

## Dictionary Techniques

- Huffman and arithmetic coding assumed i.i.d. sources
- Most sources are correlated
- Main idea:
  - 1 Build a list of commonly occurring patterns
  - 2 Transmit index in the list.
- Exploits fact that certain patterns recur frequently.
- Consider 2 cases
  1. Static Dictionary
  2. Dynamic Dictionary

## Static Dictionary:

- Suppose Five letter alphabet source:

$$A = \{a, b, c, d, r\}$$

- Using statistics of source, build dictionary

ad	ac	ab	r	d	c	b	a	Entry
111	110	101	100	011	010	001	000	Code

- Encode a b r a c a d a b r a

1. Read ab -----> 101

2. Read ra -----> not in dictionary

3. Read r -----> 100

4. Read ac -----> ~~110~~

5. ....

101 100 110 111 101 101 000  
ab r ac ad ab r a

- Opposite of Huffman coding:

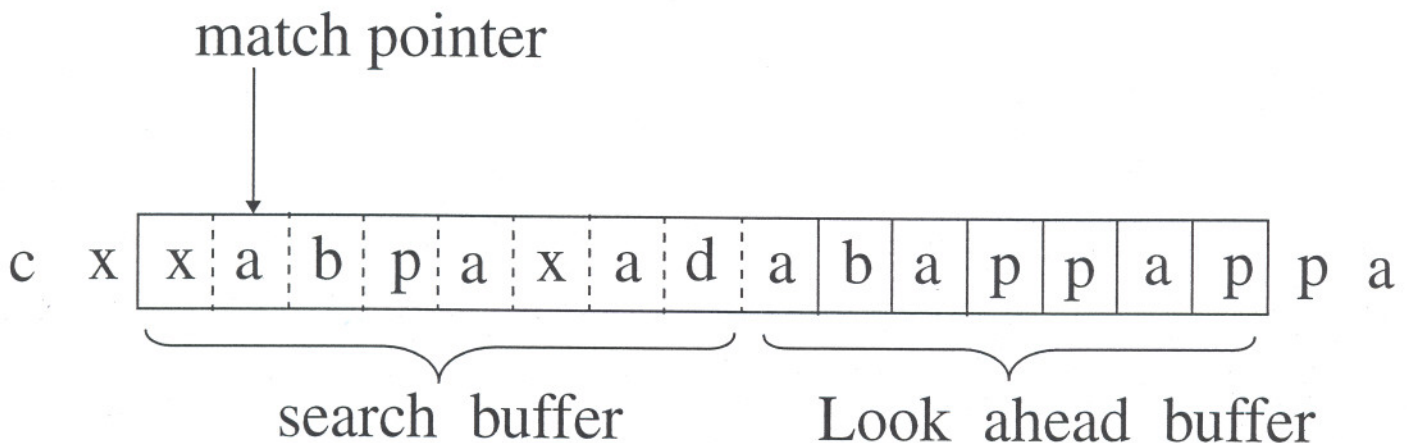
Ziv

## Adaptive Dictionary:

- ~~Ziv~~ Lempel 1977 + 1978
- LZ1 -----> 1977    LZ2 -----> 1978
- LZ1 discussed Here.
- Basic idea:

Dictionary portion of previously encoded sequence

- Sliding window:
  - 1) search buffer
  - 2) lookahead buffer



## Example

. . . c a b r a c a d a b r a r r a r r a d . . .

- Window = 13

- look ahead buffer = 6

- Search buffer = 7

. . . . c a b r a c a | d a b r a r | r a . . .

1. No match to d --->  $\langle \overset{0 \leftarrow \text{zero}}{\emptyset}, 0, C(d) \rangle$

. . . . c a b r a c a d | a b r a r r | a r . . .

