

信号理論

- No.9 A/D変換と量子化 -

渡辺 裕

Signal Theory

- No.9 A/D Conversion and Quantization -

Hiroshi Watanabe

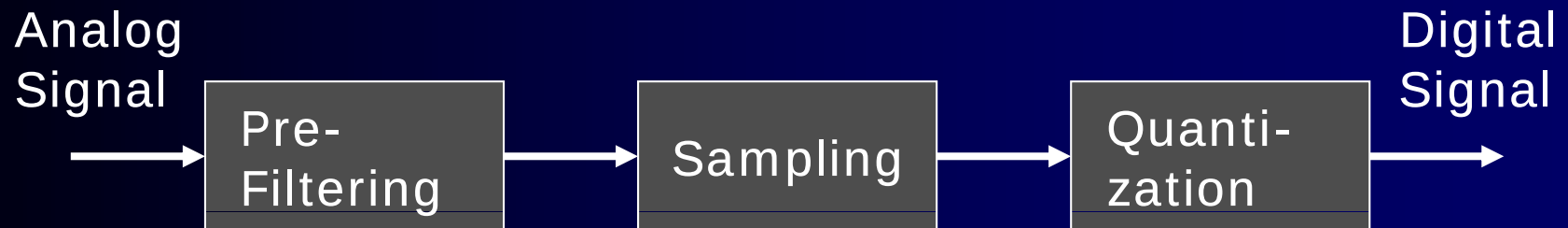
A/D 変換

- アナログ信号からデジタル信号へ
- プレフィルタ処理: 高周波成分の除去
- 標本化: 時間軸における離散化
- 量子化: レベル値の有限 n 進数への変換



A/D Conversion

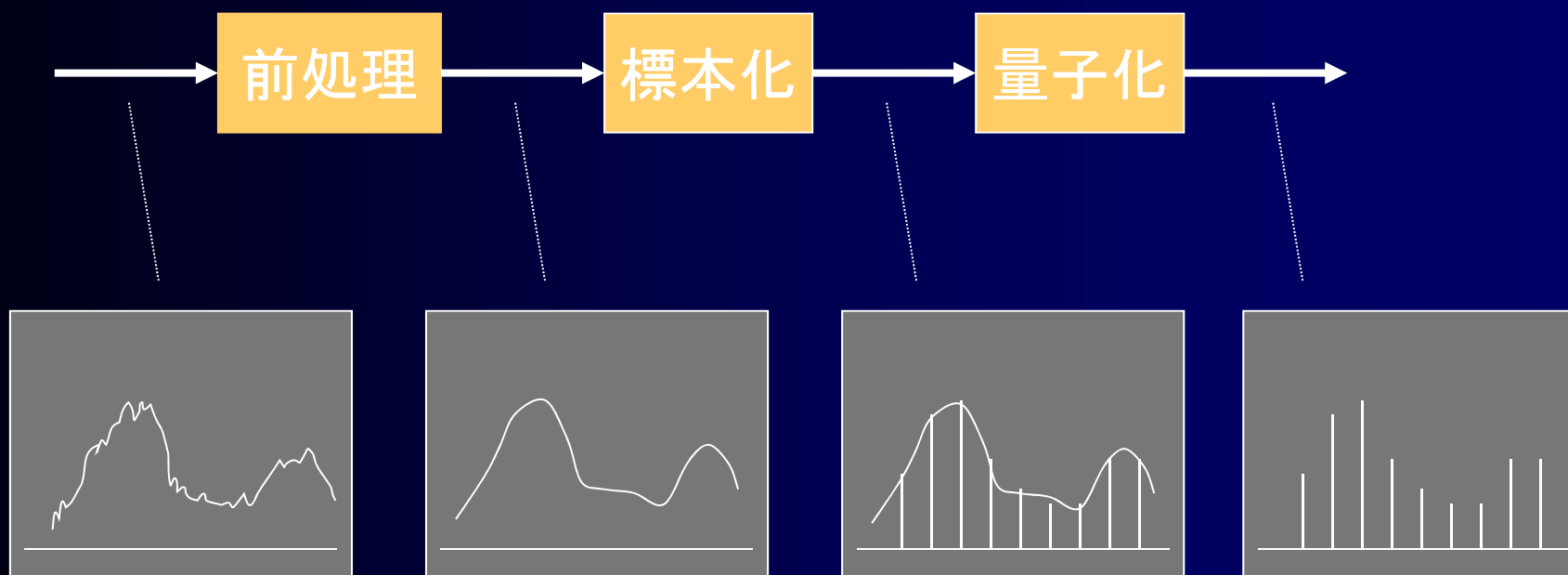
- Analog Signal to Digital Signal
- Pre-filtering: Cut High Frequency Component
- Sampling: Discrete in Time Domain
- Quantization: Convert Value to Finite Digits



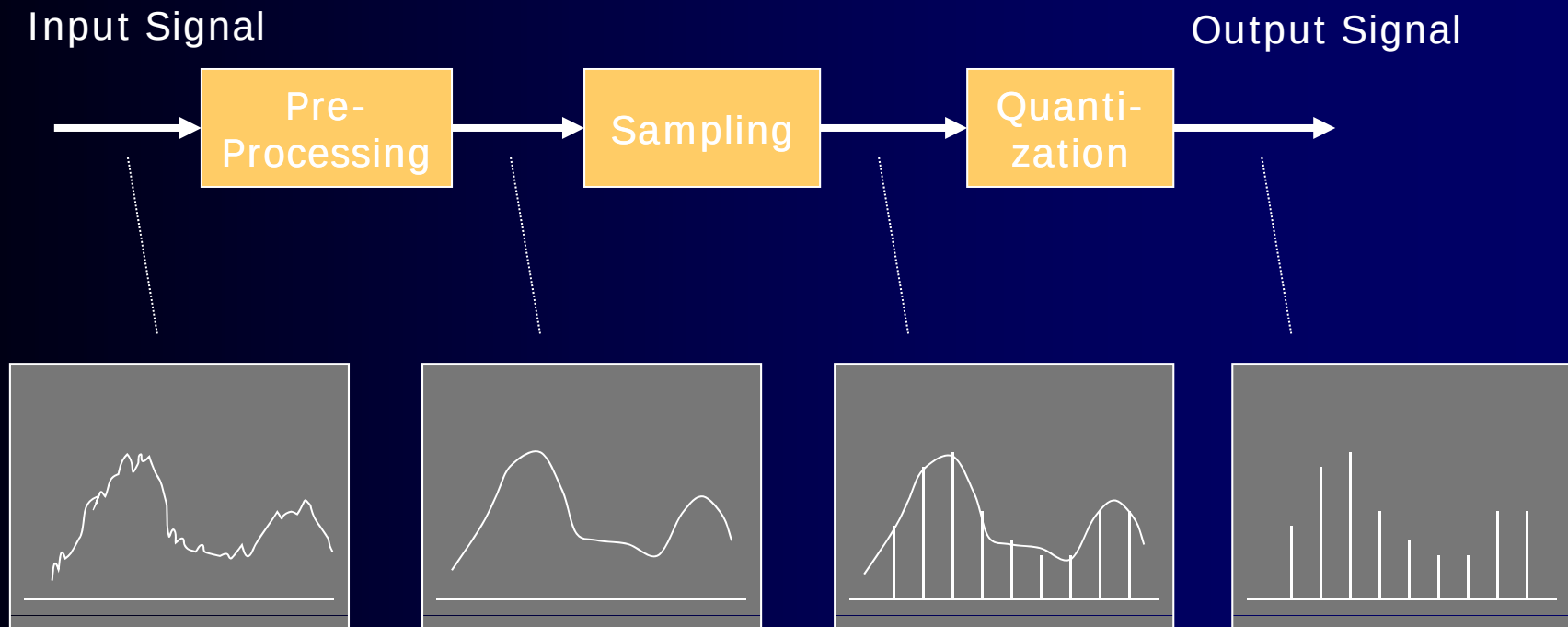
処理ダイアグラム

入力信号

出力信号

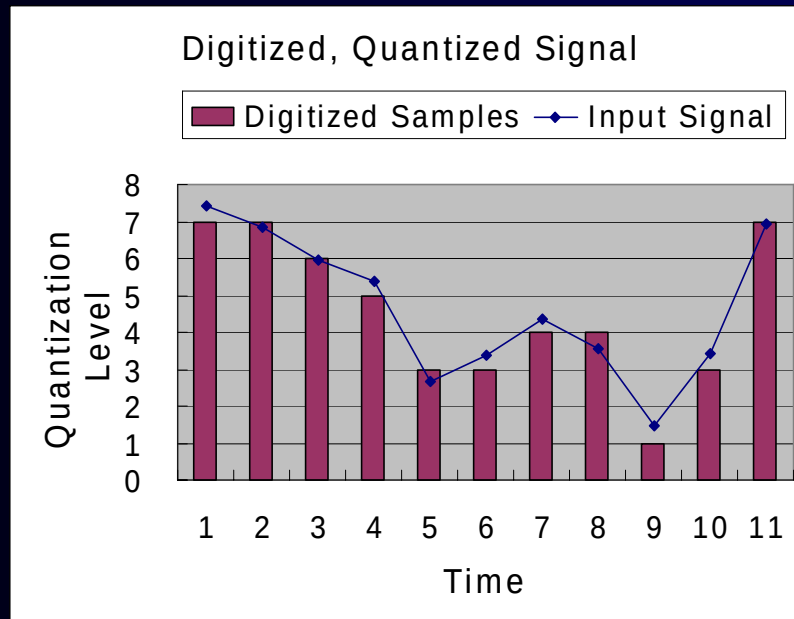


Processing Diagram

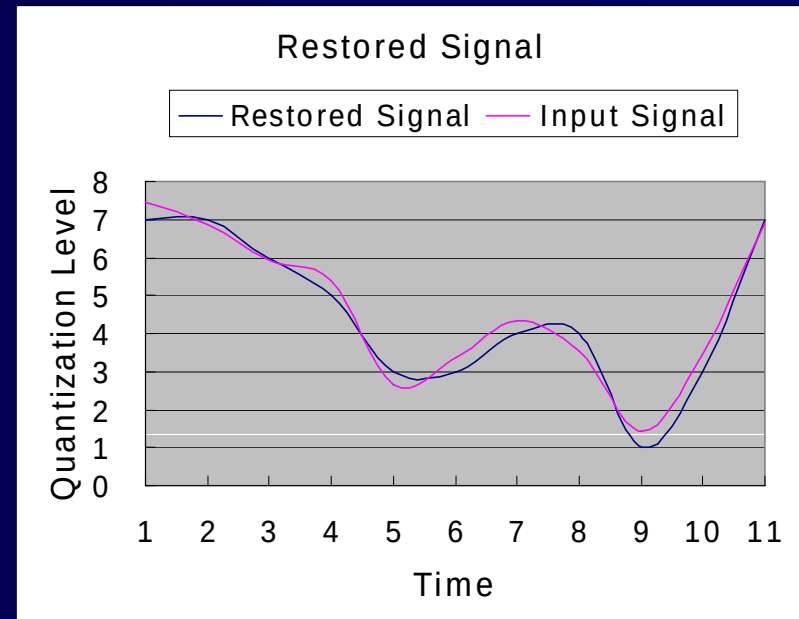


PCM

■ PAMデータへのユニーク符号の割り当て



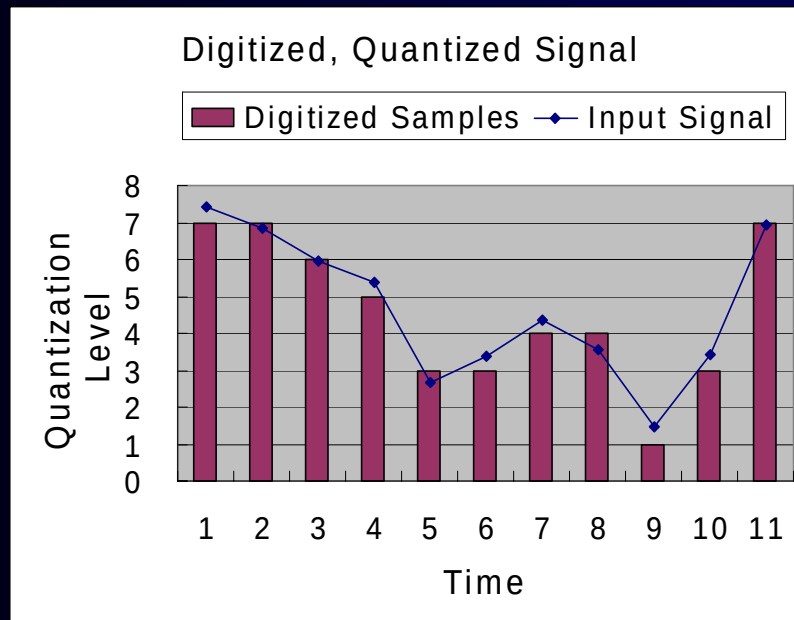
離散化・量子化信号



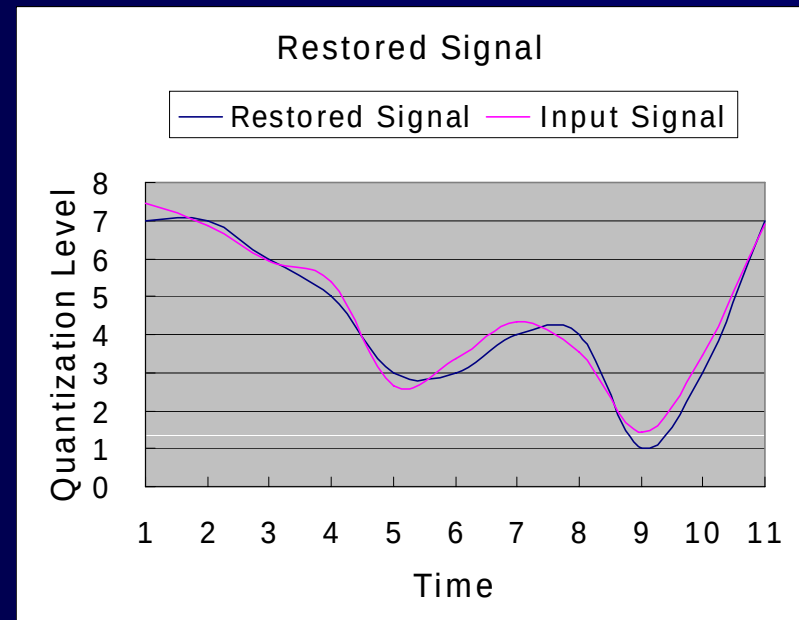
再生信号

PCM

■ Unique Code Assignment to PAM Data



Digitized and quantized signal



Restored signal

2進数と10進数

■ 2進数表現

$$\begin{aligned} b_n b_{n-1} \cdots b_1 b_0 . b_{-1} b_{-2} \cdots b_{-m} \Big|_2 &= \frac{b_n 2^n + b_{n-1} 2^{n-1} + \cdots + b_1 2^1 + b_0 2^0}{+ b_{-1} 2^{-1} + \cdots + b_{-m} 2^{-m}} \Big|_{10} \\ &= \sum_{i=-m}^n b_i 2^i \Big|_{10}, \quad b_i = \begin{cases} 0 \\ 1 \end{cases} \end{aligned}$$

■ 例

$$10110 = 2^4 + 2^2 + 2^1 = 16 + 4 + 2 = 22$$

Binary and Decimal Numbers

■ Binary Representation

$$\begin{aligned} b_n b_{n-1} \cdots b_1 b_0 . b_{-1} b_{-2} \cdots b_{-m} \Big|_2 &= \frac{b_n 2^n + b_{n-1} 2^{n-1} + \cdots + b_1 2^1 + b_0 2^0}{+ b_{-1} 2^{-1} + \cdots + b_{-m} 2^{-m}} \Big|_{10} \\ &= \sum_{i=-m}^n b_i 2^i \Big|_{10}, \quad b_i = \begin{cases} 0 \\ 1 \end{cases} \end{aligned}$$

■ Example

$$10110 = 2^4 + 2^2 + 2^1 = 16 + 4 + 2 = 22$$

2進符号

■ レベル数とビット数

Level数	Bit	Code
2	1	0, 1
4	2	00, 01, 10, 11
8	3	000, 001, 010, 011, 100, 101, 110, 111
...		
2^n	n	00...0, ...

