1. THE REAL PROJECTIVE PLANE

§1.1 The Real Affine Plane

The Euclidean Plane involves a lot of things that can be measured, such as distances, angles and areas. This is referred to as the metric structure of the Euclidean Plane. But underlying this is the much simpler structure where all we have are points and lines and the relation of a point lying on a line (or equivalently a line passing through a point). This relation is referred to as the incidence structure of the Euclidean Plane.

The Real Affine Plane is simply the Euclidean Plane stripped of all but the incidence structure. This eliminates any discussion of circles (these are defined in terms of distances) and trigonometry (these need measurement of angles). It might seem that there’s nothing left but this isn’t the case.

There is, of course, the concept of collinearity. Three or more points are collinear if there is some line that passes through them all. Certain theorems of affine geometry state that if there’s such and such a configuration and certain triples of points are collinear then a certain other triple is collinear. Another affine concept is that of concurrence – three or more lines passing through a single point.

Then there’s the concept of parallelism, though we can’t define it in terms of lines having a constant distance between them or lines having the same slope.

Definition: Two lines \( h, k \) are parallel in the Real Affine Plane if either they coincide or they don’t intersect (meaning that no point lies on them both).

Here are two basic properties of the Real Affine Plane:
(1) Given any two distinct points there is exactly one line passing through both.
(2) Given any two distinct lines there is at most one point lying on both.

There’s an obvious similarity between these two properties. But while we can say “exactly one” in property (1) the best we can do in property (2) is “at most one” because there is no common point when the lines are parallel.

This is reminiscent of the situation with the real number system, where every non-zero real number has at most two square roots. Why not exactly two? The reason is that negative numbers don’t have any square roots.

This was considered to be a defect of the real number system and so imaginary numbers were invented to provide square roots for negative numbers. The field of real numbers was expanded to form the complex numbers, and for this system we can say that every non-zero number has exactly two square roots.

We do the same thing with the Real Affine Plane. We invent “imaginary” points where parallel lines can meet. We call these extra points, ideal points. And we invent an “imaginary” line called the ideal line, and specify that all the ideal points lie on this line. The Real Affine Plane is thereby extended to the Real Projective Plane. In the Real Projective Plane there are no longer such things as distinct parallel lines since every pair of distinct lines
now intersect in exactly one point. This then mirrors exactly the fact that every pair of distinct points lie on exactly one line.

§1.2. Intuitive Construction of the Real Projective Plane
We start with the Real Affine Plane.

(This trapezium is merely a representative picture of the affine plane. If we drew an accurate picture it would cover the whole page, and then some! There would be no space for any extra points, let alone the text. You should view this as a perspective view of a very large rectangle and then pretend that it extends infinitely far in all directions.)

We call the points on the Real Affine Plane **ordinary points**.

We call the lines on the Real Affine Plane **ordinary lines**.

We sort these ordinary lines into parallel classes.

A **parallel class** consists of a line together with all lines parallel to it.
For each parallel class we invent a new point, called an ideal point.

These ideal points don’t lie on the Real Affine Plane. Where are they then? The answer is simply “in our minds”. However, to assist our imagination, we can put these ideal points on our diagram outside of the shape that represents the Real Affine Plane.

As well as ordinary points lying on ordinary lines in the usual way we decree that all lines in a given parallel class (and no others) pass through the corresponding ideal point.

We also invent a new line called the ideal line and decree that this line passes through all the ideal points (and no others).
The resulting geometry is called the **Real Projective Plane**. It contains all of the real affine plane, as well as the ideal points and the ideal line. Any theorem that we can prove for the real projective plane will be true for the real affine plane simply by taking the points and lines to be ordinary ones.

§1.3. The Real Projective Plane is Complete

**Theorem 1A:** In the real projective plane:

(i) any two distinct points lie on exactly one line;
(ii) any two distinct lines intersect in exactly one point.

**Proof:** (i) **Case I:** two ordinary points lie on an ordinary line, as in the affine plane.

**Case II:** an ordinary point P and an ideal point Q lie on the line through P parallel to the lines in the parallel class corresponding to Q. (Remember that in Euclidean Geometry there’s a unique line which passes through a given point and is parallel to a given line.)

**Case III:** Two ideal points lie on the ideal line.

(ii) **Case I: Two ordinary lines:**
If these lines are non-parallel they intersect in an ordinary point.

If they’re parallel they intersect in the ideal point that corresponds to their parallel class.
Case II: An ordinary line and the ideal line:
An ordinary line intersects the ideal line in the ideal point corresponding to the parallel class in which it lies.

§1.4. The Artist’s View of the Real Projective Plane

Renaissance artists had no problem with the concept of parallel lines meeting at a point. This happens all the time in a perspective drawing.

Consider what an artist does when he sketches a scene. (If this sounds sexist – why not “she” – allow me to point out this is a Renaissance artist, and they were mostly male!) You might think that he represents points in the scene by points on the canvas, but it would be more accurate to say that he represents rays not points. Every ray emanating from his eye (it was once thought that sight traveled from the eye to the objects being seen) corresponds to a single point on his canvas.

This leads to the next way of thinking about the Real Projective Plane. Consider the 3-dimensional Euclidean space, $\mathbb{R}^3$.

Definition: A projective point is a line through the origin. A projective line is a plane through the origin.

It might seem strange at first to call a line a point and to call a plane a line but to the artist, a line in 3-dimensional space, passing through his viewpoint, is what he depicts as a
point on his canvas. [This is not completely true, in that an artist (unless he has eyes in the back of his head) only depicts rays (half-lines) not whole lines.]

\[ \text{rays} \quad \text{lines} \]

The Real Projective Plane is defined as the set of all projective points and all projective lines. It can be thought of as a sort of porcupine in 3-dimensional space, bristling with lines going in all directions through the origin. To complete the description we must define incidence, that is, we must explain what it means for a projective point to lie on a projective line.

