

# 学习FastICA算法的代码实现

---

Alex / 2022-07-16 / [free\\_learner@163.com](mailto:free_learner@163.com) / [AlexBrain.cn](http://AlexBrain.cn)

更新于2023-09-21，主要是文字排版上的更新，内容基本保持不变。

学习FastICA算法的MATLAB实现细节。

## 一、背景

---

独立成分分析和FastICA算法原理请参考我以前的[博客](#)或下面的文献：

Hyvärinen, A., & Oja, E. (2000). Independent component analysis: algorithms and applications. *Neural networks*, 13(4-5), 411-430.

FastICA的MATLAB代码来自于[这里](#)。

我下面的描述的方式是，在原始代码上添加自己的注释，代码原本的注释以 % 开头，我添加的注释以 %% 开头。

## 二、整体结构

---

在 `Contents.m` 文件里可以看到每个函数的用途，和FastICA算法有关的只有5个函数文件：

`fastica.m` 作为主函数来解析参数和调用其他函数，`fpica.m` 用于实现核心算法（approximation of negentropy & fixed-point iteration）、`whitenenv.m` 用于白化数据、`pcamat.m` 用于主成分分析、`remmean.m` 用于去除均值。个人学习感受是，`fpica.m` 比较复杂，其他函数相对简单。

## 三、具体函数

---

### 1. fastica.m

```

function [Out1, Out2, Out3] = fastica(mixedsig, varargin)
%FASTICA - Fast Independent Component Analysis
%
% FastICA for Matlab 7.x and 6.x
% Version 2.5, October 19 2005
% Copyright (c) Hugo Gåvert, Jarmo Hurri, Jaakko Särelä, and Aapo Hyvärinen.
%
% FASTICA(mixedsig) estimates the independent components from given
% multidimensional signals. Each row of matrix mixedsig is one
% observed signal. FASTICA uses Hyvarinen's fixed-point algorithm,
% see http://www.cis.hut.fi/projects/ica/fastica/. Output from the
% function depends on the number output arguments:
%
% [icasig] = FASTICA (mixedsig); the rows of icasig contain the
% estimated independent components.
%
% [icasig, A, W] = FASTICA (mixedsig); outputs the estimated separating
% matrix W and the corresponding mixing matrix A.
%
% [A, W] = FASTICA (mixedsig); gives only the estimated mixing matrix
% A and the separating matrix W.
%
% Some optional arguments induce other output formats, see below.
%
% A graphical user interface for FASTICA can be launched by the
% command FASTICAG
%
% FASTICA can be called with numerous optional arguments. Optional
% arguments are given in parameter pairs, so that first argument is
% the name of the parameter and the next argument is the value for
% that parameter. Optional parameter pairs can be given in any order.
%
% OPTIONAL PARAMETERS:
%
% Parameter name      Values and description
%
%=====%
% --Basic parameters in fixed-point algorithm:
%
% 'approach'          (string) The decorrelation approach used. Can be
%                      symmetric ('symm'), i.e. estimate all the
%                      independent component in parallel, or
%                      deflation ('defl'), i.e. estimate independent
%                      component one-by-one like in projection pursuit.
%                      Default is 'defl'.
%
% 'numOfIC'            (integer) Number of independent components to

```

```

%
% be estimated. Default equals the dimension of data.
%
%=====
%
% --Choosing the nonlinearity:
%
% 'g'          (string) Chooses the nonlinearity g used in
%               the fixed-point algorithm. Possible values:
%
%               Value of 'g':      Nonlinearity used:
%               'pow3' (default)   g(u)=u^3
%               'tanh'           g(u)=tanh(a1*u)
%               'gauss'          g(u)=u*exp(-a2*u^2/2)
%               'skew'            g(u)=u^2
%
% 'finetune'    (string) Chooses the nonlinearity g used when
%               fine-tuning. In addition to same values
%               as for 'g', the possible value 'finetune' is:
%               'off'             fine-tuning is disabled.
%
% 'a1'          (number) Parameter a1 used when g='tanh'.
%               Default is 1.
%
% 'a2'          (number) Parameter a2 used when g='gaus'.
%               Default is 1.
%
% 'mu'          (number) Step size. Default is 1.
%               If the value of mu is other than 1, then the
%               program will use the stabilized version of the
%               algorithm (see also parameter 'stabilization').
%
%
% 'stabilization' (string) Values 'on' or 'off'. Default 'off'.
% This parameter controls weather the program uses
% the stabilized version of the algorithm or
% not. If the stabilization is on, then the value
% of mu can momentarily be halved if the program
% senses that the algorithm is stuck between two
% points (this is called a stroke). Also if there
% is no convergence before half of the maximum
% number of iterations has been reached then mu
% will be halved for the rest of the rounds.
%
%=====
%
% --Controlling convergence:
%
% 'epsilon'     (number) Stopping criterion. Default is 0.0001.
%
% 'maxNumIterations' (integer) Maximum number of iterations.
%                     Default is 1000.

```

```

%
% 'maxFinetune'          (integer) Maximum number of iterations in
%                         fine-tuning. Default 100.
%
%
% 'sampleSize'           (number) [0 - 1] Percentage of samples used in
%                         one iteration. Samples are chosen in random.
%                         Default is 1 (all samples).
%
%
% 'initGuess'            (matrix) Initial guess for A. Default is random.
%                         You can now do a "one more" like this:
%                         [ica, A, W] = fastica(mix, 'numOfIC',3);
%                         [ica2, A2, W2] = fastica(mix, 'initGuess', A, 'numOfIC', 4);
%
%=====
%
% --Graphics and text output:
%
%
% 'verbose'               (string) Either 'on' or 'off'. Default is
%                         'on': report progress of algorithm in text format.
%
%
% 'displayMode'           (string) Plot running estimates of independent
%                         components: 'signals', 'basis', 'filters' or
%                         'off'. Default is 'off'.
%
%
% 'displayInterval'       Number of iterations between plots.
%                         Default is 1 (plot after every iteration).
%
%=====
%
% --Controlling reduction of dimension and whitening:
%
%
% Reduction of dimension is controlled by 'firstEig' and 'lastEig', or
% alternatively by 'interactivePCA'.
%
%
% 'firstEig'               (integer) This and 'lastEig' specify the range for
%                         eigenvalues that are retained, 'firstEig' is
%                         the index of largest eigenvalue to be
%                         retained. Default is 1.
%
%
% 'lastEig'                (integer) This is the index of the last (smallest)
%                         eigenvalue to be retained. Default equals the
%                         dimension of data.
%
%
% 'interactivePCA'         (string) Either 'on' or 'off'. When set 'on', the
%                         eigenvalues are shown to the user and the
%                         range can be specified interactively. Default
%                         is 'off'. Can also be set to 'gui'. Then the user
%                         can use the same GUI that's in FASTICAG.
%
%
% If you already know the eigenvalue decomposition of the covariance

```

```

% matrix, you can avoid computing it again by giving it with the
% following options:
%
% 'pcaE'           (matrix) Eigenvectors
% 'pcaD'           (matrix) Eigenvalues
%
% If you already know the whitened data, you can give it directly to
% the algorithm using the following options:
%
% 'whiteSig'        (matrix) Whitened signal
% 'whiteMat'        (matrix) Whitening matrix
% 'dewhiteMat'      (matrix) deweighting matrix
%
% If values for all the 'whiteSig', 'whiteSig' and 'dewhiteMat' are
% supplied, they will be used in computing the ICA. PCA and whitening
% are not performed. Though 'mixedsig' is not used in the main
% algorithm it still must be entered - some values are still
% calculated from it.
%
% Performing preprocessing only is possible by the option:
%
% 'only'            (string) Compute only PCA i.e. reduction of
%                   dimension ('pca') or only PCA plus whitening
%                   ('white'). Default is 'all': do ICA estimation
%                   as well. This option changes the output
%                   format accordingly. For example:
%
% [whitesig, WM, DWM] = FASTICA(mixedsig,
% 'only', 'white')
% returns the whitened signals, the whitening matrix
% (WM) and the deweighting matrix (DWM). (See also
% WHITENV.) In FastICA the whitening matrix performs
% whitening and the reduction of dimension. Deweighting
% matrix is the pseudoinverse of whitening matrix.
%
% [E, D] = FASTICA(mixedsig, 'only', 'pca')
% returns the eigenvector (E) and diagonal
% eigenvalue (D) matrices containing the
% selected subspaces.
%
%=====
%
% EXAMPLES
%
% [icasig] = FASTICA (mixedsig, 'approach', 'symm', 'g', 'tanh');
% Do ICA with tanh nonlinearity and in parallel (like
% maximum likelihood estimation for supergaussian data).
%
% [icasig] = FASTICA (mixedsig, 'lastEig', 10, 'numOfIC', 3);