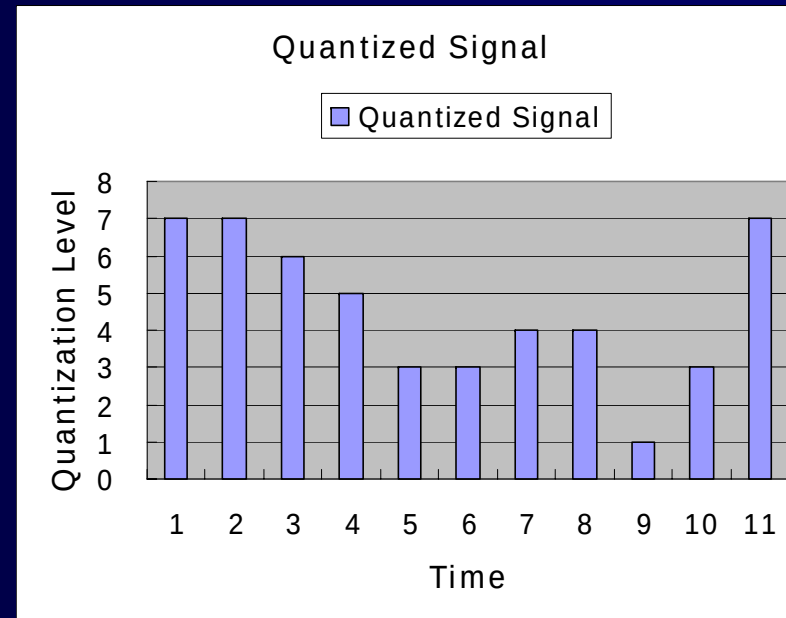
Binary Code

■ Level and Bit Numbers

Level#s	Bit	Code
2	1	0, 1
4	2	00, 01, 10, 11
8	3	000, 001, 010, 011, 100, 101, 110, 111
...		
2^n	n	00...0, ...

PCM符号の割り当て

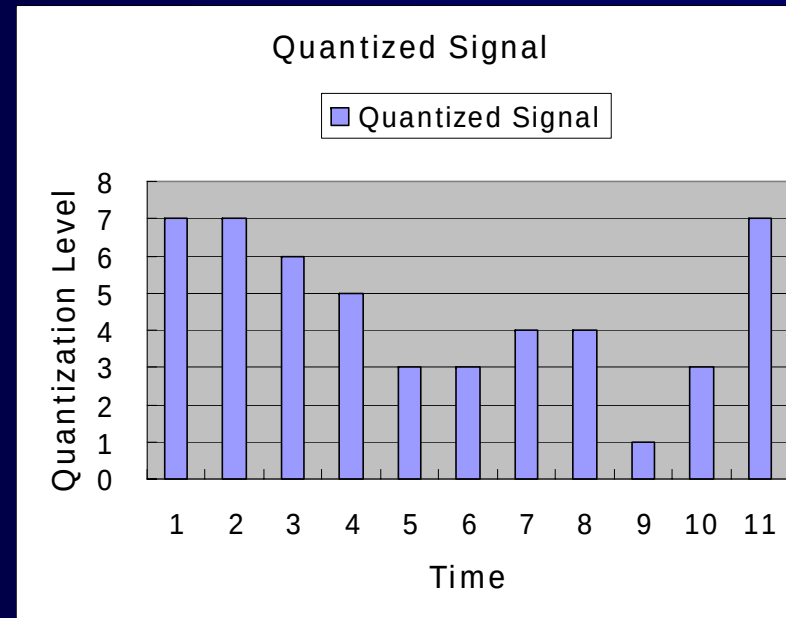
Level	Codeword
0	000
1	001
2	010
3	011
4	100
5	101
6	110
7	111



時刻:	1	2	3	4	5	6	7	8	9	10	11
レベル:	7	7	6	5	3	3	4	4	1	3	7
符号:	111	111	110	101	011	011	100	100	001	011	111 / 33 bit

Assignment of PCM Code

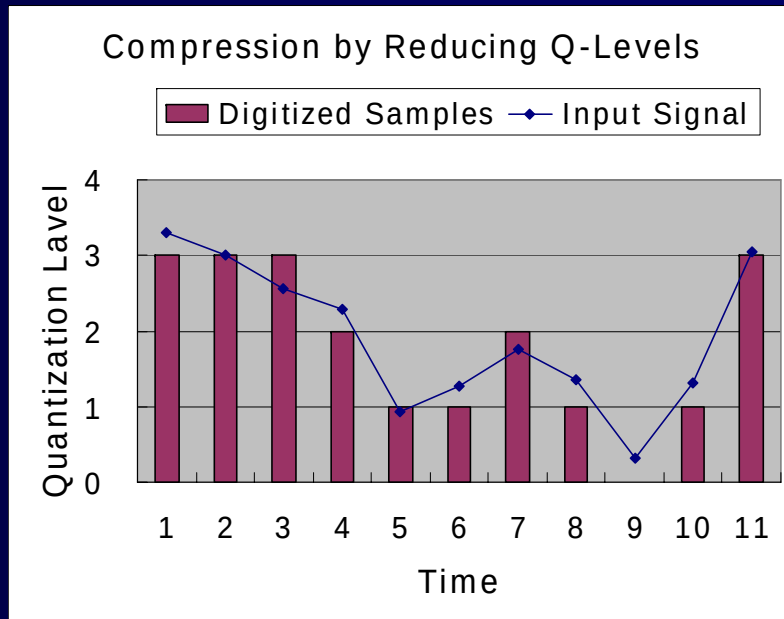
Level	Codeword
0	000
1	001
2	010
3	011
4	100
5	101
6	110
7	111



Time:	1	2	3	4	5	6	7	8	9	10	11
Level:	7	7	6	5	3	3	4	4	1	3	7
Code:	111	111	110	101	011	011	100	100	001	011	111 / 33 bit

基本的なデータ圧縮 — 量子化レベル数の削減

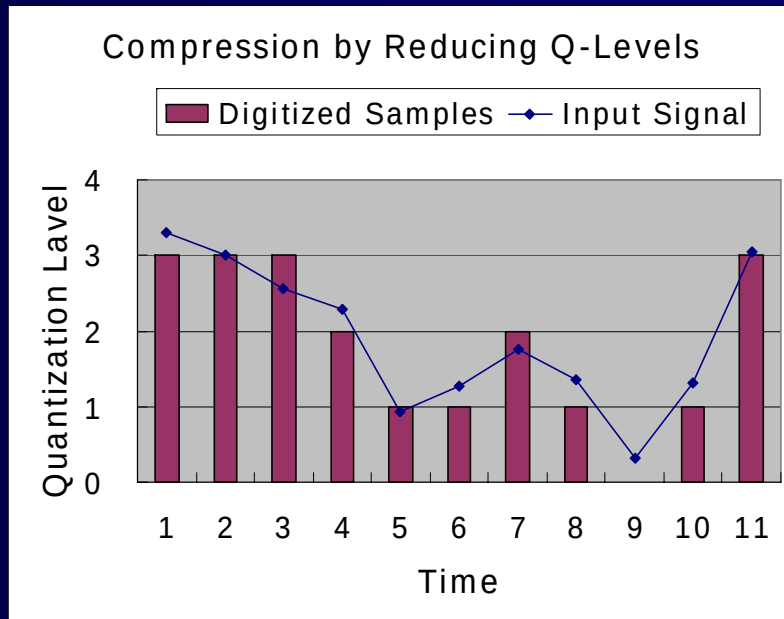
Level	Codeword
0	00
1	01
2	10
3	11



時刻:	1	2	3	4	5	6	7	8	9	10	11
レベル:	3	3	3	2	1	1	2	1	0	1	3
符号:	11	11	11	10	01	01	10	10	00	10	11 / 22bit

Basic Data Compression – Reducing Number of Quantization Levels

Level	Codeword
0	00
1	01
2	10
3	11

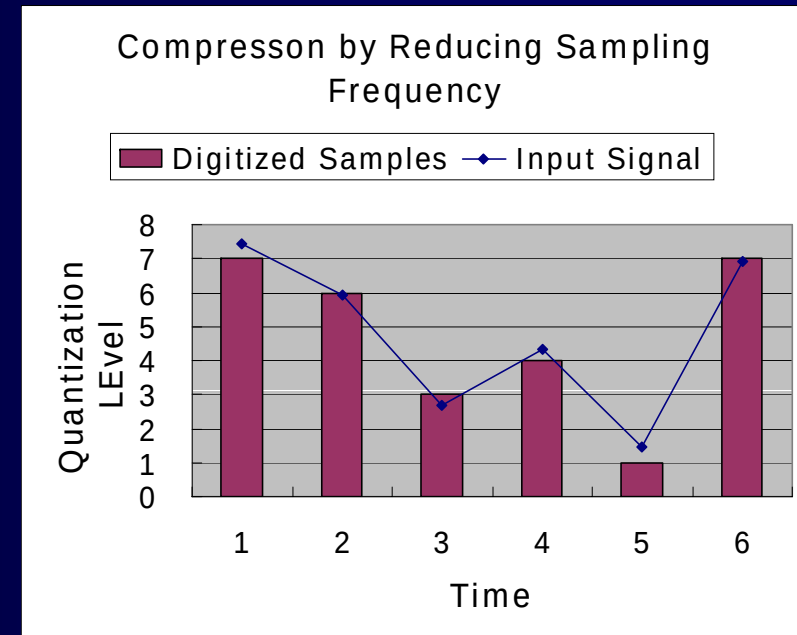


Time:	1	2	3	4	5	6	7	8	9	10	11
Level:	3	3	3	2	1	1	2	1	0	1	3
Code:	11	11	11	10	01	01	10	10	00	10	11 / 22bit

基本的なデータ圧縮 — 標本点数の削減

■	Level	Codeword
	0	000
	1	001
	2	010
	3	011
	4	100
	5	101
	6	110
	7	111

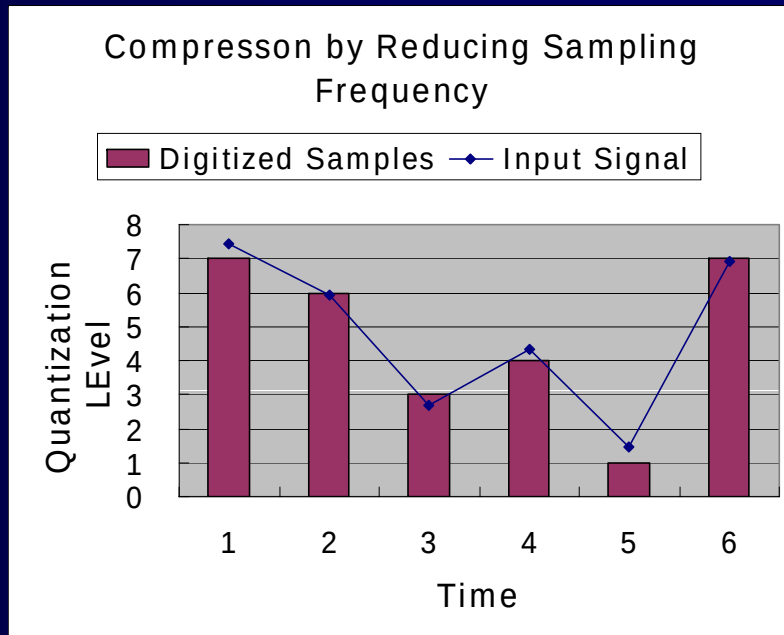
時刻:	1	2	3	4	5	6
レベル:	7	6	3	4	1	7
符号:	111	110	011	100	001	111 / 18 bit



Basic Data Compression – Reducing Number of Samples

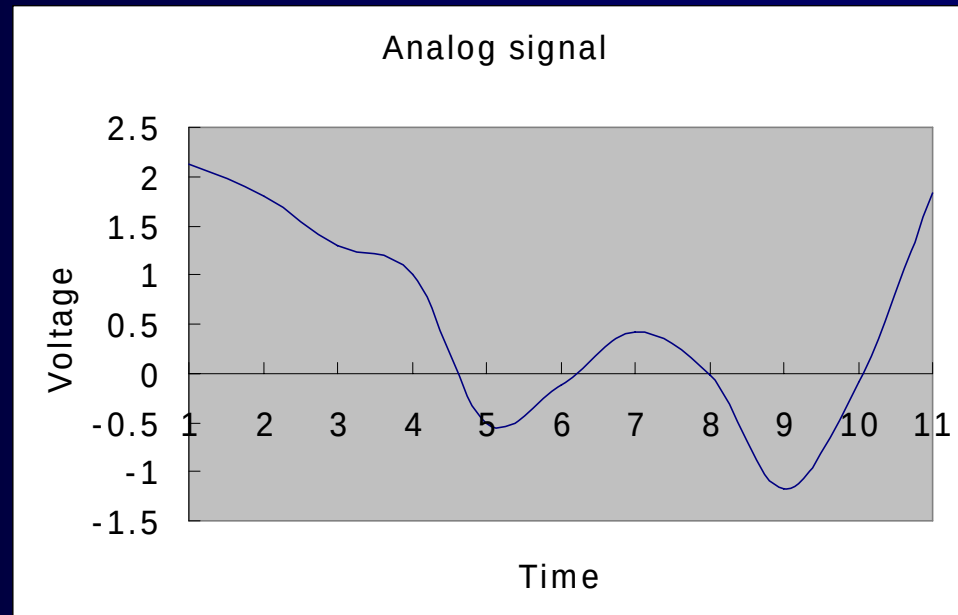
Level	Codeword
0	000
1	001
2	010
3	011
4	100
5	101
6	110
7	111

