Computer classification of linear codes

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Abstract

We present algorithms for classification of linear codes over finite fields, based on canonical augmentation and on lattice point enumeration. We apply these algorithms to obtain classification results over fields with 2, 3 and 4 elements. We validate a correct implementation of the algorithms with known classification results from the literature, which we partially extend to larger ranges of parameters.

Index Terms

Linear code, classification, enumeration, code equivalence, lattice point enumeration, canonical augmentation.

I. INTRODUCTION

INEAR codes play a central role in coding theory for several reasons. They permit a compact representation via generator matrices as well as efficient coding and decoding algorithms. Due to their simplicity, i.e., they are just subspaces of a vector space, they have numerous applications in e.g. algebraic geometry and other branches of mathematics as well as biology or physics. Also multisets of points in the projective space $PG(k-1,\mathbb{F}_q)$ of cardinality *n* correspond to linear $[n,k]_q$ codes, see e.g. [1]. So, let q be a prime power and \mathbb{F}_q be the field of order q. A q-ary linear code of length n, dimension k, and minimum (Hamming) distance d is called an $[n,k,d]_q$ code. If we do not want to specify the minimum distance d, then we also speak of an $[n,k]_q$ code or of an $[n,k,\{w_1,\ldots,w_l\}]_q$ if the non-zero codewords have weights in $\{w_1, \ldots, w_l\}$. If for the binary case q = 2 all weights w_i are divisible by 2, we also speak of an even code. We can also look at those codes as k-dimensional subspaces of the Hamming space \mathbb{F}_{q}^{n} . Two linear codes of the same length and dimension are equivalent if one can be obtained from the other by a sequence of the following transformations: (1) a permutation of the coordinate positions of all codewords; (2) a multiplication of a coordinate of all codewords with a nonzero element from \mathbb{F}_q ; (3) a field automorphism. An $[n,k]_q$ code can be represented by a generator matrix $G \in \mathbb{F}_q^{k \times n}$ whose row space gives the set of all q^k codewords of the code. In the remaining part of the paper we always assume that the length n of a given linear code equals its effective length, i.e., for every coordinate there exists a codeword with a non-zero entry in that coordinate. While a generator matrix gives a compact representation of a linear code it is far from being unique. Special generator matrices are so-called systematic generator matrices, which contain a $k \times k$ unit matrix in the first k columns. If we apply row operations of the Gaussian elimination algorithm onto a generator matrix we do not change the code itself but just its representation via a generator matrix. Applying the transformations (1)-(3), mentioned above, we can easily see that each $[n,k]_q$ code admits an equivalent code with a systematic generator matrix. Already in 1960 Slepian has enumerated binary linear codes for small parameters up to equivalence (or isometry) [2]. The general classification problem for $[n,k]_q$ codes has not lost its significance since then, see e.g. [3]. In [4] all optimal binary linear $[n, k, d]_2$ codes up to length 30 have been completely classified, where in this context optimal means that no $[n-1,k,d]_2$, $[n+1,k+1,d]_2$, or $[n+1,k,d+1]_2$ code exists. Classification algorithms for linear codes have been presented in [5], see also [6, Section 7.3]. A software package Q-Extension is publicly available, see [7] for a description.

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The aim of this paper is to present two algorithmic variants for the classification problem for linear codes. The first one is implemented in the program Generation which is a part of the software package QextNewEdition, and the second one is implemented in the program LinCode. As the implementation of such a software is a delicate issue, we exemplarily verify several classification results from the literature and partially extend them. Both algorithms are well suited for parallelization. As mentioned in [5], one motivation for the exhaustive enumeration of linear codes with some specific parameters is that afterwards the resulting codes can be easily checked for further properties. Exemplarily we do here so for the number of minimal codewords of a linear code, see Subsection IV-B.

The remaining part of the paper is organized as follows. In Section II we present two versions of our algorithm for canonical augmentation - extension of a generator matrix column by column or row by row. The details and the theoretical foundation of the other algorithm is given in Section III. Numerical enumeration and classification results for linear codes are listed in Section IV. Finally, we draw a brief conclusion in Section V.

II. CLASSIFICATION OF LINEAR CODES USING CANONICAL AUGMENTATION

The concept of canonical augmentation is introduced in [6] and [8]. The main idea is to construct only nonequivalent objects (in our case - inequivalent linear codes) and in this way to have a classification of these objects. The construction is recursive, it consists of steps in which the nonequivalent objects are obtained from smaller objects by expanding in a special way. The canonical augmentation uses a canonical form to check the so called "parent test" and considers only objects that have passed the test.

The technique of canonical augmentation has been used for classification of special types of codes and related combinatorial objects in [9]–[12], etc. The algorithms in the pointed works construct objects with the needed parameters recursively starting from the empty set. In this way, to classify all linear $[n,k]_q$ codes column by column, the codes of lengths 1, 2, ..., n and dimensions $\leq k$ are also constructed in the generation process. One of the important differences of the algorithm presented here is that we consider only codes with dimension k.

A. Preliminaries

To construct all inequivalent codes with given parameters means to have one representative of each equivalence class. To do this, we use the concept for a canonical representative, selected on the base of some specific conditions. This canonical representative is intended to make easily a distinction between the equivalence classes.

Let G be a group acting on a set Ω . This action defines an equivalence relation such that the equivalence classes are the G-orbits in Ω . We wish to find precisely one representative of each G-orbit and therefore we use a so-called canonical representative map.

Definition 1 [6] A canonical representative map for the action of the group G on the set Ω is a function $\rho: \Omega \to \Omega$ that satisfies the following two properties:

1) for all $X \in \Omega$ it holds that $\rho(X) \cong X$,

2) for all $X, Y \in \Omega$ it holds that $X \cong Y$ implies $\rho(X) = \rho(Y)$.

For $X \in \Omega$, $\rho(X)$ is the canonical form of X with respect to ρ . Analogously, X is in canonical form if $\rho(X) = X$. The configuration $\rho(X)$ is the canonical representative of its equivalence class with respect to ρ . We can take for a canonical representative of one equivalence class a code which is more convenient for our purposes.

We take Ω to be the set of all linear $[n,k, \geq d]_q$ codes with dual distance at least d^{\perp} , and *G* be the semidirect product $(\mathbb{F}_q^* \wr S_n) \rtimes_{\theta} \operatorname{Aut}(\mathbb{F}_q)$ where $\theta : \operatorname{Aut}(\mathbb{F}_q) \to \operatorname{Aut}(\mathbb{F}_q^* \wr S_n)$ is a homomorphism such that $\theta_{\alpha}((z,h)) = (\alpha(z),h)$ for all $\alpha \in \operatorname{Aut}(\mathbb{F}_q)$ and $(z,h) \in \mathbb{F}_q^* \wr S_n$ (for more details see [6]). The elements of *G* fix the minimum and the dual distance of the codes. Using that $\mathbb{F}_q^* \wr S_n \cong Mon_n(\mathbb{F}_q)$ where $Mon_n(\mathbb{F}_q)$

We use one more group action. The automorphism group of the code *C* acts on the set of coordinate positions and partitions them into orbits. The canonical representative map ρ induces an ordering of these orbits. The all-zero coordinates, if there are any, form an orbit which we denote by O_a . If the code contains codewords of weight 1 then their supports form one orbit, say O_b . The orbits for the canonical representative code $\rho(C)$ are ordered in the following way: $O_1^{(\rho)}$ contains the smallest integer in the set $\{1, 2, \ldots, n\} \setminus (O_a^{(\rho)} \cup O_b^{(\rho)}), O_2^{(\rho)}$ contains the smallest integer which is not in the set $O_a^{(\rho)} \cup O_b^{(\rho)} \cup O_1^{(\rho)}$, etc. If ϕ maps the code *C* into its canonical form $\rho(C)$ then the permutational part π_{ϕ} of ϕ maps the orbits of $\rho(C)$. Obviously, $\phi(O_a) = O_a^{(\rho)}$ and $\phi(O_b) = O_b^{(\rho)}$. If $\pi_{\phi}(O_{i_s}) = O_s^{(\rho)}$ then $O_{i_1} \prec O_{i_2} \prec \cdots \prec O_{i_m}$. We call the first orbit O_{i_1} special and denote it by $\sigma(C)$. If $\{1, 2, \ldots, n\} = O_a \cup O_b$ then the code contains only codewords with weights 0 and 1, and in this case we do not define a special orbit.

