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IC.m

[illegible]

```

16 E    = var.e_ols;
17 N    = size(S,1);
18 llf = - (T * N / 2) * (1 + log(2 * pi)) - T / 2 * log(det(S));
19 llf = llf - 1 / 2 * trace( iS * E' * E);
20
21 AIC = - 2 * llf / T + 2 * K / T;
22 % SIC = - 2 * llf / T + K * log(T) / T;
23 HQIC = - 2 * llf / T + 2 * K * log(log(T)) / T;
24 BIC = - 2 * llf / T + K * log(T) / T;

```

OrderingIndexes.m

```

1  function [sindex] = OrderingIndexes(varordering,varnames,newstrng)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'OrderingIndexes' find the specified order of variables
5
6  % input:
7  % - varnames      = the list (names) of all the variables in the database
8  % - varordering   = the names and the order of the variables in the VAR
9
10 % output:
11 % - sindex = the index in the varnames that correspond to the var ordering
12 % - index_XX = the index that the variable XXX has in the VAR
13
14 % Filippo Ferroni ,
15 % Revised , 3/21/2018
16 % Revised , 9/11/2019
17 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
18 if nargin<3
19     newstrng='';
20 end
21
22 [~,sindex] = ismember(varordering,varnames);
23 for ii = 1 : size(varordering,2)
24     check=0;
25     for jjj = 1: size(varnames,2)
26         if strcmp(deblank(varnames{jjj}),deblank(varordering{ii})) == 1,
27             check = 1;
28         end
29     end
30     if check == 1
31         assignin('base',[newstrng 'index_' varordering{ii} ], ii);
32     else
33         warning(['I did not find ' varordering{ii} ' in varnames'])

```

```

34     end
35 end

```

YXB_.m

```

1  function [YYact,XXact] = YXB_(YY,lags,constant_timetrend)
2  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
3  % 'YXB_' organizes the data in the form
4  % of  $Y = XB+E$ 
5  % NO dummy observations
6
7  % Filippo Ferroni, 6/1/2015
8  % Revised, 2/15/2017
9  % Revised, 3/21/2018
10 % Revised, 9/11/2019
11 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
12 if nargin<3
13     constant = 1;
14     timetrend = 0;
15 else
16     constant = constant_timetrend(1);
17     timetrend = constant_timetrend(2);
18 end
19
20 nlags_ = lags;           % number of lags */
21 T0 = lags;               % size of pre-sample */
22
23 nv = size(YY,2);         %* number of variables */
24 nobs = size(YY,1)-T0;    %* number of observations */
25
26 % Actual observations
27
28 YYact = YY(T0+1:T0+nobs,:);
29 XXact = zeros(nobs,nv*nlags_);
30 i = 1;
31
32 while (i <= nlags_)
33     XXact(:,(i-1)*nv+1:i*nv) = YY(T0-(i-1):T0+nobs-i,:);
34     i = i+1;
35 end
36
37 if constant
38     % last column of XXact = constant
39     XXact = [XXact ones(nobs,1)];
40 end

```

```

41
42 if timetrend
43     % last column of XXact = constant
44     XXact = [XXact (1:nobs)'];
45 end

```

blp_ml.m

```

1  % function [log_dnsty] = blp_ml(shrinkage, hh, prior, olsreg_, F, G, Fo,
    positions_nylags, position_constant)
2  function [log_dnsty] = blp_ml(shrinkage, hh, prior, olsreg_, F, G, Fo,
    positions_nylags, position_constant)
3
4  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
5  % 'blp_ml' computes the marginal likelihood for the NMIW LP
6
7  % Inputs:
8  % — hyperpara, shrinkage hyperpara over which maximize the marginal
9  % likelihood
10
11 % Output: marginal data density
12
13 % Filippo Ferroni, 3/21/2020
14 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
15
16 %*****
17 % SETTINGS
18 %*****
19 ny = size(G,2);
20 lags = size(G,2)/ny;
21 %*****
22 % Conjugate Prior: MN-IW for LP
23 %*****
24 [posterior1, prior1] = p2p(hh, shrinkage, prior, olsreg_, F, G, Fo, positions_nylags,
    position_constant);
25
26 %*****
27 %* Compute the log marginal data density for the VAR model with MNIW
28 %*****
29 var.y = olsreg_.Y;
30 var.X = olsreg_.X(:, [positions_nylags, position_constant]);
31 prior1.Sigma.df = prior1.df;
32 prior1.Sigma.scale = prior1.S;
33 prior1.Phi.mean = prior1.BetaMean;
34 prior1.Phi.cov = prior1.BetaVar;

```

```

35
36 log_dnsty      = mniw_log_dnsty(prior1,posterior1,var);

```

blp_opt_hyperpara.m

```

1  function minus_log_dnsty = blp_opt_hyperpara(hyperpara,hh,prior,olsreg_,F,G,Fo
    ,positions_nylags,position_constant)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'blp_opt_hyperpara' computes the marginal likelihood over the
5  % hyperparameters of for the LP
6
7  % Inputs:
8  % — hyperpara, prior shrinkage
9  % likelihood
10
11 % Output: marginal data density
12
13 % Filippo Ferroni, 3/21/2020
14 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
15
16 hyperpara      = exp(hyperpara);
17 log_dnsty      = blp_ml(hyperpara,hh,prior,olsreg_,F,G,Fo,positions_nylags,
    position_constant);
18 minus_log_dnsty = -log_dnsty;

```

bvar.m

```

1  function [BVAR] = bvar(y,lags,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
5  % 'bvar' generates draws from the paramters of VAR model with error
6  % distributed as a multivariate normal
7
8  % Core Inputs:
9  % — y, data columns variables
10 % — lags, lag order of the VAR
11
12 % Additonal Inputs collected options:
13 % — options are not mandatory. if nothing is specified then a Jeffrey prior is
14 % assumed
15 % — options.K, number of draws from the posterior distribution

```

```

16 % — options.hor, horizon to compute the impulse response
17 % — options.fhor, horizon to compute the out-of-sample forecast
18 % — options.priors, a string with the priors for the autoregressive parameters
    and
19 %   for the scaling matrix.
20 % (...) see below
21
22 % Output: Draws from the conditional distribution of Phi, Sigma and
23 % Omega, impulse response with the cholesky decomposition and long run
24 % restrictions, forecast and marginal likelihood.
25
26 % Filippo Ferroni, 6/1/2015
27 % Revised, 2/15/2017
28 % Revised, 3/21/2018
29 % Revised, 9/11/2019
30 % Revised, 27/02/2020
31 % Revised, 27/04/2020
32 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
33 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
34
35 if nargin < 2
36     error('the_BVAR_toolbox_needs_at_least_two_inputs:_data_and_number_of_lags
        ');
37 end
38 if lags < 1
39     error('lags_cannot_be_zero_or_negative');
40 end
41 % number of observable variables
42 ny = size(y, 2);
43
44 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
45 %* DEFAULT SETTINGS
46 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
47 % Control random number generator
48 if isOctave == 0
49     isMatlab = 1;
50     rng('default');
51     rng(999);
52 else
53     isMatlab = 0;
54     % pkg load optim
55     randn('state',999);
56     rand('state',999);
57 end
58
59
60 % Default Settings (they can all be changed in 'options' see below)

```



```

61 K = 5000; % number of draws from the posterior
62 hor = 24; % horizon for the IRF
63 fhor = 12; % horizon for the forecasts
64 firstobs = lags+1; % first observation
65 presample = 0; % using a presample for setting the hyper-
    parameter of the Minnesota prior
66 noconstant = 0; % when 0, includes a constant in the VAR
67 timetrend = 0; % when 1, includes a time trend in the VAR
68 minn_prior_tau = 3; % Minnesota prior Hyper-Param: Overall
    Tightness
69 minn_prior_decay = 0.5; % Minnesota prior Hyper-Param: Tightness on
    lags > 1
70 minn_prior_lambda = 5; % Minnesota prior Hyper-Param: Sum-of-
    Coefficient
71 minn_prior_mu = 2; % Minnesota prior Hyper-Param: Co-
    Persistence
72 minn_prior_omega = 2; % Minnesota prior Hyper-Param: Shocks
    Variance
73 long_run_irf = 0; % when 0, it does not compute long run IRF
74 irf_1STD = 1; % when 1, IRF are computed as 1SD increase
    . Else, IRF are computed as unitary increase in the shock
75 cfrfst_yes = 0; % no conditional forecast unless defined
    in options
76
77 signs_irf = 0;
78 narrative_signs_irf = 0;
79 zeros_signs_irf = 0;
80 proxy_irf = 0;
81 noprint = 0;
82 nexogenous = 0;
83 exogenous = [];
84
85 % for mixed frequency / irregularly sampled data.
86 % Interpolate the missing values of each times series.
87 if any(any(isnan(y)))==1
88     warning('Activating the Mixed-Frequency BVAR')
89
90     mixed_freq_on = 1;
91     index_nan_var = find(sum(isnan(y),1) ~= 0);
92     yoriginal = y;
93     % interpolate the variables that have nans
94     T = 1:length(y);
95     for kk = 1 : length(index_nan_var)
96         v = y(isfinite(y(:,index_nan_var(kk))),
            index_nan_var(kk));
97         x = find(isfinite(y(:,index_nan_var(kk))));
98         if isMatlab == 1

```

```

99         y(:,index_nan_var(kk)) = interp1(x,v,T','spline');
100     else
101         y(:,index_nan_var(kk)) = spline(x,v,T');
102     end
103     yinterpol = y;
104 end
105 else
106     mixed_freq_on = 0;
107 end
108 if mixed_freq_on == 1 && nargin < 3
109     warning(['You did not specified the aggregation of the mixed_freq...'
110         'I will treat them as stocks.']);
111     index = zeros(size(y,2),1);
112 end
113
114 % Priors declaration: default Jeffrey prior
115 dummy = 0;
116 flat = 1;
117 priors = priors_( );
118
119 % declaring the names for the observable variables
120 for v = 1 : ny
121     eval(['varnames{', num2str(v) '} = ''Var', num2str(v) ''';'])
122 end
123
124
125 %*****
126 %* CUSTOMIZED SETTINGS
127 %*****
128 if nargin > 2
129     if isfield(options,'vnames')==1
130         varnames = options.vnames;
131     end
132     %=====
133     % Inference options
134     %=====
135     if isfield(options,'K')==1
136         K = options.K;
137     end
138     if isfield(options,'firstobs')==1
139         firstobs = options.firstobs;
140         if firstobs < lags + 1
141             error('firstobs need to be larger than lags+1')
142         end
143     end
144     if isfield(options,'presample')==1
145         presample = options.presample;

```

```

146     end
147     if isfield(options,'noconstant')==1
148         noconstant = options.nocostant;
149     end
150     if isfield(options,'timetrend')==1
151         timetrend = options.timetrend;
152         noconstant = 0;
153     end
154     %=====
155     % Exogenous Variables options
156     %=====
157     if isfield(options,'exogenous')==1
158         exogenous = options.exogenous;
159         nexogenous = size(exogenous,2);
160         if size(exogenous,1) ~= size(y,1)+ fhor && size(exogenous,1) ~= size(y
            ,1)
161             error('Size_Mismatch_between_endogenous_and_exogenous_variables;_
                exo_must_be_either_T_or_T+fhor');
162         end
163     end
164     %=====
165     % Minnesota prior options
166     %=====
167     if (isfield(options,'priors')==1 && strcmp(options.priors.name,'Minnesota'
        )==1) || (isfield(options,'priors')==1 && strcmp(options.priors.name,'
        minnesota')==1) || ...
168     (isfield(options,'prior')==1 && strcmp(options.prior.name,'Minnesota')
        ==1) || (isfield(options,'prior')==1 && strcmp(options.prior.name,'
        minnesota')==1)
169         % MINNESOTA PRIOR
170         dummy = 1;
171         flat = 0;
172         timetrend = 0;
173         priors.name= 'Minnesota';
174     end
175     if isfield(options,'minn_prior_tau')==1 || isfield(options,'bvar_prior_tau
       ')==1
176         dummy = 1;
177         flat = 0;
178         priors.name= 'Minnesota';
179         % MINNESOTA PRIOR: tightness
180         if isfield(options,'bvar_prior_mu')==1
181             minn_prior_tau = options.bvar_prior_tau;
182         else
183             minn_prior_tau = options.minn_prior_tau;
184         end
185     end

```

```

186     if isfield(options,'minn_prior_decay')==1 || isfield(options,'
        bvar_prior_decay')==1
187         dummy = 1;
188         flat = 0;
189         priors.name= 'Minnesota';
190         % MINNESOTA PRIOR: decay
191         if isfield(options,'bvar_prior_mu')==1
192             minn_prior_decay = options.bvar_prior_decay;
193         else
194             minn_prior_decay = options.minn_prior_decay;
195         end
196     end
197     if isfield(options,'minn_prior_lambda')==1 || isfield(options,'
        bvar_prior_lambda')==1
198         dummy = 1;
199         flat = 0;
200         priors.name= 'Minnesota';
201         % MINNESOTA PRIOR: sum-of-coeff
202         if isfield(options,'bvar_prior_mu')==1
203             minn_prior_lambda = options.bvar_prior_lambda;
204         else
205             minn_prior_lambda = options.minn_prior_lambda;
206         end
207     end
208     if isfield(options,'minn_prior_mu')==1 || isfield(options,'bvar_prior_mu')
        ==1
209         dummy = 1;
210         flat = 0;
211         priors.name= 'Minnesota';
212         % MINNESOTA PRIOR: co-persistence
213         if isfield(options,'bvar_prior_mu')==1
214             minn_prior_mu = options.bvar_prior_mu;
215         else
216             minn_prior_mu = options.minn_prior_mu;
217         end
218     end
219     if isfield(options,'minn_prior_omega')==1 || isfield(options,'
        bvar_prior_omega')==1
220         dummy = 1;
221         flat = 0;
222         priors.name= 'Minnesota';
223         % MINNESOTA PRIOR: variance
224         if isfield(options,'bvar_prior_omega')==1
225             minn_prior_omega = options.bvar_prior_omega;
226         else
227             minn_prior_omega = options.minn_prior_omega;
228         end

```

```

229     end
230     if isfield(options,'max_minn_hyper')==1 && options.max_minn_hyper ==1 &&
        mixed_freq_on ==0
231         % maximize the hyper parameters of the minnesota prior
232         hyperpara(1) = minn_prior_tau;
233         hyperpara(2) = minn_prior_decay;
234         hyperpara(3) = minn_prior_lambda;
235         hyperpara(4) = minn_prior_mu;
236         hyperpara(5) = minn_prior_omega;
237
238         dummy = 1;
239         flat = 0;
240         priors.name= 'Minnesota';
241         try
242             [postmode,lm,~] = bvar_max_hyper(hyperpara,y,lags,options);
243
244             minn_prior_tau      = postmode(1);
245             minn_prior_decay    = postmode(2);
246             minn_prior_lambda   = postmode(3);
247             minn_prior_mu       = postmode(4);
248             minn_prior_omega    = postmode(5);
249             disp('~~~~~')
250             disp('Maximization Successful: I will use the the mode values.')
251
252         catch
253             warning('Maximization NOT Successful')
254             disp('Using hyper parameter default values')
255         end
256     end
257     %=====
258     % Conjugate/Hierachical prior options
259     %=====
260     if (isfield(options,'priors')==1 && strcmp(options.priors.name,'Conjugate'
        )==1) || (isfield(options,'priors')==1 && strcmp(options.priors.name,'
        conjugate')==1) || ...
261     (isfield(options,'prior')==1 && strcmp(options.prior.name,'Conjugate')
        ==1) || (isfield(options,'prior')==1 && strcmp(options.prior.name,'
        conjugate')==1)
262
263         if isfield(options,'prior')==1
264             options.priors = options.prior;
265         end
266         if dummy == 1
267             warning('You have set both the Conjugate and Minnesota (perhaps
                via options.minn_prior_XX)');
268             warning('I will consider the Conjugate prior only');
269         end

```

```

270     dummy = 2;
271     % warning('The Conjugate prior is still under construction ... ');
272     flat = 0;
273     priors.name= 'Conjugate';
274     % Priors for the AR parameters
275     if isfield(options.priors,'Phi') == 1
276         % mean
277         if isfield(options.priors.Phi,'mean') == 1
278             prior.Phi.mean = options.priors.Phi.mean;
279             if max(size(prior.Phi.mean) ~= [ny*lags+(1-noconstant)+
                timetrend+nexogenous ny]) ~= 0
280                 error('Size_mismatch')
281             end
282         else
283             warning(['You_did_not_provide_a_prior_mean_for_the_AR_coeff._'
                ...
                'Assume_zeros_everywhere.'])
284             prior.Phi.mean = zeros(ny*lags+(1-noconstant)+timetrend+
                nexogenous , ny);
285         end
286         % variance
287         if isfield(options.priors.Phi,'cov') == 1
288             prior.Phi.cov = options.priors.Phi.cov;
289             % if length(prior.Phi.cov) ~= (ny*lags+(1-noconstant) +
290             % timetrend ) * ny
291             if length(prior.Phi.cov) ~= (ny*lags+(1-noconstant) +timetrend
                +nexogenous ) || size(prior.Phi.cov,1)~=size(prior.Phi.
                cov,2)
292                 error('Size_mismatch:_Covariance_Phi_should_be_square,_e.g
                _size(Phi.mean,1)x_size(Phi.mean,1)x')
293             end
294         else
295             warning(['You_did_not_provide_a_Covariance_for_the_AR_coeff._'
                ...
                'Assume_10_times_Identity_Matrix.'])
296             % prior.Phi.cov = 10 * eye((ny*lags+(1-noconstant) +
297             % timetrend) * ny);
298             prior.Phi.cov = 10 * eye((ny*lags+(1-noconstant) + timetrend+
                nexogenous));
299         end
300     else
301         warning(['You_did_not_provide_prior_mean_and_covariance_for_the_AR
                _coeff_ ...
302             'Assume_zeros_everywhere_with_covariance_10_times_Identity_
                Matrix.'])
303         % prior.Phi.cov = 10 * eye((ny*lags+(1-noconstant) +timetrend )
                * ny);