A projective point \( P \) (line through the origin) \text{lies on} a projective line \( h \) (plane through the origin) if, when considered as a line, it lies on \( h \), considered as a plane.
A projective line \( h \) \text{passes through} a projective point \( P \) if \( P \) lies on \( h \).
Three (or more) projective points are \text{collinear} if they lie on a common projective line.
Three (or more) projective lines are \text{concurrent} if they pass through the same projective point.

\textbf{Theorem 1B:} In the Real Projective Plane:
(i) any two distinct projective points lie on exactly one projective line;
(ii) any two distinct projective lines intersect in exactly one projective point.

\textbf{Proof:} Translated into ordinary terms these state that:
(i) any two lines through the origin lie on exactly one plane through the origin and
(ii) any two planes through the origin intersect in exactly one line.
Both of these are clearly true statements for 3-dimensional Euclidean Space.

\textbf{§1.5. Embedding the Real Affine Plane in the Real Projective Plane}

It’s all very well to have constructed a geometry whose incidence structure is complete, but is it really the same structure that we created earlier, by extending the Real Affine Plane by adding ideal points? After all, with this new version, which projective points are the ideal points? With this bunch of lines through the origin they all seem pretty much the same.

Well, suppose we take an ordinary Real Affine Plane \( \pi \) in \( \mathbb{R}^3 \), one that doesn’t pass through the origin. We can think of the origin as the artist’s eye and the plane as his (infinite) canvas. Every (ordinary) point \( P \) on \( \pi \) corresponds to a unique line through the origin, namely the line \( OP \). But this is a projective point according to our second point of view.
But if we take the totality of affine points the corresponding projective points (lines through O) do not use up all the available lines through the origin. Only a projective point that, viewed as a line through the origin, cuts \( \pi \) will correspond to a point on \( \pi \) (viz. the intersection of that line with \( \pi \)). Left over will be the projective points (lines through O) that are parallel to \( \pi \). These are in fact the ideal points in the real projective plane.

Suppose we have two ordinary parallel lines on \( \pi \). They don’t intersect on \( \pi \). But the corresponding projective lines do intersect. These corresponding projective lines will be the planes through the origin that pass through the respective lines on \( \pi \). These lines will intersect in a line (just as two pages of a book intersect in the spine of the book).

This line will pass through O so it can be considered as a projective point. But it will be parallel to \( \pi \) and hence won’t correspond to an ordinary point. So what were two parallel, non-intersecting lines in the ordinary plane \( \pi \), correspond to two projective lines (planes through O) that do intersect – in an ideal point (line through O). The ideal points are in fact the lines through O that are parallel to \( \pi \) and the ideal line is thus the plane through O that is parallel to \( \pi \).

In this way we have, in effect, completed the Real Affine Plane to produce the Real Projective Plane. But notice that the plane that becomes the ideal line depends on which affine plane we took in \( \mathbb{R}^3 \). In fact any projective line can become the ideal line if we take a suitable affine plane \( \pi \) and a suitable origin O.

This is important. The distinction between ordinary point and ideal point is not one that is intrinsic to the Real Projective Plane itself. It only reflects the way we view the Real Projective Plane from the point of view of a Real Affine Plane.

The beauty of the Real Projective Plane is that it is wonderfully uniform. We can prove Euclidean theorems by projective means without the messiness of having to consider cases where lines are parallel and where lines are not parallel. In the Real Projective Plane, all points are equivalent and all lines are equivalent.
Example 1: Suppose we take the plane \( \pi \) to be \( z = 1 \).

(i) Which of the axes (x-, y- and z-) are ideal points and which are ordinary (relative to \( \pi \))? 
(ii) What is the projective point corresponding to the ordinary point \((1, 1, 1)\) on \( \pi \)?
(iii) What point on \( \pi \) corresponds to the projective point \( x = 2y = 3z \)?
(iv) Relative to which planes is \( x + y + z = 0 \) the ideal line?

Solution: (i) The x- and y-axes are ideal, the z-axis is ordinary.
(ii) The line \( x = y = z \);
(iii) \((3, 3/2, 1)\) (this is where the line \( x = 2y = 3z \) intersects \( \pi \));
(iv) any plane of the form \( x + y + z = c \) for \( c \neq 0 \).

§1.6. Review of Relevant Linear Algebra

We’re going to make this last view of the Real Projective Plane more rigorous by introducing linear algebra. Instead of talking about projective points as lines through the origin we shall think of them as one-dimensional subspaces of \( \mathbb{R}^3 \). This will free us from needing geometric intuition, and allow us to use algebraic methods. But also we can change the underlying field of our vector space, giving rise to other projective planes – even finite ones. Before we do this we will review the relevant linear algebra.

A vector space over a field \( F \) is a set (with the elements being called vectors) together with operations of addition and scalar multiplication (a scalar being an element of \( F \)) such that certain axioms are satisfied. Vectors are usually printed in bold type, such as \( \mathbf{v} \), to distinguish them from scalars (but there are times when this distinction can’t be maintained).

Practically any mathematical object can be an element of a useful vector space. You can have vector spaces of numbers, of matrices, and of functions. But here we need consider only those vectors of the form \((x_1, \ldots, x_n)\), and then only for the case \( n = 3 \).

So our vectors will all have the form \((x, y, z)\) with the components \( x, y, z \) coming from some field. We’ll mainly use the field \( \mathbb{R} \), of real numbers, though at one point we’ll use various finite fields such as \( \mathbb{Z}_p \), the field of integers modulo a prime \( p \).

Vector addition and scalar multiplication are defined in the usual way, component by component.

A linear combination of the vectors \( \mathbf{v}_1, \ldots, \mathbf{v}_n \) is a vector of the form \( \lambda_1 \mathbf{v}_1 + \cdots + \lambda_n \mathbf{v}_n \) for some scalars \( \lambda_1, \ldots, \lambda_n \).

