

# Division Algebras and Wireless Communication

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**T**he aim of this note is to bring to the attention of a wide mathematical audience the recent application of division algebras to wireless communication. The application occurs in the context of communication involving multiple transmit and receive antennas, a context known in engineering as MIMO—short for multiple input, multiple output. While the use of multiple receive antennas goes back to the time of Marconi, the basic theoretical framework for communication using multiple transmit antennas was only published about ten years ago. The progress in the field has been quite rapid, however, and MIMO communication is widely credited with being one of the key emerging areas in telecommunication. Our focus here will be on one aspect of this subject: the formatting of transmit information for optimum reliability.

Recall that a division algebra is an (associative) algebra with a multiplicative identity in which every nonzero element is invertible. The center of a division algebra is the set of elements in the algebra that commute with every other element in the algebra; the center is itself just a commutative field, and the division algebra is naturally a vector space over its center. We consider only division

algebras that are finite-dimensional as such vector spaces. Commutative fields are trivial examples of these division algebras, but they are by no means the only ones: for instance, class-field theory tells us that over any algebraic number field  $K$ , there is a rich supply of noncommutative division algebras whose center is  $K$  and are finite-dimensional over  $K$ .

Interest in MIMO communication began with the papers [21, 10, 24, 11], in which it was established that MIMO wireless transmission could be used both to decrease the probability of error and to increase the amount of information that can be transmitted. This result caught the attention of telecommunication operators, particularly since MIMO communication does not require additional resources in the form of either a larger slice of the radio spectrum or increased transmitted power.

The basic setup is as follows: Complex numbers  $Re^{i\phi}$ , encoded as the amplitude ( $R$ ) and phase ( $\phi$ ) of a radio wave, are sent from  $t$  transmit antennas (one number from each antenna), and the encoded signals are then received by  $r$  receive antennas. The presence of obstacles in the environment, such as buildings, causes attenuation of the signals; in addition, the signals are reflected several times and interfere with one another. The combined degradation of the signals is commonly referred to as fading, and achieving reliable communication in the presence of fading has been the most challenging aspect of wireless communication. The received and transmitted signals are modeled by the relation

$$Y_{r \times 1} = \theta H_{r \times t} X_{t \times 1} + W_{r \times 1}$$

where  $X$  is a  $t \times 1$  vector of information signals,  $Y$  is an  $r \times 1$  vector of received signals,  $W$  is an  $r \times 1$  vector of additive noise,  $H$  is an  $r \times t$  matrix

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that models the fading, and  $\theta$  is a real number chosen to multiply the information signals so as to fit the power available for transmission. Under the most commonly adopted model, the entries of the noise vector  $W$  and the channel matrix  $H$  are assumed to be Gaussian complex random variables that are independent and identically distributed with zero mean. (A Gaussian complex random variable is one of the form  $w = x + iy$ , where  $x$  and  $y$  are real Gaussian random variables that are independent and have the same mean and variance. The modulus of such a random variable, and in particular the magnitude of each fading coefficient  $h_{ij}$ , is then Rayleigh distributed. This model is hence also known as the Rayleigh fading channel model.) It is the presence of fading in the channel that distinguishes this model from more classical channels, where the primary source of disturbance is the additive Gaussian noise  $W$ .

A common engineering model is to assume that the channel characteristics (i.e., the fading coefficients  $h_{ij}$ ) stay constant in some fixed but small time interval and that these characteristics are known to the receiver but not the transmitter. (This is known as *coherent* transmission.) If each antenna can transmit  $n$  times during such an interval, then the transmission process is compartmentalized into blocks of length  $n$ : each antenna transmits  $n$  times, and each receiver waits to receive all  $n$  transmissions before processing them. A common simplifying assumption is to take  $r = t = n$ , and the equation above is accordingly modified to read

$$(1) \quad Y_{n \times n} = \theta H_{n \times n} X_{n \times n} + W_{n \times n}.$$

Thus the  $i$ th column of  $Y$ ,  $\theta X$ , and  $W$  represent (respectively) the received vectors, the transmitted information, and the additive noise from the  $i$ th transmission. A measure of the power available during a single transmission from all  $n$  antennas, i.e., a single use of the telecommunication channel, is the *signal-to-noise ratio* (SNR)  $\rho$ . Recall that the Frobenius norm  $\|X\|_F$  of  $X = (x_{i,j})$  equals  $\sqrt{\sum_{i,j} |x_{i,j}|^2}$ . Since the power required to send a complex number varies as the square of its modulus, the normalization constant  $\theta$  must satisfy  $\theta^2 \|X\|_F^2 \leq n\rho$ .

A subset  $S$  of the nonzero complex numbers known as the *signal set* is selected as the alphabet (a common situation is that  $S$  is a finite subset of size  $q$  of the nonzero Gaussian integers  $\mathbb{Z}[i] - \{0\}$ ), and a  $k$ -tuple  $(s_1, s_2, \dots, s_k)$ ,  $s_i \in S$ , constitutes the message that the transmitter wishes to convey to the receiver. Thus there are  $q^k$  messages in all, and it is assumed that each message is equally likely to be transmitted. A *space-time code* is then a one-to-one map  $X: S^k \rightarrow M_n(\mathbb{C})$ ; we write  $\mathcal{X}$  for  $X(S^k)$ . The transmitted matrix  $\theta X_{n \times n}$  in Equation (1) is thus drawn from the set  $\theta \mathcal{X}$  as  $(s_1, s_2, \dots, s_k)$  vary

in  $S^k$ . Often  $\mathcal{X}$  itself is referred to as the space-time code. It is typically assumed that the map  $X$  is “linear in  $S^k$ ”, that is, it is the restriction to  $S^k$  of a group homomorphism  $\langle S \rangle^k \rightarrow M_n(\mathbb{C})$ , where  $\langle S \rangle$  is the additive subgroup of  $\mathbb{C}$  generated by  $S$ . (The term “space-time” refers to the fact that information  $(s_1, s_2, \dots, s_k)$  is packaged in the spatial direction by sending it out through several physically separated transmit antennas and in the time direction by sending it out in  $n$  consecutive transmissions.)

