

Modern Digital Satellite Television: How It Works

Matthew C. Valenti

Lane Department of Computer Science and Electrical Engineering
West Virginia University
U.S.A.

Nov, 1, 2011

Outline

- 1 Satellite Television Standards
- 2 DVB-S2 Modulation
- 3 LDPC Coding
- 4 Constellation Shaping
- 5 Conclusion

Outline

- 1 Satellite Television Standards
- 2 DVB-S2 Modulation
- 3 LDPC Coding
- 4 Constellation Shaping
- 5 Conclusion

Digital Satellite Television in the United States

DirecTV

- Spinoff of Hughes Network Systems.
- Began operations in 1994.
- 19.2 million U.S. subscribers at end of 2010.
- 23,000 employees in U.S. and Latin America.
- \$ 33.6 billion market cap.

Dish Network.

- Spinoff of EchoStar.
- Began operations in 1996.
- 14.1 subscribers in 2010.
- 22,000 employees.
- \$10.9 Billion market cap.

The DVB Family



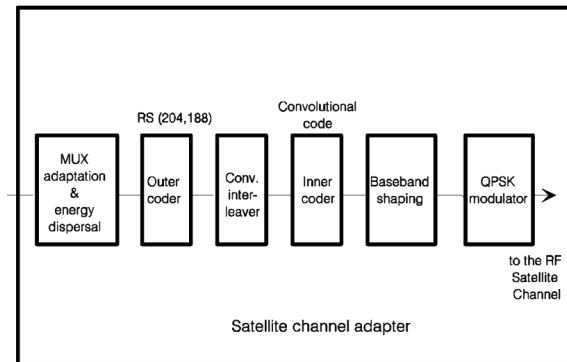
DVB is a family of open standards for digital video broadcasting.

- Maintained by 270-member industry consortium.
- Published by ETSI.

Modes of transmission

- Satellite: DVB-S, DVB-S2, and DVB-SH
- Cable: DVB-C, DVB-C2
- Terrestrial: DVB-T, DVB-T2, DVB-H

DVB-S



- Modulation: QPSK with $\alpha = 0.35$ rolloff.
- Channel coding: Concatenated Reed Solomon and convolutional.

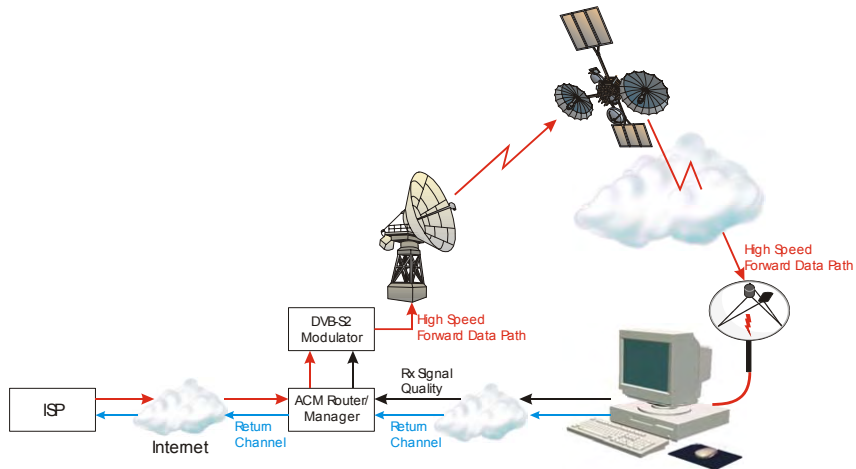
DVB-S2

DVB-S2 was introduced in 2003 with the following goals:

- Improve spectral efficiency by 30% through better modulation and coding.
 - Modulation: QPSK, 8PSK, 16/32 APSK.
 - Channel coding: LDPC with outer BCH code.
- Offer a more diverse range of services.
 - HDTV broadcast television.
 - Backhaul applications, e.g., electronic news gathering.
 - Internet downlink access.
 - Large-scale data content distribution, e.g., electronic newspapers.

Ratified 2005.

Adaptive Internet Downlink

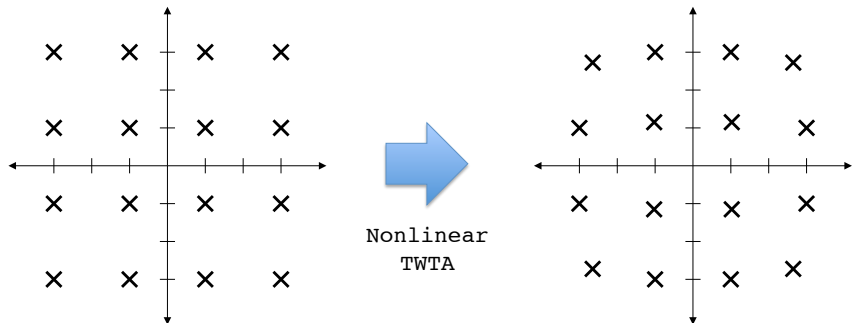


Outline

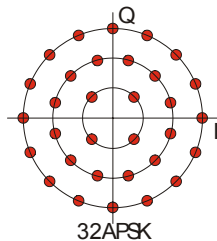
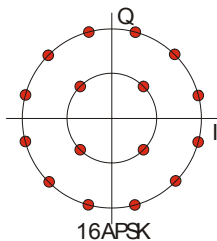
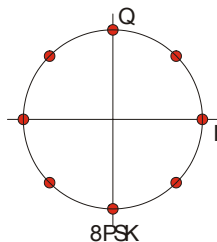
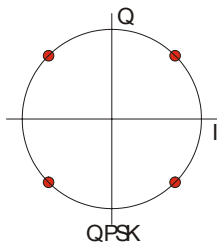
- 1 Satellite Television Standards
- 2 DVB-S2 Modulation**
- 3 LDPC Coding
- 4 Constellation Shaping
- 5 Conclusion

Why Not Use QAM?

