

# Simplicial Complexes for Topological Data Analysis

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# 1 Simplicial complexes

## 1.1 Simplices [1]

Before defining what a **simplex** is, we need to discuss the elements which are used to construct them. For this purpose, we define a subset of  $R^n$ ,  $V = \{v_0, \dots, v_n\}$  such that the vectors

$$\{v_i - v_0 \mid v_i \in V \wedge v_i \neq v_0\}$$

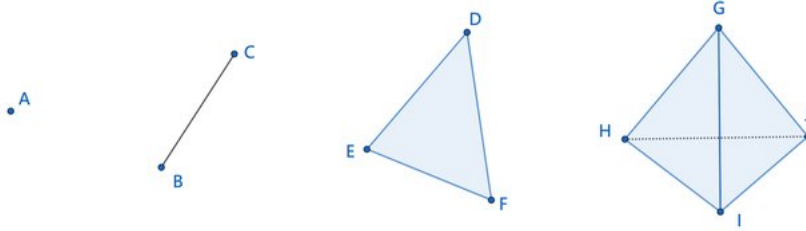
are linearly independent. Then we call  $V$  a **vertex set** with **geometrically independent** elements  $v_i \in V$ .

We also need to define the **convex hull**  $\text{conv}(X)$  of a subset  $X$  of a real or complex vector space  $S$ .

$$\text{conv}(X) := \bigcap_{\substack{X \subseteq K \subseteq S \\ K \text{ convex}}} K$$

with a set  $K$  being **convex** if, given any two points in  $K$ , their direct line segment is contained in  $K$ .

Now we define a (geometric) **n-simplex**  $\theta$  as the **convex hull** of a **geometrically independent** vertex set  $V$  with  $n + 1$  elements. Some examples can be seen in *Figure 1*.



**Figure 1** – From left to right: 0-simplex, 1-simplex, 2-simplex, 3-simplex. [5]

**Definition:** Let  $\{v_0, \dots, v_n\}$  be a **geometrically independent** set in  $R^n$ . We define the **n-simplex**  $\theta$  spanned by  $v_0, \dots, v_n$  to be the set of all points  $x \in R^n$  such that

$$x = \sum_{i=0}^n t_i v_i, \text{ where } \sum_{i=0}^n t_i = 1 \text{ and } t_i \geq 0$$

The numbers  $t_i$  are called the **barycentric coordinates** of the point  $x$  with respect to  $v_0, \dots, v_n$  (denoted as  $t_i(x)$ ). They can be uniquely determined by  $x$ . Given this definition, we describe the simplices spanned by subsets of  $V$  as

**faces**. For example, the 1-simplex in *Figure 1* has 3 **faces**: the 1-simplex itself and the two vertices it consists of.

The **interior**  $Int(\theta)$  of a **complex**  $\theta$  is defined as

$$Int(\theta) := \{x \in \theta \mid t_i(x) > 0\}$$

Intuitively, they are the points of a simplex, excluding its boundary. Looking at the 2-simplex in *Figure 1*, the **interior** is the filled out, light blue area on the inside of the triangle. The 1-simplices between the three vertices are not part of the **interior**.

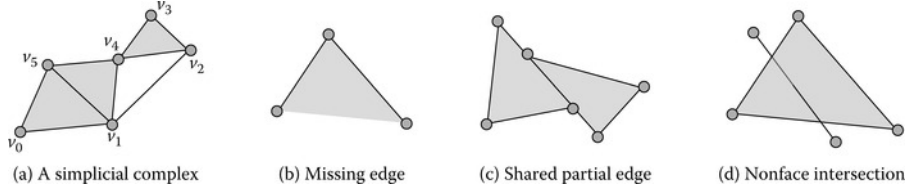
## 1.2 Simplicial Complexes [1]

The goal is to represent a space through the usage of **simplices**. Therefore we need a way of combining multiple **simplices**, to construct more complex structures.

**Definition:** Given a collection  $K$  of **simplices** in  $R^n$ , two properties have to be fulfilled:

- Any **face** of a **simplex** of  $K$  is a **simplex** of  $K$ .
- An intersection of two **simplices** must be empty or a common **face** of both.

Then we call  $K$  a **simplicial complex**.



**Figure 2** – One valid complex (a); Three invalid complexes (b,c,d). [4]

Subcollections of a complex  $K$  that contain all faces of its elements are simplicial complexes themselves. They are called **subcomplexes** of  $K$ .