2. Match for a:  $\left\{ \begin{array}{l} \overset{D_1 \text{ (letter)}}{\emptyset} = 2 \text{ ---} \rightarrow l = 1 \\ \emptyset = 4 \text{ ---} \rightarrow l = 1 \\ \boxed{\emptyset = 7 \text{ ---} \rightarrow l = 4} \end{array} \right.$

$\langle 7, 4, C(r) \rangle$

a d a b r a r | r a r r a d

3. Match for  $r$ :

$$\begin{cases} 0 = 1 \text{ ---} \rightarrow l = 1 \\ 0 = 3 \text{ ---} \rightarrow l = 5 \end{cases}$$

$\langle 3, 5, C(d) \rangle$



exceeds the  
boundary between  
search and  
look ahead buffer

## Encoding steps

1. Move search pointer back until match in search buffer
2. Offset  $\triangleq$  Distance of pointer from look ahead buffer
3. Do consecutive symbols of pointer match also?
4. Search the search buffer for the longest match
5. length of match  $\triangleq$  # of consecutive symbol match.
6. Send  $\langle o, l, c \rangle$ 
  - $o$  = offset
  - $l$  = match length
  - $c$  = codeword of symbol in LA buffer, following match

Example:  $\langle 7, 2, \text{codeword for } a \rangle$

## Adaptive Dictionary

- Why send  $\mathcal{C}$ ?
  - just in case no match

- Total # of bits:

$$\lceil \log_2 S \rceil + \lceil \log_2 W \rceil + \lceil \log_2 A \rceil$$

$W \rightarrow$  Capital  $W$

$S$  = size of search buffer

$W$  = Size of window (search + LA)

$A$  = size of source alphabet

## What to Code (Classification of Image Coding Systems)

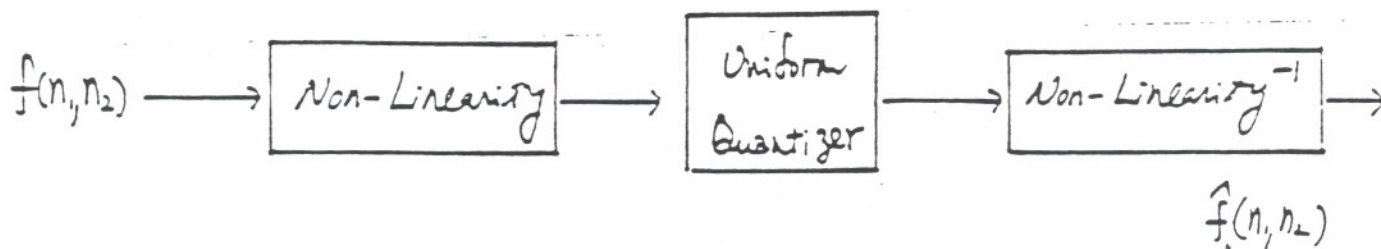
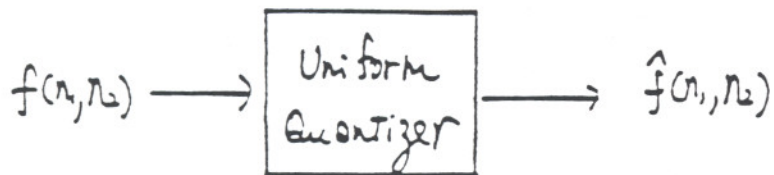
1. Waveform Coder (code the intensity)
  - PCM (Pulse Code Modulation) and its improvements
  - DM (Delta Modulation)
  - DPCM (Differential Pulse Code Modulation)
  - Two-channel Coder
  
2. Transform Coder (code transform coefficients of an image)
  - Karhunen-Loeve Transform
  - Discrete Fourier Transform
  - Discrete Cosine Transform
  
