# Finding the Complete Path and Weight Enumerators of Convolutional Codes 

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A method for obtaining the complete path enumerator $T(D, L, I)$ of a convolutional code is described. A system of algebraic equations is solved, using a new algorithm for computing determinants, to obtain $T(D, L, I)$ for the ( $7,1 / 2$ ) NASA standard code. Generating functions, derived from $T(D, L, I)$, are used to upper bound Viterbi decoder error rates. This technique is currently feasible for constraint length $K<10$ codes.

A practical, $\overline{\text { fast algorithm is presented for computing the leading nonzero }}$ coefficients of the generating functions used to bound the performance of constraint length $K<20$ codes. Code profiles with about 50 nonzero coefficients are obtained with this àlgorithm for the experimental $K=15$, rate $1 / 4$, code in the Galileo mission and for the proposed $K=15$, rate $1 / 6,{ }^{\prime} 2-d B^{*}$ code.

## I. Introduction

Convolutional codes such as the ( $7,1 / 2$ ) NASA standard have been used for satellite and deep-space communications during the past 20 years. In 1971, Viterbi [1] defined generating functions for upper bounding error probabilities of convolutional codes on memoryless channels. In practice, for codes with more than eight states, these functions are still unknown, so error bounds have been evaluated using numerical matrix multiplications $[2,8]$, which require extensive computations for each channel noise level.

[^0]As an alternative, algorithms have also been developed to calculate the first few coefficients (the distance and bit error profiles) of these enigmatic generating functions [4,5]. Unfortunately, lists of these numbers are bulky (see Tables 2 and 3 ), and the minimum number of terms required to approximate the decoder error bounds depends upon the code rate and channel noise level.

A code's complete path enumerator $T(D, L, I)$ contains the number of paths having identical triples: weight, length, number of input 1 s . The least-magnitude pole of the weight enumerator $T(D)=T(D, 1,1)$ determines the point at which the union bounds [1] diverge, while additional poles and residues yield the dominant terms in the
partial fraction expansion of $T(D)$. Concise, yet very accurate, analytic approximations to several generating functions' coefficients may be obtained from these dominant terms. Furthermore, the poles and residues of $T(D)$ may eventually help unlock the structure of convolutional codes.

The generating functions $T(D)$ and $T(D, L, I)$ for the ( $7,1 / 2$ ) NASA standard code (among many others) were obtained with a simple determinant algorithm described in this article. In addition, a new algorithm is presented for computing the initial coefficients of $T(D), \partial T(D, L, I) / \partial L$ and $\partial T(D, L, I) / \partial I$ at $L=I=1$. These numbers are used to upper bound a Viterbi decoder's event, node, bit, and symbol error rates. The techniques explained here may be adapted, with modifications for code nonlinearity and Euclidean instead of Hamming distances, to find weight enumerators of trellis codes [3,5].

## II. The Complete Path Enumerator

A binary, rate $k / n$, feed-forward convolutional encoder is defined by $k n$ binary generator polynomials $g_{i j}(x)$, each representing the transfer function from the $i$ th input to the $j$ th output ( $x$ is a delay operator). Let $g_{i j}$ be the vector whose $r$ th component is the coefficient of $x^{r}$ in $g_{i j}(x)$. The encoder's memory is $m=$ $\sum_{i=1}^{k} \max _{j}\left[\operatorname{deg} g_{i j}(x)\right]$, which for rate $1 / n$ codes is $K-1$. If the $i$ th encoder memory cell contains $s_{i}$ then the encoder is in state $s=\sum_{i=1}^{m} s_{i} 2^{i-1}$. The encoder's state diagram is a directed graph whose edges (corresponding to branches in the associated trellis diagram) have labels $a_{i, j}=D^{d} L I^{b}$ if there is an edge from state $j$ into state $i$ while $a_{i, j}=0$ otherwise. During a transition from state $j$ into state $i, b$ is the number of ones input to the encoder and the number of ones output by the encoder is $d=\sum_{h=1}^{n} \underline{g}_{1 h} \cdot[2 j+(i \bmod 2)]_{2}$, where $\cdot$ is a modulo-2 inner product and $[k]_{2}$ denotes the binary representation of the integer $k$.

Let $X_{s}$ be the trivariate generating function of all simple paths: those from state 0 into state $s$ via nonzero states. $X_{0}$ counts all simple paths into state 0 , called fundamental paths. Note that $X_{s}$ is indexed by the destination state (s) while the source state (0) is constant. Now define $\mathbf{A}$ as the $2^{m}-1$ by $2^{m}-1$ adjacency matrix of the encoder graph with state 0 and its edges removed. The entry in row $i$ and column $j$ of $\mathbf{A}$ is $a_{i, j}$, which is nonzero only if there is a directed edge connecting state $j$ to state $i$ (they are adjacent). Now the following set of linear equations is constructed:

$$
\underline{\mathbf{A}}\left[X_{1}, X_{2}, \ldots, X_{2^{m}-1}\right]^{\mathbf{T}}=\left[a_{1,0}, a_{2,0}, \ldots, a_{2 m-1,0}\right]^{\mathrm{T}}
$$

By Cramer's Rule,

$$
X_{\mathbf{i}}=\frac{\operatorname{det}\left(\mathbf{A}_{\mathbf{i}}\right)}{\operatorname{det}(\underline{\mathbf{A}})}
$$

for $i>0$, where $\underline{\mathbf{A}}_{\mathbf{i}}$ is $\mathbf{A}$ with all column $i$ entries $a_{r, i}$ replaced by $a_{r, 0}$ for all rows $r$. A code's complete path enumerator (by weight, length, and number of input 1s)

$$
T(D, L, I)=X_{0}=\sum_{j=1}^{2^{k}} a_{0, z_{j}} X_{z_{j}}
$$

where $\left\{z_{j}\right\}$ are the $2^{k}$ states having edges into state 0 (so $a_{0, z_{j}} \neq 0$ ).