```

```

304         prior.Phi.cov    = 10 * eye((ny*lags+(1-noconstant) + timetrend+
           nexogenous ));
305         prior.Phi.mean    = zeros(ny*lags+(1-noconstant) + timetrend+
           nexogenous , ny);
306     end
307     % Priors for the Residual Covariance
308     if isfield(options.priors,'Sigma') == 1
309         % scale
310         if isfield(options.priors.Sigma,'scale') == 1
311             prior.Sigma.scale = options.priors.Sigma.scale;
312             if size(prior.Sigma.scale) ~= [ny ny]
313                 error('Size_mismatch')
314             end
315         else
316             warning(['You did not provide a prior mean for the Residual _
           Covariance. _' ...
           'Assume identity matrix.'])
317             prior.Sigma.scale = eye(ny);
318         end
319         % degrees of freedom
320         if isfield(options.priors.Sigma,'df') == 1
321             prior.Sigma.df = options.priors.Sigma.df;
322             if length(prior.Sigma.df) ~= 1
323                 error('Size_mismatch')
324             end
325             if prior.Sigma.df/2 <= ny-1
326                 error('Too few degrees of freedom _ _ Increase prior df')
327             end
328         else
329             warning(['You did not provide the degrees of freedom for the _
           Residual Covariance. _' ...
           'Assume ny+1.'])
330             prior.Sigma.df = ny + nexogenous + timetrend + 1;
331             while prior.Sigma.df/2 <= ny-1 % too few df
332                 prior.Sigma.df = prior.Sigma.df + 1;
333             end
334         end
335     end
336     else
337         warning(['You did not provide prior mean and variance for the _
           Residual Covariance. _' ...
           'Assume an identity matrix with N+1 degrees of freedom. _
           '])
338         prior.Sigma.scale = eye(ny);
339         prior.Sigma.df    = ny + nexogenous + timetrend + 1;
340         while prior.Sigma.df/2 <= ny-1 % too few df
341             prior.Sigma.df = prior.Sigma.df + 1;
342         end
343     end
344 end

```

```

345         end
346     end
347     %=====
348     % IRF options
349     %=====
350     if isfield(options,'hor')==1
351         hor = options.hor;
352     end
353     if isfield(options,'long_run_irf')==1
354         % Activating Long run IRF
355         long_run_irf = options.long_run_irf;
356     end
357     if isfield(options,'signs')==1
358         % Activating IRF with sign restrictions (multiple horizons allowed)
359         signs_irf = 1;
360         signs = options.signs;
361         if iscellstr(signs) == 0
362             error(['options.signs should be a cell array. Each cell must contain a string with the format \'...
363                 \'n\'y(a,b,c)<0\' or \'y(a,b,c)>0\' where a, b and c are integers.',...
364                 \'na=index of the variable',...
365                 \'nb=horizon',...
366                 \'nc=index of the shock'],class(zeros))
367         end
368     end
369     if isfield(options,'narrative')==1
370         if signs_irf == 0
371             warning('You did not provide any sign restrictions.')
372             signs{1} = 'isempty(y(1,1,1))==0';
373         end
374         narrative_signs_irf = 1;
375         narrative = options.narrative ;
376     end
377     if isfield(options,'zeros_signs')==1
378         if signs_irf == 1
379             signs_irf = 0; % deactivating signs
380         end
381         if narrative_signs_irf == 1
382             narrative_signs_irf = 0; % deactivating narrative
383         end
384         % Activating IRF with zeros and sign restrictions (multiple horizons NOT allowed)
385         zeros_signs_irf = 1;
386         zeros_signs = options.zeros_signs;
387         if iscellstr(zeros_signs) == 0
388             error(['options.zeros_signs should be a cell array. \'...

```



```

389         '\nEach cell must contain a string with the following format'
390         ...
391         '\nFor sign restrictions 'y(a,b)=1' or 'y(a,b)=-1'',',...
392         '\nFor short run zero restriction 'ys(a,b)=0'',',...
393         '\nFor long run restriction 'yr(a,1,b)=0' where a and b are
394         integers.',...
395         '\na=_index_of_the_variable',...
396         '\nb=_index_of_the_shock'],class(zeros_signs))
397
398     end
399     [f,sr] = sign2matrix(zeros_signs,ny);
400     if isfield(options,'var_pos')==0
401         var_pos = ones(1,ny);
402     else
403         var_pos = options.var_pos;
404     end
405 end
406
407 if isfield(options,'proxy')==1
408     % Activating IRF with proxy
409     proxy_irf = 1;
410     in.proxies = options.proxy;
411     in.vars = y;
412     in.p = lags;
413     if isfield(options,'proxy_end') == 1
414         in.T_m_end = options.proxy_end;
415     else
416         in.T_m_end = 0; %if the times series of the instrument ends when
417         VAR data ends
418     end
419     in.irhor = hor;
420     if isnumeric(in.proxies) == 0
421         error(['options.proxy should be a numeric array (nans or inf not
422         allowed)'],class(in.proxies))
423     end
424 end
425
426 if isfield(options,'irf_1STD')==1
427     % Activating of unitary IRF, i.e. a unitary increase in the shocks
428     % (instead of 1 STD)
429     irf_1STD = options.irf_1STD;
430 end
431
432 %=====
433 % (Un) Conditional Forecasts options
434 %=====
435
436 if isfield(options,'fhor')==1
437     fhor = options.fhor;
438     if fhor < 1
439         error('Forecast horizon must be positive')
440     end
441 end

```

```

432 end
433 if isfield(options,'endo_index')==1
434     % Forecast conditional on the path of an endogenous var
435     cfrfst_yes = 1;
436     if isfield(options,'endo_path')== 0
437         error('You need to provide the path for the endogenous variable')
438     end
439     % rows forecasts , column variables
440     endo_path = options.endo_path;
441     endo_path_index = options.endo_index;
442     if length(endo_path_index) ~= size(endo_path)
443         error(['Mismatch between the number of endogenous paths and the
444             number of conditioned variables'...
445             '\nE.g. the # of conditioned variables must coincide with the
446             # of column in 'options.endo_path'],class(
447                 endo_path_index));
448     end
449     if isfield(options,'exo_index')==1
450         % Forecast conditional on the path of an endo var using only a
451         % subset of shocks. notice that the # of endo and # exo must
452         % coincide
453         cfrfst_yes = 2;
454         exo_index = options.exo_index;
455         Omega = eye(ny);
456         % if isfield(options,'Omegaf')==1
457         % Omegaf = options.Omegaf;
458         % end
459         if length(exo_index) ~= length(endo_path_index)
460             error('the # of conditioned endogenous and # exogenous shocks
461                 used must coincide');
462         end
463     end
464 end
465 %=====
466 % Missing Values options
467 %=====
468 if mixed_freq_on == 1 && isfield(options,'mixed_freq_index')==1
469     index = options.mixed_freq_index;
470     if length(index) ~= size(y,2)
471         error(['You have to specify as many index as observables.'...
472             '\nIf no missing values for var j index(j)=0'...
473             '\nIf missing values for var j, and var j is a stock index(j)
474             =0'...
475             '\nIf missing values for var j, and var j is a real flow index
476             (j)=2'],class(index))
477     end

```

```

472     elseif mixed_freq_on == 1 && isfield(options,'mf_varindex')== 1
473         index = zeros(ny,1);
474         index(options.mf_varindex) = 2;
475     elseif (mixed_freq_on == 1 && isfield(options,'mixed_freq_index')== 0) ||
         (mixed_freq_on == 1 && isfield(options,'mf_varindex')== 0)
476         warning(['You did not specified the aggregation of the mixed_freq.'...
477             '\nI will treat them as stocks.']);
478         index = zeros(size(y,2),1);
479     end
480     if isfield(options,'noprint')==1
481         noprint = options.noprint;
482     end
483 end
484
485 %*****
486 %* Consistency Checks
487 %*****
488 nob = size(y,1)-firstobs+1;
489 if (firstobs+ nob-1)> size(y,1)
490     fprintf('Incorrect or missing specification of the number of observations.
         nob can be at most %4u\n',size(y,1)-firstobs+1);
491     error('Inconsistent number of observations.')
492 end
493 if firstobs + presample + lags >= nob
494     error('presample too large')
495 end
496 if firstobs + presample <= lags
497     error('firstobs+presample should be > #lags (for initializing the VAR)'
         )
498 end
499 if dummy == 1 && nexogenous > 0
500     warning('I will not use exogenous variables with Minnesota');
501     nexogenous = 0;
502 end
503 if mixed_freq_on == 1 && nexogenous > 0
504     warning('I will not use exogenous variables with missing observations');
505     nexogenous = 0;
506 end
507 if cfrst_yes ~= 0 && nexogenous > 0
508     warning('I will not use exogenous variables with conditional forecasts');
509     nexogenous = 0;
510 end
511 if size(exogenous,1) == size(y,1) && nexogenous > 0
512     warning('For forecast purposes, I will assume that exo are zero out-of-
         sample.')
513     fprintf('To change this, include the exogenous forecasts in options.
         exogenous.\n')

```

```

514     exogenous = [exogenous; zeros(fhor,nexogenous)];
515 end
516
517 %*****
518 %* Priors and Posterior Distributions
519 %*****
520
521 idx = firstobs+presample-lags:firstobs+nobs-1;
522 nx = 1;
523 if noconstant
524     nx = 0;
525 end
526
527 % organize data as  $yy = XX B + E$ 
528 [yy,XX] = YXB_(y(idx, :),lags,[nx timetrend]);
529
530 ydum = [];
531 xdum = [];
532 pbreaks = 0;
533 lambda = 0;
534 mu = 0;
535
536 ydata = y(idx, :);
537 T = size(ydata, 1);
538 if T-lags < lags*ny + nx %+ flat*(ny+1)
539     error('Less observations than regressors: increase the # of obs or decrease the # of lags.')
540 end
541 xdata = ones(T,nx);
542 if timetrend == 1
543     % xdata = [xdata [1:T]'];
544     xdata = [xdata [1-lags : T-lags]'];
545 end
546 if nexogenous > 0
547     xdata = [xdata exogenous(idx,:)];
548     XX = [XX exogenous(idx(1)+lags : idx(end),:)];
549 end
550
551 % OLS estimate [NO DUMMY]:
552 varols = rfvar3(ydata, lags, xdata, [T; T], 0, 0);
553
554
555 % specify the prior
556 if dummy == 1
557     % MINNESOTA PRIOR:
558     mnprior.tight = minn_prior_tau;
559     mnprior.decay = minn_prior_decay;

```

```

560 % Use only initializations lags for the variance prior
561 % vprior.sig = std(y(firstobs+presample-lags : firstobs+
    presample, :))';
562 vprior.sig = std(y(firstobs-lags : firstobs+presample-1,:))';
563 vprior.w = minn_prior_omega;
564 lambda = minn_prior_lambda;
565 mu = minn_prior_mu;
566 [ydum, xdum, pbreaks] = varprior(ny, nx, lags, mnprior, vprior);
567 % Prior density
568 Tp = presample + lags;
569 if nx
570     xdata = xdata(1:Tp, :);
571 else
572     xdata = [];
573 end
574 % varp = rfvar3([ydum(1:Tp, :); ydum], lags, [xdata; xdum], [
    Tp; Tp + pbreaks], lambda, mu);
575 varp = rfvar3([y(firstobs-lags : firstobs+presample-1, :); ydum
    ], lags, [xdata; xdum], [Tp; Tp + pbreaks], lambda, mu);
576 Tup = size(varp.u, 1);
577 prior.df = Tup - ny*lags - nx - flat*(ny+1);
578 prior.S = varp.u' * varp.u;
579 prior.XXi = varp.xxi;
580 prior.PhiHat = varp.B;
581 prior.YYdum = varp.y;
582 prior.XXdum = varp.X;
583 if prior.df < ny
584     error('Too few degrees of freedom in the Inverse-Wishart part of prior
        _distribution. You should increase training sample size.')
585 end
586 prior.minn_prior_tau = minn_prior_tau;
587 prior.minn_prior_decay = minn_prior_decay;
588 prior.minn_prior_lambda = minn_prior_lambda;
589 prior.minn_prior_mu = minn_prior_mu;
590 prior.minn_prior_omega = minn_prior_omega;
591 elseif dummy == 0
592     % JEFFREY OR UNIFORMATIVE PRIOR:
593     % [priors] = jeffrey(y, lags);
594     prior.name = 'Jeffrey';
595
596 end
597
598 % specify the posterior (the varols agin on actual+dummy)
599 [posterior] = posterior(y);
600
601
602 %*****

```

```

603  /* Compute the log marginal data density for the VAR model
604  *****
605
606  posterior_int = matricint(posterior.S, posterior.df, posterior.XXi);
607  if dummy == 1 % only for minnesota dummy
608      prior_int = matricint(prior.S, prior.df, prior.XXi);
609      lik_nobs   = posterior.df - prior.df;
610      log_dnsty  = posterior_int - prior_int - 0.5*ny*lik_nobs*log(2*pi);
611
612  elseif dummy == 0 % jeffrey prior
613      lik_nobs   = posterior.df;
614      log_dnsty  = posterior_int - 0.5*ny*lik_nobs*log(2*pi);
615
616  elseif dummy == 2 % conjugate MN-IW prior
617      log_dnsty  = mniw_log_dnsty(prior,posterior,varols);
618
619  end
620
621
622  *****
623  /* Generating draws form the Posterior Distribution
624  *****
625
626  % Preallocation of memory
627  % Matrices for collecting draws from Posterior Density
628  % Last dimension corresponds to a specific draw
629  Phi_draws      = zeros(ny*lags+nx+timetrend + nexogenous, ny, K); %
        Autoregressive Parameters
630  Sigma_draws    = zeros(ny,ny,K); % Shocks Covariance
631  ir_draws       = zeros(ny,hor,ny,K); % variable , horizon , shock
        and draws - Cholesky IRF
632  irlr_draws     = zeros(ny,hor,ny,K); % variable , horizon , shock
        and draws - Long Run IRF
633  Qlr_draws      = zeros(ny,ny,K); % long run impact matrix
634  e_draws        = zeros(size(yy,1), ny,K); % residuals
635  yhatfut_no_shocks      = NaN(fhor, ny, K); % forecasts with shocks
636  yhatfut_with_shocks    = NaN(fhor, ny, K); % forecast without the shocks
637  yhatfut_cfrfst        = NaN(fhor, ny, K); % forecast conditional on
        endogenous path
638  logL            = NaN(K,1);
639
640  if signs_irf == 1
641      irsign_draws = ir_draws;
642      Omega_draws  = Sigma_draws;
643  end
644  if narrative_signs_irf == 1
645      irnarrsign_draws = ir_draws;

```

```

646     Omegan_draws      = Sigma_draws;
647 end
648 if zeros_signs_irf == 1
649     irzerosign_draws   = ir_draws;
650     Omegaz_draws       = Sigma_draws;
651 end
652 if proxy_irf == 1
653     irproxy_draws = ir_draws;
654 end
655
656 % Settings for the forecasts
657 forecast_data.xdata      = ones(fhor, nx);
658 if timetrend
659     forecast_data.xdata = [forecast_data.xdata (T-lags+1 : T-lags+fhor)'];
660 end
661 if nexogenous>0
662     forecast_data.xdata = [forecast_data.xdata exogenous(T-lags+1 : T-lags+
        fhor,:)];
663 end
664 forecast_data.initval     = ydata(end-lags+1:end, :);
665
666 try
667     S_inv_upper_chol      = chol(inv(posterior.S));
668 catch
669     warning('POSTERIOR_MEAN_of_SIGMA_IS_ILL-BEHAVED_(NON-POSITIVE_DEFINITE)')
670     try
671         warning('I_will_try_with_the_LDL_decomposition')
672         iS              = inv(posterior.S);
673         [L, D, P]        = ldl(iS);
674         S_inv_upper_chol = sqrt(D) * L' * P';%chol()
675     catch
676         warning('I_will_add_1e-05_to_the_diagonal')
677         S_inv_upper_chol = chol(inv(posterior.S + 1e-05*eye(ny)));
678     end
679 end
680
681
682 XXi_lower_chol          = chol(posterior.XXi)';
683
684 nk                      = ny*lags+nx+timetrend + nexogenous;
685
686 % Declaration of the companion matrix
687 Companion_matrix = diag(ones(ny*(lags-1),1),-ny);
688
689 waitbar_yes = 0;
690 if K > 99
691     waitbar_yes = 1;

```

```

692     wb = waitbar(0, 'Generating draws from the Posterior Distribution');
693 end
694
695 for d = 1 : K
696
697     %=====
698     % Inference: Drawing from the posterior distribution
699     % Step 1: draw from the Covariance
700     Sigma = randinversewishart(ny, posterior.df, S_inv_upper_chol);
701
702     % Step 2: given the Covariance Matrix, draw from the AR parameters
703     Sigma_lower_chol = chol(Sigma)';
704     Phi1 = randn(nk * ny, 1);
705     Phi2 = kron(Sigma_lower_chol, Xxi_lower_chol) * Phi1;
706     Phi3 = reshape(Phi2, nk, ny);
707     Phi = Phi3 + posterior.PhiHat;
708
709
710     % store the draws
711     Phi_draws(:, :, d) = Phi;
712     Sigma_draws(:, :, d) = Sigma;
713     errors = yy - XX * Phi;
714     e_draws(:, :, d) = errors;
715
716
717     %=====
718     % IRF
719     % Compute the impulse response functions
720     % with cholesky
721     if irf_1STD == 1
722         % one STD increase
723         ir_draws(:, :, :, d) = iresponse(Phi, Sigma, hor, eye(ny));
724     else
725         % one percent increase
726         ir_draws(:, :, :, d) = iresponse(Phi, Sigma, hor, eye(ny), 0);
727     end
728     % with long run restrictions
729     if long_run_irf == 1
730         [irlr, Qlr] = iresponse_longrun(Phi, Sigma, hor, lags);
731         irlr_draws(:, :, :, d) = irlr;
732         Qlr_draws(:, :, d) = Qlr;
733     end
734     % with sign restrictions
735     if signs_irf == 1
736         [irsign, Omega] = iresponse_sign(Phi, Sigma, hor, signs);
737         irsign_draws(:, :, :, d) = irsign;
738         Omega_draws(:, :, d) = Omega;