Under the information-theoretic framework developed by Shannon in 1948 ([18]) and adopted ever since within the telecommunication community, the amount of information conveyed by a message in this setting is equal to  $\log_2(q^k)$  “bits”. Since this amount of information is conveyed in  $n$  transmissions over the MIMO channel, the rate of information transmission is then given by  $\frac{k}{n} \log_2(q)$  bits per channel use. When  $q$  and  $n$  are fixed a priori, the quantity  $k$  serves as a measure of information rate.

Reliability of communication is commonly measured by the probability  $P_e$  of incorrectly decoding the transmitted message at the receiver. The pairwise error probability  $P_e(i, j)$  (for  $i \neq j$ ) is the probability that message  $i$  is transmitted and message  $j$  is decoded. Performance analysis of MIMO communication systems typically focuses on the pairwise error probability, as it is easier to estimate and also because the error probability  $P_e$  can be upper and lower bounded in terms of the pairwise error probability.

It was shown in [21, 11] that for a fixed SNR (i.e., power)  $\rho$ , in order to keep the pairwise error probability low, the space-time code  $\mathcal{X}$  must meet the two criteria below, of which the first is primary:

- (1) *Rank Criterion*: For  $\underline{s} := (s_1, s_2, \dots, s_k)$  and  $\underline{s}' := (s'_1, s'_2, \dots, s'_k)$  with  $\underline{s} \neq \underline{s}'$ , the difference matrix

$$X(\underline{s}) - X(\underline{s}')$$

must have full rank  $n$ , i.e., it must be invertible.

- (2) *Coding Gain Criterion*: For  $\underline{s}$  and  $\underline{s}'$  as above,  $\underline{s} \neq \underline{s}'$ , the modulus of the determinant of difference

$$|\det(\theta X(\underline{s}) - \theta X(\underline{s}'))|$$

must be as large as possible.

Clearly, the second criterion comes into play only when the first criterion has been met but then subsumes it. Each criterion impacts a different communication parameter, and the two are hence stated independently. Note that one cannot arbitrarily scale the matrices  $X$  to increase the coding gain because the assumption of fixed  $\rho$ , along with the relation  $\theta^2 \|X\|_F^2 \leq n\rho$ , would simply cause a corresponding decrease in  $\theta$ . Note too that one cannot increase the quantity  $k$  (a proxy for the rate

of information) arbitrarily, as this would create a larger set of matrices  $\theta X$  all circumscribed to lie within a sphere of radius  $\sqrt{n\rho}$ , which would then cause the determinant of their differences to get smaller, thereby going against the second criterion.

### Satisfying the Rank Criterion

The earliest space-time code, for two antennas, was given by an engineer, Alamouti ([1]): given an arbitrary signal set  $S$ , he chose  $X: S^2 \rightarrow M_2(\mathbb{C})$  to be

$$(2) \quad X(s_1, s_2) = \begin{pmatrix} s_1 & -\overline{s_2} \\ s_2 & \overline{s_1} \end{pmatrix}$$

(where  $\overline{s_i}$  stands for complex conjugation). It is easy to see that the rank criterion is immediately met. Writing  $s_1 = u_1 + iu_2, s_2 = u_3 + iu_4$ , each such matrix can be expressed in the form  $X(s_1, s_2) = \sum_{i=1}^4 u_i A_i$ . The  $2 \times 2$  complex matrices  $A_j$  are such that for any complex  $2 \times 2$  channel matrix  $H$ , the collection of  $2 \times 2$  matrices  $\{HA_i\}$  is pairwise orthogonal when regarded as vectors in  $\mathbb{R}^8$  by writing out sequentially the real and imaginary parts of each entry of the  $\{HA_i\}$ . The expansion above makes it possible to do a least squares estimation of the  $u_j$  from the received matrix  $Y$ , also considered as a vector in  $\mathbb{R}^8$  as above, by projecting onto the respective matrices  $HA_j$  (we will consider this in more detail later). It is this property that makes the Alamouti code so easy to decode, and, not surprisingly, the code has since been adopted into the IEEE 802.11n "Wireless LAN" standard. In applications, the  $\{s_1, s_2\}$  are typically drawn from a subset of  $\mathbb{Z}[i] \times \mathbb{Z}[i]$ .

Alamouti's code led to a furious search among engineers and coding theorists for generalizations for higher numbers of antennas. Much of the early work (see [22], for example) focused on combinatorial methods. The matrix  $X$  in Equation (2) is almost unitary: it satisfies  $XX^\dagger = (s_1\overline{s_1} + s_2\overline{s_2})I_2$ , where the superscript  $\dagger$  stands for transpose conjugate, and  $I_2$  stands for the  $2 \times 2$  identity matrix. Not surprisingly, early workers (see [22], for example) sought  $n \times n$  matrices  $X(s_1, \dots, s_k)$  whose entries come from the set  $\{\pm s_j, \pm \overline{s_j}, \pm i s_j, \pm i \overline{s_j}, j = 1, \dots, k\}$  and satisfy

$$(3) \quad XX^\dagger = (s_1\overline{s_1} + \dots + s_k\overline{s_k})I_n.$$

This quickly leads to a necessary condition: the existence of  $2k - 1$  complex  $n \times n$  matrices  $A_i$  satisfying  $A_i^\dagger A_i = I_n, A_i^\dagger = -A_i$ , and  $A_i A_j = -A_j A_i$  for  $1 \leq i < j \leq 2k - 1$ . These are, of course, the Hurwitz-Radon-Eckmann matrices, and classical results of Hurwitz-Radon-Eckmann (see [6], for instance) severely limits the values of  $k$  for which such matrices can exist. If  $n = 2^a(2b + 1)$ , then the Hurwitz-Radon-Eckmann result says that the maximum possible value of  $k$  equals  $(a + 1)$ . Thus  $k = n$  if and only if  $n = 2, k \leq \frac{3n}{4}$  for  $n > 2$ , and

$k \leq \frac{n}{2}$  for  $n > 4$ . It follows that these generalizations of the Alamouti code transmit too few information symbols for more than two transmit antennas. (A similar analysis of the matrices  $A_i$  using representations of Clifford algebras was made by Tirkkonen and Hottinen in [20].)