A linear combination is non-trivial if at least one of the scalars is non-zero. A set of vectors \( \{\mathbf{v}_1, \ldots, \mathbf{v}_n\} \) is linearly independent if \( \lambda_1 \mathbf{v}_1 + \cdots + \lambda_n \mathbf{v}_n = 0 \) implies that each \( \lambda_i = 0 \). Otherwise they are linearly dependent.

Note that it’s the set that’s linearly dependent or independent, not the vectors themselves. In other words it refers to the relationship between the vectors. A set of vectors is linearly dependent if there exists a non-trivial linear combination of them that is equal to zero. In such a case one of the vectors can be written as a linear combination of the others. If they’re linearly independent this isn’t possible.

Example 2: Are the vectors \((1, 5, -2), (6, -1, 9), (9, -17, 24)\) linearly independent?

Solution: Sometimes students learn to answer such questions by putting the three vectors into a determinant and concluding that, because the determinant is zero, the vectors must be
linearly dependent. That technique is valid in certain cases but it’s best to go back to the definition and use the techniques of solving systems of linear equations. Apart from the fact that the determinant method only works for \( n \) vectors in \( \mathbb{F}^n \), it’s seen by many students to be a piece of hocus-pocus – a formula to be invoked, without any understanding. The following technique reveals what’s happening, and need involve no more work.

Suppose \( a(1, 5, -2) + b(6, -1, 9) + c(9, -17, 24) = (0, 0, 0) \). This is equivalent to the system of equations:

\[
\begin{align*}
a + 6b + 9c &= 0 \\
5a - b - 17c &= 0 \\
-2a + 9b + 24c &= 0
\end{align*}
\]

which can be represented by an augmented matrix:

\[
\begin{pmatrix}
1 & 6 & 9 & | & 0 \\
5 & -1 & -17 & | & 0 \\
-2 & 9 & 24 & | & 0
\end{pmatrix}
\]

As is usual with systems of homogeneous equations we can omit the column of 0’s and simply write the system as:

\[
\begin{pmatrix}
1 & 6 & 9 \\
5 & -1 & -17 \\
-2 & 9 & 24
\end{pmatrix}
\]

To solve this system we carry out a sequence of elementary row operations to put the matrix into echelon form. This simpler matrix represents a system of homogeneous linear equations that’s equivalent to the original system (equivalent in the sense of having precisely the same solutions).

In this example we obtain:

\[
\begin{pmatrix}
1 & 6 & 9 \\
5 & -1 & -17 \\
-2 & 9 & 24
\end{pmatrix} \xrightarrow{R_2 - 5R_1, R_3 + 2R_1} \begin{pmatrix}
1 & 6 & 9 \\
0 & -31 & -62 \\
0 & 21 & 42
\end{pmatrix} \xrightarrow{R_2 + 31R_3, R_3 \leftrightarrow R_3} \begin{pmatrix}
1 & 6 & 9 \\
0 & 1 & 2 \\
0 & 0 & 0
\end{pmatrix}
\]

Clearly this system has infinitely many solutions, and hence non-trivial ones. This is not so much because of the row of zeros at the bottom as the fact that we have an equivalent system with fewer equations than variables. In solving the above equivalent system:

\[
\begin{align*}
a + 6b + 9c &= 0 \\
b + 2c &= 0
\end{align*}
\]

we may choose \( c \) arbitrarily (and hence non-zero) and use the two equations to calculate the corresponding values of \( a \) and \( b \).

The conclusion is that the vectors are linearly dependent. It’s not necessary to exhibit an explicit relationship, but if called upon to do this we simply choose a convenient non-zero value of \( c \) and calculate the corresponding values of \( a \) and \( b \).

For example if we take \( c = 1 \) we get \( b = -2 \) from the second equation and hence \( a = 12 - 9 = 3 \) from the first equation. This gives the non-trivial linear relationship:

\[
3(1, 5, -2) - 2(6, -1, 9) + (9, -17, 24) = (0, 0, 0),
\]

which can be written as:

\[
(9, -17, 24) = -3(1, 5, -2) + 2(6, -1, 9).
\]
Example 3: Are the vectors \((2, 5, 9), (3, 1, 0), (5, -2, 7)\) linearly independent?

Solution: Writing them as the columns of a matrix (to get the coefficient matrix of the corresponding system of homogeneous linear equations) we have:

\[
\begin{pmatrix}
2 & 3 & 5 \\
5 & 1 & -2 \\
9 & 0 & 7
\end{pmatrix}
\begin{pmatrix}
1 & 3 & 5 \\
2 & 2 \\
5 & 1 & -2 \\
9 & 0 & 7
\end{pmatrix}
= \begin{pmatrix}
1 & 3 & 5 \\
0 & -13/2 & -29/2 \\
0 & -27/2 & 31/2
\end{pmatrix}
\begin{pmatrix}
1 & 3 & 5 \\
2 & 2 \\
0 & -13/2 \\
0 & -27/2 & 31/2
\end{pmatrix}
= \begin{pmatrix}
1 & 3 & 5 \\
0 & -13/2 & -29/2 \\
0 & -27/2 & 31/2
\end{pmatrix}
\begin{pmatrix}
1 & 3 & 5 \\
0 & 29/13 & 29/13 \\
0 & -27/2 & 31/2
\end{pmatrix}
= \begin{pmatrix}
1 & 3 & 5 \\
0 & 1 & 29/13 \\
0 & 0 & 31/2 + 27.29/13
\end{pmatrix}.
\]

R_1 + 2, R_2 - 5R_1, R_3 - 9R_1, R_2 + (-13/2), R_3 + (27/2)R_2

Since \(\frac{31}{2} + \frac{27.29}{13} \neq 0\) this system has a unique solution (the zero one).