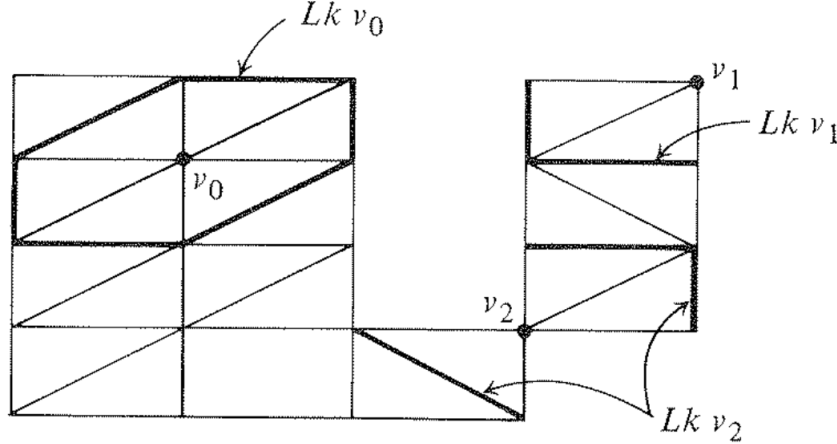
One important **subcomplex** of  $K$  is the collection of all simplices of  $K$  with dimension at most  $p$ . This is called the **p-skeleton** of  $K$  and will be denoted  $K^{(p)}$ . The most important **p-skeleton** is  $K^{(0)}$ , containing only the vertices of  $K$ .

To extend the notion of the **interior** in the context of a **complex**  $K$ , we define the following terms:

- (Star)  $St(v, K) :=$  Union of the Interiors of those simplices of  $K$  that have  $v$  as a vertex

- (Closed Star)  $\overline{St}(v, K) := \text{Union of those simplices of } K \text{ that have } v \text{ as a vertex}$
- (Link)  $Lk(v, K) := \overline{St}(v, K) - St(v, K)$

Intuitively, these three definitions describe the environment of a vertex  $v$  in a complex  $K$ . Some examples are visualized in Figure 3.



**Figure 3** – Simplicial complex with links of  $v_0$ ,  $v_1$  and  $v_2$  emphasized. [1]

Since we want to be able to utilize **simplicial complexes** in the context of topological data analysis, one needs to define the topology that is associated with a complex  $K$ . We call this the **underlying space**  $|K|$ . It is constructed by taking the subspace topology of the union of the contained simplices in  $R^n$ , and is fully characterized by the combinatorial description of the collection of simplices. In the following chapter we define such description.

### 1.3 Abstract Simplicial Complexes [1][2]

**Definition:** Let  $V$  be a vertex set. An **abstract simplicial complex**  $A$  consists of a subset of the power set of  $V$ . The elements of  $A$  are called **faces** and every subset of a **face** is also an element of  $A$ .

The dimension of an **abstract simplex** is defined by its cardinality - 1.

The dimension of an **abstract simplicial complex** is defined by the largest dimension of its simplices.

*Example:* Call the valid **geometric simplicial complex** from Figure 2  $K$ . The following **abstract simplicial complex** is the **combinatorial description** of  $K$ :

$$A = \{\{v_0, v_1, v_5\}, \{v_0, v_1\}, \{v_0, v_5\}, \{v_1, v_5\}, \{v_1, v_5, v_4\}, \{v_4, v_5\}, \{v_1, v_4\}, \\ \{v_2, v_3, v_4\}, \{v_3, v_4\}, \{v_2, v_3\}, \{v_2, v_4\}, \{v_1, v_2\}, \{v_0\}, \{v_1\}, \{v_2\}, \{v_3\}, \{v_4\}, \{v_5\}\}$$

It is important to note that the relationship between a (geometric) **simplicial complex** and its **combinatorial description** is not bijective. While there can only be one **abstract complex** per **geometric complex**, different embeddings of the same **abstract complex** can result in drastically different **geometric complexes**.

## 1.4 Duality of Simplicial Complexes [2]

These definitions give rise to a dual perspective on simplicial complexes, as we can handle them both as topological and combinatorial objects.

Given a (geometric) **simplicial complex**  $G$ , there is a unique **abstract simplicial complex**  $A$  which describes  $G$  combinatorially. We call  $G$  the **geometric realization** of  $A$ .