Example 1 If we order the codewords in a code lexicographically and then compare the codes according to a lexicographical ordering of the vectors obtained by concatenation of the ordered nonzero codewords, we can take the smallest code in any equivalence class as a canonical representative. This type of canonical map is very easy to define but computationally expensive to implement. Consider the binary code C generated by the matrix $G_C = \begin{pmatrix} 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \end{pmatrix}$ in details. The automorphism group of C is $Aut(C) = \{id, (13), (24), (13)(24)\}$. If Ω_C is the equivalence class of C then $\Omega_C = \{C_1, \ldots, C_6\}$, $C_i = \{0, v_1^{(i)}, v_2^{(i)}, v_3^{(i)}\}$, $v_1^{(i)} \prec v_2^{(i)} \prec v_3^{(i)}$. We order the codes in Ω_C in the following way:

$$C_i \prec C_j \iff (v_1^{(i)}, v_2^{(i)}, v_3^{(i)}) \prec (v_1^{(j)}, v_2^{(j)}, v_3^{(j)}).$$

Therefore, $C = \{0,0101,1011,1110\} \succ C_1 = \{0,0011,1101,1110\}$. Hence the code C_1 is the canonical form of C, $C_1 = \rho(C)$. The coordinates of C_1 are partitioned into two orbits under the action of its automorphism group, namely $O_1 = \{1,2\} \prec O_2 = \{3,4\}$. For the code C the special orbit is $\sigma(C) = \{1,3\}$.

To find a canonical form of a code, we use the algorithm described in [13]. Similarly to the McKay's program nauty [14], this algorithm gives in addition to canonical form, also generating elements of the automorphism group of the considered code. Note that if the coordinates are previously partitioned according to suitable invariants, the algorithm works much faster.

Using the concept of canonical augmentation, we have developed an algorithm in two variants.

B. Algorithm 1

The first algorithm is a canonical augmentation column by column. We are looking for all inequivalent linear codes with length *n*, dimension *k*, minimum distance $\geq d$ and dual distance at least $d^{\perp} \geq 2$. Without loss of generality we can consider the generator matrices in the form $(I_k|A)$ where *A* is a $k \times (n-k)$ matrix. To obtain the codes we use a recursive construction starting with the identity matrix I_k which generates the trivial $[k,k,1]_q$ code. In the *i*-th step we add a column to the considered generator matrices of the obtained $[k+i-1,k]_q$ codes but we take only those columns which gives codes of length k+i with minimum distance $\geq d_i = d - (n-k) + i$ and dual distance at least d^{\perp} . A strategy for effective generation of these vectors (columns) is described in [15]. Since $d \leq n-k+1$, the minimum distance in the beginning is

 ≤ 1 (it is equal to 1 as we begin with the trivial code). The codes obtained from a code *C* in this way form the set Ch(C) and they are called the children of *C*. We say that the code $\overline{C} \in Ch(C)$ passes the parent test, if the added coordinate belongs to the special orbit $\sigma(\overline{C})$. Moreover, we define an action of the automorphism group Aut(*C*) on the set of all vectors in \mathbb{F}_q^k and take only one representative from each orbit. By $Ch^*(C)$ we denote a subset of Ch(C) consisting of the codes constructed by *C* and the taken representatives.

Algorithm 1 Canonical augmentation column by column

Input: The trivial $[k,k,1]_q$ code C_k **Output:** A set U_n of linear $[n,k, \ge d]_q$ codes with dual distance $\ge d^{\perp}$ **Function Augmentation**(A: linear code of dimension k); if the length of A is equal to n then $U_n := U_n \cup \{A\}$; else for all codes $B \in Ch^*(A)$ do if B passes the parent test then Augmentation(B); end if end for end if Function Main; $U_n = \emptyset$ Augmentation(C_k);

Using some lemmas we will prove the following theorem:

Theorem 2 The set U_n obtained by Algorithm 1 consists of all inequivalent $[n,k, \ge d]_q$ codes with dual distance at least d^{\perp} .

The main idea is to prove that Algorithm 1 gives a tree of codes with root the trivial code C_k . The codes obtained in level *i* represents all inequivalent $[k+i,k]_q$ codes with minimum distance at least d_i and dual distance at least d^{\perp} . Denote the set of these codes by U_{k+i} . We have to prove that all constructed codes in U_{k+i} are inequivalent, and that any $[k+i,k]_q$ code with needed minimum and dual distance is equivalent to a code in this set.

The first lemma proves that the equivalence test for codes that pass the parent test and are obtained from non-equivalent parent codes is not necessary.

Lemma 3 If B_1 and B_2 are two equivalent linear [n,k,d] codes which pass the parent test, their parent codes are also equivalent.

PROOF. Let $B = \rho(B_1) = \rho(B_2)$ be the canonical representative of the equivalence class of the considered codes. Since both codes pass the parent test, then the added column is in the special orbit of both codes, or $n \in \sigma(B_i)$, i = 1, 2. This means that there is a map ψ that maps B_1 to B_2 and the permutational part of ψ fixes *n*-th coordinate. Hence $\psi = (M, \alpha)$, $M = \begin{pmatrix} M_1 & 0 \\ 0 & \lambda \end{pmatrix} \in Mon_n(\mathbb{F}_q)$, $\lambda \in \mathbb{F}_q^*$, $\alpha \in Aut(\mathbb{F}_q)$, and (M_1, α) maps the parent code of B_1 to the parent code of B_2 . Hence both parent codes are equivalent. \Box

Lemma 4 Let A_1 and A_2 be two equivalent linear codes of length r and dimension k. Then for any child code B_1 of A_1 which passes the parent test, there is a child code B_2 of A_2 , equivalent to B_1 , such that B_2 also passes the parent test.

PROOF. Let G_1 be a generator matrix of A_1 in systematic form, and $A_2 = \psi(A_1)$, $\psi = (M, \alpha)$, $M \in Mon_r(\mathbb{F}_q)$, $\alpha \in \operatorname{Aut}(\mathbb{F}_q)$. Let B_1 be the code generated by $(G_1|a^T)$, $a \in \mathbb{F}_q^k$, and B_2 be the code generated by the matrix $G_2 = \psi(G_1)$ and the vector $b^T = (a^\alpha)^T$, where a^α is obtained from a by applying the field automorphism α to all coordinates. Extend the map ψ to $\widehat{\psi} = \left(\begin{pmatrix} M & 0 \\ 0 & 1 \end{pmatrix}, \alpha\right) \in Mon_{r+1}(\mathbb{F}_q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$ so $\widehat{\psi}(v, v_{r+1}) = (vM, v_{r+1})^{\alpha}$. Then

$$(G_1|a^T)\begin{pmatrix} M & 0\\ 0 & 1 \end{pmatrix} \alpha = (G_1M|a^T)^\alpha = (G_2|b^T)$$

and $B_2 = \widehat{\psi}(B_1)$. Hence the codes B_1 and B_2 are equivalent and so they have the same canonical representative $B = \rho(B_1) = \rho(B_2)$.

The code B_1 passes the parent test and therefore the added column is in the special orbit. Since $\phi_1 \hat{\psi}^{-1}(B_2) = \phi_1(B_1) = \rho(B_1) = \rho(B_2)$, $\phi_2 = \phi_1 \hat{\psi}^{-1}$ maps B_2 to its canonical form B. Since ϕ_2 acts on the added coordinate in the same way as ϕ_1 , this coordinate is in the special orbit and therefore the code B_2 also passes the parent test.

