \mathbb{R}^n$ ,  $x^T Ax = x^T zz^T x = (z^T x)^2 \geq 0$ .

- (b)  $A$  is PSD. Consider the linear system  $Ax = 0$ , notice that for all  $x'$ , s.t.,  $Ax' = 0$ , then  $zz^T x' = 0$ . Note that  $z^T x'$  is a scalar, and because  $z$  is non-zero,  $z^T x' = 0$ .

Therefore, the null space of  $A$  is  $\{x \in \mathbb{R}^n \mid z^T x = 0\}$ . As the null space is a hyperplane perpendicular to non-zero vector  $z$ , the dimension of  $A$  is  $n - 1$ . Thus  $\text{rank}(A) = n - (n - 1) = 1$

- (c)  $BAB^T \in \mathbb{R}^{m \times m}$ ,  $(BAB^T)^T = BA^T B^T = BAB^T$ . For all  $x \in \mathbb{R}^m$ ,  $B^T x \in \mathbb{R}^n$ , and  $x^T B = (B^T x)^T$ . Let  $y = B^T x$ ,  $y \in \mathbb{R}^n$ , then for all such  $x$ ,  $y^T A y \geq 0$ , that is,  $x^T BAB^T x \geq 0$ . Therefore,  $BAB^T$  is PSD.

### 3. Eigenvectors, eigenvalues, and the spectral theorem

- (a) Let  $I^{(i)}$  be the  $i^{\text{th}}$  column of  $I$ . Note that  $T^{-1}T = I$ ,  $T^{-1}t^{(i)} = I^{(i)}$ . That is,

$$T^{-1}t^{(i)} = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}$$

and it's also obvious that

$$\frac{1}{\lambda_i} \Lambda = \begin{bmatrix} \frac{\lambda_1}{\lambda_i} & & & & \\ & \ddots & & & \\ & & 1 & & \\ & & & \ddots & \\ & & & & \frac{\lambda_n}{\lambda_i} \end{bmatrix}$$

therefore,

$$\frac{1}{\lambda_i} \Lambda T^{-1}t^{(i)} = T^{-1}t^{(i)}$$

i.e.  $\Lambda t^{(i)} = \lambda_i t^{(i)}$

- (b) Since  $U^T U = I$ ,  $U^{-1} = U^T$ , the statement follows immediately from the result of (a).  
(c)  $A$  is PSD, then  $A = U \Lambda U^T$  by the **spectral theorem**. For each  $u^{(i)}$ , we know that  $Au^{(i)} = \lambda_i u^{(i)}$ . And

$$(t^{(i)})^T A t^{(i)} \geq 0$$

since

$$(t^{(i)})^T A t^{(i)} = \lambda_i (t^{(i)})^T t^{(i)} = \lambda_i |t^{(i)}|^2$$

$\lambda_i \geq 0$