As an example with $k=1$ for simplicity, the $m=2$ encoder in Fig. 1 is in state $s=2 s_{1}+s_{0}$. The corresponding state diagram in Fig. 2 leads to the equations $\underline{\mathbf{A}}\left[X_{1}, X_{2}, X_{3}\right]^{\mathrm{T}}=\left[\mathrm{a}_{1,0}, 0,0\right]^{\mathrm{T}}:$

$$
\left[\begin{array}{ccc}
1 & -L I & 0 \\
-D L & 1 & -D L \\
-D L I & 0 & 1-D L I
\end{array}\right]\left[\begin{array}{l}
X_{1} \\
X_{2} \\
X_{3}
\end{array}\right]=\left[\begin{array}{c}
D^{2} L I \\
0 \\
0
\end{array}\right]
$$

Now $T(D, L, I)=X_{0}=D^{2} L X_{2}=D^{5} L^{3} I /(1-D L I-$ $D L^{2} I$ ), so the code's weight enumerator is

$$
T(D)=\frac{D^{5}}{1-2 \bar{D}}=\sum_{d=5}^{\infty} 2^{d-5} D^{d}=\sum_{d=d_{\text {tree }}}^{\infty} p(d) D^{d}
$$

where $p(d)=2^{d-5}$ is the number of weight $d$ fundamental paths, and $d_{\text {free }}=5$ is the code's free distance.

## III. Reducing an Adjacency Matrix

When $k=1$, there are only $3\left(2^{m}-1\right)-2$ nonzero out of $\left(2^{m}-1\right)^{2}$ entries in $\underline{\mathbf{A}}$, and they are located in a special pattern. Since $\underline{A}$ is sparse if $m>3$, reduction is useful before computing the determinant. The following example, different from the one in the previous section, illustrates the reduction procedure. The code with generator polynomials $g_{11}(D)=1+D+D^{3}$ and $g_{12}(D)=1+D+D^{2}+D^{3}$
has an adjacency matrix $\underline{\mathbf{A}}$ (with $L=I=1$ to simplify entries)

$$
\left[\begin{array}{ccccccc}
1 & 0 & 0 & \left(D^{2}\right)-1 & 0 & 0 & 0 \\
-D^{2} & 1 & 0 & 0 & -1 & 0 & 0 \\
-1 & 0 & 1 & 0 & -D^{2} & 0 & 0 \\
0 & -D & 0 & (0) 1 & 0 & -D & 0 \\
0 & -D & 0 & 0 & 1 & -D & 0 \\
0 & 0 & -D & 0 & 0 & 1 & -D \\
0 & 0 & -D & 0 & 0 & 0 & 1-D
\end{array}\right]
$$

The round brackets (parentheses) above indicate values in $\underline{\mathbf{A}}_{2^{m-1}}$ that are different from those in $\underline{\mathbf{A}}$. Since $k=1$, the determinants of only these two matrices are needed for $T(D, L, I)$, and this notation will lead to their simultaneous computation.

As in Gaussian elimination, $-a_{r,\lfloor r / 2\rfloor}$ times row $\lfloor r / 2\rfloor$ is added to rows $r=2$ to $2^{m}-1$ so that $\underline{\mathbf{A}}$ becomes zeroed below its main diagonal for columns 1 to $2^{m-1}-1$ :

$$
\left[\begin{array}{ccccccc}
1 & 0 & 0 & \left(D^{2}\right)-1 & 0 & 0 & 0 \\
0 & 1 & 0 & \left(D^{4}\right)-D^{2} & -1 & 0 & 0 \\
0 & 0 & 1 & \left(D^{2}\right)-1 & -D^{2} & 0 & 0 \\
0 & 0 & 0 & \left(D^{5}\right) 1-D^{3} & -D & -D & 0 \\
0 & 0 & 0 & \left(D^{5}\right)-D^{3} & 1-D & -D & 0 \\
0 & 0 & 0 & \left(D^{3}\right)-D & -D^{3} & 1 & -D \\
0 & 0 & 0 & \left(D^{3}\right)-D & -D^{3} & 0 & 1-D
\end{array}\right]
$$

Therefore, $\operatorname{det}(\underline{\mathbf{A}})$ equals the determinant of the resulting lower right $2^{m-1}$ by $2^{m-1}$ submatrix. To further reduce $\mathbf{A}$, row $\lfloor r / 2\rfloor$ times $-a_{r, 2^{m-1}+\lfloor r / 2\rfloor}$ is added to each row $r=$ $2^{m}-3$ to $2^{m-1}$ ( 4 to 5 here), so that columns $2^{m-1}+2^{m-2}$ to $2^{m}-2$ (both 6 here) are zeroed above the main diagonal. Now $\operatorname{det}(\underline{\mathbf{A}})=\operatorname{det}(\underline{\mathbf{A}})$, where $\underline{\underline{\mathbf{A}}}_{2^{m-1}}$ ) and $\underline{\overline{\mathbf{A}}}$ are the new lower right $2^{m-1}$ by $2^{m-1}$ submatrices:

$$
\left[\begin{array}{cccc}
\left(D^{4}+D^{5}\right) 1-D^{2}-D^{3} & -D-D^{4} & 0 & -D^{2} \\
\left(D^{4}+D^{5}\right)-D^{2}-D^{3} & 1-D-D^{4} & 0 & -D^{2} \\
\left(D^{3}\right)-D & -D^{3} & 1 & -D \\
\left(D^{3}\right)-D & -D^{3} & 0 & 1-D
\end{array}\right]
$$