Time:	1	2	3	4	5	6
Level:	7	6	3	4	1	7
Code:	111	110	011	100	001	111 / 18 bit



定量的解析

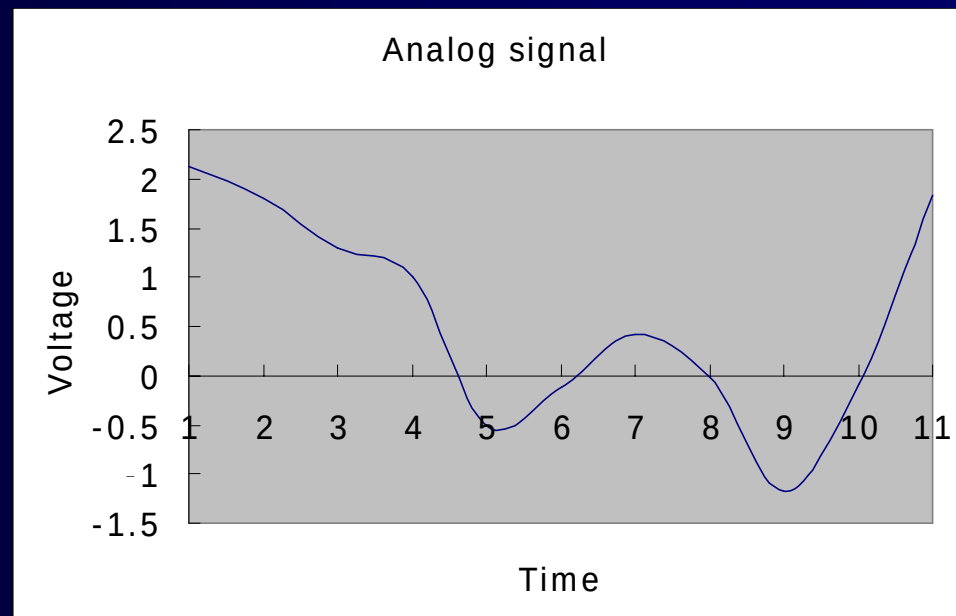
時刻	アナログ信号値
0	2.12
1	1.80
2	1.30
3	1.00
4	-0.50
5	-0.12
6	0.42
7	-0.02
8	-1.17
9	-0.08
10	1.84



Quantitative Analysis

Time Analogue signal Value

0	2.12
1	1.80
2	1.30
3	1.00
4	-0.50
5	-0.12
6	0.42
7	-0.02
8	-1.17
9	-0.08
10	1.84



定量的解析 (2)

- 量子化ステップ d の算出

$$d = \frac{2V}{L}$$

V : 入力電圧の最大値

L : 量子化レベル数

$$L = 2^n \quad n: \text{符号のビット数}$$

- 例

$V=2.2, n=4$ のとき $L=16$ であるから

$$d = 0.275$$

Quantitative Analysis (2)

- Calculation of Quantization step d

$$d = \frac{2V}{L}$$

V : Maximum value of input

L : number of quantization levels

$$L = 2^n$$

n : number of bits for code

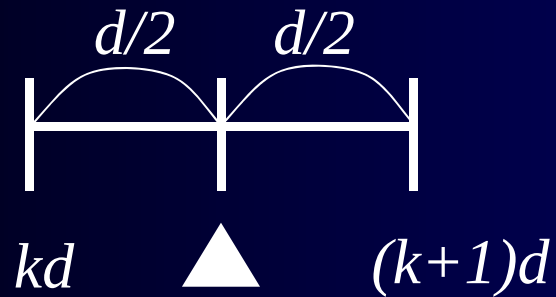
- Example

When $V=2.2$, $n=4$, L equals to 16, thus,

$$d = 0.275$$

定量的解析 (3)

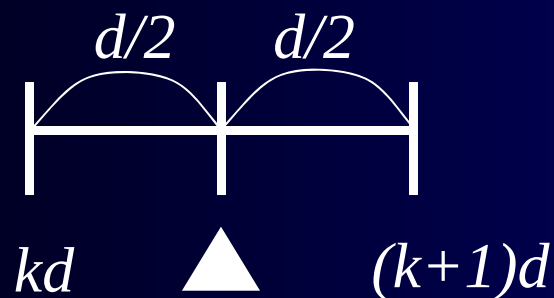
- 量子化誤差電力: σ_q^2



$$\sigma_q^2 = \frac{1}{d} \int_{-d/2}^{d/2} x^2 dx = \frac{1}{d} \left[\frac{1}{3} x^3 \right]_{-d/2}^{d/2} = d^2 / 12$$

Quantitative Analysis (3)

- Quantization Error Power: σ_q^2



$$\sigma_q^2 = \frac{1}{d} \int_{-d/2}^{d/2} x^2 dx = \frac{1}{d} \left[\frac{1}{3} x^3 \right]_{-d/2}^{d/2} = \frac{d^2}{12}$$

定量的解析 (4)

- 量子化によるSNR(雑音対信号比)

$$SNR(dB) = 20 \log_{10} \frac{\sigma_x}{\sigma_q} = 10 \log_{10} \frac{\sigma_x^2}{\sigma_q^2}$$

- 入力信号の分散: σ_x^2 (画像処理では最大値)

$$\sigma_x^2 = 4V^2$$

Quantitative Analysis (4)

- SNR (Signal to Noise Ratio) caused by Quantization

$$SNR(dB) = 20 \log_{10} \frac{\sigma_x}{\sigma_q} = 10 \log_{10} \frac{\sigma_x^2}{\sigma_q^2}$$

- Variance of Input signal: σ_x^2 (maximum value for image processing)

$$\sigma_x^2 = 4V^2$$

定量的解析 (5)

- n ビット量子化時のSNRの算出

$$\begin{aligned} \text{SNR}(dB) &= 10 \log_{10} \frac{4V^2}{d^2/12} \\ &= 10 \log_{10} \frac{4V^2}{4V^2/12L^2} \\ &= 10 \log_{10} 12L^2 \\ &= 10(\log_{10} 3 + 2 \log_{10} 2) + 20 \log_{10} 2^n \\ &= 10.79 + 6.02n \end{aligned}$$

$$\begin{aligned} \log_{10} 2 &\cong 0.301 \\ \log_{10} 3 &\cong 0.477 \end{aligned}$$

1bit 6dBの法則

Quantitative Analysis (5)

- Calculation of SNR at n bit quantization

$$\begin{aligned} \text{SNR}(dB) &= 10 \log_{10} \frac{4V^2}{d^2/12} \\ &= 10 \log_{10} \frac{4V^2}{4V^2/12L^2} \\ &= 10 \log_{10} 12L^2 \\ &= 10(\log_{10} 3 + 2 \log_{10} 2) + 20 \log_{10} 2^n \\ &= 10.79 + 6.02n \end{aligned}$$

$$\begin{aligned} \log_{10} 2 &\cong 0.301 \\ \log_{10} 3 &\cong 0.477 \end{aligned}$$

Rule: 1bit 6dB

量子化

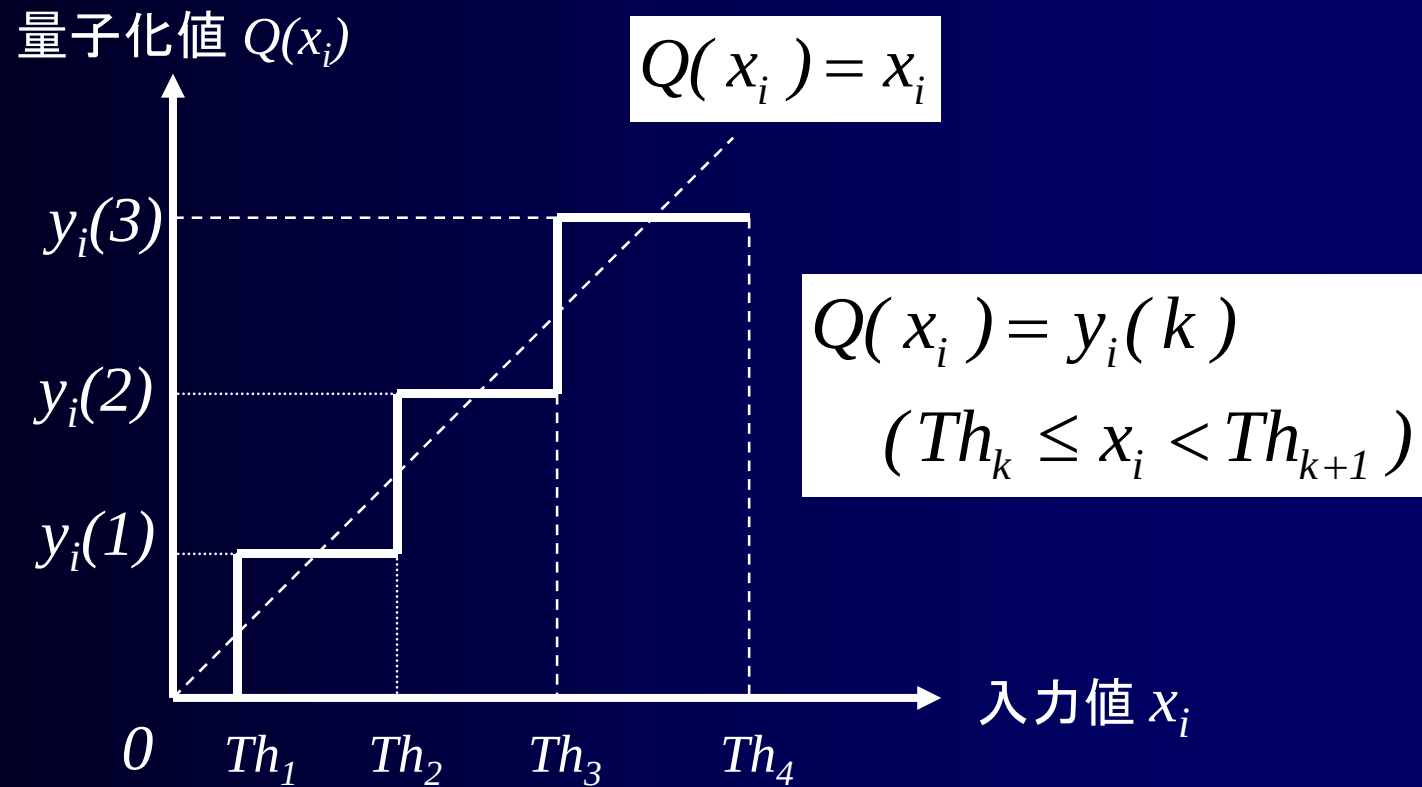
- 量子化の役割
 - 状態数の削減によるデータ圧縮
 - 量子化歪: 量子化することにより生じる誤差
- 量子化の種類
 - スカラー量子化、ベクトル量子化
 - 線形量子化、非線形量子化
- 量子化の最適化
 - Max量子化器
- 画像符号化における量子化
 - JPEG, H.261, MPEG-1, MPEG-2

Quantization

- Role of Quantization
 - Data compression by reducing the number of status
 - Quantization Error: Error caused by quantization
- Types of Quantization
 - Scalar Quantization, Vector Quantization
 - Linear Quantization, Non-linear Quantization
- Optimization of Quantization
 - Max Quantizer
- Quantizer used at Image Coding
 - JPEG, H.261, MPEG-1, MPEG-2

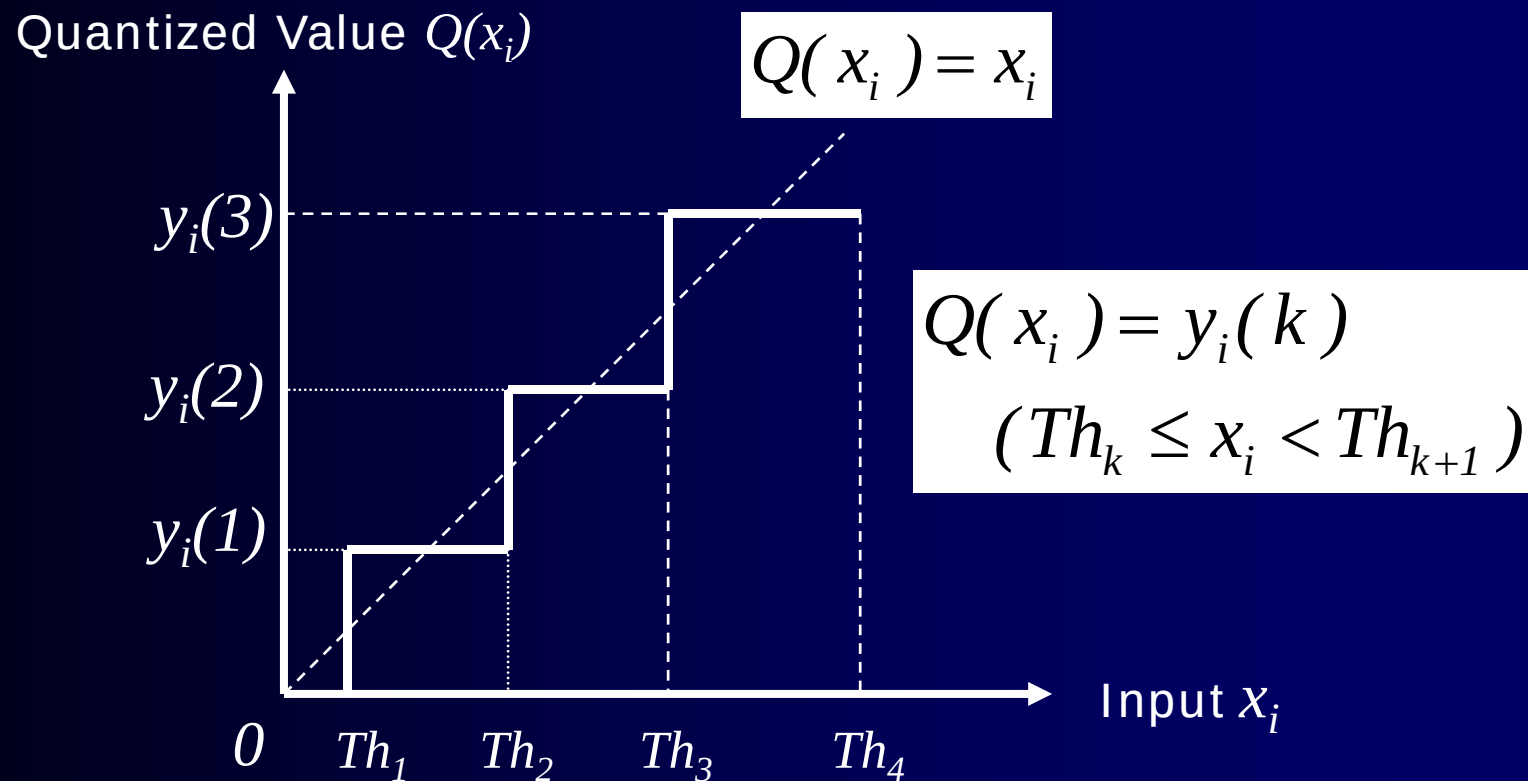
量子化とは？

■ 離散レベル化



What's Quantization?