```

```

% Reduce dimension to 10, and estimate only 3
% independent components.
%
% [icasig] = FASTICA (mixedsig, 'verbose', 'off', 'displayMode', 'off');
% Don't output convergence reports and don't plot
% independent components.
%
%
% A graphical user interface for FASTICA can be launched by the
% command FASTICAG
%
% See also FASTICAG

% @(#$)Id: fastica.m,v 1.14 2005/10/19 13:05:34 jarmo Exp $

%%%%%%%%%%%%%%%
%% check the number of inputs, the dimension of mixed signals, any NaNs and double precision or
not
% Check some basic requirements of the data
if nargin == 0,
    error ('You must supply the mixed data as input argument.');
end

if length (size (mixedsig)) > 2,
    error ('Input data can not have more than two dimensions.');
end

if any (any (isnan (mixedsig))),
    error ('Input data contains NaN''s.');
end

if ~isa (mixedsig, 'double')
    fprintf ('Warning: converting input data into regular (double) precision.\n');
    mixedsig = double (mixedsig);
end

%%%%%%%%%%%%%%%
% Remove the mean and check the data
% the first step is removing the row-wise mean

[mixedsig, mixedmean] = remmean(mixedsig);

[Dim, NumOfSampl] = size(mixedsig);

%%%%%%%%%%%%%%%
% Default values for optional parameters

```

```

% All
verbose          = 'on';

% Default values for 'pcamat' parameters
firstEig        = 1;
lastEig         = Dim;
interactivePCA  = 'off';

% Default values for 'fpica' parameters
approach        = 'defl';
numOfIC          = Dim;
g                = 'pow3';
finetune         = 'off';
a1               = 1;
a2               = 1;
myy              = 1;
stabilization   = 'off';
epsilon          = 0.0001;
maxNumIterations = 1000;
maxFinetune     = 5;
initState        = 'rand';
guess             = 0;
sampleSize        = 1;
displayMode      = 'off';
displayInterval  = 1;

%%%%%%%%%%%%%%%
% Parameters for fastICA - i.e. this file
%% jumpPCA=0, jumpWhitening=0 means that PCA and whitening would not be skipped
%% only=3 means doing preprocessing and ICA (instead of PCA or whitening only)
%% userNumOfIC=0 means the IC numbers would be automatically determined.

b_verbose = 1;
jumpPCA = 0;
jumpWhitening = 0;
only = 3;
userNumOfIC = 0;

%%%%%%%%%%%%%%
% pay attention to the way of parsing arguments and dealing with upper/lower cases
% Read the optional parameters

if (rem(length(varargin),2)==1)
    error('Optional parameters should always go by pairs');
else
    for i=1:2:(length(varargin)-1)
        if ~ischar (varargin{i}),
            error (['Unknown type of optional parameter name (parameter' ...

```

```

    ' names must be strings').']);
end

% change the value of parameter
switch lower (varargin{i})
case 'stabilization'
    stabilization = lower (varargin{i+1});
case 'maxfinetune'
    maxFinetune = varargin{i+1};
case 'samplesize'
    sampleSize = varargin{i+1};
case 'verbose'
    verbose = lower (varargin{i+1});
    % silence this program also
    if strcmp (verbose, 'off'), b_verbose = 0; end
case 'firsteig'
    firstEig = varargin{i+1};
case 'lasteig'
    lastEig = varargin{i+1};
case 'interactivepca'
    interactivePCA = lower (varargin{i+1});
case 'approach'
    approach = lower (varargin{i+1});
case 'numofic'
    numOfIC = varargin{i+1};
    % User has supplied new value for numOfIC.
    % We'll use this information later on...
    userNumOfIC = 1;
case 'g'
    g = lower (varargin{i+1});
case 'finetune'
    finetune = lower (varargin{i+1});
case 'a1'
    a1 = varargin{i+1};
case 'a2'
    a2 = varargin{i+1};
case {'mu', 'myy'}
    myy = varargin{i+1};
case 'epsilon'
    epsilon = varargin{i+1};
case 'maxnumiterations'
    maxNumIterations = varargin{i+1};
case 'initguess'
    % no use setting 'guess' if the 'initState' is not set
    initState = 'guess';
    guess = varargin{i+1};
case 'displaymode'
    displayMode = lower (varargin{i+1});
case 'displayinterval'

```

```

displayInterval = varargin{i+1};
case 'pcae'
% calculate if there are enough parameters to skip PCA
jumpPCA = jumpPCA + 1;
E = varargin{i+1};
case 'pcad'
% calculate if there are enough parameters to skip PCA
jumpPCA = jumpPCA + 1;
D = varargin{i+1};
case 'whitesig'
% calculate if there are enough parameters to skip PCA and whitening
jumpWhitening = jumpWhitening + 1;
whitesig = varargin{i+1};
case 'whitemat'
% calculate if there are enough parameters to skip PCA and whitening
jumpWhitening = jumpWhitening + 1;
whiteningMatrix = varargin{i+1};
case 'dewhitemat'
% calculate if there are enough parameters to skip PCA and whitening
jumpWhitening = jumpWhitening + 1;
dewhitenMatrix = varargin{i+1};
case 'only'
% if the user only wants to calculate PCA or...
switch lower (varargin{i+1})
case 'pca'
only = 1;
case 'white'
only = 2;
case 'all'
only = 3;
end

otherwise
% Hmm, something wrong with the parameter string
error(['Unrecognized parameter: "' varargin{i} '"']);
end;
end;
end

%%%%%%%%%%%%%
% print information about data
if b_verbose
fprintf('Number of signals: %d\n', Dim);
fprintf('Number of samples: %d\n', NumOfSampl);
end

% Check if the data has been entered the wrong way,
% but warn only... it may be on purpose

```

```

if Dim > NumOfSampl
    if b_verbose
        fprintf('Warning: ');
        fprintf('The signal matrix may be oriented in the wrong way.\n');
        fprintf('In that case transpose the matrix.\n\n');
    end
end

%%%%%%%%%%%%%%%
% Calculating PCA
% the second step is doing PCA

% We need the results of PCA for whitening, but if we don't
% need to do whitening... then we dont need PCA...
% to skip whitening (so PCA is also skipped), the whitened signals and whitening/dewhitening
matrices should be given
if jumpWhitening == 3
    if b_verbose,
        fprintf ('Whitened signal and corresponding matrises supplied.\n');
        fprintf ('PCA calculations not needed.\n');
    end;
else
    % OK, so first we need to calculate PCA
    % Check to see if we already have the PCA data
    %% if jumpPCA = 2, the eigenvalues and eigenvectors are given
    if jumpPCA == 2,
        if b_verbose,
            fprintf ('Values for PCA calculations supplied.\n');
            fprintf ('PCA calculations not needed.\n');
        end;
    else
        % display notice if the user entered one, but not both, of E and D.
        if (jumpPCA > 0) & (b_verbose),
            fprintf ('You must suply all of these in order to jump PCA:\n');
            fprintf ('''pcaE'', ''pcaD''.\n');
        end;

        % Calculate PCA
        [E, D]=pcamat(mixedsig, firstEig, lastEig, interactivePCA, verbose);
    end
end

% skip the rest if user only wanted PCA
if only > 1
% the third step is whitening the data
%%%%%%%%%%%%%%%
% Whitening the data