To zero column $2^{m-1}$ of $\underline{\tilde{A}}$ (above the diagonal entry $\left.\tilde{a}_{2^{m-1}, 2^{m-1}}=a_{2^{m}-1,2^{m}-1} \neq 0\right),-\tilde{a}_{r, 2^{m-1}} / \tilde{a}_{2^{m-1}, 2^{m-1}}$ times row $2^{m-1}$ is added to each row $r=1$ to $2^{m-1}-1$. Define
( $\underline{B}_{2^{m-1}}$ ) and $\underline{B}$ as the resulting upper left $2^{m-2}$ by $2^{m-2}$ submatrices,

$$
\left.\left.\frac{\left[\begin{array}{cc}
\left(D^{4}+D^{5}-D^{6}\right) 1-D-D^{2}-D^{3}+D^{4} & -D+D^{2}-D^{4} \\
\left(D^{4}+D^{5} .-D^{6}\right) & -D^{2}-D^{3}+D^{4}
\end{array}\right.}{1-2 D+D^{2}-D^{4}}\right]\right] .
$$

This reduction method simultaneously produces two dense $2^{m-2}$ by $2^{m-2}$ matrices $\left(\underline{\mathbf{B}}_{2^{m-1}}\right)$, $\underline{\mathrm{B}}$ with the same determinants as the corresponding original sparse $2^{m}-1$ by $2^{m}-1$ matrices $\left(\underline{\mathbf{A}}_{2^{m-1}}\right), \underline{\mathbf{A}}$.

## IV. A Determinant Algorithm

The following algorithm yields the determinant of any $N \times N$ matrix $\underline{B}$ having entries from a Euclidean domain, such as the set of all polynomials with integer coefficients. A sequence of matrices $\underline{B}^{(N)}, \underline{B}^{(N-1)}, \underline{B}^{(N-2)}, \ldots, \underline{B}^{(1)}$ is computed with each matrix having the same determinant (up to sign) as $\underline{\mathbf{B}}$. Starting with $\underline{\mathbf{B}}^{(\mathbf{N})}=\underline{\mathbf{B}}$, step $j$ in the algorithm produces the numerator $b_{i, k}^{(j-1)}$ of each entry in row $i$ and column $k$ of $\underline{B}^{(\mathbf{j}-1)}$ :

$$
\text { for } j=N \text { to } 2
$$

for $i=j-1$ to $1 \quad$ (row index)
for $k=j$ to $1 \quad$ (column index)

$$
b_{i, k}^{(j-1)}=\frac{b_{i, k}^{(j)} b_{j, j}^{(j)}-b_{i, j}^{(j)} b_{j, k}^{(j)}}{b_{j+1, j+1}^{(j+1)}}
$$

Naturally, if $b_{j+1, j+1}^{(j+1)}=0$ prior to step $j$, then any column $k$ such that $b_{j+1, k}^{(j+1)} \neq 0$ must first be interchanged with column $j+1$. This operation negates the determinant, so a counter $t$ is incremented to record the event. If no such column $k$ exists, row $j+1$ is zero, so the algorithm is stopped and $\operatorname{det}(B)=\operatorname{det}\left(B^{(j)}\right)=0$. Also note that $b_{N+1, N+1}^{(N+1)}=1$ initially.

The following example with $N=4$ illustrates the above reduction procedure. When $j=4$, after $-b_{34}^{(4)} / b_{44}^{(4)}$, $-b_{24}^{(4)} / b_{44}^{(4)}$, and $-b_{14}^{(4)} / b_{44}^{(4)}$ times row 4 are added to rows 3,2 , and 1 , respectively (corresponding to $j=N=4$ in the
algorithm), in order to zero all entries in column 4 above $b_{44}^{(4)}$,

$$
\underline{\mathbf{B}}^{(3)}=\left[\begin{array}{llll}
\frac{b_{11}^{(3)}}{b_{44}^{(4)}} & \frac{b_{12}^{(3)}}{b_{44}^{(4)}} & \frac{b_{13}^{(3)}}{b_{44}^{(4)}} & 0 \\
\frac{b_{21}^{(3)}}{b_{44}^{(4)}} & \frac{b_{22}^{(3)}}{b_{44}^{(4)}} & \frac{b_{23}^{(3)}}{b_{44}^{(4)}} & 0 \\
\frac{b_{31}^{(3)}}{b_{44}^{(4)}} & \frac{b_{32}^{(3)}}{b_{44}^{(4)}} & \frac{b_{33}^{(3)}}{b_{44}^{(4)}} & 0 \\
b_{41}^{(4)} & b_{42}^{(4)} & b_{43}^{(4)} & \\
b_{44}^{(4)}
\end{array}\right]
$$

Then, after the $j=3$ step during which $-b_{23}^{(3)} / b_{33}^{(3)}$ and $-b_{13}^{(3)} / b_{33}^{(3)}$ times row 3 are added to rows 2 and 1 ,

$$
\underline{\mathbf{B}}^{(2)}=\left[\begin{array}{cccc}
\frac{b_{11}^{(2)}}{b_{33}^{(3)}} & \frac{b_{12}^{(2)}}{b_{33}^{(3)}} & 0 & 0 \\
\frac{b_{21}^{(2)}}{b_{33}^{(3)}} & \frac{b_{22}^{(2)}}{b_{33}^{(3)}} & 0 & 0 \\
\frac{b_{31}^{(3)}}{b_{44}^{(4)}} & \frac{b_{32}^{(3)}}{b_{44}^{(4)}} & \frac{b_{33}^{(3)}}{b_{44}^{(4)}} & 0 \\
b_{41}^{(4)} & b_{42}^{(4)} & b_{43}^{(4)} & b_{44}^{(4)}
\end{array}\right]
$$

