

# An introduction to coding theory

Adrish Banerjee

Department of Electrical Engineering  
Indian Institute of Technology Kanpur  
Kanpur, Uttar Pradesh  
India

Feb. 13, 2017



## Lecture #9A: Convolutional codes: Classification, Realization



# Outline of the lecture

- Classification of convolutional encoder



# Outline of the lecture

- Classification of convolutional encoder
  - Feedforward encoder



# Outline of the lecture

- Classification of convolutional encoder
  - Feedforward encoder
  - Feedback encoder



# Outline of the lecture

- Classification of convolutional encoder
  - Feedforward encoder
  - Feedback encoder
- Equivalent encoder



# Outline of the lecture

- Classification of convolutional encoder
  - Feedforward encoder
  - Feedback encoder
- Equivalent encoder
- Catastrophic encoder



# Outline of the lecture

- Classification of convolutional encoder
  - Feedforward encoder
  - Feedback encoder
- Equivalent encoder
- Catastrophic encoder
- Controller canonical form realization



# Outline of the lecture

- Classification of convolutional encoder
  - Feedforward encoder
  - Feedback encoder
- Equivalent encoder
- Catastrophic encoder
- Controller canonical form realization
- Observer canonical form realization



# Outline of the lecture

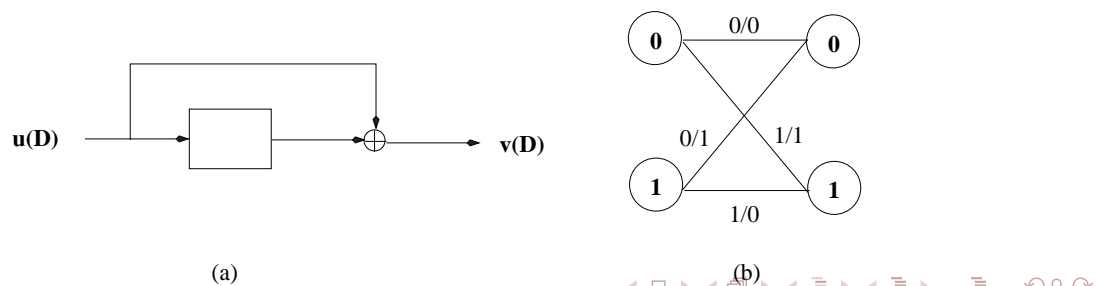
- Classification of convolutional encoder
  - Feedforward encoder
  - Feedback encoder
- Equivalent encoder
- Catastrophic encoder
- Controller canonical form realization
- Observer canonical form realization
- Minimal encoder



# Classification of convolutional encoders

Feedforward Encoder:

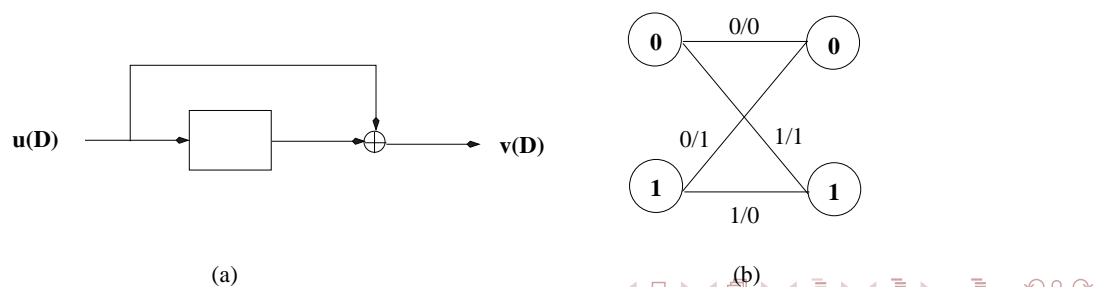
- The encoder corresponding to a polynomial generator matrix does not contain any feedback path, and hence it is known as a feedforward encoder.



