Absdet-Pseudo-Codewords and Perm-Pseudo-Codewords: Definitions and Properties

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Abstract—The linear-programming decoding performance of a binary linear code crucially depends on the structure of the fundamental cone of the parity-check matrix that describes the code. Towards a better understanding of fundamental cones and the vectors therein, we introduce the notion of absdet-pseudo-codewords and perm-pseudo-codewords: we give the definitions, we discuss some simple examples, and we list some of their properties.

Index Terms—Absdet-pseudo-codeword, fundamental cone, low-density parity-check code, message-passing iterative decoding, perm-pseudo-codeword, pseudo-codeword, Tanner graph.

I. INTRODUCTION

In [1], MacKay and Davey discussed a simple technique for upper bounding the minimum Hamming distance of a binary linear code that is described by a parity-check matrix. Their technique was based on explicitly constructing codewords and using the fact that the Hamming weight of a non-zero codeword is an upper bound on the minimum Hamming distance of the code. This approach was subsequently extended and refined in the papers [2] and [3]. (Note that [1]–[3] focused mostly on quasi-cyclic binary linear codes, however, the technique is more generally applicable since any binary linear code of length $n$ can trivially be considered to be a quasi-cyclic code with period $n$.)

In the technique by MacKay and Davey, the constructed codewords are binary vectors whose entries stem from certain determinants that are computed over the binary field. One wonders what happens if these determinants are not computed over the binary field but over the ring of integers. Do the resulting integer vectors still say something useful about the code under investigation? In this paper we answer this question affirmatively by showing that the resulting vectors (after replacing the components by their absolute value) are pseudo-codewords, i.e., vectors that lie in the fundamental cone of the parity-check matrix of the code. These pseudo-codewords, in the following called absdet-pseudo-codewords, are therefore important in the characterization of the performance of linear programming decoding [4], [5] and message-passing iterative decoding [6], [7].

The remainder of the paper is structured as follows. In Section [II] we formally define the class of absdet-pseudo-codewords and a closely related class of pseudo-codewords, so-called perm-pseudo-codewords. In order to get some initial understanding of these pseudo-codewords, in Section [IV] we construct them for some small codes. Afterwards, in Section [V] we discuss properties of these pseudo-codewords. We conclude the paper in Section [VI].

II. BASIC NOTATIONS AND DEFINITIONS

Let $\mathbb{Z}$, $\mathbb{R}$, and $\mathbb{F}_2$ be the ring of integers, the field of real numbers, and the finite field of size 2, respectively. If $a$ is some vector with integer entries, then $a \pmod{2}$ will denote an equally long vector whose entries are reduced modulo 2. Rows and columns of matrices and entries of vectors will be indexed starting at 0. If $M$ is some matrix and if $R$ and $S$ are subsets of the row and column index sets, respectively, then $M_{R,S}$ is the sub-matrix of $M$ that contains only the rows of $M$ whose index appears in the set $R$ and only the columns of $M$ whose index appears in the set $S$. If $R$ equals the set of all row indices of $M$, we will simply write $M_S$ instead of $M_{R,S}$. Moreover, we will use the short-hand $S \setminus i$ for $S \setminus \{i\}$.

Definition 1 Let $M = (m_{ij})_{i,j}$ be an $n \times n$-matrix over some ring. Its determinant is defined to be

$$\det(M) = \sum_{\sigma} \sgn(\sigma) \prod_{j=0}^{n-1} m_{j,\sigma(j)},$$

where the summation is over all $n!$ permutations of the set $\{0, 1, \ldots, n-1\}$, and where $\sgn(\sigma)$ equals $+1$ if $\sigma$ is an even permutation and equals $-1$ if $\sigma$ is an odd permutation. Similarly, the permanent of $M$ is defined to be

$$\perm(M) = \sum_{\sigma} \prod_{j=0}^{n-1} m_{j,\sigma(j)}.$$

Clearly, for any matrix $M$ with elements from a ring or field of characteristic 2 it holds that $\det(M) = \perm(M)$.