Since only elementary row operations on $\mathbf{B}$ have been performed, $\operatorname{det}(\underline{\mathbf{B}})=\operatorname{det}\left(\underline{B}^{(3)}\right)=\operatorname{det}\left(\underline{\mathbf{B}}^{(\mathbf{B})}\right)$. The final step ( $j=2$ ) produces a $\underline{\mathbf{B}}^{(\mathbf{1})}$ matrix identical to $\underline{\mathbf{B}}^{(\mathbf{2})}$ except with $b_{11}^{(1)} / b_{22}^{(2)} \quad 0 \quad 0 \quad 0$ in the first row. Therefore, $\operatorname{det}(\underline{\mathbf{B}})=\operatorname{det}\left(\underline{B}^{(1)}\right)=b_{11}^{(1)}$. This result generalizes to any nonsingular $N \times N$ matrix $\mathbf{B}$.

Lemma. Entry $b_{1,1}^{(1)}=\operatorname{det}\left(\underline{\mathbf{B}}^{(1)}\right)=(-1)^{t} \operatorname{det}(\underline{B})$, where $t$ is the total number of column interchanges performed by the algorithm in computing $b_{1,1}^{(1)}$.

Proof. For each value of $j$ from $N$ to 2 , when $k=j$, the algorithm makes $b_{i, j}^{(j-1)}=0$ for all $1 \leq i \leq j-1$ so that all column $j$ entries above the main diagonal become 0 . Since $\underline{B}^{(1)}$ is zero above its main diagonal, $\operatorname{det}(\underline{\mathbf{B}})$ is the product of $(-1)^{t}$ and the diagonal entries in $\underline{\mathbf{B}}^{(1)}$.

$$
\begin{aligned}
\operatorname{det}(\underline{\mathbf{B}}) & =(-1)^{t} \operatorname{det}\left(\underline{\mathbf{B}}^{(\mathbf{1})}\right) \\
& =(-1)^{t} \prod_{j=1}^{N} \frac{b_{j, j}^{(j)}}{b_{j+1, j+1}^{(j+1)}}
\end{aligned}
$$

$$
=(-1)^{t} b_{1,1}^{(1)}
$$

Expanding $b_{i, k}^{(j)} b_{j, j}^{(j)}-b_{i, j}^{(j)} b_{j, k}^{(j)}$ using the equation in the algorithm shows that $b_{j+1, j+1}^{(j+1)}$ divides this expression so there are never any remainders. This is expected because the algorithm implements a recursive factorization of the determinant written as a sum of products of matrix entries. The algorithm differs from standard Gaussian elimination because the diagonal entries in the reduced matrices are not made equal to 1. Also, after calculations with a particular value of $j$ are completed, all entries' denominators are previous pivot numerators.

## V. Path Distance, Length, and Bit Error Approximations

The complete path enumerator $T(D, L, I)$ for the $m=6$, rate $1 / 2$, NASA standard code was obtained by using the preceding algorithm to simultaneously compute determinants of the two 16 by 16 reduced matrices $\underline{B}_{2^{m-1}}$ and $\underline{B}$. The 76 poles of $T(D)$ for this code are plotted on the complex plane (Fig. 3) along with the unit circle for reference. Using only the six least-magnitude poles (indicated by the large points in Fig. 3), an approximation to the partial fraction expansion of $T(D)$ is

$$
\begin{aligned}
T(D) \approx & D^{10}\left[\frac{r_{1}}{1-\alpha_{d} D}+\frac{r_{1}}{1+\alpha_{d} D}+\frac{r_{2}}{1-\alpha_{b}}\right. \\
& \left.+\frac{r_{2}}{1+\alpha_{b} D}+\frac{r_{2}^{*}}{1-\alpha_{b}^{*} D}+\frac{r_{2}^{*}}{1+\alpha_{b}^{*} D}\right] \\
= & D^{10} \sum_{k=0}^{\infty}\left[r_{1}\left(\alpha_{d}\right)^{k}+r_{1}\left(-\alpha_{d}\right)^{k}+r_{2}\left(\alpha_{b}\right)^{k}\right. \\
& \left.+r_{2}\left(-\alpha_{b}\right)^{k}+r_{2}^{*}\left(\alpha_{b}^{*}\right)^{k}+r_{2}^{*}\left(-\alpha_{b}^{*}\right)^{k}\right] D^{k} \\
= & \sum_{j=0}^{\infty}\left[2 r_{1}\left(\alpha_{d}\right)^{2 k}+4 \operatorname{Re}\left\{r_{2}\left(\alpha_{b}\right)^{2 k}\right\}\right] D^{2 k+10}
\end{aligned}
$$

where $\alpha_{d}=2.3876225$ is the reciprocal of the leastmagnitude pole of $T(D), \alpha_{b}=1.657193 e^{-0.983418 j}$ is the reciprocal of the pole with next smallest magnitude, * means complex conjugate, and $j=\sqrt{-1}$. The residues are

$$
r_{1}=\frac{-P\left(\alpha_{d}^{-1}\right)}{\alpha_{d}^{-1} Q^{\prime}\left(\alpha_{d}^{-1}\right)} \quad \text { and } \quad r_{2}=\frac{-P\left(\alpha_{b}^{-1}\right)}{\alpha_{b}^{-1} Q^{\prime}\left(\alpha_{b}^{-1}\right)}
$$

where $D^{10} P(D) / Q(D)=T(D)$ and $Q^{\prime}(D)$ is the derivative of $Q(D)$. The other poles in Fig. 3 may be ignored because their magnitudes are greater than 0.8 , and the corresponding residues have magnitudes less than 0.07 .

