

## Weekly Exercises 2

Room: 02.09.023

Wed, 29.04.2015, 14:00-16:00

Submission deadline: Tue, 28.04.2015, 23:59 to windheus@in.tum.de

### Mathematics: Metric Spaces

**Exercise 1** (From the lecture, One point). *Show that...*

1. ... every isometry is an homeomorphism.
2. ... every surjective, distance-preserving map is injective (and thus an isometry).
3. ... being isometric is an equivalence relation.
4. ...  $f : X \rightarrow \mathbb{R}, f(x) \mapsto \text{dist}_X(x, S)$  is a nonexpanding function, where  $S$  is some fixed subset of  $X$ .

Let  $f : X \rightarrow Y, g : Y \rightarrow Z$  be bi-Lipschitz homeomorphisms. *Show that...*

5. ...  $g \circ f : X \rightarrow Z$  is a bi-Lipschitz homeomorphism.
6. ...  $\text{dil}(g \circ f) \leq \text{dil}(f) \cdot \text{dil}(g)$
7. ... the Lipschitz distance is a metric.

**Exercise 2** (Continuity, One point). *Let  $X \subset \mathbb{R}^2, Y \subset \mathbb{R}^3$ . Find metrics  $d_X, d_Y$ , such that...*

1. ... every function  $f : X \rightarrow Y$  is continuous.
2. ... the only continuous functions  $f : X \rightarrow Y$  are the constant functions.

*Show that...*

3. ... every Lipschitz function is continuous.
4. ... there exist continuous functions that are not Lipschitz.

# Programming: Farthest Point Sampling

**Exercise 3** (One point). Download and expand the file `exercise2.zip` from the lecture website. Modify the files `euclideanfps.m`, `euclideanvoronoi.m` to implement the functions as explained below. You can run the script `exercise.m` to test and visualize your solutions.

1. Given a triangle mesh  $\mathcal{M} = (\mathcal{V}, \mathcal{F})$ , a metric  $d : \mathcal{V} \times \mathcal{V} \rightarrow \mathbb{R}_{\geq 0}$ , a number  $K \in \mathbb{N}$  and an initial vertex  $v_1 \in \mathcal{V}$ , the farthest-point-sampling method computes a set of  $K$  vertices by the following algorithm:

```
for  $i = 2$  to  $K$  do  
     $v_i \leftarrow \arg \max_{v \in \mathcal{V}} (\min_{j \in \{1, \dots, i-1\}} d(v, v_j))$   
end for  
return  $\{v_1, \dots, v_K\}$ 
```

Implement function `euclideanfps` that computes the farthest point sampling of a mesh with respect to the euclidean metric. The resulting set of vertices should be returned as a vector of indices.

2. Given a triangle mesh  $\mathcal{M} = (\mathcal{V}, \mathcal{F})$  and a metric  $d : \mathcal{V} \times \mathcal{V} \rightarrow \mathbb{R}_{\geq 0}$ , the Voronoi cells with respect to a set of  $K \in \mathbb{N}$  vertices  $\{v_1, \dots, v_K\} \subset \mathcal{V}$  are defined as the partition  $\mathcal{V}_1 \dot{\cup} \dots \dot{\cup} \mathcal{V}_K \dot{\cup} \mathcal{B} = \mathcal{V}$  such that

$$v \in \mathcal{V}_i \iff (\forall j \neq i : d(v, v_i) < d(v, v_j)),$$

for all  $i \in \{1, \dots, K\}, v \in \mathcal{V}$ .

If we assume  $\mathcal{B} = \emptyset$  the partition into Voronoi cells can be represented by a function  $f : \mathcal{V} \rightarrow \{1, \dots, K\}$ . Implement function `euclideanvoronoi` that computes this partition of vertices into Voronoi cells with respect to the euclidean metric. (If you encounter vertices lying on the boundary  $\mathcal{B}$ , just pick the index of one of the neighbouring cells.)

**Exercise 4** (One point). Modify the matlab files `metricfps.m`, `metricvoronoi.m`, `distortion.m` and `ghdistance.m` to implement the functions as explained below. You can run the script `exercise.m` to test and visualize your solutions.

1. Implement function `metricfps` that computes the farthest point sampling of a mesh with respect to a metric given by matrix  $D \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$ . The resulting set of vertices should be returned as a vector of indices.
2. Implement function `metricvoronoi` that computes the partition of vertices into Voronoi cells with respect to a metric given by matrix  $D \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$ .
3. Now we want to match two shapes, i.e. two set of vertices  $\mathcal{V}_1, \mathcal{V}_2$ , both equipped with a metric represented by  $D_1 \in \mathbb{R}^{|\mathcal{V}_1| \times |\mathcal{V}_1|}, D_2 \in \mathbb{R}^{|\mathcal{V}_2| \times |\mathcal{V}_2|}$ . To make the task computationally tractable we will use farthest point sampling to obtain two smaller subsets  $S_1 \subset \mathcal{V}_1, S_2 \subset \mathcal{V}_2$ , such that  $|S_1| = |S_2|$ . We will represent the metric  $D_1$  restricted to  $S_1$  by matrix  $\hat{D}_1 \in \mathbb{R}^{|S_1| \times |S_1|}$ . Analogically define

$\hat{D}_2 \in \mathbb{R}^{|S_2| \times |S_2|}$ . A surjective correspondence can be represented by function  $f : \{1, \dots, |S_1|\} \rightarrow \{1, \dots, |S_2|\}$ , where  $f(i) = j$  means that vertex  $v_i \in S_1$  is set into correspondence with  $v_j \in S_2$ .

Implement function `distortion` that computes the distortion of given function  $f : \{1, \dots, |S_1|\} \rightarrow \{1, \dots, |S_2|\}$  with respect to metrics  $\hat{D}_1 \in \mathbb{R}^{|S_1| \times |S_1|}$ ,  $\hat{D}_2 \in \mathbb{R}^{|S_2| \times |S_2|}$ . Recall that the distortion is defined by

$$\text{dis}(f) = \sup_{v, w \in S_1} |\hat{D}_1(v, w) - \hat{D}_2(f(v), f(w))|.$$

4. Implement function `ghdistance` that computes the Gromov-Hausdorff distance between the two sets  $S_1, S_2$  and also returns the optimal permutation  $p^*$ . You can use exhaustive search over the space of all permutations  $p : \{1, \dots, |S_1|\} \rightarrow \{1, \dots, |S_2|\}$ .