

## Section 5.3 Projection

2) We use that  $A^T A = I_n$  and  $\vec{v} \cdot A\vec{w} = A^T \vec{v} \cdot \vec{w}$  so conclude

$$A\vec{v} \cdot A\vec{w} = A^T A\vec{v} \cdot \vec{w} = \vec{v} \cdot \vec{w}.$$

This is an important property of orthogonal transformations. It implies that both angle and length are preserved if  $A^T A = I_n$ .

6) Yes, we have to show that if  $A^T A = I_n$ , then also  $AA^T = B^T B = I_n$ , where  $B = A^T$ .

From  $A^T A = I_n$ , we get  $A^T = A^{-1}$  which can be read as  $B = (B^T)^{-1}$ . Taking inverses shows  $B^{-1} = B^T$ . By multiplying both sides from the right with  $B$ , we end up with  $B^T B = I_n$ .

(It is useful to remember from this that also the rows of an orthogonal matrix form an orthonormal basis.)

8) a) No, take just the example on the same page, where  $A^T A$  is  $I_2$  while  $AA^T$  is a projection matrix.

b) Yes, it is the case because we have seen in problem 6) that  $B = A^T$  is also  $AA^T = I_n$ .

16) Yes:  $A = A^T$  implies  $(A^2)^T = (AA)^T = A^T A^T = (A^T)^2$ .

Note that if  $A, B$  are different symmetric matrices, then  $AB$  is not necessarily symmetric any more.

20) The two vectors are not yet orthonormal, but a Gram-Schmidt orthonormalisation gives

to normal vectors which can be used to define  $A = \begin{bmatrix} 1/2 & -1/10 \\ 1/2 & 7/10 \\ 1/2 & -7/10 \\ 1/2 & 1/10 \end{bmatrix}$  and get the projection

$$P = AA^T = \frac{1}{50} \begin{bmatrix} 13 & 9 & 16 & 12 \\ 9 & 37 & -12 & 16 \\ 16 & -12 & 37 & 9 \\ 12 & 16 & 9 & 13 \end{bmatrix}.$$

$$18) a) A = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, A^2 = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

b) Actually,  $A^2$  is symmetric, because  $(A^2)^T = (AA)^T = (-A^T)(-A^T) = (A^T)^2$ .

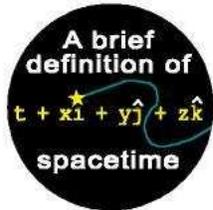
44) We can write any quaternion matrix as  $M = p1 + qi + sj + rk$ , where

$$1 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, i = \begin{bmatrix} 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{bmatrix}, j = \begin{bmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}, k = \begin{bmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}.$$

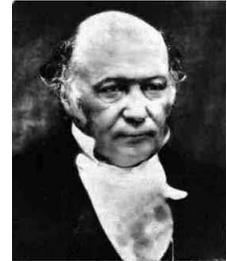
From this it is obvious that  $M$  is a linear space of dimension 4. Because  $i \cdot j = -j \cdot i = k$ ,  $j \cdot k = -k \cdot j = i$ ,  $k \cdot i = -i \cdot k = j$ ,  $i \cdot i = j \cdot j = k \cdot k = -1$ , the product of quaternion matrices is a quaternion matrix. We compute  $M^T M = (p^2 + q^2 + r^2 + s^2)I_4$  and see from this that  $M$  is invertible if  $(p^2 + q^2 + r^2 + s^2)$  is not zero. In the other case, we have the zero matrix. The inverse is in

$M$  too because  $M^{-1} = M^T / (p^2 + q^2 + r^2 + s^2)$  and  $M^T = -M + pI_4$  is a quaternion matrix too. Quaternions do not commute because  $ij = -ji$ .

**ROTATION WITH QUATERNIONS.** One can calculate with quaternions directly without the matrix representation. One often writes  $q = (s, \vec{p}) = s + ip_1 + jp_2 + kp_3$ . Quaternions are useful to compute rotations in space. If one wants to rotate a vector  $\vec{v}$  in space by an angle  $\phi$  around an axis which contains the unit vector  $\vec{u}$ , one can form the quaternions  $q = (\phi, \vec{u})$  and  $p = (0, \vec{v})$  and form the new quaternion  $p' = qpq^{-1}$ . It has the form  $p' = (0, \vec{v}')$ , where  $\vec{v}'$  is the rotated vector. This algebraic manipulation is useful in physics or computer graphics.



**DISCOVERY.** Quaternions were discovered by William Rowan Hamilton while walking along the Royal Canal. He was so excited about his "invention" that he wrote the properties  $i^2 = j^2 = k^2 = ijk = -1$  of the quaternions into the stone of the Brougham bridge. Quaternions are also called **hypercomplex numbers**. They are not only used in computer graphics, also physicists find them handy.