3. Image Model Coder
  - Auto-regressive Model for texture
  - Modelling of a restricted class of images

**Note:** Each of the above can be made to be adaptive



# Waveform Coder

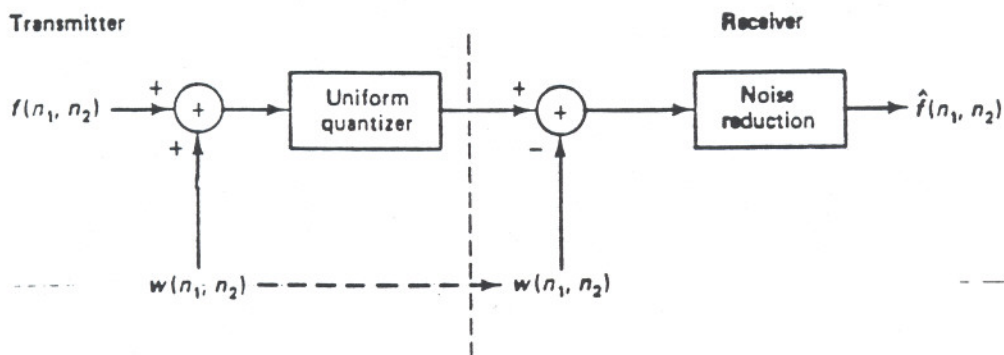
## PCM Coding



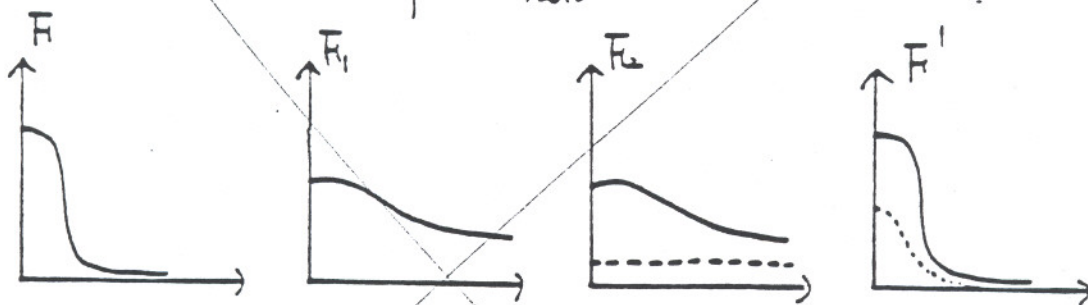
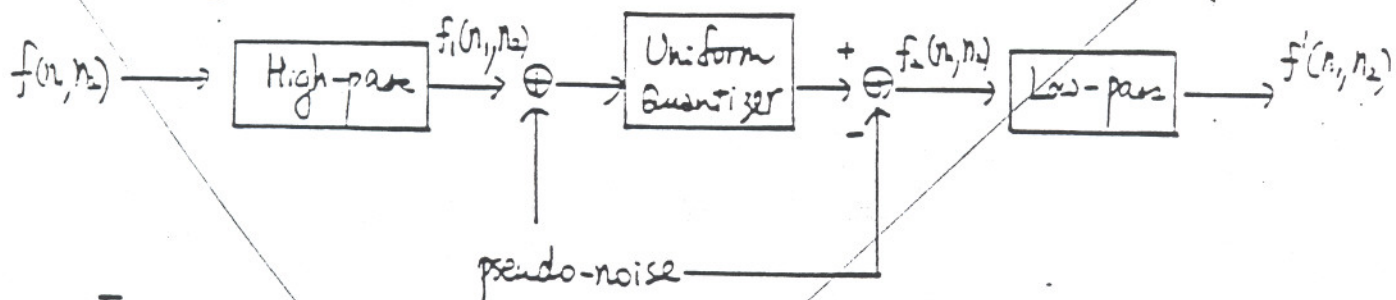
- very simple
- typically requires over 5-6 bits/pixel for good quality
- false contours for low-bit rate case

