Sample 4-5

線形シフト不変システム

畳み込み行列

画像処理特論

村松 正吾

動作確認: MATLAB R2023a

Linear shift-invariant systems

Convolution matrix

Advanced Topics in Image Processing

Shogo MURAMATSU

Verified: MATLAB R2023a

準備

(Preparation)

close all

単変量畳み込み

(Univariate convolution)

有限インパルス応答(FIR) $\{h[n] \in \mathbb{R}\}_{n \in \Omega_{h} \subset \mathbb{Z}}$ を有する線形シフト不変システム $T(\cdot)$ を仮定する. (Let us assume a linear shift-invariant system $T(\cdot)$ with a finite impulse response (FIR) $\{h[n] \in \mathbb{R}\}_{n \in \Omega_{h} \subset \mathbb{Z}}$.)

 $T(\cdot)$ の配列 $\{u[n] \in \mathbb{R}\}_{n \in \Omega_{\mathbf{u}} \subset \mathbb{Z}}$ に対する応答 $\{v[n] \in \mathbb{R}\}_{n \in \Omega_{\mathbf{v}} \subset \mathbb{Z}}$ は畳み込み (The response $\{v[n] \in \mathbb{R}\}_{n \in \Omega_{\mathbf{v}} \subset \mathbb{Z}}$ to a sequence $\{u[n] \in \mathbb{R}\}_{n \in \Omega_{\mathbf{v}} \subset \mathbb{Z}}$ of $T(\cdot)$ can be represented by convolution)

$$\{v[n]\}_n = T(\{u[n]\}_n) = \sum_{k \in \Omega_{\rm h}} h[k] \{u[n-k]\}_n.$$

により表現できる.

信号の生成

(Signal generation)

% Generating an input sequence u[n] of finite support region Q = 6;

インパルス応答の設定

(Setting the impulse response)

```
% Setting the shift amount
h = [1 0 -1];
```

写像の定義

(Definition of a map)

```
% Definition of map T as a convolution with h[n] mapT = @(x) conv(x,h);
```

写像の結果

(Result of mapping)

単変量畳み込みの行列表現

(Matrix representation of the univariate convolution)

FIR システムの畳み込み演算は (The convolution of an FIR system can be represented as a matrix as) $\mathbf{v} = \mathbf{T}\mathbf{u}$,

のように行列表現できる.

インパルス応答 $\{h[n]\}_n$ のサポート領域が $\Omega_h = \{0,1,2\}$, 入力信号 $\{u[n]\}_n$ のサポート領域が $\Omega_u = \{0,1,2,3,4,5\}$ のとき,出力信号 $\{v[n]\}_n$ のサポート領域は $\Omega_v = \{0,1,2,3,4,5,6,7\}$ となり, (When the support region of the impulse response $\{h[n]\}_n$ is $\Omega_h = \{0,1,2\}$ and the support region of the input signal $\{u[n]\}_n$ is $\Omega_u = \{0,1,2,3,4,5\}$, the support region of the output signal $\{v[n]\}_n$ is $\Omega_v = \{0,1,2,3,4,5,6,7\}$, and we have)

$$\mathbf{v} = \text{vec}(\{v[n]\}_n) = \begin{pmatrix} v[0] \\ v[1] \\ v[2] \\ v[3] \\ v[4] \\ v[5] \\ v[6] \\ v[7] \end{pmatrix}, \mathbf{u} = \text{vec}(\{u[n]\}_n) = \begin{pmatrix} u[0] \\ u[1] \\ u[2] \\ u[3] \\ u[4] \\ u[5] \end{pmatrix}$$

$$\mathbf{T} = \begin{pmatrix} h[0] & 0 & 0 & 0 & 0 & 0 \\ h[1] & h[0] & 0 & 0 & 0 & 0 \\ h[2] & h[1] & h[0] & 0 & 0 & 0 \\ 0 & h[2] & h[1] & h[0] & 0 & 0 \\ 0 & 0 & h[2] & h[1] & h[0] & 0 \\ 0 & 0 & 0 & h[2] & h[1] & h[0] \\ 0 & 0 & 0 & 0 & h[2] & h[1] \\ 0 & 0 & 0 & 0 & 0 & h[2] \end{pmatrix}.$$

と表現できる. なお、出力のサポート領域が $\Omega_{\rm v}=\{(n+k)\in\mathbb{Z}|n\in\Omega_{\rm u},k\in\Omega_{\rm h}\}$ となり T の行数が (Note that the support area of the output is $\Omega_{\rm v}=\{(n+k)\in\mathbb{Z}|n\in\Omega_{\rm u},k\in\Omega_{\rm h}\}$ and the number of rows in T is

$$|\Omega_v| = |\Omega_u| + |\Omega_h| - 1 = 6 + 3 - 1 = 8.$$

となることに注意する.

