Chapter 5 Convolutional Codes and Trellis Coded Modulation



- 5.1 Encoder Structure and Trellis Representation
- 5.2 Systematic Convolutional Codes
- 5.3 Viterbi Decoding
- 5.4 Soft-Decision Viterbi Decoding
- 5.5 BCJR Decoding
- 5.6 Trellis Coded Modulation

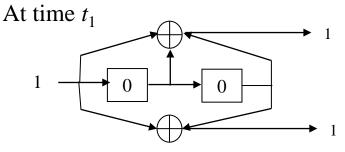


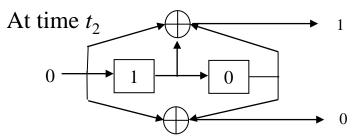
Introduction

- Encoder: contains memory (order m: m memory units);
- Output: encoder output at time unit t depends on the input and the memory units status at time unit t;
- By increasing the memory order m, one can increase the convolutional code's minimum distance (d_{\min}) and achieve low bit error rate performance (P_b) ;
- Decoding Methods:
 - Viterbi algorithm [1]: Maximum Likelihood (ML) decoding algorithm;
 - Bahl, Cocke, Jelinek, and Raviv (BCJR) [2] algorithm: Maximum *A Posteriori* Probability (MAP) decoding algorithm, used for iterative decoding process, e.g. turbo decoding.
- [1] A. J. Viterbi, "Error bounds for convolutional codes and an asymptotically optimum decoding algorithm," IEEE Trans. Inform. Theory, IT-13, 260-269, April, 1967.
- [2] L. R. Bahl, J. Cocke, F. Jelinek, and J. Raviv, "Optimal decoding of linear codes for minimizing symbol error rate," IEEE Trans, Inform. Theory, IT-20; 284-287, March, 1974.



- The $(7, 5)_8$ conv. code
- Encoder structure: c_1 Input u S_0 S_1
- Encoding Process: (Initialized state $S_0S_1 = 00$)





Code rate: ½;

Memory: m = 2;

Constraint length: m + 1 = 3;

Encoding:

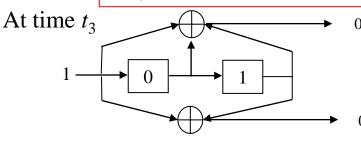
$$c_1 = u \oplus S_0 \oplus S_1;$$

$$c_2 = u \oplus S_1$$
;

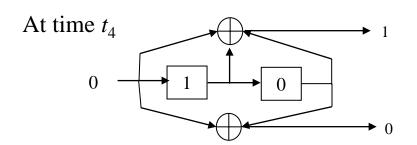
Registers update:

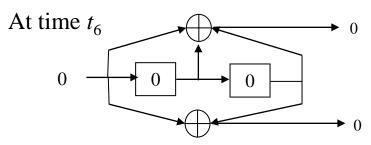
$$S_1' = S_0.$$

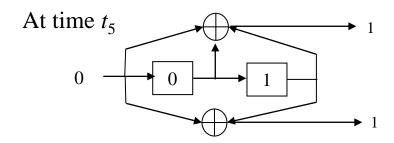
$$S_0' = u$$
.









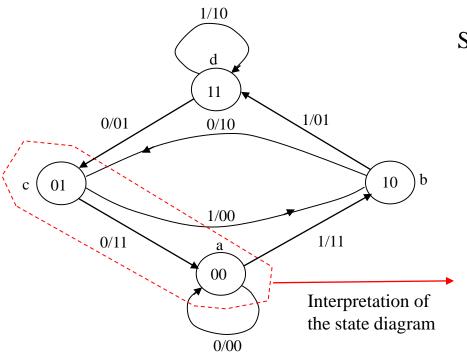


Input sequence $[u_1 \ u_2 \ u_3 \ u_4 \ u_5 \ u_6] = [1 \ 0 \ 1 \ 0 \ 0]$

Output sequence $\begin{bmatrix} c_1^1 c_1^2 & c_2^1 c_2^2 & c_3^1 c_3^2 & c_4^1 c_4^2 & c_5^1 c_5^2 & c_6^1 c_6^2 \end{bmatrix} = \begin{bmatrix} 11 & 10 & 00 & 10 & 11 & 00 \end{bmatrix}$



A state transition diagram of the $(7, 5)_8$ conv. code



State definition $(S_0 S_1)$

$$a = 00$$

$$b = 10$$

$$c = 01$$

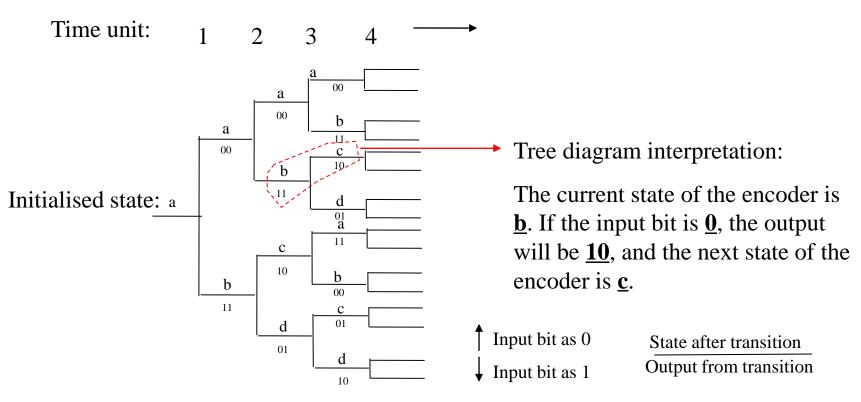
$$d = 11$$

01

Input bit (0) / output bits (11)The current state of the encoder is $\underline{\mathbf{c}}$. If the input bit is $\underline{\mathbf{0}}$, it will output $\underline{\mathbf{11}}$ and the next state of the encoder is \mathbf{a} .



Tree Representation of the $(7, 5)_8$ conv. code



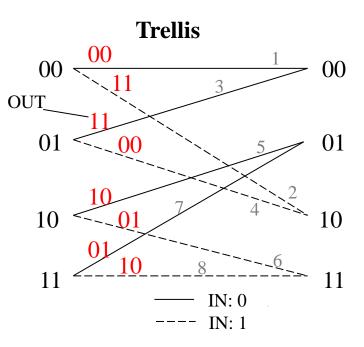
Example 5.1 Determine the codeword that corresponds to message [0 1 1 0 1]



Trellis of the $(7, 5)_8$ conv. code

State Table

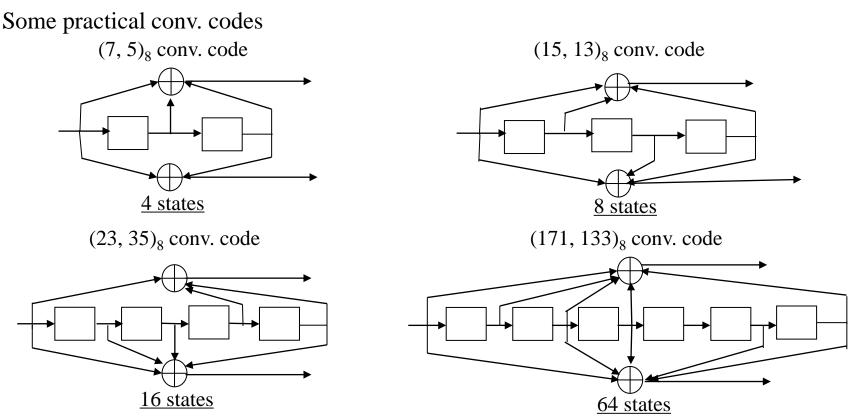
IN	Current State	Next State	Out	ID
0	00	00	00	1
1	00	10	11	2
0	01	00	11	3
1	01	10	00	4
0	10	01	10	5
1	10	11	01	6
0	11	01	01	7
1	11	11	10	8



Remark: A trellis tells the state transition and IN/OUT relationship. It can be used to yield a convolutional codeword of a sequential input.