In 2001 Sundar Rajan, a professor of communication engineering at the Indian Institute of Science, introduced the problem of designing matrices  $X(s_1, \dots, s_k)$  satisfying the rank criterion to this author. Given his algebraic background, this author could recognize easily that matrices arising from embeddings of fields and division algebras can be utilized to solve this problem. Let  $f: D \rightarrow M_n(\mathbb{C})$  be an embedding, i.e., an (injective) ring homomorphism of a division algebra  $D$  into the  $n \times n$  matrices over  $\mathbb{C}$ . Then for  $X_1 = f(d_1)$  and  $X_2 = f(d_2)$  ( $X_1 \neq X_2$ ),  $X_1 - X_2$  must necessarily be invertible. This is because  $d_1 - d_2$ , being a nonzero element of the division algebra  $D$ , is automatically invertible, and since  $f$  is a homomorphism, the same must also be true of  $X_1 - X_2$ . Thus the matrices in  $f(D)$  automatically satisfy the rank criterion. Using this observation, Sundar Rajan, his Ph.D. student Shashidhar, and this author ([19]) proposed several schemes for constructing space-time codes from various signal sets. For each signal set  $S$  and for each  $n$ , they constructed suitable division algebras  $D$ , suitable embeddings  $f: D \rightarrow M_n(\mathbb{C})$ , and suitable injective maps  $X: S^k \rightarrow f(D)$ , for suitable  $k$ .

For simplicity of construction in the noncommutative case, the authors of [19] used *cyclic division algebras* for their codes. A cyclic division algebra is constructed from two data: a field extension  $K/F$  of degree  $n$  that is Galois with cyclic Galois group  $\langle \sigma \rangle$ , and a nonzero element  $y \in F$  that satisfies the property that for any  $i = 1, \dots, n - 1, y^i$  is not a norm<sup>1</sup> from  $K$  to  $F$ . As a  $K$ -vector space, the algebra is expressible as

$$D = \bigoplus_{i=0}^{n-1} K u^i$$

where  $u$  is a symbol. The multiplication in this algebra is given by the relations  $uk = \sigma(k)u$  for all  $k \in K$ , and  $u^n = y$ . The bilinearity of multiplication, along with these relations, then allows us to determine the product of any two elements of  $D$ . One can prove that this construction indeed yields a division algebra with center  $F$ . (Such a division algebra is said to be of *index  $n$* .)

There is a well-known embedding of such a  $D$  into  $M_n(K)$  that sends  $k_0 + k_1 u + \dots + k_{n-1} u^{n-1}$  to

<sup>1</sup>This is a sufficient condition to obtain a cyclic division algebra.

$$(4) \begin{bmatrix} k_0 & \gamma\sigma(k_{n-1}) & \gamma\sigma^2(k_{n-2}) & \dots & \gamma\sigma^{n-1}(k_1) \\ k_1 & \sigma(k_0) & \gamma\sigma^2(k_{n-1}) & \dots & \gamma\sigma^{n-1}(k_2) \\ k_2 & \sigma(k_1) & \sigma^2(k_0) & \dots & \gamma\sigma^{n-1}(k_3) \\ k_3 & \sigma(k_2) & \sigma^2(k_1) & \dots & \gamma\sigma^{n-1}(k_4) \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ k_{n-2} & \sigma(k_{n-3}) & \sigma^2(k_{n-4}) & \dots & \gamma\sigma^{n-1}(k_{n-1}) \\ k_{n-1} & \sigma(k_{n-2}) & \sigma^2(k_{n-3}) & \dots & \sigma^{n-1}(k_0) \end{bmatrix}$$

By taking  $F$  to be various subfields of  $\mathbb{C}$  containing  $\mathbb{Q}(S)$  (the field generated by the elements of  $S$  over  $\mathbb{Q}$ ) in this formulation, and for each such  $F$  taking various  $K$  and  $\gamma$ , a wide variety of space-time codes can be constructed for a wide range of signal sets. For further simplicity of construction, particularly in the selection of the element  $\gamma$  above, the authors of [19] chose all their base fields  $F$  to contain transcendental elements; in most cases, their cyclic extensions  $K/F$  were of the form  $K_0(x)/F_0(x)$ , where  $K_0/F_0$  is a cyclic extension of number fields, and  $x$  is a transcendental. In these cases, the authors' construction yielded codes  $X: S^{n^2} \rightarrow M_n(\mathbb{C})$ , i.e., with  $k = n^2$ .

Alamouti's original code above arises as a special case of this formulation: the matrices of Equation (2) are just the matrices of Equation (4) above specialized to the cyclic algebra  $(\mathbb{C}/\mathbb{R}, \sigma, -1)$ , where  $\sigma$  stands for complex conjugation. This is nothing other than Hamilton's quaternions: the four-dimensional  $\mathbb{R}$  algebra  $\mathbb{R} \oplus \mathbb{R}i \oplus \mathbb{R}j \oplus \mathbb{R}k$  subject to the relations  $i^2 = j^2 = -1$ ,  $ij = -ji = k$ . (The signal set in Alamouti's construction is contained in  $K$  instead of  $F$ , unless of course  $S$  is real.)

### Satisfying the Coding Gain Criterion

The coding community immediately recognized the potential of cyclic division algebras as a fundamental tool for constructing space-time codes and began to work with the coding paradigm introduced in [19]. However, there was still a drawback. Although the specific codes of [19] certainly satisfied the rank criterion, their performance was not satisfactory. The reason for this became clear: the specific division algebras of [19] were proposed only for mathematical simplicity—merely as easy examples of the larger paradigm of division algebras—and were not optimized for the coding gain performance criterion above. The use of transcendental numbers in the codes in [19] caused the determinants of the difference matrices to come arbitrarily close to zero and limited their performance.