```



```

739 end
740 % with narrative and sign restrictions
741 if narrative_signs_irf == 1
742     [irnarrsign, Omega] = iresponse_sign.narrative(errors, Phi, Sigma
743         , hor, signs, narrative);
744     irnarrsign_draws(:, :, :, d) = irnarrsign;
745     Omegan_draws(:, :, d) = Omega;
746 end
747 % with zeros and sign restrictions
748 if zeros_signs_irf == 1 % = iresponse_zeros_signs(Phi, Sigma, bvar1.
749     hor, lags, var_pos, f, sr);
750     [irzerosign, Omega] = iresponse_zeros_signs(Phi, Sigma, hor, lags
751         , var_pos, f, sr);
752     irzerosign_draws(:, :, :, d) = irzerosign;
753     Omegaz_draws(:, :, d) = Omega;
754 end
755 % with proxy
756 if proxy_irf == 1
757     in.res = e_draws(:, :, d);
758     in.Phi = Phi_draws(:, :, d);
759     in.Sigma = Sigma;
760     tmp_ = iresponse_proxy(in);
761     irproxy_draws(:, :, 1, d) = tmp_.irs';
762     clear tmp_
763 end
764 %=====
765 % Forecasts
766 % compute the out of sample forecast (unconditional)
767 [frcst_no_shock, frcsts_with_shocks] = forecasts(forecast_data, Phi, Sigma,
768     fhor, lags);
769 yhatfut_no_shocks(:, :, d) = frcst_no_shock;
770 yhatfut_with_shocks(:, :, d) = frcsts_with_shocks;
771
772 if cfrfst_yes == 1
773     % Forecast conditional on the path of an endo var using all shocks
774     [sims_with_endopath, EPS(:, :, d)] = ...
775         cforecasts(endo_path, endo_path_index, forecast_data, Phi, Sigma);
776     yhatfut_cfrfst(:, :, d) = sims_with_endopath;
777
778 elseif cfrfst_yes == 2
779     % Forecast conditional on the path of an endo var using only a
780     % subset of shocks. notice that the # of endo and # exo must coincide
781     % Omega is the structural orthonormal matrix
782     [sims_with_endopath, EPS(:, :, d)] = ...
783         cforecasts2(endo_path, endo_path_index, exo_index, forecast_data, Phi,
784             Sigma, Omega);

```

```

781     yhatfut_cfrfst(:, :, d) = sims_with_endopath;
782 end
783
784
785 %=====
786 % Mixed Frequency
787 if mixed_freq_on
788     % Checking the eigenvalues of the companion matrix (on or inside the
789     % unit circle). Needed for the initial of KF
790     Companion_matrix(1:ny, :) = Phi(1:ny*lags, :)';
791     test = (abs(eig(Companion_matrix)));
792     if any(test > 1.000000000000001)
793         KFoptions.initialCond = 1;
794     end
795     KFoptions.index = index;
796     KFoptions.noprint = noprint;
797     % Forward Kalman Filter and Smoother
798     [KFout] = kfilternan(Phi, Sigma, yoriginal, KFoptions);
799     yfill(:, :, d) = KFout.smoothSt_plus_ss(:, KFout.index_var);
800     yfilt(:, :, d) = KFout.filteredSt_plus_ss(:, KFout.index_var);
801     % recompute the posterior with smoothed data
802     [posterior1] = posterior_(yfill(:, :, d));
803     S_inv_upper_chol = chol(inv(posterior1.S));
804     XXi_lower_chol = chol(posterior1.XXi)';
805     posterior.PhiHat = posterior1.PhiHat;
806     forecast_data.initval = yfill(end-lags+1:end, :, d);
807     logL(d) = KFout.logL;
808 end
809 if waitbar_yes, waitbar(d/K, wb); end
810 end
811 if waitbar_yes, close(wb); end
812
813
814 %*****
815 %* Storing the results
816 %*****
817
818 % classical inference: OLS estimator
819 BVAR.Phi_ols = varols.B;
820 BVAR.e_ols = varols.u;
821 BVAR.Sigma_ols = 1/(nobs-nk)*varols.u'*varols.u;
822 [BVAR.InfoCrit.AIC, BVAR.InfoCrit.HQIC, BVAR.InfoCrit.BIC] = IC(BVAR, nobs, nk
    );
823 % the model with the lowest IC is preferred
824
825 % OLS irf
826 BVAR.ir_ols = iresponse(BVAR.Phi_ols, BVAR.Sigma_ols, hor, eye(ny));

```

```

827 if long_run_irf == 1
828     [irlr,Qlr]           = iresponse_longrun(BVAR.Phi_ols,BVAR.Sigma_ols,
        hor,lags);
829     BVAR.irlr_ols        = irlr;
830     BVAR.Qlr_ols(:, :)   = Qlr;
831 end
832
833 % test the normality of the ols VAR residuals (matlab stat toolbox needed)
834 if exist('kstest')==2
835     for gg = 1 : ny
836         [H,Pv] = kstest(BVAR.e_ols(:,gg)/BVAR.Sigma_ols(gg,gg));
837         BVAR.HP(gg,:) = [H,Pv];
838     end
839 else
840     BVAR.HP = [];
841 end
842
843 % bayesian inference :
844 % the last dimension of these objects corresponds to a draw from the posterior
845
846 % inference and IRFs
847 BVAR.Phi_draws      = Phi_draws;           % draws from the autoregressive part
848 BVAR.Sigma_draws    = Sigma_draws;         % draws from the covariance matrix
849 BVAR.alpha_draws    = Phi_draws;         % Older name: draws from the
        autoregressive part
850 BVAR.sigma_draws    = Sigma_draws;         % Older name: draws from the covariance
        matrix
851
852 BVAR.ir_draws       = ir_draws;           % draws from the IRF with cholesky
853 BVAR.irlr_draws     = irlr_draws;         % draws from the IRF with Long Run
854 BVAR.Qlr_draws      = Qlr_draws;         % Long Run Rotation matrix
855 BVAR.lags           = lags;              % lags
856 BVAR.N              = ny;               % number of variables
857 BVAR.e_draws        = e_draws;           % residuals
858 BVAR.e              = e_draws;           % backward compatible with earlier
        versions
859
860 BVAR.posterior      = posterior;
861 BVAR.prior          = prior;             % priors used
862 BVAR.logmlike       = log_dnsty;
863 BVAR.X              = var.X;             % regressors (including dummy if
        Minnesota)
864 BVAR.y              = var.y;             % dependent (including dummy if
        Minnesota)
865 BVAR.XX             = XX;               % regressors (no dummy)
866 BVAR.yy             = yy;               % dependent (no dummy)
867

```

```

868 % prediction
869 BVAR.fhor      = fhor;          % forecast horizon
870 BVAR.hor       = hor;          % IRF horizon
871 BVAR.forecasts.no_shocks      = yhatfut_no_shocks;      % trajectories of
      forecasts without shocks
872 BVAR.forecasts.with_shocks   = yhatfut_with_shocks;    % trajectories of
      forecasts with shocks
873 BVAR.forecasts.conditional   = [];                    % trajectories of conditional
      forecasts
874 BVAR.forecasts.EPScond       = [];                    % shocks of conditional
      forecasts
875 if cfrfst_yes ~= 0
876     BVAR.forecasts.conditional = yhatfut_cfrfst;        % trajectories
      of forecasts
877     BVAR.forecasts.EPScond     = EPS;                  % shocks of
      forecasts
878 end
879 BVAR.forecast_data            = forecast_data;
880
881 %
882 BVAR.varnames                = varnames;
883 BVAR.ndraws                  = K;
884
885 if signs_irf == 1 && narrative_signs_irf == 0
886     BVAR.irsign_draws = irsign_draws;
887     BVAR.Omegas       = Omega_draws(:, :, d);
888 else
889     BVAR.irsign_draws = [];
890     BVAR.Omegas       = [];
891 end
892 if narrative_signs_irf == 1
893     BVAR.irnarrsign_draws = irnarrsign_draws;
894     BVAR.Omeganz          = Omeganz_draws(:, :, d);
895 else
896     BVAR.irnarrsign_draws = [];
897     BVAR.Omeganz          = [];
898 end
899 if zeros_signs_irf == 1
900     BVAR.irzerosign_draws = irzerosign_draws;
901     BVAR.Omegaz           = Omegaz_draws(:, :, d);
902 else
903     BVAR.irzerosign_draws = [];
904     BVAR.Omegaz           = [];
905 end
906 if proxy_irf == 1
907     BVAR.irproxy_draws = irproxy_draws;
908 else

```

```

909     BVAR.irproxy_draws= [];
910 end
911
912 % missing observations
913 BVAR.data      = y;                % raw data
914 if mixed_freq_on
915     BVAR.yfill = yfill;
916     BVAR.yfilt = yfilt;
917     BVAR.yinterpol = yinterpol;
918     BVAR.logL   = logL;
919 end
920
921 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
922 % end of bvar.m
923 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
924
925 %*****
926 %*****
927     function [posterior] = posterior_(y)
928
929         % This part is needed in the case of missing values. The posterior
930         % needs to be reevaluated given a new value of the Kalman smoothed
931         % observables
932         %=====
933         ydata = y(idx, :);
934         %=====
935         T      = size(ydata, 1);
936         xdata = ones(T,nx);
937         if timetrend ==1
938             xdata = [xdata [1-lags:T-lags]'];
939             % xdata = [xdata [1:T]'];
940         end
941         if nexogenous >0
942             xdata = [xdata exogenous(idx,:)];
943         end
944         % posterior density
945         var = rfvar3([ydata; ydum], lags, [xdata; xdum], [T; T+pbreaks],
946                     lambda, mu);
947         Tu = size(var.u, 1);
948
949         if dummy == 1
950             %*****
951             % Minnesota Prior
952             %*****
953             % Prior density
954             Tp = presample + lags;
955             if nx

```

```

955         xdata = xdata(1:Tp, :);
956     else
957         xdata = [];
958     end
959     % varp = rfvar3([ydata(1:Tp, :); ydum], lags, [xdata;
          xdum], [Tp; Tp + pbreaks], lambda, mu);
960     varp = rfvar3([y(firstobs-lags : firstobs+presample-1,
          :); ydum], lags, [xdata; xdum], [Tp; Tp + pbreaks], lambda, mu
          );
961     Tup = size(varp.u, 1);
962     prior.df = Tup - ny*lags - nx - flat*(ny+1);
963     prior.S = varp.u' * varp.u;
964     prior.XXi = varp.xxi;
965     prior.PhiHat = varp.B;
966     priors.YYdum = varp.y;
967     priors.XXdum = varp.X;
968     if prior.df < ny
969         error('Too few degrees of freedom in the Inverse-Wishart part
          of prior distribution. You should increase training sample
          size.')
970     end
971     posterior.df = Tu - ny*lags - nx - flat*(ny+1);
972     posterior.S = var.u' * var.u;
973     posterior.XXi = var.xxi;
974     posterior.PhiHat = var.B;
975
976 elseif dummy == 2
977     %*****
978     % Conjugate Prior
979     %*****
980     Ai = inv(prior.Phi.cov);
981     posterior.df = Tu - ny*lags - nx - nexogenous - prior.Sigma.df;
982     posterior.XXi = inv(var.X'*var.X + Ai);
983     posterior.PhiHat = posterior.XXi * (var.X' * var.y + Ai * prior.
          Phi.mean);
984     %posterior.S = var.u' * var.u + prior.Sigma.scale ;
985     %posterior.S = var.u' * var.u + prior.Sigma.scale + (var.B -
          prior.Phi.mean)' * Ai * (var.B - prior.Phi.mean);
986     %posterior.S = var.u' * var.u + prior.Sigma.scale + ...
987     % (posterior.PhiHat - prior.Phi.mean)' * Ai * (var.B - prior.
          Phi.mean);
988     % check
989     posterior.S = var.u' * var.u + prior.Sigma.scale + ...
990         prior.Phi.mean' * Ai * prior.Phi.mean + ...
991         var.B' * (var.X'*var.X) * var.B ...
992         - posterior.PhiHat' * (var.X'*var.X + Ai) * posterior.PhiHat;
993

```

```

994         else
995             %*****
996             % Flat Jeffrey Prior
997             %*****
998             posterior.df      = Tu - ny*lags - nx + flat*(ny+1) - nexogenous;
999             posterior.S       = var.u' * var.u;
1000             posterior.XXi     = var.xxi;
1001             posterior.PhiHat = var.B;
1002         end
1003
1004     end
1005     %*****
1006     %*****
1007     function [priors] = priors_( )
1008
1009         priors.name = 'N/A';
1010
1011     end
1012     %*****
1013     %*****
1014
1015 end

```

bvar_max_hyper.m

[illegible]

```

21 lb          = -1e10*ones(length(hyperpara),1);
22 ub          = 1e10*ones(length(hyperpara),1);
23 objective_function = 'bvar_opt_hyperpara';
24 dummy_      = 1;
25 x0          = log(hyperpara);
26 index_fixed = [];
27 index_est   = 1:5;
28 hyperpara_fidex = [];
29
30 if nargin > 3
31     if isfield(options,'index_est') ==1
32         clear x0 index_fixed index_est hyperpara_fidex;
33         index_est      = options.index_est;
34         index_fixed    = setdiff(1:5,index_est);
35         x0              = log(hyperpara(index_est));
36         hyperpara_fidex = log(hyperpara(index_fixed));
37     end
38     if isfield(options,'max_compute') ==1
39         max_compute     = options.max_compute;
40     end
41     if isfield(options,'lb') ==1
42         lb              = options.lb;
43         if length(lb) ~= length(x0)
44             error('Mismatch between the size of lower bounds and the param_
                    vector');
45         end
46     end
47     if isfield(options,'ub') ==1
48         ub              = options.ub;
49         if length(ub) ~= length(x0)
50             error('Mismatch between the size of upper bounds and the param_
                    vector');
51         end
52     end
53     if isfield(options,'objective_function') ==1
54         objective_function = options.objective_function;
55         dummy_ =0;
56     end
57 end
58
59
60 options.index_est      = index_est;
61 options.index_fixed    = index_fixed;
62 options.hyperpara_fidex = hyperpara_fidex;
63
64
65 switch max_compute

```



```

66     case 1 % unconstraint
67         % Set default optimization options for fminunc.
68         optim_options = optimset('display','iter','MaxFunEvals',100000,'TolFun',
        '1e-8','TolX',1e-6);
69         [xh,fh,~,~,~,H] = ...
70             fminunc(objective_function,x0,optim_options,y,lags,options);
71         %=====
72     case 2 % constraint
73         % Set default optimization options for fmincon.
74         optim_options = optimset('display','iter','LargeScale','off',' ',
        'MaxFunEvals',100000,'TolFun',1e-8,'TolX',1e-6);
75         [xh,fh,~,~,~,~,H] = ...
76             fmincon(objective_function,x0,[],[],[],[],lb,ub,[],optim_options,y
        ,lags,options);
77         %=====
78     case 3 % sims
79         crit = 10e-5;
80         nit = 10e-4;
81         [fh, xh, ~, H, ~, ~, ~] = csminwel(objective_function,x0,.1*eye(length
        (x0)),[],crit,nit,y,lags,options);
82         %=====
83     case 7 % Matlab's simplex (Optimization toolbox needed).
84     %         optim_options = optimset('display','iter','MaxFunEvals',30000,'
        MaxIter',10000,'TolFun',1e-3,'TolX',1e-3,'OutputFcn',@outsavefun);
85         optim_options = optimset('display','iter','MaxFunEvals',30000,'MaxIter
        ',10000,'TolFun',1e-3,'TolX',1e-3);
86         [xh,fh,~,~] = fminsearch(objective_function,x0,optim_options,y,lags,
        options);
87         H = zeros(length(x0));
88     end
89
90     if dummy_ % Print the output for Minnesoty prior with dummy
91         postmode = zeros(1,5);
92         postmode(options.index_est) = exp(xh);
93         postmode(options.index_fixed) = exp(options.hyperpara_fidex);
94
95         % processing the output of the maximization
96         log_dnsty = -fh;
97         JJ = jacob_bvar(xh);
98         HH = JJ * H * JJ';
99
100        disp('=====');
101        disp('___');
102        disp('**_Initial_Hyperpara_Values_and_Log_Density_**');
103        disp('___');
104        rowname = {'tau','decay','lambda','mu','omega','log_density'};
105        minus_log_dnsty_0 = bvar_opt_hyperpara(x0,y,lags,options);

```

```

106         x                = [hyperpara'; -minus_log_dnsty_0];
107     for jj =1: length(x)
108         X = sprintf('%s_=%0.5g',rownam{jj},x(jj));
109         disp(X)
110     end
111     disp(' _ ');
112     disp('**_Posterior_Mode:_ (Minimization_of _Log_Density)_**');
113     disp(' _ ');
114     x      = [postmode'; log_dnsty];
115     for jj =1: length(x)
116         X = sprintf('%s_=%0.5g',rownam{jj},x(jj));
117         disp(X)
118     end
119 else % other maximization e.g. conjugate MNIW
120     postmode      = exp(xh);
121     % processing the output of the maximization
122     log_dnsty     = -fh;
123     JJ            = jacob_bvar(xh);
124     HH            = JJ * H * JJ';
125 end
126
127
128 end

```

bvar_ml.m

[illegible]

```

20 %*****
21 % SETTINGS
22 %*****
23
24
25 % randn('state',999);
26 % rand('state',999);
27 %
28 ny          = size(y, 2);
29 first_obs   = lags+1;
30 presample   = 0;
31 noconstant  = 0;
32
33 bvar_prior_tau   = hyperpara(1);%3;
34 bvar_prior_decay = hyperpara(2);%0.5;
35 bvar_prior_lambda = hyperpara(3);%5;
36 bvar_prior_mu    = hyperpara(4);%2;
37 bvar_prior_omega = hyperpara(5);%1;
38 unit_root_      = ones(ny,1);
39 flat            = 0;
40
41
42 if nargin > 3
43 %     fields = fieldnames(options);
44 %     nf     = size(fieldnames(options),1);
45 %     for i = 1 : nf
46 %         if strcmp(fields{i},'bvar_prior_train ')
47 %             bvar_prior_train = options.bvar_prior_train;
48 %         elseif strcmp(fields{i},'first_obs ')
49 %             first_obs = options.first_obs;
50 %         elseif strcmp(fields{i},'presample ')
51 %             presample = options.presample;
52 %         elseif strcmp(fields{i},'train ')
53 %             train = options.train;
54 %         elseif strcmp(fields{i},'noconstant ')
55 %             noconstant = options.nocostant;
56 %         end
57 %     end
58     if isfield(options,'first_obs')==1
59         first_obs = options.first_obs;
60     end
61     if isfield(options,'presample')==1
62         presample = options.presample;
63     end
64     if isfield(options,'noconstant')==1
65         noconstant = options.nocostant;
66         % MINNESOTA PRIOR

