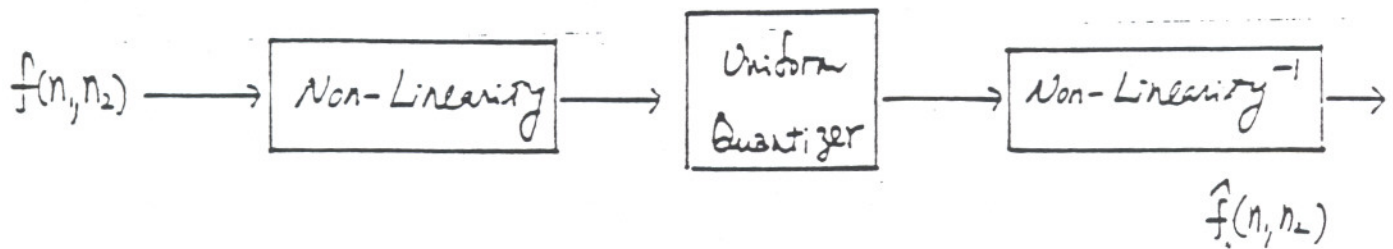
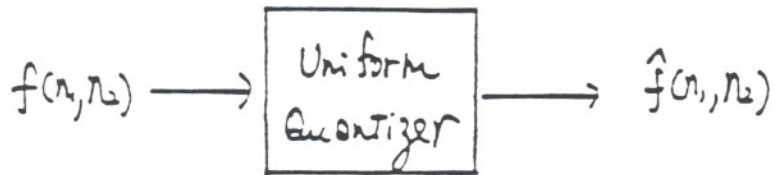


Waveform Coder

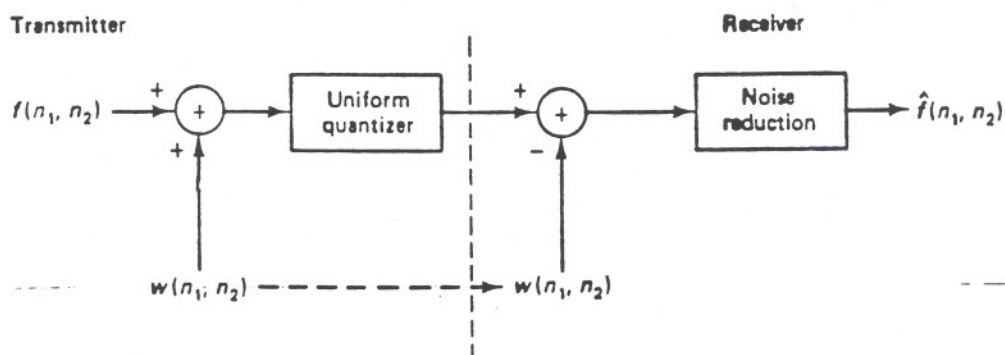
PCM Coding



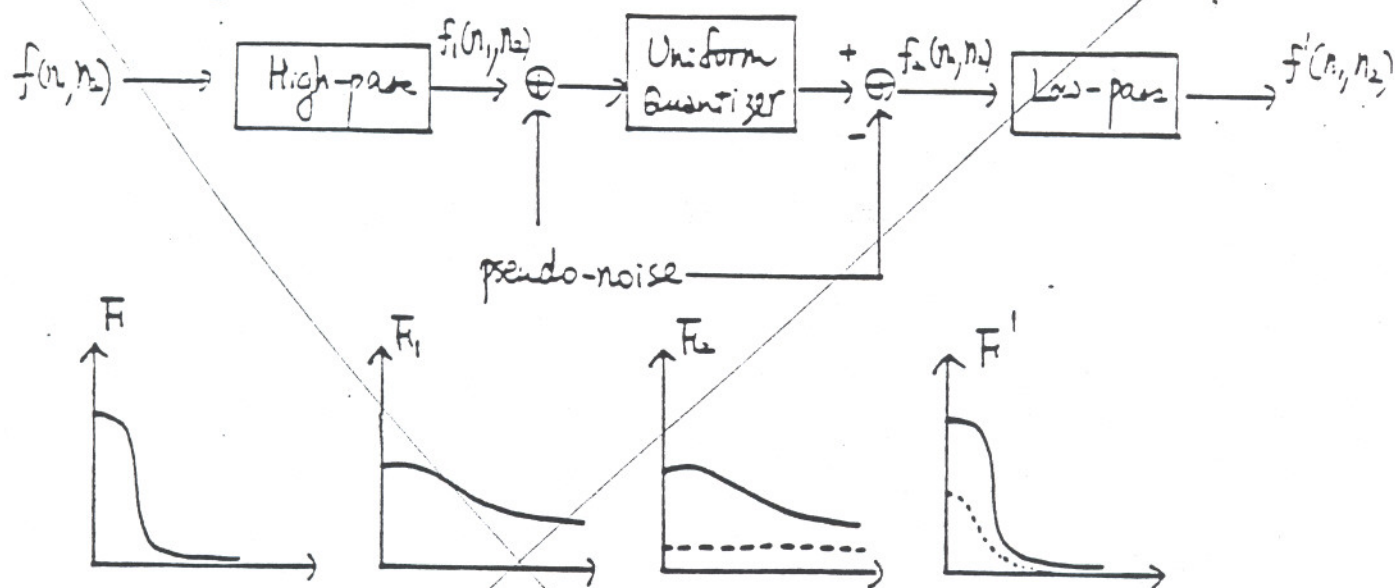
- very simple
- typically requires over 5-6 bits/pixel for good quality
- false contours for low-bit rate case