Define $p(d), i(d)$, and $\ell(d)$ as the coefficients of $D^{d}$ in $T(D), \partial T(D, L, I) / \partial L$, and $\partial T(D, L, I) / \partial L$, respectively, at $L=I=1$ (these generating functions are shown in the Appendix). The number of fundamental paths having weight $2 k+10$ is

$$
\begin{aligned}
p(2 k+10) \approx & 6.82(2.3876225)^{2 k} \\
& +4.25(1.657193)^{2 k} \cos (0.310-1.967 k)
\end{aligned}
$$

Similarly, the terms corresponding to the six leastmagnitude poles of $T(D)$ in the partial fraction expansions of $\partial T(D, L, I) / \partial L$ and $\partial T(D, L, I) / \partial I$ at $L=I=1$ were used to obtain the approximations

$$
\begin{aligned}
\ell(2 k+10) & \approx(77.725+22.625 k)(2.3876225)^{2 k} \\
& +39.3(1.657193)^{2 k} \cos (0.485-1.967 k) \\
& +(2 k+1) 7.2676(1.657193)^{2 k} \cos (0.383-1.967 k) \\
i(2 k+10) & \approx(24.474+12.018 k)(2.3876225)^{2 k} \\
& +9.942(1.657193)^{2 k} \cos (0.575-1.967 k) \\
& +(2 k+1) 2.8723(1.657193)^{2 k} \cos (0.366-1.967 k)
\end{aligned}
$$

which have a relative error $<0.0001$ for $k>4$.

A rate $1 / n$ Viterbi decoder's bit error rate (BER) on a binary-input, output-symmetric [2], discrete memoryless channel is bounded by

$$
\mathrm{BER} \leq \sum_{d=d_{\mathrm{free}}}^{\infty} i(d) P_{d}
$$

where $P_{d}$ is the probability that the decoder outputs a fundamental path having distance $d$ from the one transmitted. The probability that a $b$-bit symbol is decoded incorrectly is bounded by [9]

$$
\mathrm{SER}_{b} \leq \sum_{d=d_{\mathrm{free}}}^{\infty}[[b-1-m] p(d)+\ell(d)] P_{d}
$$

For the additive white Gaussian noise (AWGN) channel with bit signal-to-noise ratio $E_{b} / N_{0}$,

$$
P_{d}=Q\left(\sqrt{2 d R E_{b} / N_{0}}\right)
$$

where $Q(x)$ is the Gaussian integral function [2, p. 62]. On a binary symmetric channel with crossover probability $p$,
$P_{2 i}=P_{2 i-1}=\sum_{e=i}^{2 i-1}\binom{2 i-1}{e} p^{e}(1-p)^{2 i-1-e} \leq\binom{ 2 i-1}{i} p^{i}$
[6]. For decoders using integer metrics, as for example on a binary-input, output-quantized AWGN channel, $P_{d}$ can also be computed exactly [2, p. 291].

## VI. Algorithms for Profiles of Convolutional Codes

Finding the complete path enumerator of codes with memory greater than 8 currently seems infeasible. In these cases, an algorithm for distance profiles [4,5] may be used to calculate the first few nonzero coefficients of the generating functions used for error bounds. However, these methods are fairly complex and some require extra computation to ensure that the output is correct. In this section, Viterbi's algorithm, with survivors replaced by vectors of integers that count paths, lengths, or bits, is applied on a noiseless channel to compute $p(d), \ell(d)$, and $b(d)$ values. Rate $1 / n$ codes are treated first to simplify the discussion. Define out $[s]$ and out ${ }_{1}[s]$ as the number of ones that the encoder outputs going from state $s 0=\lfloor s / 2\rfloor$ and $s 1=s 0+2^{m-1}$ into state $s$. Analogous to a state metric, the entry in row $s>0$ and column 0 of a matrix $W$, referred to as $W[s][0]$, will contain the least weight of any simple path with length $\leq T$ trellis branches. For $t=1$ to coeffs (a parameter described later), $W[s][t]$ will be the number of simple paths of weight $W[s][0]+t-1$ and length $\leq T$ branches into state $s>0$. For state $0, W[0][t]$ is always kept at 0 , except $W[0][1]=1$. The entries in a second matrix $B$ count either the total number of ones input to the encoder (when the variable len $=0$ ) or the total length in trellis branches (when len $=1$ ), of all simple paths having length $\leq T$ (again $B[0][t]=0$ always). These matrices are obtained for successive values of $T$ starting with 1 by extending, one branch length at a time (an algorithm 'step'), the code trellis starting from state 0 only. Thus the longest pathlength $(T)$ explored by the algorithm equals the number of 'steps' executed. The algorithm terminates after step $T^{*}$ when $W$ has reached values that will never change, which also forces $B$ to remain constant.

Then since $W[s][0]$ is the least weight of any simple path into state $s>0, d_{\text {free }}=W\left[2^{m-1}\right][0]+$ out $_{1}[0]$. Also,

$$
\begin{aligned}
& W\left[2^{m-1}\right]\left[d-d_{\text {free }}+1\right]=p(d) \\
& B\left[2^{m-1}\right]\left[d-d_{\text {free }}+1\right]=b(d) \quad(\ell(d) \text { if len }=1)
\end{aligned}
$$

for $d=d_{\text {free }}$ to $d_{\text {free }}+$ coeffs -1 .

Two versions of the basic algorithm above are presented in C language format in Algorithm 1 and Table 1. In Algorithm 1, matrices $P$ and $A$ store previous $W$ and $B$ entries corresponding to simple paths of length $\leq T$, which are used to compute new $W$ and $B$ matrices for length $\leq T+1$ simple paths. When change remains 0 after step $T^{*}, W$ (and thus $B$ ) will never change because $W[s][t]=P[s][t]$ for all $s$ and $t . P[s][0]$ is initialized to 999 for $s>0, P[0][1]=1$, and all other array values are initialized to 0 . If any second array index $t+$ offset is $\leq 0$ in the $W[s][t]$ and $B[s][t]$ instructions, the array referenced is simply ignored.

