## Symmetric Product Codes

# Henry D. Pfister ${ }^{1}$, Santosh Emmadi ${ }^{2}$, and Krishna Narayanan ${ }^{2}$ 

${ }^{1}$ Department of Electrical and Computer Engineering Duke University<br>${ }^{2}$ Department of Electrical and Computer Engineering<br>Texas A\&M University

Coding: From Practice to Theory Simons Institute<br>UC Berkeley

## Prologue

- Let $\mathcal{C}$ be an $(n, k, d)$ linear code over $\mathbb{F}$
- generator / parity-check matrix: $G \in \mathbb{F}^{k \times n} / H \in \mathbb{F}^{(n-k) \times n}$
- product code given by $n \times n$ arrays with rows/columns in $\mathcal{C}$ :

$$
\mathcal{P}=\left\{G^{\top} U G \mid U \in \mathbb{F}^{k \times k}\right\}
$$

- well-known that $\mathcal{P}$ is an $\left(n^{2}, k^{2}, d^{2}\right)$ linear code


## Prologue

- Let $\mathcal{C}$ be an $(n, k, d)$ linear code over $\mathbb{F}$
- generator / parity-check matrix: $G \in \mathbb{F}^{k \times n} / H \in \mathbb{F}^{(n-k) \times n}$
- product code given by $n \times n$ arrays with rows/columns in $\mathcal{C}$ :

$$
\mathcal{P}=\left\{G^{\top} U G \mid U \in \mathbb{F}^{k \times k}\right\}
$$

- well-known that $\mathcal{P}$ is an $\left(n^{2}, k^{2}, d^{2}\right)$ linear code
- Let $\mathcal{U}$ be the symmetric subcode of $\mathcal{P}$ :

$$
\mathcal{U}=\left\{X \in \mathcal{P} \mid X^{\top}=X\right\}
$$

- if char $(\mathbb{F}) \neq 2$, then $\mathcal{U}=\left\{2^{-1}\left(X^{\top}+X\right) \mid X \in \mathcal{P}\right\}$
- puncturing the lower triangle gives $\left.\binom{n+1}{2},\binom{k+1}{2},\binom{d+1}{2}\right)$ code


## Prologue (2)

Product Code


## Prologue (2)

Symmetric Subcode


## Prologue (2)

## Punctured Symmetric Subcode



## Prologue (3)

- Benefits
- for moderate $k$ and $n$, length and dimension reduced by $\sim 2$
- same component code: roughly same rate and half the length


## Prologue (3)

- Benefits
- for moderate $k$ and $n$, length and dimension reduced by $\sim 2$
- same component code: roughly same rate and half the length
- Drawbacks
- minimum distance also drops by $\sim 2$. Can one do better?


## Prologue (3)

- Benefits
- for moderate $k$ and $n$, length and dimension reduced by $\sim 2$
- same component code: roughly same rate and half the length
- Drawbacks
- minimum distance also drops by $\sim 2$. Can one do better?
- Let $\mathcal{V}$ be the anti-symmetric subcode of $\mathcal{P}$ :

$$
\mathcal{V}=\left\{X \in \mathcal{P} \mid X^{\top}=-X, \operatorname{diag}(X)=0\right\}
$$

- if $\operatorname{char}(\mathbb{F}) \neq 2$, then $\mathcal{V}=\left\{2^{-1}\left(X^{\top}-X\right) \mid X \in \mathcal{P}\right\}$
- Justesen suggested puncturing the lower triangle to get an

$$
\left(\binom{n}{2},\binom{k}{2}, D\right) \quad \text { Half-Product Code } \mathcal{H}
$$

## Prologue (4)

Product Code


## Prologue (4)

Anti-Symmetric Subcode


## Prologue (4)

Punctured Anti-Symmetric Subcode


## Outline

- Background
- Applications
- Half-Product Codes
- Symmetric Product Codes


## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding



## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Received block

## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Row decoding

## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Row decoding

## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Column decoding

## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Column decoding

## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding



## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Decoding successful

## Background

- Product Codes
- introduced by Elias in 1954
- hard-decision "cascade decoding" by Abramson in 1968
- "GLDPC" introduced by Tanner in 1981
- Example: 2-error-correcting codes, bounded distance decoding