- Discrete Level



量子化歪

- スカラの場合の量子化歪と距離尺度 (L^p ノルム)

$$d(x, y) = \|x - y\|_p$$
$$\|x - y\|_p = \left(|x - y|^p \right)^{1/p}$$

- ベクトルの場合の量子化歪とノルム

$$d(\mathbf{X}, \mathbf{Y}) = \|\mathbf{X} - \mathbf{Y}\|_p$$
$$\|\mathbf{X} - \mathbf{Y}\|_p = \left(|x_1 - y_1|^p + |x_2 - y_2|^p + \cdots + |x_n - y_n|^p \right)^{1/p}$$

Quantization Distortion

- Quantization Error and Distortion Measure for Scalar's Case (L^p -norm)

$$d(x, y) = \|x - y\|_p$$
$$\|x - y\|_p = \left(|x - y|^p \right)^{1/p}$$

- Quantization Error and Distortion Measure for Vector's Case (L^p -norm)

$$d(\mathbf{X}, \mathbf{Y}) = \|\mathbf{X} - \mathbf{Y}\|_p$$
$$\|\mathbf{X} - \mathbf{Y}\|_p = \left(|x_1 - y_1|^p + |x_2 - y_2|^p + \cdots + |x_n - y_n|^p \right)^{1/p}$$

ノルム

■ 代表的なノルム

$$\|x\|_p = \left(|x_1|^p + |x_2|^p + \cdots + |x_n|^p \right)^{1/p} \quad (\infty > p \geq 1)$$

$$\|x\|_1 = |x_1| + |x_2| + \cdots + |x_n|$$

$$\|x\|_2 = \left(|x_1|^2 + |x_2|^2 + \cdots + |x_n|^2 \right)^{1/2} \quad (Euclidean\ norm)$$

$$\|x\|_\infty \equiv \max_i |x_i|$$

Norm

■ Typical Norms

$$\|x\|_p = \left(|x_1|^p + |x_2|^p + \cdots + |x_n|^p \right)^{1/p} \quad (\infty > p \geq 1)$$

$$\|x\|_1 = |x_1| + |x_2| + \cdots + |x_n|$$

$$\|x\|_2 = \left(|x_1|^2 + |x_2|^2 + \cdots + |x_n|^2 \right)^{1/2} \quad (\text{Euclidean norm})$$

$$\|x\|_\infty \equiv \max_i |x_i|$$

スカラー量子化とベクトル量子化

■ スカラー量子化

- N 個の要素を持つ入力信号 (N 次元ベクトル) X に対して、 N 個の量子化値を要素として持つ出力信号は、個々の要素の量子化値を並べたもの

$$X = [x_1 \quad x_2 \quad \cdots \quad x_N]^t$$

$$Q_s(X) = [Q_s(x_1) \quad Q_s(x_2) \quad \cdots \quad Q_s(x_N)]^t$$

$$Q_s(x_i) = (y_i(k_i) | k_i = \arg \min_j d(x_i, y_i(j)) \quad (j = 1, 2, \dots, r_i))$$

$$d(x, y) = \|x - y\|_p$$

Scalar and Vector Quantization

■ Scalar Quantization

- For input signals (N -dimensional vector) having N elements, output signals have N elements which are quantized individually

$$\mathbf{X} = [x_1 \quad x_2 \quad \cdots \quad x_N]^t$$

$$Q_s(\mathbf{X}) = [Q_s(x_1) \quad Q_s(x_2) \quad \cdots \quad Q_s(x_N)]^t$$

$$Q_s(x_i) = (y_i(k_i) | k_i = \arg \min_j d(x_i, y_i(j)) \quad (j = 1, 2, \dots, r_i))$$

$$d(x, y) = \|x - y\|_p$$

スカラー量子化とベクトル量子化 (2)

■ ベクトル量子化

- N 個の要素を持つ入力信号 (N 次元ベクトル) X に対して、準備した代表ベクトル群 (コードブック) $Y(i)$ ($i=1, \dots, M$) の中から最も距離の近いものを選び出す

$$\mathbf{X} = [x_1 \quad x_2 \quad \cdots \quad x_N]^t$$

$$Q_V(\mathbf{X}) = [y_1(k) \quad y_2(k) \quad \cdots \quad y_N(k)]^t$$

$$Q_V(\mathbf{X}) = (\mathbf{Y}(k) | k = \arg \min_i d(\mathbf{X}, \mathbf{Y}(i))) \quad (i = 1, 2, \dots, M)$$

$$\mathbf{Y}(i) = [y_1(i) \quad y_2(i) \quad \cdots \quad y_N(i)]^t$$

$$d(\mathbf{X}, \mathbf{Y}(i)) = \|\mathbf{X} - \mathbf{Y}(i)\|_p$$

Scalar and Vector Quantization (2)

■ Vector Quantization

- For input signals (N -dimensional vector) having N elements, a set of output signals giving the minimum distortion is selected from codevectors stored in a codebook $Y(i)$ ($i=1,\dots,M$)

$$\mathbf{X} = [x_1 \quad x_2 \quad \cdots \quad x_N]^t$$

$$Q_V(\mathbf{X}) = [y_1(k) \quad y_2(k) \quad \cdots \quad y_N(k)]^t$$

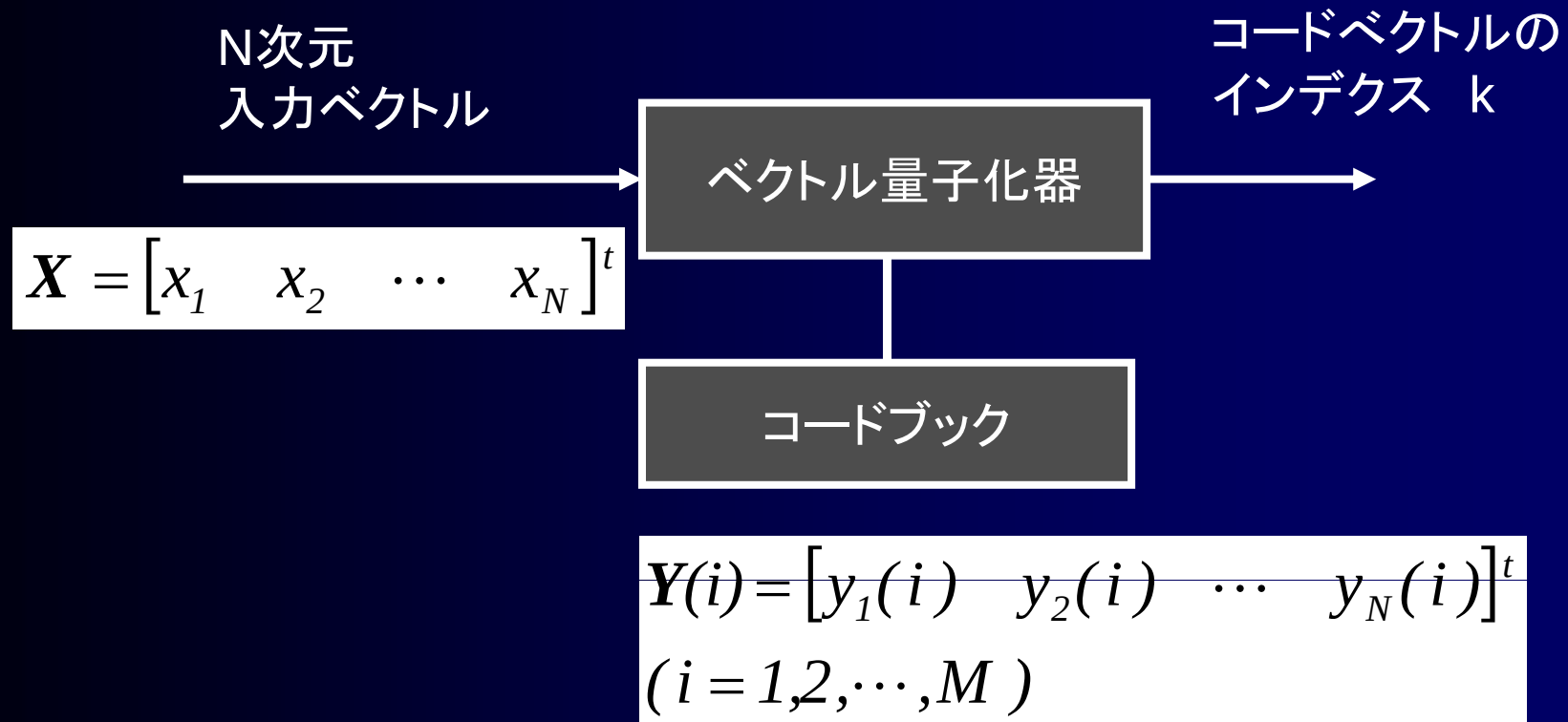
$$Q_V(\mathbf{X}) = (\mathbf{Y}(k) | k = \underset{i}{\arg \min} d(\mathbf{X}, \mathbf{Y}(i))) \quad (i = 1, 2, \dots, M)$$

$$\mathbf{Y}(i) = [y_1(i) \quad y_2(i) \quad \cdots \quad y_N(i)]^t$$

$$d(\mathbf{X}, \mathbf{Y}(i)) = \|\mathbf{X} - \mathbf{Y}(i)\|_p$$

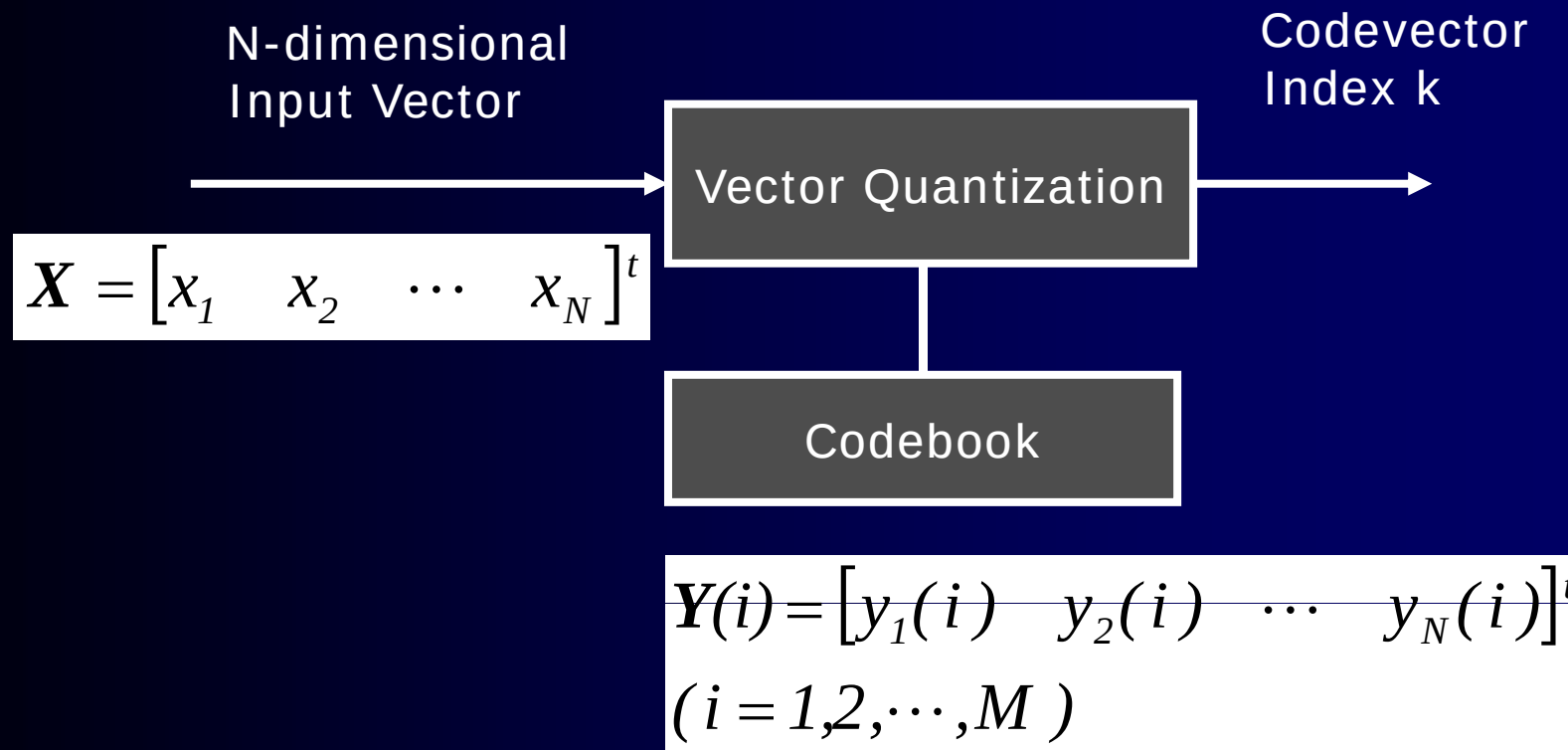
スカラー量子化とベクトル量子化 (3)

■ ベクトル量子化の概念図



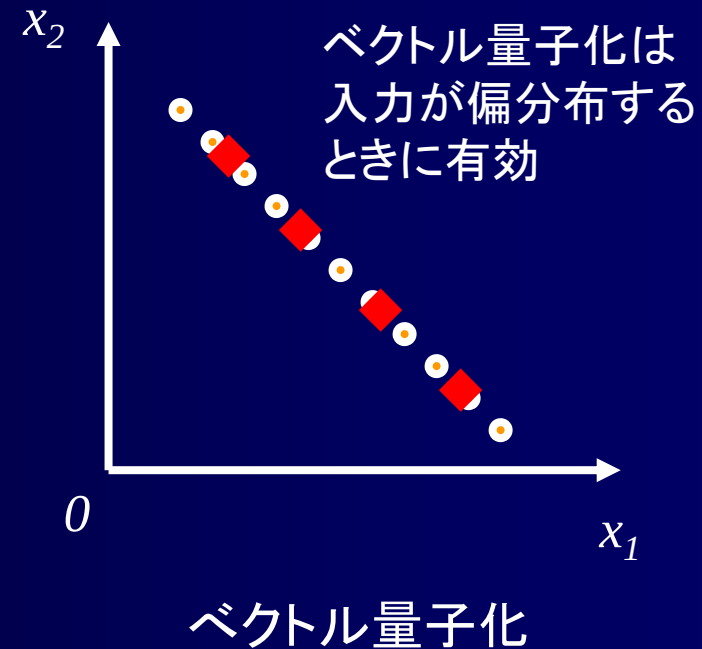
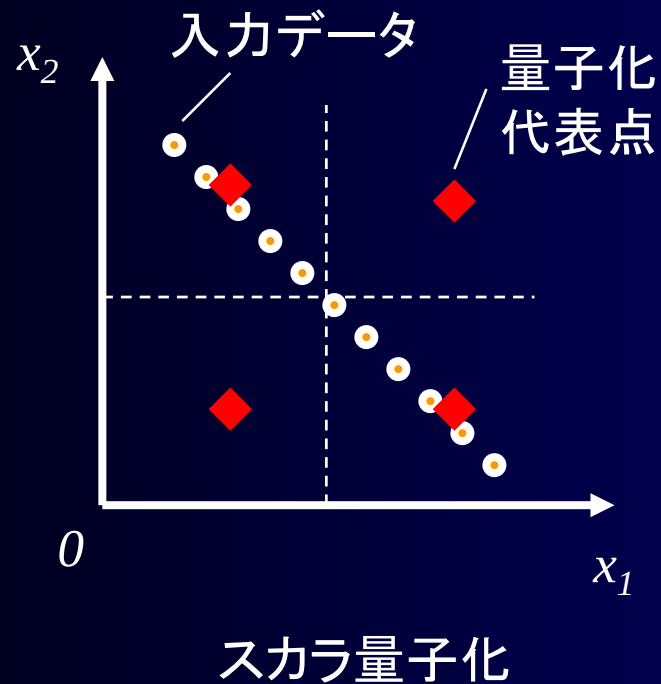
Scalar and Vector Quantization (3)

