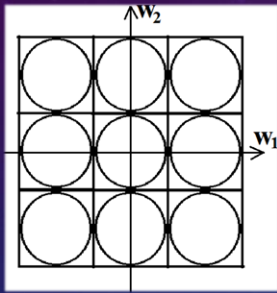


SAMPLING EFFICIENCY

- Rectangular sampling



$$U = \begin{bmatrix} 2f_m & 0 \\ 0 & 2f_m \end{bmatrix}$$

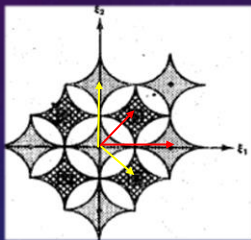
$$U^T V = 2\pi I$$

$$V = \begin{bmatrix} \frac{\pi}{f_m} & 0 \\ 0 & \frac{\pi}{f_m} \end{bmatrix} = \begin{bmatrix} \frac{T}{2} & 0 \\ 0 & \frac{T}{2} \end{bmatrix}$$

$$\frac{2\pi}{T} = f_m$$

SAMPLING EFFICIENCY

- A different sampling strategy more efficient for other spectral shapes
- Rhombus



$$U = \begin{bmatrix} 2f_m & f_m \\ 0 & f_m \end{bmatrix}$$

$$U^T V = 2\pi I$$

$$\frac{2\pi}{T} = f_m$$

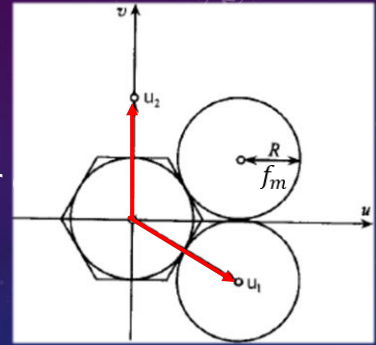
$$V = \begin{bmatrix} \frac{f_m}{\pi} & 0 \\ \frac{f_m}{2\pi} & \frac{f_m}{2\pi} \end{bmatrix} = I$$

$$V = \begin{bmatrix} \frac{\pi}{f_m} & 0 \\ \frac{\pi}{f_m} & \frac{2\pi}{f_m} \end{bmatrix} = \begin{bmatrix} \frac{T}{2} & 0 \\ -\frac{T}{2} & T \end{bmatrix}$$

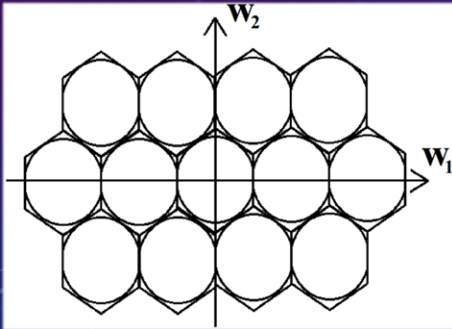
$$\begin{bmatrix} \frac{\sqrt{3}f_m}{2\pi} & -\frac{f_m}{2\pi} \\ 0 & \frac{f_m}{\pi} \end{bmatrix} V = I$$

SAMPLING EFFECIENCY

- A different sampling strategy more efficient for shapes
 - Disk



$$U^T V = 2\pi I$$

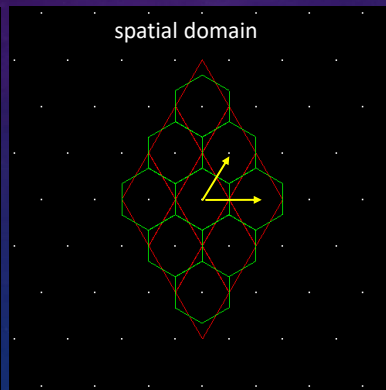
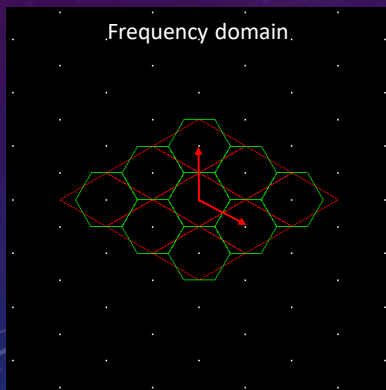