```

```

% Check to see if the whitening is needed...
if jumpWhitening == 3,
    if b_verbose,
        fprintf ('Whitening not needed.\n');
    end;
else

    % Whitening is needed
    % display notice if the user entered some of the whitening info, but not all.
    if (jumpWhitening > 0) & (b_verbose),
        fprintf ('You must suply all of these in order to jump whitening:\n');
        fprintf ('''whiteSig'', ''whiteMat'', ''dewhiteMat''.\n');
    end;

    % Calculate the whitening
    [whitesig, whiteningMatrix, dewhitenMatrix] = whitenv ...
        (mixedsig, E, D, verbose);
end

end % if only > 1

% skip the rest if user only wanted PCA and whitening
if only > 2
    %% the final step is fixed-point method, the core of FastICA
    %%%%%%%%%%%%%%
    % Calculating the ICA

    % Check some parameters
    % The dimension of the data may have been reduced during PCA calculations.
    % The original dimension is calculated from the data by default, and the
    % number of IC is by default set to equal that dimension.

    Dim = size(whitesig, 1);

    % The number of IC's must be less or equal to the dimension of data
    if numOfIC > Dim
        numOfIC = Dim;
        % Show warning only if verbose = 'on' and user supplied a value for 'numOfIC'
        if (b_verbose & userNumOfIC)
            fprintf('Warning: estimating only %d independent components\n', numOfIC);
            fprintf('(Can''t estimate more independent components than dimension of data)\n');
        end
    end

    % Calculate the ICA with fixed point algorithm.
    [A, W] = fpica (whitesig, whiteningMatrix, dewhitenMatrix, approach, ...
        numOfIC, g, finetune, a1, a2, myy, stabilization, epsilon, ...

```

```

    maxNumIterations, maxFinetune, initState, guess, sampleSize, ...
    displayMode, displayInterval, verbose);

% Check for valid return
if ~isempty(W)
    % Add the mean back in.
    if b_verbose
        fprintf('Adding the mean back to the data.\n');
    end
    icasig = W * mixedsig + (W * mixedmean) * ones(1, NumOfSampl);
    %icasig = W * mixedsig;
    %% max(abs(W * mixedmean)) > 1e-9 means the mean is non-zero
    if b_verbose & ...
        (max(abs(W * mixedmean)) > 1e-9) & ...
        (strcmp(displayMode,'signals') | strcmp(displayMode,'on'))
        fprintf('Note that the plots don''t have the mean added.\n');
    end
else
    icasig = [];
end

end % if only > 2

%%%%%%%%%%%%%%%
% The output depends on the number of output parameters
% and the 'only' parameter.

if only == 1      % only PCA
    Out1 = E;
    Out2 = D;
elseif only == 2  % only PCA & whitening
    if nargout == 2
        Out1 = whiteningMatrix;
        Out2 = dewhiteningMatrix;
    else
        Out1 = whitesig;
        Out2 = whiteningMatrix;
        Out3 = dewhiteningMatrix;
    end
else      % ICA
    if nargout == 2
        Out1 = A;
        Out2 = W;
    else
        Out1 = icasig;
        Out2 = A;
        Out3 = W;
    end
end

```

```
    end  
end
```

## 2. remmean.m

```
function [newVectors, meanValue] = remmean(vectors);  
%REMMEAN - remove the mean from vectors  
%  
% [newVectors, meanValue] = remmean(vectors);  
%  
% Removes the mean of row vectors.  
% Returns the new vectors and the mean.  
%  
% This function is needed by FASTICA and FASTICAG  
  
% @(#)$Id: remmean.m,v 1.2 2003/04/05 14:23:58 jarmo Exp $  
%% the vectors are mixed signals, where the row means variable, column means sample  
  
newVectors = zeros (size (vectors));  
meanValue = mean (vectors');  
newVectors = vectors - meanValue * ones (1,size (vectors, 2));
```

## 3. pcamat.m

```

function [E, D] = pcamat(vectors, firstEig, lastEig, s_interactive, ...
    s_verbose);
%PCAMAT - Calculates the pca for data
%
% [E, D] = pcamat(vectors, firstEig, lastEig, ...
%                 interactive, verbose);
%
% Calculates the PCA matrices for given data (row) vectors. Returns
% the eigenvector (E) and diagonal eigenvalue (D) matrices containing the
% selected subspaces. Dimensionality reduction is controlled with
% the parameters 'firstEig' and 'lastEig' - but it can also be done
% interactively by setting parameter 'interactive' to 'on' or 'gui'.
%
% ARGUMENTS
%
% vectors      Data in row vectors.
% firstEig     Index of the largest eigenvalue to keep.
%              Default is 1.
% lastEig      Index of the smallest eigenvalue to keep.
%              Default is equal to dimension of vectors.
% interactive   Specify eigenvalues to keep interactively. Note that if
%                you set 'interactive' to 'on' or 'gui' then the values
%                for 'firstEig' and 'lastEig' will be ignored, but they
%                still have to be entered. If the value is 'gui' then the
%                same graphical user interface as in FASTICAG will be
%                used. Default is 'off'.
% verbose       Default is 'on'.
%
%
% EXAMPLE
%     [E, D] = pcamat(vectors);
%
% Note
%     The eigenvalues and eigenvectors returned by PCAMAT are not sorted.
%
% This function is needed by FASTICA and FASTICAG
%
% For historical reasons this version does not sort the eigenvalues or
% the eigen vectors in any ways. Therefore neither does the FASTICA or
% FASTICAG. Generally it seems that the components returned from
% whitening is almost in reversed order. (That means, they usually are,
% but sometime they are not - depends on the EIG-command of matlab.)
%
% @(#) $Id: pcamat.m,v 1.5 2003/12/15 18:24:32 jarmo Exp $
%%%%%
%% the parameters are determined by positions instead of name-value pair in the main function

```

```

% Default values:
if nargin < 5, s_verbose = 'on'; end
if nargin < 4, s_interactive = 'off'; end
if nargin < 3, lastEig = size(vectors, 1); end
if nargin < 2, firstEig = 1; end

%%%%%%%%%%%%%
% Check the optional parameters;
switch lower(s_verbose)
case 'on'
    b_verbose = 1;
case 'off'
    b_verbose = 0;
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''verbose''\n', s_verbose));
end

switch lower(s_interactive)
case 'on'
    b_interactive = 1;
case 'off'
    b_interactive = 0;
case 'gui'
    b_interactive = 2;
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''interactive''\n', ...
        s_interactive));
end

oldDimension = size (vectors, 1);
%% if the interactive is off, check whether the firstEig & lastEig are valid
if ~(b_interactive)
    if lastEig < 1 | lastEig > oldDimension
        error(sprintf('Illegal value [ %d ] for parameter: ''lastEig''\n', lastEig));
    end
    if firstEig < 1 | firstEig > lastEig
        error(sprintf('Illegal value [ %d ] for parameter: ''firstEig''\n', firstEig));
    end
end

%%%%%%%%%%%%%
% Calculate PCA

% Calculate the covariance matrix.
%% cov(X, 1) means the covariance matrix is normalized by N instead of N-1
if b_verbose, fprintf ('Calculating covariance...\n'); end
covarianceMatrix = cov(vectors', 1);