単変量畳み込みの行列生成

(Matrix generation of univariate convolution)

```
% Find the matrix representation of the univariate convolution
T = zeros(length(u)+length(h)-1,length(u));
for idx = 1:length(u)
    % Generating a standard basis vector
    e = zeros(size(u),'like',u);
    e(idx) = 1;
    % Response to the standard basis vector
    T(:,idx) = mapT(e);
end
% Matrix representation of the univariate convolution
T
```

```
0 0 0 -1 0 1
0 0 0 0 -1 6
0 0 0 0 -1
```

関数 CONVMTX の利用

(Using the CONVMTX function)

単変量畳み込み行列の生成に関数 CONVMTX も利用できる.

(The function CONVMTX can also be used to generate univariate convolutional matrices.)

% Generating the matrix representation of the univariate convolution by CONVMTX H = convmtx(h(:), length(u)) %#ok

```
H = 8×6

1 0 0 0 0 0 0
0 1 0 0 0 0
-1 0 1 0 0 0
0 -1 0 1 0 0
0 0 -1 0 1 0
0 0 0 -1 0 1
0 0 0 0 -1 0
0 0 0 0 -1 0
```

行列演算による単変量畳み込み

(Univariate convolution by matrix operation)

畳み込みは可換図に沿って (Convolution can be computed as)

$$\{v[n]\}_n = \operatorname{vec}_{\Omega_v}^{-1}(\mathbf{v}) = \operatorname{vec}_{\Omega_v}^{-1}(\mathbf{T}\mathbf{u}) = \operatorname{vec}_{\Omega_v}^{-1} \circ \mathbf{T} \operatorname{vec}_{\Omega_u}(\{u[n]\}_n)$$

のように行列演算が可能である. すなわち, (along the commutative diagram. That is, we have)

$$T = \operatorname{vec}_{\Omega_{\cdot \cdot}}^{-1} \circ \operatorname{Tvec}_{\Omega_{\cdot \cdot}}$$

と表現できる.

```
% Column vectorization of sequence u[n]
vecu = u(:);

% Matrix operation
vecv = T*vecu;

% Reshaping the result into the row sequence
recv = reshape(vecv,[1 (length(u)+length(h)-1)])
```

recv = 1×8 1 2 2 2 2 2 -5 -6

行列演算による単変量畳み込みの評価

(Evaluation of univariate convolution by matrix operation)

```
% Comparizon between mapping and matrix operation
mymse = @(x,y) mean((double(x)-double(y)).^2,'all');
mymse(v,recv)
```

ans = 0

2変量畳み込み

(Bivariate convolution)

2 変量の有限インパルス応答(FIR) $\{h[\mathbf{n}]\in\mathbb{R}\}_{\mathbf{n}\in\Omega_\mathbf{n}\subset\mathbb{Z}^2}$ を有する線形シフト不変システム $T(\cdot)$ を仮定する.

(Assume a linear shift-invariant system T with a bivariate finite impulse response (FIR) $\{h[\mathbf{n}] \in \mathbb{R}\}_{\mathbf{n} \in \Omega_h \subset \mathbb{Z}^2}$.)