Note that if we’re doing the calculations by hand we can avoid most fractions if we get leading 1’s by subtracting suitable multiples of rows from others rather than by division. This greatly simplifies the arithmetic. Re-doing the above example we can obtain:

\[
\begin{pmatrix}
2 & 3 & 5 \\
5 & 1 & -2 \\
9 & 0 & 7
\end{pmatrix}
\begin{pmatrix}
1 & 3 & 5 \\
2 & 2 \\
5 & 1 & -2 \\
9 & 0 & 7
\end{pmatrix}
= \begin{pmatrix}
1 & -5 & -12 \\
2 & 3 & 5 \\
9 & 0 & 7
\end{pmatrix}
\begin{pmatrix}
1 & -5 & -12 \\
0 & 13 & 29 \\
0 & 45 & 115
\end{pmatrix}
= \begin{pmatrix}
1 & -5 & -12 \\
0 & 13 & 29 \\
0 & 6 & 28
\end{pmatrix}
= \begin{pmatrix}
1 & 5 & -12 \\
0 & 1 & -27 \\
0 & 0 & 190
\end{pmatrix}.
\]

R_3 - 6R_2

As before this system has only the trivial solution so the vectors are linearly independent.

Example 4: Is the vector \((3, 18, -4)\) a linear combination of the vectors \((1, 5, -3)\) and \((4, 22, 1)\)?

Solution: The question is whether there exist scalars \(x, y\) such that:

\[
(3, 18, -4) = x (1, 5, -3) + y (4, 22, 1).
\]

We can write this equation as a non-homogeneous system of linear equations:

\[
\begin{align*}
x + 4y &= 3 \\
5x + 22y &= 18 \\
-3x + y &= -4
\end{align*}
\]

Now we can write this system as an augmented matrix:

\[
\begin{pmatrix}
1 & 4 & 3 \\
5 & 22 & 18 \\
-3 & 1 & -4
\end{pmatrix}
\begin{pmatrix}
1 & 4 & 3 \\
0 & 2 & 3 \\
0 & 1 & -13
\end{pmatrix}
\begin{pmatrix}
1 & 4 & 3 \\
0 & 1 & -13 \\
0 & 2 & 3
\end{pmatrix}
= \begin{pmatrix}
1 & 4 & 3 \\
0 & 1 & -13 \\
0 & 0 & 29
\end{pmatrix}.
\]

R_2 - 5R_1, R_3 + 3R_1, R_3 - 6R_2, R_2 \leftrightarrow R_3, R_3 - 3R_2

This echelon form represents an equivalent system that’s clearly inconsistent. In other words there’s no solution for \(x, y\) and so \((3, 1, -4)\) is not a linear combination of the vectors \((1, 5, 9)\) and \((4, -2, 1)\).
A set of vectors spans a vector space if every vector in that space is a linear combination of them. A vector space is finite-dimensional if it has a finite spanning set. A set of vectors is a basis for a vector space if it’s linearly independent and also spans it.

If we have a finite spanning set for a vector space, we can remove suitable vectors one by one, as long as the set is linearly dependent, without affecting the spanning, until we reach a basis. That is, every spanning set contains a basis. In a similar way, every linearly independent subset of a vector space can be suitably extended to a basis. But any two bases of a finite-dimensional vector space contain the same number of vectors. This unique number of vectors in a basis is called the dimension of the vector space. The vector space $\mathbb{R}^3$ clearly has dimension 3 since we have the basis $\{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$.

If we have a vector space of dimension $n$, then any set of more than $n$ vectors must clearly be linearly dependent (otherwise we could extend them to a basis with more vectors than the dimension). And if we have a set of less than $n$ vectors they can’t possibly span $V$ (otherwise the dimension would be less than $n$).

**Example 5:** Are the vectors $(1, 4, 6), (2, 5, 1), (6, 1, 9), (4, 8, 3)$ linearly independent?

**Solution:** The answer is clearly NO since you can’t have a set of four linearly independent vectors inside a 3-dimensional vector space. There’s no need to do any calculation in this case other than to count the number of vectors!

A vector space $V$ is said to be the sum of subspaces $U$ and $W$ if every vector $v \in V$ can be expressed as $v = u + w$ where $u \in U$ and $w \in W$. It is said to be the direct sum if as well, $U \cap W = 0$. In the first case we write $V = U + W$ and in the case of direct sum we write $V = U \oplus W$. The special feature of a direct sum is that if $V = U \oplus W$ then every vector $v \in V$ can be expressed uniquely as $v = u + w$ where $u \in U$ and $w \in W$.

If we take a basis for each of the subspaces in a direct sum, and combine them, we’ll produce a basis for the whole space. This means that $\dim(U \oplus V) = \dim U + \dim V$.

A Euclidean space is a vector space over $\mathbb{R}$ on which there is defined an inner product, that is a function that assigns to every pair of vectors $u, v \in V$ a real number $\langle u, v \rangle$ such that the following axioms hold:

1. $\langle u, v \rangle = \langle v, u \rangle$ for all $u, v$;
2. $\langle u, v + w \rangle = \langle u, v \rangle + \langle u, w \rangle$, for all $u, v, w$;
3. $\langle u, u \rangle > 0$ when $u \neq 0$.

Where the space is a space of functions the inner product is often defined as a certain definite integral, but in the simple case of row vectors over $\mathbb{R}$ we usually take the dot product:

$$\langle x_1, \ldots, x_n \rangle \cdot \langle y_1, \ldots, y_n \rangle = x_1y_1 + \ldots + x_ny_n.$$  

Vectors in a Euclidean space are orthogonal if their dot product is zero. The orthogonal complement of a subspace $U$ of a vector space $V$ is defined to be the set of all vectors that are orthogonal to every vector in $U$, that is $U^\perp = \{ v \in V \mid \langle u, v \rangle = 0 \text{ for all } u \in U \}$. The orthogonal complement $U^\perp$ is also a subspace of $V$ and it’s easy to show that every vector in $V$ can be expressed uniquely as the sum of a vector in $U$ and a vector orthogonal to every vector in $U$. In symbols this can be expressed as $V = U \oplus U^\perp$. One can also easily show that for all subspaces of a Euclidean space $U^{\perp\perp} = U$ (the orthogonal complement of the orthogonal complement is the original subspace).
For the purpose of Projective Geometry we’ll be dealing with $\mathbb{R}^3$ with the usual dot product. This means that we’ll have to consider (apart from the zero subspace and the whole of $\mathbb{R}^3$) only subspaces of dimension 1 and 2. Moreover the orthogonal complement of a 1-dimensional subspace will be a 2-dimensional one and vice-versa.