```

```

67     end
68     if isfield(options,'unit_root')==1
69         unit_root_ = options.unit_root_;
70         % MINNESOTA PRIOR: unit root assumption only for a subset of var
71     end
72
73 end
74
75 nobs = size(y,1)-first_obs+1;
76
77 if (first_obs+ nobs-1)> size(y,1)
78     fprintf('Incorrect or missing specification of the number of observations.
79         _nobs can be at most %4u\n',size(y,1)-first_obs+1);
80     error('Inconsistent number of observations.')
81 end
82 % Parameters for prior
83 if first_obs + presample <= lags
84     error('first_obs+presample should be > lags (for initializing the VAR)')
85 end
86
87 if first_obs + presample <= lags
88     error('first_obs+presample should be > nlags (for initializing the VAR)'
89         )
90 end
91
92
93
94 if noconstant
95     nx = 0;
96 else
97     nx = 1;
98 end
99
100
101 mnprior.tight = bvar_prior_tau;
102 mnprior.decay = bvar_prior_decay;
103 mnprior.unit_root_ = unit_root_;
104 % Use only initializations lags for the variance prior
105 vprior.sig = std(y(first_obs+presample-lags : first_obs+presample,:))';
106 vprior.w = bvar_prior_omega;
107 lambda = bvar_prior_lambda;
108 mu      = bvar_prior_mu;
109 [ydum, xdum, pbreaks] = varprior(ny, nx, lags, mnprior, vprior);
110
111 ydata = y(idx, :);

```

```

112 T      = size(ydata, 1);
113 xdata = ones(T,nx);
114 % posterior density
115 var = rfvar3([ydata; ydum], lags, [xdata; xdum], [T; T+pbreaks], lambda, mu);
116 Tu = size(var.u, 1);
117
118 % Prior density
119 Tp = presample + lags;
120 if nx
121     xdata = xdata(1:Tp, :);
122 else
123     xdata = [];
124 end
125 varp = rfvar3([ydata(1:Tp, :); ydum], lags, [xdata; xdum], [Tp; Tp + pbreaks],
    lambda, mu);
126 Tup = size(varp.u, 1);
127
128 prior.df      = Tup - ny*lags - nx - flat*(ny+1);
129 prior.S       = varp.u' * varp.u;
130 prior.XXi     = varp.xxi;
131 prior.PhiHat  = varp.B;
132
133 priors.YYdum = varp.y;
134 priors.XXdum = varp.X;
135
136 if prior.df < ny
137     error('Too few degrees of freedom in the Inverse-Wishart part of prior_
    distribution. You should increase training sample size.')
138 end
139 posterior.df   = Tu - ny*lags - nx - flat*(ny+1);
140 posterior.S    = var.u' * var.u;
141 posterior.XXi  = var.xxi;
142 posterior.PhiHat = var.B;
143
144
145 %*****
146 %* Compute the log marginal data density for the VAR model
147 %*****
148
149 posterior_int  = matricint(posterior.S, posterior.df, posterior.XXi);
150 prior_int     = matricint(prior.S, prior.df, prior.XXi);
151 lik_nobs      = posterior.df - prior.df;
152 log_dnsty     = posterior_int - prior_int - 0.5*ny*lik_nobs*log(2*pi);
153 % minus_log_dnsty = -log_dnsty;
154 end

```

bvar_opt_hyperpara.m

```

1  function minus_log_dnsty = bvar_opt_hyperpara(hyperpara,y,lags,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'bvar_opt_hyperpara' computes the marginal likelihood over the
5  % hyperparameters of for the Minnesota prior
6  % Bridge function between 'bvar_ml' and the minimization routine
7  % 'bvar_max_hyper'
8
9  % Inputs:
10 % — hyperpara, Minnesota hyperpara over which maximize the marginal
11 % likelihood
12 % — y, data columns variables
13 % — lags, lag order of the VAR
14 % — options, see below for details
15
16 % Output: marginal data density
17
18 % Filippo Ferroni, 6/1/2017
19 % Revised, 3/21/2018
20 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
21
22 if nargin < 4
23     hyperpara = exp(hyperpara);
24     log_dnsty = bvar_ml(hyperpara,y,lags);
25 else
26     hyperpara(options.index_est) = exp(hyperpara);
27     if isempty(options.index_fixed) == 0
28         hyperpara(options.index_fixed) = exp(options.hyperpara_fidex);
29     end
30     log_dnsty = bvar_ml(hyperpara,y,lags,options);
31 end
32
33 minus_log_dnsty = -log_dnsty;

```

cforecasts.m

```

1  function [sims_with_endopath,EPSn] = cforecasts(endo_path,endo_index,
2         forecast_data,Phi,Sigma)
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'cforecasts' performs the forecast conditional on a path for one or more
5  % endogenous variables using all the shocks of the VAR
6  % references: Waggoner and Zha (1999) and Jarocinski (2010)

```

```

7
8 % Inputs:
9 % — endo_path, path for the endogenous variables
10 % — endo_index, order of the endogenous variables with a contional path
11 % — forecast_data, last data
12 % — Phi, AR parameters of the VAR
13 % — Sigma, Covariance matrix of the reduced form VAR shocks
14
15 % Output:
16 % — sims_with_endopath
17 % 1st dimension: horizon
18 % 2nd dimension: variable
19 % — EPSn, shocks generating the path
20
21 % Filippo Ferroni, 6/1/2015
22 % Revised, 2/15/2017
23 % Revised, 3/21/2018
24 % Revised, 9/11/2019
25 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
26
27 [fhor,Ncondvar] = size(endo_path);
28 Nvar           = size(Sigma,2);
29
30 if size(Ncondvar,2) ~= length(endo_index)
31     error('something went wrong: the number of path and the number of vars are
32           not consistent. ')
33 end
34 if size(endo_index,1) > size(endo_index,2)
35     error('the index needs to be a row vector');
36 end
37 Nres = size(endo_index,2)*fhor;
38 [F,~,~,const,~,lags] = var2ss(Phi,Sigma);
39
40 % endo_path_deviation = endo_path — repmat(const',1,fhor);
41
42 % no shocks forecasts
43 [sims_no_shock,~] = forecasts(forecast_data,Phi,Sigma,fhor,lags);
44
45 % restrictions
46 % err = endo_path_deviation — sims_no_shock(:,endo_path_index)
47 % ;
48 % err = zeros(Nvar*fhor);
49 err = endo_path — sims_no_shock(:,endo_index);
50 err = reshape(err',size(endo_index,1)*fhor,1);
51
52 % 1 standard deviation increase (if unit = eye(N))

```

```

52 [C] = chol(Sigma,'lower');
53
54 R      = zeros(Nvar * fhor);
55 tmp0 = [];
56 for ff = 1 :fhor
57     tmp  = F^(ff-1);
58     tmp0 = [tmp(1:Nvar,1:Nvar) * C tmp0];
59     R((ff-1)*Nvar + 1 : (ff)*Nvar, 1: size(tmp0,2) ) = tmp0;
60     index(ff,:) = (ff-1)*Nvar + endo_index;
61 end
62 index = reshape(index',Nres,1);
63 % index= [4 10];
64 Rtilde = R(index,:);
65
66 [U,D,V] = svd(Rtilde);
67 V1 = V(:,1:Nres);
68 V2 = V(:,Nres+1:end);
69
70 eps = V1*inv(D(1:Nres,1:Nres))*U'*err + V2 * randn(size(R,1)-Nres,1);
71
72 % orthogonal shocks
73 EPSn = reshape(eps,Nvar,fhor);
74 % reduced form shocks
75 EPS  = C*EPSn;
76
77 [~,sims_with_endopath] = forecasts(forecast_data,Phi,Sigma,fhor,lags,EPS');

```

cforecasts2.m

```

1 function [sims_with_endopath,EPsN] = cforecasts2(endo_path,endo_index,
        exo_index,forecast_data,Phi,Sigma,Omega,EPsi,epslags_)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % 'cforecasts' performs the forecast conditional on a path for one or more
5 % endogenous variables using some of teh STRUCTURAL shocks of the VAR
6 % references: Waggoner and Zha (1999), Maih (2010), Dynare reference manual
7 % Inputs:
8 % - endo_path, path for the endogenous variables
9 % - endo_index, order of the endogenous variables with a contional path
10 % - exo_index, order of the structural VAR shocks to use
11 % - forecast_data, last data
12 % - Phi, AR parameters of the VAR
13 % - Sigma, Covariance matrix of the reduced form VAR shocks
14 % - Omega, Rotation for identification
15

```



```

16 % Output:
17 % — sims_with_endopath
18 % 1st dimension: horizon
19 % 2nd dimension: variable
20 % — EPSn, shocks generating the path
21
22 % Filippo Ferroni, 6/1/2015
23 % Revised, 2/15/2017
24 % Revised, 3/21/2018
25 % Revised, 9/11/2019
26 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
27
28 [fhor,Ncondvar] = size(endo_path);
29 Nvar           = size(Sigma,2);
30
31 if nargin < 7
32     Omega = eye(Nvar);
33 end
34 if nargin < 8
35     % drawing the shocks
36     % 1 standard deviation increase (if unit = eye(N))
37     EPSi           = randn(fhor,Nvar);
38     EPSi(:,exo_index) = zeros(fhor,length(exo_index));
39 end
40 if nargin < 9
41     epslags_ = 0;
42 end
43
44 if size(Ncondvar,2) ~= length(endo_index)
45     error('something_went_wrong:_the_number_of_path_and_the_number_of_vars_are_not_consistent.')
```

```

61 % % restrictions
62 % % err = endo_path_deviation - sims_no_shock(:,
    endo_path_index);
63 % % err = zeros(Nvar*fhor);
64 % % err = (endo_path - sims_no_shock(:, endo_index))';
65 % % err = reshape(err', size(endo_index,1)*fhor,1);
66
67 [C] = chol(Sigma,'lower');
68 R = C*Omega;
69 EPS = (R*EPSi')';
70
71 if epslags_ == 1
72     tmp = A*B*R;
73     RR = tmp(1:Nvar,1:Nvar);
74
75 elseif epslags_ == 2
76     tmp = A^2*B*R;
77     RR = tmp(1:Nvar,1:Nvar);
78
79 elseif epslags_ == 3
80     tmp = A^3*B*R;
81     RR = tmp(1:Nvar,1:Nvar);
82 else
83     RR = R;
84 end
85
86 % With shocks but exo_index
87 lags_data = forecast_data.initval;
88 e = zeros(length(exo_index),fhor);
89
90 for t = 1 : fhor
91     X = [ reshape(flipdim(lags_data, 1)', 1, Nvar*lags) forecast_data.xdata(t,
        :) ];
92     % unconditional forecasts with all shocks but 'exo_index'
93     shock = EPS(t,:);
94     y = X * Phi + shock;
95     e(:,t) = inv(RR(endo_index,exo_index))*(endo_path(t,:) - y(endo_index));
96     y = y + (RR(:,exo_index)*e(:,t))';
97     lags_data(1:end-1,:) = lags_data(2:end, :);
98     lags_data(end,:) = y;
99     sims_with_endopath(t, :) = y;
100 end
101
102 EPSn = EPSi;
103 EPSn(:,exo_index) = e';

```

checkrestrictions.m

```

1  function d = checkrestrictions(restriction,y,v)
2
3  % Check the restrictions
4
5  if nargin<3
6      v = [];
7  end
8
9  d=0;
10 count = 0;
11 for ii = 1 : size(restriction,2)
12     tmp = eval(restriction{ii});
13     count = count + min(tmp);
14 end
15 if isempty(count)==1
16     error('There_is_a_nan_in_the_narrative_restrictions.');
```

colospace.m

```

1  function varargout = colospace(Conversion,varargin)
2  %COLORSPACE Transform a color image between color representations.
3  % B = COLORSPACE(S,A) transforms the color representation of image A
4  % where S is a string specifying the conversion. The input array A
5  % should be a real full double array of size Mx3 or MxNx3. The output B
6  % is the same size as A.
7  %
8  % S tells the source and destination color spaces, S = 'dest<-src', or
9  % alternatively, S = 'src->dest'. Supported color spaces are
10 %
11 % 'RGB' sRGB IEC 61966-2-1
12 % 'YCbCr' Luma + Chroma ("digitized" version of Y'PbPr)
13 % 'JPEG-YCbCr' Luma + Chroma space used in JFIF JPEG
14 % 'YDbDr' SECAM Y'DbDr Luma + Chroma
15 % 'YPbPr' Luma (ITU-R BT.601) + Chroma
16 % 'YUV' NTSC PAL Y'UV Luma + Chroma
17 % 'YIQ' NTSC Y'IQ Luma + Chroma
18 % 'HSV' or 'HSB' Hue Saturation Value/Brightness
19 % 'HSL' or 'HLS' Hue Saturation Luminance
20 % 'HSI' Hue Saturation Intensity
```

```

21 %      'XYZ'          CIE 1931 XYZ
22 %      'Lab'         CIE 1976 L*a*b* (CIELAB)
23 %      'Luv'         CIE L*u*v* (CIELUV)
24 %      'LCH'         CIE L*C*H* (CIELCH)
25 %      'CAT02 LMS'   CIE CAT02 LMS
26 %
27 % All conversions assume 2 degree observer and D65 illuminant.
28 %
29 % Color space names are case insensitive and spaces are ignored. When
30 % sRGB is the source or destination, it can be omitted. For example
31 % 'yuv<-' is short for 'yuv<-rgb'.
32 %
33 % For sRGB, the values should be scaled between 0 and 1. Beware that
34 % transformations generally do not constrain colors to be "in gamut."
35 % Particularly, transforming from another space to sRGB may obtain
36 % R'G'B' values outside of the [0,1] range. So the result should be
37 % clamped to [0,1] before displaying:
38 %     image(min(max(B,0),1)); % Clamp B to [0,1] and display
39 %
40 % sRGB (Red Green Blue) is the (ITU-R BT.709 gamma-corrected) standard
41 % red-green-blue representation of colors used in digital imaging. The
42 % components should be scaled between 0 and 1. The space can be
43 % visualized geometrically as a cube.
44 %
45 % Y'PbPr, Y'CbCr, Y'DbDr, Y'UV, and Y'IQ are related to sRGB by linear
46 % transformations. These spaces separate a color into a grayscale
47 % luminance component Y and two chroma components. The valid ranges of
48 % the components depends on the space.
49 %
50 % HSV (Hue Saturation Value) is related to sRGB by
51 %     H = hexagonal hue angle      (0 <= H < 360),
52 %     S = C/V                      (0 <= S <= 1),
53 %     V = max(R',G',B')            (0 <= V <= 1),
54 % where C = max(R',G',B') - min(R',G',B'). The hue angle H is computed on
55 % a hexagon. The space is geometrically a hexagonal cone.
56 %
57 % HSL (Hue Saturation Lightness) is related to sRGB by
58 %     H = hexagonal hue angle      (0 <= H < 360),
59 %     S = C/(1 - |2L-1|)           (0 <= S <= 1),
60 %     L = (max(R',G',B') + min(R',G',B'))/2 (0 <= L <= 1),
61 % where H and C are the same as in HSV. Geometrically, the space is a
62 % double hexagonal cone.
63 %
64 % HSI (Hue Saturation Intensity) is related to sRGB by
65 %     H = polar hue angle          (0 <= H < 360),
66 %     S = 1 - min(R',G',B')/I      (0 <= S <= 1),
67 %     I = (R'+G'+B')/3             (0 <= I <= 1).

```

```

68 % Unlike HSV and HSL, the hue angle H is computed on a circle rather than
69 % a hexagon.
70 %
71 % CIE XYZ is related to sRGB by inverse gamma correction followed by a
72 % linear transform. Other CIE color spaces are defined relative to XYZ.
73 %
74 % CIE L*a*b*, L*u*v*, and L*C*H* are nonlinear functions of XYZ. The L*
75 % component is designed to match closely with human perception of
76 % lightness. The other two components describe the chroma.
77 %
78 % CIE CAT02 LMS is the linear transformation of XYZ using the MCAT02
79 % chromatic adaptation matrix. The space is designed to model the
80 % response of the three types of cones in the human eye, where L, M, S,
81 % correspond respectively to red ("long"), green ("medium"), and blue
82 % ("short").
83
84 % Pascal Getreuer 2005–2010
85
86
87 %%% Input parsing %%%
88 if nargin < 2, error('Not enough input arguments.');
```

end

```

89 [SrcSpace, DestSpace] = parse(Conversion);
90
91 if nargin == 2
92     Image = varargin{1};
93 elseif nargin >= 3
94     Image = cat(3, varargin{:});
95 else
96     error('Invalid number of input arguments.');
```

end

```

97
98
99 FlipDims = (size(Image, 3) == 1);
100
101 if FlipDims, Image = permute(Image, [1, 3, 2]); end
102 if ~isa(Image, 'double'), Image = double(Image)/255; end
103 if size(Image, 3) ~= 3, error('Invalid input size.');
```

end

```

104
105 SrcT = gettransform(SrcSpace);
106 DestT = gettransform(DestSpace);
107
108 if ~ischar(SrcT) && ~ischar(DestT)
109     % Both source and destination transforms are affine, so they
110     % can be composed into one affine operation
111     T = [DestT(:, 1:3)*SrcT(:, 1:3), DestT(:, 1:3)*SrcT(:, 4)+DestT(:, 4)];
112     Temp = zeros(size(Image));
113     Temp(:, :, 1) = T(1)*Image(:, :, 1) + T(4)*Image(:, :, 2) + T(7)*Image(:, :, 3) + T
        (10);
```

```

114     Temp(:,:,2) = T(2)*Image(:,:,1) + T(5)*Image(:,:,2) + T(8)*Image(:,:,3) + T
        (11);
115     Temp(:,:,3) = T(3)*Image(:,:,1) + T(6)*Image(:,:,2) + T(9)*Image(:,:,3) + T
        (12);
116     Image = Temp;
117 elseif ~ischar(DestT)
118     Image = rgb(Image,SrcSpace);
119     Temp = zeros(size(Image));
120     Temp(:,:,1) = DestT(1)*Image(:,:,1) + DestT(4)*Image(:,:,2) + DestT(7)*
        Image(:,:,3) + DestT(10);
121     Temp(:,:,2) = DestT(2)*Image(:,:,1) + DestT(5)*Image(:,:,2) + DestT(8)*
        Image(:,:,3) + DestT(11);
122     Temp(:,:,3) = DestT(3)*Image(:,:,1) + DestT(6)*Image(:,:,2) + DestT(9)*
        Image(:,:,3) + DestT(12);
123     Image = Temp;
124 else
125     Image = feval(DestT,Image,SrcSpace);
126 end
127
128 %%% Output format %%%
129 if nargout > 1
130     varargout = {Image(:,:,1),Image(:,:,2),Image(:,:,3)};
131 else
132     if FlipDims, Image = permute(Image,[1,3,2]); end
133     varargout = {Image};
134 end
135
136 return;
137
138
139 function [SrcSpace, DestSpace] = parse(Str)
140 % Parse conversion argument
141
142 if ischar(Str)
143     Str = lower(strrep(strrep(Str, '—', ''), '=', ''));
144     k = find(Str == '>');
145
146     if length(k) == 1 % Interpret the form 'src->dest'
147         SrcSpace = Str(1:k-1);
148         DestSpace = Str(k+1:end);
149     else
150         k = find(Str == '<');
151
152         if length(k) == 1 % Interpret the form 'dest<-src'
153             DestSpace = Str(1:k-1);
154             SrcSpace = Str(k+1:end);
155         else

