# Uniform Theory of Geometric Spaces 



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To my wife Raisa and my mother Tamara.

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## Introduction

The first documented attempt to construct the geometry theory in an axiomatic way was made, as we know, by Euclid (III cent BC) in his Elements. And while the word 'geometry' literally means 'earth measuring', Euclidean geometry doesn't describe elliptic space, as more proper for measuring of our planet. New axiomatic approach was revolutionary one, however the axiomatic has limitations. Euclid study what can be constructed calculated or demonstrated starting with compass and straightedge. It was sufficient for that time. However today, despite the fact Euclidean geometry is studied in the school, many people, including geometriests, can't remember its axioms. Exception makes famous Euclid's V-th postulate, which many of us remember in the form: "At most one line can be drawn through any point not on a given line parallel to the given line in a plane". Euclid decided to formulate it so: "If a line segment intersects two straight lines forming two interior angles on the same side that sum to less than two right angles, then the two lines, if extended indefinitely, meet on that side on which the angles sum to less than two right angles".

For modern geometry Euclid's axiomatic has several limitations:

- Euclid's axiomatic theory covered the only geometry system and only two-dimensional case. The axiomatic of Euclidian geometry used today was developed by David Hilbert (1862 - 1943), has 20 axioms and covers two and three dimensions.
- The four-dimensional case uses much more axioms. Development of axiomatic for spaces of further dimensions is non-trivial.
- Except hyperbolic geometry, construction of good axiomatic for other geometries is also non-trivial. Usually, an axiomatic is constructed after the geometry is well studied with aim of some model (for example, [3] describes the space-time axiomatic).
- Undefined notions in geometry (point, line, between) differ very much from undefined notions in other mathematic disciplines (number, function, space). Undefined notions of different geometries differ from each other.
- Mathematicians successful study Euclidean space of any dimension using analytic geometry and forget Euclid's axiomatic.

Euclid's axiomatic played one important role. Its V-th postulate is so hard expressed and creates so artificial feeling that urged mathematicians to create the hyperbolic geometry. Sad, when Nikolai Lobachevsky (1792 - 1856) and János Bolyai (1802 - 1860) published their results, the new geometry was slow in acceptance. Only after decades it was demonstrated that hyperbolic geometry is interior geometry of surfaces with constant negative curvature.

After next several years some models of hyperbolic geometry were elaborated. Due to that fact the new geometry became accessible.

Author of a model, Felix Klein (1849 - 1925) proposed "Erlangen Program" [2] - the unified view over different geometries as complex of different transformation groups of space. The invariants of these groups are figures of the geometries. In such way, Klein presented 9 two-dimensional spaces. However, 6 of them he considered practic unaplicable [1]. Till now speaking about "non-euclidean geometry", elliptic or hyperbolic geometry is primarily understood. Obviously, in order to make all geometries to be taken seriously, an accessible model is required. One of such model for two-dimensional case proposed [4] Isaak Moiseevich Yaglom (1921 - 1988), using the notion of generalized complex number. Among more recent results you can refer to [5, 6, 7].

In this work, supposed to your attention an uniform model of geometric spaces and based on it general analytic geometry are described. Among its advantages there are its universality and linearity, hence easyness to use. It isn't limited to specific dimension.

The first chapter describes different types of distance and angular measure and their models. Different variants of axioms valid for different geometries are analyzed, as well as one variant of them, depending on some parameter and universally valid. A analytic model depending on some parameters is constructed. Lengths and angles are defined as parameters of corresponding motions.

In the second chapter you can find triangle equations valid for all geometries. The chapter describes generalized orthogonal matrix as general form of motion matrix. A vector approach will be shown for description of points, lines and planes, and for linear calculus of lengths and angles. At the end of chapter, the reader will find a linear way to calculate volumes.

The third chapter has more philosophical character then practical one. Your attention will be set on proper terminology and several well known spaces will be described in terms of constructed theory.

Uniform model of geometric spaces becomes the background of the GeomSpace project ${ }^{1}$.

[^0]
## Chapter 1

## Geometric Space Model Construction

### 1.1 Three Kinds of Plane Rotations. Rotation Characteristic

Consider real plane $\mathbb{R}^{2}$. Consider three different transformations of $\mathbb{R}^{2}$ : rotation $\mathfrak{R}^{\prime}(\phi)$, Galilean transformation $\mathfrak{R}^{\prime \prime}(\phi)$ and Lorintz transformation $\mathfrak{R}^{\prime \prime \prime}(\phi)$ defined by matrices:

$$
\begin{gathered}
\mathfrak{R}^{\prime}(\phi)=\left(\begin{array}{cc}
\cos \phi & -\sin \phi \\
\sin \phi & \cos \phi
\end{array}\right), \\
\mathfrak{R}^{\prime \prime}(\phi)=\left(\begin{array}{ll}
1 & 0 \\
\phi & 1
\end{array}\right)
\end{gathered}
$$

and

$$
\mathfrak{R}^{\prime \prime \prime}(\phi)=\left(\begin{array}{cc}
\cosh \phi & \sinh \phi \\
\sinh \phi & \cosh \phi
\end{array}\right),
$$

where $\phi \in \mathbb{R}$.
Transformations $\mathfrak{R}^{\prime}(\phi)$, $\mathfrak{R}^{\prime \prime}(\phi)$ and $\mathfrak{R}^{\prime \prime \prime}(\phi)$ have several common properties. The determinant of all their matrices is 1 , all them have the only fixed point - origin $O=(0,0), \mathfrak{R}(0)=I$ - unit matrix, $\mathfrak{R}(x) \mathfrak{R}(y)=\mathfrak{R}(x+y)=\mathfrak{R}(y) \mathfrak{R}(x)$ and the trajectory of point $P=(1,0)$ verifies equations:

$$
\begin{gathered}
x_{0}^{2}+x_{1}^{2}=1, \text { for } \mathfrak{R}^{\prime}, \\
x_{0}=1, \text { for } \mathfrak{R}^{\prime \prime}, \\
x_{0}^{2}-x_{1}^{2}=1, \text { for } \mathfrak{R}^{\prime \prime \prime}
\end{gathered}
$$

More general, the trajectory equation can be written as (Figure 1.1):

$$
x_{0}^{2}+k x_{1}^{2}=1, k=-1,0,1 .
$$

We will name $\mathfrak{R}^{\prime}$ elliptic rotation, $\mathfrak{R}^{\prime \prime}$ parabolic rotation, $\mathfrak{R}^{\prime \prime \prime}$ hyperbolic rotation and $\phi$ respective angle. We will name the coefficient $k$ characteristic of a rotation. $k=1$ corresponds to elliptic, $k=0$ to parabolic and $k=-1$ to hyperbolic rotation.


Figure 1.1: Trajectory of point $P$ on transformations $\mathfrak{R}^{\prime}, \mathfrak{R}^{\prime \prime}$ and $\mathfrak{R}^{\prime \prime \prime}$.

### 1.2 Functions $C(x), S(x)$ and $T(x)$

We can see, that the matrices $\mathfrak{R}^{\prime}, \mathfrak{R}^{\prime \prime}$ and $\mathfrak{R}^{\prime \prime \prime}$ have elements $r_{11}=r_{22}$ and $r_{12}=-k r_{21}$. We can write:

$$
\mathfrak{R}(\phi)=\left(\begin{array}{cc}
C(\phi) & -k S(\phi)  \tag{1.1}\\
S(\phi) & C(\phi)
\end{array}\right)
$$

where

$$
C(x)= \begin{cases}\cos x, & k=1 \\ 1, & k=0 \\ \cosh x, & k=-1\end{cases}
$$

and

$$
S(x)= \begin{cases}\sin x, & k=1 \\ x, & k=0 \\ \sinh x, & k=-1\end{cases}
$$

Finally, we can define formally the functions $C(x)$ and $S(x)$ as $^{1}$ :

$$
\begin{gather*}
C(x)=C(x, k)=\sum_{n=0}^{\infty}(-k)^{n} \frac{x^{2 n}}{(2 n)!}  \tag{1.2}\\
S(x)=S(x, k)=\sum_{n=0}^{\infty}(-k)^{n} \frac{x^{2 n+1}}{(2 n+1)!} . \tag{1.3}
\end{gather*}
$$

We will introduce one more function:

$$
\begin{equation*}
T(x)=\frac{S(x)}{C(x)} \tag{1.4}
\end{equation*}
$$

Note, that always has place the equality:

[^1]
### 1.3. REPRESENTATION OF TRANSLATION AS ROTATION. ITS CHARACTERISTICS9

$$
\begin{equation*}
C^{2}(x)+k S^{2}(x)=1, \forall x \in \mathbb{R} . \tag{1.5}
\end{equation*}
$$

We will name transformation (1.1) generalized rotation. Note, that along the angle it has one more parameter - its characteristic.

### 1.3 Representation of Translation as Rotation. Its Characteristics

Generalized rotation has a fixed point. However, the translation usually doesn't have a fixed point ${ }^{2}$. We will define the translation through generalized rotation using an extra-dimension, as it is done in projective geometry.

For $n$-dimensional space consider vector space $\mathbb{R}^{n+1}$. Rotation matrices have two different rows compared to unit matrix. When one of them is the first row, we will consider them translations. When none of them is the first one, we will consider them rotations. Therefore, the first coordinate (we will count it 0 ) will be additional.

Let $n=1$. We will name the vector $o=\{1,0\}$ origin. If $a=\mathfrak{R}(\phi) o$, then the angle $\phi$ between $o$ and $a$ we can name distance $o a$. Different values of $k$ correspond to different translations: when $k=1$ translations are elliptic, when $k=0$ they are parabolic, and when $k=-1$ they are hyperbolic.

There is a kind of distance measure for each kind of translation: elliptic, parabolic and hyperbolic. The difference between them can be seen in variants of V postulate of Euclid for elliptic, linear and hyperbolic geometry (Figure 1.2).


Figure 1.2: Variants of Euclid's V postulate - elliptic a), linear b) and hyperbolic c).

Elliptic postulate (Figure 1.2 a ) is ${ }^{3}$ : For a given line $l$ and a point $P \notin l$, exists no line $p \ni P \mid l \cap p=\varnothing$. It is identical to the following: For a given line $l$ and a point $P \notin l$, all lines $p \ni P$ intersect $l$.

The linear postulate (Figure 1.2 b ) is: For a given line $l$ and a point $P \notin l$, exists one line $p \ni P \mid l \cap p=\varnothing$.

The hyperbolic postulate (Figure 1.2 c ) is: For a given line $l$ and a point $P \notin l$, exist at least two lines $p^{\prime}, p^{\prime \prime} \ni P \mid l \cap p^{\prime}=\varnothing, l \cap p^{\prime \prime}=\varnothing$.

[^2]Generally, V postulate of Euclid can be formulated as: For a given line $l$ and a point $P \notin l$, exist $0^{k_{1}}$ lines $p \ni P \mid l \cap p=\varnothing$. It should be mentioned that $0^{k_{1}}$ is a symbol, not a number used in calculus. Its value equals to 0 for $k_{1}=1,1$ for $k_{1}=0$ and $\infty$ for $k_{1}=-1$.

### 1.4 Kinds of Space Rotations. Bundles of Unconnectable Points

It's easy to see that classic rotations in Euclidean geometry, as well as in the elliptic (Riemannian) geometry and the hyperbolic (Bolyai-Lobachevsky) geometry has the characteristic $k=1$. We can extend the notion of space rotation to generalized space rotation with some characteristic. The best way to illustrate difference between them is to formulate angular equivalent of V Postulate of Euclid - axiom of points connectability (Figure 1.3). In order to do this we will change the following phrases between them:

$$
\begin{aligned}
\text { line } l & \longleftrightarrow \text { point } L, \\
P \in l & \longleftrightarrow p \ni L, \\
P \notin l & \longleftrightarrow p \not \supset L \\
A B=\phi & \longleftrightarrow \angle a b=\phi, \\
a \cap b=C & \longleftrightarrow c=A B, \\
a \cap b=\varnothing & \longleftrightarrow A \text { is unconnectable with } B .
\end{aligned}
$$

The last statement is unusual for the above three geometries ${ }^{4}$. It makes sense in geometries with angular characteristic 0 or -1 . The unconnectable property of points is similar to parallel property of lines.


Figure 1.3: Different variants of points unconnectability axiom - elliptic a), linear b) and hyperbolic c).

The angle equivalent of V Postulate for elliptic characteristic (Figure 1.3 a ) is: On a line $l \not \supset P$ exist no points $L$ unconnectable with $P$.

For parabolic characteristic (Figure 1.3 b ) it is: On a line $l \not \supset P$ exists the only point $L$ unconnectable with $P$.

For the hyperbolic characteristic (Figure 1.3) it is: On a line $l \not \supset P$ exist at least two points $L^{\prime}$ and $L^{\prime \prime}$ unconnectable with $P$.

[^3]Generally this axiom can be formulated as: On a line $l \not \supset P$ exist $0^{k_{2}}$ points $L$ unconnectable with $P$. As in case of parallel lines, symbol $0^{k_{2}}$ isn't used in calculus.

Similar to bundles of lines - intersected, parallel or divergent we can speak about bundles of points. More exactly, let $X, Y \in \mathbb{R}^{n+1}$. All linear combinations $Z=\alpha X+\beta Y, \alpha, \beta \in \mathbb{R}$ form a set we will name bundle of points. As we will see, this set has one constraint. Therefore, it has one free parameter. As every two lines define a bundle of lines, every two points ( $X$ and $Y$ ) define bundle of points. If $X$ is connectable with $Y$ this bundle is a line (similar to intersection point of bundle of intersected lines). Lines has blue color on figure 1.3. If $X$ and $Y$ are unconnectable, this bundle of points isn't a line (similar to bundle of parallel or divergent lines). Bundles of unconnectable points are green and red on figure 1.3.

For any angle characteristic there are infinity of bundles of connectable points. For angle characteristic 1 all point bundles are lines. For angle characteristic 0 for any point there is the only bundle of unconnectable points (green). For angle characteristic -1 there are infinity bundles of unconnectable points (red). In thes case the bundles of connectable points and the bundles of unconnectable points for some point form two categories of bundles. The limit (marginal) bundles of unconnectable points (green) can be viewed as the third category (similar to differencee between parallel and divergent lines). There are exactly two limit bundles. Note that bundles of connectable points intersect all circles with centre in the centre of bundle, all bundles of unconnectable points don't intersect these circles and limit bundles are asymptotic to circles (Figure 1.4).


Figure 1.4: Mutual position of different bundles and circles a) elliptic angular characteristic, b) linear angular characteristic and c) hyperbolic angular characteristic.

Emphasize that the angle between two lines and the angle between two two-dimensional planes are the different measures. The angle between two threedimensional planes is different from them both and so on. Thus, the angle between lines can have the different characteristic then the angle between two-dimensional planes and so on.

### 1.5 Main Space Rotations

Consider $\mathbb{R}^{n+1}$ and $k_{1}, k_{2}, \ldots k_{n} \in\{-1,0,1\}$. We will note $C_{i}(x)=C\left(x, k_{i}\right), S_{i}(x)=S\left(x, k_{i}\right)$ and $T_{i}(x)=S_{i}(x) / C_{i}(x)$. Let

$$
\begin{aligned}
& \Re_{1}(\phi)=\left(\begin{array}{ccccc}
C_{1}(\phi) & -k_{1} S_{1}(\phi) & 0 & \ldots & 0 \\
S_{1}(\phi) & C_{1}(\phi) & 0 & \ldots & 0 \\
0 & 0 & 1 & \ldots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \ldots & 1
\end{array}\right), \\
& \Re_{2}(\phi)=\left(\begin{array}{cccccc}
1 & 0 & 0 & \ldots & 0 \\
0 & C_{2}(\phi) & -k_{2} S_{2}(\phi) & \ldots & 0 \\
0 & S_{2}(\phi) & C_{2}(\phi) & \ldots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \ldots & 1
\end{array}\right), \\
& \Re_{n}(\phi)=\left(\begin{array}{ccccccc}
1 & 0 & \ldots & 0 & & \\
0 & 1 & \ldots & 0 & & 0 \\
\vdots & \vdots & \ddots & \vdots & & \\
0 & 0 & \ldots & C_{n}(\phi) & -k_{n} S_{n}(\phi) \\
0 & 0 & \ldots & S_{n}(\phi) & C_{n}(\phi)
\end{array}\right) .
\end{aligned}
$$

We will name $\mathfrak{R}_{1}, \ldots \Re_{n}$ main space rotations.