$$U = \begin{bmatrix} \sqrt{3}f_m & 0 \\ -f_m & 2f_m \end{bmatrix}$$

$$V = \begin{bmatrix} \frac{2\pi}{\sqrt{3}f_m} & \frac{\pi}{\sqrt{3}f_m} \\ 0 & \frac{\pi}{f_m} \end{bmatrix} = \begin{bmatrix} \frac{T}{\sqrt{3}} & \frac{T}{2\sqrt{3}} \\ 0 & \frac{T}{2} \end{bmatrix}$$

SAMPLING EFFICIENCY

- Hexagonal sampling



$$U = \begin{bmatrix} \sqrt{3}f_m & 0 \\ -f_m & 2f_m \end{bmatrix}$$

$$V = \begin{bmatrix} \frac{2\pi}{\sqrt{3}f_m} & \frac{\pi}{\sqrt{3}f_m} \\ 0 & \frac{\pi}{f_m} \end{bmatrix} = \begin{bmatrix} \frac{T}{\sqrt{3}} & \frac{T}{2\sqrt{3}} \\ 0 & \frac{T}{2} \end{bmatrix}$$

SAMPLING EFFICIENCY

- **Sampling density**
 - Lossless condition
 - Number of samples per unit area
- Sampling density is inversely proportional to $\det(V)$
- For rectangular sampling

$$S.D. = \frac{1}{\det(V)} = \frac{\det(U)}{4\pi^2}$$

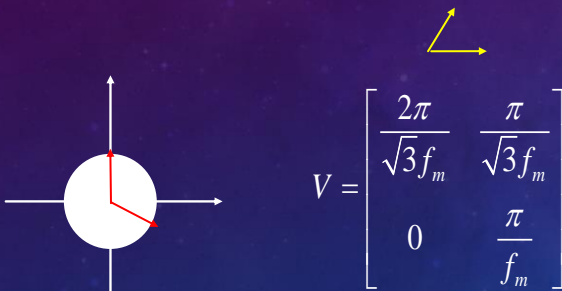
$$V = \begin{bmatrix} \frac{\pi}{f_m} & 0 \\ 0 & \frac{\pi}{f_m} \end{bmatrix}$$

$$\det(V) = |V| = \frac{\pi^2}{f_m^2}$$

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SAMPLING EFFICIENCY

- Hexagonal sampling



$$V = \begin{bmatrix} \frac{2\pi}{\sqrt{3}f_m} & \frac{\pi}{\sqrt{3}f_m} \\ 0 & \frac{\pi}{f_m} \end{bmatrix}$$

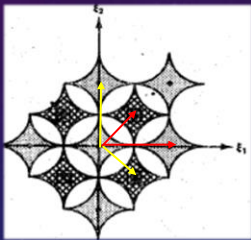
$$\det(V) = |V| = \frac{2\pi^2}{\sqrt{3}f_m^2}$$

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$$\rho_{H/R} = \frac{|V_R|}{|V_H|} = \frac{\sqrt{3}}{2} \approx 0.866 \quad \frac{2}{\sqrt{3}} \approx 1.155$$

SAMPLING EFFICIENCY

- Interlaced sampling



$$V = \begin{bmatrix} \frac{\pi}{f_m} & 0 \\ -\frac{\pi}{f_m} & \frac{2\pi}{f_m} \end{bmatrix}$$