■ Block Diagram of Vector Quantization



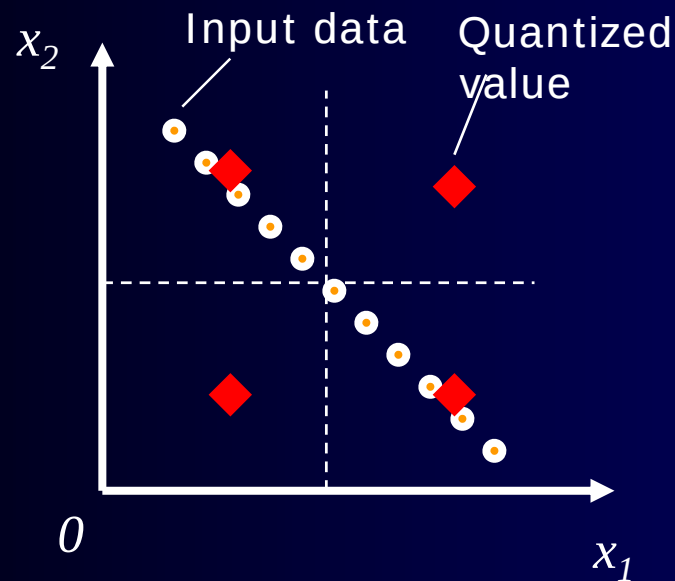
スカラ量子化とベクトル量子化 (4)

■ 2次元入力のスカラ量子化とベクトル量子化

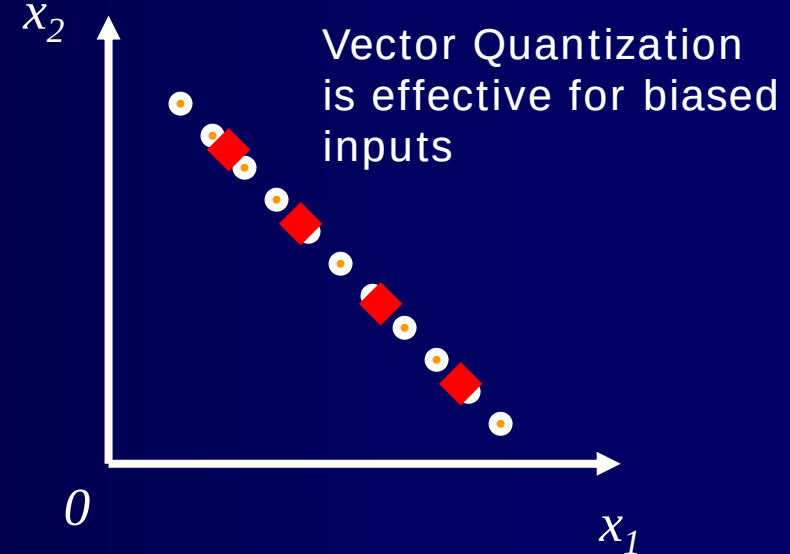


Scalar and Vector Quantization (4)

- Scalar and Vector Quantization for 2-D input



Scalar Quantization



Vector Quantization

ベクトル量子化

- コードブックの作成... LBGアルゴリズム (Linde, Buzo, Gray)
 - 初期コードブック $C_0 = \{Y_0(i)\} (i=1,2,\dots,N)$
 - トレーニング集合 T を N 個の集合 $R_0(i)$ に分割 (最近傍探索)

$$R_0(i) = \{ X \in T : d(X, Y_0(i)) < d(X, Y_0(j)); \text{ for } j \neq i \}$$

- セントロイドベクトルの計算

$$Y_1(i) = E[X | X \in R_0(i)]$$

- 全体での誤差が一定以下になるまで, この処理の繰り返し

Vector Quantization

- Codebook Generation ... LBG Algorithm (Linde, Buzo, Gray)
 - Initial codebook $C_0 = \{Y_0(i)\} (i=1,2,\dots,N)$
 - Divide training set T to N sets $R_0(i)$ (Nearest Neighbor Search)

$$R_0(i) = \{ \mathbf{X} \in T : d(\mathbf{X}, Y_0(i)) < d(\mathbf{X}, Y_0(j)); \text{ for } j \neq i \}$$

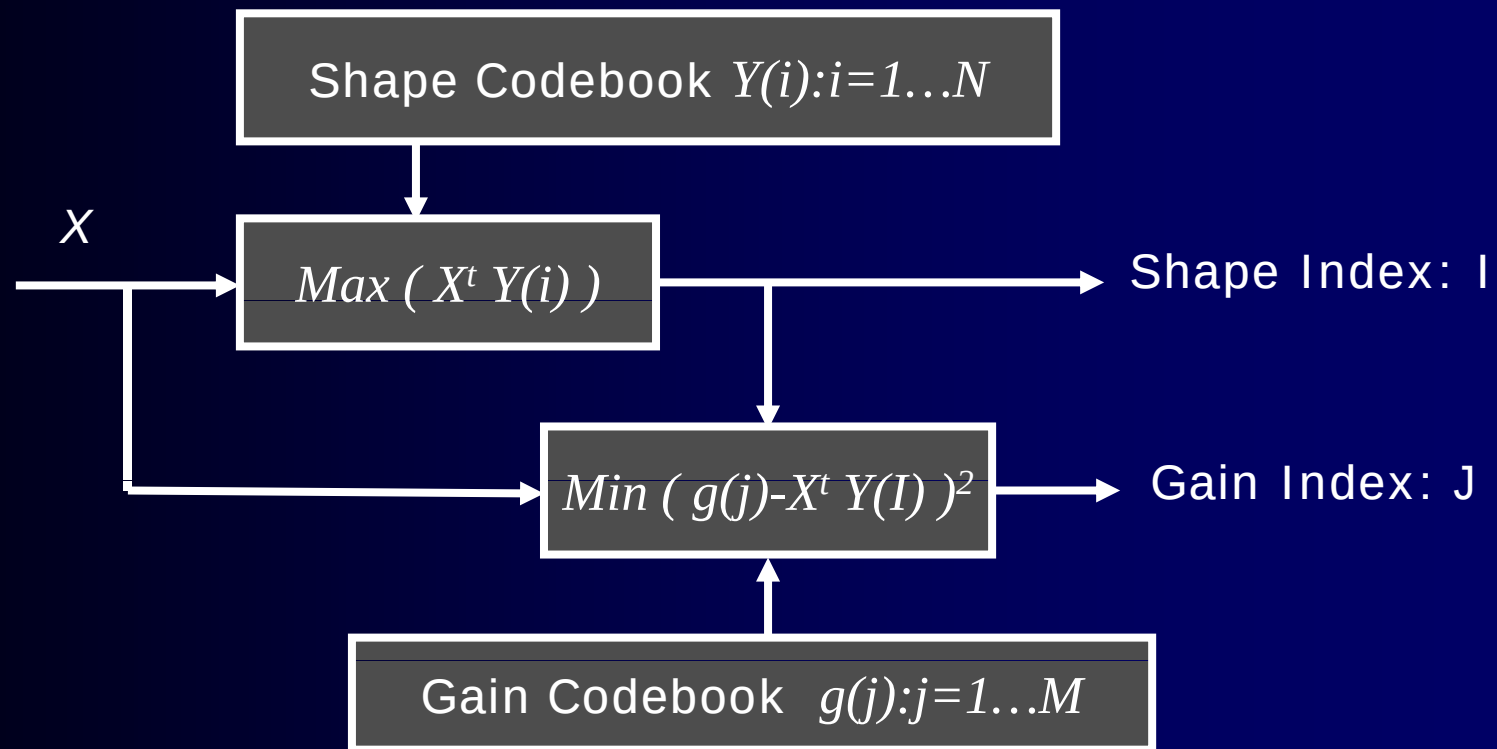
- Calculation of Centroid Vector

$$Y_1(i) = E[\mathbf{X} \mid \mathbf{X} \in R_0(i)]$$

- Repeat this procedure until total error will reach to a certain level

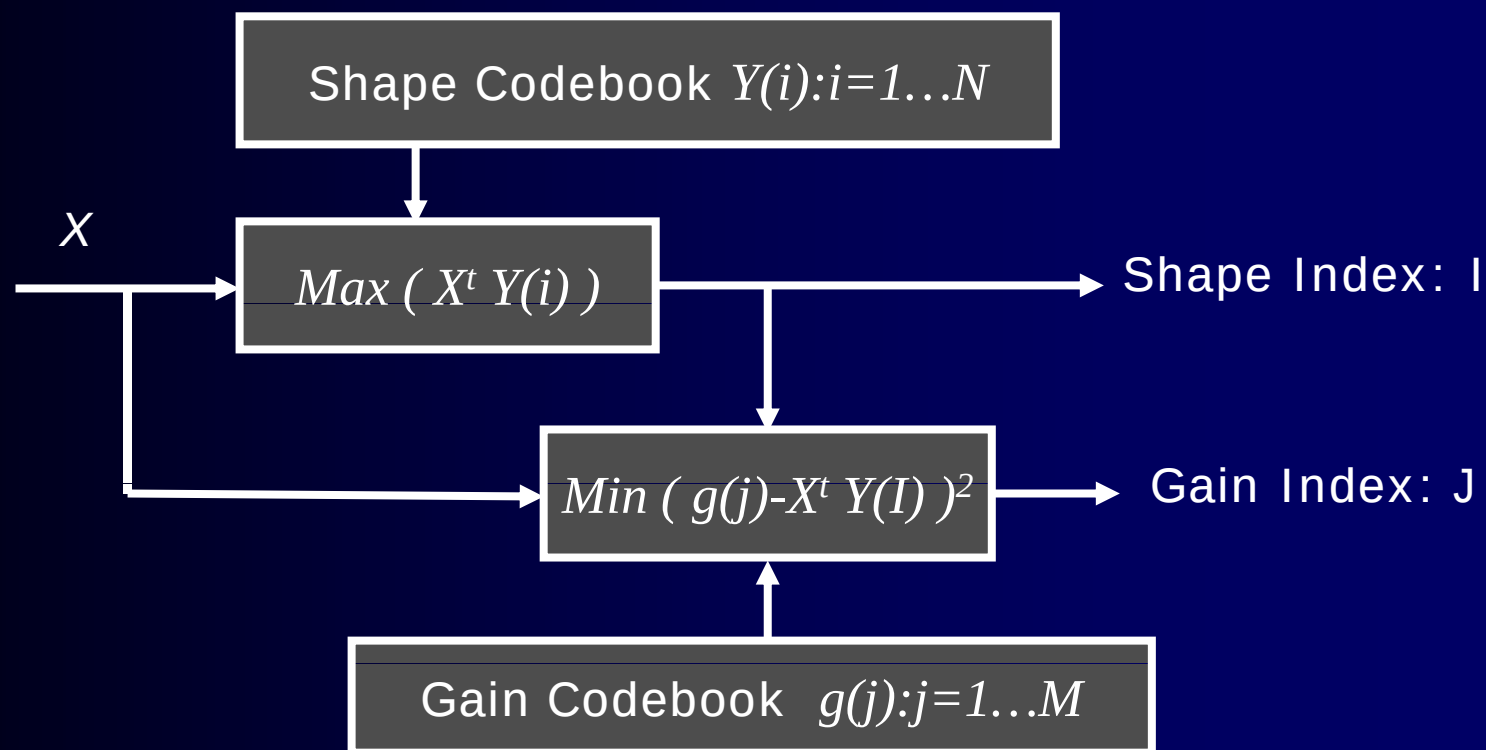
ベクトル量子化 (2)

■ Shape-Gain ベクトル量子化



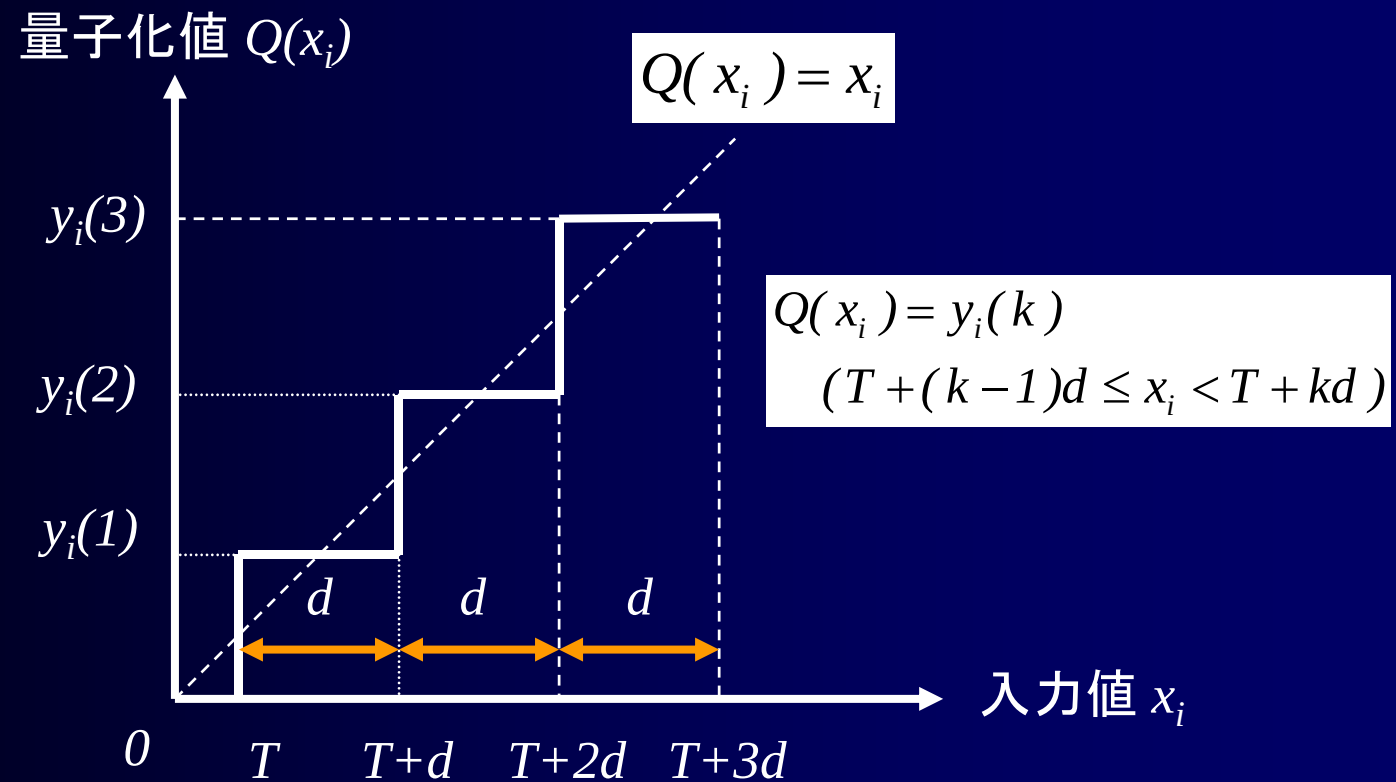
Vector Quantization (2)