### 1.6 Vector Product. Invariant Quadric Form

Let

$$
\begin{equation*}
K_{m}=\prod_{i=1}^{m} k_{i}, \forall m=\overline{0, n} \tag{1.6}
\end{equation*}
$$

We can see that $K_{m} \in\{-1,0,1\}, \forall m=\overline{0, n}$ as well as $k_{m}$. Let define vector product $\odot$ as

$$
\begin{equation*}
x \odot y=\sum_{i=0}^{n} K_{i} x_{i} y_{i} \tag{1.7}
\end{equation*}
$$

For some vectors $x=\left\{x_{0}, x_{1}, \ldots x_{n}\right\}$ and $y=\left\{y_{0}, y_{1}, \ldots y_{n}\right\}, x^{\prime}=\mathfrak{R}_{m}(\phi) x=\left\{x_{0}, \ldots x_{m-2}, x_{m-1}\right.$
$\left.C_{m}(\phi)-k_{m} x_{m} S_{m}(\phi), x_{m-1} S_{m}(\phi)+x_{m} C_{m}(\phi), x_{m+1}, \ldots x_{n}\right\}$ and $y^{\prime}=\mathfrak{R}_{m}(\phi) y$. We can see that

$$
\begin{aligned}
x^{\prime} \odot y^{\prime} & =\sum_{i=0}^{n} K_{i} x_{i}^{\prime} y_{i}^{\prime} \\
& =\sum_{i=0}^{m-2} K_{i} x_{i} y_{i} \\
& +\left(\left(x_{m-1} C_{m}(\phi)-k_{m} x_{m} S_{m}(\phi)\right)\left(y_{m-1} C_{m}(\phi)-k_{m} y_{m} S_{m}(\phi)\right)\right. \\
& \left.+k_{m}\left(x_{m-1} S_{m}(\phi)+x_{m} C_{m}(\phi)\right)\left(y_{m-1} S_{m}(\phi)+y_{m} C_{m}(\phi)\right)\right) K_{m-1} \\
& +\sum_{i=m+1}^{n} K_{i} x_{i} y_{i} \\
& =\sum_{i=0}^{m-2} K_{i} x_{i} y_{i} \\
& +\left(\left(x_{m-1} y_{m-1} C_{m}^{2}(\phi)-k_{m}\left(x_{m-1} y_{m}+x_{m} y_{m-1}\right) S_{m}(\phi) C_{m}(\phi)+k_{m}^{2} x_{m} y_{m} S_{m}^{2}(\phi)\right.\right. \\
& \left.\left.+k_{m} x_{m-1} y_{m-1} S_{m}^{2}(\phi)+k_{m}\left(x_{m-1} y_{m}+x_{m} y_{m-1}\right) S_{m}(\phi) C_{m}(\phi)+k_{m} x_{m} y_{m} C_{m}^{2}(\phi)\right)\right) K_{m-1} \\
& +\sum_{i=m+1}^{n} K_{i} x_{i} y_{i} \\
& =\sum_{i=0}^{m-2} K_{i} x_{i} y_{i} \\
& +\left(x_{m-1} y_{m-1}\left(C_{m}^{2}(\phi)+k_{m} S_{m}^{2}(\phi)\right)+k_{m} x_{m} y_{m}\left(C_{m}^{2}(\phi)+k_{m} S_{m}^{2}(\phi)\right)\right) K_{m-1} \\
& +\sum_{i=m+1}^{n} x_{i} y_{i} K_{i} \\
& =\sum_{i=0}^{n} K_{i} x_{i} y_{i}=x \odot y
\end{aligned}
$$

This is true for all $m=\overline{1, n}$. So the quadric form $x \odot y$ is invariant in respect to main rotations of $\Re_{m}$.

### 1.7 Space Definition by its Specification

Consider $\mathbb{R}^{n}$ projective space and $k_{i} \in\{-1,0,1\}, \forall i=\overline{1, n}$. We can now introduce a geometric space 'unit sphere' $\mathbb{B}^{n}=\left\{x \in \mathbb{R}^{n} \mid x \odot x=1\right\}$ (Figure 1.5). As all main rotations preserves the quadric form defined by product $\odot$, they also preserves $\mathbb{B}^{n}$. We will name $k_{i}, i=\overline{1, n}$ space specification. We will name 'point' $X \in \mathbb{B}^{n}$ the corresponding vector $x \in \mathbb{R} \mathbb{P}^{n}$ and will use homogeneous coordinates normalized in order to $x \odot x=1$.

We will name 'origin' of $\mathbb{B}^{n}$ the point $O=[1: 0: \ldots: 0] \in \mathbb{B}^{n}$. It isn't origin of $\mathbb{R}^{n+1}$, $(0,0, \ldots 0) \notin \mathbb{B}^{n}$ and we will refer to $O$ as origin if isn't specified otherwise.

It's easy to see that for any $k_{1}, k_{2}, \ldots k_{n}, O=\mathbb{B}^{0} \subset \mathbb{B}^{1} \subset \ldots \subset \mathbb{B}^{n}$.
We will define motions of $\mathbb{B}^{n}$ all transformations that result on finite product of main rotations.


Figure 1.5: Sphere of space with specification $\{-1,-1\}$.

We will define 'lines' all images of $\mathbb{B}^{1}$ on any motion of $\mathbb{B}^{n}$. Similarly, we define ' $m$ dimensional' planes all images of $\mathbb{B}^{m}$ on any motions of $\mathbb{B}^{n}$ for any $m \in \overline{0, n-1}$.

For each characteristic parameter $k_{i}$ we can introduce a scale parameter $r_{i} \in \mathbb{R}_{+}, i=\overline{1, n}$. The $k_{1} / r_{1}^{2}$ is exactly the gaussian curvature of space. Others have no representation since finite angle measure doesn't require scaling. In this case the radian measure is native. An example of angle scale is degree measure which has scale $180 / \pi$. However when the angle is not bounded a scale introduction has sense. All scales can be easy embedded in functions $C_{i}(x), S_{i}(x)$ and $T_{i}(x)$ by using instead $C_{i}\left(\frac{x}{r_{i}}\right), S_{i}\left(\frac{x}{r_{i}}\right)$ and $T_{i}\left(\frac{x}{r_{i}}\right)$ respectively, $i=\overline{1, n}$.

### 1.8 Definition of Measure Using Motions

A traditional way of definition the measures and motions is to provide a way to calculate the distances as is and then to define motions in such way that all maps $M: \mathbb{R}^{n} \rightarrow \mathbb{R}^{n}$ preserve the distance. We go another way. We provide motions as is and then search for a way to define measures in such way that motions preserve them.

We will say point $A \in \mathbb{B}^{1} \subset \mathbb{B}^{n}$ has the distance $\phi$ from origin $O$ if $A=\mathfrak{R}_{1}(\phi) O$. Having $O=[1: 0: \ldots: 0], A=\left[C_{1}(\phi): S_{1}(\phi): 0: \ldots: 0\right], O \odot A=C_{1}(\phi)$. We will say one-dimensional (planar) angle between $\mathbb{B}^{1}$ and some one-dimensional line $\mathbb{B}^{\prime 1} \subset \mathbb{B}^{2}$ equals $\phi$ if $\mathbb{B}^{11}=\mathfrak{R}_{2}(\phi) \mathbb{B}^{1}$. Similarly, we will define the $m$-dimensional angle $\phi$ between $\mathbb{B}^{m}$ and $m$-dimensional plane $\mathbb{B}^{\prime m} \subset \mathbb{B}^{m+1}$ if $\mathbb{B}^{\prime m}=\mathfrak{R}_{m+1}(\phi) \mathbb{B}^{m}, \forall m=\overline{0, m-1}$. Note, that $n$-dimensional angle between any planes is 0 since all them are subset of $B^{n}$.

Let $X, Y \in \mathbb{B}^{n}$. If there exists a motion that maps $\mathbb{B}^{1}$ to $X Y$ we will name points $X$ and $Y$ connectable and distance $X Y$ measurable. If not, we will name points $X$ and $Y$ unconnectable (just as lines can be parallel) and strictly speaking the distance $X Y$ doesn't exists ${ }^{5}$.

We can find a motion $\mathfrak{M}$ of space $\mathbb{B}^{n}$ that maps origin $O$ to $X$ and some point $A \in \mathbb{B}^{1} \subset \mathbb{B}^{n}$ to $Y$. As motion $\mathfrak{M}$ preserves the quadric form $\odot$, we can see that $X \odot Y=O \odot A$. We can define the distance $\phi$ between $X$ and $Y$ as

$$
\begin{equation*}
C_{1}(\phi)=X \odot Y . \tag{1.8}
\end{equation*}
$$

It's easy to see that all motions preserve the distance. In case of elliptic, Euclidian and hyperbolic space it is sufficient, because all other measures can be calculated from distances. However, in some spaces angles can be scaled in a manner distances are scaled in Euclidean space. So we should find the way to measure all the measures in general case.

[^4]
## Chapter 2

## Measure Calculus

### 2.1 General Triangle Equations

Consider triangle $A B C \in \mathbb{B}^{2}$ with the edges $a, b, c$, interior angles $\alpha, \gamma$ and exterior angle $\beta^{\prime}$ (Figure 2.1). Let $A=[1: 0: 0]$ the origin, $C=\mathfrak{R}_{1}(b) A=\left[C_{1}(b): S_{1}(b): 0\right]$ and $B=$ $\mathfrak{R}_{2}(\alpha) \mathfrak{R}_{1}(c) A=\left[C_{1}(c): S_{1}(c) C_{2}(\alpha): S_{1}(c) S_{2}(\alpha)\right]$. Note, that the interior angle $\beta$ does not exist in case of $k_{2}=0$ or $k_{2}=-1$. The exterior angle $\beta^{\prime}$ always exists.


Figure 2.1: General triangle equations deduction.

Now, let $A^{\prime} B^{\prime} C^{\prime}=\mathfrak{R}(-b)(A B C)$ (Figure 2.1, cyan). The point we are interested in is $B^{\prime}=\left[C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha):-S_{1}(b) C_{1}(c)+C_{1}(b) S_{1}(c) C_{2}(\alpha): S_{1}(c) S_{2}(\alpha)\right]$. At the other hand, now $B^{\prime}=\left[C_{1}(a):-S_{1}(a) C_{2}(\gamma): S_{1}(a) S_{2}(\gamma)\right]$. It means:

$$
\begin{aligned}
C_{1}(a) & =C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha), \\
-S_{1}(a) C_{2}(\gamma) & =-S_{1}(b) C_{1}(c)+C_{1}(b) S_{1}(c) C_{2}(\alpha), \\
S_{1}(a) S_{2}(\gamma) & =S_{1}(c) S_{2}(\alpha) .
\end{aligned}
$$

The first equation is the form of the Cosine I law. Similarly we have

$$
C_{1}(c)=C_{1}(a) C_{1}(b)+k_{1} S_{1}(a) S_{1}(b) C_{2}(\gamma) .
$$

The third equation is equivalent to

$$
\frac{S_{1}(a)}{S_{2}(\alpha)}=\frac{S_{1}(c)}{S_{2}(\gamma)}
$$

which is the form of the Sine law.
Let now $A^{\prime \prime} B^{\prime \prime} C^{\prime \prime}=\mathfrak{R}_{1}(-c) \mathfrak{R}_{2}(-\alpha)(A B C)$ (Figure 2.1, brown). Now we are interested in vertex $C^{\prime \prime}=\left[C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha):-C_{1}(b) S_{1}(c)+S_{1}(b) C_{1}(c) C_{2}(\alpha):-S_{1}(b) S_{2}(\alpha)\right]$. At the other hand, $C^{\prime \prime}=\left[C_{1}(a): S_{1}(a) C_{2}\left(\beta^{\prime}\right):-S_{1}(a) S_{2}\left(\beta^{\prime}\right)\right]$. From here we have:

$$
\begin{aligned}
C_{1}(a) & =C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha), \\
S_{1}(a) C_{2}\left(\beta^{\prime}\right) & =-C_{1}(b) S_{1}(c)+S_{1}(b) C_{1}(c) C_{2}(\alpha), \\
-S_{1}(a) S_{2}\left(\beta^{\prime}\right) & =-S_{1}(b) S_{2}(\alpha) .
\end{aligned}
$$

The first one is the Cosine I law, the third one is equivalent to:

$$
\begin{equation*}
\frac{S_{1}(a)}{S_{2}(\alpha)}=\frac{S_{1}(b)}{S_{2}\left(\beta^{\prime}\right)}=\frac{S_{1}(c)}{S_{2}(\gamma)} \tag{2.1}
\end{equation*}
$$

which is the Sine law. Note that in case $k_{2}=1$ we have $\beta=\pi-\beta^{\prime}, S_{2}(\beta)=S_{2}\left(\beta^{\prime}\right)$. Let calculate the value of $C_{2}(\alpha)$ from the first equation and put it to the second one:

$$
\begin{aligned}
S_{1}(a) C_{2}\left(\beta^{\prime}\right) & =-C_{1}(b) S_{1}(c)+S_{1}(b) C_{1}(c) \frac{C_{1}(a)-C_{1}(b) C_{1}(c)}{k_{1} S_{1}(b) S_{1}(c)} \\
& =-C_{1}(b) S_{1}(c)+C_{1}(c) \frac{C_{1}(a)-C_{1}(b) C_{1}(c)}{k_{1} S_{1}(c)}, \\
k_{1} S_{1}(a) S_{1}(c) C_{2}\left(\beta^{\prime}\right) & =-k_{1} S_{1}(c)^{2} C_{1}(b)+C_{1}(a) C_{1}(c)-C_{1}(b) C_{1}(c)^{2} \\
& =C_{1}(a) C_{1}(c)-C_{1}(b)\left(C_{1}(c)^{2}+k_{1} S_{1}(c)^{2}\right) \\
& =C_{1}(a) C_{1}(c)-C_{1}(b) \\
C_{1}(b) & =C_{1}(a) C_{1}(c)-k_{1} S_{1}(a) S_{1}(c) C_{2}\left(\beta^{\prime}\right)
\end{aligned}
$$

Note the ' - ' sign in the right part of the equation. It is so because the $\beta^{\prime}$ angle is external. For the case $k_{2}=1$, the internal angle $\beta=\pi-\beta^{\prime}, C_{2}(\beta)=-C_{2}\left(\beta^{\prime}\right)$.