 $T(\cdot)$ の配列 $\{u[\mathbf{n}] \in \mathbb{R}\}_{\mathbf{n} \in \Omega_\mathbf{u} \subset \mathbb{Z}^2}$ に対する応答 $\{v[\mathbf{n}] \in \mathbb{R}\}_{\mathbf{n} \in \Omega_\mathbf{u} \subset \mathbb{Z}^2}$ は畳み込み (The response $\{v[\mathbf{n}] \in \mathbb{R}\}_{\mathbf{n} \in \Omega_\mathbf{u} \subset \mathbb{Z}^2}$ to the array $\{u[\mathbf{n}] \in \mathbb{R}\}_{\mathbf{n} \in \Omega_\mathbf{u} \subset \mathbb{Z}^2}$ of $T(\cdot)$ can be represented by convolution)

$$\{v[\mathbf{n}]\}_{\mathbf{n}} = T(\{u[\mathbf{n}]\}_{\mathbf{n}}) = \sum_{k \in \Omega_{\mathbf{h}}} h[\mathbf{k}]\{u[\mathbf{n} - \mathbf{k}]\}_{\mathbf{n}}.$$

により表現できる.

信号の生成

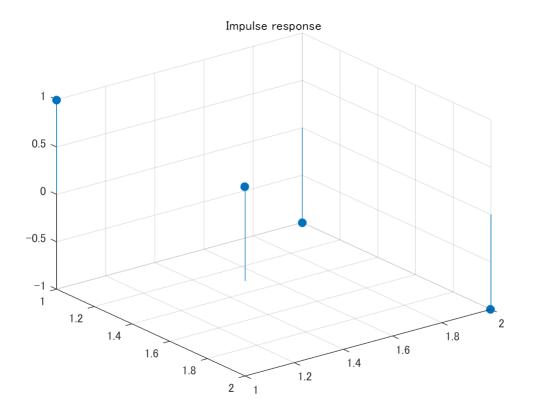
(Signal generation)

```
% Generating an input array u[n1,n2] of finite support region
N1 = 2; % # of rows
N2 = 3; % # of columns
ugen = "reshape((0:N1*N2-1),[N1 N2])";
u = eval(ugen)
```

u = 2×3 0 2 4 1 3 5

インパルス応答の設定

(Setting the impulse response)



写像の定義

(Definition of a map)

```
% Definition of map T as a convolution with h[n]
mapT = @(x) imfilter(x,h,'conv','full'); % or conv2(x,h);
```

写像の結果

(Result of mapping)

```
% Mapping with the circular shift T(.)
v = mapT(u)
```

```
v = 3 \times 4
0 2 2 -4
1 4 4 -9
1 2 2 -5
```

2変量畳み込みの行列表現

(Matrix representation of the bivariate convolution)

2 変量 FIR システムの畳み込み演算は (The convolution of a bivariate FIR system can be represented as a matrix as)

```
v = Tu,
```

のように行列表現できる.

インパルス応答 $\{h[\mathbf{n}]\}_{\mathbf{n}}$ のサポート領域が $\Omega_h = \{0,1\} \times \{0,1\}$, 入力信号 $\{u[\mathbf{n}]\}_{\mathbf{n}}$ のサポート領域が $\Omega_u = \{0,1\} \times \{0,1,2\}$ のとき,出力信号 $\{v[\mathbf{n}]\}_{\mathbf{n}}$ のサポート領域は $\Omega_v = \{0,1,2\} \times \{0,1,2,3\}$ となり,(If the support region of the impulse response $\{h[\mathbf{n}]\}_{\mathbf{n}}$ is $\Omega_h = \{0,1\} \times \{0,1\}$, and the support region of the input signal $\{u[\mathbf{n}]\}_{\mathbf{n}}$ is $\Omega_u = \{0,1\} \times \{0,1,2\}$, then the support region of the output signal $\{v[\mathbf{n}]\}_{\mathbf{n}}$ is $\Omega_v = \{0,1,2\} \times \{0,1,2,3\}$ and can be expressed as)

$$\mathbf{v} = \text{vec}(\{v[\mathbf{n}]\}_{\mathbf{n}}) = \begin{pmatrix} v[0,0] \\ v[1,0] \\ v[2,0] \\ v[0,1] \\ \vdots \\ v[2,3] \end{pmatrix}, \mathbf{u} = \text{vec}(\{u[\mathbf{n}]\}_{\mathbf{n}}) = \begin{pmatrix} u[0,0] \\ u[1,0] \\ u[0,1] \\ u[1,1] \\ \vdots \\ u[1,2] \end{pmatrix}$$

$$\mathbf{T} = \begin{pmatrix} h[0,0] & 0 & 0 & 0 & \cdots & 0 \\ h[1,0] & h[0,0] & 0 & 0 & \cdots & 0 \\ 0 & h[1,0] & h[0,0] & 0 & \cdots & 0 \\ h[1,0] & 0 & h[1,0] & h[0,0] & \cdots & 0 \\ h[1,1] & h[1,0] & 0 & h[1,0] & \cdots & 0 \\ 0 & h[1,1] & h[1,0] & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & h[1,1] \end{pmatrix}.$$