To see what happens with the children of the same code *C*, we have to consider the automorphism group of *C* and the group $G = Mon_n(\mathbb{F}_q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$ which acts on all linear $[n,k]_q$ codes (for more details on this group see [16]). A monomial matrix *M* can be written either in the form *DP* or the form *PD*₁, where *D* and *D*₁ are diagonal matrices and *P* is a permutation matrix, $D_1 = P^{-1}DP$. The multiplication in the group $Mon_n(\mathbb{F}_q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$ is defined by $(D_1P_1\alpha_1)(D_2P_2\alpha_2) = (D_1(P_1D_2^{\alpha_1^{-1}}P_1^{-1})P_1P_2\alpha_1\alpha_2)$, where B^{α} denotes the matrix obtained by *B* after the action of the field automorphism α on its elements. Obviously, $(AB)^{\alpha} = A^{\alpha}B^{\alpha}$ and $P^{\alpha} = P$ for any permutation matrix *P*. Let see now what happens if we take different vectors $a, b \in \mathbb{F}_q^k$ and use them in the construction extending the same linear $[n,k]_q$ code *C* with a generator matrix *G_C*. We define an action of the automorphism group Aut(*C*) of the code *C* on the set of all vectors in \mathbb{F}_q^k . To any automorphism $\phi \in \operatorname{Aut}(C)$ we can correspond an invertible matrix $A_{\phi} \in \operatorname{GL}(k,q)$ such that $G' = G_C \phi = A_{\phi} G_C$, since *G'* is another generator matrix of *C*. Using this connection, we obtain a homomorphism *f* : $\operatorname{Aut}(C) \longrightarrow \operatorname{GL}(k,q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$, $f(M, \alpha) = (A_{\phi}, \alpha)$. We have

$$G_C \phi_1 \phi_2 = (A_{\phi_1} G_C) \phi_2 = (A_{\phi_1} G_C) M_2 \alpha_2 = (A_{\phi_1} G_C)^{\alpha_2} M_2^{\alpha_2} = A_{\phi_1}^{\alpha_2} G_C^{\alpha_2} M_2^{\alpha_2} = A_{\phi_1}^{\alpha_2} A_{\phi_2} G_C.$$

Hence $A_{\phi_1\phi_2} = A_{\phi_1}^{\alpha_2}A_{\phi_2}$ and so $f(\phi_1\phi_2) = f(\phi_1)f(\phi_2)$, when the operation in the group $GL(k,q) \rtimes Aut(\mathbb{F}_q)$ is $(A, \alpha) \circ (B, \beta) = (A^{\beta}B, \alpha\beta)$. Consider the action of Im(f) on the set \mathbb{F}_q^k defined by $(A, \alpha)(x) = (Ax^T)^{\alpha^{-1}}$ for every $x \in \mathbb{F}_q^k$.

Lemma 5 Let $a, b \in \mathbb{F}_q^k$. Suppose that a^T and b^T belong to the same Im(f)-orbit, where a^T denotes the transpose of a. Then the $[n+1,k]_q$ codes with generator matrices $(G_C a^T)$ and $(G_C b^T)$ are equivalent and if one of them passes the parent test, the other also passes the test. Moreover, if the codes with generator matrices $(G_C a^T)$ and $(G_C b^T)$ are equivalent and pass the parent test, the vectors a^T and b^T belong to the same Im(f)-orbit.

PROOF. Let the matrices $(G_C|a^T)$ and $(G_C|b^T)$ generate the codes C_1 and C_2 , respectively, and $b^T = (A_{\phi}a^T)^{\alpha^{-1}}$, where $\phi = (M, \alpha) \in \operatorname{Aut}(C)$. Then

$$\widehat{\phi}(G_C|b^T) = (G_C M|b^T)^{\alpha} = ((G_C M)^{\alpha}|(b^T)^{\alpha}) = (A_{\phi}G A_{\phi}a^T) = A_{\phi}(G a^T),$$

where $\widehat{\phi} = \begin{pmatrix} M & 0 \\ 0 & 1 \end{pmatrix}, \alpha \in Mon_{n+1}(\mathbb{F}_q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$. Since $A_{\phi}(G \ a^T)$ is another generator matrix of the code C_1 , both codes are equivalent. Moreover, the permutational part of $\widehat{\phi}$ fixes the last coordinate position,

hence if n+1 is in the special orbit of C_1 , it is in the special orbit of C_2 and so both codes pass (or don't pass) the parent test.

Conversely, let $C_1 \cong C_2$ and both codes pass the parent test. It turns out that there is a map $\psi = (M_{\psi}, \beta) \in G$ such that $\psi(C_1) = C_2$ and $\pi_{\psi}(n+1) = n+1$ where π_{ψ} is the permutational part of ψ . Hence $M_{\psi} = \begin{pmatrix} M_1 & 0 \\ 0 & \mu \end{pmatrix}$ and

$$(G_C|a^T)M_{\psi}\beta = (G_C M_1|\mu a^T)\beta = (G_C M_1\beta|(\mu a^T)^\beta) = A(G_C|b^T).$$

It follows that $G_C M_1 \beta = A G_C$ which means that $(M_1, \beta) \in \operatorname{Aut}(C)$, and $(\mu a^T)^{\beta} = A b^T$, so $a^T = ((\mu^{-1})^{\beta} A b^T)^{\beta^{-1}}$. Since

$$G(\mu^{-1}M_1,\beta) = (\mu^{-1}GM_1)\beta = (\mu^{-1})^{\beta}(GM_1)^{\beta} = (\mu^{-1})^{\beta}AG,$$

we have $((\mu^{-1})^{\beta}A,\beta) = f(\mu^{-1}M_1,\beta)$. Hence $(\mu^{-1}M_1,\beta) \in Aut(C)$ and a^T and b^T belong to the same orbit under the defined action.

Proof of Theorem 2:

The algorithm starts with the trivial $[k,k,1]_q$ code $C_k = \mathbb{F}_q^k$. In this case $\operatorname{Aut}(C_k) = Mon_k(\mathbb{F}_q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$ and the group partitions the set \mathbb{F}_q^k into k+1 orbits as two vectors are in the same orbit iff they have the same weight. We take exactly one representative of each orbit (instead the zero vector) and extend I_k with these column-vectors. If $d_1 = 2$, we take only the obtained $[k+1,k,2]_q$ code, otherwise we take all constructed codes and put them in the set $ch^*(C)$. All obtained codes pass the parent test.

Suppose that U_{k+i} contains inequivalent $[k+i,k, \ge d_i]_q$ codes with dual distance $\ge d^{\perp}$, $d_i = d - n + k + i$, and any code with these parameters is equivalent to a code in U_{k+i} . We will show that the set U_{k+i+1} consists only of inequivalent codes, and any linear $[k+i+1,k, \ge d_{i+1}]_q$ code is equivalent to a code in the set U_{k+i+1} .

Suppose that the codes $B_1, B_2 \in U_{k+i+1}$ are equivalent. Since these two codes have passed the parent test, their parent codes are also equivalent according to Lemma 3. These parent codes are linear codes from the set U_{k+i} which consists only in inequivalent codes. The only option for both codes is to have the same parent. But as we take only one vector of each orbit under the considered group action, we obtain only inequivalent children from one parent code (Lemma 5). Hence B_1 and B_2 cannot be equivalent.

Take now a linear $[k+i+1,k, \ge d_{i+1}]_q$ code *C* with a canonical representative *B*. If $\sigma(C)$ is the special orbit, we can reorder the coordinates of *C* such that one of the coordinates in $\sigma(C)$ to be the last one. So we obtain a code C_1 that is permutational equivalent to *C* and passes the parent test. Removing this coordinate, we obtain a parent code C_P of C_1 . Since U_{k+i} consists of all inequivalent $[k+i,k,\ge d_i]_q$ codes with dual distance $\ge d^{\perp}$, the parent code C_P is equivalent to a code $A \in U_{k+i}$. According to Lemma 4, to any child code of C_P that passes the parent test, there is a child code of *A* that also passes the test. So there is a child code C_A of *A* that passes the test, so $C_A \in U_{k+i+1}$, and C_A is equivalent to *C*. In this way we find a code in U_{k+i+1} which is equivalent to *C*.

Hence in the last step we obtain all inequivalent $[n,k, \geq d]_q$ codes with the needed dual distance.

Our goal is to get all linear $[n,k]_q$ codes with given dual distance starting from the $k \times k$ identity matrix. We can also start with all already constructed $[n' < n,k]_q$ codes to get all $[n,k]_q$ codes with the needed properties. Similar algorithms are developed in [11], [12] but these algorithms start from the empty set and generate all inequivalent codes of length $\leq n$ and dimensions 1, 2, ..., k.

C. Algorithm 2

The second algorithm is a canonical augmentation row by row. We start from the empty set (or set of already given codes with parameters $[n-i,k-i,d]_q$, $1 \le i \le k$) and aim to construct all $[n,k,\ge d]_q \ge d^{\perp}$ codes. In any step we add one row and one column to the considered generator matrix. In the *i*-th step we extend the $[n-k+i-1,i-1,\ge d]_q$ codes to $[n-k+i,i,\ge d]_q$ codes.