This situation was quickly remedied in [2] by a very clever technique. To provide a lower bound on the moduli of the determinants of the difference of code matrices, the authors Belfiore, Rekaya, and Viterbo first constructed division algebras from cyclic extensions  $K/\mathbb{Q}(i)$  and  $\gamma \in \mathbb{Z}[i]$ , but then restricted the various  $k_i$  in the matrix (4) above to entries in  $\mathcal{O}_K$ , the ring of integers of  $K$ .

The net result, as can easily be seen, is that the determinant of the difference of any two such matrices will live in  $\mathbb{Z}[i]$  and therefore will have modulus bounded below by 1. Moreover, this will be true no matter how large a subset of  $\mathbb{Z}[i]$  is used as the signal set. They called this last property the “nonvanishing determinant property”, and they called the specific code they proposed the Golden Code. It was so named for the Golden Ratio that appears naturally: it is derived from the division algebra  $(\mathbb{Q}(i, \sqrt{5})/\mathbb{Q}(i), \sigma, i)$ . Here,  $\sigma$  is the automorphism of  $K = \mathbb{Q}(i, \sqrt{5})$  that sends  $\sqrt{5}$  to  $-\sqrt{5}$  and acts as the identity on  $\mathbb{Q}(i)$ . A  $\mathbb{Z}[i]$ -basis for  $\mathcal{O}_K$  is given by 1 and  $\phi = \frac{1+\sqrt{5}}{2}$ . Write  $\psi$  for  $\sigma(\phi) = \frac{1-\sqrt{5}}{2}$ . For a signal set  $S \subset \mathbb{Z}[i] \subset \mathbb{Q}(i)$  (the most common kind of signal set), this code sends  $S^4$  to  $M_n(\mathbb{C})$  via the matrix

$$(5) \frac{1}{\sqrt{5}} \begin{pmatrix} s_{0,1}\alpha + s_{0,2}\alpha\phi & i(s_{1,1}\theta + s_{1,2}\theta\psi) \\ s_{1,1}\alpha + s_{1,2}\alpha\phi & s_{0,1}\theta + s_{0,2}\theta\psi \end{pmatrix}.$$

Here, the  $\frac{1}{\sqrt{5}}$  scale factor,  $\alpha = 1 + i(1 - \phi)$ , and  $\theta = \sigma(\alpha) = 1 + i(1 - \psi)$  are used to shape the code (more on this later). Comparing with the matrix (4) above and ignoring the scale factor, we see that  $k_0 = s_{0,1}\alpha + s_{0,2}\alpha\phi$  and  $k_1 = s_{1,1}\alpha + s_{1,2}\alpha\phi$ . Note that this code encodes four information symbols in each matrix. (A variant of this code, also based on the division algebra  $(\mathbb{Q}(i, \sqrt{5})/\mathbb{Q}(i), \sigma, i)$ , also incorporating the shaping criterion described later, is currently part of the IEEE 802.16e “WiMAX” standard. The Alamouti code based on the quaternions is also part of this standard.)

With the introduction of cyclic division algebras as a fundamental construction paradigm and with the use of codes constructed with entries from  $\mathcal{O}_K$  for suitable extensions of  $\mathbb{Q}(i)$ , the subject of space-time coding took off. It is harmless and very often actually useful to assume that the signal set  $S$  is infinite: typically,  $S$  is assumed to be one of the standard lattices  $\mathbb{Z}$ ,  $\mathbb{Z}[i]$  or the Eisenstein lattice  $\mathbb{Z}[\omega]$ , where  $\omega$  stands for the primitive third root of unity  $\frac{-1+\sqrt{-3}}{2}$ . (Under these assumptions the code forms an additive group, so one only needs to consider the rank of  $X(s_1, \dots, s_k)$  and the modulus of the determinant  $|\det \theta X(s_1, \dots, s_k)|$  in the rank criterion and the coding gain criterion.) Coding theorists immediately looked for specific constructions of division algebras of the form  $(K/F, \sigma, \gamma)$  for the cases where  $F = \mathbb{Q}$ ,  $F = \mathbb{Q}(i)$ , and  $F = \mathbb{Q}(\sqrt{-3})$ , corresponding to signal sets equaling one of the three lattices above. While such constructions have been known in principle to mathematicians working with division algebras, the coding theorists absorbed the necessary number-theoretic background in very short order and explicitly constructed division algebras over such fields for all indices  $n$  ([15] and [7]). (The

hard task here is to select  $y \in \mathcal{O}_F$  so that it has the property that  $y^i$  is not a norm from  $K$  to  $F$  for  $i = 1, \dots, n-1$ .) In all such cases, an  $\mathcal{O}_F$ -basis  $\beta_j$  of  $\mathcal{O}_K$  is chosen, and each  $k_i$  is written as  $\sum_{j=1}^n s_{i,j} \beta_j$  for  $s_{i,j}$  in the signal set. Thus  $n^2$  elements from the signal set are coded in each matrix, and by construction, the determinant of each matrix is nonzero and lies in one of the discrete lattices above. The modulus of the determinant will therefore be bounded below by the length of the shortest vector in the lattice so the code will have the nonvanishing determinant property.

### Other Performance Measures

In parallel, as the subject became better understood, several additional performance criteria started to be imposed on codes. In a fundamental paper ([25]), Zheng and Tse provided a precise quantification of the trade-off (known as the diversity-multiplexing gain or “DMG” trade-off) between information rate and reliability. They defined numerical measures for each of the benefits and showed that the pair of benefits lie in a region of the first quadrant whose upper boundary is a piecewise linear concave up curve. In the paper [7], Vijay Kumar and his students showed that all codes constructed from cyclic division algebras with the additional nonvanishing determinant property will automatically perform at the upper boundary of this region and will hence be “DMG optimal”. This of course further cemented the use of cyclic division algebras for code construction.