- Higher spectral-efficiencies require larger signal constellations.
- However, nonlinear satellite channels are not well suited to square QAM.



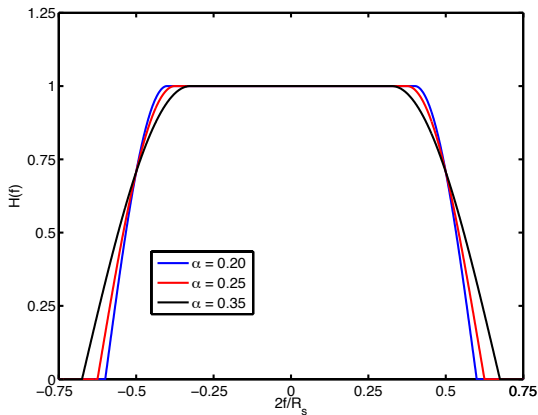
The DVB-S2 Signal Constellations



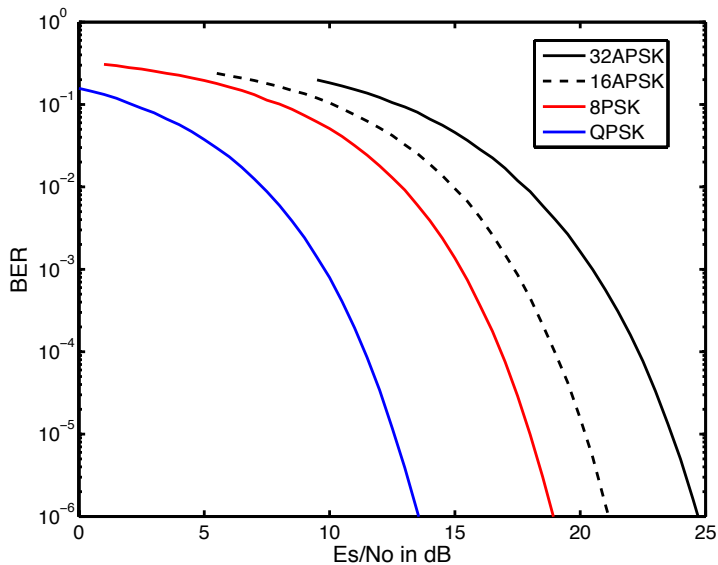
Raised-Cosine Rolloff Filtering

DVB-S2 uses a tighter root RC-rolloff filter.

- $B = R_s(1 + \alpha)$
- Assuming a 6 MHz transponder channel...
- DVB-S Example:
 - $\alpha = 0.35$.
 - QPSK: $R_b = 2R_s$
 - $2(6)/(1.35) = 8.9$ Mbps
- DVB-S2 Example:
 - $\alpha = 0.20$.
 - 32-APSK: $R_b = 5R_s$
 - $5(6)/(1.2) = 25$ Mbps



BER in AWGN

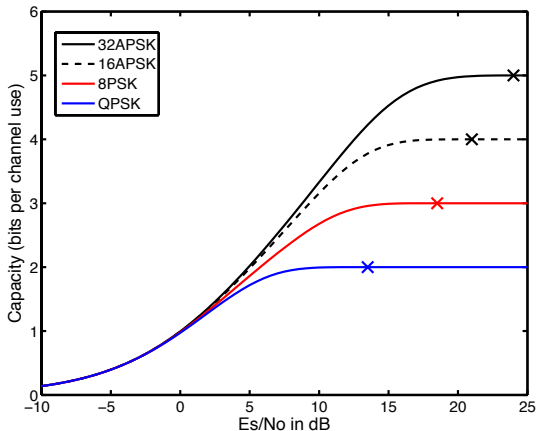


Outline

- 1 Satellite Television Standards
- 2 DVB-S2 Modulation
- 3 LDPC Coding**
- 4 Constellation Shaping
- 5 Conclusion

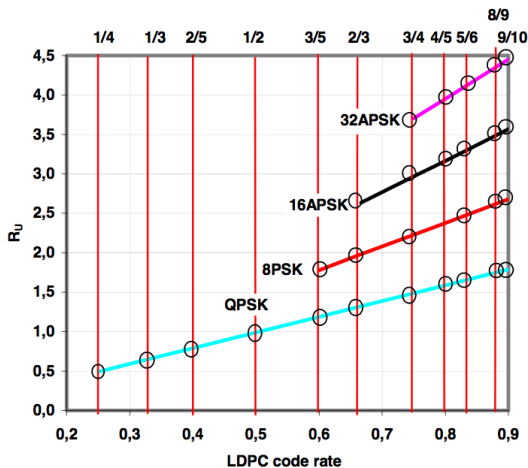
Capacity

- Performance can be improved by using error control coding.
- Gains are limited by the modulation-constrained capacity.
- LDPC codes are capable of approaching capacity.

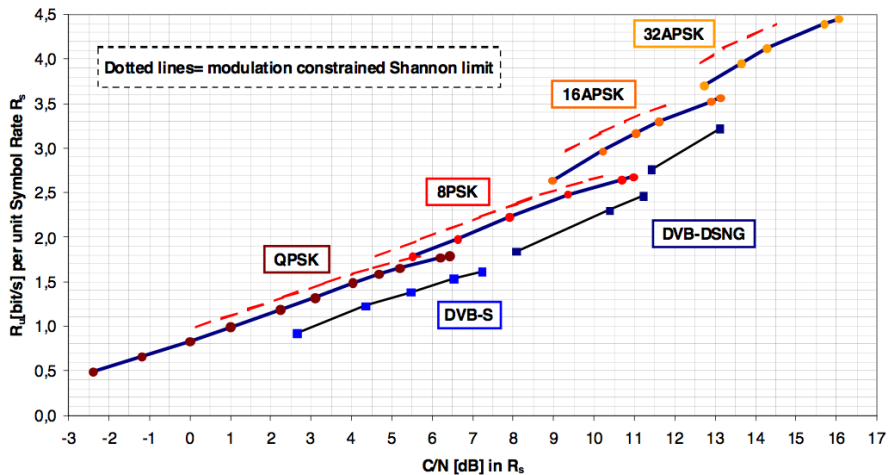


Available Code Rates

- The encoder maps length- k messages to length- n codewords.
- The code rate is $R = k/n$.
- Useful bit rate is $R_u = R \log_2(M)$.
- Two codeword lengths:
 - 16, 200.
 - 64, 800.