# Classification of convolutional encoders

Feedforward Encoder:

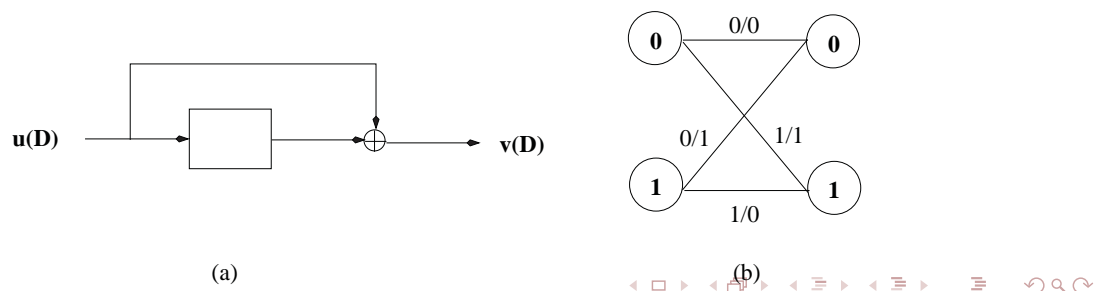
- The encoder corresponding to a polynomial generator matrix does not contain any feedback path, and hence it is known as a feedforward encoder.
- The output of a feedforward encoder can be represented as a linear combination of the current input and a finite number of past inputs. This is also referred as nonrecursive encoder.



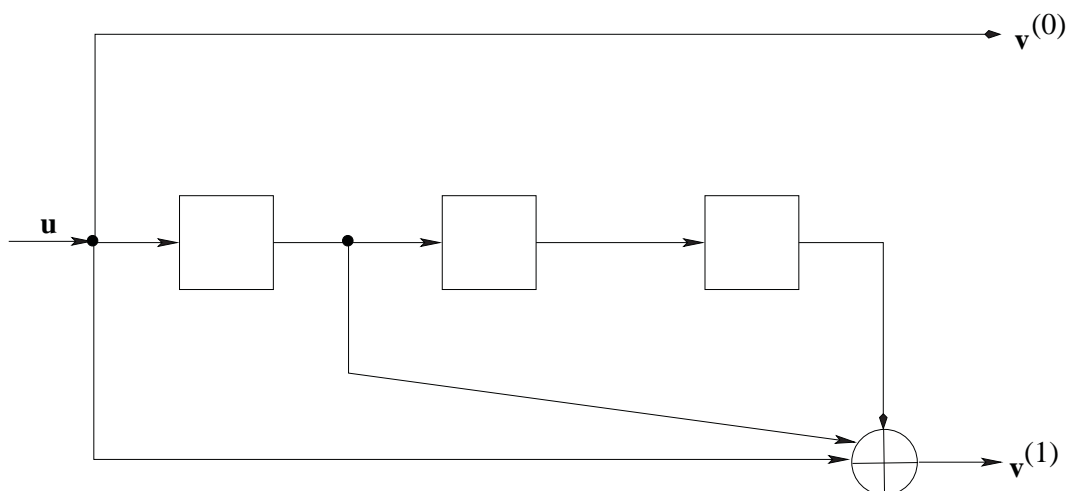
# Classification of convolutional encoders

## Feedforward Encoder:

- The encoder corresponding to a polynomial generator matrix does not contain any feedback path, and hence it is known as a feedforward encoder.
- The output of a feedforward encoder can be represented as a linear combination of the current input and a finite number of past inputs. This is also referred as nonrecursive encoder.
- In figure, the encoder diagram of a rate  $R = 1$ , 2-state feedforward encoder with generator matrix  $\mathbf{G}(D) = [1 + D]$  is shown using a shift register implementation.



# Classification of convolutional encoders

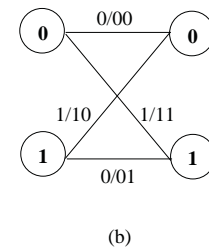
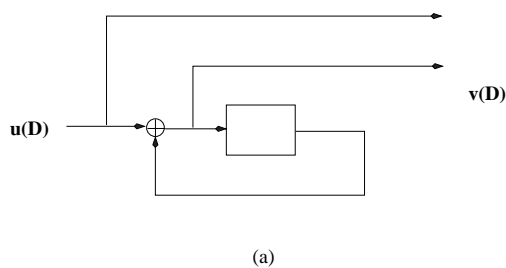


Example: Feedforward encoder

# Classification of convolutional encoders

## Feedback Encoder:

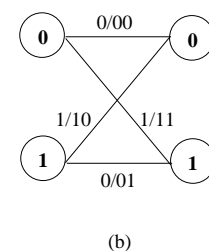
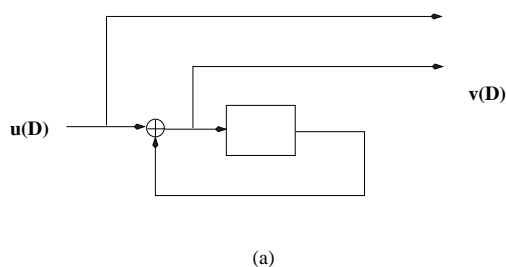
- The encoder corresponding to a rational generator matrix with atleast one nonpolynomial transfer function contains a feedback path and is known as a feedback encoder.