Another set of criteria was proposed by Oggier and coworkers in the paper [17]. One first rewrites the matrix (4) as a single  $n^2 \times 1$  vector. When  $k_i = \sum_{j=1}^n s_{i,j} \beta_j$  for  $s_{i,j}$  in the signal set and  $\beta_j$  an  $\mathcal{O}_F$  basis for  $\mathcal{O}_K$ , this  $n^2 \times 1$  vector can be expressed as  $M \cdot \bar{v}$ , where  $M$  is an  $n^2 \times n^2$  matrix and  $\bar{v}$  is the column vector  $(s_{0,1}, s_{0,2}, \dots, s_{i,j}, \dots, s_{n-1,n})^T$ . One now requires that the matrix  $M$  be *unitary* and that  $|y| = 1$ . The first condition is called good shaping, and the idea behind it is that this forces the average energy needed to send the vector  $\bar{v}$  without coding to be the same as that needed to send it in the coded matrix form (4). The condition  $|y| = 1$  causes the average energy transmitted per antenna to be equal for all transmissions. Oggier and coworkers called such codes “perfect” and constructed perfect codes for  $n = 2, 3, 4$ , and 6. This was followed by work of Elia and coworkers ([8]), who constructed perfect codes for all values of  $n$  and additionally showed that perfect codes satisfy other information-theoretic properties such as information-losslessness (a concept introduced by Damen and coworkers in [4]) and approximate universality (a concept introduced by Tavildar and Viswanath in [23]).

The mathematics needed for the work on perfect codes is quite interesting. Analyzing the condition that  $M$  be unitary, we find that it is sufficient to make the following matrix unitary:

$$U(\{\beta_1, \dots, \beta_n\}) = \begin{bmatrix} \beta_1 & \cdots & \beta_n \\ \sigma(\beta_1) & \cdots & \sigma(\beta_n) \\ \vdots & & \vdots \\ \sigma^{n-1}(\beta_1) & \cdots & \sigma^{n-1}(\beta_n) \end{bmatrix}.$$

Here, it is not necessary that the  $\beta_j$  be an  $\mathcal{O}_F$  basis of  $\mathcal{O}_K$ ; it is sufficient that they be an  $\mathcal{O}_F$  linearly independent subset of  $\mathcal{O}_K$ . (So, for example, in the Golden Code (5) above,  $\alpha$  is chosen so that with  $\beta_1 = \alpha$  and  $\beta_2 = \alpha\phi$ , the matrix

$$\begin{pmatrix} \alpha & \alpha\phi \\ \theta & \theta\psi \end{pmatrix}$$

is unitary after being multiplied by the  $\frac{1}{\sqrt{5}}$  scale factor.) So the question is: how to find  $\mathcal{O}_F$  submodules of  $\mathcal{O}_K$  that satisfy this unitary condition? For  $n = 2^b$ , it is easy to see that for the field  $K = \mathbb{Q}(\zeta)$  and  $F = \mathbb{Q}(i)$ , where  $\zeta$  is a primitive  $2^{b+2}$ th root of unity, the various powers of  $\zeta$  are  $\mathbb{Z}[i]$ -linearly independent and satisfy the unitary condition above. For odd  $n$ , Elia and coworkers use a construction due to B. Erez [9] that was needed in a different context: Erez was showing that for certain cyclic extensions  $K/\mathbb{Q}$  with Galois group  $G$ , the square root of the inverse different is a free  $\mathbb{Z}[G]$  module that has an orthogonal basis with respect to the usual trace form on  $K$  that sends  $x, y$  to  $\text{Tr}_{K/\mathbb{Q}}(xy)$ .

The most recent performance criteria for space-time codes, and in some sense the most mathematically exciting, have come from Lahtonen and coworkers ([13]). For the usual cases in which  $S$  is one of  $\mathbb{Z}$ ,  $\mathbb{Z}[i]$ ,  $\mathbb{Z}[\omega]$ , it is easy to see from the linearity of the code matrices  $X$  that on writing each  $X$  as an  $n^2 \times 1$  vector as above and separating the real and imaginary parts, one gets a full lattice in  $\mathbb{R}^{2n^2}$ , i.e., the additive group generated by  $2n^2$  linearly independent vectors in  $\mathbb{R}^{2n^2}$ . We refer to this lattice as the *code lattice*. After normalizing all code matrices so that  $\inf_{X \in \mathcal{X}} |\det(X)| = 1$ , they postulate that codes whose lattice points are the most dense in  $\mathbb{R}^{2n^2}$  will have the best performance, and, indeed, they find this is borne out in several circumstances by simulations. To obtain a suitable numerical measure for the relative density, they invert the situation: they normalize the code lattice to have fundamental volume 1 instead. Thus they define the normalized minimum determinant of a code lattice  $\Lambda$  of rank  $2n^2$  in a  $\mathbb{Q}(i)$  division algebra of index  $n$  (embedded in  $M_n(\mathbb{C})$ ) as the minimum of the moduli of the determinants  $|\det(X(s_1, \dots, s_{n^2}))|$  as  $X(s_1, \dots, s_{n^2})$  runs through the lattice, divided by the fundamental volume of  $\Lambda$ . Since a smaller fundamental volume represents

a higher density, the goal is to construct codes whose code lattice  $\Lambda$  would maximize this ratio among all full lattices in the division algebra.