Now, if a **geometric complex**  $G$  is the **geometric realization** of an **abstract complex**  $A$ , their underlying spaces are homeomorphic.

Therefore, whenever a **geometric simplicial complex** is given, we can handle it in a purely **combinatorial** way (profiting from the computational efficiency) while not losing out on any **topological** attributes.

## 2 Nerve Theorem

Having introduced the notion of a **simplicial complex**, the question arises how one can apply these complexes in data analysis. Naturally, we can use a given dataset  $X$  as **vertex set** and construct **complexes** on top of them. One example for such **complex** is the **Čech complex**.

### 2.1 Čech Complex [2]

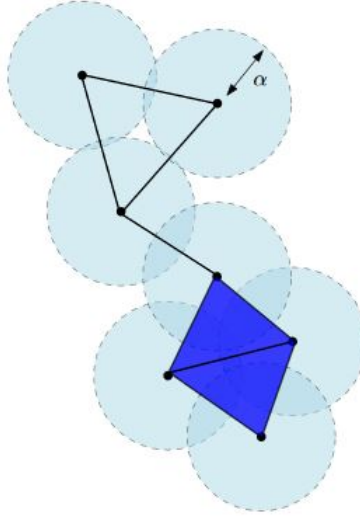
**Definition:** Let  $X$  be a set of points in  $R^n$ . The **Čech complex**  $\check{Cech}_\alpha(X)$  consists of all **n-simplices**  $\{x_0, \dots, x_n\}$  with  $x_i \in X$ , such that the  $n + 1$  closed balls  $B(x_i, \alpha)$  have a non-empty intersection.

*Example:* Given a dataset consisting of 7 points in the  $R^2$  (Figure 4), we can construct the **closed ball** around each **vertex**(data point). Whenever two or more balls intersect, a new **simplex** is generated. It has to be emphasized that only the simultaneous intersection of multiple balls give rise to higher dimensional **simplices**.

### 2.2 Nerve [2]

The **Čech complex** is a particular case of a family of complexes associated to covers of a (data) set  $X$ . Given a cover  $D = (U_i)_{i \in I}$  of  $X$ , the **nerve** of  $D$  is the abstract simplicial complex  $C(D)$  whose vertices are the  $U_i$ 's, such that

$$\theta = [U_{i_0}, \dots, U_{i_n}] \in C(D) \iff \bigcap_{j=0}^n U_{i_j} \neq \emptyset$$



**Figure 4** – A Čech complex of dimension 2. [2]

It is implied that the **nerve** can vary heavily, depending solely on the **cover** that is chosen.

*Example:* Figure 5 visualizes how a **nerve** can be constructed out of an arbitrary **cover**. In this case, the **cover**  $D$  consists of the sets  $U_1, U_2, U_3$ . These sets give rise to one **0-simplex** each, while their pairwise intersections give rise to **1-simplices**.

We want to make sure that the underlying space of a **nerve** is topologically equivalent to the space of our data. For this purpose we introduce the notion of a **homotopy equivalence**.

### 2.3 Homotopy [3]

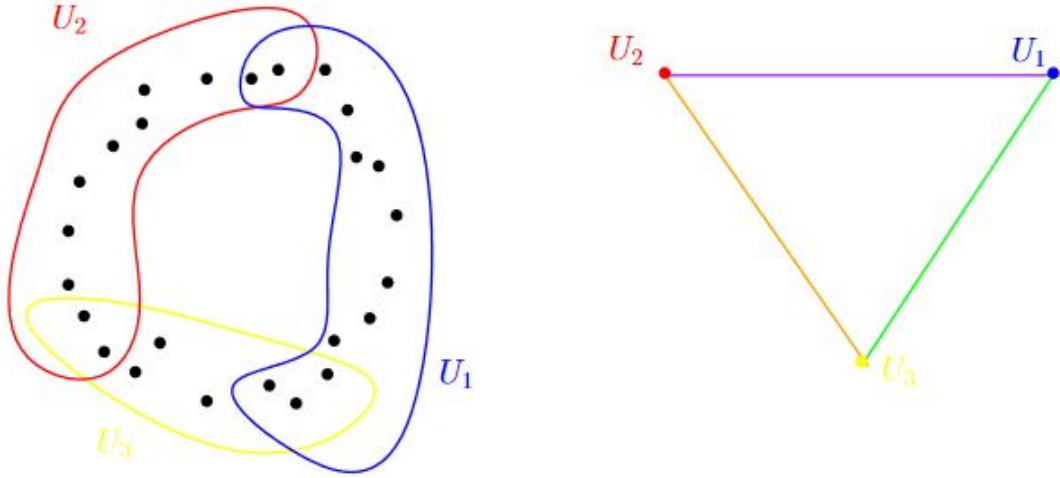
**Definition:** Given two continuous maps  $f, g : X \rightarrow Y$ , a **Homotopy** between  $f$  and  $g$  is defined as the continuous function