Example 5.2 Use the above trellis to determine the codeword that corresponds to message [0 1 1 0 1].





Remark: A convolutional code's error-correction capability improves by increasing the number of the encoder states.

Remark: m tailing bits are needed to force the encoder back to the all-zero state. For the above nonsystematic conv. codes, m 0s are needed.



The encoder structure can also be represented by generator sequences or transfer functions.

Example 5.3

- The $(7,5)_8$ conv. code can also be written as:

A rate ½ conv. code with generator sequences

$$g^{(1)} = [1 \ 1 \ 1], \ g^{(2)} = [1 \ 0 \ 1]$$

A rate ½ conv. code with transfer functions

$$g^{(1)}(x) = 1 + x + x^2, \ g^{(2)}(x) = 1 + x^2$$

- The $(15,13)_8$ conv. code can also be written as:

A rate ½ conv. code with generator sequences:

$$g^{(1)} = [1 \ 1 \ 0 \ 1], \ g^{(2)} = [1 \ 0 \ 1 \ 1]$$

A rate ½ conv. code with transfer functions:

$$g^{(1)}(x) = 1 + x + x^3, \ g^{(2)}(x) = 1 + x^2 + x^3$$



When the codeword (message) length is finite, conv. code is also a linear block code. That says its encoding can be defined by $\bar{c} = \bar{u} \cdot \mathbf{G}$

Example 5.4 For the (7,5) convolutional code with message $\bar{u} = (u_1, u_2, u_3, u_4, u_5)$, we have

$$u_{1} + \begin{cases} c_{1}^{1} = 0 \oplus 0 \oplus u_{1} \\ c_{1}^{2} = 0 \oplus u_{1} \end{cases} \qquad u_{2} + \begin{cases} c_{2}^{1} = 0 \oplus u_{1} \oplus u_{2} \\ c_{2}^{2} = 0 \oplus u_{2} \end{cases} \qquad u_{3} + \begin{cases} c_{3}^{1} = u_{1} \oplus u_{2} \oplus u_{3} \\ c_{3}^{2} = u_{1} \oplus u_{3} \end{cases}$$
$$u_{4} + \begin{cases} c_{4}^{1} = u_{2} \oplus u_{3} \oplus u_{4} \\ c_{4}^{2} = u_{2} \oplus u_{4} \end{cases} \qquad u_{5} + \begin{cases} c_{5}^{1} = u_{3} \oplus u_{4} \oplus u_{5} \\ c_{5}^{2} = u_{3} \oplus u_{5} \end{cases}$$

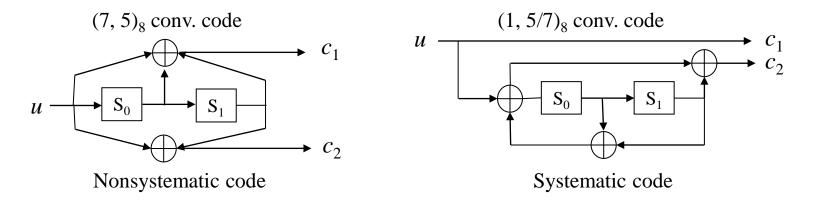


In general, given a rate $\frac{1}{2}m = 2$ conv. code can be defined by $g^{(1)}(x) = g_0^{(1)} + g_1^{(1)}x + g_2^{(1)}x^2$ and $g^{(2)}(x) = g_0^{(2)} + g_1^{(2)}x + g_2^{(2)}x^2$. Its generator matrix **G** is



§ 5.2 Systematic Convolutional Codes

- The $(7, 5)_8$ conv. code's systematic counterpart is:



Encoding and Registers' updating rules: $[S_0 S_1]$ are initialized as [0 0];

$$c_1 = u$$
; (systematic feature) feedback = $S_0 \oplus S_1$; $c_2 = u \oplus \text{feedback} \oplus S_1$; $S_1' = S_0$; $S_0' = u \oplus \text{feedback}$;

Remark: Systematic encoding structure is often used to constitute Turbo codes (Chapter 6).

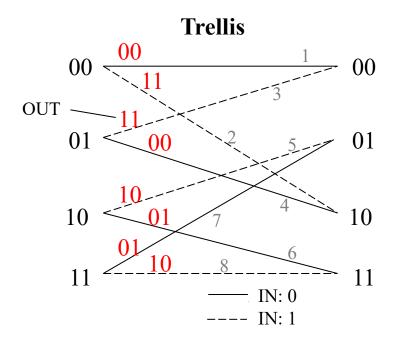


§ 5.2 Systematic Convolutional Codes

For the $(1, 5/7)_8$ conv. code

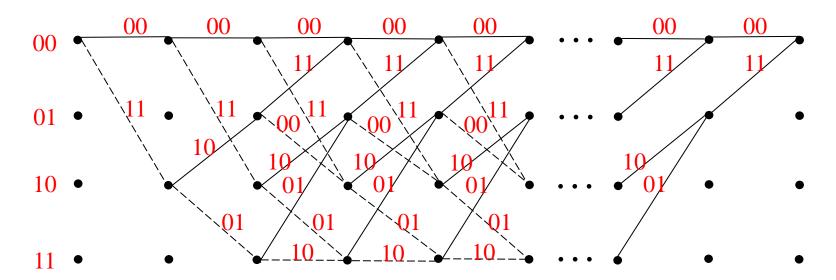
State Table

IN	Current State	Next State	Out	ID
0	00	00	00	1
1	00	10	11	2
0	01	10	00	3
1	01	00	11	4
0	10	11	01	5
1	10	01	10	6
0	11	01	01	7
1	11	11	10	8



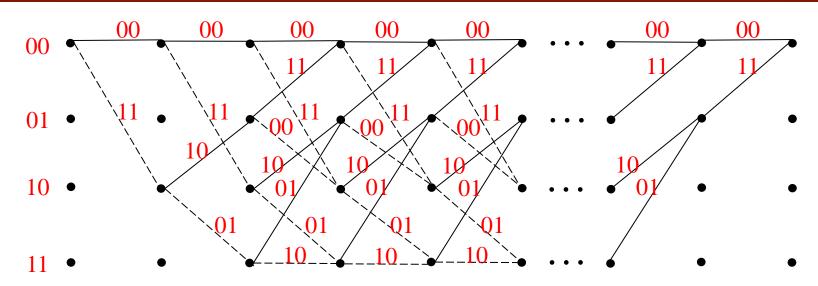


Let us extend the trellis of the $(7, 5)_8$ conv. code as if there is a sequential input.



- Such an extension results in a Viterbi trellis
- A path in the Viterbi trellis represents a convolutional codeword that corresponds to a sequential input (message).





- Decoding motivation: Given a received word \bar{r} , find the mostly likely codeword \hat{c} such that the Hamming distance $d(\bar{r}, \hat{c})$ is minimized.
- Since \hat{c} corresponds to a path in the Viterbi trellis, tracing back the path of \hat{c} enables us to find out the message.
- Branch metrics: Hamming distance between a transition branch's output and the corresponding received symbol (or bits).
- Path metrics: Accumulated Hamming distance of the previous branch metrics.