We consider generator matrices in the form $(A|I_k)$. If *C* is a linear $[n-k+s,s, \ge d]_q$ code with a generator matrix $(A|I_s)$, we extend the matrix to $\begin{pmatrix} A & I_s & 0^T \\ a & 0 \dots 0 & 1 \end{pmatrix} = \begin{pmatrix} A & I_{s+1} \end{pmatrix}$, where $a \in \mathbb{F}_{n-k}$. If our aim is to construct codes with dual distance $d_k^{\perp} \ge d^{\perp}$, in the *s*-th step we need codes with dual distance $d_s^{\perp} \ge d^{\perp} - (k-s)$. The obtained $[n-k+s+1,s+1,\ge d]_q$ codes with dual distance $\ge d^{\perp} - (k-s)$ are the children of *C* and the set of all such codes is denoted by Ch(C). The parent test for these codes is the same as in Algorithm II-B. We take a canonical representative for the dual code of *C* such that $\rho(C^{\perp}) = \rho(C)^{\perp}$. The orbits of *C* are ordered in the same way as the orbits of C^{\perp} and the special orbit for both codes is the same. The only difference is that if *C* is a code with zero coordinates then the orbit consisting of these coordinates coincides with the orbit of C^{\perp} consisting of the supports of the codewords with weight 1. As in the previous algorithm, we define a group action but now on the vectors in \mathbb{F}_q^{n-k} and take one representative from each orbit for the construction. The corresponding set of codes is denoted by $Ch^*(C)$. Lemma 3 and Lemma 4 hold in this case, too.

If $(A|I_k)$ is a generator matrix of *C* then $(I_{n-k}|-A^T)$ generates C^{\perp} . So in the extension in the *s*-th step the vector $\underline{-a^T}$ expands the considered generator matrix of C^{\perp} to give a generator matrix of the extended code $\overline{C^{\perp}} \in Ch(C^{\perp})$. Moreover, $\operatorname{Aut}(C^{\perp}) = \{(D^{-1}P, \alpha) | (DP, \alpha) \in \operatorname{Aut}(C)\}$. Therefore, for the action of $\operatorname{Aut}(C)$ on the vectors in \mathbb{F}_q^{n-k} , we use the elements of $\operatorname{Aut}(C^{\perp})$. If $\phi = (DP, \alpha) \in \operatorname{Aut}(C)$ then $\phi' = (D^{-1}P, \alpha) \in \operatorname{Aut}(C^{\perp})$ and so we have an invertible matrix $B_{\phi} \in \operatorname{GL}(n-k,q)$ such that $G' = (I_k| - A^T)\phi' = B_{\phi}(I_k| - A^T)$, since G' is another generator matrix of C^{\perp} . In this way we obtain a homomorphism f': $\operatorname{Aut}(C) \longrightarrow \operatorname{GL}(n-k,q) \rtimes \operatorname{Aut}(\mathbb{F}_q)$, $f(DP, \alpha) = (B_{\phi}, \alpha)$. Then we consider the action of $\operatorname{Im}(f')$ on the set \mathbb{F}_q^{n-k} defined by $(B, \alpha)(x) = (Bx^T)^{\alpha^{-1}}$ for every $x \in \mathbb{F}_q^{n-k}$. This action is similar to the action defined in Subsection II-B. The proof of the following lemma for an [n,k] code *C* with a generator matrix $(A|I_k)$ is similar to the proof of Lemma 5.

Lemma 6 Let $a, b \in \mathbb{F}_q^{n-k}$. Suppose that a and b belong to the same Im(f')-orbit. Then the $[n+1,k+1]_q$ codes with generator matrices $\begin{pmatrix} A \\ a \end{pmatrix} I_{k+1}$ and $\begin{pmatrix} A \\ b \end{pmatrix} I_{k+1}$ are equivalent and if one of them passes the

parent test, the other also passes the test. Moreover, if the codes with generator matrices $\begin{pmatrix} A \\ a \end{pmatrix} I_{k+1}$ and $\begin{pmatrix} A \\ a \end{pmatrix}$

 $\begin{pmatrix} A \\ b \end{pmatrix}$ are equivalent and pass the parent test, the vectors a and b belong to the same Im(f')-orbit.

The proof that Algorithm 2 gives the set U_n of all inequivalent $[n,k, \ge d]_q$ codes with dual distance $\ge d^{\perp}$ is similar to the proof of Theorem 2, therefore we skip it.

D. Some details

The parent test is an expensive part of the algorithms. That's way we use invariants to take information about the orbits $\{O_1, \ldots, O_m\}$ after the action of $\operatorname{Aut}(C)$ on the set of coordinate positions. An invariant of the coordinates of *C* is a function $f: N \to \mathbb{Z}$ such that if *i* and *j* are in the same orbit with respect to $\operatorname{Aut}(C)$ then f(i) = f(j), where $N = \{1, 2, \ldots, n\}$ is the set of the coordinate positions. The code *C* and the invariant *f* define a partition $\pi = \{N_1, N_2, \ldots, N_l\}$ of the coordinate set *N*, such that $N_i \cap N_j = \emptyset$ for $i \neq j$, $N = N_1 \cup N_2 \cup \cdots \cup N_l$, and two coordinates *i*, *j* are in the same subset of $N \iff f(i) = f(j)$. So the subsets N_i are unions of orbits, therefore we call them pseudo-orbits. We can use the fact that if we take two coordinates from two different subsets, for example $s \in N_i$ and $t \in N_j$, $N_i \cap N_j = \emptyset$, they belong to different orbits under the action of $\operatorname{Aut}(C)$ on the coordinate set *N*. Moreover, using an invariant *f*, we can define a new canonical representative and a new special orbit of *C* in the following way. If $f_i = f(j_i)$ for $j_i \in N_i$, $i = 1, 2, \ldots, l$, we can order the pseudo-orbits in respect to the integers f_i . We take for a canonical representative a code for which $f_1 < f_2 < \cdots < f_l$. Moreover, we order the orbits in one pseudo-orbit as it is described in Section II-A. So the orbits in the canonical representative are ordered according this new ordering. The special orbit for a code C is defined in the same way as in Section II-A (only the canonical map and the canonical representative may be different).

In the step "if B passes the parent test", using a given generator matrix of the code B we have to calculate invariants, and in some cases also canonical form and the automorphism group Aut(B). Finding a canonical form and the automorphism group is necessary when the used invariants are not enough to prove whether the code B pass or not the parent test. If the code B passes the parents test, the algorithm needs a set of generators of Aut(B) for the next step (finding the child codes). Description of some very effective invariants and the process of their applications are described in details in [13] and [14].

Similar algorithms can be used to construct linear codes with a prescribed fixed part - a residual code or a subcode.

III. EXTENDING LINEAR CODES VIA LATTICE POINT ENUMERATION

As mentioned in the introduction, we represent an $[n,k]_q$ code by a systematic generator matrix $G \in \mathbb{F}_q^{k \times n}$, i.e., *G* is of the form $G = (I_k | R)$, where I_k is the $k \times k$ unit matrix and $R \in \mathbb{F}_q^{k \times (n-k)}$. While this representation is quite compact, it nevertheless can cause serious storage requirements if the number of codes get large.

Our general strategy to enumerate linear codes is to start from a (systematic) generator matrix G of a code and to extend G to a generator matrix G' of a "larger" code. Of course, there are several choices how the shapes of the matrices G and G' can be chosen, see e.g. [5] for some variants. Here we assume the form

$$G' = \begin{pmatrix} I_k & 0 \dots 0 & R \\ 0 & \underbrace{1 \dots 1}_r & \star \end{pmatrix}$$

where $G = (I_k|R)$ and $r \ge 1$. Note that if G is a systematic generator matrix of an $[n,k]_q$ code, then G' is a systematic generator matrix of an $[n+r,k+1]_q$ code. Typically there will be several choices for the \star s and some of these can lead to equivalent codes. So, in any case we will have to face the problem that we are given a set \mathscr{C} of linear codes and we have to sift out all equivalent copies. A classical approach for this problem is to reformulate the linear code as a graph, see [13], and then to compare canonical forms of graphs using the software package Nauty [14], see also [5]. In our software we use the implementation from Q-Extension as well as another direct algorithmic approach implemented in the software CodeCan [17]. In our software, we can switch between these two tools to sift out equivalent copies and we plan to implement further variants. The reason to choose two different implementations for the same task is to independently validate results.