What about the Cosine II law? We will use these two equations:

$$
\begin{aligned}
-S_{1}(a) C_{2}(\gamma) & =-S_{1}(b) C_{1}(c)+C_{1}(b) S_{1}(c) C_{2}(\alpha) \\
S_{1}(a) C_{2}\left(\beta^{\prime}\right) & =-C_{1}(b) S_{1}(c)+S_{1}(b) C_{1}(c) C_{2}(\alpha)
\end{aligned}
$$

First, replace $S_{1}(b)$ with $S_{1}(a) S_{2}\left(\beta^{\prime}\right) / S_{2}(\alpha)$ and $S_{1}(c)$ with $S_{1}(a) S_{2}(\gamma) / S_{2}(\alpha)$ :

$$
\begin{aligned}
-S_{1}(a) C_{2}(\gamma) & =-S_{1}(a) \frac{S_{2}(\beta)}{S_{2}(\alpha)} C_{1}(c)+C_{1}(b) S_{1}(a) \frac{S_{2}(\gamma)}{S_{2}(\alpha)} C_{2}(\alpha), \\
-S_{2}(\alpha) C_{2}(\gamma) & =-C_{1}(c) S_{2}\left(\beta^{\prime}\right)+C_{1}(b) S_{2}(\gamma) C_{2}(\alpha), \\
S_{2}\left(\beta^{\prime}\right) C_{1}(c) & =S_{2}(\alpha) C_{2}(\gamma)+C_{2}(\alpha) S_{2}(\gamma) C_{1}(b),
\end{aligned}
$$

and

$$
\begin{aligned}
S_{1}(a) C_{2}\left(\beta^{\prime}\right) & =-C_{1}(b) S_{1}(a) \frac{S_{2}(\gamma)}{S_{2}(\alpha)}+S_{1}(a) \frac{S_{2}\left(\beta^{\prime}\right)}{S_{2}(\alpha)} C_{1}(c) C_{2}(\alpha), \\
S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right) & =-C_{1}(b) S_{2}(\gamma)+C_{1}(c) S_{2}\left(\beta^{\prime}\right) C_{2}(\alpha) \\
S_{2}(\gamma) C_{1}(b) & =-S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)+C_{2}(\alpha) S_{2}\left(\beta^{\prime}\right) C_{1}(c)
\end{aligned}
$$

Now from the first equation let calculate $C_{1}(c)$ and put it in the second one:

$$
\begin{aligned}
S_{2}(\gamma) C_{1}(b) & =-S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)+C_{2}(\alpha) S_{2}\left(\beta^{\prime}\right) \frac{S_{2}(\alpha) C_{2}(\gamma)+C_{2}(\alpha) S_{2}(\gamma) C_{1}(b)}{S_{2}\left(\beta^{\prime}\right)} \\
& =-S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)+C_{2}(\alpha) S_{2}(\alpha) C_{2}(\gamma)+C_{2}(\alpha)^{2} S_{2}(\gamma) C_{1}(b), \\
S_{2}(\gamma) C_{1}(b)\left(1-C_{2}(\alpha)^{2}\right) & =S_{2}(\alpha)\left(C_{2}(\alpha) C_{2}(\gamma)-C_{2}\left(\beta^{\prime}\right)\right), \\
k_{2} S_{2}(\gamma) C_{1}(b) S_{2}(\alpha)^{2} & =S_{2}(\alpha)\left(C_{2}(\alpha) C_{2}(\gamma)-C_{2}(\beta)\right), \\
k_{2} S_{2}(\alpha) S_{2}(\gamma) C_{1}(b) & =C_{2}(\alpha) C_{2}(\gamma)-C_{2}\left(\beta^{\prime}\right), \\
C_{2}\left(\beta^{\prime}\right) & =C_{2}(\alpha) C_{2}(\gamma)-k_{2} S_{2}(\alpha) S_{2}(\gamma) C_{1}(b) .
\end{aligned}
$$

When $k_{2}=1$ we have

$$
\begin{aligned}
-\cos \beta & =\cos \alpha \cos \gamma-k_{2} \sin \alpha \sin \gamma C_{1}(b) \\
\cos \beta & =-\cos \alpha \cos \gamma+k_{2} \sin \alpha \sin \gamma C_{1}(b)
\end{aligned}
$$

Similarly, calculating $C_{1}(b)$ form the second equation and putting it in the first one, obtain:

$$
\begin{aligned}
S_{2}\left(\beta^{\prime}\right) C_{1}(c) & =S_{2}(\alpha) C_{2}(\gamma)+C_{2}(\alpha) S_{2}(\gamma) \frac{C_{2}(\alpha) S_{2}\left(\beta^{\prime}\right) C_{1}(c)-S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)}{S_{2}(\gamma)} \\
& =S_{2}(\alpha) C_{2}(\gamma)+C_{2}(\alpha)^{2} S_{2}\left(\beta^{\prime}\right) C_{1}(c)-C_{2}(\alpha) S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right), \\
S_{2}\left(\beta^{\prime}\right) C_{1}(c)\left(1-C_{2}(\alpha)^{2}\right) & =S_{2}(\alpha)\left(C_{2}(\gamma)-C_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)\right), \\
k_{2} S_{2}\left(\beta^{\prime}\right) C_{1}(c) S_{2}(\alpha)^{2} & =S_{2}(\alpha)\left(C_{2}(\gamma)-C_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)\right), \\
k_{2} S_{2}(\alpha) S_{2}\left(\beta^{\prime}\right) C_{1}(c) & =C_{2}(\gamma)-C_{2}(\alpha) C_{2}\left(\beta^{\prime}\right), \\
C_{2}(\gamma) & =C_{2}(\alpha) C_{2}\left(\beta^{\prime}\right)+k_{2} S_{2}(\alpha) S_{2}\left(\beta^{\prime}\right) C_{1}(c) .
\end{aligned}
$$

When $k_{2}=1$ we have as above

$$
\cos \gamma=-\cos \alpha \cos \beta+k_{2} \sin \alpha \sin \beta C_{1}(c)
$$

Similarly, we have

$$
C_{2}(\alpha)=C_{2}\left(\beta^{\prime}\right) C_{2}(\gamma)+k_{2} S_{2}\left(\beta^{\prime}\right) S_{2}(\gamma) C_{1}(a)
$$

We will find the form of the Cosine I and II law that does not contain $C_{1}$ or $C_{2}$ functions in the left part. However, it contains these functions in the right part. It makes sense since in the case $k_{1} \neq 0$ (for the Cosine I law) and $k_{2} \neq 0$ (for the Cosine II law) when we can calculate their respective $C^{-1}$ functions, but when $k_{1}=0$, the space admit distance scaling and the angle values does not determine the distances (Cosine II law is a equality which doesn't contain $C_{1}$ function), while when $k_{2}=0$, the space admit the angular scaling and distances does not determine angles (Cosine I law is a equality which doesn't contain $C_{2}$ function).

Note also that we can deduce one form of Cosine I and one form of Cosine II law if we introduce a (may be virtual) angle $\beta$ so as:

$$
\begin{aligned}
S_{2}(\beta) & =S_{2}\left(\beta^{\prime}\right) \\
C_{2}(\beta) & =-C_{2}\left(\beta^{\prime}\right), \\
T_{2}(\beta) & =-T_{2}\left(\beta^{\prime}\right)
\end{aligned}
$$

Then both Cosine I and II law have identical form. Now let calculate

$$
\begin{aligned}
k_{1} S_{1}^{2}(a) & =1-C_{1}^{2}(a) \\
& =\left(C_{1}^{2}(b)+k_{1} S_{1}^{2}(b)\right)\left(C_{1}^{2}(c)+k_{1} S_{1}^{2}(c)\right) \\
& -\left(C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha)\right)^{2} \\
& =C_{1}^{2}(b) C_{1}^{2}(c)+k_{1} C_{1}^{2}(b) S_{1}^{2}(c)+k_{1} S_{1}^{2}(b) C_{1}^{2}(c)+k_{1}^{2} S_{1}^{2}(b) S_{1}^{2}(c) \\
& -C_{1}^{1}(b) C_{1}^{2}(c)-2 k_{1} C_{1}(b) C_{1}(c) S_{1}(b) S_{1}(c) C_{2}(\alpha)-k_{1}^{2} S_{1}^{2}(b) S_{1}^{2}(c) C_{2}^{2}(\alpha) \\
& =k_{1}\left(C_{1}^{2}(b) S_{1}^{2}(c)+S_{1}^{2}(b) C_{1}^{2}(c)-2 C_{1}(b) C_{1}(c) S_{1}(b) S_{1}(c) C_{2}(\alpha)\right) \\
& +k_{1}^{2} S_{1}^{2}(b) S_{1}^{2}(c)\left(1-C_{2}^{2}(\alpha)\right) \\
& =k_{1}\left(C_{1}^{2}(b) S_{1}^{2}(c)+S_{1}^{2}(b) C_{1}^{2}(c)-2 C_{1}(b) C_{1}(c) S_{1}(b) S_{1}(c) C_{2}(\alpha)\right) \\
& +k_{1}^{2} k_{2} S_{1}^{2}(b) S_{1}^{2}(c) S_{2}^{2}(\alpha), \\
S_{1}^{2}(a) & =C_{1}^{2}(b) S_{1}^{2}(c)+S_{1}^{2}(b) C_{1}^{2}(c)-2 C_{1}(b) C_{1}(c) S_{1}(b) S_{1}(c) C_{2}(\alpha) \\
& +k_{1} k_{2} S_{1}^{2}(b) S_{1}^{2}(c) S_{2}^{2}(\alpha),
\end{aligned}
$$

or, having:

$$
\begin{align*}
C_{1}(a) & =C_{1}(b) C_{1}(c)\left(1+k_{1} T_{1}(b) T_{1}(c) C_{2}(\alpha)\right) \\
T_{1}^{2}(a) & =\frac{T_{1}^{2}(b)+T_{1}^{2}(c)-2 T_{1}(b) T_{1}(c) C_{2}(\alpha)+k_{1} k_{2} T_{1}^{2}(b) T_{1}^{2}(c) S_{1}^{2}(\alpha)}{\left(1+k_{1} T_{1}(b) T_{1}(c) C_{2}(\alpha)\right)^{2}} \tag{2.2}
\end{align*}
$$

Similarly,

$$
\begin{align*}
S_{1}^{2}(b) & =C_{1}^{2}(a) S_{1}^{2}(c)+S_{1}^{2}(a) C_{1}^{2}(c)+2 C_{1}(a) C_{1}(c) S_{1}(a) S_{1}(c) C_{2}\left(\beta^{\prime}\right) \\
& +k_{1} k_{2} S_{1}^{2}(a) S_{1}^{2}(c) S_{2}^{2}\left(\beta^{\prime}\right) \\
T_{1}^{2}(b) & =\frac{T_{1}^{2}(a)+T_{1}^{2}(c)+2 T_{1}(a) T_{1}(c) C_{2}\left(\beta^{\prime}\right)+k_{1} k_{2} T_{1}^{2}(a) T_{1}^{2}(c) S_{1}^{2}\left(\beta^{\prime}\right)}{\left(1-k_{1} T_{1}(a) T_{1}(c) C_{2}\left(\beta^{\prime}\right)\right)^{2}}, \tag{2.3}
\end{align*}
$$

and

$$
\begin{align*}
S_{1}^{2}(c) & =C_{1}^{2}(a) S_{1}^{2}(b)+S_{1}^{2}(a) C_{1}^{2}(b)-2 C_{1}(a) C_{1}(b) S_{1}(a) S_{1}(b) C_{2}(\gamma) \\
& +k_{1} k_{2} S_{1}^{2}(a) S_{1}^{2}(b) S_{2}^{2}(\gamma) \\
T_{1}^{2}(c) & =\frac{T_{1}^{2}(a)+T_{1}^{2}(b)-2 T_{1}(a) T_{1}(b) C_{2}(\gamma)+k_{1} k_{2} T_{1}^{2}(a) T_{1}^{2}(b) S_{1}^{2}(\gamma)}{\left(1+k_{1} T_{1}(a) T_{1}(b) C_{2}(\gamma)\right)^{2}} \tag{2.4}
\end{align*}
$$

Now, let calculate

$$
\begin{aligned}
k_{2} S_{2}^{2}(\alpha) & =1-C_{2}^{2}(\alpha) \\
& =\left(C_{2}^{2}\left(\beta^{\prime}\right)+k_{2} S_{2}^{2}\left(\beta^{\prime}\right)\right)\left(C_{2}^{2}(\gamma)+k_{2} S_{2}^{2}(\gamma)\right) \\
& -\left(C_{2}\left(\beta^{\prime}\right) C_{2}(\gamma)+k_{2} S_{2}\left(\beta^{\prime}\right) S_{2}(\gamma) C_{1}(a)\right)^{2} \\
& =C_{2}^{2}\left(\beta^{\prime}\right) C_{2}^{2}(\gamma)+k_{2} C_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma)+k_{2} S_{2}^{2}\left(\beta^{\prime}\right) C_{2}^{2}(\gamma)+k_{2}^{2} S_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma) \\
& -C_{2}^{2}\left(\beta^{\prime}\right) C_{2}^{2}(\gamma)-2 k_{2} C_{2}\left(\beta^{\prime}\right) S_{2}\left(\beta^{\prime}\right) C_{2}(\gamma) S_{2}(\gamma) C_{1}(a)-k_{2}^{2} S_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma) C_{1}^{2}(a) \\
& =k_{2}\left(C_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma)+S_{2}^{2}\left(\beta^{\prime}\right) C_{2}^{2}(\gamma)-2 C_{2}\left(\beta^{\prime}\right) S_{2}\left(\beta^{\prime}\right) C_{2}(\gamma) S_{2}(\gamma) C_{1}(a)\right) \\
& +k_{2}^{2} S_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma)\left(1-C_{1}^{2}(a)\right) \\
& =k_{2}\left(C_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma)+S_{2}^{2}\left(\beta^{\prime}\right) C_{2}^{2}(\gamma)-2 C_{2}\left(\beta^{\prime}\right) S_{2}\left(\beta^{\prime}\right) C_{2}(\gamma) S_{2}(\gamma) C_{1}(a)\right) \\
& +k_{1} k_{2}^{2} S_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma) S_{1}^{2}(a), \\
S_{2}^{2}(\alpha) & =C_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma)+S_{2}^{2}\left(\beta^{\prime}\right) C_{2}^{2}(\gamma)-2 C_{2}\left(\beta^{\prime}\right) S_{2}\left(\beta^{\prime}\right) C_{2}(\gamma) S_{2}(\gamma) C_{1}(a) \\
& +k_{1} k_{2} S_{2}^{2}\left(\beta^{\prime}\right) S_{2}^{2}(\gamma) S_{1}^{2}(a),
\end{aligned}
$$

or

$$
\begin{equation*}
T_{2}^{2}(\alpha)=\frac{T_{2}^{2}\left(\beta^{\prime}\right)+T_{2}^{2}(\gamma)-2 T_{2}\left(\beta^{\prime}\right) T_{2}(\gamma) C_{1}(a)+k_{1} k_{2} T_{2}^{2}\left(\beta^{\prime}\right) T_{2}^{2}(\gamma) S_{1}^{2}(a)}{\left(1+k_{2} T_{2}\left(\beta^{\prime}\right) T_{2}(\gamma) C_{1}(a)\right)^{2}} \tag{2.5}
\end{equation*}
$$