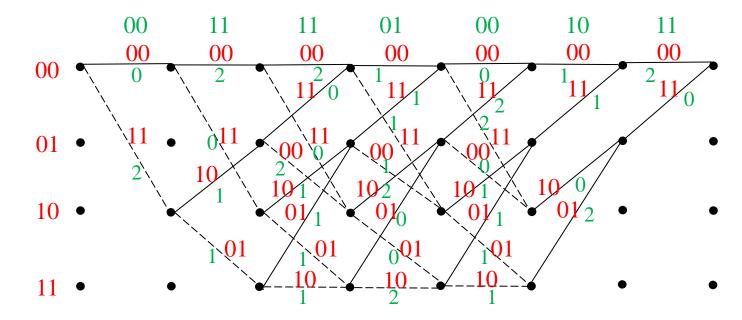


Example 5.5 Given the $(7, 5)_8$ conv. code as in *Examples 5.1 - 5.3*. The transmitted codeword is $\bar{c} = [00 \ 11 \ 01 \ 01 \ 01 \ 11]$ (codeword produced by adding tailing bits). After channel, the received word is

$$\bar{r} = [00\ 11\ 1\ 1\ 01\ 00\ 10\ 11]$$

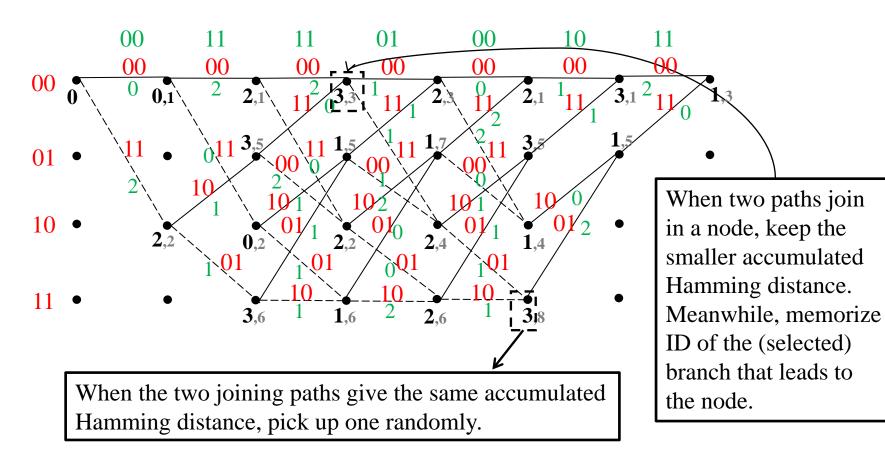
Try to use the following Viterbi trellis to decode it.

Step 1: Calculate all the branch metrics.





Step 2: Calculate the path metrics and memorize the path IDs.

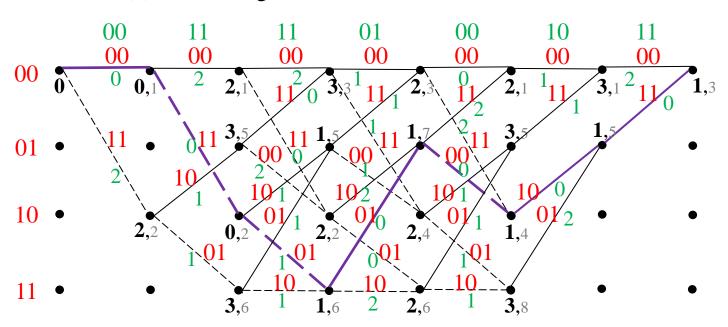




Step 3: Pick up the minimal path metric and trace back to determine the message.

Tracing rules: (1) Trellis connection;

(2) The tracing route should match the trellis transition ID.



Decoding output: 0 1 1 0 1 0 0



Metrics of the Viterbi Decoding Process

Branch Metrics Table

ID	Branch Metric									
1	0	2	2	1	0	1	2			
2	2	0	0	1	2	∞	∞			
3	∞	∞	0	1	2	1	0			
4	∞	∞	2	1	0	∞	∞			
5	8	1	1	2	1	0	∞			
6	8	1	1	0	1	8	∞			
7	∞	8	1	0	1	2	∞			
8	8	8	1	2	1	8	∞			

Path Metrics Table

00	0	0	2	3	2	2	3	1
01	8	8	3	1	1	3	1	8
10	8	2	0	2	2	1	8	8
11	8	8	3	1	2	3	8	8

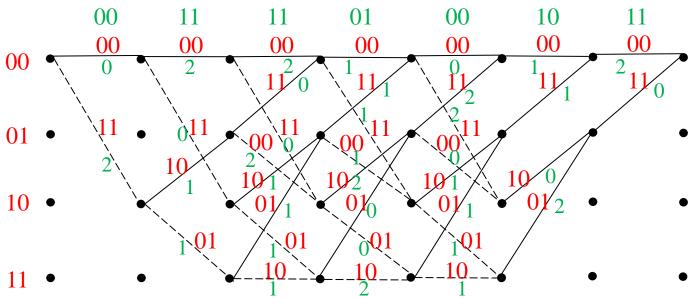
Trellis Transition ID Table

00	1	1	3	3	1	1	3
01	×	5	5	7	5	5	×
10	2	2	2	4	4	×	×
11	×	6	6	6	8	×	×

Remark: With tailing bits, the backward trace always starts from the all-zero state.



Free distance of a convolutional code



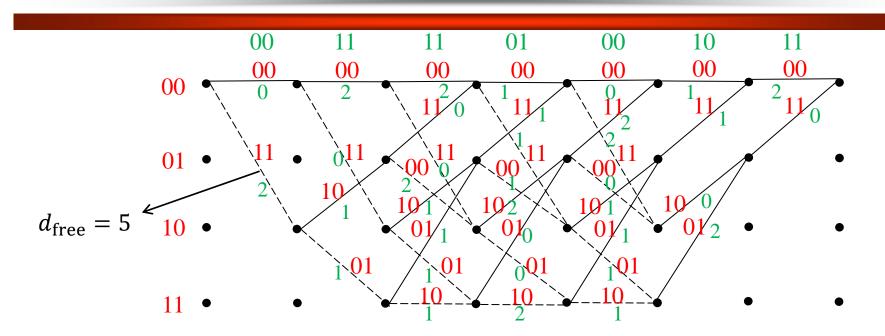
- A conv. code's performance is determined by its free distance.
- Free distance

$$d_{\text{free}} = \min\{d_{\text{Ham}}(\overline{c_1}, \overline{c_2}), \overline{c_1} \neq \overline{c_2}\}$$

- With knowing $\mathbf{0} = [0 \ 0 \ 0 \ \dots 0]$ is also a convolutional codeword

$$d_{\text{free}} = \min\{weight(\bar{c}), \bar{c} \neq \mathbf{0}\}.$$





Hence, it is the minimum weight of all finite length paths in the Viterbi trellis that diverge from and emerge with the all zero state.

- The tailing bits are needed to ensure a greater free distance for the code.
- Convolutional code with a large number of states will have a great $d_{\rm free}$, and hence stronger error-correction capability



Remark: Convolutional code is more competent in correcting spread errors, but not burst errors.

