

Applied Optimization

Exercise 2 - Convex Functions

Heng Liu

Nicolas Gallego

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Hand-in instructions:

Please hand-in **only one** compressed file named after the following convention: `Exercisen-GroupMemberNames.zip`, where n is the number of the current exercise sheet. This file should contain:

- **Only** the files you changed (headers and source). It is up to you to make sure that all files that you have changed are in the zip.
- A `readme.txt` file containing a description on how you solved each exercise (use the same numbers and titles) and the encountered problems.
- Other files that are required by your `readme.txt` file. For example, if you mention some screenshot images in `readme.txt`, these images need to be submitted too.
- Submit your solutions to ILIAS before the submission deadline.

Convex Functions (6 pts)

First-order condition (1 pt)

Prove the first-order condition: A differentiable function f is convex if and only if $\text{dom} f$ is convex and

$$f(y) \geq f(x) + \nabla f(x)^T(y - x)$$

holds for all $x, y \in \text{dom} f$.

Hint: First consider $f : \mathbb{R} \rightarrow \mathbb{R}$ and the definition of the derivative. Then for $f : \mathbb{R}^n \rightarrow \mathbb{R}$ consider f constrained to a line.

Second-Order Condition (1 pt)

Show that a function f , twice differentiable, is convex if and only if $\text{dom} f$ is convex and its Hessian is positive semidefinite for all $x \in \text{dom} f$

$$\nabla^2 f \succeq 0$$

Hint: Consider also first $f : \mathbb{R} \rightarrow \mathbb{R}$ and you can also use the first order condition.

Log-sum-exp (2 pt)

Show that the function

$$f(\mathbf{x}) = \log(e^{x_1} + \dots + e^{x_n})$$

is convex on \mathbb{R}^n .

Hint: The proof is outlined in the Boyd's book p. 74. Develop by it yourself and provide the intermediate steps.

Bonus (2 pts)**Geometric mean**

Show that the geometric mean

$$f(\mathbf{x}) = (\prod_{i=1}^n x_i)^{1/n}$$

is concave on \mathbb{R}_{++}^n .

Programming Exercise: Mass Spring System (6 pts)

Consider a system of springs connecting nodes of an m by n grid with connectivity as illustrated in Figure ???. The coordinate \mathbf{x} of a node indexed a at grid (i, j) , where $a \in (0, \dots, (m+1) * (n+1) - 1)$, $i \in (0, \dots, m)$ and $j \in (0, \dots, n)$, is not fixed in \mathbb{R}^2 . Each edge of the grid has an elastic constant k and each node is subject to an elastic force proportional to the distance between the nodes. Let \mathbf{x}_a and \mathbf{x}_b denote the positions of nodes a and b respectively connected by an edge with elastic constant $k_{a,b}$. Compute the total potential energy stored in the system for a given set of positions under the following modeling assumptions: (1) ideal springs without length and (2) springs with length. Denote with $\|\mathbf{x}_a - \mathbf{x}_b\|$ the euclidean distance between \mathbf{x}_a and \mathbf{x}_b .

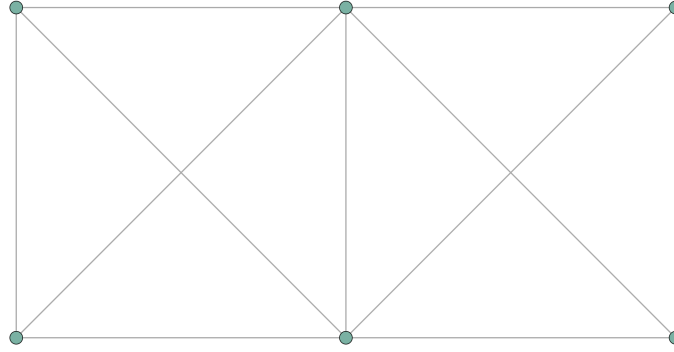


Figure 1: Example: 2×1 grid

1. Springs without length In this case magnitude of the force between this nodes will be given by:

$$F_{a,b} = k_{a,b} \|\mathbf{x}_a - \mathbf{x}_b\|,$$

and the potential energy between nodes

$$E_{a,b} = \frac{1}{2} k_{a,b} \|\mathbf{x}_a - \mathbf{x}_b\|^2,$$

Thus the total energy in the system is the sum of the energies of the individual edges $e(a,b)$

$$E = \sum_{e(a,b)} E_{a,b}$$

2. Springs with length Now let the rest length of the spring be $l_{i,j}$. The magnitude of the force will be in this case

$$\hat{F}_{a,b} = k_{a,b} (\|\mathbf{x}_a - \mathbf{x}_b\| - l_{a,b}),$$

and the potential energy between nodes

$$\hat{E}_{a,b} = \frac{1}{2} k_{a,b} (\|\mathbf{x}_a - \mathbf{x}_b\|^2 - l_{a,b}^2)^2,$$

Thus the total energy in the system is the sum of the energies of the individual edges $e(i,j)$

$$\hat{E} = \sum_{e(a,b)} \hat{E}_{a,b}$$

In this exercise, you are request to check if these two energy functions E and \hat{E} are convex with the second-order condition. The first step is to implement the `setup_spring_graph()` function in `MassSpringSystem.cc` to construct the spring graph. You can make use of the data structure provided in `SpringGraph.hh`. Make sure the spring graph is correctly setup.

The system can be viewed as a combination of elements, with each spring being an element. In `SpringElement2D.hh` and `SpringElement2DWithLength.hh`, with the energy functions $E_{a,b}$ and $\hat{E}_{a,b}$ given above, the functions `f(...)`, `grad_f(...)` and `hess_f(...)` should be implemented accordingly. Then the local data of each element is assembled into the `MassSpringProblem`. There are two implementations of this problem, one with dense hessian matrix and the other with sparse hessian matrix. You should fill out the corresponding functions in the two files `MassSpringProblem2D.hh` and `MassSpringProblem2DSparse.hh`. From what we've learned so far, one natural way to store the hessian matrix is to use eigen dense matrix. With the `eigenvalues()` function shipped with dense matrix type, you can easily check if the hessian matrix is positive semi-definite by examine the smallest eigenvalue. However, since the hessian matrix of the energy function is sparse, a smarter way is to employ the eigen sparse matrix type. Together with the `spectra` library, the eigenvalue calculation would be much more efficient.

In the end, you should be able to implement the `is_convex()` function in `MassSpringSystem.cc` with the `MassSpringProblem` classes. Regarding the elastic constant k and the length l , for simplicity all k is set to 1 and the diagonal edge length l is set to $\sqrt{2}$ and the rest is 1.