```

```

% Calculate the eigenvalues and eigenvectors of covariance matrix.
[E, D] = eig (covarianceMatrix);

% The rank is determined from the eigenvalues - and not directly by
% using the function rank - because function rank uses svd, which
% in some cases gives a higher dimensionality than what can be used
% with eig later on (eig then gives negative eigenvalues).
rankTolerance = 1e-7;
maxLastEig = sum (diag (D) > rankTolerance);
if maxLastEig == 0,
    fprintf (['Eigenvalues of the covariance matrix are' ...
        ' all smaller than tolerance [ %g ].\n' ...
        'Please make sure that your data matrix contains' ...
        ' nonzero values.\nIf the values are very small,' ...
        ' try rescaling the data matrix.\n'], rankTolerance);
    error ('Unable to continue, aborting.');
end

% Sort the eigenvalues - decending.
eigenvalues = flipud(sort(diag(D)));

%%%%%%%%%%%%%
% Interactive part - command-line
if b_interactive == 1

    % Show the eigenvalues to the user
    hndl_win=figure;
    bar(eigenvalues);
    title('Eigenvalues');

    % ask the range from the user...
    % ... and keep on asking until the range is valid :-
    areValuesOK=0;
    while areValuesOK == 0
        firstEig = input('The index of the largest eigenvalue to keep? (1) ');
        lastEig = input(['The index of the smallest eigenvalue to keep? (' ...
            int2str(oldDimension) ') ']);
        % Check the new values...
        % if they are empty then use default values
        if isempty(firstEig), firstEig = 1;end
        if isempty(lastEig), lastEig = oldDimension;end
        % Check that the entered values are within the range
        areValuesOK = 1;
        if lastEig < 1 | lastEig > oldDimension
            fprintf('Illegal number for the last eigenvalue.\n');
            areValuesOK = 0;
        end
        if firstEig < 1 | firstEig > lastEig

```

```

fprintf('Illegal number for the first eigenvalue.\n');
areValuesOK = 0;
end
end
% close the window
close(hndl_win);
end

%%%%%%%%%%%%%
% Interactive part - GUI
% this part is removed because the GUI is of no interest at present

%%%%%%%%%%%%%
% See if the user has reduced the dimension enough

if lastEig > maxLastEig
lastEig = maxLastEig;
if b_verbose
fprintf('Dimension reduced to %d due to the singularity of covariance matrix\n',...
    lastEig-firstEig+1);
end
else
% Reduce the dimensionality of the problem.
if b_verbose
if oldDimension == (lastEig - firstEig + 1)
fprintf ('Dimension not reduced.\n');
else
fprintf ('Reducing dimension...\n');
end
end
end
end

%%%%%%%%%%%%%
% Drop the smaller eigenvalues
if lastEig < oldDimension
lowerLimitValue = (eigenvalues(lastEig) + eigenvalues(lastEig + 1)) / 2;
else
lowerLimitValue = eigenvalues(oldDimension) - 1;
end

lowerColumns = diag(D) > lowerLimitValue;

%%%%%%%%%%%%%
% Drop the larger eigenvalues
if firstEig > 1
higherLimitValue = (eigenvalues(firstEig - 1) + eigenvalues(firstEig)) / 2;
else
higherLimitValue = eigenvalues(1) + 1;

```

```

end
higherColumns = diag(D) < higherLimitValue;

% Combine the results from above
selectedColumns = lowerColumns & higherColumns;

%%%%%%%%%%%%%%%
% print some info for the user
if b_verbose
    fprintf ('Selected [ %d ] dimensions.\n', sum (selectedColumns));
end
if sum (selectedColumns) ~= (lastEig - firstEig + 1),
    error ('Selected a wrong number of dimensions.');
end

if b_verbose
    fprintf ('Smallest remaining (non-zero) eigenvalue [ %g ]\n', eigenvalues(lastEig));
    fprintf ('Largest remaining (non-zero) eigenvalue [ %g ]\n', eigenvalues(firstEig));
    fprintf ('Sum of removed eigenvalues [ %g ]\n', sum(diag(D) .* ...
        (~selectedColumns)));
end

%%%%%%%%%%%%%%
% Select the columns which correspond to the desired range of eigenvalues.
E = selcol(E, selectedColumns);
D = selcol(selcol(D, selectedColumns)', selectedColumns);

%%%%%%%%%%%%%%
% Some more information
if b_verbose
    sumAll=sum(eigenvalues);
    sumUsed=sum(diag(D));
    retained = (sumUsed / sumAll) * 100;
    fprintf('[ %g ] % of (non-zero) eigenvalues retained.\n', retained);
end

%%%%%%%%%%%%%%
function newMatrix = selcol(oldMatrix, maskVector);

% newMatrix = selcol(oldMatrix, maskVector);
%
% Selects the columns of the matrix that marked by one in the given vector.
% The maskVector is a column vector.

% 15.3.1998
% based on my understanding, there is no need to write a function to extract specified

```

```
columns.  
%% maybe the authors have other considerations  
  
if size(maskVector, 1) ~= size(oldMatrix, 2),  
    error ('The mask vector and matrix are of incompatible size.');  
end  
  
numTaken = 0;  
  
for i = 1 : size (maskVector, 1),  
    if maskVector(i, 1) == 1,  
        takingMask(1, numTaken + 1) = i;  
        numTaken = numTaken + 1;  
    end  
end  
  
newMatrix = oldMatrix(:, takingMask);
```

#### 4. whitenv.m

```

function [newVectors, whiteningMatrix, dewhitenMatrix] = whitenv ...
    (vectors, E, D, s_verbose);
%WHITENV - Whitenv vectors.

%
% [newVectors, whiteningMatrix, dewhitenMatrix] = ...
%             whitenv(vectors, E, D, verbose);
%
% Whitens the data (row vectors) and reduces dimension. Returns
% the whitened vectors (row vectors), whitening and dewhitenning matrices.
%
% ARGUMENTS
%
% vectors      Data in row vectors.
% E            Eigenvector matrix from function 'pcamat'
% D            Diagonal eigenvalue matrix from function 'pcamat'
% verbose     Optional. Default is 'on'
%
% EXAMPLE
%     [E, D] = pcamat(vectors);
%     [nv, wm, dwm] = whitenv(vectors, E, D);
%
%
% This function is needed by FASTICA and FASTICAG
%
% See also PCAMAT

% @(#) $Id: whitenv.m,v 1.3 2003/10/12 09:04:43 jarmo Exp $

% =====
% Default value for 'verbose'
if nargin < 4, s_verbose = 'on'; end

% Check the optional parameter verbose;
switch lower(s_verbose)
case 'on'
    b_verbose = 1;
case 'off'
    b_verbose = 0;
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''verbose''\n', s_verbose));
end

% =====
% In some cases, rounding errors in Matlab cause negative
% eigenvalues (elements in the diagonal of D). Since it
% is difficult to know when this happens, it is difficult
% to correct it automatically. Therefore an error is

```

```

% signalled and the correction is left to the user.

if any (diag (D) < 0),
    error (sprintf (['[ %d ] negative eigenvalues computed from the' ...
        ' covariance matrix.\nThese are due to rounding' ...
        ' errors in Matlab (the correct eigenvalues are\n' ...
        ' probably very small).\nTo correct the situation,' ...
        ' please reduce the number of dimensions in the' ...
        ' data\nby using the ''lastEig'' argument in' ...
        ' function FASTICA, or ''Reduce dim.'' button\nin' ...
        ' the graphical user interface.'], ...
    sum (diag (D) < 0)));
end

% =====
% Calculate the whitening and dewhitening matrices (these handle dimensionality
simultaneously).

% note that the whitening matrix is different from the formula given in the Hyvärinen's paper,
but it is

% easy to check that the whitened data has an identity covariance matrix
whiteningMatrix = inv (sqrt (D)) * E' ;
dewhiteningMatrix = E * sqrt (D);

% Project to the eigenvectors of the covariance matrix.
% Whiten the samples and reduce dimension simultaneously.

if b_verbose, fprintf ('Whitening...\n'); end
newVectors = whiteningMatrix * vectors;

% =====
% Just some security...
if ~isreal(newVectors)
    error ('Whitened vectors have imaginary values.');
end

% Print some information to user
if b_verbose
    fprintf ('Check: covariance differs from identity by [ %g ].\n', ...
        max (abs (cov (newVectors', 1) - eye (size (newVectors, 1)))));
end