The algorithm requires storage for $2^{m+1}$ (coeffs +2 ) integers and the amount of work per step is proportional to this number. The number of steps executed, $\bar{T}^{*}$, equals the length, in trellis branches, of the longest fundamental path(s) having weight $d_{\text {free }}+$ coeffs -1 . The parameter coeffs should be set equal to $\lceil 10 n / k\rceil$ because using this many nonzero terms in the union bounds gives results with three significant digits of precision when the bounds are tight enough to be useful.

Setting coeffs $=0$ and ignoring offset0, offset1, $A$, and $B$ yields a simple and fast algorithm for finding $d_{\text {free }}$. About $2^{m+1}$ bytes of storage and 2.1 CPU seconds (on a computer executing 12 million instructions per second) were required to obtain $d_{\text {free }}$ for the $m=14$ Galileo and " 2 -dB" codes.

For rate $k / n$ codes with $k>1$, a state $s$ is partitioned into $s_{1}, s_{2}, \ldots, s_{k}$ where $s_{i}$ corresponds to the contents of the $i$ th encoder shift register. The output weight as the encoder enters state $s$ is out ${ }_{j}[s]$ and the input is
$j \in\left[0 \ldots 2^{k}-1\right]$. New $W[s][t]$ and $B[s][t]$ values are computed using at most $2^{k}$ entries from each of the $P$ and $A$ matrices.

For codes with $2^{k} \ll 2^{m}$ (such as rate $1 / n$ with $m>3$ ), Algorithm 1 may be improved by looping through groups of $2^{k}$ states called butterflies (see Fig. 4) instead of individual states and by computing $W[s][t]$ and $B[s][t]$ in place [7]. This reduces the storage memory required by almost one-half because the "double-buffering" matrices $P$ and $A$ are eliminated. If the entire vector $W\left[2^{m-1}\right]$ remains the same for $m$ consecutive steps, it will never change because there is a trellis path of length $\leq m$ branches between any two states. The algorithm shown in Table 1 , which incorporates these improvements, produced the first $10 n$ nonzero coeffs in the profiles of two $K=15$ codes: the rate $1 / 4$ Galileo code (Table 2) and the proposed " 2 -dB", rate $1 / 6$ code (Table 3). These profiles took only a few minutes of CPU to generate and required storage of 7 and 8.5 Mbytes, respectively, when 4 bytes were used for each integer. These memory requirements could be further reduced by storing each integer in the smallest number of bytes needed ( 1 for $W[s][0]$ to $W[s][15]$ in the $K=15$ codes above).

## Algorithm 1. A simple profile algorithm

```
do \{
        for \(\left(s=1\right.\) to \(\left.2^{m}-1\right)\{\)
            \(s 0=\lfloor s / 2\rfloor ; s 1=s 0+2^{m-1} ;\)
            bit \(=\) len \(+(1-\mathrm{len}) *(s \bmod 2)\);
            \(W[s][0]=\min \left(P[s 0][0]+\right.\) out \(_{0}[s], P[s 1][0]+\) out \(\left._{0}[s]\right) ;\)
            offset \(0=W[s][0]-P[s 0][0]-\) out \(_{0}[s] ;\)
            offset1 \(=W[s][0]-P[s 1][0]-\) out \(_{0}[s] ;\)
            for ( \(t=1\) to coeffs) \{
                \(W[s][t]=P[s 0][t+\) offset 0\(]+P[s 1][t+\) offset 1\(] ;\)
            \(B[s][t]=A[s 0][t+\) offset 0\(]+\)
                        \(A[s 1][t+\) offset 1\(]+\) bit * \(W[s][t] ;\}\)
        \(\}\) change \(=0\);
    for \(\left(s=1\right.\) to \(\left.2^{m}-1\right)\)
            for ( \(t=0\) to coeffs)
            if \((P[s][t] \neq W[s][t])\{\) change \(=1\);
                            \(P[s][t]=W[s][t] ; B[s][t]=A[s][t] ;\}\)
\} while (change \(\neq 0\) );
```


## References

[1] A. J. Viterbi, "Convolutional Codes and their Performance in Communication Systems," IEEE Trans. Comm., COM-19, pp. 751-772, October 1971.
[2] A.J. Viterbi and J.K. Omura, Principles of Digital Communication and Coding. New York: McGraw-Hill, 1979.
[3] E. Zehavi and J.K. Wolf, "On the Performance Evaluation of Trellis Codes," IEEE Trans. Info. Theory, vol. IT-33, pp. 196-202, March 1987.
[4] L. Chevillat and R. Johannesson, "A Fast Algorithm for Finding the Distance Spectrum of Convolutional Codes," IEEE Trans. Info. Theory, to appear.
[5] M. Rouanne and D.J. Costello, Jr., "An Algorithm for Computing the Distance Spectrum of Trellis Codes," IEEE Journal on Selected Areas of Comm., vol. SAC-21, August 1989.
[6] L. Van de Meeburg, "A Tightened Upper Bound on the Error Probability of Binary Convolutional Codes with Viterbi Decoding," IEEE Trans. Info. Theory, vol. IT-20, pp. 389-391, May 1974.
[7] C. M. Rader, "Memory Management in a Viterbi Decoder," IEEE Trans. Comm., COM-29, pp. 1399-1401, September 1981.
[8] P. J. Lee, "A Very Efficient Transfer Function Bounding Technique on Bit Error Rate for Viterbi Decoded, Rate 1/N Convolutional Codes," TDA Progress Report 42-79, vol. July-September, pp. 114-123, November 15, 1984.
[9] R. J. McEliece and I. M. Onyszchuk, "A Symbol Error Upper Bound for Convolutional Codes," Proc. 27th Allerton Conf., Monticello, Illinois, September 26-29, 1989.