Recall that if  $D$  is a division algebra with center  $F$  and if  $R$  is a subring of  $F$  whose quotient field is  $F$ , then an  $R$ -order in  $D$  is a subring  $T$  of  $D$  containing  $R$  that is finitely generated as an  $R$ -module and satisfies  $TF = D$ . A maximal  $R$ -order is one that is maximal with respect to inclusion. In the typical situation where  $S$  is one of  $\mathbb{Z}$ ,  $\mathbb{Z}[i]$ , or  $\mathbb{Z}[\omega]$ , so  $F$  is one of  $\mathbb{Q}$ ,  $\mathbb{Q}(i)$ , or  $\mathbb{Q}(\sqrt{-3})$ , and where the  $k_i$  of the matrices in (4) are constrained to lie in  $\mathcal{O}_K$  and  $\gamma \in \mathcal{O}_F$ , the code matrices of (4) naturally form an  $S$ -order. Thus the code matrices have a dual structure of an  $S$ -order and a full lattice in  $\mathbb{R}^{2n^2}$ . Lahtonen and coworkers investigate the interplay between these two structures. They ask: How will the code's performance as measured by its normalized minimum determinant vary if, in addition to carrying its natural structure of a full  $\mathbb{Z}$ -lattice in  $\mathbb{R}^{2n^2}$ , we choose our code to form an arbitrary  $S$ -order inside an  $F$ -division algebra? In these cases, the minimum modulus of the determinants of the code matrices is 1, so it follows from the definition of the normalized minimum determinant that the smaller the fundamental volume of the lattice the better the code. If  $T_1$  and  $T_2$  are  $S$ -orders and  $\Lambda_{T_1}$  and  $\Lambda_{T_2}$  the corresponding lattices with fundamental volumes  $V_{T_1}$  and  $V_{T_2}$ , then  $T_1 \subseteq T_2$  implies  $\Lambda_{T_1} \subseteq \Lambda_{T_2}$ , which in turn means that  $V_{T_2} \leq V_{T_1}$ . It follows therefore that the best normalized minimum determinant will arise when a maximal order is used for the code. The authors then relate the fundamental volume of the code lattice to the  $\mathbb{Z}$ -discriminant of the maximal order and then invoke known formulas for discriminants of maximal orders to compute the best normalized minimum determinant of codes arising from  $\mathcal{O}_F$  orders inside a given division algebra. In particular, they show (for the fields  $\mathbb{Q}(i)$ ,  $\mathbb{Q}(\sqrt{-3})$  and  $\mathbb{Q}$ ) that the best division algebras to use will be ones that are ramified at precisely two of the "smallest" primes of the field (where the size of a prime  $P = \langle \pi \rangle$  is defined to be the modulus  $|\pi|$ ). Thus, for  $\mathbb{Q}(i)$  for example, one needs to transmit on a code arising from a maximal order inside a division algebra ramified only at  $(1+i)$  and  $(2+i)$  (or  $(2-i)$ ). (Much of this was part of Vehkalahti's Ph.D. thesis.)

One of the drawbacks of using maximal orders is that the corresponding code lattice may not have good shape. Thus optimizing a code for minimum normalized determinant may destroy any optimization for shape. The recent work of Raj Kumar and Caire ([3]) proposes a very clever technique of mapping lattice points to certain *cosets* of a suitably chosen sublattice of a standard cubic lattice; this smooths out an irregular lattice and gives it better shape. In particular, their technique

applies to codes from lattices from maximal orders and provides a further performance boost in such cases.

### Key Challenge: Decoding

What are some of the key problems that need to be solved in space-time codes? Perhaps the biggest engineering challenge in the subject is the issue of decoding. The problem quite simply is the following: given the received vectors in  $Y$  (see Equation (1)), determine the entries of the matrix  $X$  that represent the original information. Assume that  $k$  symbols are coded in the matrix  $X$  and that the entries of  $X$  are linear in the signal entries  $s_1, \dots, s_k$  (typically arising from  $\mathbb{Z}$ ,  $\mathbb{Z}[i]$ , or  $\mathbb{Z}[\omega]$ ). By writing out sequentially the real and imaginary parts of each entry of  $Y$ ,  $W$ , and  $s_1, \dots, s_k$ , we may rewrite Equation (1) as  $\tilde{Y} = Z\bar{v} + \tilde{W}$ . Here  $Z$  is a  $2n^2 \times 2k$  real matrix that depends on  $H$ ,  $\theta$ , and the parameters of the code matrix  $X$ ,  $\bar{v}$  is the signal vector  $(x_1, y_1, \dots, x_i, y_i, \dots, x_k, y_k)^T$  with  $x_i$  and  $y_i$  being the real and imaginary parts of  $s_i$ , and similarly for  $\tilde{Y}$  and  $\tilde{W}$ . If the columns of  $Z$  were orthonormal, decoding would be quite simple: we would have  $Z^T \tilde{Y} = \bar{v} + Z^T \tilde{W}$  with  $Z^T \tilde{W}$  also having independent, identically distributed Gaussian entries. Hence, under maximum likelihood estimation,  $\bar{v}$  can be taken to be the closest vector in  $S^k$  (viewed inside the Euclidean space  $\mathbb{R}^{2k}$ ) to  $Z^T \tilde{Y}$ . This is a very simple and computationally fast scheme: we march through  $Z^T \tilde{Y}$  component pair by component pair, and we find the element of the signal lattice  $S$  closest to that component pair.

The process above is called *single symbol decoding*. (For  $k < n^2$  this is the same as orthogonal projection onto the subspace of  $\mathbb{R}^{2n^2}$  determined by the columns of  $Z$ .) There are some nice situations in which the matrix  $Z$  is (essentially) orthogonal; this happens in the case of the Alamouti code, and more generally, in the codes satisfying Equation (3). The matrix  $Z$  for such codes satisfies  $ZZ^T = \theta^2 \text{Tr}(HH^+) I_{2k}$ . We may divide the relation  $\tilde{Y} = Z\bar{v} + \tilde{W}$  by  $\theta\sqrt{\text{Tr}(HH^+)}$ . The entries of the new noise vector  $1/(\theta\sqrt{\text{Tr}(HH^+)})\tilde{W}$  are still independent identically distributed Gaussian, while the columns of the matrix  $1/(\theta\sqrt{\text{Tr}(HH^+)})Z$  are now orthonormal. Thus single symbol decoding can be employed in all these cases.

But for other codes  $Z$  is rarely orthogonal! In general, given that the entries of  $W$  are independent identically distributed Gaussian, for maximum likelihood estimation one needs to search in  $S^{n^2}$  (viewed inside the Euclidean space  $\mathbb{R}^{2n^2}$ ) for that vector  $\bar{v} = (x_1, y_1, \dots, x_{n^2}, y_{n^2})^T$  such that  $Z\bar{v}$  is closest to  $\tilde{Y}$ . (Here we will assume that  $k = n^2$ , as is usually the case for codes from cyclic division algebras.) This can no longer be

accomplished symbol by symbol, and one needs to search in the full space  $S^{n^2}$  instead of just in  $S$ . There is an algorithm called the *sphere decoding algorithm* (see [5], for instance) that accomplishes this search in an intelligent manner, but as is to be expected of any search in  $S^{n^2}$ , even this algorithm gets very cumbersome once  $n$  exceeds 2. (One must keep in mind that the search for  $\bar{v}$  such that  $Z\bar{v}$  is closest to  $\tilde{Y}$  is essentially a closest lattice point search, and this is known to be NP-hard. What saves the day is that the received vectors  $\tilde{Y}$  are not random but have a Gaussian distribution about the lattice vectors  $Z\bar{v}$ . In [12], Hassibi and Vikalo show that under certain technical assumptions, the expected complexity of the sphere decoding algorithm is polynomial, although the worst-case complexity is exponential.)