```

## 5. fpica.m

```

function [A, W] = fpica(X, whiteningMatrix, dewhitenMatrix, approach, ...
    numOfIC, g, finetune, a1, a2, myy, stabilization, ...
    epsilon, maxNumIterations, maxFinetune, initState, ...
    guess, sampleSize, displayMode, displayInterval, ...
    s_verbose);
%FPICA - Fixed point ICA. Main algorithm of FASTICA.
%
% [A, W] = fpica(whitesig, whiteningMatrix, dewhitenMatrix, approach,
%     numOfIC, g, finetune, a1, a2, mu, stabilization, epsilon,
%     maxNumIterations, maxFinetune, initState, guess, sampleSize,
%     displayMode, displayInterval, verbose);
%
% Perform independent component analysis using Hyvarinen's fixed point
% algorithm. Outputs an estimate of the mixing matrix A and its inverse W.
%
% whitesig           :the whitened data as row vectors
% whiteningMatrix    :can be obtained with function whitenv
% dewhitenMatrix     :can be obtained with function whitenv
% approach          :the approach used (deflation or symmetric)
% numOfIC           :number of independent components estimated
% g                 :the nonlinearity used
%                  ['pow3' | 'tanh' |
%                   'gaus' | 'skew']
% finetune          :the nonlinearity used in finetuning.
% a1                :parameter for tuning 'tanh'
% a2                :parameter for tuning 'gaus'
% mu               :step size in stabilized algorithm
% stabilization     :if mu < 1 then automatically on
% epsilon           :stopping criterion
% maxNumIterations   :maximum number of iterations
% maxFinetune       :maximum number of iteretions for finetuning
% initState         :initial guess or random initial state. See below
% guess             :initial guess for A. Ignored if initState = 'rand'
% sampleSize        :percentage of the samples used in one iteration
% displayMode       :plot running estimate
%                  ['signals' | 'basis' |
%                   'filters' | 'off']
% displayInterval   :number of iterations we take between plots
% verbose           :report progress in text format
%
% EXAMPLE
%      [E, D] = pcamat(vectors);
%      [nv, wm, dwm] = whitenv(vectors, E, D);
%      [A, W] = fpica(nv, wm, dwm);
%
%
% This function is needed by FASTICA and FASTICAG
%
% See also FASTICA, FASTICAG, WHITENV, PCAMAT

```

```
% @(#) $Id: fpica.m,v 1.7 2005/06/16 12:52:55 jarmo Exp $

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Global variable for stopping the ICA calculations from the GUI
global g_FastICA_interrupt;
if isempty(g_FastICA_interrupt)
    clear global g_FastICA_interrupt;
    interruptible = 0;
else
    interruptible = 1;
end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Default values

if nargin < 3, error('Not enough arguments!'); end
[vectorSize, numSamples] = size(X);
if nargin < 20, s_verbose = 'on'; end
if nargin < 19, displayInterval = 1; end
if nargin < 18, displayMode = 'on'; end
if nargin < 17, sampleSize = 1; end
if nargin < 16, guess = 1; end
if nargin < 15, initState = 'rand'; end
if nargin < 14, maxFinetune = 100; end
if nargin < 13, maxNumIterations = 1000; end
if nargin < 12, epsilon = 0.0001; end
if nargin < 11, stabilization = 'on'; end
if nargin < 10, myy = 1; end
if nargin < 9, a2 = 1; end
if nargin < 8, a1 = 1; end
if nargin < 7, finetune = 'off'; end
if nargin < 6, g = 'pow3'; end
if nargin < 5, numOfIC = vectorSize; end      % vectorSize = Dim
if nargin < 4, approach = 'defl'; end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Checking the data

if ~isreal(X)
    error('Input has an imaginary part.');
end

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Checking the value for verbose

switch lower(s_verbose)
case 'on'
```

```

b_verbose = 1;
case 'off'
b_verbose = 0;
otherwise
error(sprintf('Illegal value [ %s ] for parameter: ''verbose''\n', s_verbose));
end

%%%%%%%%%%%%%%%
% Checking the value for approach

switch lower(approach)
case 'symm'
approachMode = 1;
case 'defl'
approachMode = 2;
otherwise
error(sprintf('Illegal value [ %s ] for parameter: ''approach''\n', approach));
end
if b_verbose, fprintf('Used approach [ %s ].\n', approach); end

%%%%%%%%%%%%%%
% Checking the value for numOfIC

if vectorSize < numOfIC
error('Must have numOfIC <= Dimension! ');
end

%%%%%%%%%%%%%%
% Checking the sampleSize
if sampleSize > 1
sampleSize = 1;
if b_verbose
fprintf('Warning: Setting ''sampleSize'' to 1.\n');
end
elseif sampleSize < 1
% if the final sample size < 1000, change the proportion so that the final sample size is
1000 or use the
% full sample size
if (sampleSize * numSamples) < 1000
sampleSize = min(1000/numSamples, 1);
if b_verbose
fprintf('Warning: Setting ''sampleSize'' to %0.3f (%d samples).\n', ...
sampleSize, floor(sampleSize * numSamples));
end
end
end
if b_verbose
if b_verbose & (sampleSize < 1)

```

```

fprintf('Using about %0.0f%% of the samples in random order in every
step.\n',sampleSize*100);
end
end

%%%%%%%%%%%%%
% Checking the value for nonlinearity.

switch lower(g)
case 'pow3'
gOrig = 10;
case 'tanh'
gOrig = 20;
case {'gaus', 'gauss'}
gOrig = 30;
case 'skew'
gOrig = 40;
otherwise
error(sprintf('Illegal value [ %s ] for parameter: ''g''\n', g));
end
% if not use full sample size, the nonlinearity index + 2, e.g., 10 -> 12
if sampleSize ~= 1
gOrig = gOrig + 2;
end
% if use stabilization, the nonlinearity index + 1, e.g., 10 -> 11
if myy ~= 1
gOrig = gOrig + 1;
end

if b_verbose,
fprintf('Used nonlinearity [ %s ].\n', g);
end

finetuningEnabled = 1;
% for fine-tuning, the stabilized version was used
switch lower(finetune)
case 'pow3'
gFine = 10 + 1;
case 'tanh'
gFine = 20 + 1;
case {'gaus', 'gauss'}
gFine = 30 + 1;
case 'skew'
gFine = 40 + 1;
case 'off'
% if myy ~=1, it means the nonlinearity index is already +1
if myy ~= 1
gFine = gOrig;

```

```

else
    gFine = gOrig + 1;
end
finetuningEnabled = 0;
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''finetune''\n', ...
        finetune));
end

if b_verbose & finetuningEnabled
    fprintf('Finetuning enabled (nonlinearity: [ %s ]).\\n', finetune);
end

switch lower(stabilization)
case 'on'
    stabilizationEnabled = 1;
case 'off'
    if myy ~= 1
        stabilizationEnabled = 1;
    else
        stabilizationEnabled = 0;
    end
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''stabilization''\n', ...
        stabilization));
end

if b_verbose & stabilizationEnabled
    fprintf('Using stabilized algorithm.\\n');
end

%%%%%%%%%%%%%
% Some other parameters
myyOrig = myy;
% When we start fine-tuning we'll set myy = myyK * myy
myyK = 0.01;
% How many times do we try for convergence until we give up.
failureLimit = 5;
% these codes seem redundant, because they are initialized again below
usedNlinearity = gOrig;
stroke = 0;
notFine = 1;
long = 0;

%%%%%%%%%%%%%
% Checking the value for initial state.

switch lower(initState)