When we want to emphasize that the matrix $M$, of which we are computing the determinant or the permanent, is to be considered to be a matrix over the ring of integers, then we will write $\det_\mathbb{Z}(M)$ and $\perm_\mathbb{Z}(M)$, respectively. Note that $\det_\mathbb{Z}(M) \pmod{2} = \perm_\mathbb{Z}(M) \pmod{2}$. $\square$
Let $H = (h_{j,i})_{j,i}$ be a parity-check matrix of some binary linear code. We define the sets $J(H)$ and $I(H)$ to be the set of row and column indices of $H$. Moreover, we will use the sets $J_i(H) \triangleq \{ j \in J \mid h_{j,i} = 1 \}$ and $I_j(H) \triangleq \{ i \in I \mid h_{j,i} = 1 \}$. The Tanner graph that is associated to $H$ will be denoted by $\Gamma(H)$; the graph distance of bit nodes $X_i$ and bit nodes $X_{j'}$ in $\Gamma(H)$ will then be denoted by $d_\Gamma(X_i, X_{j'})$. (Note that this latter quantity is always a non-negative even integer.) In the following, when no confusion can arise, we will sometimes omit the argument $H$ in the preceding expressions.

**Definition 2** The fundamental cone $K(H)$ of $H$ is the set of all vectors $\omega \in \mathbb{R}^n$ that satisfy

$$\omega_i \geq 0 \quad \text{(for all } i \in I(H)), \quad \omega_i \leq \sum_{i' \in J \setminus i} \omega_{i'} \quad \text{(for all } j \in J(H), \text{ for all } i \in I_j(H)). \quad (1, 2)$$

A vector $\omega \in K(H)$ is called a pseudo-codeword. If such a vector lies on an edge of $K(H)$, it is called a minimal pseudo-codeword. Moreover, if $\omega \in K(H) \cap \mathbb{Z}^n$ and $\omega \mod 2 \in C$, then $\omega$ is called an unscaled pseudo-codeword. (For a motivation of these definitions, see [7], [8]).

Although the region in the log-likelihood ratio vector space where linear-programming decoding decides for the all-zero codeword is completely characterized by the minimal pseudo-codewords of $K(H)$, the knowledge of non-minimal pseudo-codewords is also valuable since such pseudo-codewords can be used to bound this decision region.

**III. DEFINITION OF ABSDET-PSUEDOCODEWORDS AND PERM-PSUEDOCODEWORDS**

We start with the definition of det-vectors, absdet-vectors, and perm-vectors. As we will see, the properties of these vectors will then allow us to rename absdet-vectors and perm-vectors into absdet-pseudo-codewords and perm-pseudo-codewords, respectively.

**Definition 3** Let $C$ be a binary linear code described by a parity-check matrix $H \in F_2^{m \times n}$, $m < n$. For a size-$\nu$ subset $S$ of $I(H)$ we define the det-vector based on $S$ to be the vector $\nu \in \mathbb{Z}^n$ with components

$$\nu_i \triangleq \begin{cases} (-1)^{\eta_S(i)} \det(H_{S \setminus i}) & \text{if } i \in S \\ 0 & \text{otherwise} \end{cases},$$

where $\eta_S(i) \in \{0, 1, \ldots, |S|-1\}$ is the index of $i$ within the set $S$.

**Definition 4** Let $C$ be a binary linear code described by a parity-check matrix $H \in F_2^{m \times n}$, $m < n$. For a size-$\nu$ subset $S$ of $I(H)$ we define the absdet-vector based on $S$ to be the vector $\omega \in \mathbb{Z}^n$ with components

$$\omega_i \triangleq \begin{cases} \det(H_{S \setminus i}) & \text{if } i \in S \\ 0 & \text{otherwise} \end{cases}.$$
and noting that this latter matrix is singular (because at least two rows are equal), we obtain the result that \( s = 0 \), as promised.