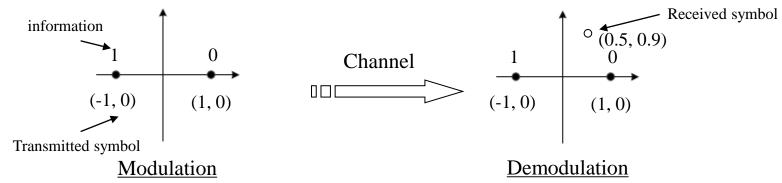
E.g., with
$$\bar{r}_1 = [0 \ 1 \ e \ 1 \ e \ 1 \ 0 \ 1 \ 0 \ e \ 1]$$
 and $\bar{r}_2 = [0 \ 1 \ 0 \ e \ e \ e \ 0 \ 1 \ 0 \ 0 \ 1 \ 1]$

Viterbi algorithm is more competent in correcting received vector \bar{r}_1 .



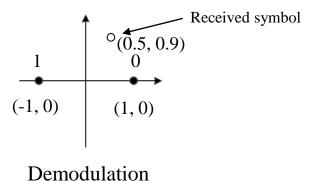
- Soft-decision Viterbi decoding
 - While we are performing the hard-decision Viterbi decoding, we have the scenario that two joining paths yield the same accumulated Hamming distance. This would cause decoding 'ambiguity' and performance penalty;
 - Such a performance loss can be compensated by utilizing soft-decision decoding, e.g., soft-decision Viterbi decoding
- Modulation and Demodulation
 - Modulation: mapping the coded symbol into a transmitted symbol;
 - Demodulation: determining the codeword symbol with a received symbol;

BPSK





Modulation and Demodulation (e.g., BPSK)



Hard-decision: the information bit is 0. The Hamming distance becomes the Viterbi decoding metrics;

Soft-decision: the information bit has Pr. of 0.7 being 0 and Pr. of 0.3 bing 1. The *Euclidean distance (or probability)* becomes the Viterbi decoding metrics;

Euclidean Distance

Definition: The Euclidean distance between points p and q is the length of the line segment connecting them.

$$p(x_1, y_1)$$
 • $d_{Eud} = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$



Example 5.6. Given the $(7, 5)_8$ conv. code as in **Examples 5.5**.

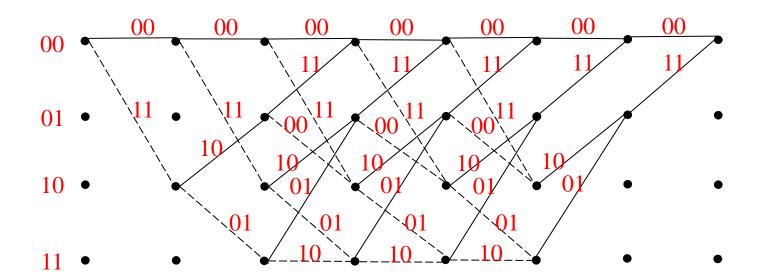
The transmitted codeword is $\bar{c} = [00 \ 11 \ 01 \ 01 \ 00 \ 10 \ 11]$

After BPSK modulation, the transmitted symbols are:

$$(1,0),(1,0),(-1,0),(-1,0),(1,0),(-1,0),(1,0),(-1,0),(1,0),(1,0),(-1,0),(1,0),(-1,0),(-1,0)$$

After the channel, the received symbols are:

$$(0.8, 0.2), (1.2, -0.4), (-1.3, 0.3), (-0.9, -0.1), (-0.5, 0.4), (-1.0, 0.1), (1.1, 0.4), (-0.7, -0.2), (1.2, 0.2), (0.9, 0.3), (-0.9, -0.2), (1, 0.2), (-1.1, 0), (-0.8, 0.1).$$





Step 1: Calculate all the branch metrics.

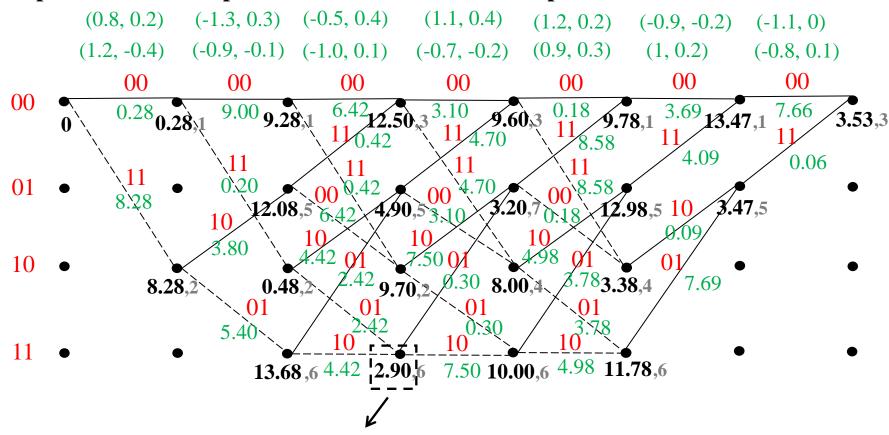
$$0 \rightarrow (1, 0)$$

$$1 \rightarrow (-1, 0)$$

Branch metric: $(-1.3 - 1)^2 + (0.3 - 0)^2 + (-0.9 + 1)^2 + (-0.1 - 0)^2 = 5.40$



Step 2: Calculate the path metrics and memorize the path IDs.



When two paths join in a node, keep the smaller accumulated squared Euclidean distance.



Step 3: Pick up the minimal path metric and trace back to determine the message.

Tracing rules: The same as hard-decision Viterbi decoding algorithm.

Decoding output: 0 1 1 0 1 0 0



Metrics of the Soft-Decision Viterbi Decoding Process

Path Metrics Table

00	0	0.28	0.98	12.50	9.60	9.78	13.47	3.53
01	∞	∞	12.08	4.90	3.20	12.98	3.47	8
10	∞	8.28	0.48	9.70	8.00	3.38	8	8
11	∞	∞	13.68	2.90	10.00	11.78	∞	∞

Branch Metrics Table _____

ID	Branch Metric									
1	0.28	9.00	6.42	3.10	0.18	3.69	7.66			
2	8.28	0.20	0.42	4.70	8.58	8	8			
3	∞	8	0.42	4.70	8.58	4.09	0.06			
4	∞	∞	6.42	3.10	0.18	8	8			
5	∞	3.80	4.42	7.50	4.98	0.09	∞			
6	∞	5.40	2.42	0.30	3.78	∞	∞			
7	∞	∞	2.42	0.30	3.78	7.69	8			
8	∞	∞	4.42	7.50	4.98	8	8			

Trellis Transition ID Table

00	1	1	3	3	1	1	3
01	×	5	5	7	5	5	×
10	2	2	2	4	4	×	×
11	×	6	6	6	6	×	×



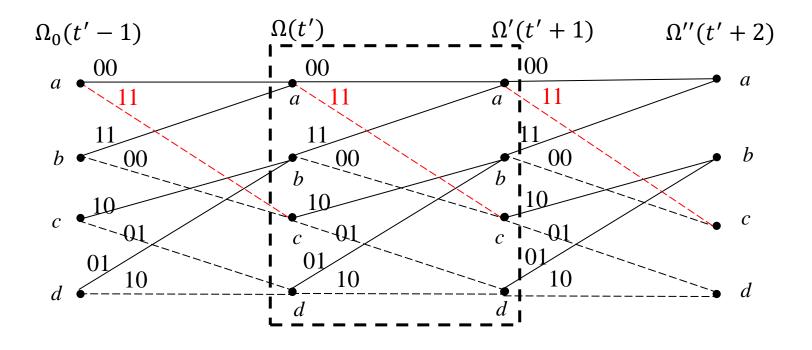
- Hard-decision Viterbi Decoding: a Hard-In-Hard-Out (HIHO) decoding.
 Soft-decision Viterbi Decoding: a Soft-In-Hard-Out (SIHO) decoding.
 BCJR Decoding: a Soft-In-Soft-Out (SISO) decoding.
- A Soft-In-Soft-Out (SISO) decoding algorithm that takes probabilities as the input and delivers probabilities as the output.
- With an attempt to deliver both the *a posteriori probabilities* of $P(c_t|y_t)$ and $P(u_{t'}|y_t)$, it is also called the maximum *a posteriori* (MAP) algorithm.