It remains to solve the extension problem from a given generator matrix G to all possible extension candidates G'. To this end we utilize the geometric description of the linear code generated by G as a multiset \mathcal{M} of points in PG $(k-1, \mathbb{F}_q)$, where

$$\mathscr{M} = \left\{ \left\{ \langle g^i \rangle : 1 \le i \le n \right\} \right\},^1$$

 g^i are the *n* columns of *G*, and $\langle v \rangle$ denotes the row span of a column vector *v*. In general, the 1-dimensional subspaces of \mathbb{F}_q^k are the points of $PG(k-1,\mathbb{F}_q)$. The (k-1)-dimensional subspaces of \mathbb{F}_q^k are called the hyperplanes of $PG(k-1,\mathbb{F}_q)$. By m(P) we denote the multiplicity of a point $P \in \mathcal{M}$. We also say that a column g^i of the generator matrix has multiplicity m(P), where $P = \langle g^i \rangle$ is the corresponding point, noting that the counted columns can differ by a scalar factor. Similarly, let \mathcal{M}' denote the multiset of points in $PG((k+1)-1,\mathbb{F}_q)$ that corresponds to the code generated by the generator matrix G'. Note that our notion of equivalent linear codes goes in line with the notion of equivalent multisets of points in projective spaces, see [1]. Counting column multiplicities indeed partially takes away the inherent symmetry of the generator matrix of a linear code, i.e., the ordering of the columns and multiplications of columns with

¹We use the notation $\{\{\cdot\}\}$ to emphasize that we are dealing with multisets and not ordinary sets. A more precise way to deal with a multiset \mathscr{M} in $PG(k-1,\mathbb{F}_q)$ is to use a characteristic function χ which maps each point *P* of $PG(k-1,\mathbb{F}_q)$ to an integer, which is the number of occurrences of *P* in \mathscr{M} . With this, the cardinality $\#\mathscr{M}$ can be written as the sum over m(P) for all points *P* of $PG(k-1,\mathbb{F}_q)$.

non-zero field elements is not specified explicitly any more. If the column multiplicity of every column is exactly one, then the code is called projective.

Our aim is to reformulate the extension problem $G \to G'$ as an enumeration problem of integral points in a polyhedron. Let $W \subseteq \{i\Delta : a \le i \le b\} \subseteq \mathbb{N}_{\ge 1}$ be a set of feasible weights for the non-zero codewords, where we assume $1 \le a \le b$ and $\Delta \ge 1.^2$ Linear codes where all weights of the codewords are divisible by Δ are called Δ -divisible and introduced by Ward, see e.g. [18], [19].

The non-zero codewords of the code generated by the generator matrix *G* correspond to the non-trivial linear combinations of the rows of *G* (over \mathbb{F}_q). In the geometric setting, i.e., where an $[n,k]_q$ code *C* is represented by a multiset \mathcal{M} , each non-zero codeword $c \in C$ corresponds to a hyperplane *H* of the projective space $PG(k-1,\mathbb{F}_q)$. (More precisely, $\mathbb{F}_q^* \cdot c$ is in bijection to *H*, where $\mathbb{F}_q^* = \mathbb{F}_q \setminus \{0\}$.) With this, the Hamming weight of a codeword *c* is given by

$$n - \sum_{P \in \mathrm{PG}(k-1,\mathbb{F}_q): P \in \mathcal{M}, P \leq H} m(P),$$

see [1]. By \mathscr{P}_k we denote the set of points of $PG(k-1,\mathbb{F}_q)$ and by \mathscr{H}_k the set of hyperplanes.

Lemma 7 Let G be a systematic generator matrix of an $[n,k]_q$ code C whose non-zero weights are contained in $\{i\Delta : a \le i \le b\} \subseteq \mathbb{N}_{\ge 1}$. By c(P) we denote the number of columns of G whose row span equals P for all points P of $PG(k-1,\mathbb{F}_q)$ and set $c(\mathbf{0}) = r$ for some integer $r \ge 1$. With this let $\mathscr{S}(G)$ be the set of feasible solutions of

$$\Delta y_H + \sum_{P \in \mathscr{P}_{k+1} : P \le H} x_P = n - a\Delta \qquad \forall H \in \mathscr{H}_{k+1}$$
(1)

$$\sum_{q \in \mathbb{F}_q} x_{\langle (u|q) \rangle} = c(\langle u \rangle) \qquad \forall \langle u \rangle \in \mathscr{P}_k \cup \{\mathbf{0}\}$$
⁽²⁾

$$x_{\langle e_i \rangle} \ge 1 \qquad \forall 1 \le i \le k+1 \tag{3}$$

$$x_P \in \mathbb{N} \quad \forall P \in \mathscr{P}_{k+1}$$
 (4)

$$y_H \in \{0, \dots, b-a\} \qquad \forall H \in \mathscr{H}_{k+1},\tag{5}$$

where e_i denotes the ith unit vector in \mathbb{F}_q^{k+1} . Then, for every systematic generator matrix G' of an $[n+r,k+1]_q$ code C' whose first k rows coincide with G and whose weights of its non-zero codewords are contained in $\{i\Delta : a \leq i \leq b\}$, we have a solution $(x,y) \in \mathscr{S}(G)$ such that G' has exactly x_P columns whose row span is equal to P for each $P \in \mathscr{P}_{k+1}$.

PROOF. Let such a systematic generator matrix G' be given and x_P denote the number of columns of G' whose row span is equal to P for all points $P \in \mathscr{P}_{k+1}$. Since G' is systematic, Equation (3) is satisfied. As G' arises by appending a row to G, also Equation (2) is satisfied for all $P \in \mathscr{P}_k$. For $P = \mathbf{0}$ Equation (2) is just the specification of r. Obviously, the x_P are non-negative integers. The conditions (1) and (5) correspond to the restriction that the weights are contained in $\{i\Delta : a \le i \le b\}$.

We remark that some of the constraints (1) are automatically satisfied since the subcode C of C' satisfies all constraints on the weights. If there are further forbidden weights in $\{i\Delta : a \le i \le b\}$ then, one may also use the approach of Lemma 7, but has to filter out the integer solutions that correspond to codes with forbidden weights. Another application of this first generate, then filter strategy is to remove some of the constraints (1), which speeds up, at least some, lattice point enumeration algorithms. In our implementation we use Solvediophant [20], which is based on the LLL algorithm [21], to enumerate the integral points of the polyhedron from Lemma 7.

Noting that each $[n',k',W]_q$ code, where $W \subseteq \mathbb{N}$ is a set of weights, can indeed be obtained by extending³

²Choosing $\Delta = 1$ such a representation is always possible. Moreover, in many applications we can choose $\Delta > 1$ quite naturally. I.e., for optimal binary linear $[n,k,d]_2$ codes with even minimum distance d, i.e., those with maximum possible d, we can always assume that there exists an *even* code, i.e., a code where all weights are divisible by 2.

³This operation is also called *lengthening* in the coding theoretic literature, i.e., both the effective length *n* and the dimension *k* is increased, while one usually assumes that the redundancy n-k remains fix. The reverse operation is called *shortening*.

all possible $[n'-r,k'-1,W]_q$ codes via Lemma 7, where $1 \le r \le n'-k'+1$, already gives an algorithm for enumerating and classifying $[n',k',W]_q$ codes. (For k'=1 there exists a unique code for each weight $w \in W$, which admits a generator matrix consisting of w ones.) However, the number of codes C with generator matrix G that yield the same $[n',k',W]_q$ code C' with generator matrix G' can grow exponentially with k'. We can limit this growth a bit by studying the effect of the extension operation and its reverse on some code invariants.

Lemma 8 Let C' be an $[n',k',W]_q$ code with generator matrix G'. If G' contains a column g' of multiplicity $r \ge 1$, then there exists a generator matrix G of an $[n'-r,k'-1,W]_q$ code C such that the extension of G via Lemma 7 yields at least one code that is equivalent to C'. Moreover, if Λ is the maximum column multiplicity of G', without counting the columns whose row span equals $\langle g' \rangle$, then the maximum column multiplicity of G is at least Λ .

PROOF. Consider a transform \hat{G} of G' such that the column g' of G' is turned into the *j*th unit vector e_j for some integer $1 \le j \le k'$. Of course also \tilde{G} is a generator matrix of C'. Now let \hat{G} be the $(k'-1) \times (n'-r)$ matrix over \mathbb{F}_q that arises from \tilde{G} after removing the *r* occurrences of the columns with row span $\langle e_j \rangle$ and additionally removing the *j*th row. Note that the non-zero weights of the linear code generated by \hat{G} are also contained in *W*. If *G* is a systematic generator matrix of the the linear code *C* generated by \hat{G} , then Lemma 7 applied to *G* with the chosen parameter *r* yields especially a linear code with generator matrix *G'* as a solution. By construction the effective length of *C* is indeed n' - r. Finally, note that removing a row from a generator matrix does not decrease column multiplicities.