```

```

case 'rand'
    initialStateMode = 0;
case 'guess'
    if size(guess,1) ~= size(whiteningMatrix,2)
        initialStateMode = 0;
        if b_verbose
            fprintf('Warning: size of initial guess is incorrect. Using random initial guess.\n');
        end
    else
        initialStateMode = 1;
        if size(guess,2) < numOfIC
            if b_verbose
                fprintf('Warning: initial guess only for first %d components. Using random initial guess
for others.\n', size(guess,2));
            end
            guess(:, size(guess, 2) + 1:numOfIC) = ...
                rand(vectorSize,numOfIC-size(guess,2))- .5;
        elseif size(guess,2)>numOfIC
            guess=guess(:,1:numOfIC);
            fprintf('Warning: Initial guess too large. The excess column are dropped.\n');
        end
        if b_verbose, fprintf('Using initial guess.\n'); end
    end
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''initState''\n', initState));
end

%%%%%%%%%%%%%
% Checking the value for display mode.
% mode 1/2/3 plot the sources, mixing matrix and unmixing matrix separately
switch lower(displayMode)
case {'off', 'none'}
    usedDisplay = 0;
case {'on', 'signals'}
    usedDisplay = 1;
    if (b_verbose & (numSamples > 10000))
        fprintf('Warning: Data vectors are very long. Plotting may take long time.\n');
    end
    if (b_verbose & (numOfIC > 25))
        fprintf('Warning: There are too many signals to plot. Plot may not look good.\n');
    end
case 'basis'
    usedDisplay = 2;
    if (b_verbose & (numOfIC > 25))
        fprintf('Warning: There are too many signals to plot. Plot may not look good.\n');
    end
case 'filters'
    usedDisplay = 3;

```

```

if (b_verbose & (vectorSize > 25))
    fprintf('Warning: There are too many signals to plot. Plot may not look good.\n');
end
otherwise
    error(sprintf('Illegal value [ %s ] for parameter: ''displayMode''\n', displayMode));
end

% The displayInterval can't be less than 1...
if displayInterval < 1
    displayInterval = 1;
end

%%%%%%%%%%%%%
if b_verbose, fprintf('Starting ICA calculation...\n'); end

%%%%%%%%%%%%%
% SYMMETRIC APPROACH
if approachMode == 1,

    % set some parameters more...
    usedNlinearity = gOrig;
    stroke = 0;
    notFine = 1;
    long = 0;
    %% vectorSize means the number of rows of whitened data X
    A = zeros(vectorSize, numOfIC); % Dewhitened basis vectors.
    if initialStateMode == 0
        % Take random orthonormal initial vectors.
        B = orth (randn (vectorSize, numOfIC));
    elseif initialStateMode == 1
        % Use the given initial vector as the initial state
        %% M=AS, whiteningMatrix * M = whiteningMatrix * AS, let B=whiteningMatrix * A, M means
        unwhitened data
        %% whiteningMatrix * M = BS, S=WM, let W=B-1 * whiteningMatrix, B-1=BT, as B is an
        orthogonal matrix
        B = whiteningMatrix * guess;
    end
    %% Bold and Bold2 store the estimates at n-1th and n-2th iterations, which could be used to
    detect stroke.
    Bold = zeros(size(B));
    Bold2 = zeros(size(B));

    % This is the actual fixed-point iteration loop.
    for round = 1:maxNumIterations + 1,
        %% if the maximum iteration is reached, exit the loop
        if round == maxNumIterations + 1,
            fprintf('No convergence after %d steps\n', maxNumIterations);
            fprintf('Note that the plots are probably wrong.\n');

```

```

% if the final estimate is not empty, use it as output
if ~isempty(B)
% Symmetric orthogonalization.
B = B * real(inv(B' * B)^(1/2));

W = B' * whiteningMatrix;
A = dwhiteningMatrix * B;
else
W = [];
A = [];
end
return;
end

if (interruptible & g_FastICA_interrupt)
if b_verbose
    fprintf('\n\nCalculation interrupted by the user\n');
end
if ~isempty(B)
W = B' * whiteningMatrix;
A = dwhiteningMatrix * B;
else
W = [];
A = [];
end
return;
end

% Symmetric orthogonalization.
%% Formula (45) in Hyvarinen's paper
B = B * real(inv(B' * B)^(1/2));

% Test for termination condition. Note that we consider opposite
% directions here as well.
%% the inner product of two vectors with unit length is the cosine of the angle between the
two vectors
%% the inner product is equal to 1/-1 when two vectors are parallel
%% use the absolute minimum of vector pairs to represent the cost
minAbsCos = min(abs(diag(B' * Bold)));
minAbsCos2 = min(abs(diag(B' * Bold2)));

if (1 - minAbsCos < epsilon)
    %% fine-tuning means using stabilized updating rule on the basis of the current estimate
    %% fine-tuning is only considered after initial convergence is reached
    %% if fine-tuning is enabled and fine-tuning is not started
    if finetuningEnabled & notFine
        if b_verbose, fprintf('Initial convergence, fine-tuning: \n'); end;
        notFine = 0;
    end
end

```

```

usedNlinearity = gFine;
myy = myyK * myyOrig;
B0ld = zeros(size(B));
B0ld2 = zeros(size(B));

else
    if b_verbose, fprintf('Convergence after %d steps\n', round); end

    % Calculate the de-whitened vectors.
    A = dewhiteingMatrix * B;
    break;
end
elseif stabilizationEnabled
    % if convergence is not achieved, consider stabilization
    % stroke is defined that the round-th and (round-2)-th estimates are parallel
    % if stroke is detected, set the myy to the half momentarily (see below)
    if (~stroke) & (1 - minAbsCos2 < epsilon)
        if b_verbose, fprintf('Stroke!\n'); end;
        stroke = myy;
        myy = .5*myy;
        % for stabilization, use stabilized updating rule (indexed by odd numbers)
        if mod(usedNlinearity,2) == 0
            usedNlinearity = usedNlinearity + 1;
        end
        elseif stroke
            %% if stroke is detected in the last iteration, restore the myy value (which is saved in
            %% stroke)
            %% also reset the nonlinearity into the original one if myy == 1
            myy = stroke;
            stroke = 0;
            if (myy == 1) & (mod(usedNlinearity,2) ~= 0)
                usedNlinearity = usedNlinearity - 1;
            end
            %% if iterations > 500, set the myy into the half in the rest of iterations
            %% if no stroke, the stabilized version will only be enabled after the half of max
            iterations
            elseif (~long) & (round>maxNumIterations/2)
                if b_verbose, fprintf('Taking long (reducing step size)\n'); end;
                long = 1;
                myy = .5*myy;
                if mod(usedNlinearity,2) == 0
                    usedNlinearity = usedNlinearity + 1;
                end
                end
            end
        Bold2 = Bold;
        Bold = B;

```

```

%%%%%%%%%%%%%%%
% Show the progress...
if b_verbose
    if round == 1
        fprintf('Step no. %d\n', round);
    else
        fprintf('Step no. %d, change in value of estimate: %.3g \n', round, 1 - minAbsCos);
    end
end

%%%%%%%%%%%%%%%
% Also plot the current state...
switch usedDisplay
    case 1
        if rem(round, displayInterval) == 0,
% There was and may still be other displaymodes...
% 1D signals
% plot the estimated components or sources
icaplot('dispsig',(X'*B)');
drawnow;
    end
    case 2
        if rem(round, displayInterval) == 0,
% ... and now there are :-)
% 1D basis
% plot the mixing matrix
A = dwhiteningMatrix * B;
icaplot('dispsig',A');
drawnow;
    end
    case 3
        if rem(round, displayInterval) == 0,
% ... and now there are :-)
% 1D filters
% plot the unmixing matrix
W = B' * whiteningMatrix;
icaplot('dispsig',W);
drawnow;
    end
otherwise
end

switch usedNonlinearity
    % pow3
    %% 10 means the original fixed iteration method;
    %% 11 means the stabilized version, Formula (47) in Hyvarinen's paper
    %% 12 means using part of the sample size in each iteration