Vectors in $\mathbb{R}^3$ have a simple geometric interpretation whereby $(x, y, z)$ denotes the point with coordinates $x, y, z$ in 3-dimensional Euclidean space. A 1-dimensional subspace of $\mathbb{R}^3$ will then represent a line through the origin and a 2-dimensional subspace will represent a plane through the origin.

Often we interpret $(x, y, z)$ as representing not just the point, but the directed line segment from the origin to that point. In this case the vector has both magnitude (length) and direction. With this interpretation, non-zero vectors are orthogonal if and only if they’re perpendicular.

The orthogonal complement of a 1-dimensional subspace of $\mathbb{R}^3$ (a line through the origin) is a 2-dimensional subspace (a plane through the origin), being the plane perpendicular to that line. The orthogonal complement of a plane through the origin is a line through the origin.

In this special case of $\mathbb{R}^3$ we have another product of vectors. The cross product $\mathbf{u} \times \mathbf{v}$ of two vectors $\mathbf{u}, \mathbf{v}$ in a 3-dimensional vector space is a vector that is orthogonal to both $\mathbf{u}$ and $\mathbf{v}$. It is defined by:

$$(x_1, x_2, x_3) \times (y_1, y_2, y_3) = (x_2y_3 - x_3y_2, -(x_1y_3 - x_3y_1), x_1y_2 - x_2y_1).$$

It’s more easily remembered as the $3 \times 3$ “determinant”:

$$
\begin{vmatrix}
i & j & k \\
x_1 & x_2 & x_3 \\
y_1 & y_2 & y_3
\end{vmatrix}
$$

The top row are the standard basis vectors $\mathbf{i} = (1, 0, 0), \mathbf{j} = (0, 1, 0), \mathbf{k} = (0, 0, 1)$ and not scalars like the second and third rows. Nevertheless, using the usual rules for expanding determinants, we obtain a valid expression for the cross-product.

**Example 6:** If $\mathbf{u} = (1, 5, -3)$ and $\mathbf{v} = (4, -1, 2)$ find $\mathbf{u} \times \mathbf{v}$.

**Solution:**

$$\begin{vmatrix}
i & j & k \\
1 & 5 & -3 \\
4 & -1 & 2
\end{vmatrix} = [5.2 - (-3)(-1)] \mathbf{i} - [1.2 - (-3).4] \mathbf{j} + [1.(1) - 5.4] \mathbf{k}$$

$$= 7\mathbf{i} - 14\mathbf{j} - 21\mathbf{k} = (7, -14, -21).$$

**Check:**

$$(7, -14, -21).(1, 5, -3) = 7 - 70 + 63 = 0.$$  
$$(7, -14, -21).(4, -1, 2) = 28 + 14 - 42 = 0.$$  

Since it’s possible to make an error in sign at some stage it’s important that you check your answer to a cross-product against at least one of the two vectors.

Another concept from linear algebra that we’ll find useful is that of a linear transformation. A function $f : U \to V$ is a **linear transformation** if

$$f(\lambda\mathbf{u} + \mu\mathbf{v}) = \lambda f(\mathbf{u}) + \mu f(\mathbf{v})$$

for all $\mathbf{u}, \mathbf{v}$ in $U$ and all scalars $\lambda$ and $\mu$. 

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The kernel of such a linear transformation is \( \ker f = \{ u \in U \mid f(u) = 0 \} \) and its image is \( \operatorname{im} f = \{ f(u) \mid u \in U \} \). The kernel is a subspace of \( U \) and its dimension is called the nullity of \( f \). The image is a subspace of \( V \) and its dimension is called the rank of \( f \). One can show that
\[
\text{rank}(f) + \text{nullity}(f) = \dim U.
\]

§1.7. The Algebraic Version of the Real Projective Plane

The description of the Real Projective Plane as lines and planes through the origin is better than the intuitive one, where we simply invented ideal points, in that it frees us from having to maintain the artificial distinction between the ordinary and the ideal. But it’s still somewhat difficult to work with because it involves 3-dimensional geometry. The effective way of dealing with this is to use the linear algebra of \( \mathbb{R}^3 \). This third development of the Real Projective Plane is essentially the same as the previous one, except that things are expressed algebraically instead of geometrically.

We work in the 3-dimensional real vector space \( \mathbb{R}^3 \). The vectors have the form \((x, y, z)\) where \(x, y\) and \(z\) are real numbers.

Definitions:
A **projective point** is a 1-dimensional subspace of \( \mathbb{R}^3 \).
A **projective line** is a 2-dimensional subspace of \( \mathbb{R}^3 \).
A projective point \( P \) lies on a projective line \( h \) if the 1-dimensional subspace \( P \) is a subspace of the 2-dimensional subspace \( h \).

The **Real Projective Plane** is the set of all projective points and lines, together with the above incidence structure (relation of point lying on a line).

You should be able to see that this is simply an algebraic restatement of the previous description of the artist’s real projective plane.

Now a projective point \( P \), being a 1-dimensional subspace, has a basis consisting of one non-zero vector \( p \) so \( P = \langle p \rangle \). The elements of \( P \) are scalar multiples of \( p \).