```

```

156         error(['Invalid conversion, ', Str, '']);
157     end
158 end
159
160 SrcSpace = alias(SrcSpace);
161 DestSpace = alias(DestSpace);
162 else
163     SrcSpace = 1; % No source pre-transform
164     DestSpace = Conversion;
165     if any(size(Conversion) ~= 3), error('Transformation matrix must be 3x3.');
```

end

```

166 end
167 return;
168
169
170 function Space = alias(Space)
171 Space = strrep(strrep(Space, 'cie', ''), '_ ', '');
172
173 if isempty(Space)
174     Space = 'rgb';
175 end
176
177 switch Space
178 case {'ycbcr', 'ycc'}
179     Space = 'ycbcr';
180 case {'hsv', 'hsb'}
181     Space = 'hsv';
182 case {'hsl', 'hsi', 'hls'}
183     Space = 'hsl';
184 case {'rgb', 'yuv', 'yiq', 'ydbdr', 'ycbcr', 'jpegycbcr', 'xyz', 'lab', 'luv', 'lch'}
185     return;
186 end
187 return;
188
189
190 function T = gettransform(Space)
191 % Get a colorspace transform: either a matrix describing an affine transform,
192 % or a string referring to a conversion subroutine
193 switch Space
194 case 'ypbpr'
195     T =
196         [0.299, 0.587, 0.114, 0; -0.1687367, -0.331264, 0.5, 0; 0.5, -0.418688, -0.081312, 0];
197
198 case 'yuv'
199     % sRGB to NTSC/PAL YUV
200     % Wikipedia: http://en.wikipedia.org/wiki/YUV
201     T = [0.299, 0.587, 0.114, 0; -0.147, -0.289, 0.436, 0; 0.615, -0.515, -0.100, 0];

```

```

200 case 'ydbdr'
201     % sRGB to SECAM YDbDr
202     % Wikipedia: http://en.wikipedia.org/wiki/YDbDr
203     T = [0.299,0.587,0.114,0;-0.450,-0.883,1.333,0;-1.333,1.116,0.217,0];
204 case 'yiq'
205     % sRGB in [0,1] to NTSC YIQ in
206     % Wikipedia: http://en.wikipedia.org/wiki/YIQ
207     T =
208         [0.299,0.587,0.114,0;0.595716,-0.274453,-0.321263,0;0.211456,-0.522591,0.311135,0];

208 case 'ycbcr'
209     % sRGB (range [0,1]) to ITU-R BRT.601 (CCIR 601) Y'CbCr
210     % Wikipedia: http://en.wikipedia.org/wiki/YCbCr
211     % Poynton, Equation 3, scaling of R'G'B to Y'PbPr conversion
212     T =
213         [65.481,128.553,24.966,16;-37.797,-74.203,112.0,128;112.0,-93.786,-18.214,128];

213 case 'jpegycbcr'
214     % Wikipedia: http://en.wikipedia.org/wiki/YCbCr
215     T =
216         [0.299,0.587,0.114,0;-0.168736,-0.331264,0.5,0.5;0.5,-0.418688,-0.081312,0.5]*255;

216 case {'rgb','xyz','hsv','hsl','lab','luv','lch','cat02lms'}
217     T = Space;
218 otherwise
219     error(['Unknown color space, ',Space,'']);
220 end
221 return;
222
223
224 function Image = rgb(Image,SrcSpace)
225 % Convert to sRGB from 'SrcSpace'
226 switch SrcSpace
227 case 'rgb'
228     return;
229 case 'hsv'
230     % Convert HSV to sRGB
231     Image = huetorgb((1 - Image(:,:,2)).*Image(:,:,3),Image(:,:,3),Image(:,:,1)
232         );
233 case 'hsl'
234     % Convert HSL to sRGB
235     L = Image(:,:,3);
236     Delta = Image(:,:,2).*min(L,1-L);
237     Image = huetorgb(L-Delta,L+Delta,Image(:,:,1));
238 case {'xyz','lab','luv','lch','cat02lms'}
239     % Convert to CIE XYZ

```



```

239     Image = xyz(Image,SrcSpace);
240     % Convert XYZ to RGB
241     T = [3.2406, -1.5372, -0.4986; -0.9689, 1.8758, 0.0415; 0.0557, -0.2040,
          1.057];
242     R = T(1)*Image(:,:,1) + T(4)*Image(:,:,2) + T(7)*Image(:,:,3); % R
243     G = T(2)*Image(:,:,1) + T(5)*Image(:,:,2) + T(8)*Image(:,:,3); % G
244     B = T(3)*Image(:,:,1) + T(6)*Image(:,:,2) + T(9)*Image(:,:,3); % B
245     % Desaturate and rescale to constrain resulting RGB values to [0,1]
246     AddWhite = -min(min(min(R,G),B),0);
247     R = R + AddWhite;
248     G = G + AddWhite;
249     B = B + AddWhite;
250     % Apply gamma correction to convert linear RGB to sRGB
251     Image(:,:,1) = gammacorrection(R); % R'
252     Image(:,:,2) = gammacorrection(G); % G'
253     Image(:,:,3) = gammacorrection(B); % B'
254 otherwise % Conversion is through an affine transform
255     T = gettransform(SrcSpace);
256     temp = inv(T(:,1:3));
257     T = [temp,-temp*T(:,4)];
258     R = T(1)*Image(:,:,1) + T(4)*Image(:,:,2) + T(7)*Image(:,:,3) + T(10);
259     G = T(2)*Image(:,:,1) + T(5)*Image(:,:,2) + T(8)*Image(:,:,3) + T(11);
260     B = T(3)*Image(:,:,1) + T(6)*Image(:,:,2) + T(9)*Image(:,:,3) + T(12);
261     Image(:,:,1) = R;
262     Image(:,:,2) = G;
263     Image(:,:,3) = B;
264 end
265
266 % Clip to [0,1]
267 Image = min(max(Image,0),1);
268 return;
269
270
271 function Image = xyz(Image,SrcSpace)
272 % Convert to CIE XYZ from 'SrcSpace'
273 WhitePoint = [0.950456,1,1.088754];
274
275 switch SrcSpace
276 case 'xyz'
277     return;
278 case 'luv'
279     % Convert CIE L*uv to XYZ
280     WhitePointU = (4*WhitePoint(1))./(WhitePoint(1) + 15*WhitePoint(2) + 3*
          WhitePoint(3));
281     WhitePointV = (9*WhitePoint(2))./(WhitePoint(1) + 15*WhitePoint(2) + 3*
          WhitePoint(3));
282     L = Image(:,:,1);

```

```

283     Y = (L + 16)/116;
284     Y = invf(Y)*WhitePoint(2);
285     U = Image(:,:,2)./(13*L + 1e-6*(L==0)) + WhitePointU;
286     V = Image(:,:,3)./(13*L + 1e-6*(L==0)) + WhitePointV;
287     Image(:,:,1) = -(9*Y.*U)./((U-4).*V - U.*V); % X
288     Image(:,:,2) = Y; % Y
289     Image(:,:,3) = (9*Y - (15*V.*Y) - (V.*Image(:,:,1)))./(3*V); % Z
290     case {'lab','lch'}
291         Image = lab(Image,SrcSpace);
292         % Convert CIE L*a*b to XYZ
293         fY = (Image(:,:,1) + 16)/116;
294         fX = fY + Image(:,:,2)/500;
295         fZ = fY - Image(:,:,3)/200;
296         Image(:,:,1) = WhitePoint(1)*invf(fX); % X
297         Image(:,:,2) = WhitePoint(2)*invf(fY); % Y
298         Image(:,:,3) = WhitePoint(3)*invf(fZ); % Z
299     case 'cat02lms'
300         % Convert CAT02 LMS to XYZ
301         T = inv([0.7328, 0.4296, -0.1624; -0.7036, 1.6975, 0.0061; 0.0030, 0.0136,
302                 0.9834]);
303         L = Image(:,:,1);
304         M = Image(:,:,2);
305         S = Image(:,:,3);
306         Image(:,:,1) = T(1)*L + T(4)*M + T(7)*S; % X
307         Image(:,:,2) = T(2)*L + T(5)*M + T(8)*S; % Y
308         Image(:,:,3) = T(3)*L + T(6)*M + T(9)*S; % Z
309     otherwise % Convert from some gamma-corrected space
310         % Convert to sRGB
311         Image = rgb(Image,SrcSpace);
312         % Undo gamma correction
313         R = invgammacorrection(Image(:,:,1));
314         G = invgammacorrection(Image(:,:,2));
315         B = invgammacorrection(Image(:,:,3));
316         % Convert RGB to XYZ
317         T = inv([3.2406, -1.5372, -0.4986; -0.9689, 1.8758, 0.0415; 0.0557,
318                 -0.2040, 1.057]);
319         Image(:,:,1) = T(1)*R + T(4)*G + T(7)*B; % X
320         Image(:,:,2) = T(2)*R + T(5)*G + T(8)*B; % Y
321         Image(:,:,3) = T(3)*R + T(6)*G + T(9)*B; % Z
322     end
323     return;
324
325 function Image = hsv(Image,SrcSpace)
326 % Convert to HSV
327 Image = rgb(Image,SrcSpace);
328 V = max(Image,[],3);

```

```

328 S = (V - min(Image,[],3))./(V + (V == 0));
329 Image(:,:,1) = rgbtohue(Image);
330 Image(:,:,2) = S;
331 Image(:,:,3) = V;
332 return;
333
334
335 function Image = hsl(Image,SrcSpace)
336 % Convert to HSL
337 switch SrcSpace
338 case 'hsv'
339     % Convert HSV to HSL
340     MaxVal = Image(:,:,3);
341     MinVal = (1 - Image(:,:,2)).*MaxVal;
342     L = 0.5*(MaxVal + MinVal);
343     temp = min(L,1-L);
344     Image(:,:,2) = 0.5*(MaxVal - MinVal)./(temp + (temp == 0));
345     Image(:,:,3) = L;
346 otherwise
347     Image = rgb(Image,SrcSpace); % Convert to sRGB
348     % Convert sRGB to HSL
349     MinVal = min(Image,[],3);
350     MaxVal = max(Image,[],3);
351     L = 0.5*(MaxVal + MinVal);
352     temp = min(L,1-L);
353     S = 0.5*(MaxVal - MinVal)./(temp + (temp == 0));
354     Image(:,:,1) = rgbtohue(Image);
355     Image(:,:,2) = S;
356     Image(:,:,3) = L;
357 end
358 return;
359
360
361 function Image = lab(Image,SrcSpace)
362 % Convert to CIE L*a*b* (CIELAB)
363 WhitePoint = [0.950456,1,1.088754];
364
365 switch SrcSpace
366 case 'lab'
367     return;
368 case 'lch'
369     % Convert CIE L*CH to CIE L*ab
370     C = Image(:,:,2);
371     Image(:,:,2) = cos(Image(:,:,3)*pi/180).*C; % a*
372     Image(:,:,3) = sin(Image(:,:,3)*pi/180).*C; % b*
373 otherwise
374     Image = xyz(Image,SrcSpace); % Convert to XYZ

```

```

375     % Convert XYZ to CIE L*a*b*
376     X = Image(:,:,1)/WhitePoint(1);
377     Y = Image(:,:,2)/WhitePoint(2);
378     Z = Image(:,:,3)/WhitePoint(3);
379     fX = f(X);
380     fY = f(Y);
381     fZ = f(Z);
382     Image(:,:,1) = 116*fY - 16;      % L*
383     Image(:,:,2) = 500*(fX - fY);    % a*
384     Image(:,:,3) = 200*(fY - fZ);    % b*
385 end
386 return;
387
388
389 function Image = luv(Image,SrcSpace)
390 % Convert to CIE L*u*v* (CIELUV)
391 WhitePoint = [0.950456,1,1.088754];
392 WhitePointU = (4*WhitePoint(1))./(WhitePoint(1) + 15*WhitePoint(2) + 3*
    WhitePoint(3));
393 WhitePointV = (9*WhitePoint(2))./(WhitePoint(1) + 15*WhitePoint(2) + 3*
    WhitePoint(3));
394
395 Image = xyz(Image,SrcSpace); % Convert to XYZ
396 Denom = Image(:,:,1) + 15*Image(:,:,2) + 3*Image(:,:,3);
397 U = (4*Image(:,:,1))./(Denom + (Denom == 0));
398 V = (9*Image(:,:,2))./(Denom + (Denom == 0));
399 Y = Image(:,:,2)/WhitePoint(2);
400 L = 116*f(Y) - 16;
401 Image(:,:,1) = L;                      % L*
402 Image(:,:,2) = 13*L.*(U - WhitePointU); % u*
403 Image(:,:,3) = 13*L.*(V - WhitePointV); % v*
404 return;
405
406
407 function Image = lch(Image,SrcSpace)
408 % Convert to CIE L*ch
409 Image = lab(Image,SrcSpace); % Convert to CIE L*ab
410 H = atan2(Image(:,:,3),Image(:,:,2));
411 H = H*180/pi + 360*(H < 0);
412 Image(:,:,2) = sqrt(Image(:,:,2).^2 + Image(:,:,3).^2); % C
413 Image(:,:,3) = H;                                     % H
414 return;
415
416
417 function Image = cat02lms(Image,SrcSpace)
418 % Convert to CAT02 LMS
419 Image = xyz(Image,SrcSpace);

```

```

420 T = [0.7328, 0.4296, -0.1624;-0.7036, 1.6975, 0.0061; 0.0030, 0.0136, 0.9834];
421 X = Image(:,:,1);
422 Y = Image(:,:,2);
423 Z = Image(:,:,3);
424 Image(:,:,1) = T(1)*X + T(4)*Y + T(7)*Z; % L
425 Image(:,:,2) = T(2)*X + T(5)*Y + T(8)*Z; % M
426 Image(:,:,3) = T(3)*X + T(6)*Y + T(9)*Z; % S
427 return;
428
429
430 function Image = huetorgb(m0,m2,H)
431 % Convert HSV or HSL hue to RGB
432 N = size(H);
433 H = min(max(H(:),0),360)/60;
434 m0 = m0(:);
435 m2 = m2(:);
436 F = H - round(H/2)*2;
437 M = [m0, m0 + (m2-m0).*abs(F), m2];
438 Num = length(m0);
439 j = [2 1 0;1 2 0;0 2 1;0 1 2;1 0 2;2 0 1;2 1 0]*Num;
440 k = floor(H) + 1;
441 Image = reshape([M(j(k,1)+(1:Num).'),M(j(k,2)+(1:Num).'),M(j(k,3)+(1:Num).')]
    , [N,3]);
442 return;
443
444
445 function H = rgbtohue(Image)
446 % Convert RGB to HSV or HSL hue
447 [M,i] = sort(Image,3);
448 i = i(:,:,3);
449 Delta = M(:,:,3) - M(:,:,1);
450 Delta = Delta + (Delta == 0);
451 R = Image(:,:,1);
452 G = Image(:,:,2);
453 B = Image(:,:,3);
454 H = zeros(size(R));
455 k = (i == 1);
456 H(k) = (G(k) - B(k))./Delta(k);
457 k = (i == 2);
458 H(k) = 2 + (B(k) - R(k))./Delta(k);
459 k = (i == 3);
460 H(k) = 4 + (R(k) - G(k))./Delta(k);
461 H = 60*H + 360*(H < 0);
462 H(Delta == 0) = nan;
463 return;
464
465

```

```

466 function Rp = gammacorrection(R)
467 Rp = zeros(size(R));
468 i = (R <= 0.0031306684425005883);
469 Rp(i) = 12.92*R(i);
470 Rp(~i) = real(1.055*R(~i).^0.41666666666666667 - 0.055);
471 return;
472
473
474 function R = invgammacorrection(Rp)
475 R = zeros(size(Rp));
476 i = (Rp <= 0.0404482362771076);
477 R(i) = Rp(i)/12.92;
478 R(~i) = real(((Rp(~i) + 0.055)/1.055).^2.4);
479 return;
480
481
482 function fY = f(Y)
483 fY = real(Y.^(1/3));
484 i = (Y < 0.008856);
485 fY(i) = Y(i)*(841/108) + (4/29);
486 return;
487
488
489 function Y = invf(fY)
490 Y = fY.^3;
491 i = (Y < 0.008856);
492 Y(i) = (fY(i) - 4/29)*(108/841);
493 return;

```

demean.m

```

1 function [x]=demean(y)
2 T=size(y,1);
3 my= repmat(nanmean(y),T,1);
4 x=(y-my);

```

directmethods.m

```

1 function [dm] = directmethods(y,lags,options)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
5 % 'directmethods' generates local projeciton IRF and direct forecasts

```

```

6
7 % Core Inputs :
8 % — y, data columns variables
9 % — lags, lag order of the VAR
10
11 % Additonal Inputs collected options: see below
12
13 % Filippo Ferroni, 27/02/2020
14 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
15 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
16
17 if nargin < 2
18     error('the_BVAR_toolbox_needs_at_least_two_inputs:_data_and_number_of_lags
19         ');
19 end
20 if lags < 1
21     error('lags_cannot_be_zero_or_negative');
22 end
23 % number of observable variables
24 [T,ny] = size(y);
25
26 %*****
27 %* DEFAULT SETTINGS
28 %*****
29 % Control random number generator
30 if isOctave == 0
31     isMatlab = 1;
32     rng('default');
33     rng(999);
34 else
35     isMatlab = 0;
36     % pkg load optim
37     pkg load statistics
38     randn('state',999);
39     rand('state',999);
40 end
41
42 hor = 24;
43 conf_sig = 0.9;
44 controls_ = 0;
45 robust_se_ = 1; % by default robust SE
46 proxy_ = 0;
47 noconstant = 0;
48 ns = 0;
49 Q = eye(ny);
50 dummy = 0;
51 K = 5000;