Corollary 9 Let C' be an $[n',k',W]_q$ code with generator matrix G' and minimum column multiplicity r. Then there exists a generator matrix G of an $[n'-r,k'-1,W]_q$ code C with minimum column multiplicity at least r such that the extension of G via Lemma 7 yields at least one code that is equivalent C'.

Corollary 9 has multiple algorithmic implications. If we want to classify all $[n,k,W]_q$ codes, then we need the complete lists of $[\leq n-1, k-1, W]_q$ codes, where $[\leq n', k', W'_q]$ codes are those with an effective length of at most n'. Given an $[n', k-1, W]_q$ code with $n' \leq n-1$ we only need to extend those codes which have a minimum column multiplicity of at least n-n' via Lemma 7. If n-n' > 1 this usually reduces the list of codes, where an extensions needs to be computed. Once the set $\mathscr{S}(G)$ of feasible solutions is given, we can also sift out some solutions before applying the equivalence sifting step. Corollary 9 allows us to ignore all resulting codes which have a minimum column multiplicity strictly smaller than n-n'. Note that when we know $x_P > 0$, which we do know e.g. for $P = \langle e_i \rangle$, where $1 \leq i \leq k+1$, then we can add the valid inequality $x_P \geq n-n'$ to the inequality system from Lemma 7. We call the application of the extension step of Lemma 7 under these extra assumptions *canonical length extension* or *canonical lengthening*.

As an example we consider the $[7,2]_2$ code that arises from two codewords of Hamming weight 4 whose support intersect in cardinality 1, i.e., their sum has Hamming weight 6. A direct construction gives the generator matrix

$$G_1 = \begin{pmatrix} 1 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{pmatrix}$$

which can be transformed into

 $G_2 = \begin{pmatrix} 1 & 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 1 & 1 & 1 \end{pmatrix}.$

Now column permutations are necessary to obtain a systematic generator matrix

$$G_3 = \begin{pmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix}.$$

Note that G_2 and G_3 do not generate the same but only equivalent codes. Using the canonical length extension the systematic generator matrix

$$G_0 = \begin{pmatrix} 1 & 1 & 1 & 1 \end{pmatrix}$$

of a single codeword of Hamming weight 4 cannot be extended to G_3 , since we would need to choose r = 3 to get from a $[4,1]_2$ code to a $[7,2]_2$ code, while the latter code has a minimum column multiplicity of 1. However, the unique codeword with Hamming weight 6 and systematic generator matrix

$$G = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix}$$

can be extended to

$$G_4 = \begin{pmatrix} 1 & 0 & 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 & 1 & 1 & 1 \end{pmatrix},$$

which generates the same code as G_3 . So, we needed to consider an extension of a $[6,1]_2$ code to a $[7,2]_2$ code. Now let us dive into the details of the integer linear programming formulation of Lemma 7. In our example we have k = 1 and q = 2, so that $\mathscr{P}_1 = \{\langle (1) \rangle\}$, and

$$\mathscr{P}_2 = \left\{ \left\langle \begin{pmatrix} 1\\0 \end{pmatrix} \right\rangle, \left\langle \begin{pmatrix} 0\\1 \end{pmatrix} \right\rangle, \left\langle \begin{pmatrix} 1\\1 \end{pmatrix} \right\rangle \right\}.$$

The multiplicities corresponding to the columns of G and r are given by

$$c(\langle (1) \rangle) = 6$$
 and $c(\langle (0) \rangle) = 1$.

Due to constraint (2) we have

$$x_{\langle e_1 \rangle} + x_{\langle e_1 + e_2 \rangle} = 6$$
 and $x_{\langle e_2 \rangle} = 1$.

Constraint (3) reads

$$x_{\langle e_1 \rangle} \geq 1$$
 and $x_{\langle e_2 \rangle} \geq 1$.

In order to write down constraint (1), we need to specify the set W of allowed weights. Let us choose $W = \{4, 6\}$, i.e., $\Delta = 2$, a = 2, and b = 3. If we label the hyperplanes by $\mathcal{H} = \{1, 2, 3\}$, for the ease of notation, we obtain

$$\begin{array}{rcl} 2y_1 + x_{\langle e_2 \rangle} &=& 3, \\ 2y_2 + x_{\langle e_1 + e_2 \rangle} &=& 3, \text{ and} \\ 2y_3 + x_{\langle e_1 \rangle} &=& 3. \end{array}$$

Since the y_i are in $\{0,1\}$ we have $x_{\langle e_1 \rangle} \leq 3$ and $x_{\langle e_1+e_2 \rangle} \leq 3$, so that $x_{\langle e_1 \rangle} = 3$ and $x_{\langle e_1+e_2 \rangle} = 3$. The remaining variables are given by $x_{\langle e_2 \rangle} = 1$, $y_1 = 1$, $y_2 = 0$, and y_3 . Thus, in our example there is only one unique solution, which then corresponds to generator matrix G_4 (without specifying the exact ordering of the columns of G_4).

Note that for the special situation k+1=2, every hyperplane of \mathscr{P}_2 consists of a unique point. The set of column or point multiplicities is left invariant by every isometry of a linear code. For hyperplanes in $PG(k+1,\mathbb{F}_q)$ or non-zero codewords of C' a similar statement applies. To this end we introduce the weight enumerator $w_C(x) = \sum_{i=0}^n A_i x^i$ of a linear code C, where A_i counts the number of codewords of Hamming weight exactly i in C. Of course, the weight enumerator $w_C(x)$ of a linear code C does not depend on the chosen generator matrix C. The geometric reformulation uses the number a_i of hyperplanes $H \in \mathscr{H}_k$ with $\#H \cap \mathscr{M} := \sum_{P \in \mathscr{P}_k: P \in \mathscr{M}, P \leq H} m(P) = i$. The counting vector (a_0, \ldots, a_n) is left unchanged by isometries. One application of the weight enumerator in our context arises when we want to sift out equivalent copies from a list \mathscr{C} of linear codes. Clearly, two codes whose weight enumerators do not coincide, cannot be equivalent. So, we can first split \mathscr{C} according to the occurring different weight enumerators and then apply one of the mentioned algorithms for the equivalence filtering on the smaller parts separately. We can even refine this invariant a bit more. For a given $[n,k]_q$ code *C* with generator matrix *G* and corresponding multiset \mathscr{M} let $\widetilde{\mathscr{M}}$ be the set of different elements in \mathscr{M} , i.e., $\#\mathscr{M} = \sum_{P \in \mathscr{M}} m(P)$, which means that we ignore the multiplicities in $\widetilde{\mathscr{M}}$. With this we can refine Lemma 8:

Lemma 10 Let C be an $[n,k,W]_q$ code with generator matrix G and \mathcal{M} , $\widetilde{\mathcal{M}}$ as defined above. For each $P \in \widetilde{\mathcal{M}}$ there exists a generator matrix G_P of an $[n-m(P),k-1]_q$ code such that the extension of G_P via Lemma 7 yields at least one code that is equivalent to C.

Now we can use the possibly different weight enumerators of the subcodes generated by G_P to distinguish some of the extension paths.

Corollary 11 Let C' be an $[n',k',W]_q$ code with generator matrix G', minimum column multiplicity r, and \mathcal{M} , $\widetilde{\mathcal{M}}$ as defined above. Then there exists a generator matrix G of an $[n'-r,k'-1,W]_q$ code C such that the extension of G via Lemma 7 yields at least one code that is equivalent to C' and the weight enumerator $w_C(x)$ is lexicographically minimal among the weight enumerators $w_{C_P}(x)$ for all $P \in \widetilde{\mathcal{M}}$ with column multiplicity r in C', where C_P is the linear code generated by the generator matrix G_P from Lemma 10.

We remark that the construction for subcodes, as described in Lemma 10, can also be applied for points $P \in \mathscr{P}_k \setminus \mathscr{M}$. And indeed, we obtain an $[n - m(P), k - 1]_q = [n, k - 1]_q$ code, i.e., the effective length does not decrease, while the dimension decreases by one.