Since  $Z$  is rarely orthogonal, we may ask whether we can take advantage of the obvious algebraic structure of the code and simplify the closest vector problem for our particular application. A very clever set of ideas of Luzzi et al. [16] does just that and gives an approximate solution to the decoding problem for the Golden Code (Equation (5)) by reducing the situation to the action of  $SL_2(\mathbb{C})$  on three-dimensional hyperbolic space  $\mathbb{H}^3$ . Their work is a veritable tour de force of the application of abstract mathematics to engineering problems. Their goal is to approximate the channel matrix  $H$  (normalized to have determinant 1) by an element  $U$  of determinant 1 in the  $\mathbb{Z}[i]$ -order  $R = (\mathcal{O}_K/\mathbb{Z}[i], \sigma, \iota)$ . Writing  $H = EU$  with  $E$  simply being the error  $HU^{-1}$ , they argue that choosing  $U$  so that the Frobenius norm of  $E^{-1} = UH^{-1}$  is minimized approximates the original problem by the following: given a vector  $Y$  in  $\mathbb{C}^{n^2}$  and an unknown vector  $S$  in  $\mathbb{Z}[i]^{n^2}$  determine a “best” estimate of  $S$  if the difference vector  $W = Y - S$  is known to be *approximately* independent identically distributed Gaussian (in a suitable sense). Given this assumption about the noise vector  $W$ , a reasonable way to proceed is to assume that  $W$  is actually independent identically distributed Gaussian. In this situation, the maximum-likelihood estimate of  $S$  is obtained by taking the  $i$ th entry of  $S$  to be the lattice point in  $\mathbb{Z}[i]$  closest to the  $i$ th entry of  $Y$ . The authors find that their scheme gives a fast and acceptably accurate decoding.

What is fascinating is the mathematics behind their choice of  $U$ . First, they need to determine generators and relations for the group of norm 1 units  $\mathcal{U}_1(R)$  of  $R$  (i.e., the set of multiplicatively invertible elements of  $R$  whose determinant as a code matrix is 1). In general, it is very difficult to find these for orders in division algebras, but in the case of certain special quaternion algebras over number fields, generators and relations for  $\mathcal{U}_1(R)$  are known. Much of the idea behind this goes back to Poincaré. The norm 1 units in the order  $R$

above (modulo the subgroup  $\{\pm 1\}$ ) turns out to be a Kleinian group, i.e., a discrete subgroup of the projective special linear group  $PSL_2(\mathbb{C})$ . As a subgroup of  $PSL_2(\mathbb{C})$ ,  $U_1(R)$  (modulo  $\{\pm 1\}$ ) acts on the upper half-space model of hyperbolic 3-space  $\mathbb{H}^3$  as a group of orientation-preserving isometries, and Poincaré’s fundamental polyhedron theorem gives a set of generators and relations for such a group in terms of certain automorphisms of a fundamental domain for the group. Given a point  $P$  in  $\mathbb{H}^3$ , the Dirichlet polyhedron centered on  $P$  is the closure of the set of points  $x$  such that  $d_H(x, P) < d_H(g(x), P)$  for all  $g \in U_1(R)$  (modulo  $\{\pm 1\}$ ),  $g \neq 1$ , where  $d_H$  is the hyperbolic metric on  $\mathbb{H}^3$ . The authors construct a Dirichlet polyhedron centered on  $J = (0, 0, 1)$ ; this is a fundamental domain for  $U_1(R)$ . From this polyhedron, using Poincaré’s theorem and a computer search, they determine a set of generators of  $\mathcal{U}_1(R)$ . They do this ahead of time and store the results. Next, in real time, given a fading matrix  $H$  (normalized to have determinant 1), they need to find an element  $U$  of  $\mathcal{U}_1(R)$  such that the Frobenius norm of  $UH^{-1}$  is minimized. They observe that viewing  $UH^{-1}$  as an element of  $PSL_2(\mathbb{C})$  acting on  $\mathbb{H}^3$ , the Frobenius norm of  $UH^{-1}$  is just  $2 \cosh d_H(J, UH^{-1}(J))$ , where  $J$  and  $d_H$  are as above. Since  $U$  is an isometry, they must find  $U \in \mathcal{U}_1(R)$  that minimizes  $\cosh d_H(U^{-1}(J), H^{-1}(J))$ . From the definition of Dirichlet polyhedra, this means that they need to find a Dirichlet polyhedron centered on some  $U^{-1}(J)$  which contains  $H^{-1}(J)$ . They use the geometry of  $\mathbb{H}^3$  relative to the action of  $U_1(R)$  to find such a  $U$ : they just need to repeatedly consider the various Dirichlet polyhedra centered on  $J$  and the various  $g_i(J)$ , where the  $g_i$  run through the generators of  $\mathcal{U}_1(R)$  that they have computed ahead of time, along with their inverses.

### Role of Mathematicians

What is the role of mathematicians in this field? The subject is clearly very mathematical; yet, unlike classical coding theory, which now has a mathematical life of its own and can, for instance, be thought of as a theory of subspaces of vector spaces over finite fields, the center of gravity of space-time codes currently lies very solidly in engineering. There is as yet no deep independent “mathematics of space-time codes”: the driving force behind the subject consists of fundamental engineering problems that need to be solved before MIMO wireless communication reaches its full practical potential, particularly for three or more antennas. This author therefore believes that, as things stand now, isolated mathematical investigations of space-time codes that are not grounded in concrete engineering questions would very likely lead to sterile results. At least for now, mathematicians can best contribute to the subject by

working in collaboration with engineers who are motivated by fundamental engineering questions. This author has found that the leading engineers in the field already have a practical and intuitive understanding of much abstract mathematics but welcome help from trained mathematicians. (This author has also found that they are a genuine pleasure to collaborate with.) There is clearly a lot of work for mathematicians to do: particularly in decoding systems with large numbers of receive and transmit antennas, but also in other areas of MIMO communication that we have not touched upon in this article, such as cooperative communication in networks, or noncoherent communication, where the matrix  $H$  is not known to either the receiver or the transmitter.