# Classification of convolutional encoders

## Feedback Encoder:

- The encoder corresponding to a rational generator matrix with atleast one nonpolynomial transfer function contains a feedback path and is known as a feedback encoder.
- The output of a feedback encoder can be represented as a linear combination of past inputs as well as past outputs.

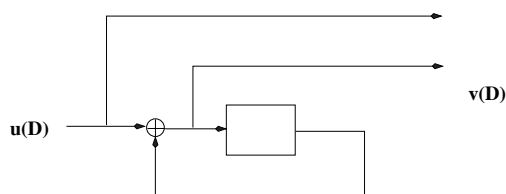




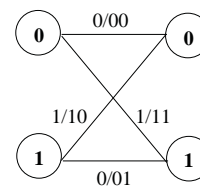
# Classification of convolutional encoders

## Feedback Encoder:

- The encoder corresponding to a rational generator matrix with atleast one nonpolynomial transfer function contains a feedback path and is known as a feedback encoder.
- The output of a feedback encoder can be represented as a linear combination of past inputs as well as past outputs.
- Hence the output depends on infinite number of past inputs. This is also sometimes referred as recursive encoder.



(a)

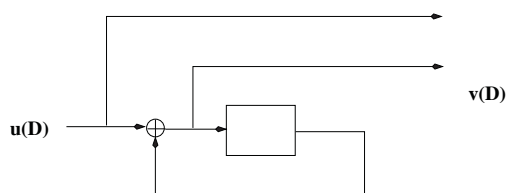


(b)

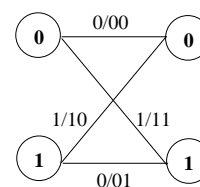
# Classification of convolutional encoders

## Feedback Encoder:

- The encoder corresponding to a rational generator matrix with atleast one nonpolynomial transfer function contains a feedback path and is known as a feedback encoder.
- The output of a feedback encoder can be represented as a linear combination of past inputs as well as past outputs.
- Hence the output depends on infinite number of past inputs. This is also sometimes referred as recursive encoder.
- In figure, the encoder diagram of a rate  $R = 1/2$ , 2-state feedback encoder with generator matrix  $\mathbf{G}(D) = [1 \quad \frac{1}{1+D}]$  is shown.

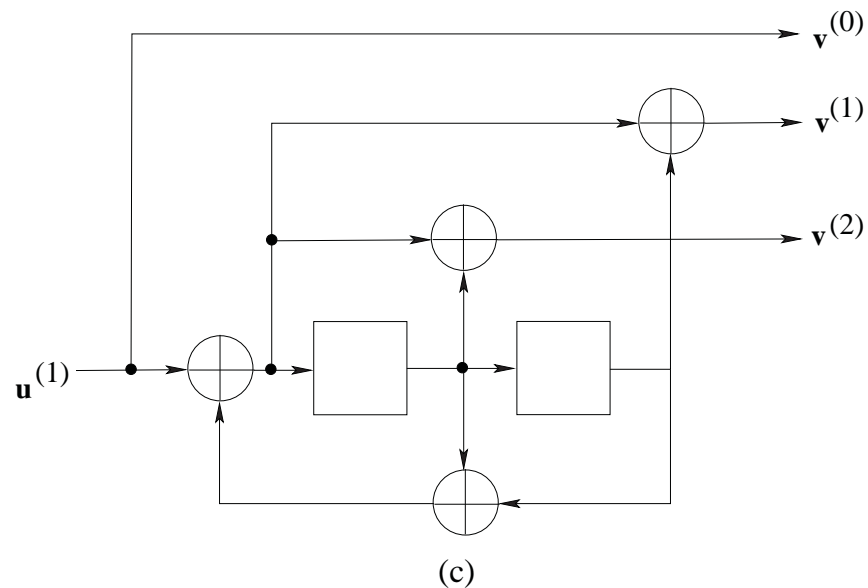


(a)



