

HWK 1  
ADVANCED MACHINE LEARNING  
DATA 442/642

**Exercise 1** (5 points)

Show that the eigenvalues of a symmetric positive matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  are all positive. (Hint: Recall that the eigenvalues of a symmetric matrix are real.)

**Exercise 2** (5 points)

Show that the determinant of an orthogonal matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is  $\pm 1$ . Next, for the rotation matrix  $\mathbf{R} \in \mathbb{R}^{2 \times 2}$ , show that the determinant equals 1.

**Exercise 3** (5 points)

Show that  $\mathbf{A}^\top \mathbf{A}$  and  $\mathbf{A} \mathbf{A}^\top$  have the same eigenvalues.

**Exercise 4** (5 points)

Show that the negative entropy function  $f(x) = x \log x$  is convex for all  $x > 0$ . (Hint: If we know that a function is twice differentiable, that is, the Hessian exists for all values in the domain of  $x$ , then the function is convex if and only if the Hessian is positive semi-definite.)

**Exercise 5** (5 points)

Show that the least-squares objective function  $f(\mathbf{x}) = \|\mathbf{A}\mathbf{x} - \mathbf{b}\|_2^2$  is convex for any invertible matrix  $\mathbf{A}$ .

**Exercise 6** (Extra credit)(5points)

Show that the spectral norm of a matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  is its largest singular value  $\sigma_1$ . This is

$$\max_{\mathbf{x}} \frac{\|\mathbf{A}\mathbf{x}\|_2}{\|\mathbf{x}\|} = \sigma_1$$