$$H : X \times [0, 1] \rightarrow Y \text{ with } H(x, 0) = f(x) \wedge H(x, 1) = g(x)$$

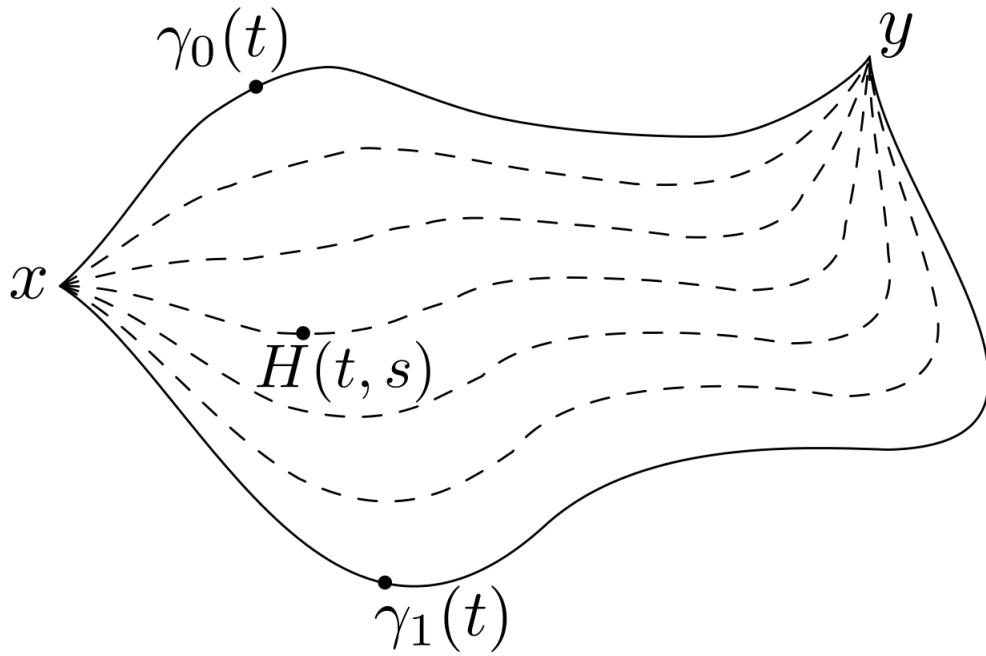
Intuitively, this means that  $f$  and  $g$  can be continuously deformed into one another. (Figure 6)

**Definition:** Two spaces  $X$  and  $Y$  are **homotopy equivalent** if there exist two continuous maps

$$f : X \rightarrow Y \text{ and } g : Y \rightarrow X$$



**Figure 5** – Generalized nerve construction. [2]



**Figure 6** – Two continuous functions  $\gamma_0(t), \gamma_1(t)$  that can be continuously deformed into each other using homotopy  $H(t, s)$ . [6]



such that  $g \circ f$  is **homotopic** to the identity map  $id_X$  and  $f \circ g$  is **homotopic** to  $id_Y$ .

One can think of such **equivalence** of spaces as a continuous squishing and stretching of one into another. This implies that **homeomorphic** spaces are always **homotopy equivalent**, while the converse doesn't apply (**homotopy equivalence** is a weaker version of equivalence). (*Figures 7, 8*)

In TDA, the **homotopy equivalence** of a space to a point in itself is especially important. We call such space **contractible**.