```

```

52 max_prior_tau_ = 0;
53 lb             = -1e10;
54 ub             = 1e10;
55
56
57 %*****
58 %* CUSTOMIZED SETTINGS
59 %*****
60 if nargin>2
61     %=====
62     % Various options
63     %=====
64     if isfield(options,'hor')==1 % horizon of IRF and Forecasts
65         hor = options.hor;
66     end
67     if isfield(options,'conf_sig')==1 % CI of OLS estimation of LP and DF
68         conf_sig = options.conf_sig;
69     end
70     if isfield(options,'Q')==1 % orthonormal matrix
71         Q = options.Q;
72     end
73     if isfield(options,'K')==1 % # of draws for the BLP
74         K = options.K;
75     end
76     %=====
77     % options: adding controls
78     %=====
79     if isfield(options,'controls')==1
80         controls_ = 1;
81         controls = options.controls;
82         if T~=size(controls,1)
83             error('Control variables and observables must have the same time length')
84         end
85     end
86     %=====
87     % options: adding proxy variable for identification
88     %=====
89     if isfield(options,'proxy')==1
90         proxy_ = 1;
91         proxy = options.proxy;
92         ns = size(proxy,2);
93         if T~=size(proxy,1)
94             error('Shocks proxies and observables must have the same time length')
95         end
96     end

```



```

97 %=====
98 % options: Computing Robust Standard Errors
99 %=====
100 if isfield(options,'robust_se_')== 1
101     robust_se_ = options.robust_se_;
102     % robust_se_ = 0      unadjusted SE
103     % robust_se_ = 1      NW Robust SE:  Hamilton (1994), Ch 10 pag 282, eq
104                             (10.5.20)
105     % robust_se_ = 5      Matlab HAC function: need Matlab Econ Toolbox
106 end
107 %=====
108 % Conjugate/Hierachical prior options
109 %=====
110 if (isfield(options,'priors')==1 && strcmp(options.priors.name,'Conjugate'
111     )==1) || (isfield(options,'priors')==1 && strcmp(options.priors.name,'
112     conjugate')==1) || ...
113     (isfield(options,'prior')==1 && strcmp(options.prior.name,'
114         Conjugate')==1) || (isfield(options,'prior')==1 && strcmp(
115             options.prior.name,'conjugate')==1)
116
117     dummy = 2;
118     prior.name= 'Conjugate';
119     % Priors for the AR parameters
120     if isfield(options.priors,'Phi')== 1
121         % mean
122         if isfield(options.priors.Phi,'mean')== 1
123             prior.Phi.mean = options.priors.Phi.mean;
124             if size(prior.Phi.mean) ~= [ny*lags+(1-noconstant)    ny]
125                 error('Size_mismatch')
126             end
127         else
128             warning(['You_did_not_provide_a_prior_mean_for_the_AR_coeff._'
129                 ...
130                 'Assume_zeros_everywhere.'])
131             prior.Phi.mean = zeros(ny*lags+(1-noconstant) , ny);
132         end
133         % variance
134         if isfield(options.priors.Phi,'cov')== 1
135             prior.Phi.cov = options.priors.Phi.cov;
136             if length(prior.Phi.cov) ~= (ny*lags+(1-noconstant) ) || size(
137                 prior.Phi.cov,1 )~=size(prior.Phi.cov,2)
138                 error('Size_mismatch:_Covariance_Phi_should_be_square,_e.g
139                     _size(Phi.mean,1)x_size(Phi.mean,1)x')
140             end
141         else
142             warning(['You_did_not_provide_a_Covariance_for_the_AR_coeff._'
143                 ...

```

```

135         'Assume_10_times_Identity_Matrix.'])
136     prior.Phi.cov = 10 * eye((ny*lags+(1-noconstant) ));
137 end
138 else
139     warning(['You did not provide prior mean and covariance for the AR
140             _coeff_ ...
141             'Assume_zeros_everywhere_with_covariance_10_times_Identity_
142             Matrix.'])
143     %           prior.Phi.cov = 10 * eye((ny*lags+(1-noconstant) +
144             timetrend ) * ny);
145     prior.Phi.cov = 10 * eye((ny*lags+(1-noconstant) ));
146     prior.Phi.mean = zeros(ny*lags+(1-noconstant) , ny);
147 end
148 % Priors for the Residual Covariance
149 if isfield(options.priors,'Sigma') == 1
150     % scale
151     if isfield(options.priors.Sigma,'scale') == 1
152         prior.Sigma.scale = options.priors.Sigma.scale;
153         if size(prior.Sigma.scale) ~= [ny ny]
154             error('Size_mismatch')
155         end
156     else
157         warning(['You did not provide a prior mean for the Residual
158                 Covariance. _' ...
159                 'Assume_identity_matrix.'])
160         prior.Sigma.scale = eye(ny);
161     end
162     % degrees of freedom
163     if isfield(options.priors.Sigma,'df') == 1
164         prior.Sigma.df = options.priors.Sigma.df;
165         if length(prior.Sigma.df) ~= 1
166             error('Size_mismatch')
167         end
168         if prior.Sigma.df < ny
169             error('Too_few_degrees_of_freedom_-_prior_on_variance_
170                 residuals')
171         end
172     else
173         warning(['You did not provide the degrees of freedom for the
174                 Residual_Covariance. _' ...
175                 'Assume_ny+1.'])
176         prior.Sigma.df = ny + 1;
177     end
178 end
179 else
180     warning(['You did not provide prior mean and variance for the
181             Residual_Covariance. _' ...
182             'Assume_an_identity_matrix_matrix_with_N+1_degrees_of_freedom.

```

```

        '])
175         prior.Sigma.scale = eye(ny);
176         prior.Sigma.df     = ny + 1;
177     end
178     if isfield(options.priors,'tau') == 1 && isnumeric(options.priors.tau)
        == 1
179         prior.tau = options.priors.tau;
180         if length(options.priors.tau) ~= hor
181             error('tau must be a vector of size hor')
182         end
183     elseif isfield(options.priors,'max_tau') == 1 && options.priors.
        max_tau ==1
184         max_prior_tau_ = 1;
185         max_compute     = 3;
186         prior.tau = ones(hor,1);
187     else
188         prior.tau = ones(hor,1);
189     end
190 end
191 % options for the maximization
192 if isfield(options,'max_compute') == 1
193     max_compute = options.max_compute;
194 end
195 if isfield(options,'ub') == 1
196     ub = options.ub;
197     if length(ub) ~= length(prior.tau)
198         error('Mismatch between the size of upper bounds and the param_
            vector');
199     end
200 end
201 if isfield(options,'lb') == 1
202     lb = options.lb;
203     if length(lb) ~= length(prior.tau)
204         error('Mismatch between the size of lower bounds and the param_
            vector');
205     end
206 end
207 end
208
209 % Confidence Interval Points and Index for confidence serts
210 alpha = 1 - conf_sig;
211 talpha = abs(tinv(alpha/2,T-ny-1));
212 sort_idx = round((0.5 + [-conf_sig, conf_sig, 0]/2) * K);
213
214 %*****
215 % Construct the RHS matrix
216 %*****

```

```

217
218 X1 = lagX(y,1:lags);
219 positions_nylags = 1 : size(X1,2);
220 nylags           = length(positions_nylags);
221
222 % forecast launching point
223 fdata_initval    = lagX(y(end-lags+1:end, :),0:lags-1);
224 fdata_initval(1:end-1,:) = [];
225 % add controls if any
226 if controls_ == 1
227     a = size(X1,2) + 1;
228     X1 = [X1 lagX(controls,1:lags)];
229     b = size(X1,2);
230     position_controls = a:b;
231     % forecast launching point
232     lastvalc = lagX(controls(end-lags+1:end, :),0:lags-1);
233     fdata_initval = [fdata_initval, controls(end,:)];
234
235 else
236     position_controls = [];
237 end
238 % add proxy shocks if any
239 if proxy_ == 1
240     a = size(X1,2) + 1;
241     X1 = [X1 proxy];
242     b = size(X1,2);
243     position_proxy = a:b ;
244     % forecast launching point (assume shocks are zero)
245     fdata_initval = [fdata_initval, zeros(1,ns)];
246 else
247     position_proxy = [];
248 end
249 X_ = X1;
250 if noconstant == 0
251     X_ = [X1 ones(T,1)];
252     position_constant = size(X_,2);
253     % forecast launching point
254     fdata_initval = [fdata_initval, 1];
255 else
256     position_constant = [];
257 end
258 nx = 1 - noconstant;
259 % other checks
260 if (lags + hor) >= size(X_,1)
261     error('More_parameters_than_observations:_consider_reducing_'_hor''_or_'_'_lags'')
262 end

```

```

263
264
265 %*****
266 % pre allocation
267 %*****
268 Omegaproxy = eye(ny);
269 Omega      = eye(ny);
270 ir_lp      = nan(ny,hor+1,ny,3); % variable , horizon , shock and mean upper
      lower
271 irproxy_lp = nan(ny,hor+1,1,3);
272 forecasts  = nan(hor,ny,3);
273 if dummy == 2
274     ir_blp      = nan(ny,hor+1,ny,K);
275     irproxy_blp = nan(ny,hor+1,1,K);
276     bforecasts_no_shocks = nan(hor,ny,K);
277     bforecasts_with_shocks = nan(hor,ny,K);
278     log_dnsty      = nan(hor,1);
279 end
280
281 %*****
282 %* Computing the LP and DF
283 %*****
284 wb = waitbar(0, 'Direct_Methods');
285 for hh = 0 : hor % iteration over horizon
286     ytmp = lagX(y, -hh);
287     % Reduced Form estimations
288     if robust_se_ ~= 0 % Robust SE
289         % options.robust_se_ = robust_se_;
290         options.L      = lags + hh + 1;
291         olsreg_(hh+1) = ols_reg(ytmp, X_, options);
292     else
293         olsreg_(hh+1) = ols_reg(ytmp, X_);
294     end
295     % Proxy IV identifications
296     if proxy_
297         irproxy_lp(:, hh+1, :, 2) = olsreg_(hh+1).beta(position_proxy, :)' ; %
            mean
298         irproxy_lp(:, hh+1, :, 3) = irproxy_lp(:, hh+1, :, 2) + talpha *
            olsreg_(hh+1).se(position_proxy, :)' ; % upper
299         irproxy_lp(:, hh+1, :, 1) = irproxy_lp(:, hh+1, :, 2) - talpha *
            olsreg_(hh+1).se(position_proxy, :)' ; % lower
300         if hh == 0
301             Omegaproxy(:,1) = olsreg_(hh+1).beta(position_proxy, :)' ;
302             Omegasproxy(:,1,:) = repmat(Omegaproxy(:,1), 1, 1, K) + ...
            olsreg_(hh+1).se(position_proxy, :)' .* randn(ny,1,K);
303         end
304     end
305 end

```

```

306 % Choleski/Rotated identification
307 if hh == 0
308     Omega = chol(olsreg_(hh+1).Serror,'Lower') * Q;
309     % generate uncertainty on S (flat prior)
310     Sbar = olsreg_(hh+1).error'*olsreg_(hh+1).error;
311     df = olsreg_(hh+1).N - olsreg_(hh+1).K;
312     [~, Omegas] = generateOmegas(Sbar,df,K,Q);
313     Omegasort = sort(Omegas,3);
314
315     ir_lp(:, hh+1, :, 2) = Omega; % mean
316     ir_lp(:, hh+1, :, 3) = Omegasort(:, :, sort_idx(2)); % UPPER
317     ir_lp(:, hh+1, :, 1) = Omegasort(:, :, sort_idx(1)); % LOWER
318 else
319     [ir] = iresponse(olsreg_(hh).beta(positions_nylags, :), eye(ny) , 2,
        Omega);
320     ir_lp(:, hh+1, :, 2) = ir(:, 2, :); % mean
321     [ir3] = iresponse(olsreg_(hh).beta(positions_nylags, :) + talpha *
        olsreg_(hh).se(positions_nylags, :), eye(ny) , 2, Omega);
322     ir_lp(:, hh+1, :, 3) = ir3(:, 2, :); % upper
323     [ir1] = iresponse(olsreg_(hh).beta(positions_nylags, :) - talpha *
        olsreg_(hh).se(positions_nylags, :), eye(ny) , 2, Omega);
324     ir_lp(:, hh+1, :, 1) = ir1(:, 2, :); % lower
325 end
326
327 % forecast part
328 forecasts(hh+1, :, 2) = (fdata_initval * olsreg_(hh+1).beta);
329 forecasts(hh+1, :, 3) = forecasts(hh+1, :, 2) + talpha * diag(olsreg_(hh
    +1).Serror)' ;
330 forecasts(hh+1, :, 1) = forecasts(hh+1, :, 2) - talpha * diag(olsreg_(hh
    +1).Serror)' ;
331
332 %=====
333 % Baysian DM
334 if dummy == 2 % activating Bayesian Direct methods.
335     if hh == 0
336         % cholesky IRF on impact
337         ir_blp(:, hh+1, :, :) = Omegas;
338         % proxy IRF on impact
339         if proxy_
340             irproxy_blp(:, hh+1, :, :) = Omegasproxy;
341         end
342         % one-step ahead forecast
343         bforecasts_no_shocks(hh+1, :, :) = repmat(forecasts(hh+1, :, 2)
            ,1,1,K);
344         fnoise = Omega * randn(ny,K);
345         bforecasts_with_shocks(hh+1, :, :) = bforecasts_no_shocks(hh+1, :,
            :) + reshape(fnoise,1,ny,K);

```

```

346
347 % computing the companion matrix
348 F = [prior.Phi.mean(1 : ny * lags, :)'; eye(ny*(lags-1), ny*
      lags)];
349 % constant
350 Fo = [prior.Phi.mean(end, :)'; zeros(ny * (lags-1), 1)];
351 % Shocks Companion
352 G = eye(ny * lags, ny);
353
354 else % hh > 0
355 %*****
356 % Conjugate Prior: MN-IW
357 %*****
358 if max_prior_tau_ == 1 % Maximize the shrinkage on the VAR
      coefficients
359     try
360         disp(['*****'])
361         disp(['*****'])
362         disp(['Optimization_at_horizon_' num2str(hh)])
363         x0 = log(prior.tau(hh));
364         switch max_compute
365             case 1
366                 optim_options = optimset('display','iter','
                     MaxFunEvals',100000,'TolFun',1e-8,'TolX',1e-6)
                     ;
367                 [xh,fh,~,~,~,~] = ...
368                     fminunc('blp-opt-hyperpara',x0,optim_options
                     ,...
369                     hh,prior,olsreg_(hh),F,G,Fo,positions_nylags,
                     position_constant);
370             %
=====
371
372 case 2 % constraint
373 % Set default optimization options for fmincon.
374 optim_options = optimset('display','iter','
      LargeScale','off','MaxFunEvals',100000,'
      TolFun',1e-8,'TolX',1e-6);
375 [xh,fh,~,~,~,~,~] = ...
      fmincon('blp-opt-hyperpara',x0,[],[],[],[],lb(
          hh),ub(hh),[],optim_options,y,lags,options
          );
376 %
=====
377
378 case 3 % Sims
379     crit = 10e-5;

```

```

379         nit = 10e-4;
380         [fh, xh, ~, ~, ~, ~] = ...
381             csminwel('blp_opt_hyperpara',x0,.1*eye(length(
                    x0)),[],crit,nit,hh,prior,olsreg_(hh),F,G,
                    Fo,positions_nylags,position_constant);
382         %
=====

383         case 7 % Matlab's simplex (Optimization toolbox needed
384             ).
385             optim_options = optimset('display','iter','
                    MaxFunEvals',30000,'MaxIter',10000,'TolFun',1e
                    -3,'TolX',1e-3);
386             [xh,fh,~,~] = fminsearch('blp_opt_hyperpara',x0,
                    optim_options,y,lags,options);
387         end
388         fprintf('%s_=%0.5g\n','Hyperparameter_Mode_',exp(xh))
389         fprintf('%s_=%0.5g\n','Marginal Likelihood_',-fh)
390         catch
391             warning('Maximization_NOT_Successful')
392             disp('Using_hyperparameter_default_values')
393             xh = x0;
394         end
395         prior.tau(hh) = exp(xh);
396         disp(['*****'])
397     end
398     % constructing the posterior moments given the (optimal) shrinkage
399     [posterior_, ~] = p2p(hh,prior.tau(hh),prior,olsreg_(hh),F,G,Fo,
400         positions_nylags,position_constant);
401     % computing the marginal likelihood
402     log_dnsty(hh) = blp_ml(prior.tau(hh),hh,prior,olsreg_(hh),F,G,
403         Fo,positions_nylags,position_constant);
404     % Second moments
405     S_inv_upper_chol = chol(inv(posterior_.S));
406     XXi_lower_chol = chol(posterior_.XXi)';
407     % number of regressors
408     nk = nylags + nx;
409     %*****
410     %* Generating draws form the Posterior Distribution
411     %*****
412     for d = 1 : K % Gibbs Sampler
413         %
=====

414         % Inference: Drawing from the posterior distribution
415         % Step 1: draw from the Covariance
416         Sigma = rand_inverse_wishart(ny, posterior_.df,

```



```

414         S_inv_upper_chol);
415         % Step 2: given the Covariance Matrix, draw from the AR
           parameters
416         Sigma_lower_chol = chol(Sigma)';
417         Phi1 = randn(nk * ny, 1);
418         Phi2 = kron(Sigma_lower_chol , Xxi_lower_chol) * Phi1;
419         Phi3 = reshape(Phi2, nk, ny);
420         Phi = Phi3 + posterior_.PhiHat;
421
422         % Step 3: compute IRF
423         blp = iresponse(Phi, eye(ny) , 2, Omega);
424         ir_blp(:, hh+1, :, d) = blp(:, 2, :); % mean
425         if proxy_
426             blpproxy = iresponse(Phi, eye(ny) , 2, Omegaproxy);
427             irproxy_blp(:, hh+1, :, d) = blpproxy(:,2,1);
428         end
429         % Step 4: compute Forecasts
430         bforecasts_no_shocks(hh+1, :, d) = (fdata_initval(1, [
           positions_nylags position_constant])) * Phi);
431         bforecasts_with_shocks(hh+1, :, d) = (fdata_initval(1, [
           positions_nylags position_constant])) * Phi) + (
           Sigma_lower_chol * randn(ny,1))';
432     end
433     end
434     end
435     waitbar(hh/hor, wb);
436 end
437 close(wb);
438
439 %*****
440 %* Storing the results
441 %*****
442 % dm.olsreg_ = olsreg_ ;
443 dm.forecasts = forecasts;
444 dm.ir_lp = ir_lp;
445 dm.irproxy_lp = irproxy_lp;
446 if dummy == 2
447     dm.ir_blp = ir_blp;
448     dm.irproxy_blp = irproxy_blp;
449     dm.bforecasts.no_shocks = bforecasts_no_shocks; % trajectories
           of forecasts without shocks
450     dm.bforecasts.with_shocks = bforecasts_with_shocks; % trajectories
           of forecasts with shocks
451     dm.logmlike = log_dnsty;
452     dm.prior = prior;
453 else