Similarly,

$$
\begin{align*}
S_{2}^{2}\left(\beta^{\prime}\right) & =C_{2}^{2}(\alpha) S_{2}^{2}(\gamma)+S_{2}^{2}(\alpha) C_{2}^{2}(\gamma)+2 C_{2}(\alpha) S_{2}(\alpha) C_{2}(\gamma) S_{2}(\gamma) C_{1}(b) \\
& +k_{1} k_{2} S_{2}^{2}(\alpha) S_{2}^{2}(\gamma) S_{1}^{2}(b), \\
T_{2}^{2}\left(\beta^{\prime}\right) & =\frac{T_{2}^{2}(\alpha)+T_{2}^{2}(\gamma)+2 T_{2}(\alpha) T_{2}(\gamma) C_{1}(b)+k_{1} k_{2} T_{2}^{2}(\alpha) T_{2}^{2}(\gamma) S_{1}^{2}(b)}{\left(1-k_{2} T_{2}(\alpha) T_{2}(\gamma) C_{1}(b)\right)^{2}} \tag{2.6}
\end{align*}
$$

and

$$
\begin{align*}
S_{2}^{2}(\gamma) & =C_{2}^{2}(\alpha) S_{2}^{2}\left(\beta^{\prime}\right)+S_{2}^{2}(\alpha) C_{2}^{2}\left(\beta^{\prime}\right)-2 C_{2}(\alpha) S_{2}(\alpha) C_{2}\left(\beta^{\prime}\right) S_{2}\left(\beta^{\prime}\right) C_{1}(c) \\
& +k_{1} k_{2} S_{2}^{2}(\alpha) S_{2}^{2}\left(\beta^{\prime}\right) S_{1}^{2}(c), \\
T_{2}^{2}(\gamma) & =\frac{T_{2}^{2}(\alpha)+T_{2}^{2}\left(\beta^{\prime}\right)-2 T_{2}(\alpha) T_{2}\left(\beta^{\prime}\right) C_{1}(c)+k_{1} k_{2} T_{2}^{2}(\alpha) T_{2}^{2}\left(\beta^{\prime}\right) S_{1}^{2}(c)}{\left(1+k_{2} T_{2}(\alpha) T_{2}\left(\beta^{\prime}\right) C_{1}(c)\right)^{2}} \tag{2.7}
\end{align*}
$$

What does it mean for triangle? From the Sine law (2.1), having function $S(x)$ monotonically increasing result that the longest side of any triangle is opposite to the largest angle and the shortest side is opposed to the smallest angle. From the Cosine I law in its form that uses $C_{1}(x)$ function, having

$$
\begin{aligned}
& C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c)=C_{1}(b-c), \\
& C_{1}(a) C_{1}(b)-k_{1} S_{1}(a) S_{1}(b)=C_{1}(a+b),
\end{aligned}
$$

and $C_{i}(x) l e 1$ and is decreasing when $k_{i}=1, C_{i}(x)=1$ and is constant when $k_{i}=0, C_{i}(x) g e 1$ is increasing when $k_{i}=-1$, we can see:

$$
C_{1}(a)=C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha)
$$

is equivalent to

$$
a\left\{\begin{array}{l}
>b-c, k_{2}=1 \\
=b-c, k_{2}=0 \\
<b-c, k_{2}=-1
\end{array}\right.
$$

Similarly,

$$
C_{1}(b)=C_{1}(a) C_{1}(c)-k_{1} S_{1}(a) S_{1}(c) C_{2}\left(\beta^{\prime}\right)
$$

is equivalent to

$$
b\left\{\begin{array}{l}
<a+c, k_{2}=1 \\
=a+c, k_{2}=0 \\
>a+c, k_{2}=-1
\end{array}\right.
$$

From the Cosine II law we can see:

$$
C_{2}(\alpha)=C_{2}\left(\beta^{\prime}\right) C_{2}(\gamma)+k_{2} S_{2}\left(\beta^{\prime}\right) S_{2}(\gamma) C_{1}(a)
$$

is equivalent to

$$
\alpha\left\{\begin{array}{l}
>\beta^{\prime}-\gamma, k_{1}=1 \\
=\beta^{\prime}-\gamma, k_{1}=0 \\
<\beta^{\prime}-\gamma, k_{1}=-1
\end{array}\right.
$$

Similarly,

$$
C_{2}\left(\beta^{\prime}\right)=C_{2}(\alpha) C_{2}(\gamma)-k_{2} S_{2}(\alpha) S_{2}(\gamma) C_{1}(b)
$$

is equivalent to

$$
\beta^{\prime}\left\{\begin{array}{l}
<\alpha+\gamma, k_{1}=1 \\
=\alpha+\gamma, k_{1}=0 \\
>\alpha+\gamma, k_{1}=-1
\end{array}\right.
$$

### 2.2 Right (Quasi)-Triangle Equations

We can define orthogonality in $\mathbb{B}^{n}$ using the orthogonality in $\mathbb{R}^{n}$. Namely, two vectors $v_{1}$ and $v_{2}$ of space $\mathbb{R}^{n}$ are orthogonal, if $v_{1} \odot v_{2}=0$.

For $\mathbb{B}^{2}$ plane and a line, the orthogonal bundle is line only if $k_{2}=1$. In this case when line rotates count-clockwise, its orthogonal line rotates count-clockwise and vice-versa (Figure 1.4 a). When $k_{2}=0$ there is the only orthogonal bundle, which doesn't rotate (Figure 1.4 b ). When $k_{2}=-1$ the orthogonal bundle rotates clockwise when the line rotates count-clockwise toward to the same limit bundle and vice-versa (Figure 1.4 c ).

Generally we can't speak about right triangle as one of its catheti is line and another isn't (when $k_{2} \neq 1$ ). However, as we will see, this figure is important. We will name it right (quasi)-triangle, which means right triangle, when $k_{2}=1$ and right quasi-triangle when $k_{2} \neq 1$.

We will construct a (quasi)-triangle as half of isosceles one (Figure 2.2). Consider a triangle $A_{0} B_{0} A_{0}^{\prime}$ with $A_{0} B_{0}=A_{0}^{\prime} B_{0}=c, A_{0} A_{0}^{\prime}=2 b, \angle A_{0}^{\prime} A_{0} B_{0}=\angle A_{0} A_{0}^{\prime} B_{0}=\alpha$ and external angle $\angle A_{0} B_{0} A_{0}^{\prime}=2 \beta^{\prime}$.


Figure 2.2: Right (quasi)-triangle equations deduction.

Let $A_{0}^{\prime}=O=[1: 0: 0]$ be origin, $A_{0}=\mathfrak{R}_{1}(2 b) A_{0}^{\prime}=\left[C_{1}(2 b): S_{1}(2 b): 0\right], B_{0}=\mathfrak{R}_{2}(\alpha) \Re_{1}(c)$ $A_{0}=\left[C_{1}(c): S_{1}(c) C_{2}(\alpha): S_{1}(c) S_{2}(\alpha)\right]$.

Let $A B A^{\prime}=\mathfrak{R}_{1}(-b)\left(A_{0} B_{0} A_{0}^{\prime}\right)$ (Figure 2.2, black). Now, $A^{\prime}=\mathfrak{R}_{1}(-b) A_{0}^{\prime}=\left[C_{1}(b):-S_{1}(b)\right.$ : $0], A=\mathfrak{R}_{1}(-b) A_{0}=\left[C_{1}(b): S_{1}(b): 0\right]$ and $B=\mathfrak{R}_{1}(-b) B_{0}=$

$$
\left(\begin{array}{ccc}
C_{1}(b) & k_{1} S_{1}(b) & 0 \\
-S_{1}(b) & C_{1}(b) & 0 \\
0 & 0 & 1
\end{array}\right)\left(\begin{array}{c}
C_{1}(c) \\
S_{1}(c) C_{2}(\alpha) \\
S_{1}(c) S_{2}(\alpha)
\end{array}\right)
$$

$=\left[C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha):-S_{1}(b) C_{1}(c)+C_{1}(b) S_{1}(c) C_{2}(\alpha): S_{1}(c) S_{2}(\alpha)\right]$.
Finally, let $C \in A A^{\prime}, A C=A^{\prime} C=b$. Then $C=[1: 0: 0]$ is origin. From figure equality $A^{\prime} B C=A B C$ result $B C \perp A^{\prime} A$. Therefore we can consider $A B C$ right (quasi)-triangle.

Having figures $A^{\prime} B C=A B C$ and $C$ is origin, result $B$ have form $B=(x, 0, y)$, where

$$
\begin{aligned}
C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) C_{2}(\alpha) & =x \\
-S_{1}(b) C_{1}(c)+C_{1}(b) S_{1}(c) C_{2}(\alpha) & =0 \\
S_{1}(c) S_{2}(\alpha) & =y .
\end{aligned}
$$

From the second equality, have

$$
\begin{equation*}
T_{1}(b)=T_{1}(c) C_{2}(\alpha) \tag{2.8}
\end{equation*}
$$

Using the value of $C_{2}(\alpha)$ from this equality and putting it in the first one, have

$$
\begin{aligned}
x & =C_{1}(b) C_{1}(c)+k_{1} S_{1}(b) S_{1}(c) \frac{T_{1}(b)}{T_{1}(c)} \\
& =\frac{C_{1}(c)}{C_{1}(b)}\left(C_{1}^{2}(b)+k_{1} S_{1}^{2}(b)\right)=\frac{C_{1}(c)}{C_{1}(b)}
\end{aligned}
$$

Let now calculate the value of

$$
\begin{aligned}
x^{2}+k_{1} k_{2} y^{2} & =\frac{C_{1}^{2}(c)}{C_{1}^{2}(b)}+k_{1} k_{2} S_{1}^{2}(c) S_{2}^{2}(\alpha) \\
& =\frac{C_{1}^{2}(c)}{C_{1}^{2}(b)}+k_{1} S_{1}^{2}(c)\left(1-C_{2}^{2}(\alpha)\right) \\
& =\frac{C_{1}^{2}(c)}{C_{1}^{2}(b)}+k_{1} S_{1}^{2}(c)-k_{1} S_{1}^{2}(c) \frac{T_{1}^{2}(b)}{T_{1}^{2}(c)} \\
& =\frac{C_{1}^{2}(c)}{C_{1}^{2}(b)}+k_{1} S_{1}^{2}(c)-k_{1} S_{1}^{2}(b) \frac{C_{1}^{2}(c)}{C_{1}^{2}(b)} \\
& =\frac{C_{1}^{2}(c)}{C_{1}^{2}(b)}\left(1-k_{1} S_{1}^{2}(b)\right)+k_{1} S_{1}^{2}(c) \\
& =\frac{C_{1}^{2}(c)}{C_{1}^{2}(b)} C_{1}^{2}(b)+k_{1} S_{1}^{2}(c) \\
& =C_{1}^{2}(c)+k_{1} S_{1}^{2}(c)=1
\end{aligned}
$$

It means that exists $a \in \mathbb{R}, C_{12}(a)=\frac{C_{1}(c)}{C_{1}(b)}, S_{12}(a)=S_{1}(c) S_{2}(\alpha)$ that has characteristic $k=$ $k_{1} k_{2}$. It is a 'distance' parameter $B C$. We have two more equations:

$$
\begin{align*}
C_{1}(c) & =C_{12}(a) C_{1}(b)  \tag{2.9}\\
S_{12}(a) & =S_{1}(c) S_{2}(\alpha) \tag{2.10}
\end{align*}
$$

From (2.8) and (2.10) have:

$$
\frac{S_{12}(a)}{T_{1}(b)}=\frac{S_{1}(c) S_{2}(\alpha)}{T_{1}(c) C_{2}(\alpha)}=C_{1}(c) T_{2}(\alpha)
$$

using the value of $C_{1}(c)$ from (2.9) have

$$
\begin{align*}
& \frac{S_{12}(a)}{T_{1}(b)}=C_{12}(a) C_{1}(b) T_{2}(\alpha) \\
& T_{12}(a)=S_{1}(b) T_{2}(\alpha) \tag{2.11}
\end{align*}
$$

The last 6 equations will include $\beta^{\prime}$. In order to be able to deduce them we will introduce translation $\mathfrak{T}(-a)$, so as $\mathfrak{T}(-a) B=C$. Having characteristic $a$ is $k_{1} k_{2}=K_{2}$,

$$
\mathfrak{T}(-a)=\left(\begin{array}{ccc}
C_{12}(a) & 0 & K_{2} S_{12}(a) \\
0 & 1 & 0 \\
-S_{12}(a) & 0 & C_{12}(a)
\end{array}\right)
$$

We can check this map preserves vector product.
Applying $\mathfrak{T}(-a)$, obtain $B_{1}=\mathfrak{T}(-a) B=[1: 0: 0], C_{1}=\mathfrak{T}(-a) C=\left[C_{12}(a): 0:-S_{12}(a)\right]$ and $A_{1}=\mathfrak{T}(-a) A=$

$$
\left(\begin{array}{ccc}
C_{12}(a) & 0 & K_{2} S_{12}(a) \\
0 & 1 & 0 \\
-S_{12}(a) & 0 & C_{12}(a)
\end{array}\right)\left(\begin{array}{c}
C_{1}(b) \\
S_{1}(b) \\
0
\end{array}\right)
$$

$=\left[C_{12}(a) C_{1}(b): S_{1}(b):-S_{12}(a) C_{1}(b)\right]=\left[C_{1}(c): S_{1}(c) C_{2}\left(\beta^{\prime}\right):-S_{1}(c) S_{2}\left(\beta^{\prime}\right)\right]$ (Figure 2.2, cyan). From here we have

$$
\begin{equation*}
S_{1}(b)=S_{1}(c) C_{2}\left(\beta^{\prime}\right) . \tag{2.12}
\end{equation*}
$$

Moreover, having (2.9), obtain

$$
\begin{align*}
S_{12}(a) C_{1}(b) & =S_{1}(c) S_{2}\left(\beta^{\prime}\right), \\
S_{12}(a) \frac{C_{1}(c)}{C_{12}(a)} & =S_{1}(c) S_{2}\left(\beta^{\prime}\right), \\
T_{12}(a) & =T_{1}(c) S_{2}\left(\beta^{\prime}\right) . \tag{2.13}
\end{align*}
$$

Combining the last 2 equalities (2.12), (2.13) with (2.9), we have

$$
\begin{align*}
\frac{T_{12}(a)}{S_{1}(b)} & =\frac{T_{1}(c) S_{2}\left(\beta^{\prime}\right)}{S_{1}(c) C_{2}\left(\beta^{\prime}\right)} \\
=\frac{T_{2}\left(\beta^{\prime}\right)}{C_{1}(c)} & =\frac{T_{2}\left(\beta^{\prime}\right)}{C_{12}(a) C_{1}(b)}, \\
S_{12}(a) & =T_{1}(b) T_{2}\left(\beta^{\prime}\right) . \tag{2.14}
\end{align*}
$$

Now, having (2.8) and (2.12):

$$
T_{1}(c) C_{2}(\alpha)=T_{1}(b)=\frac{S_{1}(b)}{C_{1}(b)}=\frac{S_{1}(c) C_{2}\left(\beta^{\prime}\right)}{C_{1}(b)}
$$

calculate with (2.9):

$$
\begin{align*}
C_{1}(b) & =\frac{S_{1}(c) C_{2}\left(\beta^{\prime}\right)}{T_{1}(c) C_{2}(\alpha)} \\
& =C_{1}(c) \frac{C_{2}\left(\beta^{\prime}\right)}{C_{2}(\alpha)} \\
& =C_{12}(a) C_{1}(b) \frac{C_{2}\left(\beta^{\prime}\right)}{C_{2}(\alpha)}, \\
C_{12}(a) \frac{C_{2}\left(\beta^{\prime}\right)}{C_{2}(\alpha)} & =1, \\
C_{2}(\alpha) & =C_{12}(a) C_{2}\left(\beta^{\prime}\right) . \tag{2.15}
\end{align*}
$$

Now from (2.11), (2.12) and (2.13),

$$
\begin{align*}
T_{12}(a)=S_{1}(b) T_{2}(\alpha) & =T_{1}(c) S_{2}\left(\beta^{\prime}\right), \\
S_{1}(c) C_{2}\left(\beta^{\prime}\right) T_{2}(\alpha) & =T_{1}(c) S_{2}\left(\beta^{\prime}\right), \\
T_{2}\left(\beta^{\prime}\right) & =C_{1}(c) T_{2}(\alpha) . \tag{2.16}
\end{align*}
$$