If \( P = \langle p \rangle \) and \( Q = \langle q \rangle \) are projective points then \( P = Q \) doesn’t necessarily imply that \( p = q \). What it does mean is that each is a non-zero scalar multiple of the other, or to use the language of vector spaces, the set \( \{p, q\} \) is linearly dependent. On the other hand if \( P, Q \) are distinct points then \( p, q \) are linearly independent.

If \( P = \langle p \rangle \) and \( Q = \langle q \rangle \) are distinct then the smallest subspace containing them both is the 2-dimensional subspace \( \langle p, q \rangle \). This is the line \( P Q \).
**Theorem 1C:** The Real Projective Plane is complete. That is:
(i) any two projective points lie on exactly one projective line;
(ii) any two projective lines intersect in exactly one projective point.

**Proof:**
(i) Let $P = \langle p \rangle$ and $Q = \langle q \rangle$ be two distinct projective points. Then $P + Q = \langle p, q \rangle$ is clearly the only 2-dimensional subspace containing both $P$ and $Q$. It is thus the only projective line containing the projective points $P, Q$.

(ii) Let $h, k$ be two distinct projective lines. Now $h + k$ contains both $h$ and $k$. If $h + k = h$ then $k \subseteq h + k = h$. Since $h, k$ both have dimension 2 we must have $h = k$. [Recall that if two subspaces have the same dimension and one is inside the other, they must be equal.] This contradicts our assumption.

Hence $h$ is a proper subspace of $h + k$, so $\dim(h + k) > 2$. On the other hand the whole thing is going on in $\mathbb{R}^3$ which has dimension 3, so $\dim(h + k) = 3$. Thus we must have $\dim(h + k) = 3$.

Now recall the theorem from linear algebra that states that:
$$\dim(h + k) = \dim(h) + \dim(k) - \dim(h \cap k).$$
Therefore $3 = 2 + 2 - \dim(h \cap k)$, and so $\dim(h \cap k) = 1$. So $h \cap k$ is a projective point, and being their intersection, it lies on both $h$ and $k$.

**Theorem 2:** Three projective points $P = \langle p \rangle, Q = \langle q \rangle$ and $R = \langle r \rangle$ are collinear if and only if the vectors $p, q$ and $r$ are linearly dependent.

**Proof:** For $\langle p \rangle, \langle q \rangle, \langle r \rangle$ to be collinear they must lie on a common projective line. That means they must lie in a 2-dimensional subspace. For three vectors to lie in a 2-dimensional subspace they must be linearly dependent. Conversely if $p, q$ and $r$ are linearly dependent they’ll span a subspace of dimension at most 2. If $\langle p, q, r \rangle$ has dimension 2 then it’s a projective line passing through the projective points $P, Q, R$.

**EXERCISES FOR CHAPTER 1**

**TRUE/FALSE QUIZ**
1. An ordinary point cannot lie on the ideal line.
2. An ideal point cannot lie on an ordinary line.
3. Every ordinary line cuts the ideal line.
4. Two non-parallel ordinary lines cannot intersect in an ideal point.
5. There are infinitely many lines parallel to the ideal line.
6. There are infinitely many parallel classes.
7. There are infinitely many ideal points.
8. There are infinitely many ideal lines.
9. There is exactly one ideal point on every projective line.
10. There is exactly one ideal line through every projective point.

**Exercise 1:** Which of the following sets of vectors are linearly dependent?
With $\mathbb{R}$ as the field of scalars:
(i) $\{(3, 4, 7), (2, 8, 3), (10, 8, 25)\}$;
(ii) $\{(4, 6, 2), (6, 0, -1), (5, 1, 3), (1, 1, 0)\}$;
(iii) \{(6, 9, 11), (2, 1, 1)\};
(iv) \{(12, 15, 9), (20, 25, 15)\}.

With \( \mathbb{Z}_2 \) as the field of scalars:
(v) \{\{(1, 1, 1), (1, 1, 0), (0, 0, 1)\};
(vi) \{\{(1, 0, 0), (1, 1, 0), (0, 1, 1), (1, 1, 1)\}.

**Exercise 2:** Write \((10, 8, 25)\) as a linear combination of the vectors \((3, 4, 7)\) and \((2, 8, 3)\).

**Exercise 3:**
(a) What is the dimension of the subspace of \( \mathbb{R}^3 \) spanned by \((3, 4, 7), (2, 8, 3)\) and \((10, 8, 25)\)?
(b) What is the dimension of the space of all vectors which are orthogonal to all three of the vectors \((3, 7), (1, 1, 0), (2, 3, -7)\)?
(c) With \( \mathbb{Z}_5 \) as the field of scalars find the dimension of the subspace spanned by the vectors \((1, 3, 2), (4, 1, 0)\) and \((0, 1, 3)\).

**Exercise 4:** If \(a = (6, 3, 2)\) and \(b = (-2, 1, 5)\), find
(i) \(a.a\);
(ii) \(a.b\);
(iii) \(b.a\);
(iv) \(b.b\);
(v) \(a \times a\);
(vi) \(a \times b\);
(vii) \(b \times a\);
(viii) \(b \times b\).

**Exercise 5:**
(a) Prove that for all \(u, v \in \mathbb{R}^3\) \(u \times v = -(v \times u)\).
(b) Hence show that for all \(v \in \mathbb{R}^3\) \(v \times v = 0\).

**Exercise 6:** Find \((\langle 4, -7, 8 \rangle)^\perp \cap \langle 3, 1, -2 \rangle)^\perp\).

**Exercise 7:** Let \(a = (1, 2, 3)\). Find the rank and nullity of the following (where they exist):
(a) \(f: \mathbb{R}^3 \to \mathbb{R}\) defined by \(f(v) = v.a\).
(b) \(g: \mathbb{R}^3 \to \mathbb{R}^3\) defined by \(g(v) = v \times a\).
(c) \(h: \mathbb{R}^3 \to \mathbb{R}\) defined by \(h(v) = v.v\).