# Improvements of PCM (cont.)

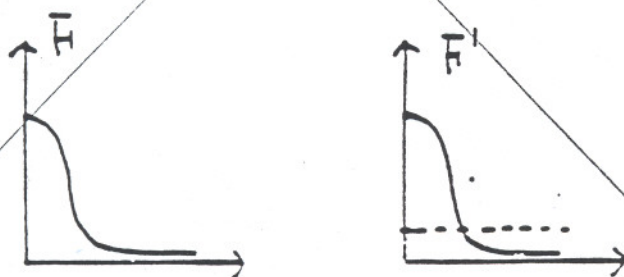
## 2. Roberts' Pseudo-Noise Technique with Noise Reduction:



## 3. Roberts' Pseudo-Noise Technique and Highpass/Lowpass Filtering:

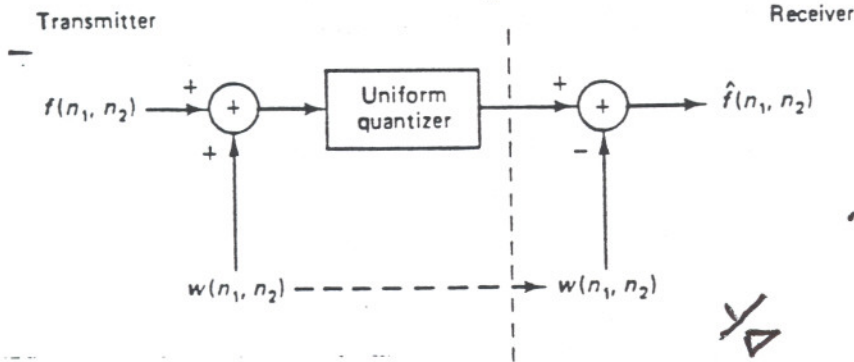


If there were not Highpass/Lowpass Filtering:



# Improvements of PCM

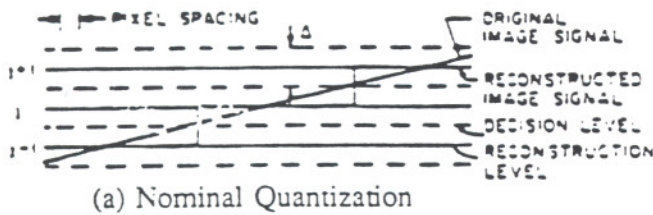
## 1. Roberts' Pseudo-Noise Technique:



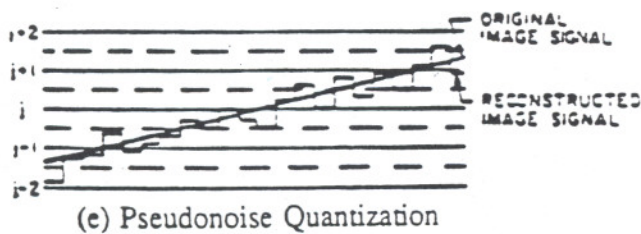
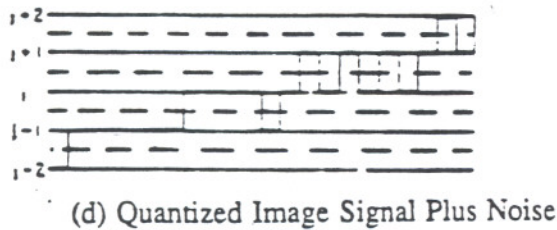
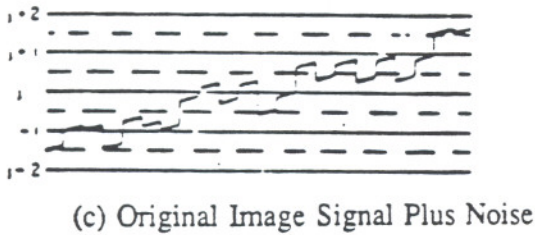
$$-\frac{\Delta}{2} \leq w_0 \leq \frac{\Delta}{2}$$

otherwise

$$P(w) = \begin{cases} \frac{1}{\Delta} & 0 \\ 0 & \text{otherwise} \end{cases}$$