(b)

# Classification of convolutional encoders



### Example: Feedback encoder

# Classification of convolutional encoders

Systematic Encoder:

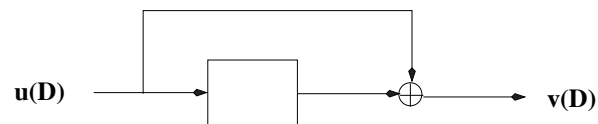
- A rate  $R = k/n$  convolutional encoder whose  $k$  information sequences appear unchanged among the  $n$  code sequences is called a systematic encoder, and its generator matrix is called a systematic generator matrix.



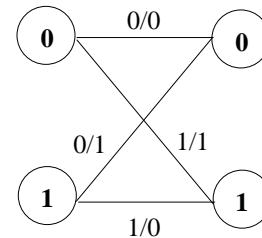
# Classification of convolutional encoders

## Nonsystematic Encoder:

- In a nonsystematic convolutional encoder, the  $k$  information sequences do not appear unchanged in the  $n$  code sequences.
- In figure, a nonsystematic rate  $R = 1/2$  feedforward convolutional encoder is shown.



(a)



(b)

## Equivalent Encoder

- Two convolutional generator matrices  $\mathbf{G}(D)$  and  $\mathbf{G}'(D)$  are equivalent if they encode the same code.

# Equivalent Encoder

- Two convolutional generator matrices  $\mathbf{G}(D)$  and  $\mathbf{G}'(D)$  are equivalent if they encode the same code.
- Two convolutional encoders are equivalent if their generator matrices are equivalent.



# Equivalent Encoder

- Two convolutional generator matrices  $\mathbf{G}(D)$  and  $\mathbf{G}'(D)$  are equivalent if they encode the same code.
- Two convolutional encoders are equivalent if their generator matrices are equivalent.
- Two generator matrices  $\mathbf{G}(D)$  and  $\mathbf{G}'(D)$  are equivalent if and only if there exists a rational invertible matrix  $\mathbf{T}(D)$  such that

$$\mathbf{G}'(D) = \mathbf{T}(D)\mathbf{G}(D)$$



## Equivalent Encoder

- Two convolutional generator matrices  $\mathbf{G}(D)$  and  $\mathbf{G}'(D)$  are equivalent if they encode the same code.
- Two convolutional encoders are equivalent if their generator matrices are equivalent.
- Two generator matrices  $\mathbf{G}(D)$  and  $\mathbf{G}'(D)$  are equivalent if and only if there exists a rational invertible matrix  $\mathbf{T}(D)$  such that

$$\mathbf{G}'(D) = \mathbf{T}(D)\mathbf{G}(D)$$

- Example: The generator matrix,  $\mathbf{G}(D) = [1 \quad \frac{1}{1+D}]$  and  $\mathbf{G}'(D) = [1+D \quad 1]$  are equivalent.



## Equivalent Encoder

- Consider the nonsystematic encoder

$$\mathbf{G}(D) = \begin{bmatrix} 1+D & D & 1+D \\ D & 1 & 1 \end{bmatrix}$$



## Equivalent Encoder

- Consider the nonsystematic encoder

$$\mathbf{G}(D) = \begin{bmatrix} 1+D & D & 1+D \\ D & 1 & 1 \end{bmatrix}$$

- Step 1: Row 1  $\Rightarrow [1/(1+D)][\text{Row 1}]$ .

$$\mathbf{G}_1(D) = \begin{bmatrix} 1 & D/(1+D) & 1 \\ D & 1 & 1 \end{bmatrix}$$



## Equivalent Encoder

- Consider the nonsystematic encoder

$$\mathbf{G}(D) = \begin{bmatrix} 1+D & D & 1+D \\ D & 1 & 1 \end{bmatrix}$$

- Step 1: Row 1  $\Rightarrow [1/(1+D)][\text{Row 1}]$ .

$$\mathbf{G}_1(D) = \begin{bmatrix} 1 & D/(1+D) & 1 \\ D & 1 & 1 \end{bmatrix}$$

- Step 2: Row 2  $\Rightarrow \text{Row 2} + [D][\text{Row 1}]$ .

$$\mathbf{G}_2(D) = \begin{bmatrix} 1 & D/(1+D) & 1 \\ 0 & (1+D+D^2)/(1+D) & 1+D \end{bmatrix}$$