Finally, by multiplying the last equations (2.15) and (2.16), have using (2.9):

$$
\begin{align*}
C_{1}(c) T_{2}(\alpha) C_{2}(\alpha) & =T_{2}\left(\beta^{\prime}\right) C_{12}(a) C_{2}\left(\beta^{\prime}\right), \\
C_{1}(c) S_{2}(\alpha) & =C_{12}(a) C_{2}\left(\beta^{\prime}\right), \\
C_{12}(a) C_{1}(b) S_{2}(\alpha) & =C_{12}(a) S_{2}\left(\beta^{\prime}\right), \\
S_{2}\left(\beta^{\prime}\right) & =C_{1}(b) S_{2}(\alpha) . \tag{2.17}
\end{align*}
$$

It is necessary to modify equations (2.9) and (2.15) in order to not contain the $C(x)$ function.

$$
\begin{aligned}
k_{1} S_{1}^{2}(c) & =1-C_{1}^{2}(c)=\left(C_{12}^{2}(a)+k_{1} k_{2} S_{12}^{2}(a)\right)\left(C_{1}^{2}(b)+k_{1} S_{1}^{2}(b)\right)-C_{12}^{2}(a) C_{1}^{2}(b) \\
& =k_{1} k_{2} S_{12}^{2}(a) C_{1}^{2}(b)+k_{1} C_{12}^{2}(a) S_{1}^{2}(b)+k_{1}^{2} k_{2} S_{12}^{2}(a) S_{1}^{2}(b), \\
S_{1}^{2}(c) & =k_{2} S_{12}^{2}(a) C_{1}^{2}(b)+C_{12}^{2}(a) S_{1}^{2}(b)+k_{1} k_{2} S_{12}^{2}(a) S_{1}^{2}(b)
\end{aligned}
$$

By dividing the last equality by its $C(x)$ form, obtain:

$$
\begin{equation*}
T_{1}^{2}(c)=k_{2} T_{12}^{2}(a)+T_{1}^{2}(b)+k_{1} k_{2} T_{12}^{2}(a) T_{1}^{2}(b) \tag{2.18}
\end{equation*}
$$

Similarly,

$$
\begin{aligned}
k_{2} S_{2}^{2}(\alpha) & =1-C_{2}^{2}(\alpha)=\left(C_{12}^{2}(a)+k_{1} k_{2} S_{12}^{2}(a)\right)\left(C_{2}^{2}\left(\beta^{\prime}\right)+k_{2} S_{2}^{2}\left(\beta^{\prime}\right)\right)-C_{12}^{2}(a) C_{2}^{2}\left(\beta^{\prime}\right) \\
& =k_{2} C_{12}^{2}(a) S_{2}^{2}\left(\beta^{\prime}\right)+k_{1} k_{2} S_{12}^{2}(a) C_{2}^{2}\left(\beta^{\prime}\right)+k_{1} k_{2}^{2} S_{12}^{2}(a) S_{2}^{2}\left(\beta^{\prime}\right) \\
S_{2}^{2}(\alpha) & =C_{12}^{2}(a) S_{2}^{2}\left(\beta^{\prime}\right)+k_{1} S_{12}^{2}(a) C_{2}^{2}\left(\beta^{\prime}\right)+k_{1} k_{2} S_{12}^{2}(a) S_{2}^{2}\left(\beta^{\prime}\right)
\end{aligned}
$$

By dividing the last equality by its $C(x)$ form, obtain:

$$
\begin{equation*}
T_{2}^{2}(\alpha)=k_{1} T_{12}^{2}(a)+T_{2}^{2}\left(\beta^{\prime}\right)+k_{1} k_{2} T_{12}^{2}(a) T_{2}^{2}\left(\beta^{\prime}\right) \tag{2.19}
\end{equation*}
$$

Note that for $k_{2}=1$ equations (2.8) - (2.19) can be used if external angle $\beta^{\prime}$ change to internal $\beta$ with the following changes:

$$
\begin{aligned}
\beta & =\frac{\pi}{2}-\beta^{\prime} \\
\cos \beta & =\sin \beta^{\prime} \\
\sin \beta & =\cos \beta^{\prime} \\
\tan \beta & =\cot \beta^{\prime} \\
\cot \beta & =\tan \beta^{\prime}
\end{aligned}
$$

### 2.3 More rotations

As we can see, transformation $\mathfrak{T}(-a)$ preserves vector product. In order to be a motion it needs to be presented as finite product of main rotations. If $a, b, c, \alpha$ and $\beta^{\prime}$ are real numbers for which have place equalities (2.8) - (2.17) then it can be checked that $\mathfrak{T}(a)=$ $\mathfrak{R}_{2}\left(\beta^{\prime}\right) \mathfrak{R}_{1}(c) \mathfrak{R}_{2}(-\alpha) \mathfrak{R}_{1}(-b)$. We will introduce new transformations as following:

$$
\Re_{i j}(\phi)=\left(\begin{array}{ccccccc}
1 & \ldots & 0 & \ldots & 0 & \ldots & 0 \\
\vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\
0 & \ldots & C_{i+1, \ldots j}(\phi) & \ldots & -\frac{K_{j}}{K_{i}} S_{i+1, \ldots j}(\phi) & \ldots & 0 \\
\vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\
0 & \ldots & S_{i+1, \ldots j}(\phi) & \ldots & C_{i+1, \ldots j}(\phi) & \ldots & 0 \\
\vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\
0 & \ldots & 0 & \ldots & 0 & \ldots & 1
\end{array}\right) .
$$

It's easy to see that all $\mathfrak{R}_{i j}(\phi)$ are motions. All they can be presented as finit product of main rotations. We will name them rotations of the space $\mathbb{B}^{n}$. In special case, $\mathfrak{R}_{0 i}(\phi)$ we will name them translations $\mathfrak{T}_{i}(\phi)$ of the space.

### 2.4 Generalized Orthogonal Matrix

For $\mathbb{R} \mathbb{P}^{n}$ with given specification $k_{p}, p=\overline{1, n}$ we will name the vector $x \in \mathbb{R}^{n} \mathbb{P}^{n}$ upper $i$ normalized, $i \in \overline{0, n}$ if $\frac{1}{K_{i}} x \odot x=1$. For $0 \leq i<j \leq n$ we will name two vectors $x$ and $y$ upper $i j$-orthogonal if $\frac{1}{K_{\min (i, j)}} x \odot y=0$. We will name the matrix $M_{(n+1) \times(n+1)}$ composed of columns $c_{i}$ upper orthogonal if all columns $c_{i}$ are upper $i$-normalized and any two columns $c_{i}$ and $c_{j}$ are upper $i j$-orthogonal.

It's easy to see that all main rotation matrixes are upper orthogonal. Moreover product of two upper orthogonal matrix is upper orthogonal. Really, let $X, Y$ are two upper orthogonal matrices. It means that $X$ is composed of $\left(x_{0}, \ldots, x_{n}\right)$ columns and $Y$ - from $\left(y_{0}, \ldots, y_{n}\right)$ columns, and $\frac{1}{K_{\min (i, j)}} x_{i} \odot x_{j}=\frac{1}{K_{\min (i, j)}} y_{i} \odot y_{j}=\delta_{i j}$ for all $i, j=\overline{0, n}$, where $\delta_{i j}=1, i=j$ and $\delta_{i j}=0, i \neq j$. Let $Z=X Y$ with elements $z_{i j}=\sum_{p=0}^{n} x_{i p} y_{p j}$. Let $z_{i}$ and $z_{j}$ be 2 columns of $Z$. Let calculate

$$
\begin{aligned}
\frac{1}{K_{\min (i, j)}} z_{i} \odot z_{j} & =\frac{1}{K_{\min (i, j)}} \sum_{p=0}^{n} K_{p} z_{p i} z_{p j} \\
& =\frac{1}{K_{\min (i, j)}} \sum_{p=0}^{n} K_{p}\left(\sum_{m_{1}=0}^{n} x_{p m_{1}} y_{m_{1} i}\right)\left(\sum_{m_{2}=0}^{n} x_{p m_{2}} y_{m_{2} j}\right) \\
& =\frac{1}{K_{\min (i, j)}} \sum_{p=0}^{n} K_{p} \sum_{m_{1}=0}^{n} \sum_{m_{2}=0}^{n} x_{p m_{1}} x_{p m_{2}} y_{m_{1} i} y_{m_{2} j} \\
& =\frac{1}{K_{\min (i, j)}} \sum_{m_{1}=0}^{n} \sum_{m_{2}=0}^{n} y_{m_{1} i} y_{m_{2} j} \sum_{p=0}^{n} K_{p} x_{p m_{1}} x_{p m_{2}} \\
& =\frac{1}{K_{\min (i, j)}} \sum_{m_{1}=0}^{n} \sum_{m_{2}=0}^{n} y_{m_{1} i} y_{m_{2} j} K_{\min \left(m_{1}, m_{2}\right)} \delta_{m_{1} m_{2}} \\
& =\frac{1}{K_{\min (i, j)}} \sum_{m=0}^{n} y_{m i} y_{m j} K_{m} \\
& =\delta_{i j}
\end{aligned}
$$

We will name the vector $x \in \mathbb{R P}^{n}$ lower $i$-normalized, $i \in \overline{0, n}$ if $K_{i} \sum_{j=0}^{n} \frac{x_{j}^{2}}{K_{j}}=1$. For $0 \leq i<j \leq n$ we will name two vectors $x$ and $y$ lower $i j$-orthogonal if $K_{\max (i, j)} \sum_{p=0}^{n} \frac{x_{p} y_{p}}{K_{p}}=0$. We will name the matrix $M_{(n+1) \times(n+1)}$ composed of rows $r_{i}$ lower orthogonal if all rows $r_{i}$ are lower $i$-normalized and any two rows $r_{i}$ and $r_{j}$ are lower $i j$-orthogonal.

It's easy to see that all main rotation matrixes are also lower orthogonal. Moreover product of two lower orthogonal matrices is lower orthogonal. Really, let $X, Y$ are two lower orthogonal matrices. It means that $X$ is composed of $\left(x_{0}, \ldots, x_{n}\right)$ rows and $Y$ - from $\left(y_{0}, \ldots, y_{n}\right)$ rows, where $K_{\max (i, j)} \sum_{p=0}^{n} \frac{x_{i p} x_{j p}}{K_{p}}=K_{\max (i, j)} \sum_{p=0}^{n} \frac{y_{i p} y_{j p}}{K_{p}}=\delta_{i j}$ for all $i, j=\overline{0, n}$. Let $Z=X Y$. Let
calculate

$$
\begin{aligned}
K_{\max (i, j)} \sum_{p=0}^{n} \frac{z_{i p} z_{j p}}{K_{p}} & =K_{\max (i, j)} \sum_{p=0}^{n} \frac{1}{K_{p}}\left(\sum_{m_{1}=0}^{n} x_{i m_{1}} y_{m_{1} p}\right)\left(\sum_{m_{2}=0}^{n} x_{j m_{2}} y_{m_{2} p}\right) \\
& =K_{\max (i, j)} \sum_{p=0}^{n} \frac{1}{K_{p}} \sum_{m_{1}=0}^{n} \sum_{m_{2}=0}^{n} x_{i m_{1}} x_{j m_{2}} y_{m_{1} p} y_{m_{2} p} \\
& =K_{\max (i, j)} \sum_{m_{1}=0}^{n} \sum_{m_{2}=0}^{n} x_{i m_{1}} x_{j m_{2}} \sum_{p=0}^{n} \frac{1}{K_{p}} y_{m_{1} p} y_{m_{2} p} \\
& =K_{\max (i, j)} \sum_{m_{1}=0}^{n} \sum_{m_{2}=0}^{n} x_{i m_{1}} x_{j m_{2}} \frac{\delta_{m_{1} m_{2}}}{K_{\max \left(m_{1}, m_{2}\right)}} \\
& =K_{\max (i, j)} \sum_{m=0}^{n} \frac{x_{i m} x_{j m}}{K_{m}} \\
& =\delta_{i j}
\end{aligned}
$$

For some upper orthogonal matrix $X$ has place the equality

$$
\begin{aligned}
\frac{1}{K_{j}} \sum_{i=0}^{n} K_{i} x_{i j}^{2} & =1, \\
\sum_{i=0}^{n} K_{i} x_{i j}^{2} & =K_{j}, \\
\sum_{i=0}^{n} x_{i j}^{2} \prod_{p=1}^{i} k_{p} & =\prod_{p=1}^{j} k_{p} .
\end{aligned}
$$

Let divide it to $K_{q}, q \leq j$ :

$$
\sum_{i=0}^{q-1} \frac{K_{i}}{K_{q}} x_{i j}^{2}+\sum_{i=q}^{n} \frac{K_{i}}{K_{q}} x_{i j}^{2}=\frac{K_{j}}{K_{q}}
$$

As $K_{q}$ divides $K_{j}$ and $K_{i}, i \geq q$, but $K_{q}$ doesn't divide $K_{i}, i<q$, result that for $0 \leq i<j \leq n$, $x_{i j}$ divide $K_{q} / K_{i}$ for all $i<q \leq j$, or $x_{i j}$ divide $K_{j} / K_{i}$.

Having for some upper orthogonal matrix $X$, elements $x_{i j}$ divide $K_{j} / K_{i}$, construct the matrix $Y$ of the same size with elements $y_{i j}=\sqrt{\frac{K_{i}}{K_{j}}} x_{i j}$. The matrix $Y$ is orthogonal one ${ }^{1}$ (if may be complex, in this case it isn't unitar, but orthogonal). Really, for some $i=\overline{0, n}$,

$$
\sum_{i=0}^{n} y_{i j}^{2}=\sum_{i=0}^{n} \frac{K_{i}}{K_{j}} x_{i j}^{2}=\frac{1}{K_{j}} \sum_{i=0}^{n} K_{i} x_{i j}^{2}=1
$$

[^5]and for some $j_{1} \neq j_{2}=\overline{0, n}$,
\[

$$
\begin{aligned}
\sum_{i=0}^{n} y_{i j_{1}} y_{i j_{2}} & =\sum_{i=0}^{n} \frac{K_{i}}{\sqrt{K_{j_{1}} K_{j_{2}}}} x_{i j_{1}} x_{i j_{2}} \\
& =\frac{K_{\min \left(j_{1}, j_{2}\right)}}{\sqrt{K_{j_{1}} K_{j_{2}}}} \frac{1}{K_{\min \left(j_{1}, j_{2}\right)}} \sum_{i=0}^{n} K_{i} x_{i j_{1}} x_{i j_{2}} \\
& =\frac{K_{\min \left(j_{1}, j_{2}\right)}}{\sqrt{K_{j_{1}} K_{j_{2}}}} 0=0
\end{aligned}
$$
\]

because $x_{i j_{1}} x_{i j_{2}}$ divides $\frac{\sqrt{K_{j_{1}} K_{j_{2}}}}{K_{i}}$.
For some orthogonal matrix always has place also the following equalities for $i_{1} \neq i_{2}=\overline{0, n}$ :

$$
\begin{aligned}
\sum_{j=0}^{n} y_{i j}^{2} & =\sum_{j=0}^{n} \frac{K_{i}}{K_{j}} x_{i j}^{2}=K_{i} \sum_{j=0}^{n} \frac{1}{K_{j}} x_{i j}^{2}=1, \\
\sum_{j=0}^{n} y_{i_{1} j} y_{i_{2} j} & =\sum_{j=0}^{n} \frac{\sqrt{K_{i_{1}} K_{i_{2}}}}{K_{j}} x_{i_{1} j} x_{i_{2} j} \\
& =\frac{\sqrt{K_{i_{1}} K_{i_{2}}}}{K_{\max \left(i_{1}, i_{2}\right)}} K_{\max \left(i_{1}, i_{2}\right)} \sum_{j=0}^{n} \frac{1}{K_{j}} x_{i_{1} j} x_{i_{2} j} \\
& =\frac{\sqrt{K_{i_{1}} K_{i_{2}}}}{K_{\max \left(i_{1}, i_{2}\right)}} 0=0
\end{aligned}
$$