message symbols: $u_1, u_2, \dots, u_{t'}, \dots$ codeword symbols: $c_1, c_2, \dots, c_t, \dots$ $c_1^1, c_1^2, c_2^1, c_2^2, \dots, c_{t'}^1, c_{t'}^2, \dots$

received symbols:
$$y_1, y_2, ..., y_t, ... y_1^1, y_1^2, y_2^1, y_2^2, ..., y_{t'}^1, y_{t'}^2, ...$$



- In a trellis (e.g., trellis of the $(7, 5)_8$ conv. code).

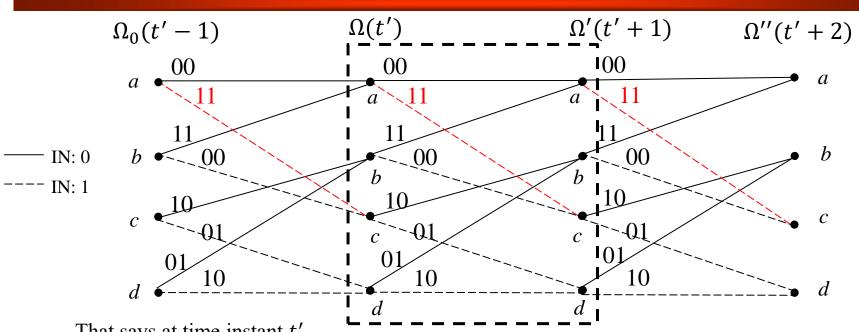


- IN: 0

The (IN, OUT, current state, next state) tuple happens as an entity.

---- IN: 1





That says at time instant t'

 \sum Prob [trellis transition w.r.t. an input θ] = Prob [$u_{t'} = \theta$], $\theta \in \{0,1\}$.

 \sum Prob [trellis transition w.r.t. an output of θ] = Prob [$c_{t'}^{1(2)} = \theta$], $\theta \in \{0,1\}$.

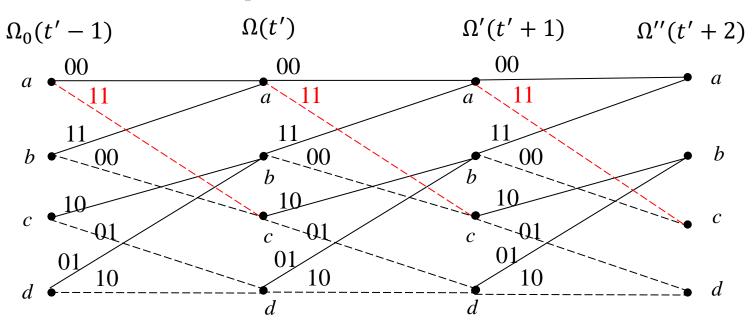
We seek to determine all the Prob[trellis transition w.r.t. an **input** θ] at time t' to know $P(u_{t'} = \theta | y_t)$

We seek to determine all the Prob[trellis transition w.r.t. an **output** θ] at time t' to know

$$P\left(c_{t'}^{1(2)} = \theta | y_t\right)$$



- Determine the state transition probabilities.

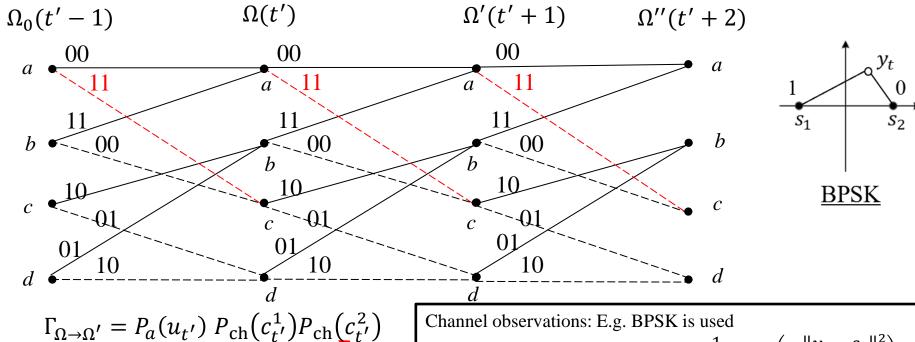


- For a rate half conv. code, $u_{t'} \rightarrow c_{t'}^1, c_{t'}^2$
- Trellis state transition probability: $(\Omega, \Omega') \in \{a, b, c, d\}$

$$\Gamma_{\Omega \to \Omega'} = P_a(u_{t'}) P_{ch}(c_{t'}^1) P_{ch}(c_{t'}^2)$$



Determine the state transition probabilities.



A priori prob. of information bit. E.g. w/o knowledge of $u_{t'}$,

$$P_a(u_{t'}=0) = P_a(u_{t'}=1) = 0.5$$

Channel observations: E.g. BPSK is used

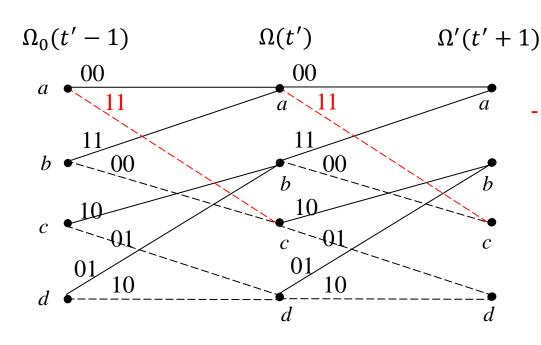
$$P_{\rm ch}(c_{t'}^1 = 0) = P(y_t | c_{t'}^1 = 0) = \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{\|y_t - s_2\|^2}{N_0}\right)$$

$$P_{\text{ch}}(c_{t'}^1 = 1) = P(y_t | c_{t'}^1 = 0) = \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{\|y_t - s_1\|^2}{N_0}\right)$$

 $P_{\rm ch}(c_{t'}^2)$ can be calculated similarly.



- Determine the probability of each beginning state.



Probability of beginning a trellis transition $(\Omega \rightarrow \Omega')$ from state Ω (Determined by a forward trace).

$$A_{t'}(\Omega) = N_A \sum_{(\Omega_0, \Omega)} A_{t'-1}(\Omega_0) \cdot \Gamma_{\Omega_0 \to \Omega},$$

$$t' = 1, 2, \dots, k.$$

 N_A : Normalization factor that ensures $A_{t'}(a) + A_{t'}(b) + A_{t'}(c) + A_{t'}(d) = 1$.