# Equivalent Encoder

- Consider the nonsystematic encoder

$$\mathbf{G}(D) = \begin{bmatrix} 1+D & D & 1+D \\ D & 1 & 1 \end{bmatrix}$$

- Step 1: Row 1  $\Rightarrow [1/(1+D)][\text{Row 1}]$ .

$$\mathbf{G}_1(D) = \begin{bmatrix} 1 & D/(1+D) & 1 \\ D & 1 & 1 \end{bmatrix}$$

- Step 2: Row 2  $\Rightarrow \text{Row 2} + [D][\text{Row 1}]$ .

$$\mathbf{G}_2(D) = \begin{bmatrix} 1 & D/(1+D) & 1 \\ 0 & (1+D+D^2)/(1+D) & 1+D \end{bmatrix}$$

- Step 3: Row 2  $\Rightarrow [(1+D)/(1+D+D^2)][\text{Row 2}]$ .

$$\mathbf{G}_3(D) = \begin{bmatrix} 1 & D/(1+D) & 1 \\ 0 & 1 & (1+D^2)/(1+D+D^2) \end{bmatrix}$$

Navigation icons: back, forward, search, etc.

# Equivalent Encoder

- Step 4: Row 1  $\Rightarrow \text{Row 1} + [D/(1+D)][\text{Row 2}]$ .

Navigation icons: back, forward, search, etc.



## Equivalent Encoder

- Step 4: Row 1  $\Rightarrow$  Row 1 +  $[D/(1+D)][$ Row 2 $]$ .
- Modified Systematic generator matrix

$$\mathbf{G}'(D) = \begin{bmatrix} 1 & 0 & 1/(1+D+D^2) \\ 0 & 1 & (1+D^2)/(1+D+D^2) \end{bmatrix}$$



## Catastrophic Encoder

- A convolutional encoder is catastrophic if it encodes some information sequence with infinitely many non-zero symbols into a code sequence with finitely many non-zero symbols.



# Catastrophic Encoder

- A convolutional encoder is catastrophic if it encodes some information sequence with infinitely many non-zero symbols into a code sequence with finitely many non-zero symbols.
- This means that a finite number of channel errors may result in infinitely many errors in the receiver.



# Catastrophic Encoder

- A convolutional encoder is catastrophic if it encodes some information sequence with infinitely many non-zero symbols into a code sequence with finitely many non-zero symbols.
- This means that a finite number of channel errors may result in infinitely many errors in the receiver.
- Example:

$$\mathbf{G}(D) = \begin{bmatrix} 1 + D & 1 + D^2 \end{bmatrix}$$



# Catastrophic Encoder

- A convolutional encoder is catastrophic if it encodes some information sequence with infinitely many non-zero symbols into a code sequence with finitely many non-zero symbols.
- This means that a finite number of channel errors may result in infinitely many errors in the receiver.
- Example:

$$\mathbf{G}(D) = [1 + D \quad 1 + D^2]$$

- If the input sequence  $\mathbf{u}(D) = \left[ \frac{1}{1+D} \right] = 1 + D + D^2 + \dots$ , then the output sequence,  $[1 \quad 1 + D]$  has only weight 3, even though the information sequence has infinite weight.

# Controller Canonical Form Realization

- In controller canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $k$  shift registers are used for input sequences, and  $n$  adders are used to form the output sequences.

# Controller Canonical Form Realization

- In controller canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $k$  shift register are used for input sequences, and  $n$  adders are used to form the output sequences.
- the  $k$  input sequences enter the shift registers at the left end of each shift register.

# Controller Canonical Form Realization

- In controller canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $k$  shift register are used for input sequences, and  $n$  adders are used to form the output sequences.
- the  $k$  input sequences enter the shift registers at the left end of each shift register.
- The  $n$  adders used to obtain output sequences are external to the shift register.

# Controller Canonical Form Realization

- In controller canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $k$  shift register are used for input sequences, and  $n$  adders are used to form the output sequences.
- the  $k$  input sequences enter the shift registers at the left end of each shift register.
- The  $n$  adders used to obtain output sequences are external to the shift register.
- In Figure 3.6(a) (next page), a rate  $R = 1$ , nonsystematic convolutional encoder with following generator function  $\mathbf{G}(D)$  is implemented in controller canonical form realization.

$$\mathbf{G}(D) = \left[ \frac{f_0 + f_1 D + \cdots + f_{m-1} D^{m-1} + f_m D^m}{1 + q_1 D + q_2 D^2 + \cdots + q_m D^m} \right]$$

Navigation icons: back, forward, search, etc.