- false contours disappear — replaced by additive random noise



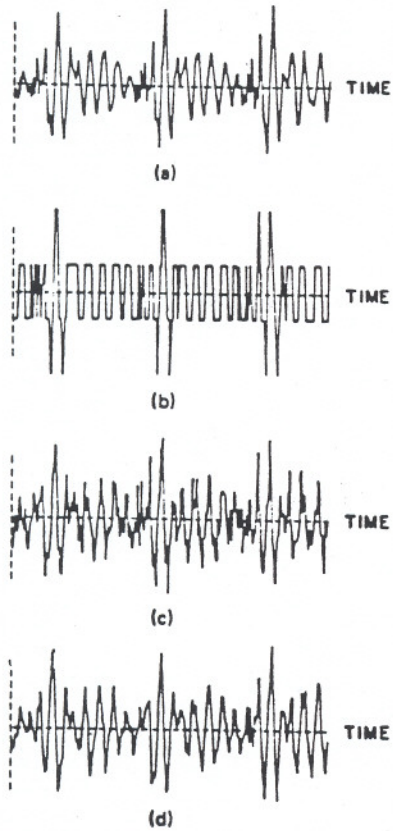


Figure 10.21 Example of quantization noise reduction in PCM speech coding. (a) Segment of noise-free voiced speech; (b) PCM-coded speech at 2 bits/sample; (c) PCM-coded speech at 2 bits/sample by Roberts's pseudonoise technique; (d) PCM-coded speech at 2 bits/sample with quantization noise reduction.

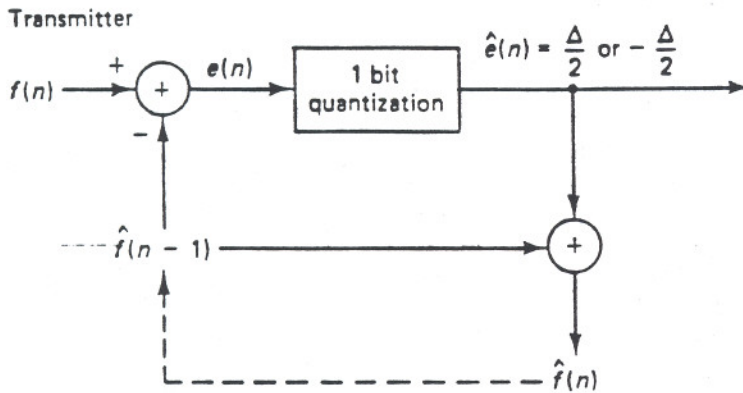


Figure 10.22 Example of quantization noise reduction in PCM image coding. (a) Original image of  $512 \times 512$  pixels; (b) PCM-coded image at 2 bits/pixel; (c) PCM-coded image at 2 bits/pixel by Robert's pseudonoise technique; (d) PCM-coded image at 2 bits/pixel with quantization noise reduction.

# Delta Modulation (DM)

$f(n_1, n_2)$  : signal,  $\hat{f}(n_1, n_2)$  : coded signal

## Transmitter



## Receiver

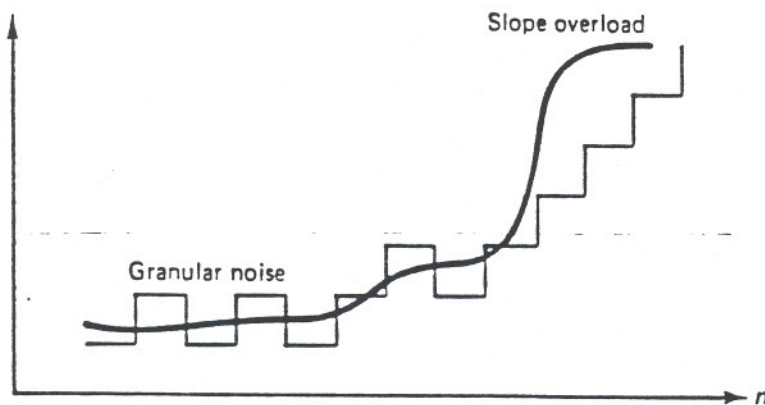
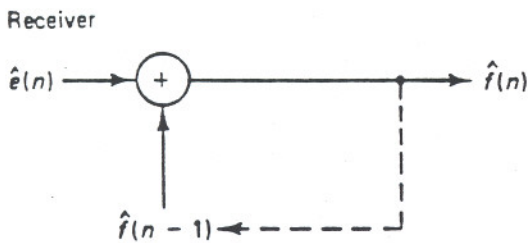