Improvements of PCM (cont.)

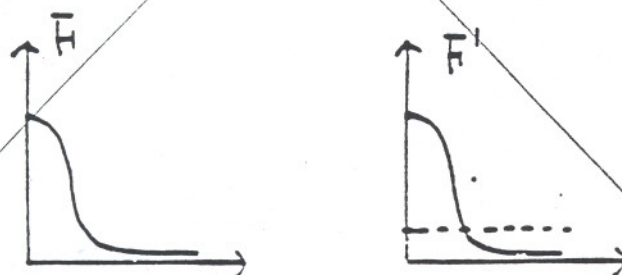
2. Roberts' Pseudo-Noise Technique with Noise Reduction:



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

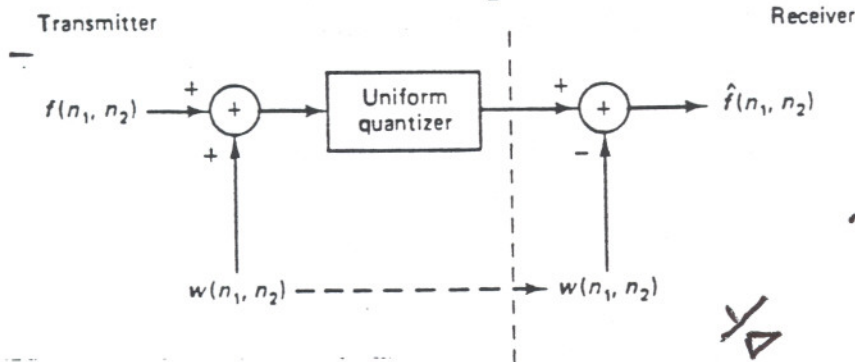


If there were not Highpass/Lowpass Filtering:



Improvements of PCM

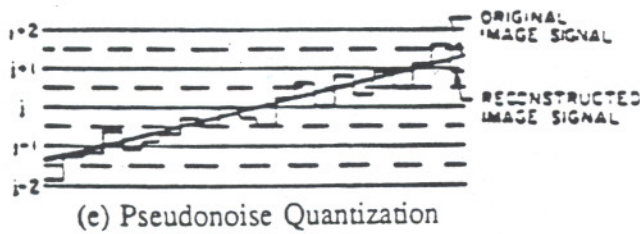
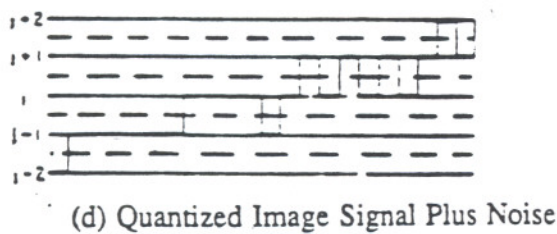
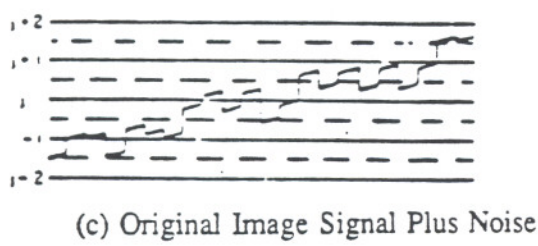
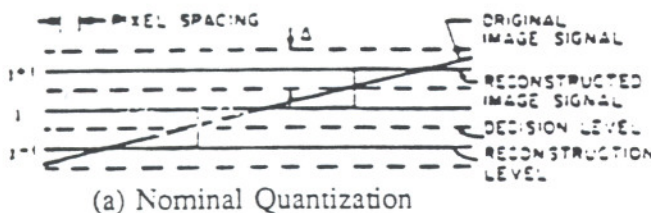
1. Roberts' Pseudo-Noise Technique:



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

otherwise

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



- false contours disappear — replaced by additive random noise

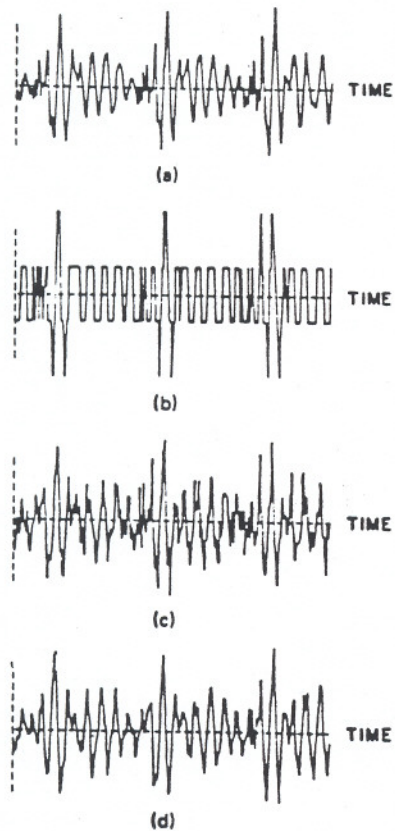


Figure 10.21 Example of quantization noise reduction in PCM speech coding. (a) Segment of noise-free voiced speech; (b) PCM-coded speech at 2 bits/sample; (c) PCM-coded speech at 2 bits/sample by Roberts's pseudonoise technique; (d) PCM-coded speech at 2 bits/sample with quantization noise reduction.



(a)



(b)



(c)



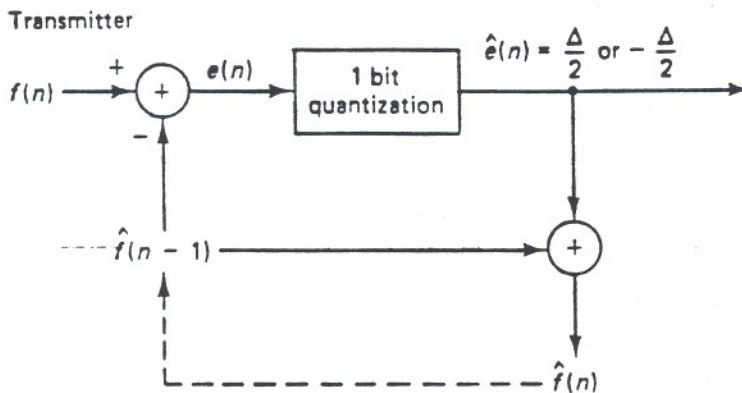
(d)

Figure 10.22 Example of quantization noise reduction in PCM image coding. (a) Original image of 512×512 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