So, $x_{i_{1} j} x_{i_{2} j}$ divides $\frac{K_{j}}{\sqrt{K_{i_{1}} K_{i_{2}}}}$. It means that $X$ is also lower orthogonal matrix.
Inverse orthogonal matrix $Y^{-1}$ is easy constructed as $y_{j i}^{\prime}=y_{i j}$. Then

$$
\begin{aligned}
\sqrt{\frac{K_{j}}{K_{i}}} x_{j i}^{\prime} & =\sqrt{\frac{K_{i}}{K_{j}}} x_{i j} \\
x_{j i}^{\prime} & =\frac{K_{i}}{K_{j}} x_{i j}
\end{aligned}
$$

The last equality isn't applicable if some characteristic $k_{i}=0$. Although is true, it isn't determinable having the form of $0 / 0$. If some characteristic $k_{m}=0, m<n$, the matrix has the form:

$$
M=\left(\begin{array}{ll}
A & O \\
B & C
\end{array}\right)
$$

Really, for the first $m$ columns the upper orthogonality condition is equivalent to:

$$
\begin{aligned}
\sum_{i=0}^{n} \frac{K_{i}}{K_{j}} x_{i j}^{2} & =\sum_{i=0}^{m-1} \frac{K_{i}}{K_{j}} x_{i j}^{2}, \forall j=\overline{0, m-1}, \\
\sum_{i=0}^{n} \frac{K_{i}}{K_{j_{1}}} x_{i j_{1}} x_{i j_{2}} & =\sum_{i=0}^{m-1} \frac{K_{i}}{K_{j_{1}}} x_{i j_{1}} x_{i j_{2}}, \forall j_{1}=\overline{0, m-1}, j_{2}=\overline{0, n}, j_{1}<j_{2},
\end{aligned}
$$

Having $K_{j} \neq 0$ and $K_{i}=0$, all terms, starting with $i=m$ equals to zero. So, matrix $A$ is upper orthogonal of size $m \times m$ and matrix $B$ is free of size $(n-m+1) \times m$. For the last $n-m+1$ columns upper orthogonality has form:

$$
\begin{aligned}
\sum_{i=0}^{n} \frac{K_{i}}{K_{j}} x_{i j}^{2} & =\sum_{i=m}^{n} \frac{\prod_{p=m}^{i} k_{p}}{\prod_{p=m}^{j} k_{p}} x_{i j}^{2}, \forall j=\overline{m, n} \\
\sum_{i=0}^{n} \frac{K_{i}}{K_{j_{1}}} x_{i j_{1}} x_{i j_{2}} & =\sum_{i=m}^{n} \frac{\prod_{p=m}^{i} k_{p}}{\prod_{p=m}^{j_{1}} k_{p}} x_{i j_{1}} x_{i j_{2}}, \forall j_{1}<j_{2}=\overline{m, n}
\end{aligned}
$$

because $K_{j}=0$. It means the matrix $C$ is upper orthogonal of size $(n-m+1) \times(n-m+1)$ and the matrix $O$ is obligatory zero one of size $m \times(n-m+1)$ (otherwise elements of $M$ aren't finite).

It's easy to verify that inverse matrix has form:

$$
M^{-1}=\left(\begin{array}{cc}
A^{-1} & O \\
-C^{-1} B A^{-1} & C^{-1}
\end{array}\right)
$$

This way of calculating the inverse matrix can easy be generalized to either number of null characteristics.

We will name upper orthogonal matrixes (which also are lower orthogonal) generalized orthogonal. We will use the term orthogonal matrix meaning generalized orthogonal matrix if isn't stated otherwise. As we can see, the orthogonal matrix set is closed in respect of multiplying, it contains the unit element and for any element it contains its inverse. So the orthogonal matrix set form isomentry group of space. All motion matrices are generalized orthogonal.

### 2.5 Orthogonal Matrix as Product of Rotations

Let $X$ will be orthogonal matrix. We will search the rotation matrices, the product of which gives $X$. Note that The matrix $X \mathfrak{R}_{i j}(\phi)$ have all columns $x_{p}$ of $X$ except $i$ and $j$ ones. These columns are $x_{i}^{\prime}=x_{i} C_{i+1 \ldots j}(\phi)+x_{j} S_{i+1 \ldots j}(\phi)$ and $x_{j}^{\prime}=-\frac{K_{j}}{K_{i}} x_{i} S_{i+1 \ldots j}(\phi)+x_{j} C_{i+1 \ldots j}$.

For the last row let separate elements $x_{i}, i=\overline{0, n}$ in three categories: having characteristics $K_{n} / K_{i}$ equals to 1,0 and -1 . Note that for $i$ row the $i$ element is always of the category 1 , because its characteristic is $K_{i} / K_{i}=1$. We will multiply $X$ on the right by $\Re_{i n}(\phi), i=\overline{0, n}$ in order to have in the $n$-th row a single element of category 1 and single element of category -1 , different from 0 . All these rotations are elliptic ones. For elements of the same characteristic $x_{n i}$ and $x_{n j}$ we can use $\cos \phi=\frac{x_{n i}}{\sqrt{x_{n i}^{2}+x_{n j}^{2}}}$ and $\sin \phi=-\frac{x_{n j}}{\sqrt{x_{n i}^{2}+x_{n j}^{2}}}$. Moreover, always $x_{n n} \neq 0$.

Now, we have one element of category 1 and one of category -1 , different from zero (the $n$-th one and, for example, the $p$-th one) and element of category 1 has absolute value greater then the element of category -1 because for this lower $n$-normalized row have place equality $x_{n n}^{2}-x_{n p}^{2}=1$. It means that exists $\phi \in \mathbb{R}$ so that $\cosh \phi=\frac{x_{n n}}{\sqrt{x_{n}^{2}-x_{n p}^{2}}}$ and $\sinh \phi=-\frac{x_{n p}}{\sqrt{x_{n n}^{n}-x_{n p}^{2}}}$ and hyperbolic rotation that transforms the element of category $-1, x_{n p}$ in $x_{n p}^{\prime}=0$ and the element of category $1, x_{n n}$ in $x_{n n}^{\prime} \neq 0$.

For category 0 there exist parabolic rotations, that preserves the element of category 1 $\left(x_{n n}\right)$ and elements of category 0 transform in 0 . For this case, if one this element is on $q$-th column, $\phi=-x_{n q}$. The last non-zero element $x_{n n}$ equals to 1 or -1 , because the last row is lower $n$-normalized.

We can consider the first $n$ columns as having $n$ elements (the last one equals to zero). They form orthogonal matrix of size $n$. The last, $(n+1)$-th column (without the last element) is upper $i n$-orthogonal to first $n$ columns, $i=\overline{0, n-1}$. As these columns have $n$ elements each, $(n+1)$-th column is obligatory null (excluding the last its element).

In this stage we can consider the resulting matrix as having size $n$ instead of $n+1$ and repeat the process for it. Finally obtain the matrix $E$ which has elements on main diagonal 1 or -1 and all rest elements 0 . It is the reflection matrix on a point or line or plane or hyperplane. Obtain the equality: $X \prod_{j=1}^{q} \mathfrak{M}_{j}=E$. It's easy to see that $X=E \prod_{j=q}^{1} \mathfrak{M}_{j}^{-1}(q=(n+1) n / 2)$. Strictly speaking, the matrix $E$ can't be presented as product of rotations. In order to identify motions of $\mathbb{B}^{n}$ with orthogonal matrices, we should name $E$ (which preserve vector product) motion. However, these motions are improper (there is no continuous parameterization of motion $\mathfrak{M}(\alpha)$ on a segment $[0,1]$ such that $\mathfrak{M}(0)=I, \mathfrak{M}(1)=E$ and all $\mathfrak{M}(\alpha)$ are motions on all $\alpha \in[0,1])$. Having in expression for $X$ determinant of matrices $\mathfrak{M}_{i}$ equals to 1 and determinant of $E$ is $\pm 1$, determinant of $X$ equals $\pm 1$.

### 2.6 Coordinate and State Matrix

Consider in some space $\mathbb{B}^{n} n+1$ vectors $v_{0}, \ldots v_{n}$. Let coordinates of $v_{i}$ are $\left[v_{0 i}: \ldots: v_{n i}\right], i=\overline{0, n}$. Let vectors $v_{i}$ are ordered and form basis of $\mathbb{B}^{n}$ (not obligatory orthonormal). Compose the matrix $V$ with elements $v_{i j}, i, j=\overline{0, n}$. Will name it coordinate matrix for vectors $v_{i}$. Construct also the matrix $M$ of size $(n+1) \times(n+1)$ with elements $m_{i, j}=\frac{v_{i} \odot v_{j}}{K_{i}}$. We will name the matrix $M$ state matrix of $v_{i}$. Having $v_{i}$ is space basis, elements $m_{i j}$ are all finite. State matrix shows how orthonormal is some vector family. It tends to unite one when vectors are more normalized and orthogonal to each other ${ }^{2}$. We will demonstrate that volume of parallelepiled constructed on vectors $v_{i}$ equals to $|\operatorname{det} V|$ and

$$
\begin{equation*}
\operatorname{det} M=(\operatorname{det} V)^{2} \tag{2.20}
\end{equation*}
$$

First, let $v_{i}, i=\overline{0, n}$ are orthonormal. Then the parallelepiped volume is 1 , the matrix $V$ is orthogonal one and $\operatorname{det} V= \pm 1$. So, $|\operatorname{det} V|$ equals to parallelepiped volume. All elements on main diagonal $m_{i i}=1$, because all vectors $v_{i}$ are upper $i$-normalized. All elements above main diagonal are $m_{i j}=0, i<j$, because all vectors $v_{i}$ and $v_{j}$ are upper $i j$-orthogonal (elements under the main diagonal may differ from 0 ). It means that the matrix $M$ is lower triangular with all elements on main diagonal equls to 1 and $\operatorname{det} M=1=(\operatorname{det} V)^{2}$.

Further, note that $x \odot(y+z)=x \odot y+x \odot z,(x+y) \odot z=x \odot z+y \odot z,(\alpha x) \odot y=\alpha(x \odot y)=$ $x \odot(\alpha y), \forall x, y, z \in \mathbb{B}^{n}, \alpha \in \mathbb{R}$. Matrix determinant equals to zero if it contain proportional columns or rows. When some row or column of a matrix is multiplied by $\alpha$, the resulting matrix determinant is $\alpha$ times original matrix determinant. When some row or column is sum

[^6]of two rows / columns, then the matrix determinant equals to sum of determinants of matrices containing the first and the second row / column.

Second, let instead of some $v_{i}$ use $v_{i}^{\prime}=\alpha v_{i}$. In this case parallelepiped volume grows in $\alpha$ times and $\left|\operatorname{det} V^{\prime}\right|=|\alpha||\operatorname{det} V|$. Moreover, $\operatorname{det} M^{\prime}=\alpha^{2} \operatorname{det} M=(\alpha \operatorname{det} V)^{2}=\left(\operatorname{det} V^{\prime}\right)^{2}$.

Third, let instead of some $v_{i}$ use $v_{i}^{\prime}=v_{i}+\alpha v_{j}$. In this case parallelepiped volume remains unchanged, as well as determinant of $V$, and $\operatorname{det} M^{\prime}=\operatorname{det} M=(\operatorname{det} V)^{2}=\left(\operatorname{det} V^{\prime}\right)^{2}$.

Finally, observe, that all matrices $V$ result from orthogonal matrices using operations form the second and the third step. It means the equation (2.20) is true for all matrices and volume of parallelepiped constructed on vectors $v_{i}$ equals to $|\operatorname{det} V|$.

When somebody calculates the parallelepiped volume it's usefull to use the state matrix. Its elements don't change on motions and it is always square, even when the number of vectors is less then the space dimension (the matrix $V$ isn't square in this case).

### 2.7 Plane definition and Specification. Lineals and their Specification

Having $\mathbb{B}^{m} \subset \mathbb{R P}^{m}$, all $m$-dimensional planes $\mathbb{B}^{n}(m<n)$ lie in $\mathbb{R} \mathbb{P}^{n}$ with one global condition for vectors $X \in \mathbb{R P}^{m} \subset \mathbb{P}^{n}: X \odot X=1$. Leaving this condition (it doesn't change on any motion), we can consider $m$-dimensional planes of $\mathbb{B}^{n}$ as $m$-dimensional planes of $\mathbb{R}^{n}$.

By definition, $m$-dimensional plane $L^{m}$ results from subspace $\mathbb{B}^{m}$ on some motion. Subspace $\mathbb{B}^{m}$ has the first $m+1$ columns of unite matrix with dimension $n+1$ as its basis. Multiplying the basis matrix of $\mathbb{B}^{m}$ by some orthogonal matrix result basis matrix of $L^{m}$ as first $m+1$ columns of orthogonal matrix. Being a subspace, specification of $\mathbb{B}^{m}$ contain the first $m$ characteristics of specification $\mathbb{B}^{n}$.

What happens if we take any $m+1$ columns of some orthogonal matrix as basis? Let column indices $i_{0}, i_{1}, \ldots i_{m}$ and $i_{p}=\overline{0, n}$. It's easy to see that motions that preserve this figure only change these columns (interior figure motions) or change no these columns (motions of $\mathbb{B}^{n}$ that preserve all its points). Thus, figure characteristics $K_{p}^{\prime}=K_{i_{p}}, p=\overline{0, m}$, or $k_{p}^{\prime}=$ $K_{p}^{\prime} / K_{p-1}^{\prime}=K_{i_{p}} / K_{i_{p-1}}=\prod_{j=i_{p-1}+1}^{i_{p}} k_{j}, p=\overline{1, m}$ is its specification. These figures generally speaking are not planes. We will name them lineals. We will name planes also lineals.

It may happen, that some lineal has $K_{0}^{\prime} \neq 1$ (space and planes have it equals to 1 ). In this case lineal may not intersect the space sphere and may not have image. We will name lineals that have no image improper. Although they have no image, their properties help studying the space geometry.

One more interesting case is when the space specification has characteristic -1 and some lineal is constructed on limit vectors for this characteristic. Such lineals can't be constructed from matrices get as finite product of motions. They can be constructed as limit of infinite products. These lineal specifications can't be deduced from space specification.

For example, let space $\mathbb{B}^{2}$ has specification $\{1,-1\}$. Vectors $[0: 1: 1]$ and $[0:-1: 1]$ can't result from coordinate vectors on finite product of motions. However, there exist translations along these vectors (interior lineal translations):

$$
\mathfrak{M}=\left(\begin{array}{ccc}
1 & -2 S_{0}\left(\frac{\phi}{2}\right) & -2 S_{0}\left(\frac{\phi}{2}\right) \\
2 S_{0}\left(\frac{\phi}{2}\right) & 1-2 S_{0}^{2}\left(\frac{\phi}{2}\right) & -2 S_{0}^{2}\left(\frac{\phi}{2}\right) \\
-2 S_{0}\left(\frac{\phi}{2}\right) & 2 S_{0}^{2}\left(\frac{\phi}{2}\right) & 1+2 S_{0}^{2}\left(\frac{\phi}{2}\right)
\end{array}\right),
$$

$$
\mathfrak{W}=\left(\begin{array}{ccc}
1 & 2 S_{0}\left(\frac{\psi}{2}\right) & -2 S_{0}\left(\frac{\psi}{2}\right) \\
-2 S_{0}\left(\frac{\psi}{2}\right) & 1-2 S_{0}^{2}\left(\frac{\psi}{2}\right) & 2 S_{0}^{2}\left(\frac{\psi}{2}\right) \\
-2 S_{0}\left(\frac{\psi}{2}\right) & -2 S_{0}^{2}\left(\frac{\psi}{2}\right) & 1+2 S_{0}^{2}\left(\frac{\psi}{2}\right)
\end{array}\right) .
$$

These motions matrices use functions $C_{0}(x)$ and $S_{0}(x)$ have characteristic 0 , despite the fact the space specification doesn't contain zero. These translations are border space motions between elliptic and hyperbolic ones.