Figure 10.26 Granular noise and slope-overload distortion in delta modulation.

- needs over 2-3 bits/pixel to get good quality

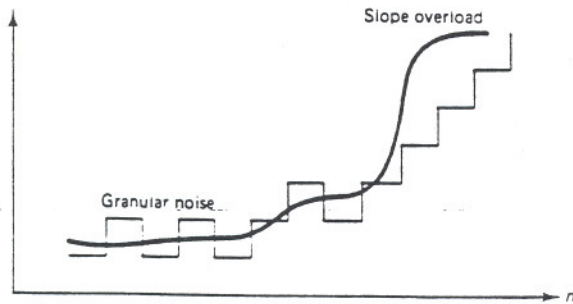


Figure 10.26 Granular noise and slope-overload distortion in delta modulation.

requirements, and the step size  $\Delta$  is chosen through some compromise between the two requirements.

Figure 10.27 illustrates the performance of a DM system. Figures 10.27(a) and (b) show the results of DM with step sizes of  $\Delta = 8\%$  and  $15\%$ , respectively, of the overall dynamic range of  $f(n_1, n_2)$ . The original image used is the  $512 \times 512$ -pixel image in Figure 10.22(a). When  $\Delta$  is small [Figure 10.27(a)], the granular noise is reduced, but the slope overload distortion problem is severe and the resulting image appears blurred. As we increase  $\Delta$  [Figure 10.27(b)], the slope overload distortion is reduced, but the graininess in the regions where the signal varies slowly is more pronounced.

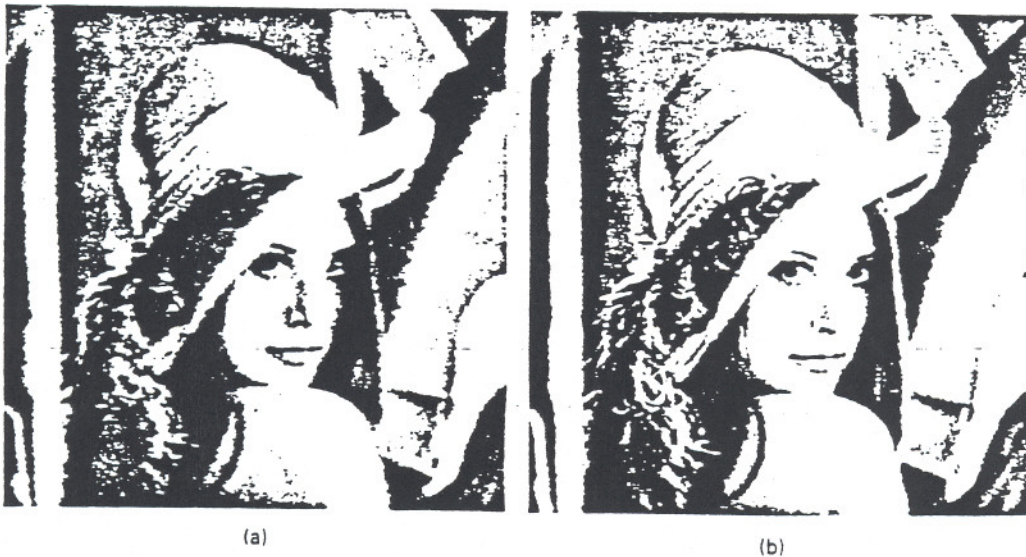


Figure 10.27 Example of delta-modulation (DM)-coded image. The original image used is the image in Figure 10.22(a). (a) DM-coded image with  $\Delta = 8\%$  of the overall dynamic range. NMSE = 14.8%, SNR = 8.3 dB; (b) DM-coded image with  $\Delta = 15\%$ . NMSE = 9.7%, SNR = 10.1 dB.