- Knowing the Viterbi trellis starts from the all-zero state, we initialize:

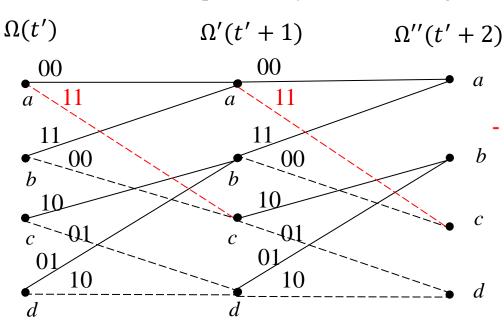
$$A_0(a) = 1$$
, and $A_0(b) = A_0(c) = A_0(d) = 0$.

- E.g., in the highlighted trellis transition

$$A_{t'}(a) = A_{t'-1}(a) \cdot \Gamma_{a \to a} + A_{t'-1}(b) \cdot \Gamma_{b \to a}.$$



- Determine the probability of each ending state.



Probability of ending the trellis transition $(\Omega \rightarrow \Omega')$ at state Ω' (Determined by a backward trace).

$$B_{t'+1}(\Omega') = N_B \sum_{(\Omega', \Omega'')} B_{t'+2}(\Omega'') \cdot \Gamma_{\Omega' \to \Omega''}.$$

 N_B : Normalization factor that ensures

$$B_{t'+1}(a) + B_{t'+1}(b) + B_{t'+1}(c) + B_{t'+1}(d) = 1.$$

By ensuring after encoding, the shift registers (encoder) are restored to the all zero state (achieved by bit tailing), we can initialize:

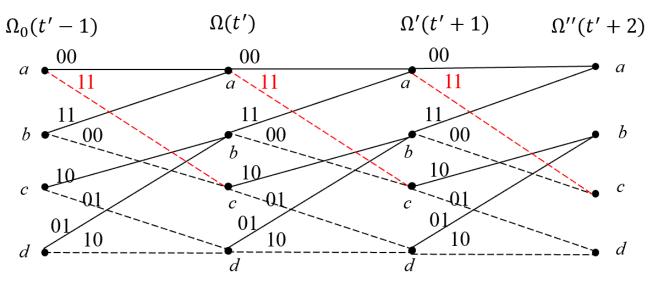
$$B_{k+2}(a) = 1$$
, and $B_{k+2}(b) = B_{k+2}(c) = B_{k+2}(d) = 0$.

- E.g., in the highlighted trellis transition

$$B_{t'+1}(c) = B_{t'+2}(b) \cdot \Gamma_{c \to b} + B_{t'+2}(d) \cdot \Gamma_{c \to d}$$



- Determine the *a posteriori* probability of each information bit



 N_p : Normalization factor that ensures $P(u_{t'} = 0|y_t) + P(u_{t'} = 1|y_t) = 1$

After the **Forward Trace** and **Backward Trace**, we obtain all the $A_{t'}(\Omega)$, $B_{t'+1}(\Omega')$ and $\Gamma_{\Omega \to \Omega'}$ of each time instant t'. We can now determine the *a posteriori* probabilities $P(u_{t'}|y_t)$ for each information bit as



- E.g.,

$$\begin{split} P(u_{t'} = 0 | y_t) &= N_P \cdot (A_{t'}(a) \cdot \Gamma_{a \to a} \cdot B_{t'+1}(a) + A_{t'}(b) \cdot \Gamma_{b \to a} \cdot B_{t'+1}(a) \\ & A_{t'}(c) \cdot \Gamma_{c \to b} \cdot B_{t'+1}(b) + A_{t'}(d) \cdot \Gamma_{d \to b} \cdot B_{t'+1}(b)). \end{split}$$

$$N_p = P(u_{t'} = 0 | y_t) + P(u_{t'} = 1 | y_t)$$

- Decision based on the *a posteriori* probabilities.

$$\hat{u}_{t'} = 0$$
, if $P(u_{t'} = 0 | y_t) \ge P(u_{t'} = 1 | y_t)$
 $\hat{u}_{t'} = 1$, if $P(u_{t'} = 1 | y_t) > P(u_{t'} = 0 | y_t)$

Similarly,
$$P(c_{t'}^1 = 0|y_t)$$
, $P(c_{t'}^1 = 1|y_t)$, $P(c_{t'}^2 = 0|y_t)$, $P(c_{t'}^2 = 1|y_t)$ can be made.



Example 5.7. With the same transmitted codeword and received symbols of **Example 5.6**, use the BCJR algorithm to decode it.

With the received symbols, we can determine

$$\begin{cases} P_{ch}(c_1^1 = 0) = 0.83 \\ P_{ch}(c_1^1 = 1) = 0.17 \end{cases} \begin{cases} P_{ch}(c_1^2 = 0) = 0.92 \\ P_{ch}(c_1^2 = 1) = 0.08 \end{cases} \begin{cases} P_{ch}(c_2^1 = 0) = 0.92 \\ P_{ch}(c_2^1 = 1) = 0.08 \end{cases} \begin{cases} P_{ch}(c_2^1 = 0) = 0.07 \\ P_{ch}(c_2^1 = 1) = 0.93 \end{cases} \begin{cases} P_{ch}(c_2^2 = 0) = 0.14 \\ P_{ch}(c_2^2 = 1) = 0.86 \end{cases}$$

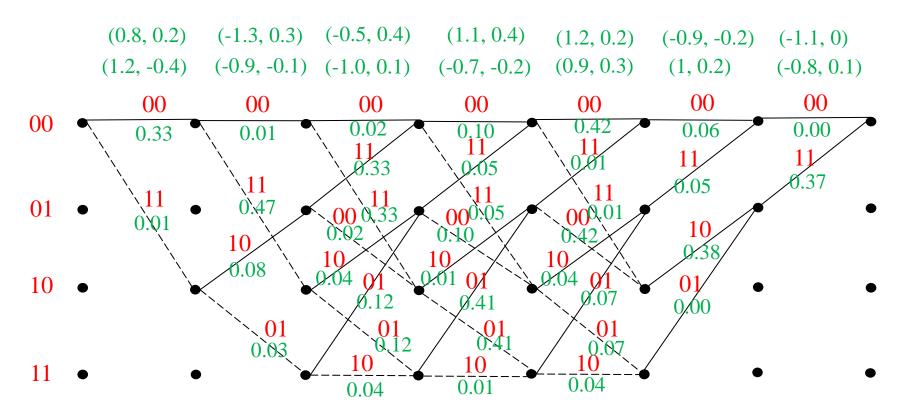
$$\begin{cases} P_{ch}(c_3^1 = 0) = 0.27 \\ P_{ch}(c_3^1 = 1) = 0.73 \end{cases} \begin{cases} P_{ch}(c_3^2 = 0) = 0.12 \\ P_{ch}(c_3^2 = 1) = 0.88 \end{cases} \begin{cases} P_{ch}(c_4^1 = 0) = 0.90 \\ P_{ch}(c_4^1 = 1) = 0.10 \end{cases} \begin{cases} P_{ch}(c_4^2 = 0) = 0.20 \\ P_{ch}(c_4^2 = 1) = 0.80 \end{cases}$$

$$\begin{cases} P_{ch}(c_5^1 = 0) = 0.92 \\ P_{ch}(c_5^1 = 1) = 0.08 \end{cases} \begin{cases} P_{ch}(c_5^2 = 0) = 0.86 \\ P_{ch}(c_5^1 = 1) = 0.14 \end{cases} \begin{cases} P_{ch}(c_6^1 = 0) = 0.14 \\ P_{ch}(c_6^1 = 1) = 0.86 \end{cases} \begin{cases} P_{ch}(c_6^2 = 1) = 0.12 \end{cases}$$

$$\begin{cases} P_{ch}(c_7^1 = 0) = 0.10 \\ P_{ch}(c_7^1 = 1) = 0.90 \end{cases} \begin{cases} P_{ch}(c_7^2 = 0) = 0.17 \\ P_{ch}(c_7^2 = 1) = 0.83 \end{cases}$$