**Exercise 8:** If \(A = \langle (1, 2, 3) \rangle\) and \(B = \langle (5, 1, 5) \rangle\) find \(AB\).

**Exercise 9:** If \(P = \langle (1, 1, 1) \rangle\) and \(Q = \langle (1, -2, 3) \rangle\) does \(R = \langle (3, 1, 4) \rangle\) lie on \(PQ\)?

**Exercise 10:** If \(A = \langle a \rangle\), \(B = \langle b \rangle\) and \(C = \langle a + b \rangle\), show that \(C\) lies on \(AB\).

**Exercise 11:** Let \(P = \langle (1, 1, 1) \rangle, Q = \langle (1, 2, 3) \rangle, R = \langle (2, 3, 1) \rangle\) and \(S = \langle (3, 1, 2) \rangle\). Find \(PQ \cap RS\)

**Exercise 12:** Let \(A = \langle (4, -10, 5) \rangle, B = \langle (0, 7, -3) \rangle, C = \langle (1, 1, 0) \rangle\) and \(D = \langle (1, 2, 2) \rangle\).
Let $M = AB \cap CD$ and let $N = AC \cap BD$. Find $MN$ and determine whether $AD \cap BC$ lies on $MN$.

**Exercise 13:** Is $\langle (1, 0, 1) \rangle$ an ideal point?

**Exercise 14:** If we embed the Euclidean plane in the projective plane by taking the plane $\pi = \{(x, y, z) \mid x + 2y + 3z = 4\}$, determine whether $\langle (1, -2, 1) \rangle$ is an ideal point.

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**SOLUTIONS FOR CHAPTER 1**

**TRUE/FALSE QUIZ:**
(1) T (2) F (3) T (4) F (5) F (6) T (7) T (8) F (9) F (don’t forget the ideal line) (10) F.

**Exercise 1:** (i) Suppose $x (3, 4, 7) + y (2, 8, 3) + z (10, 8, 25) = (0, 0, 0)$. Then $x, y, z$ satisfy the system of homogeneous linear equations with coefficient matrix:

$$
\begin{pmatrix}
3 & 2 & 10 \\
4 & 8 & 8 \\
7 & 3 & 25
\end{pmatrix}
\rightarrow
\begin{pmatrix}
3 & 2 & 10 \\
1 & 6 & -2 \\
7 & 3 & 25
\end{pmatrix}
\rightarrow
\begin{pmatrix}
1 & 6 & -2 \\
3 & 2 & 10 \\
0 & -39 & 39
\end{pmatrix}
\rightarrow
\begin{pmatrix}
1 & 6 & -2 \\
0 & 1 & -1 \\
0 & 0 & 0
\end{pmatrix}.
$$

R2 - R1 \hspace{5cm} R1 \leftrightarrow R2 \hspace{5cm} R2 - 3R1, R3 - 7R1 \hspace{5cm} R2 \leftrightarrow (-16), R3 + 39R2

This system has the non-trivial solution $z = 1$, $y = 1$, $x = -4$. Hence $4 (3, 4, 7) = (2, 8, 3) + (10, 8, 25)$ and so the set is linearly dependent.

(ii) The set is linearly dependent since there are 4 vectors in a 3-dimensional vector space.

(iii) A set of two vectors is linearly dependent if and only if one is a scalar multiple of the other. This is not the case here so the set is linearly independent.

(iv) $(12, 15, 9) = (3/5)(20, 25, 15)$ so the set is linearly dependent.

(v) The sum of the vectors is 0 so the set is linearly dependent.

(vi) The set is linearly dependent for the same reason as in (ii)

**Exercise 2:**
Let $x (3, 4, 7) + y (2, 8, 3) = (10, 8, 25)$.

Then $x, y$ satisfy the system of linear equations:

$$
\begin{pmatrix}
3 & 2 & 10 \\
4 & 8 & 8 \\
7 & 3 & 25
\end{pmatrix}
\rightarrow
\begin{pmatrix}
1 & 6 & -2 \\
3 & 2 & 10 \\
7 & 3 & 25
\end{pmatrix}
\rightarrow
\begin{pmatrix}
1 & 6 & -2 \\
0 & 1 & -1 \\
0 & 0 & 0
\end{pmatrix}
$$

we get the echelon form $1 6 | -2 0 1 | -1 0 0 | 0$ which gives $y = -1, x = 4$.

Hence $(10, 8, 25) = 4 (3, 4, 7) - (2, 8, 3)$.

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Exercise 3:
(a) Since the vectors are linearly dependent (done in Question 1 (i)) the dimension < 3. As
they are not all scalar multiples of the same vector the dimension > 1.
Hence the dimension = 2.

(b) Suppose \( x (3, 2, 7) + y (1, 1, 0) + z (2, 3, -7) = (0, 0, 0) \).
Then \( x, y, z \) satisfy the homogeneous system:
\[
\begin{pmatrix}
3 & 1 & 2 \\
2 & 1 & 3 \\
7 & 0 & -7
\end{pmatrix}
\begin{pmatrix}
x \\
y \\
z
\end{pmatrix}
= \begin{pmatrix}
1 & 0 & -1 \\
2 & 1 & 3 \\
7 & 0 & -7
\end{pmatrix}
\begin{pmatrix}
x \\
y \\
z
\end{pmatrix}.
\]
Since \( R3 = 7R1 \) this system has a non-zero solution and so the set of
vectors is linearly dependent. So they span a 2-dimensional subspace and their orthogonal
complement has dimension \( 3 - 2 = 1 \).