$$U = \begin{bmatrix} 2f_m & f_m \\ 0 & f_m \end{bmatrix}$$

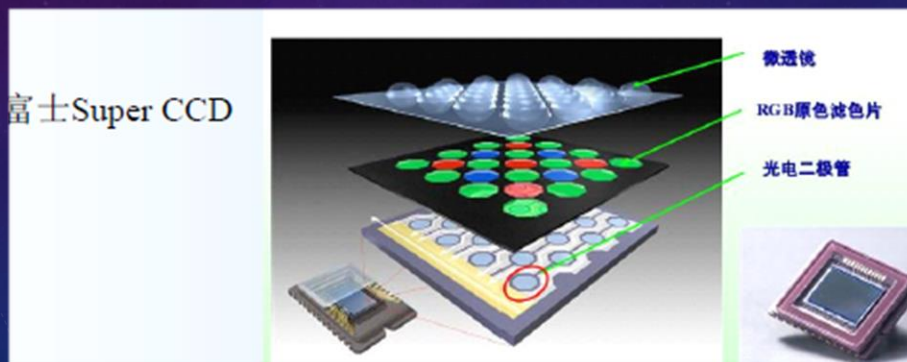
$$\det(V) = |V| = \frac{2\pi^2}{f_m^2}$$

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$$\rho_{I/R} = \frac{|V_R|}{|V_I|} = \frac{1}{2}$$

SAMPLING ON NON-RECTANGULAR LATTICES

- Non-rectangular lattices hexagonal
- Complex reconstruction function



SAMPLING & RECONSTRUCTION - LIMITS

- overlapping
- LPF before sampling

Ideal
bandlimited
signal



- Reconstruction function is impractical, since sinc function has small attenuation(衰减)
- Simple approximation techniques are employed

Ideal low pass
reconstruction
Filter



- Sampling aperture --- loss of high frequency component

Ideal 2-D
function



- Practical sampling and reconstruction of an image **is different**

QUANTIZATION

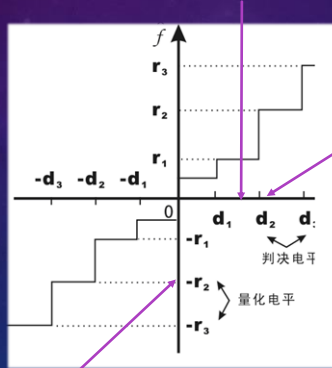
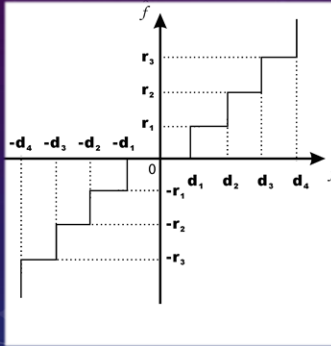
- Uniform
- Symmetric
- With Memory
- Scalar

- Non-uniform
- Asymmetric
- Memoryless
- Vector

The conversion process between analog samples and discrete-valued samples is called quantization

QUANTIZATION

quantization interval $\Delta_i = d_{i+1} - d_i$



$$d \in [d_i, d_{i+1})$$

$$d \rightarrow r_i$$

decision level d_i

quantization error $e = d - Q(d)$

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quantization noise $\varepsilon^2 = E[e^2]$

reconstruction level r_i

QUANTIZATION

- Scalar quantization ---- For a given number of reconstruction levels k , design a quantizer and minimize its loss
 - Optimal mean square error quantizer(最优均方误差量化器) ---- minimum MSE
- Random variable z , with continuous PDF $p(z)$, quantized as q_i

$$\varepsilon^2 = \sum_{i=0}^{k-1} \int_{z_i}^{z_{i+1}} (z - q_i)^2 p(z) dz \rightarrow \begin{cases} \frac{\partial \varepsilon^2}{\partial q_i} = 0 \\ \frac{\partial \varepsilon^2}{\partial z_i} = 0 \end{cases}$$

$$\text{changshu@}q_i = \frac{\int_{z_i}^{z_{i+1}} zp(z)dz}{\int_{z_i}^{z_{i+1}} p(z)dz}$$

$$z_i = \frac{q_{i-1} + q_i}{2}$$

QUANTIZATION

- Scalar quantization

- uniform (linear) quantization

Separate $[z_0, z_k]$ into k sub-intervals, each of which is of the same length, uniform quantization is the optimal quantization if z is uniformly distributed.