Figure 10.28 DM-coded image at 2 bits/pixel. The original image used is the image in Figure 10.22(a). NMSE = 2.4%, SNR = 16.2 dB.

To obtain good quality image reconstruction using DM without significant graininess or slope overload distortion, 3–4 bits/pixel is typically required. A bit rate higher than 1 bit/pixel can be obtained in DM by oversampling the original analog signal relative to the sampling rate used in obtaining  $f(n_1, n_2)$ . Oversampling reduces the slope of the digital signal  $f(n)$  so a smaller  $\Delta$  can be used without increasing the slope overload distortion. An example of an image coded by DM at 2 bits/pixel is shown in Figure 10.28. To obtain the image in Figure 10.28, the size of the original digital image in Figure 10.22(a) was increased by a factor of two by interpolating the original digital image by a factor of two along the horizontal direction. The interpolated digital image was coded by DM with  $\Delta = 12\%$  of the dynamic range of the image and the reconstructed image was undersampled by a factor of two along the horizontal direction. The size of the resulting image is the same as the image in Figure 10.27, but the bit rate in this case is 2 bits/pixel.

### 10.3.3 Differential Pulse Code Modulation

Differential pulse code modulation (DPCM) can be viewed as a generalization of DM. In DM, the difference signal  $e(n) = f(n) - \hat{f}(n-1)$  is quantized. The most recently coded  $\hat{f}(n-1)$  can be viewed as a prediction of  $f(n)$  and  $e(n)$  can be viewed as the error between  $f(n)$  and a prediction of  $f(n)$ . In DPCM, a prediction of the current pixel intensity is obtained from more than one previously coded pixel intensity. In DM, only one bit is used to code  $e(n)$ . In DPCM, more than one bit can be used in coding the error.

A DPCM system is shown in Figure 10.29. To code the current pixel intensity  $f(n_1, n_2)$ ,  $f(n_1, n_2)$  is predicted from previously reconstructed pixel intensities. The predicted value is denoted by  $\hat{f}'(n_1, n_2)$ . In the figure, we have assumed that  $\hat{f}(n_1-1, n_2)$ ,  $\hat{f}(n_1, n_2-1)$ ,  $\hat{f}(n_1-1, n_2-1)$ , ... were reconstructed prior to coding  $f(n_1, n_2)$ . We are attempting to reduce the variance of