Or trapped in a stopping set

## Applications

- Applications
- recent interest for high speed optical communication
- focus on $100 \mathrm{~Gb} / \mathrm{s}$ with $7 \%$ redundancy (i.e., $1-\frac{239}{255} \approx 0.07$ )
- high-rate generalized product codes with BCH component codes and iterative algebraic hard-decision
- many designs appeared in ITU 975.1 in 2004
- Justesen recognized the potential in 2010
- Decoding
- decoding complexity much lower than comparable LDPC codes
- for hard-decision channels, BER performance is comparable


## A Note on Decoding

- Syndrome-Based Iterative Algebraic Decoding
- Initialization
- compute and store the syndrome for each row and column
- Iteration
- run algebraic decoding on each row using syndromes
- correct errors by updating the column syndromes
- run algebraic decoding on each column using syndromes
- correct errors by updating the row syndromes
- Memory to store syndromes is $2 n(n-k)=2 n^{2}(1-R)$ vs. $n^{2}$
- $(1023,993) \mathrm{BCH}$ vs. $n=1023^{2}$ LDPC: factor 50 less memory
- Well-known trick in industry for many years...


## Symmetric Product Codes

- What are they?
- subclass of generalized product codes that use symmetry to reduce the block length while using the same component code
- one example, dubbed half-product codes (HPCs) in 2011 by Justesen, based on work by Tanner in 1981
- the minimum distance is also larger than expected
- Match the length and rate between product and HPC
- PC is $\left(n_{0}^{2}, k_{0}^{2}\right)$ and HPC is $\approx\left(n_{1}^{2} / 2, k_{1}^{2} / 2\right)$
- $n_{1} \approx \sqrt{2} n_{0}, k_{1} \approx \sqrt{2} k_{0}$, and $n_{1}-k_{1} \approx \sqrt{2}\left(n_{0}-k_{0}\right)$
- HPC component code has $n$ and $t$ larger by factor $\sqrt{2}$ !


## Minimum Distance (1)

- Support Sets and Generalized Hamming Weights
- let $\operatorname{supp}(x) \triangleq\left\{i \in[n] \mid[x]_{i} \neq 0\right\}$ denote the support set of $x$
- the 2nd generalized Hamming weight [HKY92] is

$$
\begin{aligned}
d_{2} & =\min _{\substack{x_{1}, x_{2} \in \mathcal{C} \backslash\{0\} \\
x_{1} \neq x_{2}}}\left|\operatorname{supp}\left(x_{1}\right) \cup \operatorname{supp}\left(x_{2}\right)\right| \\
& \geq\left\lceil 3 d_{\text {min }} / 2\right\rceil
\end{aligned}
$$

- measures minimal total support of two codewords
- Bound: if $d_{2}$ smaller than $\left\lceil 3 d_{\min } / 2\right\rceil$, then sum violates $d_{\text {min }}$


## Minimum Distance (2)

- Let $\mathcal{V}$ be the anti-symmetric subcode of $\mathcal{P}$


## Minimum Distance (2)

- Let $\mathcal{V}$ be the anti-symmetric subcode of $\mathcal{P}$
- For $x_{1}, x_{2} \in \mathcal{C} \backslash\{0\}$, we will show $X=x_{1}^{\top} x_{2} \notin \mathcal{V}$
- First, note $X \in \mathcal{P}$ because $H X=\left(H x_{1}^{T}\right) x_{2}=0$
- But, $\operatorname{diag}(X)=0$ for $X \in \mathcal{V}$ and, thus, $\left[x_{1}\right]_{i}\left[x_{2}\right]_{i}=0$ for all $i$
- implies $\operatorname{supp}\left(x_{1}\right) \cap \operatorname{supp}\left(x_{2}\right)=\emptyset$
- and $X_{i, j}=\left[x_{1}\right]_{i}\left[x_{2}\right]_{j} \neq 0$ implies $X_{j, i}=\left[x_{1}\right]_{j}\left[x_{2}\right]_{i}=0$
- Thus, $X^{\top} \neq-X$ and $X \notin \mathcal{V}$