**Definition:** A topological space  $X \neq \emptyset$  is **contractible** if there exists a continuous map  $H : X \times [0, 1] \rightarrow X$  and a fixed point  $p \in X$  such that

$$\forall x \in X : H(x, 0) = x \wedge \forall x \in X : H(x, 1) = p$$

Intuitively, such space can be "squished" into a single point. It is implied that **contractible** spaces must not contain "holes". (Figure 9)

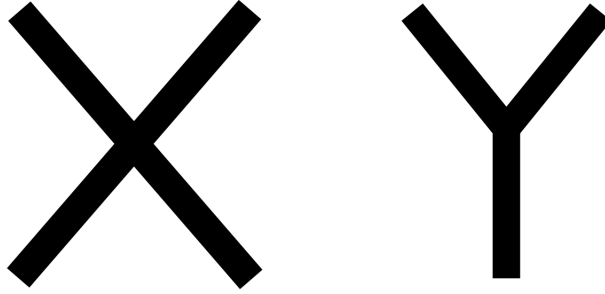
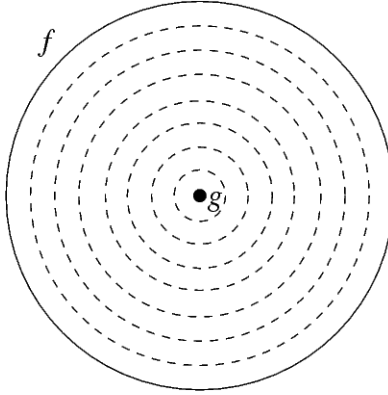


Figure 7 – Two homotopy equivalent spaces.



Figure 8 – Three spaces that are not pairwise homotopy equivalent. [3]

**Lemma:** Let  $V$  be a topological vector space over  $R$ ; let  $X \subset V$  be a **convex subset**. Then  $X$  is **contractible**.



**Figure 9 – Contractible** disc in  $R^2$ . [7]

*Proof:* Let  $x_0 \in X$ . Define continuous map  $H : X \times [0, 1] \rightarrow X$  by:

$$H(x, t) = tx_0 + (1 - t)x$$

This yields a **homotopy** between the identity map  $id_X$  and the constant map  $c_{x_0}(x) = x_0$ . Due to the assumption of **convexity** for  $X$ ,  $H$  takes values in  $X$ .  $H$  is a continuous function, since it is polynomial separately in  $x, t$  and:

$$H(x, 0) = id_X \wedge H(x, 1) = c_{x_0}(x)$$

This proves that  $H$  is a homotopy between  $id_X$  and  $c_{x_0}(x)$ . Therefore  $X$  is **contractible**.  $\square$

## 2.4 The Nerve Theorem [2]

**Theorem:** Let  $D = (U_i)_{i \in I}$  be a **cover** of a topological space  $X$  by open sets  $U_i$  such that the intersection of any subcollection of the  $U_i$ 's is either empty or contractible.

Then  $X$  and the **nerve**  $C(D)$  of  $X$  are **homotopy equivalent**.

Now, since our data is (typically) embedded in  $R^n$ , we can chose our cover  $D$  to consist of convex (and therefore contractible) subsets of  $R^n$ . It follows, that the nerve  $C(D)$  and the union of cover elements are **homotopy equivalent**.

This can be applied to the Čech complex of a dataset  $X$ , by constructing our **cover**  $D$  out of the balls  $B(x, \alpha)$  for  $\forall x \in X$ . Since the balls are **convex** subsets of  $R^n$ , the Čech complex is **homotopy equivalent** to the union of the balls.

In conclusion, we can encode the topology of a continuous space into abstract combinatorial structures for efficient computation.

### 3 Simplicial Approximation Theorem

The goal of topological data analysis is to extract information from a dataset containing (ideally) vast amounts of samples (e.g a set of images). With knowledge of **simplicial complexes** and the **nerve theorem**, one can now represent a sample (e.g an image) as a combinatorial structure, while preserving their topology. The question arises, how one can guarantee that similar samples give rise to similar **complexes**.

For this reason, we introduce an array of concepts that enable us to understand the very important **simplicial approximation theorem**. One of its use cases is the derivation of topological similarities between **complexes**.

Analogous to general topology and its introduction of similarities between spaces, we begin by defining maps between **simplicial complexes**.

#### 3.1 Simplicial Maps [1]

**Definition:** Let  $K$  and  $L$  be complexes and let  $f : K^{(0)} \rightarrow L^{(0)}$  be a map. Suppose that whenever the vertices  $v_0, \dots, v_n$  of  $K$  span a simplex of  $K$ , the points  $f(v_0), \dots, f(v_n)$  are vertices of a simplex of  $L$ . Then  $f$  can be extended to a continuous map  $g : |K| \rightarrow |L|$  such that

$$x = \sum_{i=0}^n t_i v_i \implies g(x) = \sum_{i=0}^n t_i f(v_i)$$

We call  $g$  the **simplicial map** induced by the **vertex map**  $f$ . The composite of two simplicial maps is also simplicial.