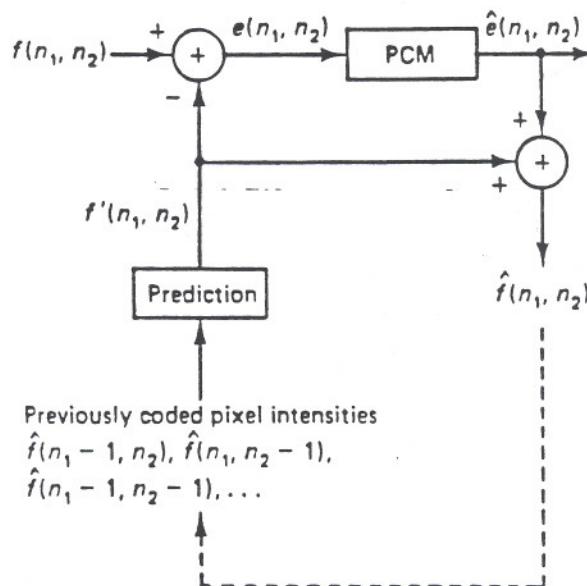


# Differential Pulse Code Modulation (DPCM)

$f(n_1, n_2)$  : original image

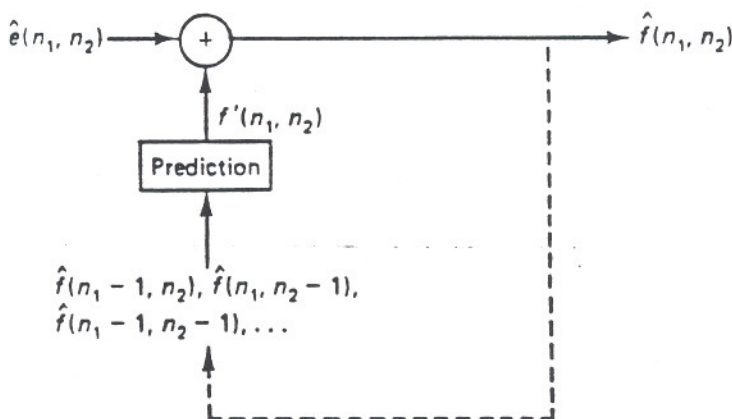
$\hat{f}(n_1, n_2)$  : reconstructed image

Transmitter



- the Auto-regressive Model parameters are obtained from the image by solving a linear set of equations or by a Markov process assumption

Receiver



- requires 2-3 bits/pixel for good quality image

where  $R_a$  is the region of  $(k_1, k_2)$  over which  $a(k_1, k_2)$  is nonzero. Typically,  $f'(n_1, n_2)$  is obtained by linearly combining  $\hat{f}(n_1 - 1, n_2)$ ,  $\hat{f}(n_1, n_2 - 1)$ , and  $\hat{f}(n_1 - 1, n_2 - 1)$ . Since the prediction of  $f(n_1, n_2)$  is made in order to reduce the variance of  $e(n_1, n_2)$ , it is reasonable to estimate  $a(k_1, k_2)$  by minimizing

$$E[e^2(n_1, n_2)] = E[(f(n_1, n_2) - \sum_{(k_1, k_2) \in R_a} a(k_1, k_2) \hat{f}(n_1 - k_1, n_2 - k_2))^2]. \quad (10.40)$$

Since  $\hat{f}(n_1, n_2)$  is a function of  $a(k_1, k_2)$  and depends on the specific quantizer used, solving (10.40) is a nonlinear problem. Since  $\hat{f}(n_1, n_2)$  is the quantized version of  $f(n_1, n_2)$ , and is therefore a reasonable representation of  $f(n_1, n_2)$ , the prediction coefficients  $a(k_1, k_2)$  are estimated by minimizing

$$E[(f(n_1, n_2) - \sum_{(k_1, k_2) \in R_a} a(k_1, k_2) f(n_1 - k_1, n_2 - k_2))^2]. \quad (10.41)$$

Since the function in (10.41) minimized is a quadratic form of  $a(k_1, k_2)$ , solving (10.41) involves solving a linear set of equations in the form of

$$R_f(l_1, l_2) = \sum_{(k_1, k_2) \in R_a} a(k_1, k_2) R_f(l_1 - k_1, l_2 - k_2) \quad (10.42)$$

where  $f(n_1, n_2)$  is assumed to be a stationary random process with the correlation function  $R_f(n_1, n_2)$ . The linear equations in (10.42) are the same as those used in the estimation of the autoregressive model parameters discussed in Chapters 5 and 6.

Figure 10.30 illustrates the performance of a DPCM system. Figure 10.30 shows the result of a DPCM system at 3 bits/pixel. The original image used is the image in Figure 10.22(a). The PCM system used in Figure 10.30 is a nonuniform quantizer. The prediction coefficients  $a(k_1, k_2)$  used to generate the example are



Figure 10.30 Example of differential pulse code modulation (DPCM)-coded image at 3 bits/pixel. Original image used is the image in Figure 10.22(a). NMSE = 2.2%, SNR = 16.6 dB.