# Controller Canonical Form Realization

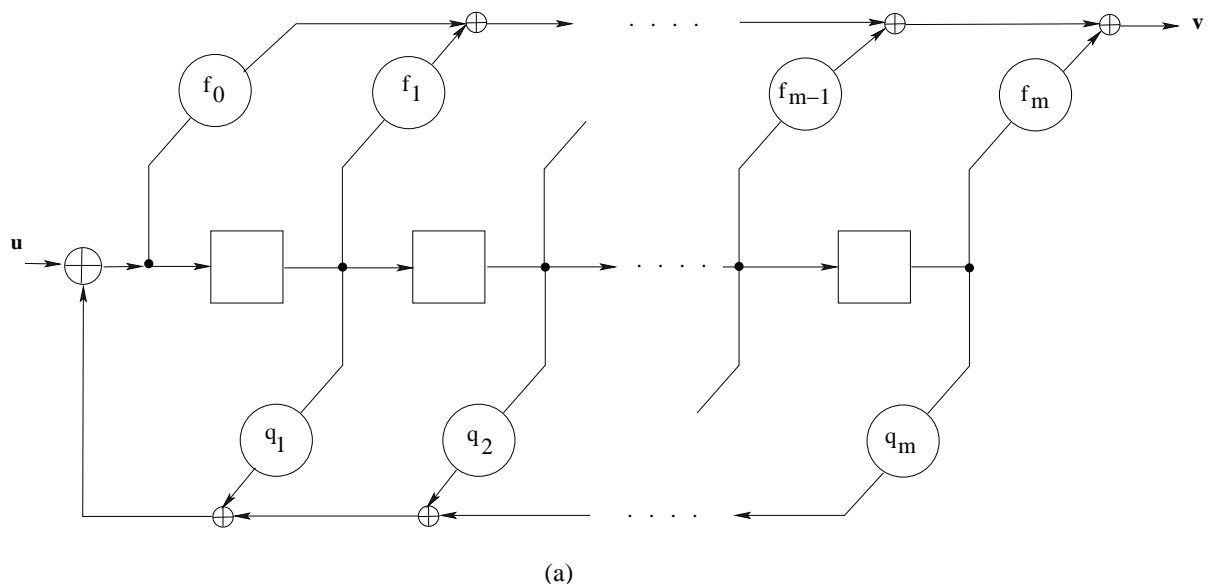


Figure 3.6

Navigation icons: back, forward, search, etc.

# Observer Canonical Form Realization

- In observer canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $n$  shift register are used for output sequences.



# Observer Canonical Form Realization

- In observer canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $n$  shift register are used for output sequences.
- The  $k$  input sequences enter the adders internal to the shift registers.



## Observer Canonical Form Realization

- In observer canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $n$  shift register are used for output sequences.
- The  $k$  input sequences enter the adders internal to the shift registers.
- The lowest degree term in the generator polynomial represent the connections to the right hand side of the shift registers.