### 2.8 Projection of Vector on Lineal and on its Orthogonal Completion

We will name some vector $v^{\prime}$ projection of vector $v$ on lineal $L^{m}$, if $v^{\prime} \in L^{m}, v-v^{\prime} \perp L^{m}$. Let lineal $L^{m}$ is constructed on vectors $l_{0}, \ldots l_{m}$. Then $v^{\prime}=\sum_{i=0}^{m} \frac{v \odot l_{i}}{K_{i}} l_{i}$. Evident, $v^{\prime} \in L_{m}$. Let's see:

$$
\begin{aligned}
\left(v-v^{\prime}\right) \odot l_{j} & =\left(v-\sum_{i=0}^{m} \frac{v \odot l_{i}}{K_{i}} l_{i}\right) \odot l_{j} \\
& =v \odot l_{j}-\sum_{i=0}^{m} \frac{v \odot l_{i}}{K_{i}}\left(l_{i} \odot l_{j}\right) \\
& =v \odot l_{j}-\frac{v \odot l_{j}}{K_{j}}\left(l_{j} \odot l_{j}\right) \\
& =v \odot l_{j}-\frac{v \odot l_{j}}{K_{j}} K_{j}=0,
\end{aligned}
$$

for all $j=\overline{0, m}$, in other words, $v^{\prime \prime}=v-v^{\prime} \perp L^{m}$. When some $K_{i}=0$, expression $\frac{v \odot l_{i}}{K_{i}}$ has undefined value. It happens when some vector direction is orthogonal to all others. In this case there is impossible to determine unique orthogonal vector. However, any value of this expression, for example 0 , is valid, as it corresponds to some orthogonal vector.

### 2.9 Basis Change in Lineal. Unique Form of Lineal

Let $L^{m} \subset \mathbb{B}^{n}$ is some space lineal, defined by matrix of size $(n+1) \times(m+1)$. The matrix columns $l_{i}, i=\overline{0, m}$ form basis of lineal. Consider vector $a=\left[a_{0}: \ldots: a_{m}\right] \in L^{m}$. Let vector coordinates in $\mathbb{B}^{m}$ are $v=\left[v_{0}: \ldots: v_{n}\right]$. Then $v=L^{m} a$. Let $\mathfrak{M}$ be interior motion of lineal $L^{m}$, defined by matrix of size $(m+1) \times(m+1)$. And let coordinates of $a$ in new basis $L^{\prime m}$ are $b=\left[b_{0}: \ldots: b_{m}\right]$. Then $b=\mathfrak{M} a$. Now $v=L^{\prime m} b$. Having the fact the coordinates of vector $v$ in $\mathbb{B}^{n}$ don't change, result matrix equality:

$$
L^{m} a=v=L^{\prime m} b=L^{\prime m}(\mathfrak{M} a)=\left(L^{\prime m} \mathfrak{M}\right) a .
$$

This equality doesn't depend on vector $a$, then

$$
\begin{equation*}
L^{m}=L^{\prime m} \mathfrak{M} \tag{2.21}
\end{equation*}
$$

is equation of basis change in lineal.

It is necessary to find the unique form of lineal definition. Consider the following algorithm for the unique basis search:

1. Let $i_{p}, p=\overline{0, n}$ is basis of $\mathbb{B}^{n}$. Start with empty basis of $L^{m}$.
2. Until new basis has less then $m+1$ elements, search for $i_{p}^{\prime}$ as projection of next $i_{p}$ on $L^{m}$.
(a) If projection isn't null, find new vector $i_{p}^{\prime \prime}$ as projection of $i_{p}^{\prime}$ on orthogonal completion of existing basis $l_{i}$.
(b) If $i_{p}^{\prime \prime}$ isn't null, find its position as free index $0 \leq q \leq m$ so that $r^{2}=\frac{1}{K_{q}} i_{p}^{\prime \prime} \odot i_{p}^{\prime \prime}>0$.
(c) Norm it and add to existing basis $l_{q}=\frac{i_{p}^{\prime \prime}}{r}$.

### 2.10 Measure Calculus Between Lineals

Let $X^{p}, Y^{q}, p \leq q \leq n$ are two lineals. Let $x_{i}, i=\overline{0, p}$ is the basis of $X^{p}$. Let $X^{\prime p}$ be


Figure 2.3: Measure calculus between lineals $X^{2}$ and $Y^{2}$.
projection of $X^{p}$ on $Y^{p}$ (Figure 2.3) and let $x_{i}^{\prime}$ be projection of $x_{i}$ on $Y^{p}$ (they are not orthonormal). If the volumes of parallelepipeds constructed on vectors $x_{i}$ and $x_{i}^{\prime}$ are equals to $V_{x}$ and $V_{x}^{\prime}$ respectively and the angle between $X^{p}$ and $Y^{q}$ is measurable and equals to $\phi$, then has place the equality:

$$
V_{x}^{\prime}=V_{x} C(\phi)
$$

This equality is a particular case of (2.8), when $k_{1}=0, T(x)=x$. In our case always $k_{1}=0$, because the space model is linear. As were discussed earlier, $V_{x}=1$ and $V_{x}^{\prime}=\sqrt{\operatorname{det} M_{x}^{\prime}}$, where $M_{x}^{\prime}$ is state matrix of vectors $x_{i}^{\prime}$ :

$$
\begin{equation*}
C(\phi)=\sqrt{\operatorname{det} M_{x}^{\prime}} \tag{2.22}
\end{equation*}
$$

It may happen that characteristic of $\phi$ equals to zero and we can't calculate $C^{-1}(\phi)$. If we project vectors $x_{i}$ on $Y_{\perp}^{q}$ orthogonal completion of $Y^{q}$ (suppose it has dimension at least $p$ ), get $X^{\prime \prime p}$ constructed on vectors $x_{i}^{\prime \prime}$ with the volume $V_{x}^{\prime \prime}$, then by (2.13) get:

$$
V_{x}^{\prime \prime}=V_{x} S(\phi)
$$

Or, having $M_{x}^{\prime \prime}$ state matrix for vectors $x_{i}^{\prime \prime}$,

$$
\begin{equation*}
S(\phi)=\sqrt{\operatorname{det} M_{x}^{\prime \prime}} \tag{2.23}
\end{equation*}
$$

If the dimension of $Y_{\perp}^{q}$ is less then $p$, then we can get $S(\phi)$ by projecting of $Y_{\perp}^{q}$ on $X^{p}$. If $\phi$ isn't measurable, then the angle $\psi$ between $X^{p}$ and $Y_{\perp}^{q}$ is measurable and:

$$
\begin{align*}
& S(\psi)=\sqrt{\operatorname{det} M_{x}^{\prime}}  \tag{2.24}\\
& C(\psi)=\sqrt{\operatorname{det} M_{x}^{\prime \prime}} \tag{2.25}
\end{align*}
$$

The angles $\phi$ and $\psi$ present measure between lineals $X^{p}$ and $Y^{q}$. Having values $C(\phi), S(\phi)$ and $C(\psi), S(\psi)$, it is possible to determine $\phi$ and $\psi$. The measure characteristic of $\phi$ and $\psi$ equals to measure characteristic between $X^{\prime p}$ and $X^{\prime \prime p}$. Depending on this characteristic, situation can be one of the following:

- If characteristic equals to 1 , then $\operatorname{det} M_{x}^{\prime}+\operatorname{det} M_{x}^{\prime \prime}=1$ and $\phi=\tan ^{-1} \sqrt{\frac{\operatorname{det} M_{x}^{\prime \prime}}{\operatorname{det} M_{x}^{\prime}}}, \psi=$ $\tan ^{-1} \sqrt{\frac{\operatorname{det} M_{x}^{\prime}}{\operatorname{det} M_{x}^{\prime \prime}}}$.
- If characteristic equals to 0 , then either $\operatorname{det} M_{x}^{\prime}=1$ and $\phi=\sqrt{\operatorname{det} M_{x}^{\prime \prime}}, \psi=\infty$, or $\operatorname{det} M_{x}^{\prime \prime}=1$, and $\phi=\infty, \psi=\sqrt{\operatorname{det} M_{x}^{\prime}}$.
- If characteristic equals to -1 , then either $\operatorname{det} M_{x}^{\prime}-\operatorname{det} M_{x}^{\prime \prime}=1$ and $\phi=\tanh ^{-1} \sqrt{\frac{\operatorname{det} M_{x}^{\prime \prime}}{\operatorname{det} M_{x}^{\prime}}}$, $\psi$ isn't measurable, or $\operatorname{det} M_{x}^{\prime \prime}-\operatorname{det} M_{x}^{\prime}=1$ and $\phi$ isn't measurable, $\psi=\tanh ^{-1} \sqrt{\frac{\operatorname{det} M_{x}^{\prime}}{\operatorname{det} M_{x}^{\prime \prime}}}$, or $\operatorname{det} M_{x}^{\prime \prime}=\operatorname{det} M_{x}^{\prime}$ and $\phi=\psi=\infty$.


### 2.11 Volume Calculation

We can see that for any $\mathbb{B}^{n}$ seen as a unit sphere in $\mathbb{R}^{n+1}$, the surface is orthogonal to radius. Let $X, Y \in \mathbb{B}^{n}$ and the distance between $X$ and $Y$ is small. Let $O=(0,0, \ldots, 0)$ is origin of $\mathbb{R}^{n+1}$. We will see that $(O-X) \odot(Y-X)=0$ when $Y \rightarrow X$ in sense of distance between them. $O-X=-X,(O-X) \odot(Y-X)=-X \odot(Y-X)=X \odot X-X \odot Y=1-C_{1}(d(X, Y))$, where $d(X, Y)$ is distance between $X$ and $Y$. When $Y \rightarrow X, d(X, Y) \rightarrow 0$ and $1-C_{1}(d(X, Y)) \rightarrow 0$.

Let $A, B \in \mathbb{B}^{1} . A=\left[C_{1}(\alpha): S_{1}(\alpha)\right], B=\left[C_{1}(\beta): S_{1}(\beta)\right]$, where $C_{1}(x), S_{1}(x)$ and $T_{1}(x)$ are defined as (1.2), (1.3) and (1.4). Let calculate in $\mathbb{R}^{2}$ the area of $\mathbb{B}^{1}$ sector between $A$ and B. In Euclidean polar system the argument $\tan \phi_{e}=y / x=S_{1}(\phi) / C_{1}(\phi)=T_{1}(\phi)$, where $\phi$ is native argument in $\mathbb{B}^{1}$. The Euclidean radius $\rho=\sqrt{x^{2}+y^{2}}=\sqrt{C_{1}^{2}(\phi)+S_{1}^{2}(\phi)}$. Having $d \phi_{e}=\frac{d \phi}{\left(1+T_{1}^{2}(\phi)\right) C_{1}^{2}(\phi)}$, the area is:

$$
\begin{aligned}
S & =\frac{1}{2} \int_{A}^{B} \rho\left(\phi_{e}\right)^{2} d \phi_{e}=\frac{1}{2} \int_{A}^{B}\left(C_{1}^{2}(\phi)+S_{1}^{2}(\phi)\right) \frac{d \phi}{C_{1}^{2}(\phi)\left(1+T_{1}^{2}(\phi)\right)} \\
& =\frac{1}{2} \int_{A}^{B} \frac{C_{1}^{2}(\phi)+S_{1}^{2}(\phi)}{C_{1}^{2}(\phi)+S_{1}^{2}(\phi)} d \phi=\frac{1}{2} \int_{A}^{B} d \phi=\left.\frac{1}{2} \phi\right|_{\alpha} ^{\beta}=\frac{\beta-\alpha}{2} .
\end{aligned}
$$

That is, $2 S$ equals to length $A B$.
Let $F \subset \mathbb{B}^{n}$ be some figure with volume (in sense of $\mathbb{B}^{n}$ ) $V_{\mathbb{B}}$. We will name $V_{\mathbb{R}}$ the volume (in sense of $\mathbb{R}^{n+1}$ ) of cone with base $F \subset \mathbb{B}^{n}$ and vertex $O \notin \mathbb{B}^{n}$ origin of $\mathbb{R}^{n+1}$ (Figure


Figure 2.4: Figure $F \subset \mathbb{B}^{n}$ volume calculation with aid of cone in $\mathbb{R}^{n+1}$.
2.4). As $\mathbb{B}^{n}$ is orthogonal to radius, $F$ also is orthogonal. The radius equals 1 , because $\forall X \in \mathbb{B}^{n}, X \odot X=1$. Then for each figure $F \subset \mathbb{B}^{n}$ have place equality:

$$
V_{\mathbb{B}}=(n+1) V_{\mathbb{R}}
$$

As motions preserve $\mathbb{B}^{n}$ and the absolute value of their matrices' determinant is 1 , all motions preserve $V_{\mathbb{R}}$ and thus, they preserves also $V_{\mathbb{B}}$.

## Chapter 3

## Theory Application

### 3.1 Space and Lineal Specification Search Algorithm

As we can see, the theory described in this book is universal and easy applicable. However, one issue stops somebody from using it. Geometric spaces are classified and defined different from the way adopted here. Therefore, in order to not loose the feeling of reality, we will describe an algorithm aimed to find specification for some geometric space. The algorithm can be applied to any space where have sense notions of points, lines, planes, subspaces, distances, angles and / or motions.

1. Let $m$ equals to the greatest number of general situated points, or same, the lowest number of vertices in a polyhedron of positive volume.
2. Count space dimension as $n=m-1$.
3. Name points 0 -dimensional planes and lines 1-dimensional planes.
4. For $i=\overline{1, n}$ do:
(a) If among $(i-1)$-dimensional planes there are non-congruent ones, then the space definition or space terminology is inconsistent. Theory still can be used, however, in order to understand it correctly, it is necessary to modify terminology or to define otherwise some space elements (about it later).
(b) If the measure between $(i-1)$-dimensional planes is bounded, then $k_{i}=1$.
(c) If the measure between $(i-1)$-dimensional planes is scalable, then $k_{i}=0$.
(d) Otherwise, $k_{i}=-1$.
5. Having space dimension $n$ and specification $\left\{k_{1}, \ldots, k_{n}\right\}$, use theory.

The necessity of proper terminology, uniform among all spaces is required by wish to have such a theory, that isn't misleading and helps us to study the space structure and to compare it with other spaces. Still, under inconsistent theory / terminology we should understand it has a contradiction, but failing it to match to theory / terminology that is common today. We assume the following here:

- All the planes of any dimension are congruent, including points and lines.
- Theory allows the duality principle of ( $m-1$ )-dimensional planes and $(n-m)$-dimensional ones.

We should mention that 'common terminology' may change over the time. In order to understand what it is consider an example of inconsistent terminology. The Minkowskii space is successfully used in physics to describe the theory of relativity. Unfortunately, from geometry point of view, it have no proper terminology. The notions of 'space-like lines', 'time-like lines' and 'light-like lines' have sense in physics, but not in geometry. Corresponding geometric notions are: 'I-st category lines', 'II-nd category lines' and 'III-rd category lines'. No space motion maps some line of a category into some line of another category. There is no contradiction here, but there is an inconsistence. What happens if somebody wants to define a space with five categories of lines ${ }^{1}$ ? Nobody defines several kinds of points. All points are congruent ${ }^{2}$. Why shouldn't be lines all congruent? At the other hand, relative position of points may differ. If we name lines only the I-st category of lines, then we should exclude II-nd and III-rd category of lines from lines. At the first look, it conflicts with the axiom that claims any two points can be connected with a line. But this axiom may have no place in other spaces. In contrast, just Euclidean geometry, where all points are connectable, gives us an example of parallel lines (that have no common point). Using the duality principle, it should exist the notion of non-connectable points (that have no common line).