## Minimum Distance (2)

- Let $\mathcal{V}$ be the anti-symmetric subcode of $\mathcal{P}$
- For $x_{1}, x_{2} \in \mathcal{C} \backslash\{0\}$, we will show $X=x_{1}^{\top} x_{2} \notin \mathcal{V}$
- First, note $X \in \mathcal{P}$ because $H X=\left(H x_{1}^{T}\right) x_{2}=0$
- But, $\operatorname{diag}(X)=0$ for $X \in \mathcal{V}$ and, thus, $\left[x_{1}\right]_{i}\left[x_{2}\right]_{i}=0$ for all $i$
- implies $\operatorname{supp}\left(x_{1}\right) \cap \operatorname{supp}\left(x_{2}\right)=\emptyset$
- and $X_{i, j}=\left[x_{1}\right]_{i}\left[x_{2}\right]_{j} \neq 0$ implies $X_{j, i}=\left[x_{1}\right]_{j}\left[x_{2}\right]_{i}=0$
- Thus, $X^{\top} \neq-X$ and $X \notin \mathcal{V}$
- Thus, no $X \in \mathcal{V}$ where n.z. rows are scalar multiples of a c.w.


## Minimum Distance (3)

- No $X \in \mathcal{V}$ where n.z. rows are scalar multiples of a c.w.
- n.z. codeword in $\mathcal{V}$ must have $\geq 2$ distinct non-zero rows
- Minimum number of n.z. columns is lower bounded by $d_{2}$
- Likewise, each column must have at least $d$ non-zero elements
- So, minimum distance of $\mathcal{V}$ must be $\geq d_{2} d \geq\lceil 3 d / 2\rceil d$
- Puncturing lower triangle gives $\mathcal{H}$
- implies $D \geq\lceil 3 d / 2\rceil d / 2$
- Or $D \geq 3 d^{2} / 4$ if $d$ even


## Minimum Distance (4)

- $\mathcal{H}$ is an $(N, K, D)$ code with $N=\binom{n}{2}, K=\binom{k}{2}$, and

$$
D \geq \begin{cases}\frac{3 d^{2}}{4} & \text { if } d \text { even } \\ \frac{(3 d+1) d}{4} & \text { if } d \bmod 4=1 \\ \frac{(3 d+1) d+2}{4} & \text { if } d \bmod 4=3\end{cases}
$$

- Also have matching upper bound if $d$ is even and there are minimum distance codewords achieving the minimum for $d_{2}$
- Basic Idea: Zeros on diagonal prevent standard square pattern codewords. Thus, support in one dimension must contain at least 2 distinct codewords. Thus, there are $d_{2}$ non-zero rows (or columns) each with weight at least $d$ and $D \geq d_{2} d$.


## Minimum Distance (5)

- Example: If $\mathcal{C}$ is an ( $8,4,4$ ) extended Hamming code
- then $d=4, d_{2}=\lceil 3 d / 2\rceil=6$, and $D \geq 12$
- there exists $x_{1}, x_{2} \in \mathcal{C}$ such that $\left|\operatorname{supp}\left(x_{1}\right) \cup \operatorname{supp}\left(x_{2}\right)\right|=6$ and $w\left(x_{1}\right)=w\left(x_{2}\right)=4$
- Half-product code is a $(28,6,12)$ binary linear code
- no $(28,6)$ binary linear code with larger $d_{\text {min }}$ exists


## Iterative Decoding Analysis (1)

- Peeling Decoder for Generalized Product Codes
- received symbols corrected sequentially without mistakes
- for the BEC and, if a genie prevents miscorrection, the BSC


## Iterative Decoding Analysis (1)

- Peeling Decoder for Generalized Product Codes
- received symbols corrected sequentially without mistakes
- for the BEC and, if a genie prevents miscorrection, the BSC
- Based on "error graph":
- vertices are code constraints
- edges connect code constraints containing same symbol
- initial observations remove fraction $1-p$ edges
- decoder peels any code constraint with $t$ or fewer errors/edges
- always reaches stopping set after finite number of iterations