The proof of (5) follows by noticing that \( \mathbf{H} \mathbf{\nu}^T = \mathbf{0}^T \) (in \( \mathbb{Z} \)) implies that \( \mathbf{H} \cdot (\mathbf{\nu} \mod 2) \mod 2 = \mathbf{0}^T \).

**Theorem 8** Let \( C \) be a binary linear code described by the parity-check matrix \( \mathbf{H} \in \mathbb{F}_2^{m \times n}, m < n \), and let \( S \) be a size-\((m+1)\) subset of \( I(\mathbf{H}) \). The absdet-vector \( \omega \) based on \( S \) is an unscaled pseudo-codeword of \( \mathbf{H} \), i.e.,

\[
\omega \in K(\mathbf{H}),
\]

\[
\omega \mod 2 \in C. \tag{7}
\]

**Proof:** Let \( \mathbf{\nu} \) be the det-vector based on \( S \). From Lemma 7 we know that \( \mathbf{\nu} \) satisfies \( \mathbf{H} \mathbf{\nu}^T = \mathbf{0}^T \) (in \( \mathbb{Z} \)). Because of this, and because \( \omega_i = |\nu_i| \) for \( i \in I \), we can use Lemma 6 to conclude that indeed \( \omega \in K(\mathbf{H}) \).

Finally, (8) is verified as follows. Lemma 7 shows that \( \mathbf{\nu} \mod 2 \in C \), which, upon noticing that \( \mathbf{\nu} \mod 2 = \omega \mod 2 \), implies that \( \omega \mod 2 \in C \).

**Theorem 9** Let \( C \) be a binary linear code described by the parity-check matrix \( \mathbf{H} \in \mathbb{F}_2^{m \times n}, m < n \), and let \( S \) be a size-\((m+1)\) subset of \( I(\mathbf{H}) \). The perm-vector \( \omega \) based on \( S \) is an unscaled pseudo-codeword of \( \mathbf{H} \), i.e.,

\[
\omega \in K(\mathbf{H}),
\]

\[
\omega \mod 2 \in C. \tag{9}
\]

**Proof:** In order to show (9), we need to verify (1) and (2). From Definition 5 it is clear that \( \omega \) satisfies (1). Therefore, let us focus on the proof that \( \omega \) satisfies (2). Fix some \( j \in J(\mathbf{H}) \) and some \( i \in I_j(\mathbf{H}) \). If \( i \notin S \) then \( \omega_i = 0 \) and (2) is clearly satisfied. Therefore, assume that \( i \in S \). Then

\[
\sum_{i' \in S \setminus i} \omega_{i'} = \sum_{i' \in S \setminus i} h_{j,i'} \omega_{i'}
\]

\[
= \sum_{i' \in S \setminus i} h_{j,i'} \cdot \text{perm}_z(\mathbf{H}_{S \setminus \iota}) + \sum_{i' \in (I \setminus S) \setminus i} h_{j,i'} \cdot 0
\]

\[
= \sum_{i' \in S \setminus i} h_{j,i'} \cdot \sum_{i'' \in S \setminus \iota} h_{j,i''} \cdot \text{perm}_z(\mathbf{H}_{j,i''} \setminus S \setminus \iota)
\]

\[
(\ast) \geq \sum_{i' \in S \setminus i} h_{j,i'} \cdot \text{perm}_z(\mathbf{H}_{j,i''} \setminus S \setminus \iota)
\]

\[
= h_{j,i} \sum_{i' \in S \setminus i} h_{j,i'} \cdot \text{perm}_z(\mathbf{H}_{j,i''} \setminus S \setminus \iota)
\]

\[
= h_{j,i} \cdot \text{perm}_z(\mathbf{H}_{S \setminus i}) = h_{j,i} \omega_i \tag{**}
\]

where at step (**) we kept only the terms for which \( i'' = i \), and where step (**) follows from \( j \in J(\mathbf{H}) \). Because \( j \in J(\mathbf{H}) \) and \( i \in I_j(\mathbf{H}) \) were arbitrary, \( \omega \) indeed satisfies (2).