DVB-S2 vs. Shannon



Single Parity-Check Codes

- Consider the following rate $R = 5/6$ single parity-check code:

$$\mathbf{c} = \left[\underbrace{1 \ 0 \ 1 \ 0 \ 1}_{\mathbf{u}} \quad \underbrace{1}_{\text{parity bit}} \right]$$

- One error in *any* position may be detected:

$$\mathbf{c} = [1 \ 0 \ X \ 0 \ 1 \ 1]$$

- Problem with using an SPC is that it can only detect a single error.

Product Codes

- Place data into a k by k rectangular array.
 - Encode each row with a SPC.
 - Encode each column with a SPC.
 - Result is a rate $R = k^2/(k+1)^2$ code.
- Example $k = 2$.

$c_1 = u_1$	$c_2 = u_2$	$c_3 = c_1 \oplus c_2$
$c_4 = u_3$	$c_5 = u_4$	$c_6 = c_4 \oplus c_5$
$c_7 = c_1 \oplus c_4$	$c_8 = c_2 \oplus c_5$	$c_9 = c_3 \oplus c_6$

 $=$

1	0	1
1	1	0
0	1	1

- A single error can be corrected by detecting its row and column location

1	0	1
0	1	0
0	1	1

 \Rightarrow

1	0	1
1	1	0
0	1	1

Linear Codes

$c_1 = u_1$	$c_2 = u_2$	$c_3 = c_1 \oplus c_2$
$c_4 = u_3$	$c_5 = u_4$	$c_6 = c_4 \oplus c_5$
$c_7 = c_1 \oplus c_4$	$c_8 = c_2 \oplus c_5$	$c_9 = c_3 \oplus c_6$

- The example product code is characterized by the set of five linearly-independent equations:

$$c_3 = c_1 \oplus c_2 \Rightarrow c_1 \oplus c_2 \oplus c_3 = 0$$

$$c_6 = c_4 \oplus c_5 \Rightarrow c_4 \oplus c_5 \oplus c_6 = 0$$

$$c_7 = c_1 \oplus c_4 \Rightarrow c_1 \oplus c_4 \oplus c_7 = 0$$

$$c_8 = c_2 \oplus c_5 \Rightarrow c_2 \oplus c_5 \oplus c_8 = 0$$

$$c_9 = c_3 \oplus c_6 \Rightarrow c_3 \oplus c_6 \oplus c_9 = 0$$

- In general, it takes $(n - k)$ linearly-independent equations to specify a *linear* code.

Parity-check Matrices

- The system of equations may be expressed in matrix form as:

$$\mathbf{c}H^T = \mathbf{0}$$

where H is a *parity-check* matrix.

- Example:

$$\begin{array}{rcl}
 c_1 \oplus c_2 \oplus c_3 & = & 0 \\
 c_4 \oplus c_5 \oplus c_6 & = & 0 \\
 c_1 \oplus c_4 \oplus c_7 & = & 0 \\
 c_2 \oplus c_4 \oplus c_8 & = & 0 \\
 c_3 \oplus c_6 \oplus c_9 & = & 0
 \end{array}
 \Leftrightarrow
 H =
 \begin{bmatrix}
 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\
 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\
 0 & 1 & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\
 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1
 \end{bmatrix}$$

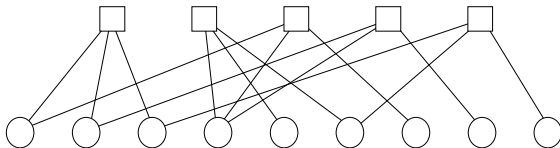
System of equations Parity-check matrix

Tanner Graphs

- The parity-check matrix may be represented by a *Tanner* graph.
- Bipartite graph:
 - Check nodes: Represent the $n - k$ parity-check equations.
 - Variable nodes: Represent the n code bits.
- If $H_{i,j} = 1$, then i^{th} check node is connected to j^{th} variable node.
- Example: For the parity-check matrix:

$$H = \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$

The Tanner Graph is:



LDPC Codes

- Observations:
 - To achieve capacity, a long code is needed.
 - The decoder's complexity depends on the number of edges in the Tanner graph.
 - The number of edges is equal to the number of zeros in H .
 - It is desirable to have a code that is long, yet has a sparse H .
- Low-density parity-check codes:
 - An LDPC code is characterized by a *sparse* parity-check matrix.
 - The row/column weights are independent of length.
 - Decoder complexity grows only linearly with block length.
- Historical note:
 - LDPC codes were the subject of Robert Gallager's 1960 dissertation.
 - Were forgotten because the decoder could not be implemented.
 - Were "rediscovered" in the mid-1990's after turbo codes were developed.

Example LDPC Code

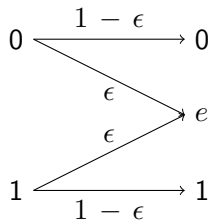
- A code from MacKay and Neal (1996):

$$\mathbf{H} = \left[\begin{array}{cccc|cccc|cccc} 1 & & & & & 1 & & 1 & & & 1 & & & \\ & 1 & & & & & & & & & & 1 & & \\ & & 1 & & & 1 & & & & & & & 1 & \\ & & & 1 & & & & 1 & & & & & & 1 \\ & & & & 1 & & & & & & & & 1 & \\ & & & & & 1 & & 1 & & & & & & \\ & & 1 & & & & & & & 1 & & 1 & & \\ & & & 1 & & 1 & & & & & 1 & & & \\ & 1 & & & 1 & & & & & & 1 & & & \\ & & 1 & & & & & 1 & & & & 1 & & \\ & & & 1 & 1 & & & & 1 & & & & 1 & \end{array} \right]$$

- The code is *regular* because:
 - The rows have constant weight (check-nodes constant degree).
 - The columns have constant weight (variable-nodes constant degree).
- This is called a $(3, 4)$ *regular* LDPC code because the variable nodes have degree 3 and the check nodes have degree 4.