Table 1. A profile algorithm for rate $1 / n$ codes

```
k=0; stop = 0;
do {
    for (t=0 to coeffs)
            { temp[t]=W[\mp@subsup{2}{}{m-1-k}[t];B[0][t]=W[0][t]=0; }
    W[0][1]=1;
    for (s=0 to 2m-1}-1)
            s0=s>>k; (cyclically)
            s1=s0+2m-1-k
            tw0[0] = min (W[s0][0] + out [2s],W[s1][0]+out [2s]);
            tw1[0]= min (W[s0][0]+out [2s+1],W[s1][0]+ outo[2s+1]);
            offset00 = tw0[0]-W[s0][0] - out0[2s];
            offset01 = tw0[0]-W[s1][0]-out, [2s];
            offet10=tw1[0]-W[s0][0]-out, [2s+1];
            ofset11 = tw1[0]-W}[s]|[0]-out\mp@subsup{t}{0}{}[2s+1]
            for (t = 1 to coeffs) {
                tw0[t]=W[s0][t+ offset00]+W[s1][t+offset10];
            tw1[t]=W[s0][t+offset10]+W[s1][t+offset11];
            tb0[t]=B[s0][t+offset00]+B[s1][t+offset10] + len *tw0[t];
            tb1[t] = B[s0][t+offset10]+B[s1][t+offset11]+tw1[t];
            }
            for (t }=0\mathrm{ to coeffs) {
            W[s0][t]=tw0[t];W[s1][t]=tw1[t];
            B[s0][t]=tb0[t]; B[s1][t]=tb1[t]; }
            k=k+1; if (k=m) k=0; change = 1;
            if (W[2m-1-k}][coeffs]>0) change = 0; 
            for ( }t=0\mathrm{ to coeffs)
                if (W[2m-1-k}][t]\not=\mathrm{ temp [t]) change = 1;
            if (change =1) stop = 0;
            if (change =0 and stop <m-1) {change =1; stop ++; }
} while (change }\not=0\mathrm{ );
```

Table 3. Rate $1 / 6$ " $2-\mathrm{dB}$ " code proflies

| Distance $d$ | Fundamental paths $p(d)$ | Bit errors $i(d)$ | Total lengths $\ell(d)$ |
| :---: | :---: | :---: | :---: |
| 56,57 | 1,5 | 2,15 | 3,19 |
| 58,60 | 1,3 | 2,12 | 2,14 |
| 61,62 | 5,12 | 25,56 | 35,84 |
| 63,64 | 11,5 | 43,24 | 67,40 |
| 65,66 | 8,11 | 44,62 | 68,95 |
| 67,68 | 8,11 | 48,62 | 76,98 |
| 69 | 27 | 167 | 267 |
| 70 | 30 | 162 | 277 |
| 71 | 36 | 216 | 363 |
| 72 | 54 | 366 | 573 |
| 73 | 74 | 464 | 785 |
| 74 | 89 | 610 | 998 |
| 75 | 94 | 670 | 1104 |
| 76 | 126 | 912 | 1524 |
| 77 | 163 | 1209 | 2022 |
| 78 | 226 | 1676 | 2814 |
| 79 | 290 | 2236 | 3785 |
| 80 | 369 | 2920 | 4993 |
| 81 | 493 | 4051 | 6846 |
| 82 | 574 | 4780 | 8168 |
| 83 | 767 | 6571 | 11236 |
| 84 | 979 | 8562 | 14687 |
| 85 | 1182 | 10474 | 18250 |
| 86 | 1574 | 14282 | 24860 |
| 87 | 1996 | 18516 | 32193 |
| 88 | 2618 | 24594 | 43183 |
| 89 | 3407 | 32955 | 57577 |
| 90 | 4238. | 41914 | 73499 |
| 91 | 5353 | 53757 | 94399 |
| 92 | 7006 | 71430 | 126401 |
| 93 | 8932 | 92712 | 164631 |
| 94 | 11418 | 120946 | 214330 |
| 95 | 14401 | 155175 | 275986 |
| 96 | 18467 | 202902 | 361135 |
| 97 | 24039 | 268439 | 479664 |
| 98 | 30325 | 344146 | 616671 |
| 99 | 38662 | 446878 | 800288 |
| 100 | 49690 | 583672 | 1048171. |
| 101 | 63930 | 762130 | 1371587 |
| 102 | 81742 | 990268 | 1785532 |
| 103 | 103839 | 1278325 | 2308219 |
| 104 | 133335 | 1666564 | 3012971 |
| 105 | 170357 | 2159215 | 3912282 |
| 106 | 217467 | 2801764 | 5081111 |
| 107 | 278512 | 3640320 | 6613934 |
| 108 | 356223 | 4721974 | 8592622 |
| 109 | 456347 | 6135943 | 11180051 |
| 110 | 583546 | 7956498 | 14517787 |
| 111 | 746528 | 10327464 | 18861798 |
| 112 | 954389 | 13376948 | 24468075 |
| 113 | 1220261 | 17333391 | 31748900 |
| 114 | 1562164 | 22493842 | 41242505 |
| 115 | 1997088 | 29126250 | 53477032 |



Fig. 1. A rate 1/2, 4-state encoder.


Fig. 2. State diagram of the encoder In Fig. 1.


Fig. 3. Poles of $T(D)$ for the $(7,1 / 2)$ NASA code.


Fig. 4. A rate $1 / n$ code butterfly (number $s$ ).