Finally, (10) is verified as follows. Let \( \mathbf{\nu} \) be the det-vector based on \( S \). Lemma 7 shows that \( \mathbf{\nu} \mod 2 \in C \), which, upon noticing that \( \mathbf{\nu} \mod 2 = \omega \mod 2 \), implies that \( \omega \mod 2 \in C \).

**Definition 10** Because of Theorems 8 and 9 absdet-vectors and perm-vectors will herefore be called absdet-pseudo-codewords and perm-pseudo-codewords, respectively.

**IV. EXAMPLES**

In order to get a better feeling of what absdet-pseudo-codewords and perm-pseudo-codewords look like, let us discuss some examples.

**Example 11** Consider the \([4, 2, 2]\) binary linear code \( C \) based on the parity-check matrix \( \mathbf{H} = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \end{bmatrix} \), where \( n = 4 \) and \( m = 2 \). Let us compute the absdet-pseudo-codewords and perm-pseudo-codewords for all possible subsets \( S \) of \( I(\mathbf{H}) \) of size \( m+1 = 3 \). We obtain the following list of absdet-pseudo-codewords: \((0, 1, 1, 0)\) (twice), \((1, 1, 0, 1), (1, 0, 1, 1)\). These happen to be all the non-zero codewords of \( C \). Moreover, this parity-check matrix yields the following list of perm-pseudo-codewords: \((2, 1, 1, 0), (1, 1, 0, 1), (1, 0, 1, 1)\). Note that, up to scaling, \((2, 1, 1, 0)\) and \((0, 1, 1, 2)\) are the only non-codeword minimal pseudo-codewords of \( K(\mathbf{H}) \).

**Example 12** Consider the dumbbell-graph-based \([7, 2, 3]\) binary linear code described by the Tanner graph in Figure 1 (left) with \( n = 7 \) bit nodes and \( m = 6 \) check nodes. Obviously, there is only one subset \( S \) of \( I(\mathbf{H}) \) of size \( m+1 = 7 = n \), i.e. \( S = I(\mathbf{H}) \). This set \( S \) yields the absdet-pseudo-codeword \((2, 2, 2, 4, 2, 2, 2)\). Note that this is the only non-codeword minimal pseudo-codeword of \( K(\mathbf{H}) \) (cf. [7], [8]). Moreover, for this example the perm-pseudo-codeword based on \( S \) happens to be also \((2, 2, 2, 4, 2, 2, 2)\).

**Example 13** Consider the dumbbell-graph-based \([9, 2, 4]\) binary linear code described by the Tanner graph in Figure 1 (right) with \( n = 9 \) bit nodes and \( m = 8 \) check nodes. It yields the absdet-pseudo-codeword \((0, 0, 0, 0, 0, 0, 0, 0, 0)\) and the perm-pseudo-codewords \((2, 2, 2, 2, 4, 2, 2, 2, 2)\). Note that this latter vector is the only non-codeword minimal pseudo-codeword of \( K(\mathbf{H}) \).

**Example 14** Consider a randomly generated \((3, 4)\)-regular \([20, 5]\) LDPC code based on a \( 15 \times 20 \) parity-check matrix \( \mathbf{H}_1 \) that potentially contains four-cycles. The blue curve (top curve) in Figure 2 shows the AWGNC pseudo-weight cumulative histogram of the absdet-pseudo-codewords of \( \mathbf{H}_1 \) based on all subsets \( S \) of \( I(\mathbf{H}) \) of size \( m+1 = 16 \).
Moreover, for any set $S$, the entries of the abdet-pseudo-codeword are always upper bounded by the corresponding entries in the perm-pseudo-codeword, and so the claim also follows for abdet-pseudo-codewords.