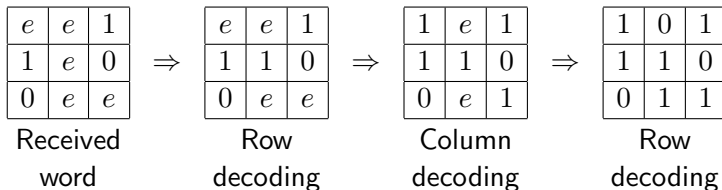
The Binary Erasure Channel

- The binary erasure channel is a conceptual model used to explain the operation of LDPC codes.
- The BEC has two inputs (data 0 and data 1) and three outputs (data 0, data 1, and *erasure* e).
- A bit is erased with probability ϵ .
- A bit is correctly received with probability $1 - \epsilon$.



Erasures Decoding of Product Codes

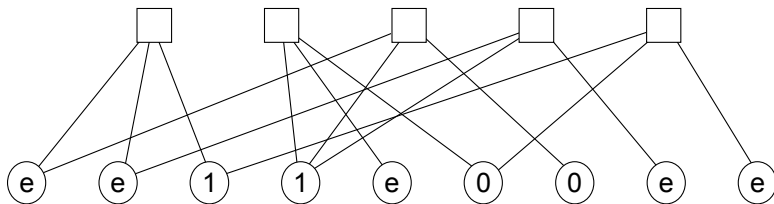
- Several erasures may be corrected by iteratively decoding the SPC on each row and column.



Erasures Decoding on the Tanner Graph

Decoding can be performed on the Tanner graph.

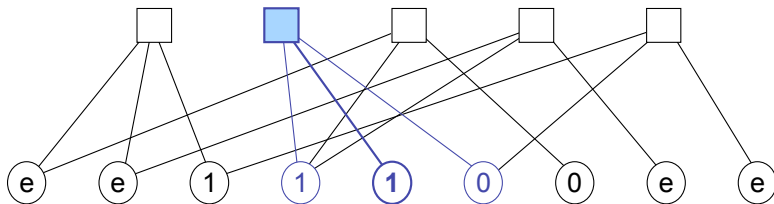
- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



Erasures Decoding on the Tanner Graph

Decoding can be performed on the Tanner graph.

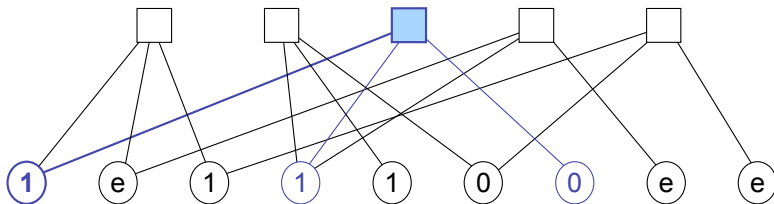
- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



Erasures Decoding on the Tanner Graph

Decoding can be performed on the Tanner graph.

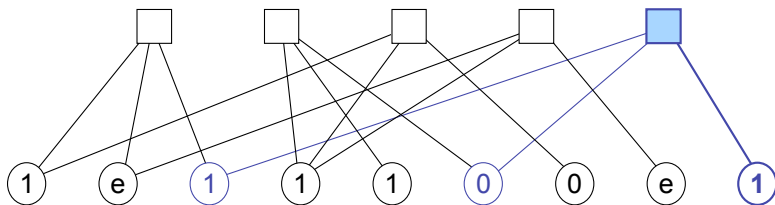
- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



Erasures Decoding on the Tanner Graph

Decoding can be performed on the Tanner graph.

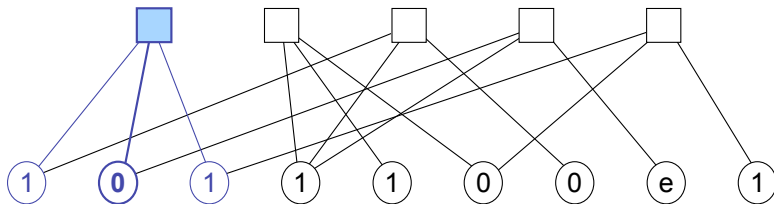
- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



Erasures Decoding on the Tanner Graph

Decoding can be performed on the Tanner graph.

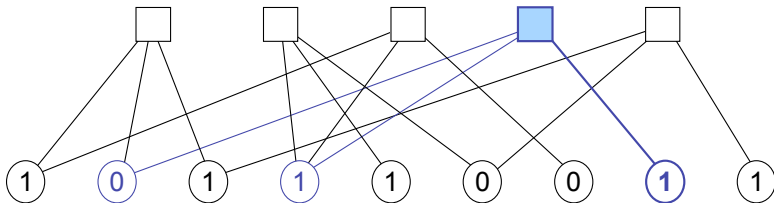
- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



Erasures Decoding on the Tanner Graph

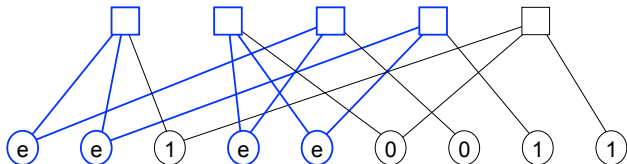
Decoding can be performed on the Tanner graph.

- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



Stopping sets

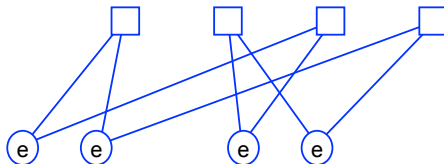
- A *stopping set* \mathcal{V} is a set of erased variable nodes that cannot be corrected, regardless of the state of the other variable nodes.



- Let \mathcal{G} be the neighbors of \mathcal{V} .
- Every check node in \mathcal{G} touches at least two variable nodes in \mathcal{V} .
- The *minimum* stopping set \mathcal{V}_{min} is the stopping set containing the fewest variable nodes.
- Let $d_{min} = |\mathcal{V}_{min}|$ be the size of the minimum stopping set.
 - There exists at least one pattern of d_{min} erasures that cannot be corrected.
 - The erasure correcting capability of the code is $d_{min} - 1$, which is the maximum number of erasures that can always be corrected.

Stopping sets

- A *stopping set* \mathcal{V} is a set of erased variable nodes that cannot be corrected, regardless of the state of the other variable nodes.



- Let \mathcal{G} be the neighbors of \mathcal{V} .
- Every check node in \mathcal{G} touches at least two variable nodes in \mathcal{V} .
- The *minimum* stopping set \mathcal{V}_{min} is the stopping set containing the fewest variable nodes.
- Let $d_{min} = |\mathcal{V}_{min}|$ be the size of the minimum stopping set.
 - There exists at least one pattern of d_{min} erasures that cannot be corrected.
 - The erasure correcting capability of the code is $d_{min} - 1$, which is the maximum number of erasures that can always be corrected.

Density Evolution

- For a (d_v, d_c) regular code, the probability that a variable-node remains erased after the ℓ^{th} iteration is

$$\epsilon_\ell = \epsilon_0 \left(1 - (1 - \epsilon_{\ell-1})^{d_c-1}\right)^{d_v-1} \quad (1)$$

where d_v is the variable-node degree, d_c is the check-node degree, and the initial condition is $\epsilon_0 = \epsilon$.

- The above result assumes independent messages, which is achieved when the girth of the Tanner graph is sufficiently large.
- If $\epsilon_\ell \rightarrow 0$ as $\ell \rightarrow \infty$ for a particular channel erasure probability ϵ , then a code drawn from the ensemble of all such (d_v, d_c) regular LDPC codes will be able to correctly decode.
- The *threshold* ϵ^* is the maximum ϵ for which $\epsilon_\ell \rightarrow 0$ as $\ell \rightarrow \infty$.
- For the $(3, 6)$ regular code, the threshold is $\epsilon^* = 0.4294$

Proof of (1), Part I/II

- Decoding involves the exchange of messages between variable nodes and check nodes.
 - Let p_{\uparrow} denote the probability of an erased message going *up* from the variable nodes to the check nodes.
 - Let p_{\downarrow} denote the probability of an erased message going *down* from the check nodes to the variable nodes.
- Consider the degree d_c check node.
 - An outgoing message sent over a particular edge is a function of the incoming messages arriving over the other $d_c - 1$ edges.
 - For the outgoing message to be correct, all $d_c - 1$ incoming messages must be correct.
 - The outgoing message will be an erasure if any of the $d_c - 1$ incoming messages is an erasure.
 - The probability of the check node sending an erasure is:

$$p_{\downarrow} = 1 - (1 - p_{\uparrow})^{d_c - 1} \quad (2)$$

Proof of (1), Part II/II

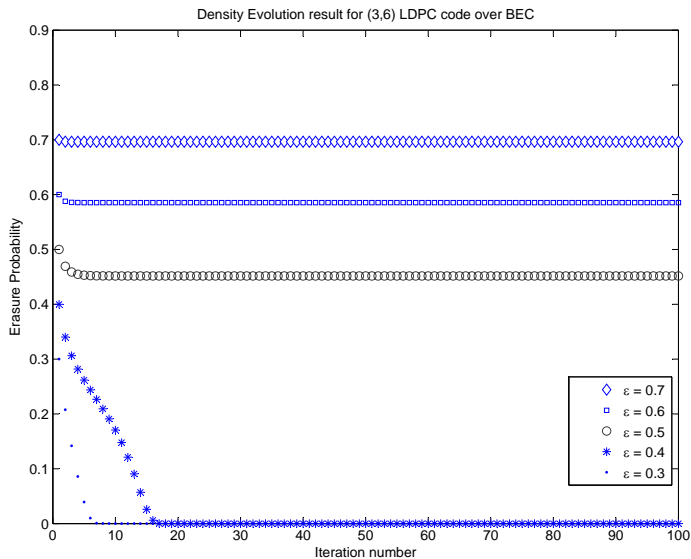
- Consider the degree d_v check node.
 - An outgoing message sent over a particular edge is a function of the incoming messages arriving over the other $d_v - 1$ edges.
 - An outgoing message will be an erasure if the variable node was initially erased *and* all of the arriving messages are erasures.
 - The probability of the variable node sending an erasure is:

$$p_{\uparrow} = \epsilon_0 p_{\downarrow}^{d_v-1} \quad (3)$$

- Letting ϵ_{ℓ} equal the value of p_{\uparrow} after the ℓ^{th} iteration, and substituting (2) into (3) yields the recursion given by (1):

$$\epsilon_{\ell} = \epsilon_0 \left(1 - (1 - \epsilon_{\ell-1})^{d_c-1} \right)^{d_v-1}$$