## Appendix

## Generating Functions for the $(7,1 / 2)$ NASA Code

Using the determinant algorithm in section IV, the complete path enumerator $T(D, L, I)$ for the $m=6$, rate $1 / 2$ NASA code was found to contain 1529 numerator and 2799 denominator trivariate terms. The code's weight enumerator, $T(D)=T(D, 1,1)$, is
$11 D^{10}-6 D^{12}-25 D^{14}+D^{16}+93 D^{18}-15 D^{20}-176 D^{22}-76 D^{24}+243 D^{26}+417 D^{28}-228 D^{30}-1156 D^{32}-49 D^{34}$ $+2795 D^{36}+611 D^{38}-5841 D^{40}-1094 D^{42}+9575 D^{44}+1097 D^{46}-11900 D^{48}-678 D^{50}+11218 D^{52}+235 D^{54}-8068 D^{56}$ $-18 D^{58}+4429 D^{60}-20 D^{62}-1838 D^{64}+8 D^{66}+562 D^{68}-D^{70}-120 D^{72}+16 D^{76}-D^{80}$
$1-4 D^{2}-6 D^{4}-30 D^{6}+40 D^{8}+85 D^{10}-81 D^{12}-345 D^{14}+262 D^{16}+844 D^{18}-403 D^{20}-1601 D^{22}+267 D^{24}+2509 D^{26}$ $+389 D^{28}-3064 D^{30}-2751 D^{32}+2807 D^{34}+8344 D^{36}-1960 D^{38}-16133 D^{40}+1184 D^{42}+21746 D^{44}-782 D^{46}-21403 D^{48}$ $+561 D^{50}+15763 D^{52}-331 D^{54}-8766 D^{56}+131 D^{58}+3662 D^{60}-30 D^{62}-1123 D^{64}+3 D^{66}+240 D^{68}-32 D^{72}+2 D^{76}$
$=11 D^{10}+38 D^{12}+193 D^{14}+1331 D^{16}+7275 D^{18}+40406 D^{20}+\cdots$
The other two generating functions used to compute error bounds, $\partial T(D, L, I) / \partial L$ and $\partial T(D, L, I) / \partial I$ at $L=I=1$, both have denominators equal to the square of $T(D)$ 's denominator above. Their numerators are, respectively,

$$
\begin{aligned}
& 121 D^{10}-387 D^{12}-706 D^{14}+1460 D^{16}+3970 D^{18}-6157 D^{20}-11643 D^{22}+8725 D^{24}+28677 D^{26}+12195 D^{28} \\
& +88 D^{30}-170654 D^{32}-306124 D^{34}+817895 D^{36}+1637616 D^{38}-2879440 D^{40}-6106837 D^{42}+8568521 D^{44}+18636083 D^{46} \\
& -22431469 D^{48}-48921504 D^{50}+52678351 D^{52}+113105887 D^{54}-112260733 D^{56}-232580537 D^{58}+217337170 D^{60} \\
& +426400859 D^{62}-379787502 D^{64}-696667758 D^{66}+592954735 D^{68}+1013294336 D^{70}-815739185 D^{72}-1311124721 D^{74} \\
& +968225450 D^{76}+1509511967 D^{78}-955561827 D^{80}-1548537967 D^{82}+721812022 D^{84}+1419185285 D^{86}-302418615 D^{88} \\
& -1166004400 D^{90}-173248817 D^{92}+861869027 D^{94}+545751792 D^{96}-574515412 D^{98}-713158178 D^{100}+345328990 D^{102} \\
& +676371119 D^{104}-186475599 D^{106}-515274530 D^{108}+89761092 D^{110}+326707300 D^{112}-38067910 D^{114}-174942675 D^{116} \\
& +14010022 D^{118}+79516060 D^{120}-4391424 D^{122}-30654965 D^{124}+1145504 D^{126}+9969013 D^{128}-241287 D^{130} \\
& -2706667 D^{132}+39344 D^{134}+603670 D^{136}-4651 D^{138}-107908 D^{140}+354 D^{142}+14883 D^{144}-13 D^{146}-1488 D^{148} \\
& +96 D^{152}-3 D^{156}
\end{aligned}
$$

and
$36 D^{10}-77 D^{12}-140 D^{14}+813 D^{16}+269 D^{18}-4414 D^{20}+321 D^{22}+14884 D^{24}-5273 D^{26}-40509 D^{28}+30344 D^{30}$ $+83884 D^{32}-177469 D^{34}-111029 D^{36}+608702 D^{38}-29527 D^{40}-1820723 D^{42}+817086 D^{44}+4951082 D^{46}-3436675 D^{48}$ $-12279246 D^{50}+10300306 D^{52}+27735007 D^{54}-25648025 D^{56}-56773811 D^{58}+55659125 D^{60}+104376199 D^{62}$ $-106695512 D^{64}-170819460 D^{66}+180836818 D^{68}+247565043 D^{70}-270555690 D^{72}-317381295 D^{74}+356994415 D^{76}$ $+360595622 D^{78}-415401723 D^{80}-364292177 D^{82}+426295756 D^{84}+328382391 D^{86}-385686727 D^{88}-264812337 D^{90}$ $+307287819 D^{92}+191225378 D^{94}-215144035 D^{96}-123515898 D^{98}+131946573 D^{100}+71124860 D^{102}-70570661 D^{104}$ $-36310569 D^{106}+32722089 D^{108}+16308558 D^{110}-13052172 D^{112}-6380604 D^{114}+4433332 D^{116}+2147565 D^{118}$ $-1265046 D^{120}-612040 D^{122}+297721 D^{124}+144665 D^{126}-56305 D^{128}-27569 D^{130}+8232 D^{132}+4066 D^{134}-874 D^{136}$ $-435 D^{138}+60 D^{140}+30 D^{142}-2 D^{144}-D^{146}$


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