The algorithmic implication of Corollary 11 is the following. Assume that we want to extend an $[n,k,W]_q$ code C with generator matrix G to an $[n+r,k+1,W]_q$ code C' with generator matrix G'. If the minimum column multiplicity of C is strictly smaller than r, then we do not need to compute any extension at all. Otherwise, we compute the set $\mathscr{S}(G)$ of solutions according to Lemma 7. If a code C' with generator matrix G', corresponding to a solution in $\mathscr{S}(G)$, has a minimum column multiplicity which does not equal r, then we can skip this specific solution. For all other candidates let $\overline{\mathscr{M}} \subseteq \mathscr{P}_{k+1}$ the set of all different points spanned by the columns of G' that have multiplicity exactly r. By our previous assumption $\overline{\mathscr{M}}$ is not the empty set. If $w_C(x)$ is the lexicographically minimal weight enumerator among all weight enumerators $w_{C_P}(x)$, where $P \in \overline{\mathscr{M}}$ and C_P is generated by the generator matrix G_P from Lemma 10, then we store C' and skip it otherwise. We call the application of the extension step of Lemma 7 under these extra assumptions *lexicographical extension* or *lexicographical lengthening*.

Lexicographical lengthening drastically decrease the ratio between the candidates of linear codes that have to be sifted out and the resulting number of non-equivalent codes. This approach also allows parallelization of our enumeration algorithm, i.e., given an exhaustive list \mathscr{C} of all $[n,k,W]_q$ codes and an integer $r \ge 1$, we can split \mathscr{C} into subsets $\mathscr{C}_1, \ldots, \mathscr{C}_l$ according to their weight enumerators. If the $[n+r,k+1,W]_q$ code C' arises by lexicographical lengthening from a code in \mathscr{C}_i and the $[n+r,k+1,W]_q$ code C'' arises by lexicographical lengthening from a code in \mathscr{C}_j , where $i \ne j$, then C' and C'' cannot be equivalent. As an example, when constructing the even $[21,8,6]_2$ codes from the 17927353 $[20,7,6]_2$ codes, we can split the construction into more than 1000 parallel jobs. If we do not need the resulting list of 1656768624 linear codes for any further computations, there is no need to store the complete list of codes during the computation.

IV. NUMERICAL RESULTS

We implemented the presented algorithms in the programs Generation and LinCode. These algorithms can be used to classify linear codes with wide-range parameters, for example, for binary codes with lengths up to 100 or more depending on the dimension. The main objectives in this section are three: (1) to show what problems in Coding Theory can be attacked with the software presented, (2) to solve given classification problems, and (3) to show what these results can be useful for (Subsection IV-B). The presented classification results are of three types: (1) binary codes with prescribed minimum distance; (2) divisible codes over fields with 2, 3 and 4 elements; (3) self-orthogonal codes.

As the implementation of a practically efficient algorithm for the classification of linear codes is a delicate issue, we exemplarily verify several classification results from the literature. Efficiency is demonstrated by partially extending some of these enumeration results.

A. Results.

In [6, Research Problem 7.2] the authors ask for the classification of $[n,k,3]_2$ codes for n > 14. In Table I we extend their Table 7.7 to $n \le 18$.

n/k	2	3	4	5	6	7	8	9	10
5	1								
6	3	1							
7	4	4	1						
8	6	10	5						
9	8	23	23	5					
10	10	42	76	41	4				
11	12	71	207	227	60	3			
12	15	115	509	1012	636	86	2		
13	17	174	1127	3813	4932	1705	110	1	
14	20	255	2340	12836	31559	24998	4467	127	1
15	23	364	4606	39750	176582	293871	132914	11507	143
16	26	505	8685	115281	896316	2955644	3048590	733778	28947
17	29	686	15797	317464	4226887	26590999	58085499	34053980	4115973
18	33	919	27907	837697	18807438	220135857	971007974	1261661451	393087258
n/k	11	12	13						
15	1								
16	144								
17	70455	129							
18	27333440	293458	226						
					TABL	E I			

The number of inequivalent $[n,k,3]_2$ codes for $n \le 18$

Blank entries correspond to the non-existence of any code with these parameters, i.e., there is no $[4, 2, 3]_2$ code and also no $[16, 12, 3]_2$ code. Obviously, there is a unique $[n, 1, 3]_2$ codes for each $n \ge 3$ and it is not too hard to show that the number of inequivalent $[n, 2, 3]_2$ codes is given by $\left[\sqrt{\frac{(n-4)(n-3)(2n-7)}{6}}\right]$ for each $n \ge 3$.

k	4	5	6	7	8	9	10	11	12	13		
#	8561	129586	1813958	16021319	60803805	73340021	22198835	1314705	11341	24		
	TABLE II											
	The number of inequivalent even $[n \leq 19, k, 4]_2$ codes for $4 \leq k \leq 13$											

There are possibilities for different restrictions for the codes in addition to the restrictions on length, dimension, minimum and dual distances. We apply also restrictions on the orthogonality and weights of the codewords in some examples.

We present the counts for the even $[\leq 19, k, 4]_2$ codes in Table II. The numbers of the inequivalent even $[n \leq 21, k, 6]_2$ codes are presented in Table III (excluding the enumeration of the even $[21, 9, 6]_2$ codes because their number is extremely huge). We have verified these results by both software programs.

Moreover, we have enumerated the divisible codes with given parameters over fields with 2, 3 and 4 elements. Recently, 8-divisible binary codes (called also triply even) have been investigated [22], [23]. In [24], it is proven that projective triply-even binary codes exist precisely for lengths 15, 16, 30, 31, 32,

k	3	4	5	6	7	8	10	11
#	726	12817	358997	11697757	246537467	1697180017	62180809	738

TARIE	ш
IADLU	ш

The number of even $[\leq 21, k, 6]_2$ codes for $3 \leq k \leq 11, k \neq 9$

$n \setminus k$	2	3	4	5	6
	-	5		5	
12	1				
13		1			
18	1				
21	1	1			
22		1	1		
24	1	1	1		
25		1	1	1	
26		1	1	1	1
27	2	3	3	1	
30	2	4	3		
31		2	3	1	
33	1	5	5	3	
34		2	5	4	1
35		1	4	4	3
36	4	10	22	13	4
37		2	7	10	3
38		1	6	12	10
39	3	15	34	41	23
40		6	25	40	30
41	0	0	0	0	0
42	2	17	52	44	15
43		6	32	40	16
44		2	14	22	17
45	5	31	141	190	72
46		6	56	122	71
47		2	29	92	89
48	5	44	297	705	468
49		15	177	613	596
50		2	39	217	295
total	28	182	958	2176	1714
		TA	BLE IV	7	

DIVISIBLE TERNARY CODES WITH $n \le 50, k \le 6, \Delta = 9$

45-51, and ≥ 60 . We have verified the enumeration of the projective 2, 4-, and 8-divisible binary linear codes from [24].

We have classified 9-divisible ternary codes and 4-divisible quaternary codes.

- q = 3, $\Delta = 9$. Table IV contains classification results for codes of this type with length $n \le 55$ and dimension $k \le 8$. The conspicuous zero row for length n = 41 has a theoretical explanation, i.e., there is no 9-divisible $[41,k]_3$ code at all, see [25, Theorem 1].⁴
- q = 4, $\Delta = 4$. Table V presents classification results for codes with $n \le 30$ and $k \le 8$. All constructed codes are Hermitian self-orthogonal.

By the program Generation, we have also classified binary, ternary and quaternary self-orthogonal codes. There are a few tables of self-orthogonal codes (see [26], [27]). Here we present classification results that are not given in these tables, namely:

• We present classification results for binary self-orthogonal $[27, k \le 12, d \ge 8]$ codes with dual distance $d^{\perp} \ge 1$ in Table VI. The codes with dimensions 11 and 12 are optimal as linear codes, and the codes

⁴More precisely, $41 = 2 \cdot 13 + 2 \cdot 12 - 1 \cdot 9$ is a certificate for the fact that such a code does not exist, see [25, Theorem 1, Example 6].