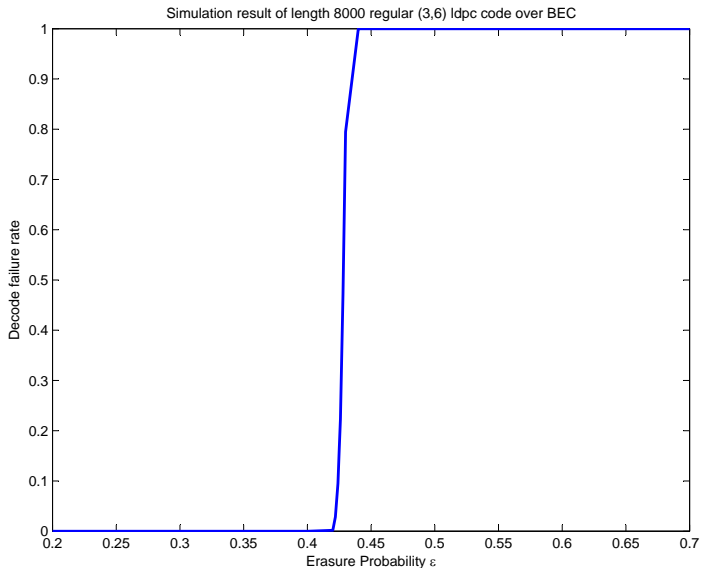
DE Example



Code Realization

- Density evolution only describes the asymptotic performance of the *ensemble* of LDPC codes.
- Implementation requires that an H matrix be generated by drawing from the ensemble of all (d_v, d_c) LDPC codes.
- Goals of good H design:
 - High girth.
 - Full rank.
 - Large minimum stopping set.
- If the girth is too low, the short cycles invalidate the iterative decoder.
- High girth achieved through girth conditioning algorithms such as progressive edge growth (PEG).
- If H is not full rank, then the rate will be reduced according to the number of dependent equations.
- Small stopping sets give rise to an *error floor*.
- A database of good regular LDPC codes can be found on MacKay's website.

Performance of an Actual Code

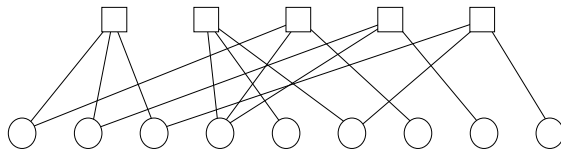


Irregular LDPC Codes

- Although regular LDPC codes perform well, they are not capable of achieving capacity.
- Properly designed irregular LDPC codes are capable of achieving capacity.
 - The degree distribution of the variable nodes is not constant.
 - The check-node degrees are often still constant (or close to it).
 - Here “designing” means picking the proper degree distribution.

Degree Distribution

- Edge-perspective degree distributions:
 - ρ_i is the fraction of *edges* touching degree i check nodes.
 - λ_i is the fraction of *edges* touching degree i variable nodes.
- For example, consider the Tanner graph:



- 15 edges.
- All are connected to degree-3 check nodes, so $\rho_3 = 15/15 = 1$.
- Four are connected to degree-1 variable nodes, so $\lambda_1 = 4/15$.
- Eight are connected to degree-2 variable nodes, so $\lambda_2 = 8/15$.
- Three are connected to the degree-3 variable node, so $\lambda_3 = 3/15$.

DE for Irregular LDPC

- The degree distributions are described in polynomial form:
 - $\rho(x) = \sum_i \rho_i x^{i-1}$ for check nodes.
 - $\lambda(x) = \sum_i \lambda_i x^{i-1}$ for variable nodes.
- For an irregular code, the probability that a variable-node remains erased after the ℓ^{th} iteration is

$$\epsilon_\ell = \epsilon_0 \lambda(1 - \rho(1 - \epsilon_{\ell-1}))$$

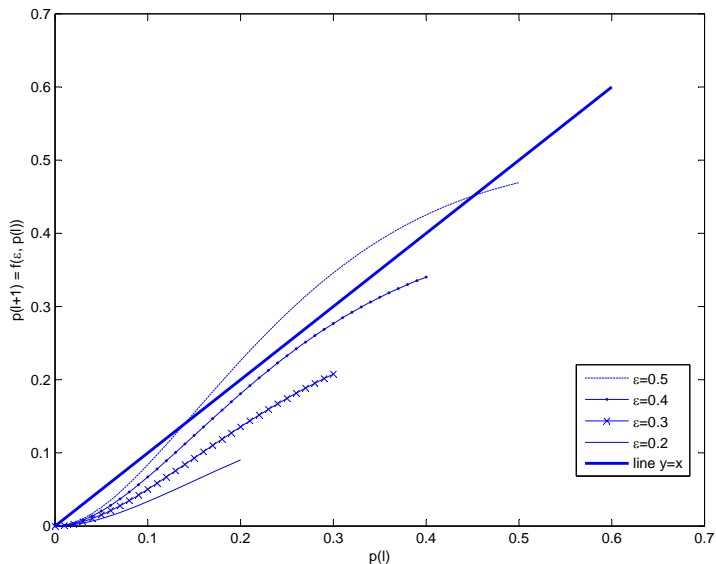
The proof follows from the Theorem on Total Probability.

- Convergence:
 - Error-free decoding requires that the erasure probability goes down from one iteration to the next.
 - Define the related function:

$$f(\epsilon, x) = \epsilon \lambda(1 - \rho(1 - x))$$

- Error-free decoding is possible iff $f(\epsilon, x) \leq x$ for all $0 \leq x \leq \epsilon$.

Convergence



Optimization

- The threshold is

$$\epsilon^* = \sup\{\epsilon : f(\epsilon, x) < x, \forall x, 0 < x \leq \epsilon\}$$

- Solving $f(\epsilon, x) = x$ for ϵ

$$\begin{aligned}x &= f(\epsilon, x) \\&= \epsilon \lambda (1 - \rho(1 - x)) \\ \epsilon &= \frac{x}{\lambda(1 - \rho(1 - x))}\end{aligned}$$

which is a function of x , and henceforth expressed as $\epsilon(x)$.

