Sample 13-3

辞書学習

K-特異值分解

画像処理特論

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動作確認: MATLAB R2020a

Dictionary learning

K-SVD

Advanced Topics in Image Processing

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Verified: MATLAB R2020a

準備

(Preparation)

```
clear
close all
import msip.download_img
msip.download_img
```

lena.png already exists in ./data/
baboon.png already exists in ./data/
goldhill.png already exists in ./data/
barbara.png already exists in ./data/

パラメータ設定

(Parameter settings)

- ・ブロックサイズ (Block size)
- 冗長率 (Redundancy ratio)
- スパース率 (Sparsity ratio)
- ・繰返し回数 (Number of iterations)

```
% Block size
szBlk = [8 8];
nDims = prod(szBlk);

% Redundancy ratio
redundancyRatio = 1.5;

% Sparsity ratio
```

```
sparsityRatio = 0.1;

% Number of iterations
nIters = 30;
```

画像の読込

(Read image)

• $\mathbf{u} \in \mathbb{R}^N$

```
% Read image
u = rgb2gray(im2double(imread('./data/lena.png')));
```

K-特異值分解

(K-Singular Value Decomposition)

問題設定 (Problem setting):

$$\{\widehat{\mathbf{\Phi}}, \{\widehat{\mathbf{x}}_n\}\} = \arg\min_{\{\mathbf{\Phi}, \{\mathbf{x}_n\}\}} \frac{1}{2S} \sum_{n=1}^{S} \|\mathbf{y}_n - \mathbf{\Phi}\widehat{\mathbf{x}}_n\|_2^2, \quad \text{s.t. } \forall n, \|\mathbf{x}_n\|_0 \le K$$

アルゴリズム (Algorithm):

スパース近似ステップと辞書更新ステップを繰返す.

• スパース近似ステップ (Sparse approximation step)

$$\widehat{\mathbf{x}}_n = \arg\min_{\mathbf{x}_n} \frac{1}{2} \|\mathbf{y}_n - \widehat{\mathbf{\Phi}} \mathbf{x}_n\|_2^2 \quad \text{s.t. } \|\mathbf{x}_n\|_0 \le K$$

• 辞書更新ステップ (Dictionary update step)

$$\widehat{\mathbf{\Phi}} = \arg\min_{\mathbf{\Phi}} \frac{1}{2S} \sum_{n=1}^{S} \|\mathbf{y}_n - \mathbf{\Phi} \widehat{\mathbf{x}}_n\|_2^2 = \arg\min_{\mathbf{\Phi}} \frac{1}{2S} \left\| \left(\mathbf{Y} - \sum_{p \neq k} \mathbf{\phi}_p \widehat{\mathbf{X}}_{p,:} \right) - \mathbf{\phi}_k \widehat{\mathbf{X}}_{k,:} \right\|_F^2$$

画像 uからのデータ行列 Y の生成 (Generation of data matrix Y of image u)

```
Y = im2col(u,szBlk,'distinct');
nSamples = size(Y,2);
```

要素画像の数 (Number of atomic images)

```
nAtoms = ceil(redundancyRatio*nDims);
```

係数の数 (Number of coefficients)

```
nCoefs = floor(sparsityRatio*nDims);
```

辞書 Φの初期化 (Initializatio of dictionary Φ)

- ・二変量離散コサイン変換(Bivariate DCT)
- ・ランダム (random)

```
Phi = randn(nDims,nAtoms);
Phi = Phi/norm(Phi,'fro');
for iAtom = 1:nDims
    delta = zeros(szBlk);
    delta(iAtom) = 1;
    Phi(:,iAtom) = reshape(idct2(delta),nDims,1);
end
```

要素ベクトルを要素画像に変換 (Reshape the atoms into atomic images)

```
atomicImages = zeros(szBlk(1),szBlk(2),nAtoms);
for iAtom = 1:nAtoms
   atomicImages(:,:,iAtom) = reshape(Phi(:,iAtom),szBlk(1),szBlk(2));
end
```

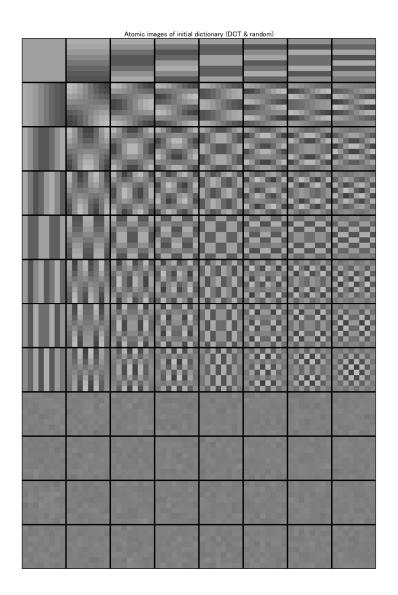
画像表示

(Image show)

```
figure(1)
imshow(u);
title('Original image u')
```



figure(2)
montage(imresize(atomicImages,8,'nearest')+.5,'BorderSize',[2 2],'Size',[ceil(nAtoms/8) 8])
title('Atomic images of initial dictionary (DCT & random)')



スパース近似ステップと辞書更新ステップの繰り返し

• スパース近似: 直交マッチング追跡 (OMP)

・辞書更新: 特異値分解と 1-ランク近似 (SVD and 1-rank approximation)

辞書更新の内容

1. $k \leftarrow 1$

2. 誤差行列 \mathbf{E}_k を定義 : \mathbf{E}_k : = $\mathbf{Y} - \sum_{p \neq k} \mathbf{\phi}_p \hat{\mathbf{X}}_{p,:}$