(c) Suppose \( x (1, 3, 2) + y (4, 1, 0) + z (0, 1, 3) = (0, 0, 0) \).
Then \( x, y, z \) satisfy the homogeneous system:
\[
\begin{pmatrix}
1 & 4 & 0 \\
3 & 1 & 1 \\
2 & 0 & 3
\end{pmatrix}
\begin{pmatrix}
x \\
y \\
z
\end{pmatrix}
= \begin{pmatrix}
1 & 4 & 0 \\
0 & 4 & 1 \\
0 & 2 & 3
\end{pmatrix}
\begin{pmatrix}
x \\
y \\
z
\end{pmatrix}.
\]
This system has only the zero solution so the set of vectors is linearly independent. So they
span a 3-dimensional subspace.

Exercise 4:
\( a.a = 36 + 9 + 4 = 49; \ a.b = -12 + 3 + 10 = 1; \ b.a = 1 \) (always \( b.a = a.b \));
\( b.b = 4 + 1 + 25 = 30; \ a \times a = 0 \) (true for all \( a \)); \( a \times b = \begin{vmatrix}
i & j & k \\
6 & 3 & 2 \\
-2 & 1 & 5
\end{vmatrix} = (13, -34, 12) \)
(check: \( 78 - 102 + 24 = 0 \)); \( b \times a = (-13, 34, -12) \) (always \( b \times a = -a \times b \)); \( b \times b = 0 \).

Exercise 5:
(a) The computations of \( u \times v \) and \( v \times u \) as \( 3 \times 3 \) determinants are identical, except for the
swapping of the 2nd and 3rd rows. This changes the sign of a determinant.
(b) By (a) \( v \times v = -v \times v \), so \( v \times v = 0 \).

Exercise 6:
\((x, y, z) \in (4, -7, 8) \perp \cap (3, 1, -2) \perp \) if and only if it is orthogonal to both vectors. Such
vectors are scalar multiples of the cross product.
\((4, -7, 8) \times (3, 1, -2) = \begin{vmatrix}
i & j & k \\
4 & -7 & 8 \\
3 & 1 & -2
\end{vmatrix} = (6, 32, 25) \)
(Check: \( 4.6 - 7.32 + 8.25 = 24 - 224 + 200 = 0 \))
So the answer is \((6, 32, 25)\).

Exercise 7:
(a) \( f \) is a linear transformation. \( \ker f \) is the orthogonal complement of \((1, 2, 3)\). Hence \( \dim \ker f = 2 \). Thus the \textbf{nullity} = 2 and the \textbf{rank is thus} 1 (ie \( 3 - 2 \)).
(b) \( g \) is a linear transformation. \( \ker g = \{ v \mid v \times a = 0 \} = \langle a \rangle \).
So the **nullity is 1** and the **rank is 2**.

(c) \( h \) is not a linear transformation and so rank and nullity are not defined.
[eg \( h(2v) = 2v.2v = 4(v.v) = 4h(v) \), not \( 2h(v) \) ]

**Exercise 8:** \( AB = \langle (1, 2, 3) \times (5, 1, 5) \rangle \perp = \langle (7, 10, -9) \rangle \perp \).

**Exercise 9:** \( PQ = \langle (5, -2, -3) \rangle \perp \). Since \( (3, 1, 4).\langle (5, -2, -3) \rangle = 15 -2 - 12 = 1 \neq 0 \), \( R \) does not lie on \( PQ \). (Alternatively we can put the vectors for \( P \), \( Q \) and \( R \) as the rows of a \( 3 \times 3 \) matrix and show that the determinant is non-zero).

**Exercise 10:** Since \( a \), \( b \), \( c \) are clearly linearly dependent.

**Exercise 11:**
(i) \( PQ = \langle (1, 1, 1) \times (1, 2, 3) \rangle \perp = \langle (1, -2, 1) \rangle \perp \),
\( RS = \langle (2, 3, 1) \times (3, 1, 2) \rangle \perp = \langle (5, -1, 7) \rangle \perp \),
\( PQ \cap RS = \langle (1, -2, 1) \times (5, -1, 7) \rangle \perp = \langle (15, 12, 9) \rangle \perp = \langle (5, 4, 3) \rangle \perp \).
(ii) \( PQ \) is the line \( y = 1 \), \( RS \) is the line \( x + y = 2 \), \( PQ \cap RS = (1, 1) \).
(iii) \( PQ = R \perp \), \( RS = N \perp \), \( PQ \cap RS = A \).

**Exercise 12:** \( AB = \langle (-5, 12, 28) \rangle \perp \) and \( CD = \langle (2, -2, 1) \rangle \perp \). If \( M = \langle m \rangle \) then \( m \) is orthogonal to both \( (-5, 12, 28) \) and \( (2, -2, 1) \) and so is parallel to \( (-5, 12, 28) \times (2, -2, 1) = (68, 61, -14) \). Hence \( M = \langle (68, 61, -14) \rangle \) and \( BD = \langle (20, -3, -7) \rangle \perp \) and \( N = \langle (7, 245, -85) \rangle \). Then \( MN = \langle (-1755, 5682, 16233) \rangle \). Now \( AD = \langle (-30, -3, 18) \rangle \perp \) and \( BC = \langle (3, -3, -7) \rangle \perp \) so \( AD \cap BC = \langle (-4155, -2424, -7329) \rangle \). It would lie on \( MN \) if the vectors \( (-1755, 5682, 16233) \) and \( (-4155, -2424, -7329) \) were orthogonal. Since their dot product is \(-125452800 \neq 0\) they are not orthogonal and so \( AD \cap BC \) does not lie on \( MN \).

**Exercise 13:** Sorry to trick you! Of course there’s no distinction between ideal points and ordinary points in the projective plane itself. The concepts only make sense when we embed a Euclidean plane in the projective plane. Depending on which embedding we take this point could be ordinary or ideal.

**Exercise 14:** Since \( (1, -2, 1) \) lies on the plane \( x + 2y + 3z = 0 \) the line joining the origin to \( (1, -2, 1) \) is parallel to \( \pi \) and so does not cut \( \pi \). Hence \( \langle (1, -2, 1) \rangle \) is an ideal point.