- This allows the threshold to be rewritten as:

$$\epsilon^* = \min\{\epsilon(x) : \epsilon(x) \geq x\}$$

Optimization with Linear Programming

- Our goal is to find the degree distribution which yields maximum threshold

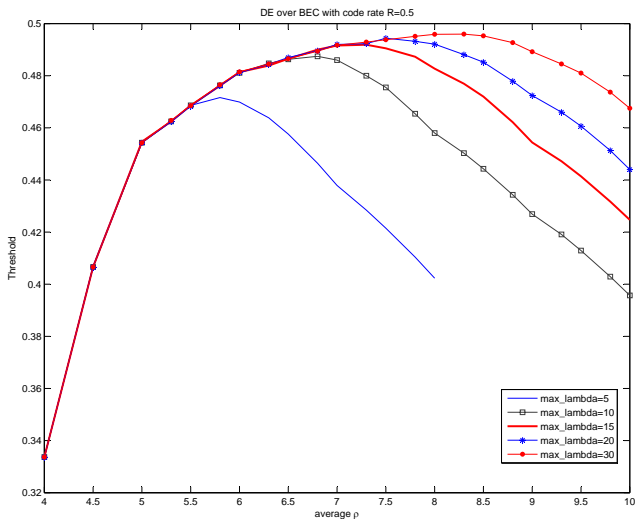
$$\max_{\varepsilon^*} \{ \varepsilon^* = \min(\varepsilon(x) : \varepsilon(x) \geq x) \};$$

- Several Constraints

$$\frac{\int_0^1 \rho(x) dx}{\int_0^1 \lambda(x) dx} = 1 - R$$
$$\sum_{i \geq 2} \lambda_i = 1; \sum_{i \geq 2} \rho_i = 1;$$
$$x \in [0, 1]$$

- Which can be modeled as a optimization problem using linear programming
 - Can use Matlab's Optimization Toolbox.

Optimization Results ($\epsilon^* = 0.49596$)



Encoding LDPC Codes

Encoding of LDPC codes is not necessarily straightforward.

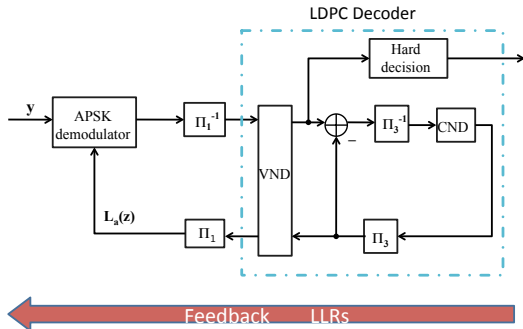
- “Systematic-form” H
 - Using Gaussian elimination, find $H = [P \ I]$.
 - Then $\mathbf{c} = \mathbf{u}G$ where $G = [I \ P^T]$.
 - However, P is likely to be high-density (complex encoding).
- Back-substitution.
 - If H is in an appropriate form, then \mathbf{c} can be encoded using *back substitution*
 - Example, $\mathbf{c}H^T = \mathbf{0}$, where

$$H = \begin{bmatrix} 1 & 0 & 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 \end{bmatrix}$$

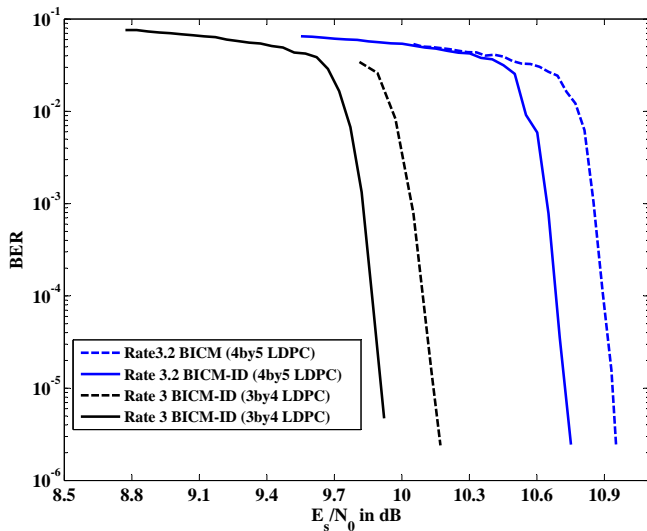
- The LDPC code in the DVB-S2 standard allows for back substitution.

Iterative Demodulation and Decoding

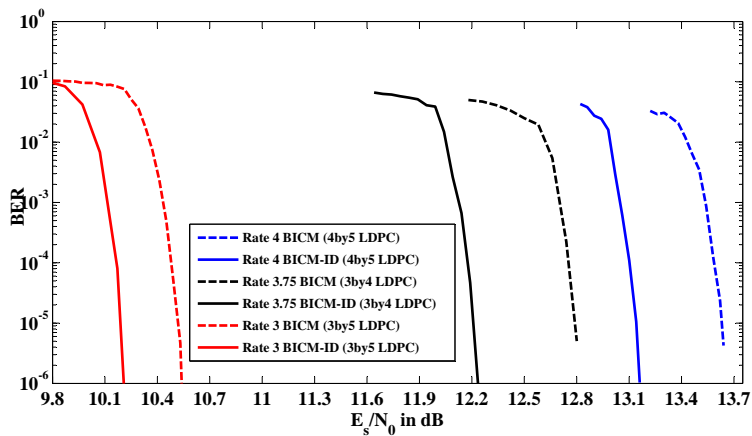
- Conventional receivers first demodulation, then decode.
- Performance is improved by iterating between the demodulator and decoder.
- BICM-ID: bit-interleaved modulation with iterative decoding.



AWGN Performance of 16APSK with BICM-ID



AWGN Performance of 32APSK with BICM-ID



Outline

- 1 Satellite Television Standards
- 2 DVB-S2 Modulation
- 3 LDPC Coding
- 4 Constellation Shaping**
- 5 Conclusion

Constellation Shaping

- Capacity curves assume equiprobable signaling.
- It is possible to increase capacity by transmitting higher-energy signals less frequently than lower-energy signals.