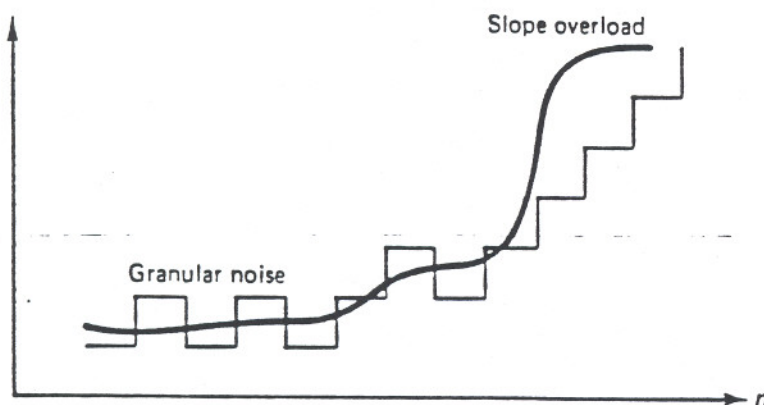
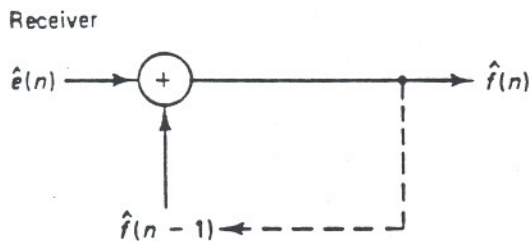


Figure 10.26 Granular noise and slope-overload distortion in delta modulation.

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

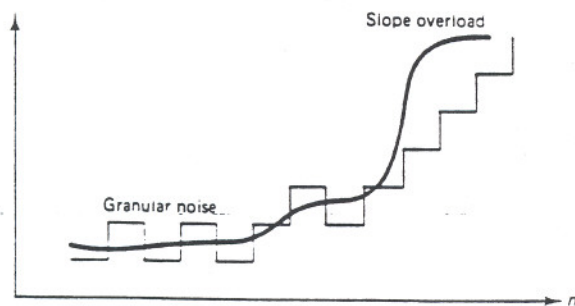


Figure 10.26 Granular noise and slope-overload distortion in delta modulation.

quirements, and the step size Δ 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 512×512 -pixel image in Figure 10.22(a). When Δ 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 Δ [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 Δ 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.

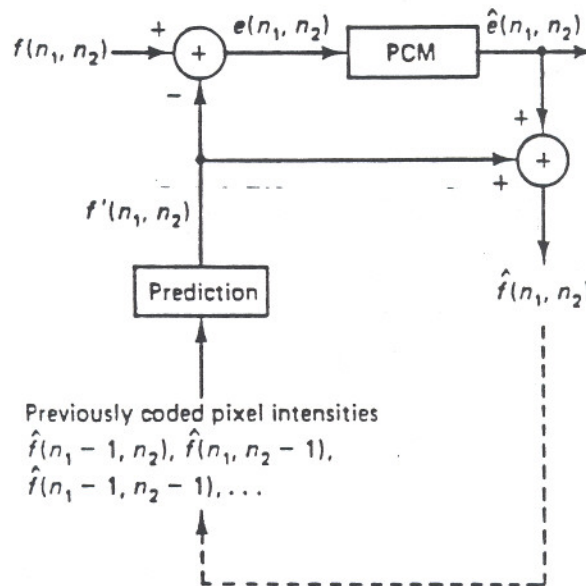
A DPCM system is shown in Figure 10.29. To code the current pixel intensity $f(n_1, n_2)$, $f(n_1, n_2)$ is predicted from previously reconstructed pixel intensities. The predicted value is denoted by $\hat{f}'(n_1, n_2)$. In the figure, we have assumed that $\hat{f}(n_1-1, n_2)$, $\hat{f}(n_1, n_2-1)$,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

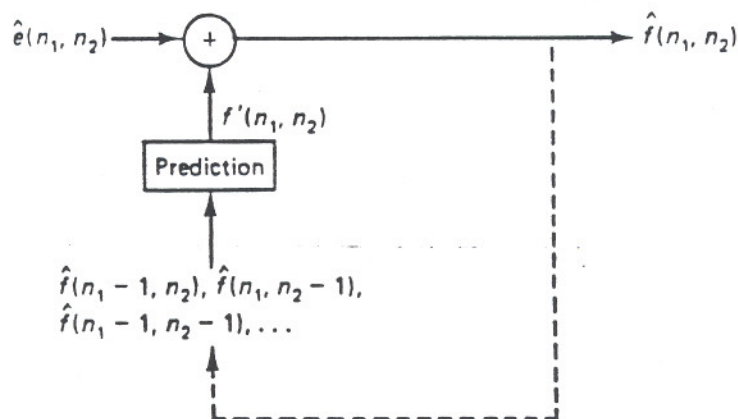
$\hat{f}(n_1, n_2)$: reconstructed image

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.