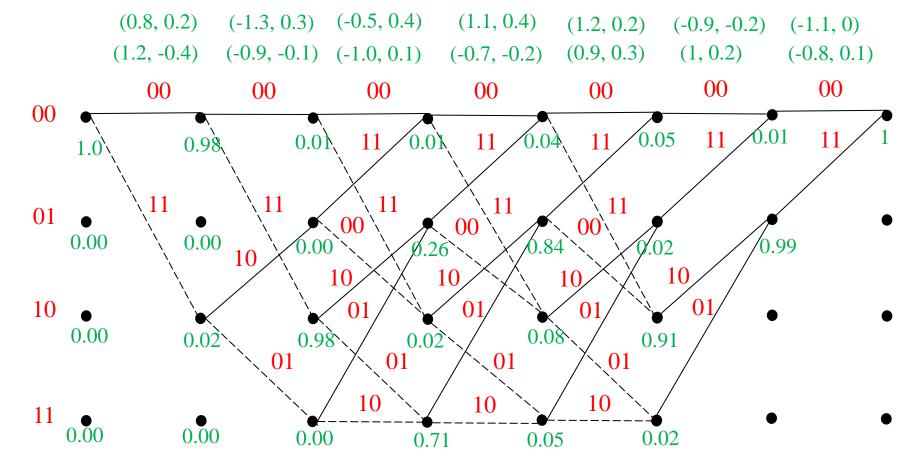


Step 1: Determine $\Gamma_{\Omega \to \Omega'}$ of all transitions.



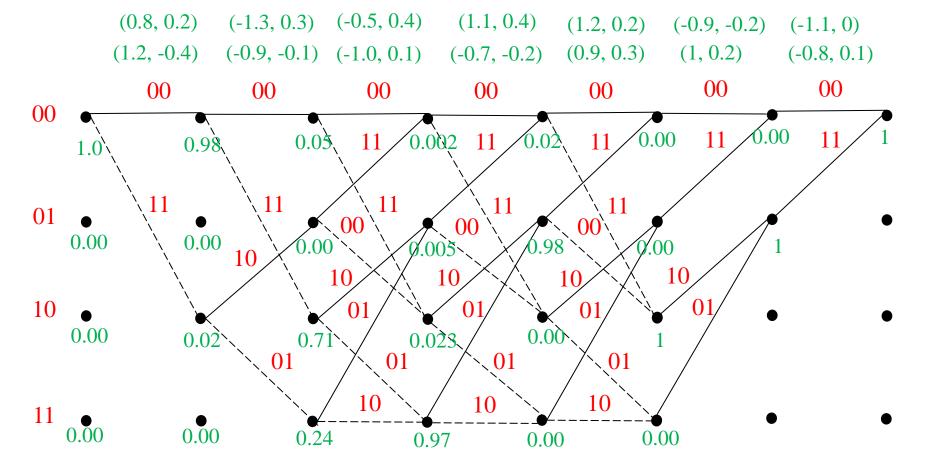


Step 2: Forward trace, determine $A_{t'}(\Omega)$ of values.





Step 3: Backward Trace, determine $B_{t'+1}(\Omega')$ of values.





Step 4: Determine the *a posteriori* probabilities of each information bit.

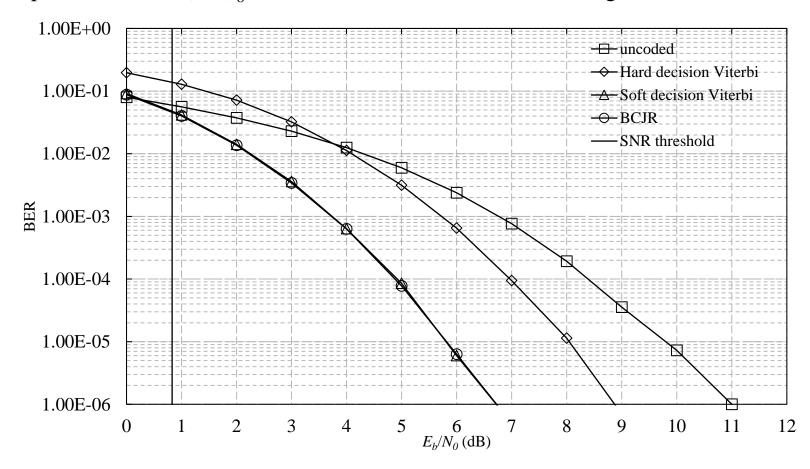
$$\begin{cases}
P(u_1 = 0 | y_1) = 1 \\
P(u_1 = 1 | y_1) = 0
\end{cases}
\qquad \widehat{u}_1 = 0
\qquad
\begin{cases}
P(u_4 = 0 | y_4) = 1 \\
P(u_4 = 1 | y_4) = 0
\end{cases}
\qquad \widehat{u}_4 = 0$$

$$\begin{cases}
P(u_2 = 0 | y_2) = 0 \\
P(u_2 = 1 | y_2) = 1
\end{cases}
\qquad \hat{u}_2 = 1
\qquad
\begin{cases}
P(u_5 = 0 | y_5) = 0 \\
P(u_5 = 1 | y_5) = 1
\end{cases}
\qquad \hat{u}_5 = 1$$

$$\begin{cases} P(u_3 = 0 | y_3) = 0 \\ P(u_3 = 1 | y_3) = 1 \end{cases} \quad \overrightarrow{u}_3 = 1 \qquad \qquad \widehat{u}_6 = \widehat{u}_7 = 0$$

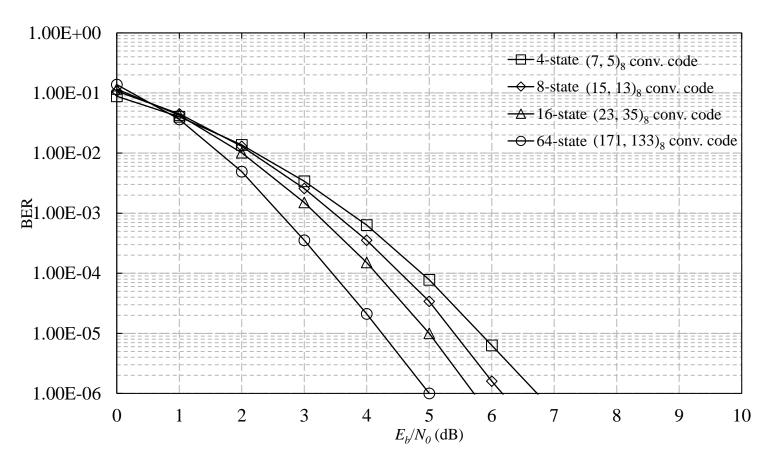


BER performance of $(7, 5)_8$ conv. code over AWGN channel using BPSK.





BER performance of different conv. code over AWGN channel using BPSK.