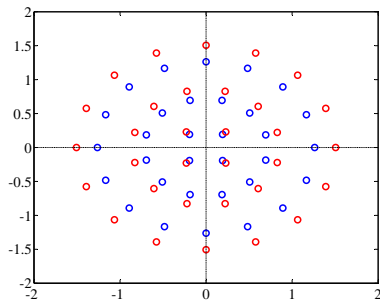


Figure: Uniform 32APSK \circ vs. shaped 32APSK \circ . Both constellations have the same energy.

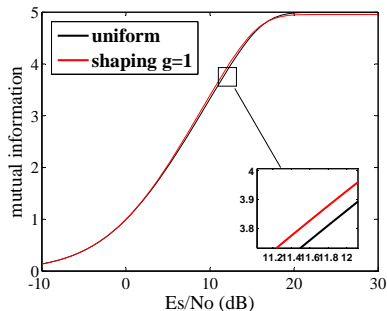


Figure: The capacity of shaped 32APSK is about 0.3 dB better than uniform 32APSK.

Sub-constellations

- The 32APSK is partitioned into two equal-sized sub-constellations.
- A *shaping bit* selects the sub-constellation, while the remaining bits select a symbol from the chosen sub-constellation.
- The lower-energy sub-constellation is selected more frequently.

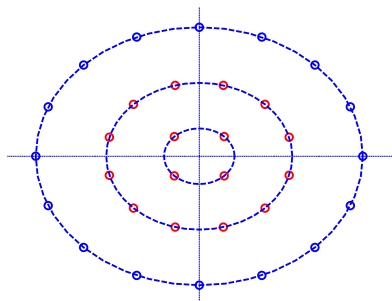


Figure: 32APSK w/ 2 sub-constellations

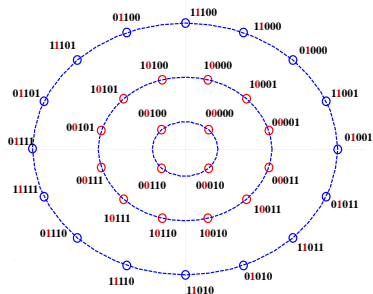


Figure: 32APSK symbol-labeling map

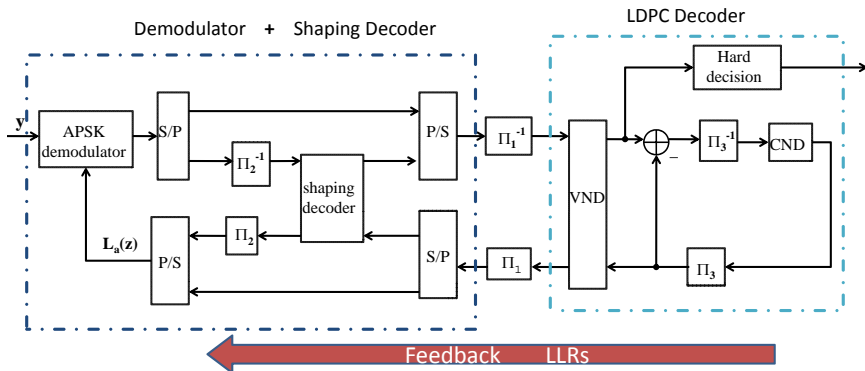
Shaping Encoder

- The shaping encoder should produce more zeros than ones.
- Example: $(n_s, k_s) = (5, 3)$

3 input data bits	5 output codeword bits
0 0 0	0 0 0 0 0
0 0 1	0 0 0 0 1
0 1 0	0 0 0 1 0
0 1 1	0 0 1 0 0
1 0 0	0 1 0 0 0
1 0 1	1 0 0 0 0
1 1 0	0 0 0 1 1
1 1 1	1 0 1 0 0

- $p_0 = 31/40$: fraction of zeros.
- $p_1 = 9/40$: fraction of ones.

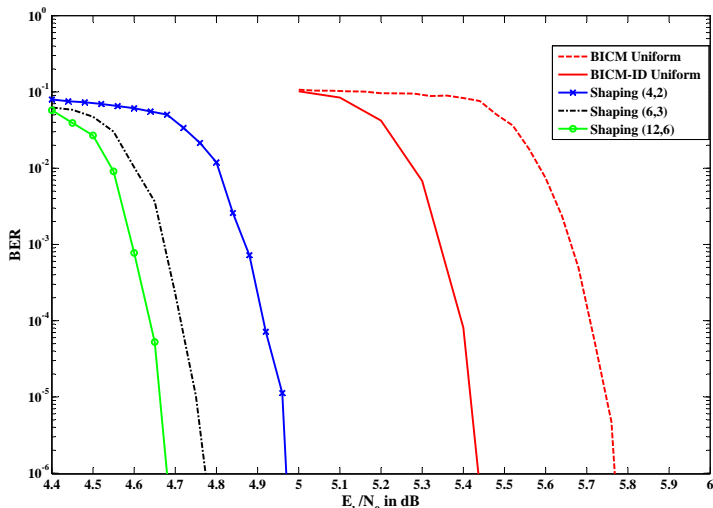
Receiver Implementation



- Additional complexity relative to BICM-ID due to shaping decoder.
- MAP shaping decoder compares against all 2^{k_s} shaping codewords.

BER of Shaping in AWGN

BER of 32-APSK in AWGN at rate $R=3$ bits/symbol



Outline

- 1 Satellite Television Standards
- 2 DVB-S2 Modulation
- 3 LDPC Coding
- 4 Constellation Shaping
- 5 Conclusion**

Conclusion

- DVB-S2 is a highly efficient system, thanks to
 - APSK modulation.
 - Tight RC-rolloff filtering.
 - Capacity-approaching LDPC codes.
- The performance of DVB-S2 can be improved by
 - BICM-ID.
 - Constellation shaping.
- Future work:
 - Using density evolution to optimize degree distribution of LDPC-coded APSK with shaping.
 - Extension to 64APSK and beyond.

Thank You.