■ Shape-Gain Vector Quantization



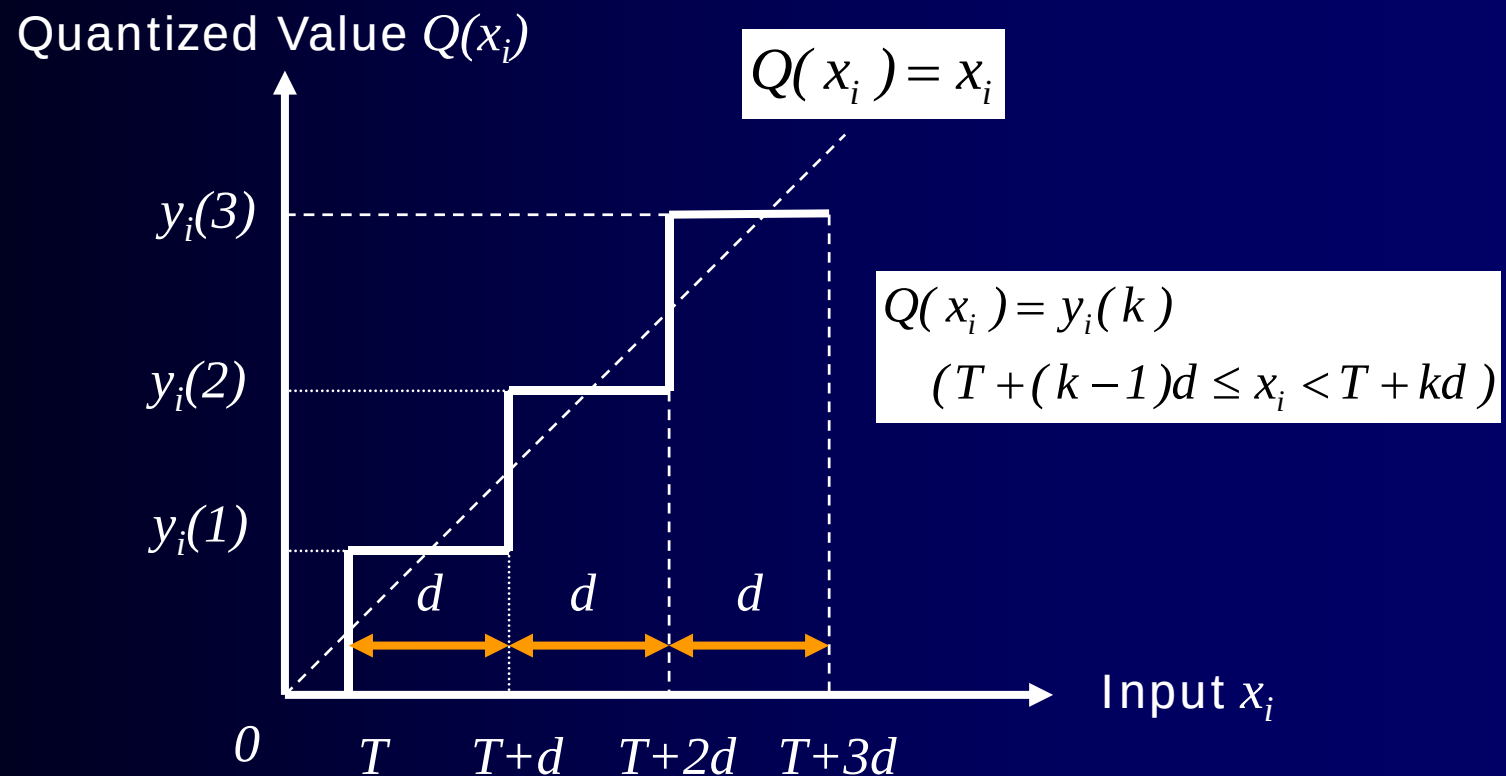
線形量子化

- 量子化間隔 d が一定 (一樣量子化とも呼ぶ)



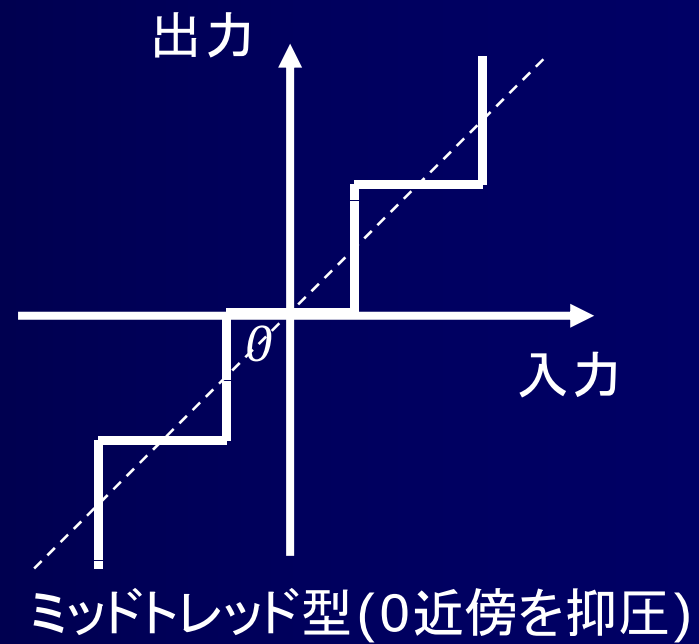
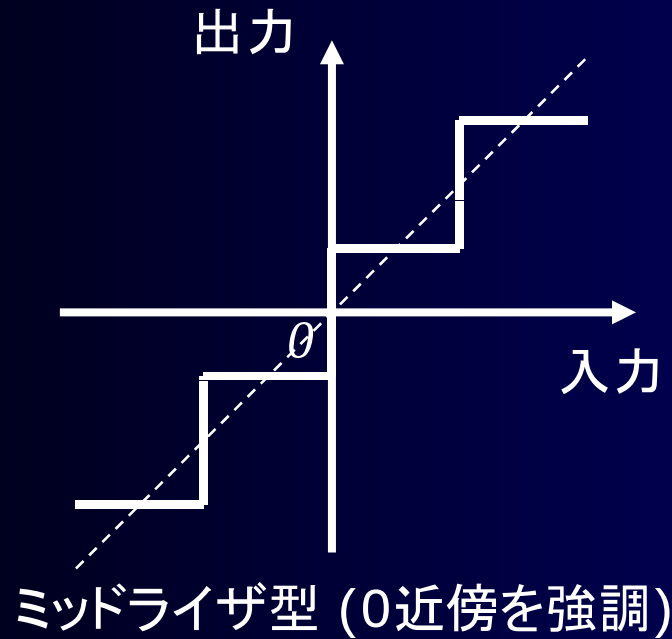
Linear Quantization

- Quantization Step d is constant (uniform quantization)



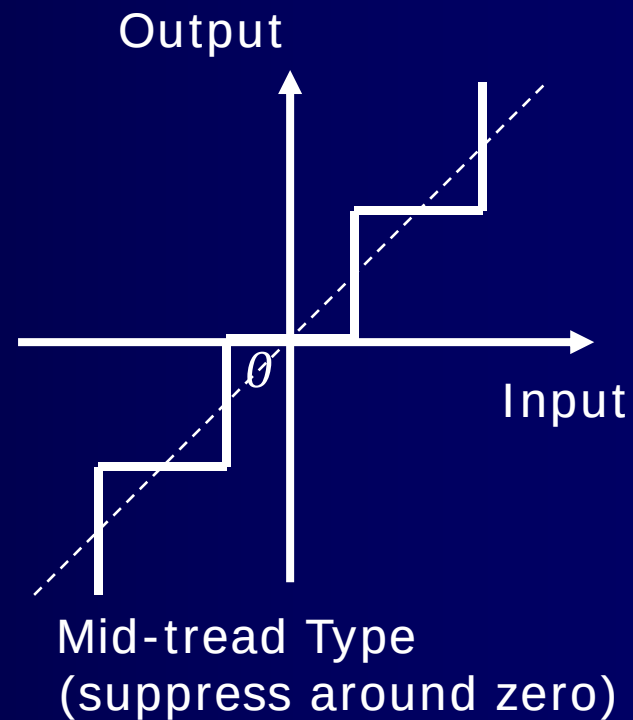
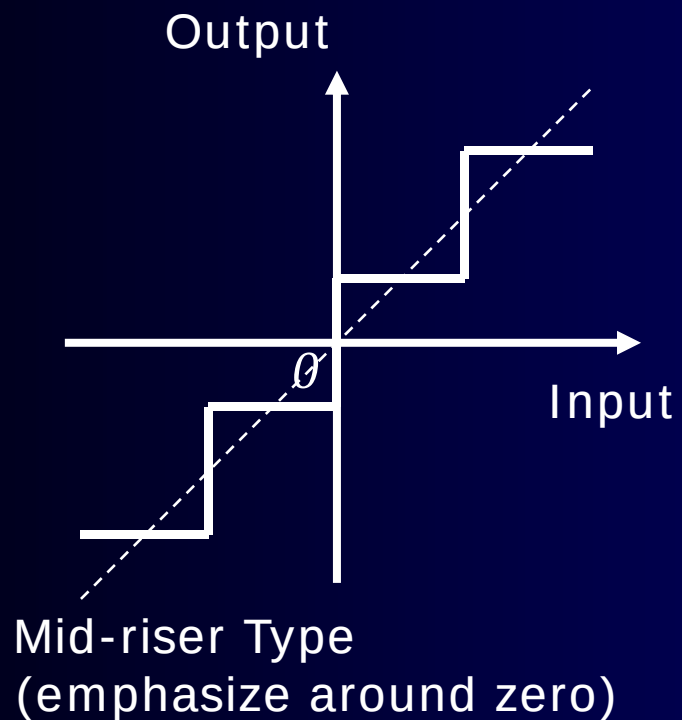
線形量子化 (2)

- ミッドライザ型とミッドトレッド型



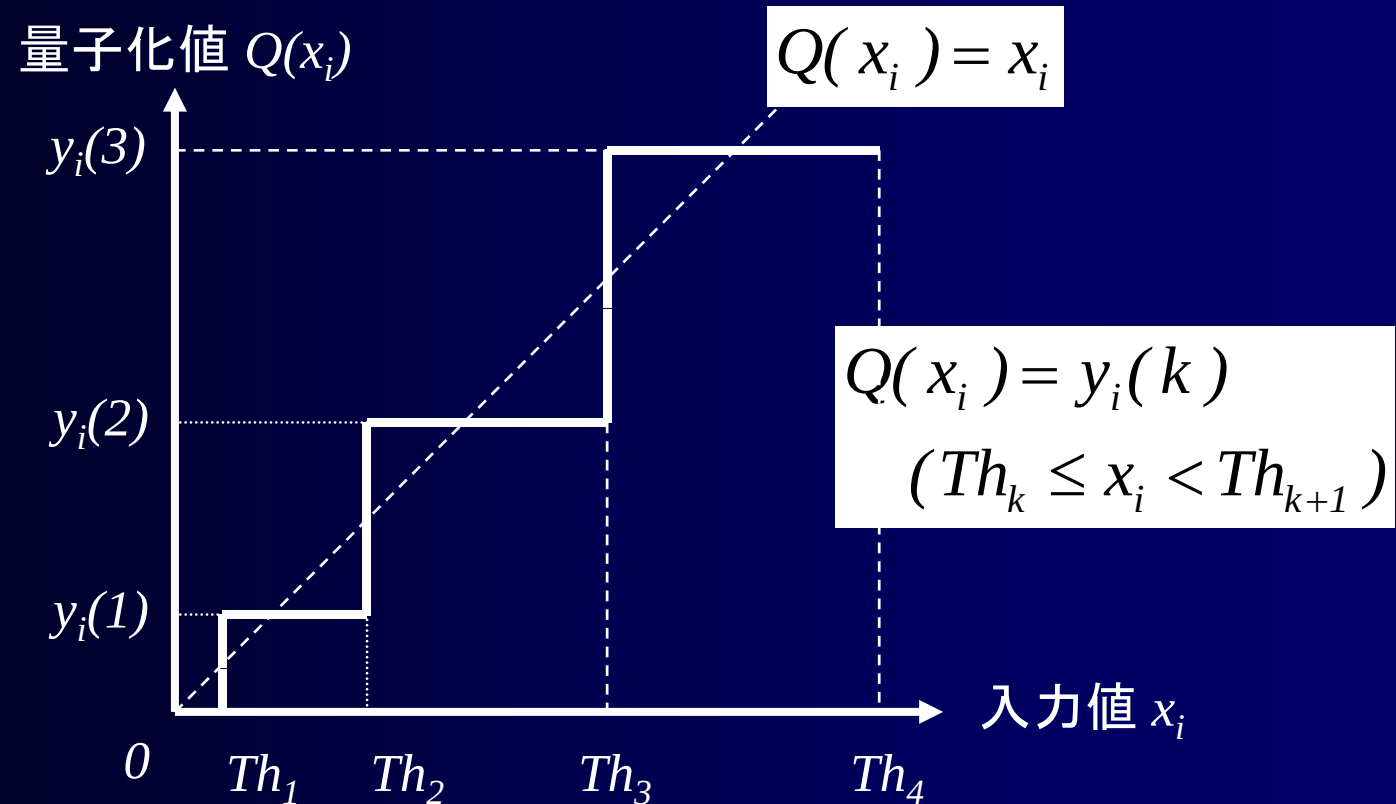
Linear Quantization (2)

- Mid-riser and Mid-tread Type



非線形量子化

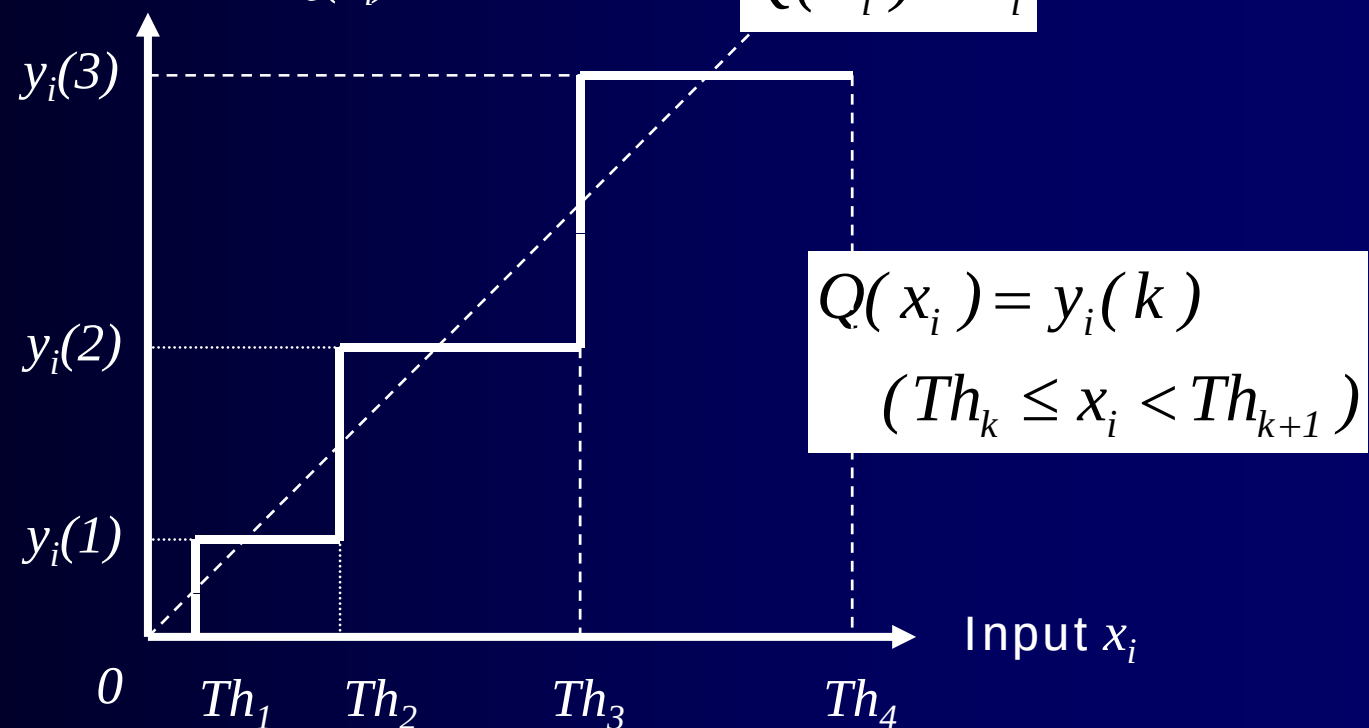
- 量子化ステップが非一様



Non-linear Quantization

- Quantization Step in non-uniform

Quantized Value $Q(x_i)$



非線形量子化 (2)