**Remark 17** For a parity-check matrix $H$, the existence of short cycles in $T(H)$ has an influence on the list of abdet-pseudo-codewords. In particular, without going into the details, four-cycles imply more abdet-pseudo-codewords that equal the all-zero codeword. Related statements can be made about six-cycles, eight-cycles, etc. As part of future research, it will be interesting to formulate probabilistic statements that will help characterizing long codes where not all abdet-pseudo-codewords can be listed by brute-force techniques.

**Remark 18** Note that the pseudo-codeword $(2, 2, 2, 4, 2, 2)$ that was found in Example 12 can be seen as a canonical completion with root $X_i$ [6], [7]. (Strictly speaking, the canonical completion was only defined for check-degree regular codes, however, it is straightforward to suitably extend the canonical completion technique to Tanner graphs where the check nodes with the same graph distance to the root node have the same degree.)

More generally, one can establish the following connection between Lemma 6 and the canonical completion. Namely, let $\omega \in \mathbb{R}^n$ be the canonical completion with root $X_i$ for some $i \in I(H)$ and define $\nu \in \mathbb{R}^n$ such that

$$
\nu_i' = \begin{cases} 
+\omega_i' & \text{(if } d_T(H)(X_i, X_{i'}) \in 4\mathbb{Z}) \\
-\omega_i' & \text{(if } d_T(H)(X_i, X_{i'}) \in 2\mathbb{Z} \setminus 4\mathbb{Z})
\end{cases}
$$

for all $i' \in I(H)$. (With this, $\omega$ obviously satisfies $\omega'_i = |\nu_i'|$ for all $i' \in I(H)$.) Let $J'(H)$ be the subset of indices of check nodes that have only one neighboring bit node that is closer (in graph distance) to the root than they are to the root. It can then easily be verified that $H_{J';T} \cdot \nu^T = 0^T$, which, with the help of Lemma 6 implies that $\omega \in K(H_{J';T})$.

The next theorem relates abdet-pseudo-codewords to quantities that appear naturally in a certain Gaussian graphical model associated to $T(H)$. In order to motivate the Gaussian graphical model in that theorem, remember that a Tanner/factor graph of a code represents the indicator function $[x \in C] = \prod_{i \in I} f'_i(x_i) \cdot \prod_{j \in J} f''_j(x_{I_j})$ with $f'_i(x_i) : [x_i \in \{0, 1\}]$ and $f''_j(x_{I_j}) : [\sum_{x_i \in I_j} x_i \equiv 0 \; (\mod 2)]$, and that the indicator function of the fundamental cone can be written as $[\omega \in K(H)] = \prod_{i \in I} k'_i(\omega_i) \cdot \prod_{j \in J} k''_j(\omega_{I_j})$ with $k'_i(\omega_i) : [x_i \geq 0]$ and some suitably defined functions $k''_j(x_{I_j})$.

**Theorem 19** Let $C$ be a binary linear code described by the parity-check matrix $H \in \mathbb{F}_2^{m \times n}$. For some arbitrary $\varepsilon > 0$, consider the Gaussian graphical model for the length-$n$ vector $U$ defined by $p_U(u) \propto \prod_{i \in I} g'_i(u_i) \cdot \prod_{j \in J} g''_j(u_{I_j})$ with

$$
g'_i(u_i) \propto \exp \left( -\frac{u_i^2}{2(1/\varepsilon)^2} \right), \quad g''_j(u_{I_j}) \propto \exp \left( -\frac{1}{2} \sum_{(i,i') \in I_j \times I_j} u_i u_{i'} \right).
$$

$$
\text{Cum. number of absdet−PCWs} = \begin{cases} 
4000 & 0 \\
5000 & 2 \end{cases}
$$

Figure 2. AWGNC pseudo-weight cumulative histograms of the abdet-PCWs (abdet-pseudo-codewords) of the parity-check matrices in Example 14. (The AWGNC pseudo-weight of a pseudo-codeword $\omega$ is defined to be $w^P_{AWGNC}(\omega) = \|\omega\|^2_2/\|\omega\|_2^2$ [6], [7].) Blue curve (top curve): $H_1$. Red curve (bottom curve): $H_2$.