$$L = (z_k - z_0) / k$$

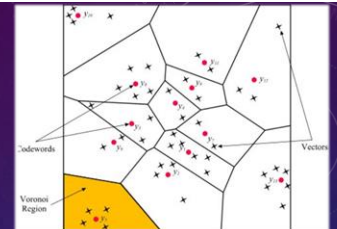
$$q_i = (z_i + z_{i+1}) / 2$$

$$z_i = (q_{i-1} + q_i) / 2$$

$$p(z) = \begin{cases} \frac{1}{(z_k - z_0)}, & z_0 \leq z < z_k \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} \varepsilon^2 &= \sum_{i=0}^{k-1} \int_{z_i}^{z_{i+1}} (z - q_i)^2 p(z) dz \\ &= \sum_{i=0}^{k-1} \int_{z_i}^{z_{i+1}} (z - q_i)^2 \frac{1}{kL} dz = \frac{L^2}{12} \end{aligned}$$

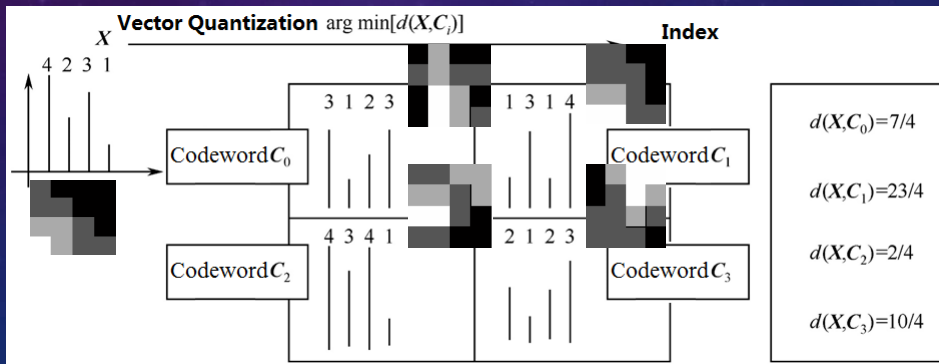
QUANTIZATION



- Vector Quantization(VQ) ---- Also called *block quantization* or *pattern matching quantization*, works by **encoding** values from a MD vector space into a **finite** set of values from a **discrete** subspace of lower dimension
- A vector in subspace requires **less storage space**, so the data is compressed. Due to the density matching property of vector quantization, the compressed data has errors that are inversely proportional to density
- The transformation is usually done by **projection** or by using a **codebook**

QUANTIZATION

- Vector Quantization(VQ) ---- works by **encoding** values from a MD vector space into a **finite** set of values from a **discrete** subspace of lower dimension.



QUANTIZATION

- Optimal matching: minimize the expected absolute/squared quantization error
 - For a random variable X with PDF $p(X)$

Expected Quantization Error: $\varepsilon = \sum_i e(X, X_i) p(X)$

$$e(X, X_i) = \|X - X_i\|^j \quad j = 1, 2, \dots, \infty$$

$$MAE: e(X, X_i) = \frac{1}{k} \sum_{m=1}^k |x(m) - \hat{x}_i(m)|$$

$$MSE: e(X, X_i) = \frac{1}{k} \sum_{m=1}^k [x(m) - \hat{x}_i(m)]^2$$

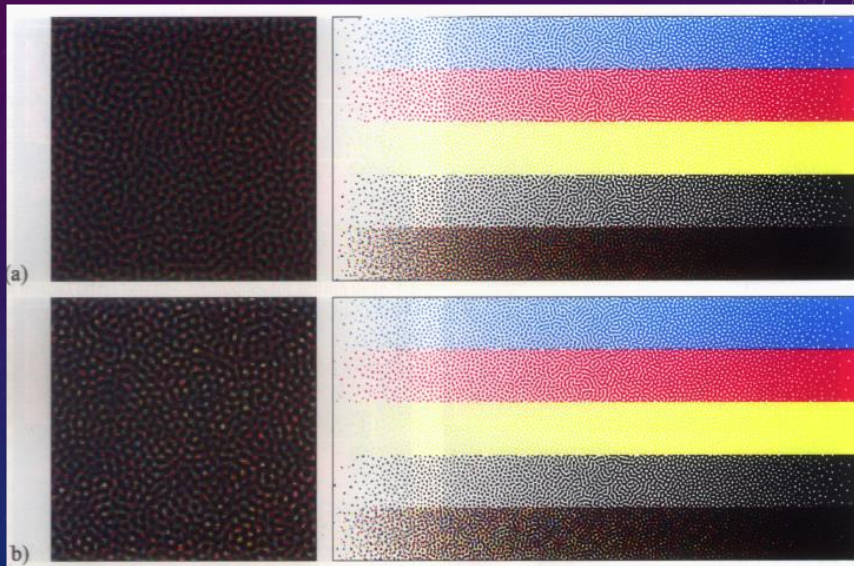
Taxicab-norm
 Euclidean-norm
 p-norm
 Infinite-norm