### Section 5.4 Orthogonality and least squares

2) The kernel of  $A^T$  is equal to the orthogonal complement of the image of  $A$ . Since the later is a plane, the kernel of  $A^T$  is a line. To find the kernel of  $B = A^T = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix}$ , we do row reduction and get  $\text{rref}(B) = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \end{bmatrix}$  from which we see that  $[1, -2, 1]^T$  is a basis for the kernel.

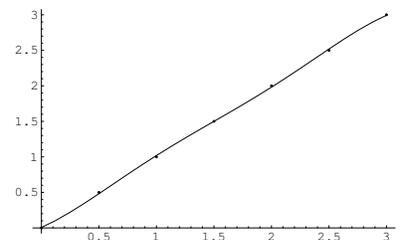
10) a) Consistent means that we have a solution  $A\vec{v} = \vec{b}$ . The hint leads the path: split up this solution  $\vec{v} = \vec{v}_h + \vec{v}_0$ , where  $\vec{v}_h$  is in the kernel of  $A$  and  $\vec{v}_0$  in the complement. Now,  $A\vec{v}_h = 0$  implies  $A\vec{v}_0 = \vec{b}$  also and we found the solution.

b) If there were two solutions  $\vec{x}_0, \vec{y}_0$  in the orthogonal complement of the kernel, then the element  $\vec{v} = \vec{x}_0 - \vec{y}_0$  in the orthogonal complement of the kernel satisfies  $A\vec{v} = \vec{0}$ . But because  $\vec{v}$  is both in the kernel and the orthogonal complement, it is  $\vec{0}$  and  $\vec{x}_0 = \vec{y}_0$ .

c) This follows from Pythagoras and the relation  $\vec{x}_1 = \vec{x}_0 + \vec{x}_h$ .

22) Use the routine formula  $(A^T A)^{-1} A^T \vec{b} = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$ .

34)  $f(t) = 1.5 - 1.53669 \cos(t) + 0.0431477 \cos(2t) + 0.108974 \sin(t) + 0.302692 \sin(2t)$  It was of course ok to use technology here.



40) We fit the data points  $(x_i, \log(y_i))$  with respect to the functions  $1, t$  and get the best linear fit  $f(s) = 0.171666 + 0.155262s$ . This means that  $g(t) = 1.18728e^{0.155262t} = 1.187281.167976^t$  is the best fit.

16) Assume  $A$  is a  $m \times n$  matrix so that  $A^T$  is a  $n \times m$  matrix. Apply dimensions to  $\text{im}(A)^\perp = \ker(A^T)$  gives  $m - \text{rank}(A) = m - \text{rank}(A^T)$ , where we have used that  $\dim \ker(A^T) + \dim \text{ran}(A^T) = m$ .

18) The rank of  $A^T A$  is smaller or equal than the rank of  $A$ . Because the kernel of  $A^T$  is perpendicular to the image of  $A$ , the rank of  $A^T A$  is indeed equal to the rank of  $A$ . and because the ranks of  $A$  and  $A^T$  agree also the rank of  $A^T A$  is equal to the rank of  $AA^T$ .

### Section 6.1 Determinants

14) This is a permutation matrix. There are 6 inversions so that the determinant is 1.

4) Do Laplace expansion with respect to the first row, then with respect to the new 3'd row.

The determinant is  $39 \det \begin{bmatrix} 1 & 1 & 1 \\ 0 & 2 & 2 \\ 7 & 4 & 1 \end{bmatrix} = -324$ .

6) After row reduction, end up with a matrix  $M_{n-1}$ . Because  $M_1 = 1$ , we have  $M_n = 1$  for all  $n$ .

12) Use the linearity to see that we have the same determinant as  $A$  which is 8.

16) We have  $t^2(b - a)$ . Because for  $a = t$  and  $b = t$ , the matrix is not invertible, we see that  $f(t) = c(t - a)(t - b)$  and from a) we get  $f(t) = (a - b)(t - a)(t - b)$ . The matrix is invertible if  $a \neq b$  and  $t$  is different from both  $a$  and  $b$ .

40) The function  $f(t)$  is a polynomial of degree 4. Now  $f(0) = (\pm 1)^2 - 1 = 0$  and similarly  $f(1) = f(2) = f(3) = f(4) = 0$  and consequently by the fundamental theorem of algebra,  $f(t) = 0$  for all  $t$ . Therefore,  $\det(A) = \pm 1$  which implies that the inverse of  $A$  has integer entries.