と表現できる. なお、出力のサポート領域が $\Omega_{\rm v}=\{(n+k)\in\mathbb{Z}|n\in\Omega_{\rm u},k\in\Omega_{\rm h}\}$ となり T の行数が (Note that the support region of the output is $\Omega_{\rm v}=\{(n+k)\in\mathbb{Z}|n\in\Omega_{\rm u},k\in\Omega_{\rm h}\}$ and the number of rows in Tis)

$$|\Omega_{\rm v}| = (2+2-1) \times (3+2-1) = 2 \times 4 = 12.$$

となることに注意する.

2 変量畳み込みの行列生成

(Matrix generation of bivariate matrix)

```
% Find the matrix representation of the convolution
T = zeros(prod(size(u)+size(h)-1),numel(u));
for idx = 1:numel(u)
    % Generating a standard basis vector
    e = zeros(size(u),'like',u);
    e(idx) = 1;
    % Response to the standard basis vector
    T(:,idx) = reshape(mapT(e),[size(T,1) 1]);
end
% Matrix representation of the convolution
```

```
T = 12 \times 6
       0
           0
               0
                   0
                       0
   1
   1
       1
           0
               0
                   0
                       0
   0
       1
           0
               0
                   0
                       0
  -1
       0
           1
               0
                   0
                       0
                  0
      -1
           1
               1
                       0
  -1
   0
      -1
          0
              1
                  0
                       0
   0
     0
          -1
              0 1
                       0
   0
     0 -1 -1 1
                       1
     0 0 -1
                  0
                       1
```

関数 CONVMTX2 の利用

(Using the CONVMTX2 function)

2変量畳み込み行列の生成に関数 CONVMTX2 も利用できる.

(The function CONVMTX2 can also be used to generate bivariate convolutional matrices.)

% Generating the matrix representation of the bivariate convolution by CONVMTX2
H = convmtx2(h,size(u)) %#ok

```
H =
   (1,1)
                1
   (2,1)
                1
   (4,1)
               -1
   (5,1)
               -1
   (2,2)
               1
   (3,2)
               1
   (5,2)
               -1
               -1
   (6,2)
   (4,3)
               1
   (5,3)
                1
   (7,3)
               -1
   (8,3)
               -1
   (5,4)
                1
   (6,4)
                1
   (8,4)
               -1
               -1
   (9,4)
   (7,5)
               1
   (8,5)
               1
  (10,5)
               -1
  (11,5)
               -1
   (8,6)
               1
               1
   (9,6)
  (11,6)
               -1
  (12,6)
               -1
```

行列演算による2変量畳み込み

(Bivariate convolution by matrix operation)

畳み込みは可換図に沿って (Convolution can be computed as)

$$\{v[n]\}_n = \operatorname{vec}_{\Omega_v}^{-1}(\mathbf{v}) = \operatorname{vec}_{\Omega_v}^{-1}(\mathbf{T}\mathbf{u}) = \operatorname{vec}_{\Omega_v}^{-1} \circ \mathbf{T} \operatorname{vec}_{\Omega_u}(\{u[n]\}_n)$$

のように行列演算が可能である. すなわち, (along the commutative diagram. That is, we have)

$$T = \operatorname{vec}_{\Omega_{v}}^{-1} \circ \operatorname{Tvec}_{\Omega_{u}}.$$

と表現できる.

```
% Column vectorization of sequence u[n]
vecu = u(:);

% Matrix operation
vecv = T*vecu;

% Reshaping the result into the row sequence
recv = reshape(vecv,(size(u)+size(h)-1))
```

recv = 3×4 0 2 2 -4 1 4 4 -9 1 2 2 -5

行列演算による単変量畳み込みの評価

(Evaluation of univariate convolution by matrix operation)

```
% Comparizon between mapping and matrix operation
mymse(v,recv)
```

ans = 0

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