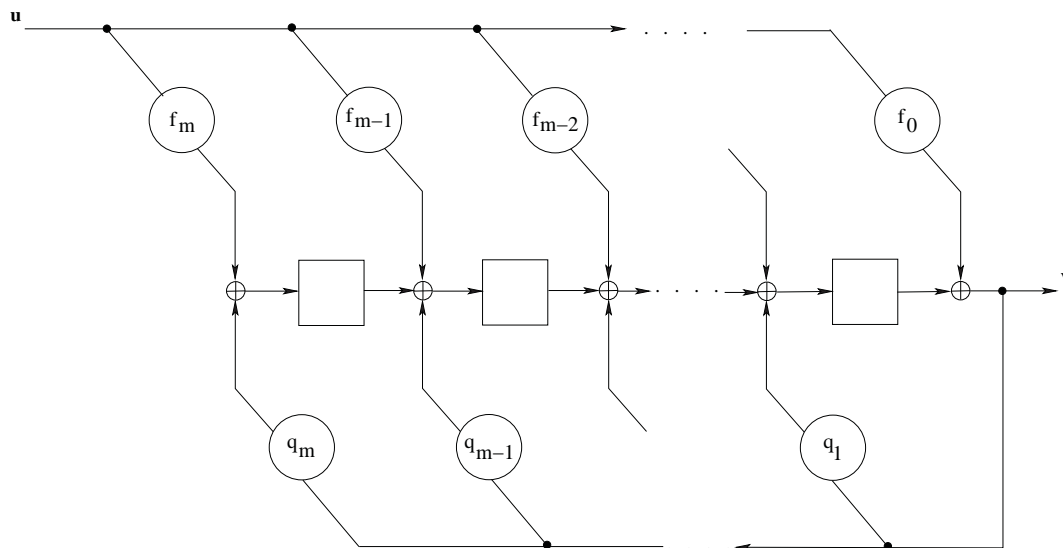
## Observer Canonical Form Realization

- In observer canonical form realization, to realize a rate  $R = k/n$  convolutional encoder,  $n$  shift register are used for output sequences.
- The  $k$  input sequences enter the adders internal to the shift registers.
- The lowest degree term in the generator polynomial represent the connections to the right hand side of the shift registers.
- In Figure 3.6(b) (next page), a rate  $R = 1$ , nonsystematic convolutional encoder with following generator function  $\mathbf{G}(D)$  is implemented in observer canonical form realization.

$$\mathbf{G}(D) = \left[ \frac{f_0 + f_1 D + \cdots + f_{m-1} D^{m-1} + f_m D^m}{1 + q_1 D + q_2 D^2 + \cdots + q_m D^m} \right]$$



# Observer Canonical Form Realization



(b)

Figure 3.6

Navigation icons: back, forward, search, etc.

## Realization of Convolutional encoder

Example 3.7:

- Let's consider a rate  $R = 2/3$  systematic feedforward encoder with generator matrix

$$\mathbf{G}(D) = \begin{bmatrix} 1 & 0 & 1 + D + D^2 \\ 0 & 1 & 1 + D \end{bmatrix}.$$

Navigation icons: back, forward, search, etc.



## Realization of Convolutional encoder

Example 3.7:

- Let's consider a rate  $R = 2/3$  systematic feedforward encoder with generator matrix

$$\mathbf{G}(D) = \begin{bmatrix} 1 & 0 & 1 + D + D^2 \\ 0 & 1 & 1 + D \end{bmatrix}.$$

- The parity check matrix can be written as

$$\mathbf{H}(D) = \begin{bmatrix} \mathbf{h}^{(0)}(D) & \mathbf{h}^{(1)}(D) & 1 \end{bmatrix} = \begin{bmatrix} 1 + D + D^2 & 1 + D & 1 \end{bmatrix}.$$

Navigation icons: back, forward, search, etc.

## Realization of Convolutional encoder

Example 3.7:

- Let's consider a rate  $R = 2/3$  systematic feedforward encoder with generator matrix

$$\mathbf{G}(D) = \begin{bmatrix} 1 & 0 & 1 + D + D^2 \\ 0 & 1 & 1 + D \end{bmatrix}.$$

- The parity check matrix can be written as

$$\mathbf{H}(D) = \begin{bmatrix} \mathbf{h}^{(0)}(D) & \mathbf{h}^{(1)}(D) & 1 \end{bmatrix} = \begin{bmatrix} 1 + D + D^2 & 1 + D & 1 \end{bmatrix}.$$

- The controller canonical form realization results in  $(3, 2, 3)$  encoder.

Navigation icons: back, forward, search, etc.

# Realization of Convolutional encoder

Example 3.7:

- Let's consider a rate  $R = 2/3$  systematic feedforward encoder with generator matrix

$$\mathbf{G}(D) = \begin{bmatrix} 1 & 0 & 1 + D + D^2 \\ 0 & 1 & 1 + D \end{bmatrix}.$$

