# Tutorial 2

## Fundamentals

CS/SWE 4/6TE3, CES 722/723 September 21, 2010

# Checking if Symmetric Matrix is PD or PSD by Computing its Eigenvalues

**Definition** Any number  $\lambda$  such that the equation  $Ax = \lambda x$  has a non-zero vector-solution x is called an eigenvalue (or a characteristic root) of the equation.

A symmetric matrix is PD if its eigenvalues  $\lambda_i > 0$  for all i = 1, 2, ..., n and PSD if  $\lambda_i \ge 0$ . How to calculate eigenvalues:  $Ax - \lambda x = 0 \Rightarrow (A - \lambda I)x = 0$ . Since x is non-zero, the determinant of  $(A - \lambda I)$  should vanish. Therefore all eigenvalues can be calculated as roots of the equation (which is often called the characteristic equation of A):

$$\det(A - \lambda I) = 0.$$

#### Example

Consider the Hessian matrix

$$\nabla^2 f(x) = \left(\begin{array}{rrrr} 3 & -1 & 0 \\ -1 & 3 & 0 \\ 0 & 0 & 5 \end{array}\right)$$

Computing eigenvalues

$$\det(\nabla^2 f(x) - \lambda I) = \begin{pmatrix} 3 - \lambda & -1 & 0\\ -1 & 3 - \lambda & 0\\ 0 & 0 & 5 - \lambda \end{pmatrix} = (5 - \lambda)(\lambda^2 - 6\lambda + 8) = (5 - \lambda)(\lambda - 2)(\lambda - 4) = 0.$$

Therefore, the eigenvalues are  $\lambda = 2$ ,  $\lambda = 4$  and  $\lambda = 5$ . As ll of them are strictly positive, the Hessian is positive definite (PD).

### **Properties of Convex Functions**

• if f is convex function, its sublevel set  $f(x) \leq \alpha$  is convex;

- positive multiple of convex function is convex:  $f \text{ convex}, \alpha \ge 0 \implies \alpha f \text{ convex}$
- sum of convex functions is convex:  $f_1, f_2 \text{ convex} \implies f_1 + f_2 \text{ convex}$
- pointwise maximum of convex functions is convex:  $f_1, f_2 \text{ convex} \implies \max\{f_1(x), f_2(x)\} \text{ convex}$ (corresponds to intersections of epigraphs)
- affine transformation of domain:  $f \text{ convex} \implies f(Ax+b) \text{ convex}$

### **Composition Rules**

Composite function

$$f(x) = h(g(x))$$

is convex if:

- g convex; h convex nondecreasing
- g concave; h convex nonincreasing

*Proof* (differentiable functions,  $x \in \Re$ ):

$$f'' = h''(g')^2 + g''h'$$

Examples:

- $f(x) = e^{g(x)}$  is convex if g is convex
- f(x) = 1/g(x) is convex if g is concave, positive
- $f(x) = g(x)^p$ ,  $p \ge 1$  is convex if g(x) is convex, positive