```

```

454     dm.ir_blp                = [];
455     dm.irproxy_blp           = [];
456     dm.bforecasts.no_shocks  = [];
457     dm.bforecasts.with_shocks = [];
458     dm.logmlike               = [];
459     dm.prior                  = [];
460 end
461
462
463
464 %*****
465 %*****
466 %*****
467 %*****
468 function [S,Omegas] = generateOmegas(Sbar,df,K,Q)
469 % generate uncertainty on S
470 ny      = size(Sbar,1);
471 S        = nan(ny,ny,K);
472 Omegas   = nan(ny,ny,K);
473
474 S_inv_upper_chol = chol(inv(Sbar));
475
476 for d = 1:K
477     S(:, :, d)      = rand_inverse_wishart(ny, df, S_inv_upper_chol);
478     Omegas(:, :, d) = chol(S(:, :, d), 'Lower') * Q;
479 end

```

distinguishable_colors.m

```

1 function colors = distinguishable_colors(n_colors,bg,func)
2 % DISTINGUISHABLE-COLORS: pick colors that are maximally perceptually distinct
3 %
4 % When plotting a set of lines, you may want to distinguish them by color.
5 % By default, Matlab chooses a small set of colors and cycles among them,
6 % and so if you have more than a few lines there will be confusion about
7 % which line is which. To fix this problem, one would want to be able to
8 % pick a much larger set of distinct colors, where the number of colors
9 % equals or exceeds the number of lines you want to plot. Because our
10 % ability to distinguish among colors has limits, one should choose these
11 % colors to be "maximally perceptually distinguishable."
12 %
13 % This function generates a set of colors which are distinguishable
14 % by reference to the "Lab" color space, which more closely matches
15 % human color perception than RGB. Given an initial large list of possible
16 % colors, it iteratively chooses the entry in the list that is farthest (in

```

```

17 % Lab space) from all previously-chosen entries. While this "greedy"
18 % algorithm does not yield a global maximum, it is simple and efficient.
19 % Moreover, the sequence of colors is consistent no matter how many you
20 % request, which facilitates the users' ability to learn the color order
21 % and avoids major changes in the appearance of plots when adding or
22 % removing lines.
23 %
24 % Syntax:
25 %     colors = distinguishable_colors(n_colors)
26 % Specify the number of colors you want as a scalar, n_colors. This will
27 % generate an n_colors-by-3 matrix, each row representing an RGB
28 % color triple. If you don't precisely know how many you will need in
29 % advance, there is no harm (other than execution time) in specifying
30 % slightly more than you think you will need.
31 %
32 %     colors = distinguishable_colors(n_colors,bg)
33 % This syntax allows you to specify the background color, to make sure that
34 % your colors are also distinguishable from the background. Default value
35 % is white. bg may be specified as an RGB triple or as one of the standard
36 % "ColorSpec" strings. You can even specify multiple colors:
37 %     bg = {'w','k'}
38 % or
39 %     bg = [1 1 1; 0 0 0]
40 % will only produce colors that are distinguishable from both white and
41 % black.
42 %
43 %     colors = distinguishable_colors(n_colors,bg,rgb2labfunc)
44 % By default, distinguishable_colors uses the image processing toolbox's
45 % color conversion functions makeform and applycform. Alternatively, you
46 % can supply your own color conversion function.
47 %
48 % Example:
49 %     c = distinguishable_colors(25);
50 %     figure
51 %     image(reshape(c,[1 size(c)]))
52 %
53 % Example using the file exchange's 'colspace':
54 %     func = @(x) colorspace('RGB->Lab',x);
55 %     c = distinguishable_colors(25,'w',func);
56
57 % Copyright 2010-2011 by Timothy E. Holy
58
59 % Parse the inputs
60 if (nargin < 2)
61     bg = [1 1 1]; % default white background
62 else
63     if iscell(bg)

```

```

64         % User specified a list of colors as a cell array
65         bgc = bg;
66         for i = 1:length(bgc)
67             bgc{i} = parsecolor(bgc{i});
68         end
69         bg = cat(1,bgc{:});
70     else
71         % User specified a numeric array of colors (n-by-3)
72         bg = parsecolor(bg);
73     end
74 end
75
76 % Generate a sizable number of RGB triples. This represents our space of
77 % possible choices. By starting in RGB space, we ensure that all of the
78 % colors can be generated by the monitor.
79 n_grid = 30; % number of grid divisions along each axis in RGB space
80 x = linspace(0,1,n_grid);
81 [R,G,B] = ndgrid(x,x,x);
82 rgb = [R(:) G(:) B(:)];
83 if (n_colors > size(rgb,1)/3)
84     error('You can''t readily distinguish that many colors');
85 end
86
87 % Convert to Lab color space, which more closely represents human
88 % perception
89 if (nargin > 2)
90     lab = func(rgb);
91     bglab = func(bg);
92 else
93     C = makecform('srgb2lab');
94     lab = applycform(rgb,C);
95     bglab = applycform(bg,C);
96 end
97
98 % If the user specified multiple background colors, compute distances
99 % from the candidate colors to the background colors
100 mindist2 = inf(size(rgb,1),1);
101 for i = 1:size(bglab,1)-1
102     dX = bsxfun(@minus,lab,bglab(i,:)); % displacement all colors from bg
103     dist2 = sum(dX.^2,2); % square distance
104     mindist2 = min(dist2,mindist2); % dist2 to closest previously-chosen
        color
105 end
106
107 % Iteratively pick the color that maximizes the distance to the nearest
108 % already-picked color
109 colors = zeros(n_colors,3);

```

```

110 lastlab = bglab(end,:); % initialize by making the "previous" color equal to
    background
111 for i = 1:n_colors
112     dX = bsxfun(@minus,lab,lastlab); % displacement of last from all colors on
        list
113     dist2 = sum(dX.^2,2); % square distance
114     mindist2 = min(dist2,mindist2); % dist2 to closest previously-chosen
        color
115     [~,index] = max(mindist2); % find the entry farthest from all previously-
        chosen colors
116     colors(i,:) = rgb(index,:); % save for output
117     lastlab = lab(index,:); % prepare for next iteration
118 end
119 end
120
121 function c = parsecolor(s)
122 if ischar(s)
123     c = colorstr2rgb(s);
124 elseif isnumeric(s) && size(s,2) == 3
125     c = s;
126 else
127     error('MATLAB:InvalidColorSpec','Color_specification_cannot_be_parsed.');
```

```

128 end
129 end
130
131 function c = colorstr2rgb(c)
132 % Convert a color string to an RGB value.
133 % This is cribbed from Matlab's whitebg function.
134 % Why don't they make this a stand-alone function?
135 rgbspec = [1 0 0;0 1 0;0 0 1;1 1 1;0 1 1;1 0 1;1 1 0;0 0 0];
136 cspec = 'rgbwcmymk';
137 k = find(cspec==c(1));
138 if isempty(k)
139     error('MATLAB:InvalidColorString','Unknown_color_string.');
```

```

140 end
141 if k~=3 || length(c)==1,
142     c = rgbspec(k,:);
143 elseif length(c)>2,
144     if strcmpi(c(1:3),'bla')
145         c = [0 0 0];
146     elseif strcmpi(c(1:3),'blu')
147         c = [0 0 1];
148     else
149         error('MATLAB:UnknownColorString','Unknown_color_string.');
```

```

150 end
151 end
152 end
```

fetchData.m

```

1
2 %% Fetch Data
3 %Fetch data sets up a function so that Haver Data can be imported directly
4 %to Matlab
5
6 %Inputs:
7 % Database — refers to Haver database such as USECON, USNA, or CAPSTOCK; USNA
8 % most commonly used to keep consistency with Jeff's previous chain
9 % aggregation code
10 % and other BEA numbers.
11 % seriesName — used for the actual names coming from Haver
12 % startDate and endDate are previously chosen for the range of data you'd like
13 % to pull
14
15 function [data,timev]=fetchData(database,seriesName,startDate,endDate,
16 frequency)
17
18     D=fetch(database,seriesName,startDate,endDate,frequency);
19     data=D(:,2);
20     timev =D(:,1);
21 end

```

fevd.m

```

1 function FEVD = fevd(hor,Phi,Sigma,Omega)
2
3 % computes the forecast error variance decomposition
4
5 N          = size(Sigma,1);
6 [m , k]    = size(Phi);
7 lags       = (m-1)/k;
8
9 if nargin < 4
10     Omega = eye(N);
11 end
12
13
14 % companion
15 F          = [Phi(1 : N * lags, :)'; eye(N*(lags-1), N*lags)];
16 G          = eye(N * lags, N);
17 %C         = [Phi(end, :)'; zeros(N*(lags-1),1)];
18
19 A          = chol(Sigma,'lower');
20 Kappa = G * A * Omega ;

```

```

21
22 tmp_=0;
23 for hh = 1 : hor
24     tmp_ = tmp_ + F^(hh-1) * Kappa * Kappa' * F^(hh-1)';
25 end
26 out_.all_var_ = diag(tmp_(1:N,1:N));
27
28 for sho =1 : N
29     tmp_1 = 0;
30     Ind = zeros(N);
31     Ind(sho,sho) = 1;
32     for hh = 1 : hor
33         tmp_1 = tmp_1 + F^(hh-1) * Kappa * Ind * Kappa' * F^(hh-1)';
34
35     end
36     out_.var_(:,sho) = diag(tmp_1(1:N,1:N));
37 end
38
39 if max(max(abs( sum(out_.var_,2) - out_.all_var_))) > 1e-10
40     error('Something_went_wrong')
41 end
42
43 for indx_sho = 1 : N
44     FEVD(:,indx_sho) = out_.var_(:,indx_sho)./out_.all_var_*100;
45
46 end

```

findP.m

```

1 % Copyright Andrew Binning 2013
2 % Please feel free to use and modify this code as you see if fit. If you
3 % use this code in any academic work, please cite
4 % Andrew Binning, 2013.
5 % "Underidentified SVAR models: A framework for combining short and long-run
6 % restrictions with sign-restrictions,"
7 % Working Paper 2013/14, Norges Bank.
8 function P = findP(C,B,Q,p,k,index)
9 %=====
10 % finds the rotation matrix that satisfies the short and long run
11 % restrictions. Based on Juan F. Rubio-Ramirez & Daniel F. Waggoner & Tao Zha
12 % , 2010.
13 % "Structural Vector Autoregressions: Theory of Identification and
14 % Algorithms for Inference," Review of Economic Studies, Oxford University
15 % Press, vol. 77(2), pages 665-696.
16 %

```

```

15 % inputs:
16 % C = initial short run impact matrix, usually from a cholesky
17 % decomposition of the forecast error variance
18 % B = Matrix of coefficients (including intercept estimates)
19 % Q = A cell containing the linear restrictions for each columnn
20 % p = number of lags
21 % k = number of dependent variables
22 % index = original column ordering in the matrix of restrictions
23 %
24 % outputs:
25 % P = orthogonal rotation matrix
26 %=====
27
28 L0 = C;
29
30 beta_temp = B(2:end,:)';
31
32 beta = zeros(k,k);
33
34 for ii = 1:p
35
36     beta = beta + beta_temp(:,(1:k)+(ii-1)*k);
37
38 end
39
40 Linf = (eye(k)-beta)\C;
41
42 F = [L0;Linf];
43
44 P = zeros(k,k);
45
46 for ii = 1:k
47     if ii == 1
48         Qtilde = Q{ii}*F;
49     else
50         Qtilde = [Q{ii}*F;P'];
51     end
52     [QQ,RR] = qr(Qtilde');
53     P_temp = QQ(:,end);
54     P(:,ii) = P_temp;
55
56 end
57
58 P = P(:,index);

```


findQs.m

```

1  % Copyright Andrew Binning 2013
2  % Please feel free to use and modify this code as you see if fit. If you
3  % use this code in any academic work, please cite
4  % Andrew Binning, 2013.
5  % "Underidentified SVAR models: A framework for combining short and long-run
    restrictions with sign-restrictions,"
6  % Working Paper 2013/14, Norges Bank.
7  function [Q,index,flag] = findQs(k,f)
8  %=====
9  % finds the Q matrices that describe the linear restrictions on the shock
10 % impact matrix. Based on Juan F. Rubio-Ramirez & Daniel F. Waggoner & Tao Zha
    , 2010.
11 % "Structural Vector Autoregressions: Theory of Identification and
12 % Algorithms for Inference," Review of Economic Studies, Oxford University
13 % Press, vol. 77(2), pages 665–696.
14 %
15 % inputs:
16 % k = number of dependent variables
17 % f = matrix of short and long run restrictions
18 %
19 % outputs:
20 % Q = a cell that contains the linear restrictions for each equation
21 % index = the original column order of the matrix of restrictions
22 % flag = indicates whether the model is over, under or exactly identified
23 %=====
24 E = eye(k);
25
26 Q_init = cell(k,2);
27
28 for ii = 1:k
29
30     Q_init{ii,1} = double(diag(f*E(:,ii))==0);
31     Q_init{ii,2} = rank(Q_init{ii,1});
32
33 end
34
35 for ii = 1:k
36
37     temp = Q_init{ii,1};
38     Q_init{ii,1} = temp(logical(sum(temp,2)),:);
39
40 end
41
42 [new,ord] = sort([Q_init{:,2}],2,'descend');
```

```

43
44 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
45 % Check identification
46
47 if any(new - (k - (1:k)) > 0) % over-identified
48     flag = 1;
49 elseif all(new - (k - (1:k)) == 0) % exactly identified
50     flag = 0;
51 elseif any(new - (k - (1:k)) < 0) % under-identified
52     flag = -1;
53 end
54
55 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
56
57 index = nan(k,1);
58
59 for ii = 1:k
60     index(ord(ii)) = ii;
61 end
62
63 Q = cell(k,1);
64
65 for ii = 1:k
66
67     Q{ii} = Q_init{ord(ii),1};
68
69 end

```

forecasts.m

```

1  function [frcst_no_shock,frcst_with_shocks]=forecasts(forecast_data,Phi,Sigma,
2      fhor,lags,EPS)
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'forecasts' computes out-of-sample forecasts
5
6  % Inputs:
7  % - forecast_data, last data
8  % - Phi, AR parameters of the VAR
9  % - Sigma, Covariance matrix of the reduced form VAR shocks
10 % - fhor, horizon of the out-of-sample forecasts
11 % - lags
12 % - EPS, a particular shock realization
13
14 % Output:

```

```

15 % — frcst_no_shock
16 % 1st dimension: horizon
17 % 2nd dimension: variable
18
19 % Filippo Ferroni, 6/1/2015
20 % Revised, 2/15/2017
21 % Revised, 3/21/2018
22 % Revised, 9/11/2019
23 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
24
25 ny = size(Sigma,1);
26 if nargin < 6
27     shock_given = 0;
28 else
29     shock_given = 1;
30 end
31
32 Sigma_lower_chol = chol(Sigma)';
33
34 % preallocating memory
35 frcst_no_shock = nan(fhor,ny);
36 frcst_with_shocks = nan(fhor,ny);
37
38 lags_data = forecast_data.initval;
39 for t = 1 : fhor
40     X = [ reshape(flip(lags_data, 1)', 1, ny*lags) forecast_data.xdata(t, :)
41           ];
42     y = X * Phi;
43     lags_data(1:end-1,:) = lags_data(2:end, :);
44     lags_data(end,:) = y;
45     frcst_no_shock(t, :) = y;
46 end
47
48 % With shocks
49 lags_data = forecast_data.initval;
50 for t = 1 : fhor
51     X = [ reshape(flip(lags_data, 1)', 1, ny*lags) forecast_data.xdata(t, :)
52           ];
53     shock = (Sigma_lower_chol * randn(ny, 1))';
54     if shock_given == 1
55         shock = EPS(t,:);
56     end
57     y = X * Phi + shock;
58     lags_data(1:end-1,:) = lags_data(2:end, :);
59     lags_data(end,:) = y;
60     frcst_with_shocks(t, :) = y;
61 end

```

generateDraw.m

```

1  % Copyright Andrew Binning 2013
2  % Please feel free to use and modify this code as you see if fit. If you
3  % use this code in any academic work, please cite
4  % Andrew Binning, 2013.
5  % "Underidentified SVAR models: A framework for combining short and long-run
    restrictions with sign-restrictions,"
6  % Working Paper 2013/14, Norges Bank.
7  function C = generateDraw(C,k)
8  %=====
9  % Generates a draw that is consistent with the shock variance/covariance
10 % matrix. Based on Juan F. Rubio-Ramirez & Daniel F. Waggoner & Tao Zha, 2010.
11 % "Structural Vector Autoregressions: Theory of Identification and
12 % Algorithms for Inference," Review of Economic Studies, Oxford University
    Press, vol. 77(2), pages 665–696.
13 %
14 % inputs:
15 % C = initial impact matrix, usually from the cholesky decomposition of the
16 % forecast error variance decomposition
17 % k = number of dependent variables
18 %
19 % outputs:
20 % C = new draw of the short run impact matrix
21 %=====
22
23 newmatrix = randn(k,k);
24
25 [Q,R] = qr(newmatrix);
26
27 for ii = 1:k
28     if R(ii,ii)<0
29         Q(:,ii) = -Q(:,ii);
30     end
31 end
32
33 C = C*Q;

```

generateQ.m

```

1  function Omega = generateQ(m)
2
3  % generate an orthonormal matrix.
4
5  G      = randn(m);

```

```

6  [Q,R] = qr(G);
7  % normalize to positive entry in the diagonal
8  In     = diag(sign(diag(R)));
9  Omega = Q * In;