*Example:* Given **simplicial complexes**  $K, L$  with a **vertex map**  $f$  (*Figure 10*), we extend  $f$  to the **simplicial map**  $g$ . Intuitively, a sense of continuity arises through the **barycentric coordinates**. Just as  $x$  lies right in the middle between **vertices**  $a$  and  $b$ , its image  $g(x)$  lies right in the middle between  $f(a)$  and  $f(b)$ .

**Definition:** Let  $h : |K| \rightarrow |L|$  be a continuous map. We say  $h$  satisfies the **star condition** with respect to  $K$  and  $L$  if for each vertex  $v$  of  $K$ , there is a vertex  $w$  of  $L$  such that

$$h(St(v, K)) \subset St(w, L)$$

*Example:* Let  $h : |K'| \rightarrow |L|$  (*Figure 11*) satisfy the **star condition**. If one chooses a vertex in  $K'$  (e.g  $k$ ), its **star** (e.g the open path between  $l$  and  $j$ ) has to lie in the **star** of a vertex in  $L$  (e.g  $St(l', L)$ ).

The condition implies a first idea of two **complexes** being "similar", by describing a **map** that always puts a **star** from  $K$  into one in  $L$ . This idea also proves to be fundamental for our requested theorem, because we can use it to "approximate" continuous maps between **simplicial complexes**.

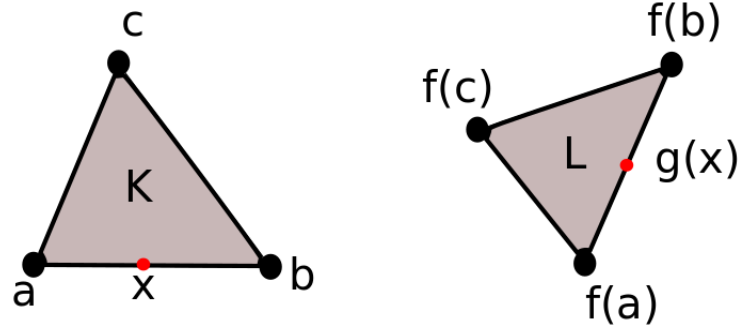


Figure 10 – Two simplicial complexes  $K, L$  with vertex map  $f$  and simplicial map  $g$ .