In this section we discuss some properties of abdet-pseudo-codewords and of perm-pseudo-codewords. Some proofs are shortened or omitted due to space restrictions.

**Remark 15** Using a well-known property of permanents of matrices with zeros and ones, it follows that the term $\text{perm}_Z(H_{S \setminus i})$, which appears in the definition of perm-vectors in Definition 5 equals the number of perfect matchings in the Tanner graph $T(H_{S \setminus i})$. Moreover, because Theorem 9 showed that perm-vectors satisfy (2), we see that for every $\omega \in H_{S \setminus i}$,\footnote{Due to space restrictions, we refer the reader to [6] for a more detailed discussion of the canonical completion.} the claim then follows from Remark 15.

**Theorem 16** Let $H$ be the parity-check matrix of a code whose associated Tanner graph is a tree, i.e., does not contain cycles. Then all entries of all abdet-pseudo-codewords and all entries of all perm-pseudo-codewords are either 0 or 1.

**Proof:** (Sketch.) A necessary condition for the Tanner graph $T(H_{S \setminus i})$ to have at least two perfect matchings is the existence of a cycle in the Tanner graph. However, if $T(H)$ is cycle-free then also $T(H_{S \setminus i})$ is cycle-free. For perm-pseudo-codewords the claim then follows from Remark 15.

Eliminating four-cycles in $T(H_1)$ by applying an edge-permutation procedure results in a Tanner graph $T(H_2)$ of a new code described by a parity-check matrix $H_2$. The red curve (bottom curve) in Figure 2 shows the AWGNC pseudo-weight cumulative histogram of the abdet-pseudo-codewords of $H_2$ based on all subsets $S$ of $I(H)$ of size $m+1=16$.

Comparing these two curves, we make the following observation: first, for $H_1$ there are more abdet-pseudo-codewords than for $H_2$ that equal the all-zero vector. As we will briefly discuss in the next section, this observation is related to the existence of four-cycles. Secondly, the curve related to $H_1$ is to the left of the curve related to $H_2$. This corroborates the common observation that codes based on Tanner graphs with four-cycles usually perform worse than codes based on Tanner graphs without four-cycles.

**V. PROPERTIES OF ABDET-PSEUDO-CODEWORDS AND PERM-PSEUDO-CODEWORDS**

In this section we discuss some properties of abdet-pseudo-codewords and of perm-pseudo-codewords. Some proofs are shortened or omitted due to space restrictions.
Let $\mathcal{S}$ be a size-$(m+1)$ subset of $\mathcal{I}(H)$ and let $\tilde{\mathcal{S}} = \mathcal{I}(H) \setminus \mathcal{S}$ be its complement. (We assume that $m < n$.) Let $\sigma_{i|\tilde{\mathcal{S}}} (\epsilon)$ be the square root of the minimum mean squared error when estimating $U_i$ (with a linear or a non-linear estimator) based on the knowledge of $U_{\tilde{\mathcal{S}}} = u_{\tilde{\mathcal{S}}}$, then the components of the absdet pseudo-codeword $\omega$ based on $\mathcal{S}$ fulfill
\[
\omega_i = \lim_{\epsilon \rightarrow 0} \gamma_{i|\tilde{\mathcal{S}}} (\epsilon) \cdot \sigma_{i|\tilde{\mathcal{S}}} (\epsilon) \tag{11}
\]
for all $i \in \mathcal{I}(H)$, where $\gamma_{i|\tilde{\mathcal{S}}} (\epsilon)$ is a function of $\epsilon$, but independent of $i \in \mathcal{I}(H)$.