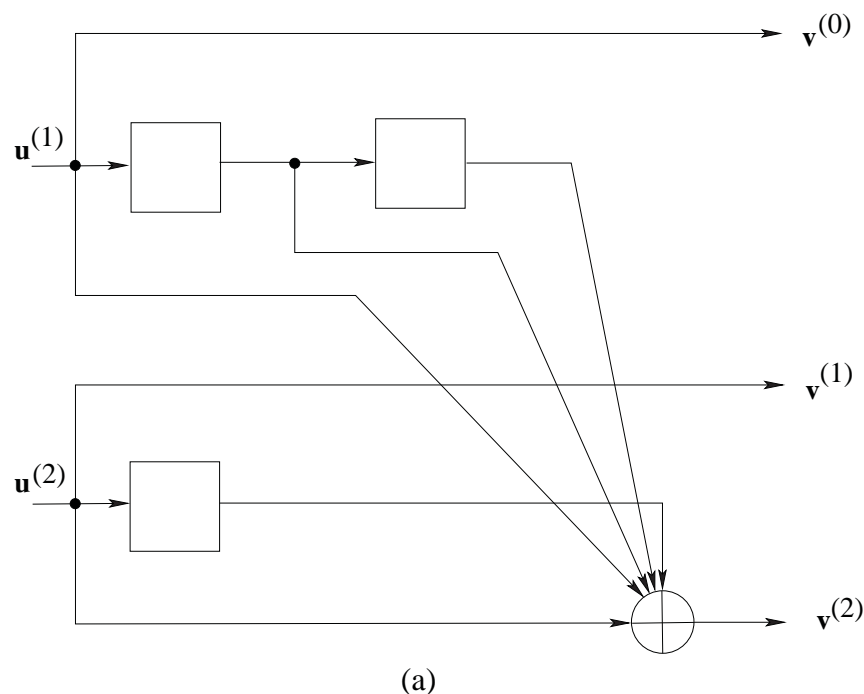
- The parity check matrix can be written as

$$\mathbf{H}(D) = \begin{bmatrix} \mathbf{h}^{(0)}(D) & \mathbf{h}^{(1)}(D) & 1 \end{bmatrix} = \begin{bmatrix} 1 + D + D^2 & 1 + D & 1 \end{bmatrix}.$$

- The controller canonical form realization results in (3, 2, 3) encoder.
- The observer canonical form realization results in (3, 2, 2) encoder.

Navigation icons: back, forward, search, etc.

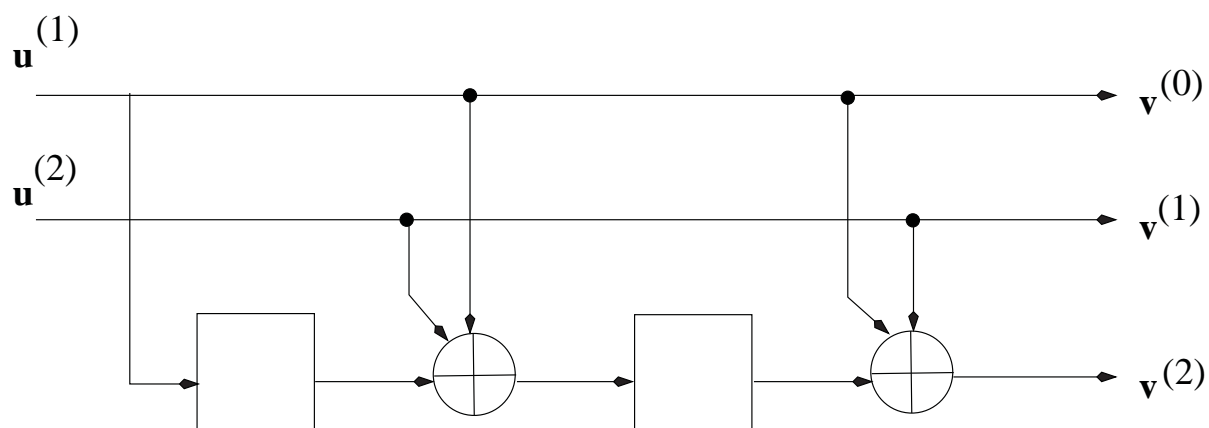
## Controller Canonical Form Realization



Example 3.7

Navigation icons: back, forward, search, etc.

# Observer Canonical Form Realization



(b)  
Example 3.7

## Minimal encoder

- A generator matrix of a convolutional code is minimal if its number of states is minimal over all equivalent generator matrices.

# Minimal encoder

- A generator matrix of a convolutional code is minimal if its number of states is minimal over all equivalent generator matrices.
- A minimal encoder is a realization of a minimal encoding matrix  $\mathbf{G}(D)$  with the minimal number of memory elements over all realizations of  $\mathbf{G}(D)$ .