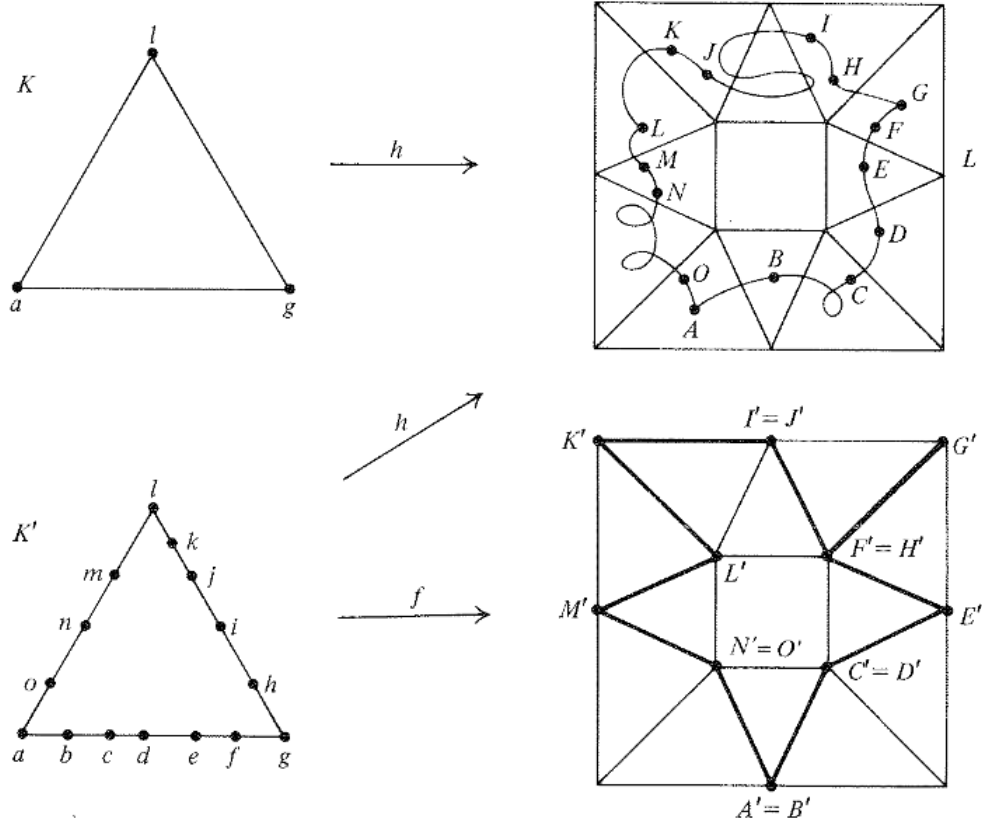


Figure 11 – Three complexes  $K, K', L$  with continuous function  $h: |K| \rightarrow |L|$  and simplicial approximation  $f: K' \rightarrow L$ .  $K, K'$  are to be seen as collections of 1-simplices, while  $L$  is a collection of 2-simplices. A capitalized vertex name in  $L$  corresponds to the image of a lowercase vertex in  $K$  or  $K'$ . (e.g  $h(d) = D$  and  $f(d) = D'$ ). [1]

### 3.2 Simplicial Approximations [1]

**Definition:** Let  $h : |K| \rightarrow |L|$  be a continuous map. If  $f : K \rightarrow L$  is a simplicial map such that

$$\forall v \in K : h(St(v, K)) \subset St(f(v), L)$$

then  $f$  is called a **simplicial approximation** to  $h$ . It also follows, that  $f$  and  $h$  are **homotopic**.

*Example:*  $f : K' \rightarrow L$  (*Figure 11*) is a **simplicial approximation** to  $h : |K'| \rightarrow |L|$ .

Intuitively, the map  $f$  is in some way "close" to  $h$ . We can formalize this in the following corollary.

**Corollary:** let  $f : K \rightarrow L$  be a simplicial approximation to  $h : |K| \rightarrow |L|$ . Given  $x \in |K|$ , there is a simplex  $\theta$  of  $L$  such that  $h(x) \in \text{Int}(\theta)$  and  $f(x) \in \theta$ .

Now one might ask, if there also exists a **simplicial approximation** to  $h : |K| \rightarrow |L|$  in *Figure 11*. Even though  $K$  and  $K'$  are rather similar, this can be disproven rather quickly. Since  $h(St(l, K))$  does not lie in any one **star** of a vertex in  $L$ , the necessary condition is not fulfilled.

Their differences in granularity are exactly the reason that only one can be used to construct an **approximation** of  $h$ . Motivated by this, we introduce the notion of a **subdivision** of a **complex**.

### 3.3 Barycentric Subdivision [1]

**Definition:** Let  $K$  be a geometric complex in  $R^n$ . A complex  $K'$  is said to be a **subdivision** of  $K$  if:

- Each simplex of  $K'$  is contained in a simplex of  $K$ .
- Each simplex of  $K$  equals the union of *finitely* many simplices of  $K'$ .

This implies, that the union of the simplices of  $K'$  equals the union of the simplices of  $K$ .  $|K|$  and  $|K'|$  are also equivalent.

It follows, that the **subdivision**  $K''$  of a **subdivision**  $K'$  is a **subdivision** of the original  $K$ .

*Example:*  $K'$  is a **subdivision** of  $K$ . (*Figure 11*)

While there are many ways of **subdividing** a **complex**, we are specifically interested in the **barycentric subdivision**.