- 信号の特性と量子化ステップの関係
 - 入力信号の出現確率に合わせる
 - 均一な量子化に比べ、平均的な量子化誤差を低減できる
- DPCMのように0付近の信号の出現確率が高い場合
 - 0付近の入力値は出現頻度が高いため細かく量子化
 - 大きな入力値は、出現頻度が低いので粗く量子化
- 入力信号の出現確率と量子化ステップの関係
 - 平均的な量子化誤差の最適化(最小化)が可能

Non-linear Quantization (2)

- Relation of signal characteristics and quantization step
 - Fit to input signal's occurrence probability
 - Can reduce average distortion than linear quantization
- In case of high occurrence probability around zero such as DPCM
 - Fine quantization around zero
 - Coarse quantization to large inputs
- Relation of input's occurrence probability and quantization steps
 - Optimization (Minimization) of the average distortion is possible

Max量子化器

- 信号 x の確率密度 $p(x)$ に関して平均2乗誤差を最小化する量子化器
- しきい値 $t_0, t_1, \dots, t_n$, 量子化値 $q_0, q_1, \dots, q_n$ に対する設計手順
 - Step.1 しきい値の設定

$$t_i = (q_{i-1} + q_i) / 2 \quad (i = 1, 2, \dots, n-1)$$

ただし $t_0 = -\infty \quad t_n = \infty$

- Step.2 量子化値の設定(重心)

$$q_i = \int_{t_i}^{t_{i+1}} x p(x) dx / \int_{t_i}^{t_{i+1}} p(x) dx \quad (i = 0, 1, \dots, n-1)$$

Max Quantizer

- Minimize MSE with regard to signal x 's probability $p(x)$
- Design procedure to threshold $t_0, t_1, \dots, t_n$ and quantized value $q_0, q_1, \dots, q_n$
 - Step.1 Set thresholds

$$t_i = (q_{i-1} + q_i) / 2 \quad (i = 1, 2, \dots, n-1)$$

where $t_0 = -\infty \quad t_n = \infty$

- Step.2 Obtain quantized values (Centroid)

$$q_i = \int_{t_i}^{t_{i+1}} xp(x)dx / \int_{t_i}^{t_{i+1}} p(x)dx \quad (i = 0, 1, \dots, n-1)$$

Max量子化器 (2)

- 量子化誤差(平均2乗誤差)の計算

$$e^2 = \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} p(x)(x - q_i)^2 dx$$

- 最適量子化を与えるしきい値は上式を t_i で偏微分して0と置けば

$$t_i = (q_i + q_{i-1}) / 2 \quad (i = 1, 2, \dots, n-1)$$

- 同様に q_i で偏微分すれば量子化値は重心で与えられる(設計手順の理論的背景)
- 確率密度 $p(x)$ が特殊な場合しか解析的に解けず、一般に数値計算で求める必要が生じる

Max Quantizer (2)

- Calculation of Quantization Error (MSE)

$$e^2 = \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} p(x)(x - q_i)^2 dx$$

- Optimum threshold is obtained by taking derivatives of t_i

$$t_i = (q_i + q_{i-1}) / 2 \quad (i = 1, 2, \dots, n-1)$$

- Similarly, by taking derivatives of q_i , quantized value is given by centroid (Theoretical background for design procedure)
- Cannot solve analytically for general probability density $p(x)$, normally use numerical calculation is employed

Max量子化器 (3)

- 入力信号の確率密度 $p(x)$ が平均0、分散1の正規分布の場合

$N=4$	しきい値	レベル
i	t_i	q_i
0		-1.510
1	-0.9816	-0.4528
2	0.0	0.4528
3	0.9816	1.510
e^2	0.1175	

$N=8$	しきい値	レベル
i	t_i	q_i
0		-2.152
1	-1.748	-1.344
2	-1.050	-0.7560
3	-0.5006	-0.2451
4	0.0	0.2451
5	0.5006	0.7560
6	1.050	1.344
7	1.748	2.152
e^2	0.03454	

Max Quantizer (3)

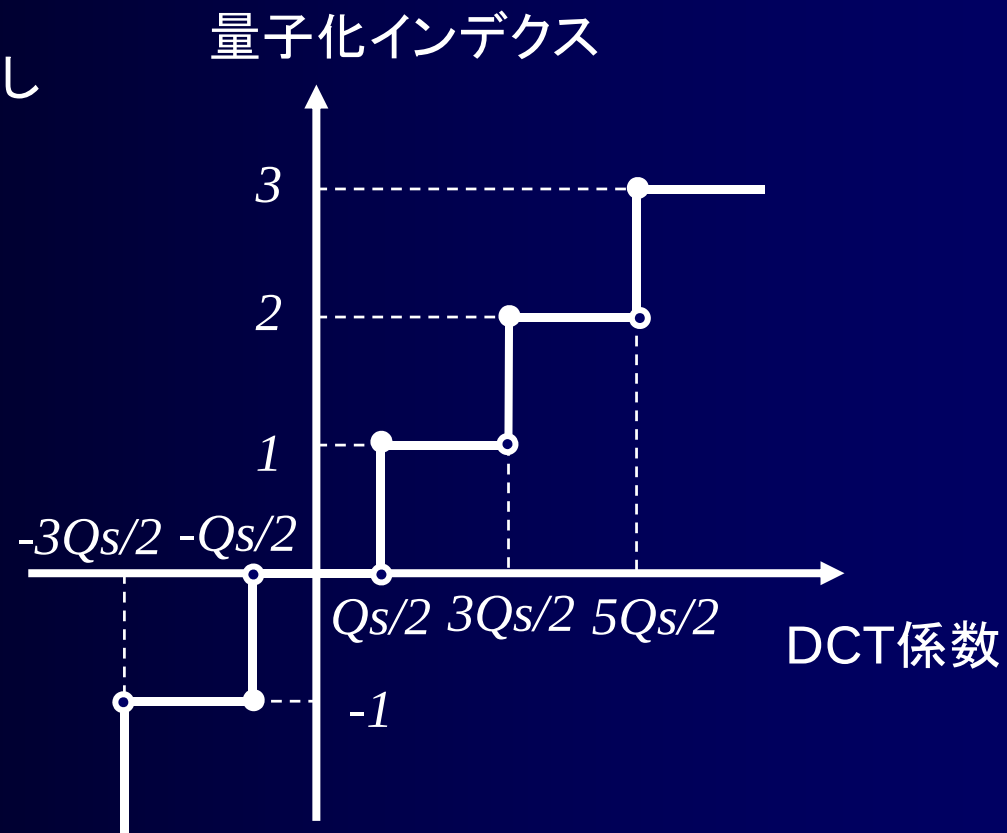
- When input signal's probability $p(x)$ is Normal distribution with average zero and variance 1

$N=4$	threshold	level
i	t_i	q_i
0		-1.510
1	-0.9816	- 0.4528
2	0.0	0.4528
3	0.9816	1.510
e^2	0.1175	

$N=8$	threshold	level
i	t_i	q_i
0		-2.152
1	-1.748	-1.344
2	-1.050	-0.7560
3	-0.5006	-0.2451
4	0.0	0.2451
5	0.5006	0.7560
6	1.050	1.344
7	1.748	2.152
e^2	0.03454	

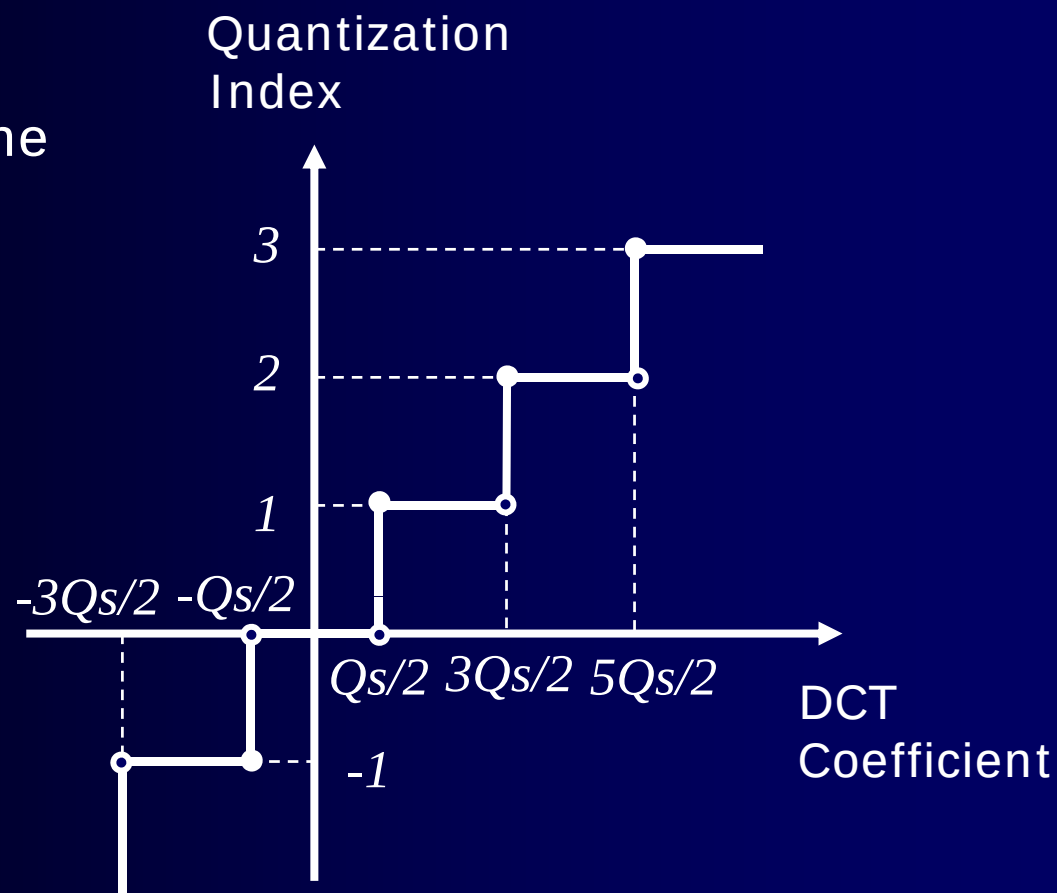
JPEG量子化

- 一様量子化器
- デッドゾーンなし
- ミッドトレッド型



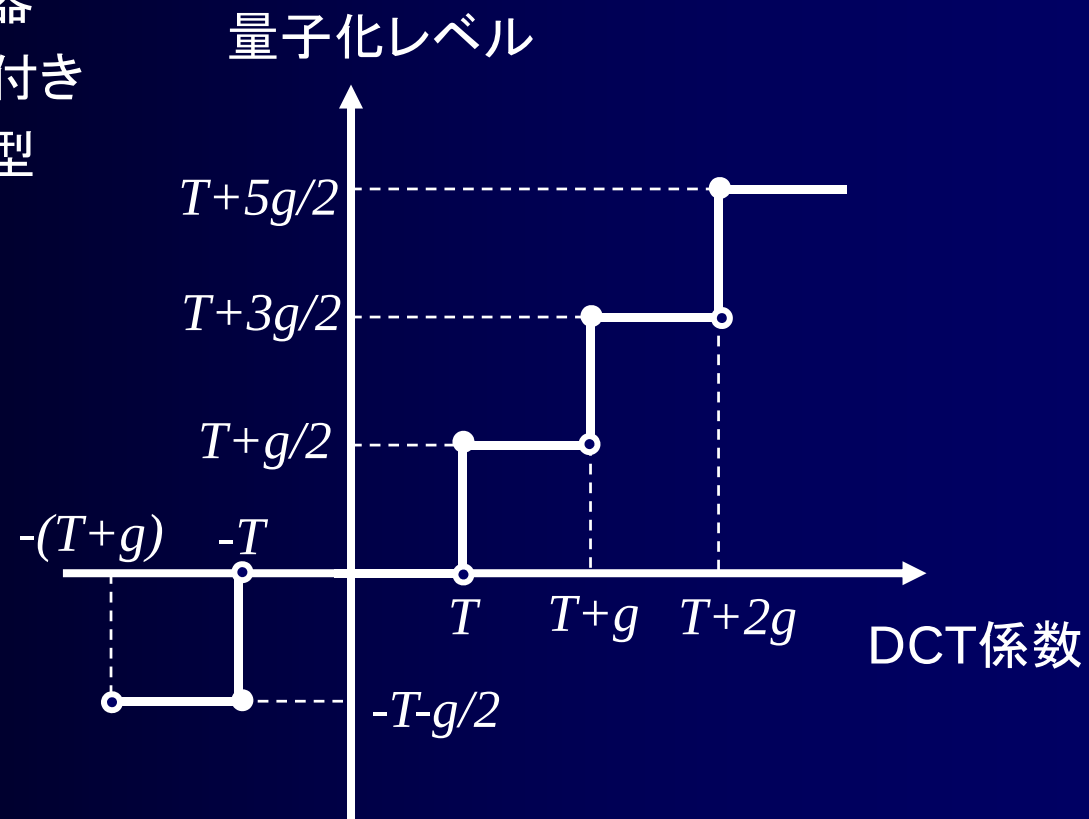
JPEG Quantization

- Uniform
- No dead-zone
- Mid-tread



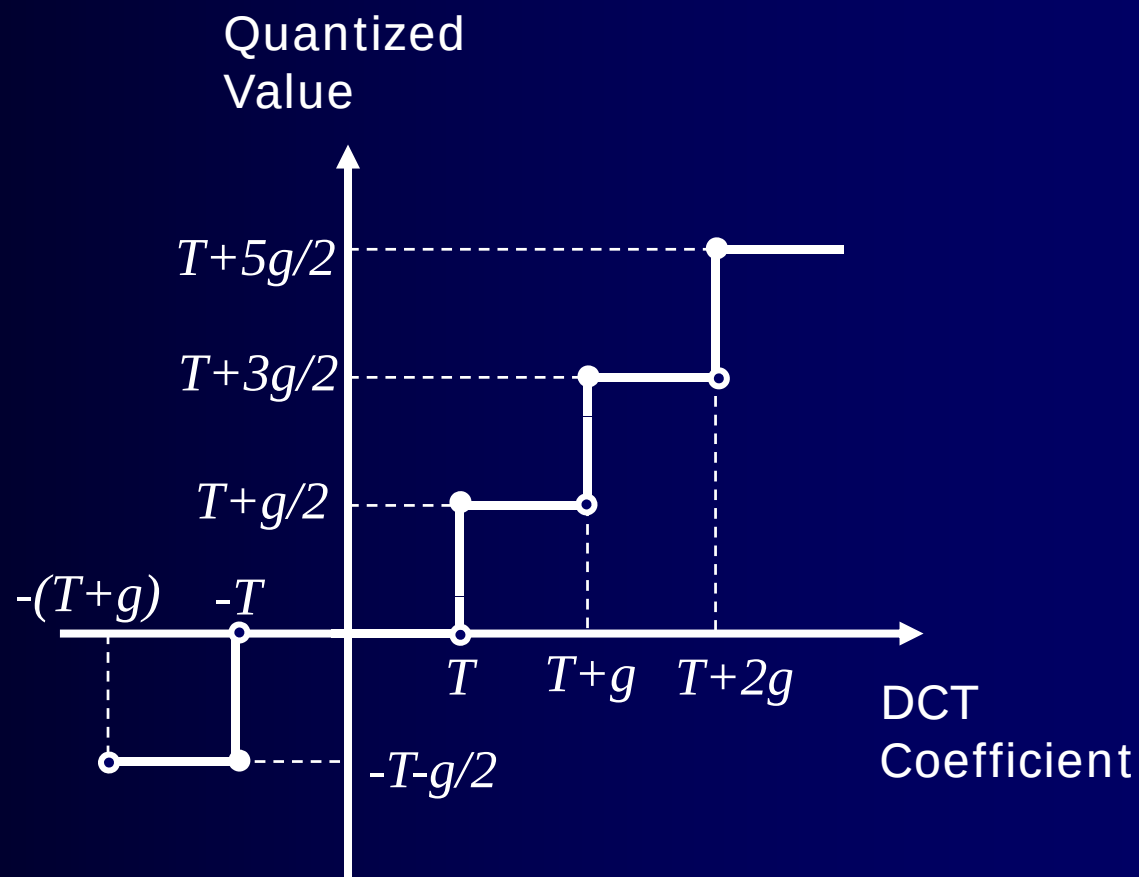
H.261量子化

- 一様量子化器
- デッドゾーン付き
- ミッドトレッド型



H.261 Quantization

- Uniform
- Dead-zone
- Mid-tread



H.261量子化 (2)