- Convolutional code enables reliable communications. But as a channel code, its error-correction function is on the expense of spectral efficiency.
- Spectral efficiency $(\eta) = \frac{Nr. of information bits}{transmitted symbol}$
- E.g., an uncoded system using BPSK

 $\eta = 1$ info bits/symbol

A rate 1/2 conv. coded system using BPSK

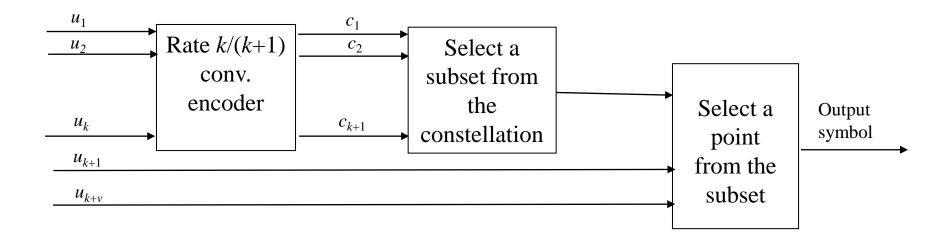
 $\eta = 0.5$ info bits/symbol

- Can we achieve reliable and yet spectrally efficient communication?

Solution: Trellis Coded Modulation (TCM) that integrates a conv. code with a high order modulation [3].

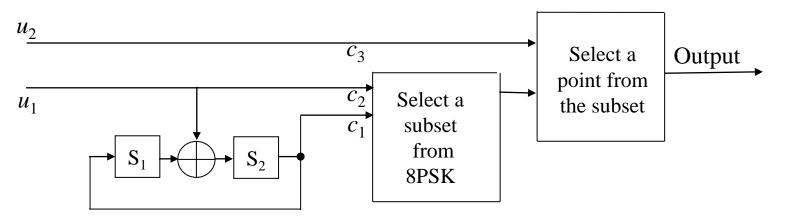


- A general structure of the TCM scheme

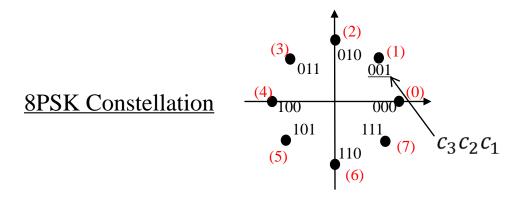




- A rate 2/3 TCM code.



Rate ½ 4-state Convolutional Code

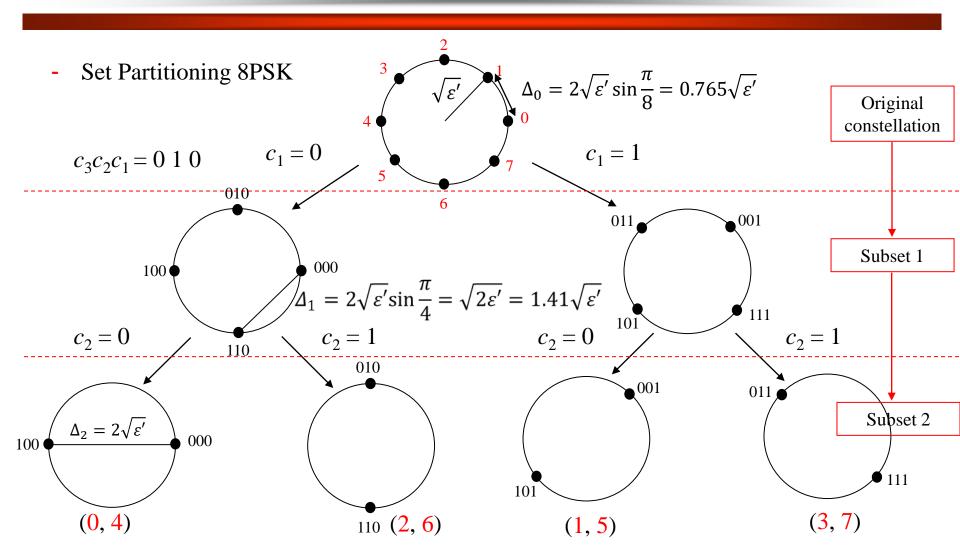




- State table of the rate 2/3 TCM code

Input		Current State		Next State		Output			Symbol
u_1	u_2	S_1	S_2	S_1	S_2	c_1	c_2	c_3	8PSK sym
0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	1	0	1	0	2
0	1	0	0	0	0	0	0	1	4
1	1	0	0	0	1	0	1	1	6
0	0	0	1	1	0	1	0	0	1
1	0	0	1	1	1	1	1	0	3
0	1	0	1	1	0	1	0	1	5
1	1	0	1	1	1	1	1	1	7
0	0	0	0	0	1	0	0	0	0
1	0	1	0	0	0	0	1	0	2
0	1	1	0	0	1	0	0	1	4
1	1	1	0	0	0	0	1	1	6
0	0	1	1	1	1	1	0	0	1
1	0	1	1	1	0	1	1	0	3
0	1	1	1	1	1	1	0	1	5
1	1	1	1	1	0	1	1	1	7







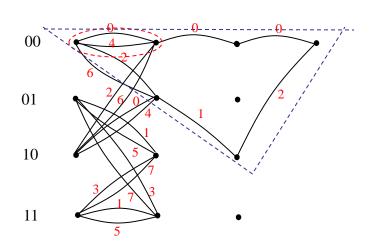
- Set Partitioning 8PSK

By doing set partitioning, the minimum distance between point within a subset is increasing as: $\Delta_0 < \Delta_1 < \Delta_2$.

Original constellation $\Delta_0 = d(0,1) = 2\sqrt{\varepsilon'}\sin\frac{\pi}{8} = 0.765\sqrt{\varepsilon'}$ Subset 1 $\Delta_1 = d(0,2) = \sqrt{2\varepsilon'} = 1.414\sqrt{\varepsilon'}$ $\Delta_2 = d(0,4) = 2\sqrt{\varepsilon'}$



- Viterbi trellis of the rate 2/3 TCM code



For diverse/remerge transition:

$$d_{\text{free}}^2 = [d^2(0,2) + d^2(0,1) + d^2(0,2)]$$

= $2\varepsilon' + (2 - \sqrt{2})\varepsilon' + 2\varepsilon' = 4.568\varepsilon'$

For parallel transition:

$$d_{\text{free}}^2 = d^2(0,4) = 4\varepsilon'$$

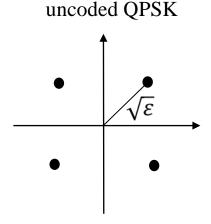
Choose the smaller one as the free distance of the code:

$$d_{\text{free}}^2 = 4\varepsilon'$$

- Bit $c_3 = 0$ and $c_3 = 1$ result in two parallel transition branches. By doing set partitioning, we are trying to maximize the Euclidean distance between the two parallel branches. So that the free distance of the TCM code can be maximized.

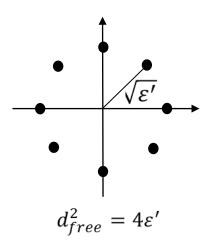


- Asymptotic coding gain over an uncoded system.
- Spectral efficiency $(\eta) = 2$ info bits/sym.



$$d_{min}^2 = 2\varepsilon$$
 Asymptotic coding gain $\gamma = \frac{\binom{d_{free}^2}{\varepsilon'}}{\binom{d_{min}^2}{\varepsilon}} = 2$.

rate 2/3 coded 8PSK



Asymptotic coding gain in $dB = 10 \log_{10} \gamma = 3 dB$.

- With the same transmission spectral efficiency of 2 info bits/sym, the TCM coded system achieves 3 dB coding gain over the uncoded system asymptotically.