**Definition:** Let  $\theta$  be a  $n$ -simplex with the vertices  $v_0, \dots, v_n$ . The **barycenter**  $\bar{\theta}$  is defined to be the point

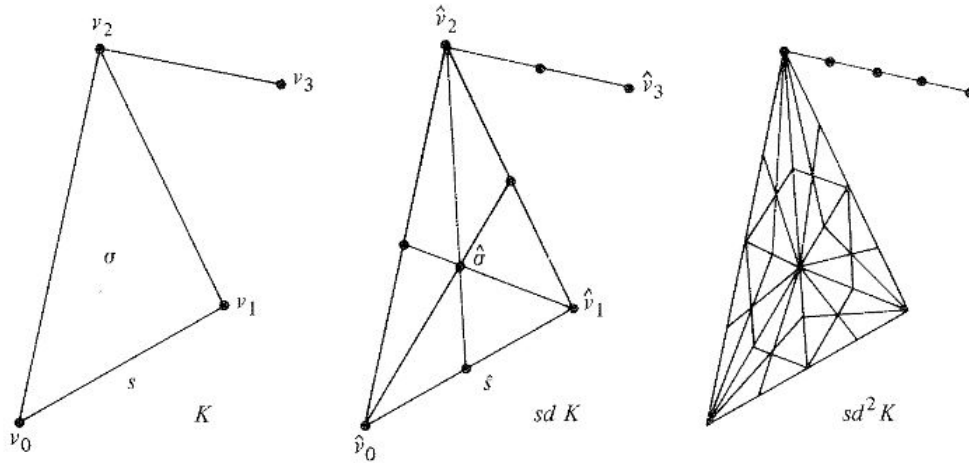
$$\bar{\theta} = \sum_{i=0}^n \frac{1}{n+1} v_i$$

It is the point of  $\text{Int}(\theta)$  all of whose barycentric coordinates with respect to the vertices of  $\theta$  are equal.

*Example:*  $\sigma$  is the **barycenter** of the **2-simplex** in  $K$ . (*Figure 12*)

**Definition** Let  $K$  be a complex. We define a sequence of subdivisions of the skeletons of  $K$  as follows: Let  $L_0 = K^{(0)}$  be the 0-skeleton of  $K$ . In general, if  $L_p$  is a subdivision of the **p-skeleton** of  $K$ , let  $L_{p+1}$  be the subdivision of the  $p+1$  skeleton obtained by starring  $L_p$  from the barycenters of the  $p+1$  simplices of  $K$ . The union of the complexes  $L_p$  is a subdivision of  $K$ . It is called the **first barycentric subdivision** of  $K$ , and denoted  $sd(K)$ . Multiple consecutive **barycentric subdivisions** are denoted as  $sd^n(K)$ , with  $n$  being the number of subdivisions.

This construct can be used to split a simplicial complex into indefinitely smaller parts. (*Figure 12*)



**Figure 12** – A complex  $K$  and its **first** and **second barycentric subdivision**.  
[1]

**Lemma:** Given a finite complex  $K$ , given a metric for  $|K|$ , and given  $\epsilon > 0$ , there is a  $n$  such that each simplex of  $sd^n(K)$  has diameter less than  $\epsilon$ .

With all of these tools, we can now define the **simplicial approximation theorem**.

### 3.4 The Simplicial Approximation Theorem [1]

**Theorem:** Let  $K$  and  $L$  be complexes; let  $K$  be finite. Given a continuous map  $h : |K| \rightarrow |L|$ , there is a  $n$  such that  $h$  has a simplicial approximation  $f : sd^n(K) \rightarrow L$ .

*Proof:* Cover  $|K|$  by the open sets  $h^{-1}(St(w, L))$ , as  $w$  ranges over the vertices of  $L$ . Now given this open covering  $D$  of the compact metric space  $K$ , there is a number  $\lambda$  such that any set of diameter less than  $\lambda$  lies in one of the elements of  $D$ ; such a number is called a **Lebesgue Number** for  $D$ . Its existence is guaranteed by the **Lebesgue number lemma**.

Choose  $n$  so that each simplex in  $sd^n(K)$  has diameter less than  $\frac{\lambda}{2}$ . Then each star of a vertex in  $sd^n(K)$  has diameter less than  $\lambda$ , so it lies in one of the sets  $h^{-1}(St(w, L))$ . Then  $h : |K| \rightarrow |L|$  satisfies the star condition relative to  $sd^n(K)$  and  $L$ , and the desired simplicial approximation exists.  $\square$

*Example:* Given two samples  $A$  and  $B$  of a dataset  $X$ , we can construct the **Čech complex** for a fixed  $\alpha$  on both samples. If there is a **continuous map**  $h : |A| \rightarrow |B|$ , we can **subdivide**  $A$  and find a **simplicial approximation**  $f : sd^n(A) \rightarrow B$ . This idea can be used to prove the **homotopy equivalence** of the **Čech complexes**.

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