HALFTONING

- Halftoning
 - Mainly used in printing/reprography(复印)
 - The halftone process reduces visual reproductions to an image that is printed with only one color of ink
 - Question: try to convert a grayscale image into a binary one(black and white).
 - This reproduction relies on a basic **optical illusion**: the tiny halftone dots are blended into smooth tones by the human eye.



CMYK color plane



The Density of tiny dots are different in these color ramps

Color printing



HALFTONING

- One color of ink -> Grayscale image
- Optic illusion
- Perceptual grayscale ----- ratio of the inked pixels within a cell
- Resolution is sacrificed

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DITHER

cyclical

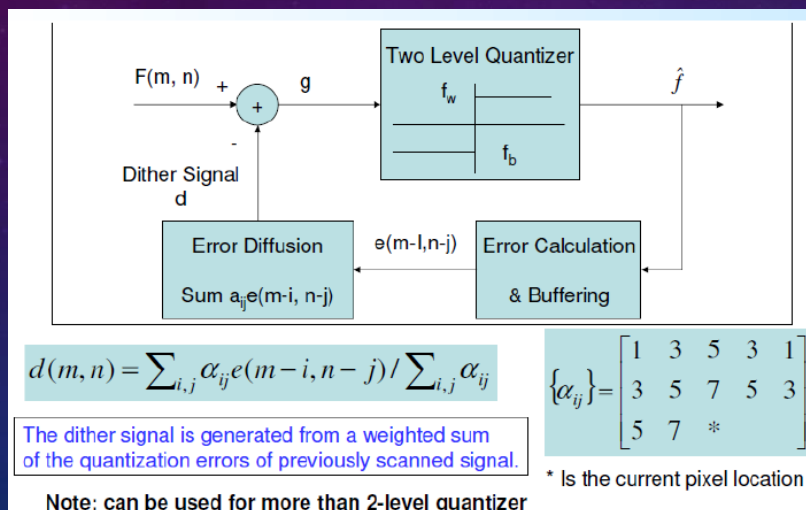
- Dither is an intentionally applied form of **noise** used to randomize quantization error
- Dither signal $d(m,n)$ could be obtained by applying dither matrix $D_k(m,n)$ repeatedly.

$$D^4 = \begin{bmatrix} 8 & 136 & 40 & 168 \\ 200 & 72 & 232 & 104 \\ 56 & 184 & 24 & 152 \\ 248 & 120 & 216 & 88 \end{bmatrix}$$

52	44	36	124	132	140	148	156
60	4	28	116	200	228	236	164
68	12	20	108	212	252	244	172
76	64	92	100	204	196	188	180
132	140	148	156	52	44	36	124
200	228	236	164	60	4	28	116
212	252	244	172	68	12	20	108
204	196	188	180	76	64	92	100

Figure 4.25 Two halftone patterns. Repeat periodically to obtain the full size array. H_2 is called 45° halftone screen because it repeats two 4×4 basic patterns at $\pm 45^\circ$ angles.

HALFTONING USING ERROR DIFFUSION



DITHER

Floyd-Steinberg error diffusion



DITHER

Original Image



Dithered Index Image



SUMMARY

- Digitalization
 - Sampling (rectangular/hexagonal, sampling density/efficiency)
 - Quantization (vector quantization, halftone, dither)
- To balance sampling and quantization, we have to take the image content into consideration

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