3. データ行 $\hat{\mathbf{X}}_{k,:}$ の非零値を抽出する行列 Ω_k を定義: $\hat{\mathbf{X}}_{k,:}^R = \hat{\mathbf{X}}_{k,:}\Omega_k \Leftrightarrow \hat{\mathbf{X}}_{k,:}^R\Omega_k^T = \hat{\mathbf{X}}_{k,:}$

4. 誤差行列 \mathbf{E}_k を行列 $\mathbf{\Omega}_k$ で縮退: $\mathbf{E}_k^R := \mathbf{E}_k \mathbf{\Omega}_k$

5. 縮退した誤差行列 \mathbf{E}_k^R を特異値分解: $\mathbf{E}_k^R = \mathbf{U}\mathbf{S}\mathbf{V}^T = (\mathbf{u}_1,\mathbf{u}_2,\cdots,\mathbf{u}_r)\mathrm{diag}(\sigma_1,\sigma_2,\cdots,\sigma_r)(\mathbf{v}_1,\mathbf{v}_2,\cdots,\mathbf{v}_r)^T$

```
6. 要素ベクトル \phi_k を更新: \mathbf{k} \leftarrow \mathbf{u}_1
7. データ行\hat{\mathbf{X}}_{k,:} を更新: \hat{\mathbf{X}}_{k,:} \leftarrow \sigma_1 \mathbf{v}_1^T
```

- 8. $k \leftarrow k + 1$
- 9. *k* ≤ *N* ならば 2. **ヘ** *k* > *N* ならば終了

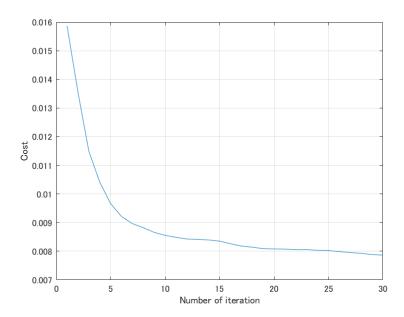
ただし、 σ_1 を最大特異値とする.

交互ステップの繰返し計算 (Iterative calculation of alternative steps)

```
cost = zeros(1,nIters);
for iIter = 1:nIters
    X = zeros(nAtoms,nSamples);
    % Sparse approximation
    for iSample = 1:nSamples
        y = Y(:,iSample);
        x = omp(y, Phi, nCoefs);
        X(:,iSample) = x;
    end
    % Dictionary update
    for iAtom = 1:nAtoms
        idxset = setdiff(1:nAtoms,iAtom);
        xk = X(iAtom,:);
        suppk = find(xk);
        Ekred = Y(:,suppk)-Phi(:,idxset)*X(idxset,suppk);
        if ~isempty(suppk)
            [U,S,V] = svd(Ekred, 'econ');
            ak = U(:,1);
            xkred = S(1,1)*V(:,1)';
            %
            Phi(:,iAtom) = ak;
            X(iAtom, suppk) = xkred;
        end
    end
    cost(iIter) = (norm(Y-Phi*X, 'fro')^2)/(2*nSamples);
end
```

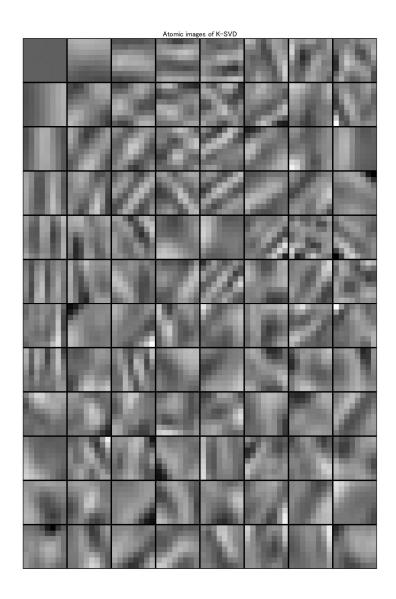
コスト評価のグラフ (Graph of cost variation)

```
figure(3)
plot(cost)
xlabel('Number of iteration')
ylabel('Cost')
grid on
```



要素ベクトルを要素画像に変換 (Reshape the atoms into atomic images)

```
atomicImages = zeros(szBlk(1),szBlk(2),nAtoms);
for iAtom = 1:nAtoms
    atomicImages(:,:,iAtom) = reshape(Phi(:,iAtom),szBlk(1),szBlk(2));
end
figure(4)
montage(imresize(atomicImages,8,'nearest')+.5,'BorderSize',[2 2],'Size',[ceil(nAtoms/8) 8])
title('Atomic images of K-SVD')
```



直交マッチング追跡関数

(Function of orthogonal matching pursuite)

```
rr = r.'*r;
        for m = setdiff(1:nAtoms, supp)
             d = Phi(:,m);
             g(m) = d.'*r; \% \gamma m = \langle dm, r \rangle
             a(m) = g(m)/(d.'*d); % Normalize \alpha m = \gamma m/||dm||^2
             e(m) = rr - g(m)*a(m); % < r-dm/||dm||^2, r>
        end
        % Minimum value search (pursuit)
        [~,mmin]= min(e);
        % Update the support
        supp = union(supp,mmin);
        subPhi = Phi(:,supp);
        x(supp) = pinv(subPhi) * y;
        % Synthesis process
        v = Phi*x;
        % Residual
        r = y - v;
        % Update
        k = k + 1;
    end
end
```

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