## Iterative Decoding Analysis (2)

- Asymptotic Results for Half-Product Codes
- $t$-error-correcting components w/bounded distance decoding
- complete graph, edges removed i.i.d. prob. $1-p$
- Assume $n \rightarrow \infty$ with fixed $t$ and $p_{n}=\frac{\lambda}{n}$
- decoding threshold $\lambda^{*}$ via $k$-core problem in graph theory
- observed in 2007 by Justesen and Høholdt
- thresholds for $t=2,3,4$ are $\lambda^{*}=3.35,5.14,6.81$
- information about finite length via $\lambda^{*}=\lim _{n \rightarrow \infty} n p_{n}^{*}$


## Simulation Results (1)

- "Fair comparison" between product and half-product codes
- can't match both rate and block length due to numerology
- we match the rate and let the block lengths differ by $<15 \%$


## Simulation Results (1)

- "Fair comparison" between product and half-product codes
- can't match both rate and block length due to numerology
- we match the rate and let the block lengths differ by $<15 \%$
- First Example
- product code from $(170,154,5)$ shortened binary BCH code
- $\left(N^{\prime}, K^{\prime}, D^{\prime}\right)=(28900,23716,25)$, rate $\approx 0.82, s_{\min }=9$
- half-product code from $(255,231,7)$ binary BCH code
- $(N, K, D)=(32385,26565,40)$, rate $\approx 0.82, s_{\text {min }}=10$


## Simulation Results (1)

- "Fair comparison" between product and half-product codes
- can't match both rate and block length due to numerology
- we match the rate and let the block lengths differ by $<15 \%$
- First Example
- product code from $(170,154,5)$ shortened binary BCH code

$$
\text { - }\left(N^{\prime}, K^{\prime}, D^{\prime}\right)=(28900,23716,25), \text { rate } \approx 0.82, s_{\min }=9
$$

- half-product code from $(255,231,7)$ binary BCH code

$$
\text { - }(N, K, D)=(32385,26565,40), \text { rate } \approx 0.82, s_{\min }=10
$$

- Iterative decoding assuming genie to prevent miscorrection
- connection to $k$-core problem allows "threshold" estimates
- For the product code, $p^{*} \approx 3.35 / 170=0.0197$
- For the half-product code, $p^{*} \approx 5.14 / 255=0.0201$


## Simulation Results (2)



- DE predicts better HPC threshold because 5.14/3.35>3/2
- Stopping set analysis predicts better HPC error floor


## Simulation Results (3)



- product code from $(383,356,7)$ shortened binary BCH code
- $(146689,126736,49)$ code, rate $\approx 0.86, s_{\text {min }}=16$
- half-product code from $(511,475,9)$ binary BCH code
- $(130305,112575,65)$ code, rate $\approx 0.86, s_{\text {min }}=15$
- DE predicts worse HPC threshold because $6.81 / 5.14<4 / 3$


## Conclusions

- Half-product codes
- Length and dimension reduced by half with same component
- Normalized minimum distance improved by $3 / 2$
- For same blocklength and rate, one can increase $t$ by $\sqrt{2}$
- Changing $t=2$ to $t=3$ generally improves performance
- More comprehensive simulations are needed
- Symmetric product codes (see ITA 2015 paper)
- Natural extension to $m$-dimensional product codes
- Length and dimension reduced roughly by $m$ factorial
- Minimum distance improves


## Conclusions

- Half-product codes
- Length and dimension reduced by half with same component
- Normalized minimum distance improved by $3 / 2$
- For same blocklength and rate, one can increase $t$ by $\sqrt{2}$
- Changing $t=2$ to $t=3$ generally improves performance
- More comprehensive simulations are needed
- Symmetric product codes (see ITA 2015 paper)
- Natural extension to $m$-dimensional product codes
- Length and dimension reduced roughly by $m$ factorial
- Minimum distance improves
- By how much is an open problem...