## References

- [1] S. ALAMOUTI, A transmitter diversity scheme for wireless communications, *IEEE J. Select. Areas Commun.* **16**, no. 10 (1998), 1451–1458.
- [2] J.-C. BELFIORE, G. REKAYA and E. VITERBO, The Golden Code: A  $2 \times 2$  full-rate space-time code with nonvanishing determinants, *Proc. IEEE Int. Symp. Inform. Th. (ISIT 2004)*.
- [3] GIUSEPPE CAIRE and RAJ KUMAR, Space-time codes from structured lattices, accepted for publication in *IEEE Trans. Inform. Theory*, available at <http://arxiv.org/abs/0804.1811>.
- [4] M. O. DAMEN, A. TEWFIK, and J.-C. BELFIORE, A construction of a space-time code based on number theory, *IEEE Trans. Inform. Theory* **48**, no. 3 (2002), 753–761.
- [5] M. O. DAMEN, H. EL GAMAL, and G. CAIRE, On maximum likelihood detection and the search for the closest lattice point, *IEEE Trans. Inform. Theory* **49**, no. 10 (2003), 2389–2402.
- [6] BENO ECKMANN, Hurwitz-Radon matrices revisited: From effective solution of the Hurwitz matrix equations to Bott periodicity, *Mathematical Survey Lectures 1943–2004*, Springer-Verlag, Berlin, 2006.
- [7] P. ELIA, K. RAJ KUMAR, S. A. PAWAR, P. VIJAY KUMAR, and H-F. LU, Explicit, minimum-delay space-time codes achieving the diversity-multiplexing gain tradeoff, *IEEE Trans. Inform. Theory* **52**, no. 9 (2006), 3869–3884.
- [8] P. ELIA, B. A. SETHURAMAN, and P. VIJAY KUMAR, Perfect space-time codes for any number of antennas, *IEEE Transactions on Information Theory* **53**, no. 11 (2007), 3853–3868.
- [9] B. EREZ, The Galois structure of the trace form in extensions of odd prime degree, *Journal of Algebra* **118** (1988), 438–446.
- [10] G. J. FOSCHINI and M. GANS, On the limits of wireless communication in a fading environment when using multiple antennas, *Wireless Personal Communication*, March 1998.
- [11] J.-C. GUEY, M. P. FITZ, M. R. BELL, and W.-Y. KUO, Signal design for transmitter diversity wireless communication systems over Rayleigh fading channels, in *Proc. IEEE Vehicular Technology Conf (VTC96)*, 2000, 136–140.
- [12] B. HASSIBI and H. VIKALO, On sphere decoding algorithm. I. Expected complexity, *IEEE Trans. Signal Process.* **53**, no. 8 (2005), 2806–2818.
- [13] C. HOLLANTI, J. LAHTONEN, K. RANTO and R. VEHKALAHTI, On the densest MIMO lattices from cyclic division algebras, *IEEE Trans. Inform. Theory* **55**, no. 8 (2009), 3751–3780.
- [14] G. IVANYOS and L. RÓNYAI, On the complexity of finding maximal orders in algebras over  $\mathbb{Q}$ , *Computational Complexity* **3** (1993), 245–261.
- [15] T. KIRAN and B. SUNDAR RAJAN, STBC-schemes with nonvanishing determinant for certain number of transmit antennas, *IEEE Trans. Inform. Theory* **51**, no. 8 (2005), 2984–2992.
- [16] LAURA LUZZI, GHAYA REKAYA-BEN OTHMAN, and JEAN-CLAUDE BELFIORE, Algebraic reduction for space-time codes based on quaternion algebras, arXiv:0809.3365v2[cs.IT]. See also Algebraic Reduction for the Golden Code, *Proc. IEEE International Conference on Communications (ICC 2009)*.
- [17] F. OGGIER, G. REKAYA, J. C. BELFIORE, and E. VITERBO, Perfect space-time block codes, *IEEE Trans. Inform. Theory* **52**, no. 9 (2006), 3885–3902.
- [18] C. E. SHANNON, A mathematical theory of communication, *Bell System Technical Journal* **27** (1948), 379–423 and 623–656.
- [19] B. A. SETHURAMAN, B. SUNDAR RAJAN, and V. SHASHIDHAR, Full-diversity, high-rate, space-time block codes from division algebras, *IEEE Trans. Inform. Theory* **49**, no. 10 (2003), 2596–2616.
- [20] OLAV TIRKKONEN and ARI HOTTINEN, Square-matrix embeddable space-time block codes for complex signal constellations, *IEEE Trans. Inform. Theory* **48**, no. 2 (2002), 384–395.
- [21] V. TAROKH, N. SESHADRI, and A. R. CALDERBANK, Space-time codes for high data rate wireless communication: Performance criterion and code construction, *IEEE Trans. Inform. Theory* **44**, no. 2 (1998), 744–765.
- [22] V. TAROKH, H. JAFARKHANI, and A. R. CALDERBANK, Space-time block codes from orthogonal designs, *IEEE Trans. Inform. Theory* **45**, no. 5 (1999), 1456–1467.
- [23] S. TAVILDAR and P. VISWANATH, Approximately universal codes over slow-fading channels, *IEEE Trans. Inform. Theory* **52**, no. 7 (2006), 3233–3258.
- [24] E. TELATAR, Capacity of multi-antenna Gaussian channels, *Europ. Trans. Telecomm.* **10**, no. 6 (1999), 585–595.
- [25] L. ZHENG and D. TSE, Diversity and multiplexing: A fundamental tradeoff in multiple-antenna channels, *IEEE Trans. Inform. Theory* **49**, no. 5 (2003), 1073–1096.