```

```

%% 13 combines 11 and 12

case 10
B = (X * (( X' * B) .^ 3)) / numSamples - 3 * B;
case 11
Y = X' * B;
Gpow3 = Y .^ 3;
Beta = sum(Y .* Gpow3);
D = diag(1 ./ (Beta - 3 * numSamples));
B = B + myy * B * (Y' * Gpow3 - diag(Beta)) * D;
case 12
Xsub=X(:, getSamples(numSamples, sampleSize));
B = (Xsub * (( Xsub' * B) .^ 3)) / size(Xsub,2) - 3 * B;
case 13
Ysub=X(:, getSamples(numSamples, sampleSize))' * B;
Gpow3 = Ysub .^ 3;
Beta = sum(Ysub .* Gpow3);
D = diag(1 ./ (Beta - 3 * size(Ysub', 2)));
B = B + myy * B * (Ysub' * Gpow3 - diag(Beta)) * D;

% tanh

case 20
hypTan = tanh(a1 * X' * B);
B = X * hypTan / numSamples - ...
ones(size(B,1),1) * sum(1 - hypTan .^ 2) .* B / numSamples * ...
a1;
case 21
Y = X' * B;
hypTan = tanh(a1 * Y);
Beta = sum(Y .* hypTan);
D = diag(1 ./ (Beta - a1 * sum(1 - hypTan .^ 2)));
B = B + myy * B * (Y' * hypTan - diag(Beta)) * D;
case 22
Xsub=X(:, getSamples(numSamples, sampleSize));
hypTan = tanh(a1 * Xsub' * B);
B = Xsub * hypTan / size(Xsub, 2) - ...
ones(size(B,1),1) * sum(1 - hypTan .^ 2) .* B / size(Xsub, 2) * a1;
case 23
Y = X(:, getSamples(numSamples, sampleSize))' * B;
hypTan = tanh(a1 * Y);
Beta = sum(Y .* hypTan);
D = diag(1 ./ (Beta - a1 * sum(1 - hypTan .^ 2)));
B = B + myy * B * (Y' * hypTan - diag(Beta)) * D;

% gauss

case 30
U = X' * B;
Usquared=U .^ 2;
ex = exp(-a2 * Usquared / 2);

```

```

gauss = U .* ex;
dGauss = (1 - a2 * Usquared) .*ex;
B = X * gauss / numSamples - ...
ones(size(B,1),1) * sum(dGauss)... .
.* B / numSamples ;
case 31
Y = X' * B;
ex = exp(-a2 * (Y .^ 2) / 2);
gauss = Y .* ex;
Beta = sum(Y .* gauss);
D = diag(1 ./ (Beta - sum((1 - a2 * (Y .^ 2)) .* ex)));
B = B + myy * B * (Y' * gauss - diag(Beta)) * D;
case 32
Xsub=X(:, getSamples(numSamples, sampleSize));
U = Xsub' * B;
Usquared=U .^ 2;
ex = exp(-a2 * Usquared / 2);
gauss = U .* ex;
dGauss = (1 - a2 * Usquared) .*ex;
B = Xsub * gauss / size(Xsub,2) - ...
ones(size(B,1),1) * sum(dGauss)... .
.* B / size(Xsub,2) ;
case 33
Y = X(:, getSamples(numSamples, sampleSize))' * B;
ex = exp(-a2 * (Y .^ 2) / 2);
gauss = Y .* ex;
Beta = sum(Y .* gauss);
D = diag(1 ./ (Beta - sum((1 - a2 * (Y .^ 2)) .* ex)));
B = B + myy * B * (Y' * gauss - diag(Beta)) * D;

% skew
case 40
B = (X * ((X' * B) .^ 2)) / numSamples;
case 41
Y = X' * B;
Gskew = Y .^ 2;
Beta = sum(Y .* Gskew);
D = diag(1 ./ (Beta));
B = B + myy * B * (Y' * Gskew - diag(Beta)) * D;
case 42
Xsub=X(:, getSamples(numSamples, sampleSize));
B = (Xsub * ((Xsub' * B) .^ 2)) / size(Xsub,2);
case 43
Y = X(:, getSamples(numSamples, sampleSize))' * B;
Gskew = Y .^ 2;
Beta = sum(Y .* Gskew);
D = diag(1 ./ (Beta));
B = B + myy * B * (Y' * Gskew - diag(Beta)) * D;

```

```

otherwise
    error('Code for desired nonlinearity not found!');
end
end

% Calculate ICA filters.
W = B' * whiteningMatrix;

%%%%%%%%%%%%%%%
% Also plot the last one...
switch usedDisplay
case 1
    % There was and may still be other displaymodes...
    % 1D signals
    icaplot('dispsig',(X'*B)');
    drawnow;
case 2
    % ... and now there are :-)
    % 1D basis
    icaplot('dispsig',A');
    drawnow;
case 3
    % ... and now there are :-)
    % 1D filters
    icaplot('dispsig',W);
    drawnow;
otherwise
end
end

%%%%%%%%%%%%%%%
% DEFLATION APPROACH
if approachMode == 2
    % vectorSize means the number of rows
    B = zeros(vectorSize);

    % The search for a basis vector is repeated numOfIC times.
    % round means the round-th IC
    round = 1;

    numFailures = 0;

    while round <= numOfIC,
        % initialize parameters when estimate one of ICs
        % myy is used for stabilized version
        % notFine means the fine-tuning is not started
        % long means whether the iterations are over 500 times

```

```

%% endFinetuning means the limit of iterations (see below)
myy = myyOrig;
usedNlinearity = g0rig;
stroke = 0;
notFine = 1;
long = 0;
endFinetuning = 0;

%%%%%%%%%%%%%
% Show the progress...
if b_verbose, fprintf('IC %d ', round); end

% Take a random initial vector of lenght 1 and orthogonalize it
% with respect to the other vectors.
%% w is one column of B, don't mix it with W, W=B' * whiteningMatrix
if initialStateMode == 0
    w = randn (vectorSize, 1);
elseif initialStateMode == 1
    w=whiteningMatrix*guess(:,round);
end
% orthogonalize by subtracting the projection of w into B. Formula (44) in Hyvarinen's
paper
w = w - B * B' * w;
w = w / norm(w);
%% wOld2 was used to detect stroke
wOld = zeros(size(w));
wOld2 = zeros(size(w));

% This is the actual fixed-point iteration loop.
% for i = 1 : maxNumIterations + 1
i = 1;
%% gabba is used to specify the number of fine-tuning iterations
%% this parameter is not available in symmetric estimation
gabba = 1;
while i <= maxNumIterations + gabba
    if (usedDisplay > 0)
        drawnow;
    end
    if (interruptible & g_FastICA_interrupt)
        if b_verbose
            fprintf('\n\nCalculation interrupted by the user\n');
        end
        return;
    end

    % Project the vector into the space orthogonal to the space
    % spanned by the earlier found basis vectors. Note that we can do
    % the projection with matrix B, since the zero entries do not

```

```

% contribute to the projection.

w = w - B * B' * w;
w = w / norm(w);

%% if fine-tuning is not started
%% fine-tuning would be only be invoked when initial convergence is reached
if notFine
if i == maxNumIterations + 1
if b_verbose
    fprintf('\nComponent number %d did not converge in %d iterations.\n', round,
maxNumIterations);
end
%% if the round-th estimation not converged, repeat the estimation until the failureLimit
is reached
    %% failureLimit is not available in symmetric approach
    %% round - 1 means the round-th IC would be re-estimated after non-convergence
    round = round - 1;
    numFailures = numFailures + 1;
    if numFailures > failureLimit
        if b_verbose
            fprintf('Too many failures to converge (%d). Giving up.\n', numFailures);
        end
        if round == 0
            A=[];
            W=[];
        end
        return;
    end
    if round == 0
        A=[];
        W=[];
    end
    return;
end
% numFailures > failurelimit
break;
end
% i == maxNumIterations + 1
else
% if notFine
%% if fine-tuning is invoked and the MaxFinetune is reached, stop fine-tuning
%% endFinetuning is equal to (initial convergence iterations + MaxFinetune)
if i >= endFinetuning
    wOld = w; % So the algorithm will stop on the next test...
end
end
% if notFine

%%%%%%%%%%%%%
% Show the progress...
if b_verbose, fprintf('.'); end;

% Test for termination condition. Note that the algorithm has
% converged if the direction of w and wOld is the same, this