$n \setminus k$	2	3	4	5	6	7	8					
5	1											
8	1											
9	1	1										
10	1	1	1									
12	2	2										
13	2	3	1									
14	1	5	3	1								
15	1	3	6	2	1							
16	4	9	7	2								
17	3	12	9	2								
18	2	18	25	8	1							
19	1	14	42	25	6	1						
20	6	34	93	70	22	4	1					
21	5	45	115	75	19	2						
22	3	64	245	131	23	2						
23	2	62	554	398	96	12	1					
24	9	123	1509	1769	491	79	9					
25		168	3189	6890	1842	334	46					
26			8420	18377	2691	360	33					
27				70147	4602	458	34					
28					36982	3075	244					
29						34180	2366					
30							24565					
total	45	564	14219	97897	46776	38507	27299					

TABLE V DIVISIBLE QUATERNARY CODES WITH $n \le 30, k \le 8, \Delta = 4$

with k = 9 and 10 are optimal only as self-orthogonal [26]. Moreover, we tried to fill some of the gaps in [26, Table 1]. We classified the *n*-optimal self-orthogonal [n,k,d] codes (the codes for which no [n-1,k,d] self-orthogonal code exists) with parameters [35,8,14], [29,9,10] and [30,10,10]. The number of codes in these cases are 376, 36504 and 573, respectively. Our program shows that no self-orthogonal [37,10,14] and [36,9,14] codes exist which means that the maximal possible minimum distance for self-orthogonal codes with these lengths and dimensions is 12.

k	2	3	4	5	6	7				
total	59	445	4615	64715	959533	8514764				
k	8	9	10	11	12					
total	21256761	7030920	159814	791	18					
TABLE VI										

Binary self-orthogonal $[27, k \le 12, d \ge 8]d^{\perp} \ge 1$ codes

- The classification results for ternary self-orthogonal [n ≤ 20, k ≤ 10, d ≥ 6] codes are given in Table VII. This table supplements [27, Table 1].
- Table VIII shows the classification of the $[n \le 21, \le 6, 12]$ quaternary Hermitian self-orthogonal codes. These results fill some of the gaps in [27, Table 2].

B. Applications

In this subsection we want to exemplarily show up, that exhaustive enumeration results of linear codes can of course be used to obtain results for special subclasses of codes and their properties by simply checking all codes. For our first example we remark that the support of a codeword is the set of its non-zero codeword c of a linear code C is called minimal if the support of no other

$\begin{array}{c c c c c c c c c c c c c c c c c c c $								
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$n \setminus k$	4	5	6	7	8	9	10
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	10	1						
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	11	1	1					
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	12	6	2	1				
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	13	10	4	1				
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	14	27	15	4				
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	15	78	73	20	2			
17 414 1466 885 86 2 18 1097 8103 10808 1401 40 19 2589 47015 167786 45950 1132 10 20 6484 285428 2851808 2121360 89670 464 6	16	181	312	121	11	1		
18 1097 8103 10808 1401 40 19 2589 47015 167786 45950 1132 10 20 6484 285428 2851808 2121360 89670 464 6	17	414	1466	885	86	2		
19 2589 47015 167786 45950 1132 10 20 6484 285428 2851808 2121360 89670 464 6	18	1097	8103	10808	1401	40		
20 6484 285428 2851808 2121360 89670 464 6	19	2589	47015	167786	45950	1132	10	
	20	6484	285428	2851808	2121360	89670	464	6

TABLE VII

Ternary self-orthogonal codes with $n \leq 20, k \leq 10$, and $d \geq 6$

$n \setminus k$	2	3	4	5	6
15	1				
16	2	1			
17	3	4	1		
18		45	12		
19			5673		
20				886576	
21					577008

TABLE VIII QUATERNARY HERMITIAN SELF-ORTHOGONAL CODES WITH $n \leq 21, \, k \leq 6, \, d=12$

non-zero codeword is contained in the support of *c*, see e.g. [28]. By $m_2(n,k)$ we denote the minimum number of minimal codewords of a projective⁵ $[n,k]_2$ code. In Table IX we state the exact values of $m_2(n,k)$ for all $2 \le k \le n \le 15$ obtained by enumerating all projective codes with these parameters.

n/k	2	3	4	5	6	7	8	9	10	11	12	13	14	15
3	3	3												
4		4	4											
5		6	5	5										
6		7	6	6	6									
7		7	8	7	7	7								
8			8	9	8	8	8							
9			12	9	9	9	9	9						
10			14	10	10	10	10	10	10					
11			14	15	11	11	11	11	11	11				
12			15	15	13	12	12	12	12	12	12			
13			15	16	14	13	13	13	13	13	13	13		
14			15	16	14	15	14	14	14	14	14	14	14	
15			15	16	17	15	16	15	15	15	15	15	15	15
						Т	ABLE	IX						

 $m_2(n,k)$ for $3 \le n \le 15, 1 \le k \le 9$

⁵Duplicating columns in a binary linear code generated by the $k \times k$ unit matrix results in exactly k minimal codewords, which is the minimum for all k-dimensional codes.

C. Verification and computational time.

We use three basic approaches to verify our programs and the results. The first one is verification by replication. We ran the programs to get already published classification results as the classification of the projective 2, 4-, and 8-divisible binary linear codes from [24], and binary projective codes with dimension 6 [9], and we obtained the same number of codes.

The second approach is to double check most of the results with the presented programs. The third one is to use enumeration of different types of codes given by theoretical methods (see [3], [16]). For self-orthogonal codes we can also use mass formulae to verify that the constructed codes represents all equivalence classes of the given length [16].

All calculations with the program Generation have been done on $2 \times$ Intel Xeon E5-2620 V4, 32 thread processor. The calculation time strictly depends on the parameters and restrictions of the considered codes that determine the size of the corresponding search tree. For example, the proof that the extended binary Golay code is unique took 0.09s on a single tread. The classification of the binary [32, 16, 8] self-dual codes took 184.49s also on a single tread.

As already mentioned, the presented algorithms are suitable for parallel implementation. The results in Table I were obtained at once in about 3 days on 32 threads. The calculations for Table III took about the same time on the same server. All other calculations took from a few minutes to several days on a single core.

For the program LinCode the most time expensive step that was executed a single computer, i.e., extending the $[19,7,6]_2$ codes to $[20,8,6]_2$ codes, took roughly 250 hours of computation time on a single core of a Intel Core i7-7600U laptop with 2.80GHz. We remark that the $[19,k,4]_2$ codes, where $k \in \{7,8,9,10\}$, and the $[21,k,6]_2$ codes, where $k \in \{7,8,10\}$, were enumerated in parallel, i.e., we have partially used the computing nodes of the *High Performance Computing Keylab* from the University of Bayreuth. We have used the oldest cluster btrzx5 that went into operation in 2009.⁶

V. CONCLUSION

The technique of canonical augmentation is used for classification of special types of codes and related combinatorial objects in [9]–[12], etc. We apply this technique to classify linear codes with given properties in the program Generation. Our algorithm expands the matrices column by column but starts from the identity $k \times k$ matrix. So it constructs all inequivalent linear $[n,k]_q$ codes without getting codes of smaller dimensions. Restrictions on the dual distance, minimum distance, etc. can be applied. The algorithm is implemented in the program Generation, which is the first module of the software package QextNewEdition. On the one hand, this program gives us the possibility to classify linear codes with given parameters over fields with q elements. On the other hand, the program can give families of inequivalent codes with certain properties that can be used for extension in length and dimension from the other modules in the package. These modules are also based on the idea of canonical augmentation, which gives the possibility for parallelization.

Moreover, we have presented an algorithm for the classification of linear codes over finite fields based on lattice point enumeration. The lattice point enumeration itself and sifting out equivalent copies is so far done with available scientific software packages. Using invariants like the weight enumerator of subcodes, see Corollary 11, the number of candidates before sifting could kept reasonably small. The advantage of the canonical augmentation that no pairs of codes have to be checked whether they are equivalent comes at the cost that the computation of the canonical form is relatively costly. Allowing not only a single canonical extension, but a relatively small number of extensions that may lead to equivalent codes, might be a practically efficient alternative. We have also demonstrated that the algorithm can be run in parallel. However, we think that our implementation can still be further improved. In some cases the used lattice point enumeration algorithm Solvediophant takes quite long to verify that a certain code does not allow an extension, while integer linear programming solvers like e.g. Cplex quickly verify infeasibility. We propose the extension of Table IV as a specific open problem. We have demonstrated that it is indeed possible to exhaustively classify sets of linear codes of magnitude 10⁹, which was not foreseeable at the time of [6].

Currently the implementation of the evolving software package LinCode is not advanced enough to be made publicly available. So, we would like to ask the readers to sent their interesting enumeration problems of linear codes directly to the third author. The software package QextNewEdition is available on the web-page

http://www.moi.math.bas.bg/moiuser/~data/Software/QextNewEdition

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