- Intra mode (フレーム内符号化モード)
 - DC係数はデッドゾーンなし、量子化ステップ値を8に固定
- Inter mode (フレーム間差分符号化モード)
 - デッドゾーンの例 $T=g$
 - 量子化ステップ $g=2q$ ($q=1, 2, \dots, 31$)
 - しきい値 $t_i=ig$ ($i= -127, \dots, -1, 1, \dots, 127$)
 - 量子化値 $q_i = (t_i + t_{i+1})/2$ ($i > 0$)
 $(t_i + t_{i-1})/2$ ($i < 0$)
 - IDCTミスマッチ処理
 - If(q_i is even) $q_i = q_i - 1$;

H.261 Quantization (2)

- Intra mode (Intra-frame coding mode)
 - No dead-zone for DC Coefficient, Quantization Step=8
- Inter mode (Inter-frame difference coding mode)
 - Example of dead-zone $T=g$
 - Quantization step $g=2q$ ($q=1, 2, \dots, 31$)
 - Threshold $t_i=ig$ ($i= -127, \dots, -1, 1, \dots, 127$)
 - Quantization value $q_i= (t_i+t_{i+1})/2$ ($i>0$)
 $(t_i+t_{i-1})/2$ ($i<0$)
 - IDCT mismatch process
 - If(q_i is even) $q_i=q_i-1$;

H.261量子化 (3)

■ IDCTミスマッチ処理

q	g	$t_i (=ig)$					q_i			
		$i=1$	$i=2$	$i=3$	$i=4$	$i=5$	$i=1$	$i=2$	$i=3$	$i=4$
1	2	2	4	6	8	10	3	5	7	9
2	4	4	8	12	16	20	6→5	10→9	14→13	18→17
3	6	6	12	18	24	30	9	15	21	27
4	8	8	16	24	32	40	12→11	20→19	28→27	36→35

H.261 Quantization (3)

■ IDCT Mismatch Process

q	g	$t_i (=ig)$					q_i			
		$i=1$	$i=2$	$i=3$	$i=4$	$i=5$	$i=1$	$i=2$	$i=3$	$i=4$
1	2	2	4	6	8	10	3	5	7	9
2	4	4	8	12	16	20	6→5	10→9	14→13	18→17
3	6	6	12	18	24	30	9	15	21	27
4	8	8	16	24	32	40	12→11	20→19	28→27	36→35

MPEG-1量子化

- Intra mode (フレーム内符号化モード)
 - DC係数はデッドゾーンなし、量子化ステップ値を8に固定
 - AC係数はJPEGと同様
 - デッドゾーンなし、一様量子化器
 - 量子化マトリクス使用
- Inter mode (フレーム間差分符号化モード)
 - DC係数、AC係数ともにH.261と同様
 - デッドゾーン付き一様量子化器
 - 量子化マトリクスを使用

MPEG-1 Quantization

- Intra mode (Intra-frame coding mode)
 - DC coeff: no dead-zone, quantization step=8
 - AC coeff is the same as JPEG
 - No dead-zone, uniform quantizer
 - Quantization matrix
- Inter mode (Inter-frame difference coding mode)
 - DC coeffs and AC coeffs are the same as H.261
 - Dead-zone, uniform quantizer
 - Quantization matrix

MPEG-1量子化 (2)

- 量子化マトリクスを用いた量子化処理
 - Intra, AC係数の量子化

$$A(u,v) = \frac{16 X(u,v) \pm W_{Intra}(u,v) / 2}{W_{Intra}(u,v)}$$

$$N(u,v) = \frac{A(u,v) \pm q}{2q}$$

$$Q_{AC-Intra}(u,v) = 2qN(u,v)W_{intra}(u,v) / 16$$

MPEG-1 Quantization (2)

- Quantization using Matrix
 - Quantization to Intra, AC Coefficients

$$A(u,v) = \frac{16 X(u,v) \pm W_{Intra}(u,v) / 2}{W_{Intra}(u,v)}$$

$$N(u,v) = \frac{A(u,v) \pm q}{2q}$$

$$Q_{AC-Intra}(u,v) = 2qN(u,v)W_{intra}(u,v) / 16$$

MPEG-1量子化 (3)

- 量子化マトリクスを用いた量子化処理
 - Inter, DC+AC係数の量子化

$$A(u,v) = \frac{16 X(u,v) \pm W_{Inter}(u,v) / 2}{W_{Inter}(u,v)}$$

$$N(u,v) = \frac{A(u,v)}{2q} \quad (q : \text{odd})$$

$$N(u,v) = \frac{A(u,v) \pm 1}{2q} \quad (q : \text{even})$$

$$Q_{Inter}(u,v) = (2N(u,v) \pm 1)qW_{inter}(u,v) / 16$$

$$Q_{Inter}(u,v) = Q_{Inter}(u,v) \mp 1 \quad (\text{if } Q_{Inter}(u,v) : \text{even})$$

MPEG-1 Quantization (3)

- Quantization using Matrix
 - Quantization to Inter, DC+AC Coefficients

$$A(u,v) = \frac{16 X(u,v) \pm W_{Inter}(u,v) / 2}{W_{Inter}(u,v)}$$

$$N(u,v) = \frac{A(u,v)}{2q} \quad (q : \text{odd})$$

$$N(u,v) = \frac{A(u,v) \pm 1}{2q} \quad (q : \text{even})$$

$$Q_{Inter}(u,v) = (2N(u,v) \pm 1)qW_{inter}(u,v) / 16$$

$$Q_{Inter}(u,v) = Q_{Inter}(u,v) \mp 1 \quad (\text{if } Q_{Inter}(u,v) : \text{even})$$

MPEG-1量子化 (4)

■ 量子化マトリクス

8	16	19	22	26	27	29	34
16	16	22	24	27	29	34	37
19	22	26	27	29	34	34	38
22	22	26	27	29	34	37	40
22	26	27	29	32	35	40	48
26	27	29	32	35	40	48	58
26	27	29	34	38	46	56	69
27	29	35	38	46	56	69	83

$$W_{intra}(u,v)$$

16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16

$$W_{inter}(u,v)$$

MPEG-1 Quantization (4)

■ Quantization Matrix

8	16	19	22	26	27	29	34
16	16	22	24	27	29	34	37
19	22	26	27	29	34	34	38
22	22	26	27	29	34	37	40
22	26	27	29	32	35	40	48
26	27	29	32	35	40	48	58
26	27	29	34	38	46	56	69
27	29	35	38	46	56	69	83

$$W_{intra}(u,v)$$

16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16
16	16	16	16	16	16	16	16

$$W_{inter}(u,v)$$

MPEG-2量子化

- Intra mode (フレーム内符号化モード)
 - DC係数はデッドゾーンなし、量子化ステップ値を8に固定
 - デッドゾーンなし、一様量子化器
 - 量子化マトリクス使用
- Inter mode (フレーム間差分符号化モード)
 - デッドゾーン付き一様量子化器
 - 量子化マトリクスを使用

MPEG-2 Quantization

- Intra mode (Intra-frame coding mode)
 - DC coeff: no dead-zone, quantization step=8
 - No dead-zone, uniform quantizer
 - Quantization matrix
- Inter mode (Inter-frame difference coding mode)
 - With dead-zone, uniform quantizer
 - Quantization matrix

MPEG-2量子化 (2)

- 量子化マトリクスを用いた量子化処理
 - Intra, AC係数の量子化

$$A(u,v) = \frac{16 X(u,v) + W_{Intra}(u,v) / 2}{W_{Intra}(u,v)}$$

$$N(u,v) = \frac{A(u,v) \pm aq / b}{2q}$$

$$Q_{Intra-AC}(u,v) = 2qN(u,v)W_{Intra}(u,v) / 16$$

MPEG-2 Quantization (2)

- Quantization by Matrix
 - Quantization to Intra, AC Coefficients

$$A(u,v) = \frac{16 X(u,v) + W_{Intra}(u,v) / 2}{W_{Intra}(u,v)}$$

$$N(u,v) = \frac{A(u,v) \pm aq / b}{2q}$$

$$Q_{Intra-AC}(u,v) = 2qN(u,v)W_{Intra}(u,v) / 16$$

MPEG-2量子化 (3)

- 量子化マトリクスを用いた量子化処理
 - Inter, DC+AC係数の量子化

$$A(u,v) = \frac{16 X(u,v) + W_{Inter}(u,v) / 2}{W_{Inter}(u,v)}$$

$$N(u,v) = \frac{A(u,v)}{2q} \quad (q : odd)$$

$$N(u,v) = \frac{A(u,v) \pm 1}{2q} \quad (q : even)$$

$$Q_{Inter}(u,v) = (2N(u,v) \pm 1) q W_{Inter}(u,v) / 16$$

MPEG-2 Quantization (3)

- Quantization by Matrix
 - Quantization to Inter, DC+AC Coefficients

$$A(u,v) = \frac{16 X(u,v) + W_{Inter}(u,v) / 2}{W_{Inter}(u,v)}$$

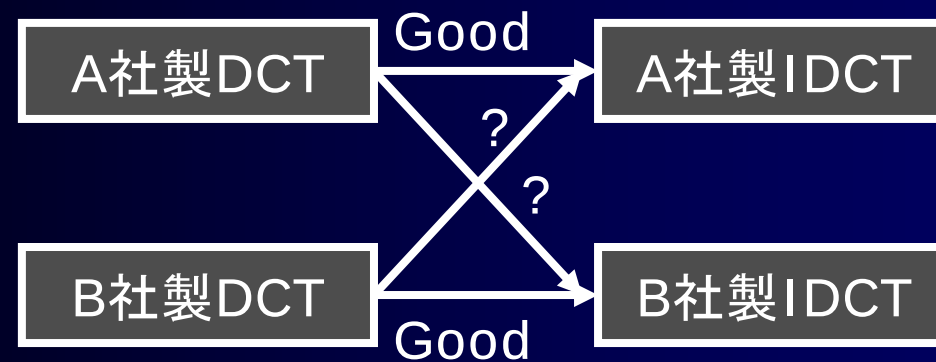
$$N(u,v) = \frac{A(u,v)}{2q} \quad (q : \text{odd})$$

$$N(u,v) = \frac{A(u,v) \pm 1}{2q} \quad (q : \text{even})$$

$$Q_{Inter}(u,v) = (2N(u,v) \pm 1)qW_{Inter}(u,v) / 16$$

IDCTミスマッチ

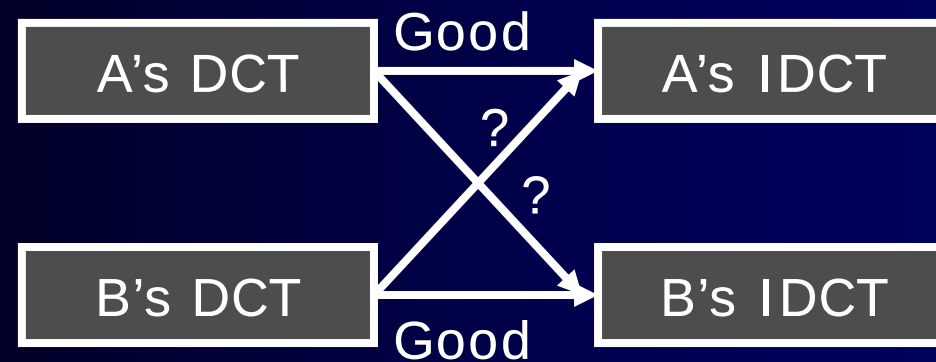
- 異なる演算精度でIDCTが実行された場合に起こる



- 8x8 DCTの計算式
 - 先頭に1/4を掛ける部分が存在する
 - $4k+2$ (k :整数)なるレベルが存在すると問題

IDCT Mismatch

- Happens when IDCT is operated different accuracy



- 8x8 DCT operation
 - Multiply $1/4$ at the head of operation
 - $4k+2$ (k :integer) level causes problem

IDCTミスマッチ (2)

- 8x8 IDCTにおける積和が $4k+2$ の場合

$$\begin{aligned}x(n, m) &= \frac{1}{4} \left(\sum_{u=0}^7 \sum_{v=0}^7 C(u) C(v) X(u, v) \cos \frac{(2n+1)u\pi}{16} \cos \frac{(2m+1)v\pi}{16} \right) \\&= \frac{1}{4} (4k+2) \\&= k+0.5\end{aligned}$$

- 演算手法, 演算精度により結果が異なる

$$\begin{aligned}x(n, m) &= k + 0.5000000000... \quad (k+1) \\&\text{or} \\&= k + 0.4999999999... \quad (k)\end{aligned}$$

IDCT Mismatch(2)

- For Multiplication and addition to $4k+2$ at 8×8 IDCT

$$\begin{aligned}x(n, m) &= \frac{1}{4} \left(\sum_{u=0}^7 \sum_{v=0}^7 C(u) C(v) X(u, v) \cos \frac{(2n+1)u\pi}{16} \cos \frac{(2m+1)v\pi}{16} \right) \\&= \frac{1}{4} (4k+2) \\&= k + 0.5\end{aligned}$$

- Result varies by operation algorithm and accuracy

$$\begin{aligned}x(n, m) &= k + 0.5000000000... \quad (k+1) \\&\text{or} \\&= k + 0.4999999999... \quad (k)\end{aligned}$$

IDCTミスマッチ対策

- 量子化値の制限
 - H.261... 偶数値の奇数化
 - MPEG-1... 偶数値の奇数化(0方向に1近づける)
逆量子化の前に処理
 - MPEG-2... 8x8DCT係数の最高周波数の係数のLSBを反転
逆量子化の後に処理
- IDCT精度の規定
 - IEEE 1180 (Dec. 1990)

IDCT Mismatch Remedy

- Restriction to Quantized Value
 - H.261... Oddification of even value
 - MPEG-1... Oddification of even value (decrease 1 towards 0)
Process before inverse quantization
 - MPEG-2... reverse LSB of a coefficient at highest frequency in 8x8DCT coefficients
Process before inverse quantization
- Standard for IDCT Accuracy
 - IEEE 1180 (Dec. 1990)