\section*{Proof:}
First, we consider the case where $i \in \tilde{\mathcal{S}}$. From Definition 4 we see that $\omega_i = 0$. On the other hand, $U_i$ can perfectly be predicted based on the knowledge of $U_{\tilde{\mathcal{S}}} = u_{\tilde{\mathcal{S}}}$, which implies $\sigma_{i|\tilde{\mathcal{S}}} (\epsilon) = 0$. Since $\gamma_{i|\tilde{\mathcal{S}}} (\epsilon)$ (defined below) is bounded for all suitably small $\epsilon > 0$, we have proven (11) for $i \in \tilde{\mathcal{S}}$.

Secondly, we consider the case where $i \in \mathcal{S}$. We start by noting that $p_{\tilde{\mathcal{S}}}(u)$ can be written as $p_{\tilde{\mathcal{S}}}(u) \propto \exp \left( -\frac{1}{2} u^T G u \right)$ with the positive definite matrix $G = \bar{\epsilon}^2 I_{n \times n} + H^T H$, where $I_{n \times n}$ is the $n \times n$ identity matrix. Then, $p_{U_{\tilde{\mathcal{S}}|U_{\mathcal{S}}} (u|u_{\mathcal{S}})} \propto \exp \left( -\frac{1}{2} u^T G_{S|\tilde{\mathcal{S}}} u + n_{S|\tilde{\mathcal{S}}} u_{\mathcal{S}} \right)$, with the positive definite matrix $G_{S|\tilde{\mathcal{S}}} = \bar{\epsilon}^2 I_{n \times n} + H^T H$ and with $n_{S|\tilde{\mathcal{S}}}$ being a linear function of $u_{\mathcal{S}}$. The inverse matrix of $G_{S|\tilde{\mathcal{S}}}$ is the covariance matrix $R_{S|\tilde{\mathcal{S}}}$ of $U_{S|\tilde{\mathcal{S}}}$. For $i \in \mathcal{S}$, a well-known property of jointly Gaussian random variables says that the $i$-th diagonal entry of $R_{S|\tilde{\mathcal{S}}}$ equals $\sigma^2_{i|\tilde{\mathcal{S}}} (\epsilon)$.

Without loss of generality, we can assume that $\mathcal{S} = \{0, 1, \ldots, m\}$ and that $i = 0$. Because $H_{S} = (H_{0|0} H_{S\setminus 0})$, we obtain
\[
G_{S|\tilde{\mathcal{S}}} = \begin{bmatrix}
\bar{\epsilon}^2 I_{1 \times 1} + H^T_{0|0} H_{0|0} & H^T_{0|0} H_{S\setminus 0} \\
H^T_{S\setminus 0} H_{0|0} & \bar{\epsilon}^2 I_{m \times m} + H^T_{S\setminus 0} H_{S\setminus 0}
\end{bmatrix}
\]

Since $\sigma^2_{0|\tilde{\mathcal{S}}} (\epsilon)$ is the $(0,0)$-entry of $R_{S|\tilde{\mathcal{S}}} = G_{S|\tilde{\mathcal{S}}}^{-1}$, we have
\[
\sigma^2_{0|\tilde{\mathcal{S}}} (\epsilon) = \gamma^2_{0|\tilde{\mathcal{S}}} (\epsilon) \cdot \det \left( \bar{\epsilon}^2 I_{m \times m} + H^T_{S\setminus 0} H_{S\setminus 0} \right),
\]
where $\gamma_{0|\tilde{\mathcal{S}}} (\epsilon) \triangleq \sqrt{\det(G_{S|\tilde{\mathcal{S}}})}$. In the limit $\epsilon \rightarrow 0$ we have
\[
\lim_{\epsilon \rightarrow 0} \left( \bar{\epsilon}^2 I_{m \times m} + H^T_{S\setminus 0} H_{S\setminus 0} \right) = \det(z) (H^T_{S\setminus 0} H_{S\setminus 0}) = \det(z) (H_{S\setminus 0})^2 \gamma_{0|\tilde{\mathcal{S}}} (\epsilon) = \omega_0^2,
\]
where step $(*)$ follows from Definition 4. Similar expressions easily follow for other $i \in \mathcal{S}$, therefore proving (11).

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