It should be mention, that even for somebody feels comfortable using this theory, the algorithm described earlier may help to determinate the specification of some exotic lineals (for example, of ones defined as limit lineals, which aren't deductable from the space specification).

### 3.2 Some Special Spaces

Many linear spaces are defined using the quadric form of distance $d^{2}(X, Y)=(Y-X) \odot(Y-X)$. As for these spaces $k_{1}=0, K_{0}=1, K_{i}=0, i>0$. In this case, the equality $1=C(d(X, Y))=$ $X \odot Y=1$ is trivial and can't be used for distance calculation. Consider one more vector product $-\otimes$ such as $(X \odot Y)^{2}+k_{1}(X \otimes Y)^{2}=1, \forall X, Y \in \mathbb{B}^{n}, k=\{-1,0,1\}$. This product

[^7]is similar to exterior vector product. Change $1=(X \odot X)(Y \odot Y)$ :
\[

$$
\begin{aligned}
(X \otimes Y)^{2} & =\frac{1}{k_{1}}\left((X \odot X)(Y \odot Y)-(X \odot Y)^{2}\right) \\
& =\frac{1}{k_{1}}\left(\left(\sum_{i=0}^{n} K_{i} x_{i}^{2}\right)\left(\sum_{j=0}^{n} K_{j} y_{j}^{2}\right)-\left(\sum_{i=0}^{n} K_{i} x_{i} y_{i}\right)\left(\sum_{j=0}^{n} K_{j} x_{j} y_{j}\right)\right) \\
& =\frac{1}{k_{1}} \sum_{i=0}^{n} \sum_{j=0}^{n} K_{i} K_{j}\left(x_{i}^{2} y_{j}^{2}-x_{i} x_{j} y_{i} y_{j}\right) \\
& =\frac{1}{k_{1}} \sum_{i<j=0}^{n} K_{i} K_{j}\left(x_{i}^{2} y_{j}^{2}+x_{j}^{2} y_{i}^{2}-2 x_{i} x_{j} y_{i} y_{j}\right) \\
& =\frac{1}{k_{1}} \sum_{i<j=0}^{n} K_{i} K_{j}\left(x_{i} y_{j}-x_{j} y_{i}\right)^{2}
\end{aligned}
$$
\]

So,

$$
\begin{equation*}
X \otimes Y=\sqrt{\frac{1}{k_{1}} \sum_{i<j=0}^{n} K_{i} K_{j}\left(x_{i} y_{j}-x_{j} y_{i}\right)^{2}} \tag{3.1}
\end{equation*}
$$

Note that from $\mathbb{B}^{n}=\left\{x \in \mathbb{R P}^{n} \mid x \odot x=x_{0}^{2}=1\right\}$ result $x_{0}=1$ or $x_{0}=-1$. As $x \in \mathbb{B}^{n}$ implies $-x \in \mathbb{B}^{n}$ we can consider $x_{0}=1$. We will use $\otimes$ operator for distance $d(X, Y)$ between points $X$ and $Y$. Note that having $C^{2}(x)+k_{1} S^{2}(X)=1, \forall x \in \mathbb{R}, k_{1}=\{-1,0,1\}$ and $(X \odot X)^{2}+k_{1}(X \otimes Y)^{2}=1, \forall X, Y \in \mathbb{B}^{n}, k_{1}=\{-1,0,1\}$ result $S(d(X, Y))=X \otimes Y, \forall X, Y \in$ $\mathbb{B}^{n}$ :

$$
d^{2}(X, Y)=S^{2}(d(X, Y))=(X \otimes Y)^{2}=\frac{1}{k_{1}} \sum_{i<j=0}^{n} K_{i} K_{j}\left(x_{i} y_{j}-x_{j} y_{i}\right)^{2}
$$

In this sum all non-zero terms are those for which $i=0$ :

$$
\begin{aligned}
d^{2}(X, Y) & =\frac{1}{k_{1}} \sum_{j=1}^{n} K_{j}\left(x_{0} y_{j}-x_{j} y_{0}\right)^{2} \\
& =\sum_{j=1}^{n} \frac{K_{j}}{k_{1}}\left(y_{j}-x_{j}\right)^{2} \\
& =\sum_{j=1}^{n}\left(y_{j}-x_{j}\right)^{2} \prod_{p=2}^{n} k_{p}
\end{aligned}
$$

In this equality don't appear $x_{0}$ or $y_{0}$. We can consider $\mathbb{B}^{n}$ a hyperplane of $\mathbb{R}^{n+1}$ with equation $x_{0}=1$ and specification $\left\{k_{2}, \ldots k_{n}\right\}$. We can identify it with $\mathbb{R}^{n}$. Then the equality above is equivalent to $(Y-X) \odot(Y-X)$.

It means firstly, that scalar product of vestors in lnear spaces $\left(k_{1}=0\right)$ induces the same metrics that is used in the model, and secondly, that non-linear spaces with specification $\left\{k_{1}, k_{2}, \ldots k_{n}\right\}\left(k_{1} \neq 0\right)$ are best approximated by linear spaces with specification $\left\{0, k_{2}, \ldots k_{n}\right\}$. Note also that from here deduce that non-linear space with specification $\left\{k_{1}, \ldots k_{n}\right\}$ is enclosed in model meta-space of greater by one dimension, of which specification is $\left\{0, k_{1}, \ldots k_{n}\right\}$.

We can use this quadric form to search for all characteristics except $k_{1}=0$. We will use this method in order to describe some special spaces by specifying their specifications.

Case 1. Elliptic, Euclidean and Hyperbolic Spaces. Elliptic, linear (Euclidean) and hyperbolic (Bolyai-Lobachevsky) spaces have characteristic $k_{1}$ equals to sign of space curvature $k_{1}=1$ for elliptic space, $k_{1}=0$ for linear space and $k_{0}=-1$ for hyperbolic space.

All these spaces are usually approximated by Euclidean one. We can calculate the rest of characteristics using euclidean quadric form. Let dimension is 3:

$$
\begin{aligned}
d(X, Y)^{2} & =\left(y_{1}-x_{1}\right)^{2}+\left(y_{2}-x_{2}\right)^{2}+\left(y_{3}-x_{3}\right)^{2} \\
& =\left(y_{1}-x_{1}\right)^{2}+k_{2}\left(y_{2}-x_{2}\right)^{2}+k_{2} k_{3}\left(y_{3}-x_{3}\right)^{2}
\end{aligned}
$$

so $k_{2}=1$ and $k_{2} k_{3}=1, k_{3}=1$.

Case 2. Minkowskii Space. The distance between $X$ and $Y$ is calculated (for time-like vectors) as

$$
\begin{aligned}
d^{2}(X, Y) & =\left(y_{1}-x_{1}\right)^{2}-\left(y_{2}-x_{2}\right)^{2}-\left(y_{3}-x_{3}\right)^{2}-\left(y_{4}-x_{4}\right)^{2} \\
& =\left(y_{1}-x_{1}\right)^{2}+k_{2}\left(y_{2}-x_{2}\right)^{2}+k_{2} k_{3}\left(y_{3}-x_{3}\right)^{2}+k_{2} k_{3} k_{4}\left(y_{4}-x_{4}\right)^{2}
\end{aligned}
$$

where coordinate 1 is time-like and coordinates 2,3 and 4 are space-like. So $k_{2}=-1$, $k_{2} k_{3}=-1, k_{3}=1$ and $k_{2} k_{3} k_{4}=-1, k_{4}=1$. As Minkowskii space is linear, $k_{1}=0$.

If we introduce curvature in space, its structure changes. For example, let $k_{1}=1$. Then

$$
\begin{aligned}
X \odot Y & =x_{0} y_{0}+k_{1} x_{1} y_{1}+k_{1} k_{2} x_{2} y_{2}+k_{1} k_{2} k_{3} x_{3} y_{3}+k_{1} k_{2} k_{3} k_{4} x_{4} y_{4} \\
& =x_{0} y_{0}+x_{1} y_{1}-x_{2} y_{2}-x_{3} y_{3}-x_{4} y_{4}
\end{aligned}
$$

so time characteristic becomes elliptic and space characteristic becomes hyperbolic. If $k_{1}=1$, then

$$
\begin{aligned}
X \odot Y & =x_{0} y_{0}+k_{1} x_{1} y_{1}+k_{1} k_{2} x_{2} y_{2}+k_{1} k_{2} k_{3} x_{3} y_{3}+k_{1} k_{2} k_{3} k_{4} x_{4} y_{4} \\
& =x_{0} y_{0}-x_{1} y_{1}+x_{2} y_{2}+x_{3} y_{3}+x_{4} y_{4}
\end{aligned}
$$

and time characteristic becomes hyperbolic and space characteristic becomes elliptic.

Case 3. Minkowskii Space with 2-dimensional Time. Consider a 4-dimensional space with distance quadric form that has 2 positive signs and 2 negative. This space is sometimes named Minkowskii space with 2-dimensional time:

$$
\begin{aligned}
d^{2}(X, Y) & =\left(y_{1}-x_{1}\right)^{2}+\left(y_{2}-x_{2}\right)^{2}-\left(y_{3}-x_{3}\right)^{2}-\left(y_{4}-x_{4}\right)^{2} \\
& =\left(y_{1}-x_{1}\right)^{2}+k_{2}\left(y_{2}-x_{2}\right)^{2}+k_{2} k_{3}\left(y_{3}-x_{3}\right)^{2}+k_{2} k_{3} k_{4}\left(y_{4}-x_{4}\right)^{2}
\end{aligned}
$$

So $k_{2}=1, k_{2} k_{3}=-1, k_{3}=-1$ and $k_{2} k_{3} k_{4}=-1, k_{4}=1$. As for all lineal spaces, $k_{1}=0$ for it.

Case 4. Spaces with Degenerate Distance Quadric Form. Consider linear 4-dimensional space ( $k_{1}=0$ ) with degenerate distance quadric form:

$$
\begin{aligned}
d^{2}(X, Y) & =\left(y_{1}-x_{1}\right)^{2}+\left(y_{2}-x_{2}\right)^{2}+\left(y_{3}-x_{3}\right)^{2} \\
& =\left(y_{1}-x_{1}\right)^{2}+k_{2}\left(y_{2}-x_{2}\right)^{2}+k_{2} k_{3}\left(y_{3}-x_{3}\right)^{2}+k_{2} k_{3} k_{4}\left(y_{4}-x_{4}\right)^{2}
\end{aligned}
$$

so $k_{2}=k_{3}=1$ and $k_{4}=0$.
It means that motions:

$$
\begin{aligned}
x_{1}^{\prime} & =x_{1} \\
x_{2}^{\prime} & =x_{2} \\
x_{3}^{\prime} & =x_{3} \\
x_{4}^{\prime} & =\phi_{1} x_{1}+\phi_{2} x_{2}+\phi_{3} x_{3}+x_{4}
\end{aligned}
$$

are all valid.
However, transformation:

$$
\begin{aligned}
x_{1}^{\prime} & =x_{1} \\
x_{2}^{\prime} & =x_{2} \\
x_{3}^{\prime} & =x_{3} \\
x_{4}^{\prime} & =\phi_{4} x_{4}
\end{aligned}
$$

is not a motion. Although it preserved distance, it doesn't preserve volume except $\phi_{4}=1$ or -1 . It is an example of angle scaling.

### 3.3 Spaces as Product of their Subspaces

Another way to define spaces is by product of their subspaces. It is necessary to be accurate here. The geometric space isn't only a structure of points. It is also the structure of all its subspaces. It is mistake to think that having $\mathbb{R}^{1}$ is isomorphic to one-dimensional Euclidean space $\mathbb{E}^{1}$, from $\mathbb{R}^{1} \times \mathbb{R}^{1}=\mathbb{R}^{2}$ results $\mathbb{E}^{1} \times \mathbb{E}^{1}=\mathbb{E}^{2}$ (using specification notation, $\{0\} \times\{0\}=\{0$, $1\})$. The problem is the product doesn't define way to measure the angle between multiplied subspaces. It can be defined in several ways, for example, $\{0,0\}$ or $\{0,-1\}$.

The situation is even worse when multiplied subspaces with different specification $X^{m}$ and $Y^{n}$. One-dimensional images can be constructed in two ways: $X^{1} \times Y^{0}$ (isomorphic to $X^{1}$ ) and $X^{0} \times Y^{1}$ (isomorphic to $Y^{1}$ ). And if $X^{1}$ and $Y^{1}$ have different specifications, these two onedimensional lines aren't congruent. For example, if somebody wants to construct geometry on a cylinder, first thing he or she thinks of is $\mathbb{S}^{1} \times \mathbb{E}^{1}(\{1\} \times\{0\})$, where $\mathbb{S}^{1}$ is one-dimensional elliptic space. In this case some lines are circles, some are lines and others are right and left helices, that may not intersect, intersect in one point or intersect in infinity of points. As an example of complete geometry on a cylinder you can take the space with specification $\{1,0\}$.

Additionally, one should not consider that if from algebraic point of view $\mathbb{E}^{1}$ is isomorphic to $\mathbb{H}^{1}$ (one-dimensional hyperbolic space), then constructions like $\mathbb{H}^{2} \times \mathbb{E}^{1}(\{-1,1\} \times\{0\})$ and $\mathbb{H}^{2} \times \mathbb{H}^{1}(\{-1,1\} \times\{-1\})$ are also isomorphic. From geometric point of view, $\mathbb{E}^{1}$ is scalable, while $\mathbb{H}^{1}$ is not (the mutual departure of points is possible, however it can't be linear).

In contrast, it is possible to construct spaces with specifications $\{-1,1,0\}$ and $\{-1,1,-1\}$, which differ one from another by the fact that in first one on a twodimensional plane doesn't containing some point there is the only point that isn't connectable with it, and for the second space the number of such a points is infinity.

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[^0]:    ${ }^{1} \mathrm{http}: / /$ sourceforge.net/projects/geomspace/. The last version of this book can also be downloaded form here.

[^1]:    ${ }^{1}$ Here and further we will consider for simplicity that $k^{0}=1$ for $k=0$ too. We will say $x$ divide $k^{i}, k=0$ if in expression $x / k^{i}$ the exponent of $k$ in numerator is not less then $i$.

[^2]:    ${ }^{2}$ A fixed point is present for example in translation on elliptic plane.
    ${ }^{3}$ For this case it is necessary to modify another two postulates, namely that from any three points on a line exactly one lies between two others, and that any line can be extended infinitely in any direction.

[^3]:    ${ }^{4}$ It conflicts with axiom which states that through any two points goes a line. This axiom should be changed by one of the following in order to consider the geometries with non-elliptic rotations.

[^4]:    ${ }^{5}$ In this case there exist a measure $X Y$, but it may have different characteristic then distance. We will name this measure also distance, keeping in mind that it is generalized distance.

[^5]:    ${ }^{1}$ consider $\sqrt{0}=0$.

[^6]:    ${ }^{2}$ If space specification contains null characteristics, some elements of state matrix can have any value, even for orthonormal vector family.

[^7]:    ${ }^{1}$ Depending on concrete space, there can exist more categories of two-dimensional planes. For further dimensions of planes the number of their categories grows.
    ${ }^{2}$ The notion of points on infinity is used in projective geometry. These points are non-congruent with others. The terminology is not common in a scope of analytic geometry.