```

```

% is why we test the two cases.
if norm(w - wOld) < epsilon | norm(w + wOld) < epsilon
    if finetuningEnabled & notFine
        if b_verbose, fprintf('Initial convergence, fine-tuning: '); end;
        notFine = 0;
    gabba = maxFinetune;
    wOld = zeros(size(w));
    wOld2 = zeros(size(w));
    usedNlinearity = gFine;
    myy = myyK * myyOrig;

endFinetuning = maxFinetune + i;

else % fine-tuning not enabled or fine-tuning has already been done
    %% if the round-th IC is converged, reset the number of Failures
    numFailures = 0;
    % Save the vector
    B(:, round) = w;

    % Calculate the de-whitened vector.
    A(:, round) = dwhiteningMatrix * w;
    % Calculate ICA filter.
    W(round, :) = w' * whiteningMatrix;

%%%%%%%%%%%%%
% Show the progress...
if b_verbose, fprintf('computed ( %d steps ) \n', i); end

%%%%%%%%%%%%%
% Also plot the current state...
switch usedDisplay
case 1
    if rem(round, displayInterval) == 0,
        % There was and may still be other displaymodes...
        % 1D signals
        temp = X'*B;
        icaplot('dispsig',temp(:,1:numOfIC));
        drawnow;
    end
case 2
    if rem(round, displayInterval) == 0,
        % ... and now there are :-)
        % 1D basis
        icaplot('dispsig',A');
        drawnow;
    end
case 3
    if rem(round, displayInterval) == 0,

```

```

% ... and now there are :-)
% 1D filters
icaplot('dispsig',W);
drawnow;
end
end
% switch usedDisplay
break; % IC ready - next...
end
%if finetuningEnabled & notFine
elseif stabilizationEnabled
if (~stroke) & (norm(w - wOld2) < epsilon | norm(w + wOld2) < ...
    epsilon)
stroke = myy;
if b_verbose, fprintf('Stroke!'); end;
myy = .5*myy;
if mod(usedNlinearity,2) == 0
    usedNlinearity = usedNlinearity + 1;
end
elseif stroke
myy = stroke;
stroke = 0;
if (myy == 1) & (mod(usedNlinearity,2) ~= 0)
    usedNlinearity = usedNlinearity - 1;
end
elseif (notFine) & (~long) & (i > maxNumIterations / 2)
if b_verbose, fprintf('Taking long (reducing step size) '); end;
long = 1;
myy = .5*myy;
if mod(usedNlinearity,2) == 0
    usedNlinearity = usedNlinearity + 1;
end
end
end
end

wOld2 = wOld;
wOld = w;

switch usedNlinearity
% pow3
case 10
w = (X * ((X' * w) .^ 3)) / numSamples - 3 * w;
case 11
EXGpow3 = (X * ((X' * w) .^ 3)) / numSamples;
Beta = w' * EXGpow3;
w = w - myy * (EXGpow3 - Beta * w) / (3 - Beta);
case 12
Xsub=X(:,getSamples(numSamples, sampleSize));

```

```

w = (Xsub * ((Xsub' * w) .^ 3)) / size(Xsub, 2) - 3 * w;
case 13
Xsub=X(:,getSamples(numSamples, sampleSize));
EXGpow3 = (Xsub * ((Xsub' * w) .^ 3)) / size(Xsub, 2);
Beta = w' * EXGpow3;
w = w - myy * (EXGpow3 - Beta * w) / (3 - Beta);
% tanh
case 20
hypTan = tanh(a1 * X' * w);
w = (X * hypTan - a1 * sum(1 - hypTan .^ 2)' * w) / numSamples;
case 21
hypTan = tanh(a1 * X' * w);
Beta = w' * X * hypTan;
w = w - myy * ((X * hypTan - Beta * w) / ...
    (a1 * sum((1-hypTan .^2)') - Beta));
case 22
Xsub=X(:,getSamples(numSamples, sampleSize));
hypTan = tanh(a1 * Xsub' * w);
w = (Xsub * hypTan - a1 * sum(1 - hypTan .^ 2)' * w) / size(Xsub, 2);
case 23
Xsub=X(:,getSamples(numSamples, sampleSize));
hypTan = tanh(a1 * Xsub' * w);
Beta = w' * Xsub * hypTan;
w = w - myy * ((Xsub * hypTan - Beta * w) / ...
    (a1 * sum((1-hypTan .^2)') - Beta));
% gauss
case 30
% This has been split for performance reasons.
u = X' * w;
u2=u.^2;
ex=exp(-a2 * u2/2);
gauss = u.*ex;
dGauss = (1 - a2 * u2) .*ex;
w = (X * gauss - sum(dGauss)' * w) / numSamples;
case 31
u = X' * w;
u2=u.^2;
ex=exp(-a2 * u2/2);
gauss = u.*ex;
dGauss = (1 - a2 * u2) .*ex;
Beta = w' * X * gauss;
w = w - myy * ((X * gauss - Beta * w) / ...
    (sum(dGauss)' - Beta));
case 32
Xsub=X(:,getSamples(numSamples, sampleSize));
u = Xsub' * w;
u2=u.^2;
ex=exp(-a2 * u2/2);

```

```

gauss = u.*ex;
dGauss = (1 - a2 * u2) .*ex;
w = (Xsub * gauss - sum(dGauss)' * w) / size(Xsub, 2);
case 33
Xsub=X(:,getSamples(numSamples, sampleSize));
u = Xsub' * w;
u2=u.^2;
ex=exp(-a2 * u2/2);
gauss = u.*ex;
dGauss = (1 - a2 * u2) .*ex;
Beta = w' * Xsub * gauss;
w = w - myy * ((Xsub * gauss - Beta * w) / ...
(sum(dGauss)' - Beta));
% skew
case 40
w = (X * ((X' * w) .^ 2)) / numSamples;
case 41
EXGskew = (X * ((X' * w) .^ 2)) / numSamples;
Beta = w' * EXGskew;
w = w - myy * (EXGskew - Beta*w)/( -Beta);
case 42
Xsub=X(:,getSamples(numSamples, sampleSize));
w = (Xsub * ((Xsub' * w) .^ 2)) / size(Xsub, 2);
case 43
Xsub=X(:,getSamples(numSamples, sampleSize));
EXGskew = (Xsub * ((Xsub' * w) .^ 2)) / size(Xsub, 2);
Beta = w' * EXGskew;
w = w - myy * (EXGskew - Beta*w)/( -Beta);

otherwise
error('Code for desired nonlinearity not found!');
end

% Normalize the new w.
w = w / norm(w);
i = i + 1;
end
round = round + 1;
end
if b_verbose, fprintf('Done.\n'); end

%%%%%%%%%%%%%
% Also plot the ones that may not have been plotted.
if (usedDisplay > 0) & (rem(round-1, displayInterval) ~= 0)
switch usedDisplay
case 1
% There was and may still be other displaymodes...
% 1D signals

```

```

temp = X'*B;
icaplot('dispsig',temp(:,1:numOfIC));
drawnow;
case 2
% ... and now there are :-
% 1D basis
icaplot('dispsig',A');
drawnow;
case 3
% ... and now there are :-
% 1D filters
icaplot('dispsig',W);
drawnow;
otherwise
end
end
end

%%%%%%%%%%%%%%%
% In the end let's check the data for some security
if ~isreal(A)
if b_verbose, fprintf('Warning: removing the imaginary part from the result.\n'); end
A = real(A);
W = real(W);
end

%%%%%%%%%%%%%%
%%%%%%%%%%%%%%
%%%%%%%%%%%%%%
% Subfunction
% Calculates tanh simplier and faster than Matlab tanh.
function y=tanh(x)
y = 1 - 2 ./ (exp(2 * x) + 1);

%%%%%%%%%%%%%%
%% max means the total sample size, percentage means the proportion of sample size
%% rand generates `max` random numbers following uniform distribution at the interval of (0,1)
%% the function returns the position or index of selected samples
function Samples = getSamples(max, percentage)
Samples = find(rand(1, max) < percentage);

```