```

histdecomp.m

```

1  function [histdec,ierror] = histdecomp(bvar_,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'histdecomp' computes decomposition of the observable variables in
5  % terms of structural VAR shocks and initial condition
6
7  % Filippo Ferroni , 6/1/2015
8  % Revised , 2/15/2017
9  % Revised , 3/21/2018
10 % Revised , 9/11/2019
11 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
12
13 % tolerance
14 tol = 10^(-7);
15 % Retrieve the initial condition
16 yo    = bvar_.XX(1,1:end-1)'; % remove the constant from X
17 % use posterior means
18 u      = mean(bvar_.e_draws,3);      % reduced form errors
19 alpha  = mean(bvar_.Phi_draws,3);    % AR
20 Sigma  = mean(bvar_.Sigma_draws,3);  % Sigma
21 % retrieve VAR settings
22 N      = bvar_.N;
23 lags   = bvar_.lags;
24 % no rotation
25 Omega = eye(N);
26 % data
27 data = bvar_.data;
28
29
30 if nargin > 1
31     if isfield(options,'tol')==1
32         tol = options.tol;
33     end
34     if isfield(options,'yo')==1 && options.yo == 0 %no initial condition
35         yo    = zeros(N*lags,1);
36     end
37     if isfield(options,'Omega')==1 % use a specific rotation
38         Omega = options.Omega;

```

```

39     end
40     if isfield(options,'draw')==1 % use a specific draw for the reduced form
        param
41         draw = options.draw;
42         u      = bvar_.e_draws(:,:,draw);          % reduced form errors
43         alpha  = bvar_.Phi_draws(:,:,draw);        % AR
44         Sigma  = bvar_.Sigma_draws(:,:,draw);      % Sigma
45     end
46     if isfield(options,'median')==1 && options.median == 1 % use median
        instead of mean
47         u      = median(bvar_.e_draws,3);          % reduced form errors
48         alpha  = median(bvar_.Phi_draws,3);        % AR
49         Sigma  = median(bvar_.Sigma_draws,3);      % Sigma
50     end
51 end
52
53 A      = chol(Sigma,'lower');
54 ierror = zeros(length(u),N);
55
56 % (1) e = A * Omega * eta
57 % e = n * 1          %reduced
58 % eta = n * 1        % structural
59 % A = n * n
60 % Omega = n * n
61 % e' = eta' * A' * Omega';
62 % (2) E = ETA * A' * Omega';
63 % where
64 % H = T * n
65 % E = T * n
66
67 % structural innovations
68 ierror = u * inv( Omega' * A'); %#ok<MINV>
69
70 % companion form
71 F      = [alpha(1 : N * lags, :)'; eye(N*(lags-1), N*lags)];
72 G      = eye(N * lags, N);
73 C      = [alpha(end, :)'; zeros(N*(lags-1),1)];
74 ystar  = zeros(length(u),N*lags,N);
75
76 % Deterministic Part
77 for t = 1 : length(u)
78     Aa = 0 ;
79     for tau = 1 : t
80         Aa = Aa + F^(tau-1)*C;
81     end
82 %     B(t,:) = (Aa + F^(t) * yo)';
83     B(t,:) = (Aa + F^(t) * yo)';

```

```

84 %      initial condition
85 Bb(t,:) = (F^(t) * yo)';
86 end
87
88 Kappa = G * A * Omega ;
89 % Stochastic part
90 for shock = 1 : N
91     Ind          = zeros(N);
92     Ind(shock,shock) = 1;
93     for t = 1 : length(u)
94         D = 0;
95         for tau = 1 : t
96             D = D + F^(t-tau) * Kappa * Ind * ierror(tau,:)';
97         end
98         E(t,:,shock) = D';
99     end
100 end
101
102 % check
103 W = B + sum(E,3);
104 yhat = W(:, 1 : N);
105 tmp = max(max(abs(data(lags+1:end,:) - yhat)));
106 if tmp > tol,
107     histdec = [];
108     warning(['Maximum Discrepancy_' num2str(tmp)])
109     return;
110 end
111
112 histdec = E(:,1:N,:);
113 histdec(:, :, N+1) = B(:,1:N);
114
115 end

```

histdecomposition.m

```

1 function [histdec,ierror] = histdecomposition(error,alpha,Sigma,data,Omega,yo,
2     tol)
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % 'histdecomposition' computes decomposition of the observable variables in
5 % terms of structural VAR shocks and initial condition
6
7 % Filippo Ferroni, 6/1/2015
8 % Revised, 2/15/2017
9 % Revised, 3/21/2018

```

```

10 % Revised , 9/11/2019
11 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
12
13
14 if nargin < 7
15     tol = 10^(-7);
16 end
17
18 N = size(Sigma,1);
19 [m , k] = size(alpha);
20 lags = (m-1)/k;
21
22 if nargin < 5
23     Omega = eye(N);
24     yo = zeros(N*lags,1);
25 end
26 if nargin < 6
27     yo = zeros(N*lags,1);
28 end
29 %
30
31 A = chol(Sigma,'lower');
32 ierror = zeros(length(error),N);
33
34 % (1) e = A * Omega * eta
35 % e = n * 1 %reduced
36 % eta = n * 1 % structural
37 % A = n * n
38 % Omega = n * n
39 % e' = eta' * A' * Omega';
40 % (2) E = ETA * A' * Omega';
41 % where
42 % H = T * n
43 % E = T * n
44
45
46
47 ierror = error * inv( Omega' * A'); %ok<MINV>
48
49
50 % companion
51 F = [alpha(1 : N * lags, :)'; eye(N*(lags-1), N*lags)];
52 G = eye(N * lags, N);
53 C = [alpha(end, :)'; zeros(N*(lags-1),1)];
54 ystar = zeros(length(error),N*lags,N);
55
56 % Deterministic Part

```



```

4  % 'iresponse' computes the impulse response functions
5
6  % Inputs:
7  % — alpha, AR parameters of the VAR
8  % — Sigma, Covariance matrix of the reduced form VAR shocks
9  % — hor, horizon of the IRF
10 % — Omega, a particular orthonormal rotation
11 % — unit, 1 shock STD or 1 percent increase
12
13 % Output:
14 % — ir contains the IRF
15 % 1st dimension: variable
16 % 2nd dimension: horizon
17 % 3rd dimension: shock
18
19 % Filippo Ferroni, 6/1/2015
20 % Revised, 2/15/2017
21 % Revised, 3/21/2018
22 % Revised, 9/11/2019
23 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
24
25 N      = size(Sigma,1);
26 ir     = zeros(N,hor,N); % variables, horizon, shock
27 [m, n] = size(alpha);
28 lags    = floor((m-1)/n);
29 [Q]     = chol(Sigma,'lower');
30
31 % units
32 if nargin < 5
33     unit = eye(N);
34 else
35     unit = inv(diag(diag(Q)));
36 end
37
38 % 1 standard deviation increase (if unit = eye(N))
39 % 100 basis point increase (else)
40 Q = Q*unit;
41
42 % companion form
43 F = [alpha(1 : N * lags, :)'; eye(N*(lags-1), N*lags)];
44 G = eye(N * lags, N);
45 Fk = eye(N * lags);
46 % compute IRFs
47 for k=1:hor
48     PHI      = Fk * G * Q * Omega;
49     ir(1:N,k,:) = G' * PHI;
50     Fk       = F * Fk;

```

51 **end**

iresponse_longrun.m

```

1  function [ir,Q]=iresponse_longrun(alpha,Sigma,hor,lags)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'iresponse_longrun' computes the impulse response functions to a long run
   shock
5  % ordered first
6
7  % Inputs:
8  % — alpha, AR parameters of the VAR
9  % — Sigma, Covariance matrix of the reduced form VAR shocks
10 % — hor, horizon of the IRF
11 % — unit, 1 shock STD or 1 percent increase
12
13 % Output:
14 % — ir contains the IRF
15 % 1st dimension: variable
16 % 2st dimension: horizon
17 % 3st dimension: shock
18
19 % Filippo Ferroni, 6/1/2015
20 % Revised, 2/15/2017
21 % Revised, 3/21/2018
22 % Revised, 9/11/2019
23 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
24
25 N          = size(Sigma,1);
26 ir         = zeros(N,hor,N); % variables, horizon, shock
27
28 % companion
29 F          = [alpha(1 : N * lags, :)'; eye(N*(lags—1), N*lags)];
30 G          = eye(N * lags, N);
31 Inp        = eye(size(F));
32 % long run sum of VMA
33 C1         = Inp — F;
34 C1         = C1 \ Inp; % = inv(C1)*Inp;
35 C1         = C1(1 : N, 1 : N);
36 PSI1       = chol(C1 * Sigma * C1')';
37 % IMPACT MATRIX
38 Q          = (C1 \ eye(size(C1))) * PSI1;
39 % normalize to positive entry in the element Q(1,1)
40 Q(1,1)     = Q(1,1) * sign(Q(1,1));

```

```

41 Fk          = eye(N * lags);
42
43 for k=1:hor
44     PHI      = Fk * G * Q;
45     ir(1:N,k,:) = G'* PHI;
46     Fk       = F * Fk;
47 end

```

iresponse_proxy.m

```

1 function in = iresponse_proxy(in)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % 'iresponse_proxy' computes the impulse response functions to shock
5 % identified via instrumental variables
6 % Codes are written based on the Mertens Ravn (2013, AER) source codes ,
7 % modified to
8 %1) handle different length of VAR and factors
9 %2) Multiple instruments to explain the same shock
10
11 % Filippo Ferroni , 6/1/2015
12 % Revised , 2/15/2017
13 % Revised , 3/21/2018
14 % Revised , 9/11/2019
15 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
16
17 Y      = in.vars(in.p+1:end,:);
18 [T,n]  = size(Y);
19 [T_m,~] = size(in.proxies);
20
21 % number of proxies
22 k      = 1;
23 %Assuming proxies start at least p periods later
24 instrument = in.proxies(1:end,:);
25
26 % Identification
27 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
28
29 % covariance of the reduced form shocks
30 in.Sigma_m = in.Sigma;
31
32 % Instrument on VAR residuals
33 Phib = [ones(T_m,1) instrument]\in.res(T-T_m-in.T_m_end+1:T-in.T_m_end,:);
34
35 % Fitted values of the identified shocks

```

```

36 % here is where the prior on beta should enter , instead of Phib(:,1)
37 % m = beta e1 + v
38 uhat1 = [ones(T_m,1) instrument]*Phib(:,1);
39
40 % regress the fitted values on the other reduced form shocks
41 b21ib11_TSL = [ones(T_m,1) uhat1]\in.res(T-T_m-in.T_m_end+1:T-in.T_m_end
    ,2:end);
42 b21ib11_TSL = b21ib11_TSL(2:end,:);
43 b21ib11 = b21ib11_TSL;
44
45 % Identification of b11 and b12 from the covariance matrix of the VAR
46 Sig11 = in.Sigma_m(1:k,1:k);
47 Sig21 = in.Sigma_m(k+1:n,1:k);
48 Sig22 = in.Sigma_m(k+1:n,k+1:n);
49 ZZp = b21ib11*Sig11*b21ib11'-(Sig21*b21ib11'+b21ib11*Sig21')+Sig22;
50 b12b12p = (Sig21-b21ib11*Sig11)*(ZZp\'(Sig21-b21ib11*Sig11));
51 b11b11p = Sig11-b12b12p;
52 b11 = sqrt(b11b11p);
53 in.b1 = [b11; b21ib11*b11];
54 in.Phib = Phib;
55
56 % Impulse Responses
57 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
58 % initial shock: eps(1,1)=1
59 irs(in.p+1,:) = in.b1(:,1);
60
61 for jj=2:in.irhor%+max(max(VAR.term_spreads_matur),max(VAR.real_rates_init+VAR
    .real_rates_matur-1))
62     lvars = (irs(in.p+jj-1:-1:jj,:))';
63     irs(in.p+jj,:) = lvars(:)'*in.Phi(1:in.p * n,:);
64 end
65
66 in.irs = irs(in.p+1:in.p+in.irhor,:);
67 in.uhat1 = uhat1;

```

iresponse_sign.m

```

1 function [ir,Omeg] = iresponse_sign(Phi,Sigma,hor,signrestriction,cont)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % 'iresponse_sign' computes the impulse response functions using sign
5 % restrictions on the endogenous variables
6 % Reference: Rubio-Ramirez, J. F., Waggoner, D. F. and Zha, T.: 2010,
7 % Structural Vector Autoregressions: Theory of Identification and Algorithms
8 % for Inference, Review of Economic Studies 77(2), 665-696.

```

```

9
10 % Inputs:
11 % — Phi, AR parameters of the VAR
12 % — Sigma, Covariance matrix of the reduced form VAR shocks
13 % — hor, horizon of the IRF
14 % — unit, 1 shock STD or 1 percent increase
15 % — signrestriction, cell array containing the restrictions. Sign
16 % restriction can be activated in the toolbox by setting the options
17 % options.signs{1} = 'y(a,b,c)>0';
18 % where a, b and c are integer. The syntax means that shock c has a
19 % positive impact on the a-th variable at horizon b.
20
21 % Output:
22 % — ir contains the IRF
23 % 1st dimension: variable
24 % 2st dimension: horizon
25 % 3st dimension: shock
26
27 % Filippo Ferroni, 6/1/2015
28 % Revised, 2/15/2017
29 % Revised, 3/21/2018
30 % Revised, 9/11/2019
31 % Revised, 4/11/2020
32 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
33
34 [m,n] = size(Sigma);
35 ir = nan(n,hor,n);
36 Omeg = nan(n);
37 d = 0;
38 tol = 0;
39 favar = 1;
40
41 if nargin < 5
42     cont = eye(n);
43     favar = 0;
44 end
45
46
47 while d==0 && tol < 30000
48     % generate a random orthonormal matrix
49     Omega = generateQ(m);
50     % compute IRF
51     y = iresponse(Phi,Sigma,hor,Omega);
52     % uncompress the factors (if favar)
53     if favar == 1
54         for jj = 1 : size(y,3)
55             yy(:, :, jj) = cont * y(:, :, jj);

```

```

56         end
57         clear y; y = yy;
58     end
59     % check restrictions
60     d = checkrestrictions(signrestriction,y);
61     if d==1
62         Omeg = Omega;
63         ir    = y;
64     else
65         tol = tol + 1;
66     end
67 end
68
69 if d==0
70     warning('I could not find a rotation satisfying the restrictions.')
71 end
72
73 end

```

iresponse_sign_narrative.m

```

1  function [ir,Omeg] = iresponse_sign_narrative(errors,Phi,Sigma,hor,
        signrestriction,narrative,cont)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'iresponse_sign_narrative' computes the impulse response functions using
        sign
5  % restrictions on the endogenous variables
6  % Reference:
7
8  % Inputs:
9  % - e, VAR reduced form errors (T x n)
10 % - Phi, AR parameters of the VAR
11 % - Sigma, Covariance matrix of the reduced form VAR shocks
12 % - hor, horizon of the IRF
13 % - unit, 1 shock STD or 1 percent increase
14 % - signrestriction, cell array containing the restrictions. Sign
15 % restriction can be activated in the toolbox by setting the options
16 % options.signs{1} = 'y(a,b,c)>0';
17 % where a, b and c are integer. The syntax means that shock c has a
18 % positive impact on the a-th variable at horizon b.
19 % - narrative, cell array containing the restrictions. Sign
20 % restriction can be activated in the toolbox by setting the options
21 % options.narrative{1} = 'v(a,b)>0';
22 % where m can be a vector of scalar and n is a scalar. The syntax means

```

```

23 % that shock n has to be positive .
24
25 % Output:
26 % — ir contains the IRF
27 % 1st dimension: variable
28 % 2st dimension: horizon
29 % 3st dimension: shock
30
31 % Filippo Ferroni, 6/1/2015
32 % Revised, 2/15/2017
33 % Revised, 3/21/2018
34 % Revised, 9/11/2019
35 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
36
37
38 [m,n] = size(Sigma);
39 ir = nan(n,hor,n);
40 Omeg = nan(n);
41 d = 0;
42 d0 = 0;
43 tol = 0;
44 favar = 1;
45 if nargin < 7
46     cont = eye(n);
47     favar = 0;
48 end
49
50
51 A = chol(Sigma,'lower');
52 v = zeros(size(errors));
53
54 while d==0 && tol < 30000
55     % generate a random orthonormal matrix
56     Omega = generateQ(m);
57     % compute IRF
58     y = iresponse(Phi,Sigma,hor,Omega);
59     % uncompress the factors (if favar)
60     if favar == 1
61         for jj = 1 : size(y,3)
62             yy(:, :, jj) = cont * y(:, :, jj);
63         end
64         clear y; y = yy;
65     end
66     % check sign restrictions
67     d0 = checkrestrictions(signrestriction,y);
68     if d0 ==1
69         v = errors / ( Omega' * A'); % structural innovations

```



```

70         % check narrative restrictions
71         d = checkrestrictions(narrative,y,v);
72     end
73     if d==1 % stop
74         Omeg = Omega;
75         ir    = y;
76     else
77         tol = tol + 1;
78     end
79 end
80
81 if d==0
82     warning('I could not find a rotation satisfying the restrictions.')
83 end
84 end

```

iresponse_zeros_signs.m

```

1  function [ir,Omeg] = iresponse_zeros_signs(Phi,Sigma,hor,lag,var_pos,f,sr,
      draws,toler)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'iresponse_zeros_sign' computes the impulse response functions using zero
5  % and sign restrictions on the endogenous variables
6  % References :
7  % Arias , J. E. , Rubio-Ramirez , J. F. and Waggoner , D. F.: 2018, Inference
8  % Based on SVARs Identified with Sign and Zero Restrictions: Theory and
9  % Applications , Econometrica 86, 685720.
10 % Binning , A.: 2013, Underidentified SVAR models: A framework for combining
11 % short and long-run restrictions with sign-restrictions , Working Paper
12 % 2013/14, Norges Bank.
13
14 % Inputs :
15 % - Phi , AR parameters of the VAR
16 % - Sigma , Covariance matrix of the reduced form VAR shocks
17 % - hor , horizon of the IRF
18 % - unit , 1 shock STD or 1 percent increase
19 % - (var_pos,f,sr) inputs for the zero and sign restrictions see bvar.m and
20 % tutorial_.m
21
22 % Output :
23 % - ir contains the IRF
24 % 1st dimension:   variable
25 % 2st dimension:   horizon
26 % 3st dimension:   shock

```

```

27
28 % Filippo Ferroni , 6/1/2015
29 % Revised , 2/15/2017
30 % Revised , 3/21/2018
31 % Revised , 9/11/2019
32 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
33
34 if nargin< 8
35     draws = 1; % Number of draws
36 end
37 if nargin< 9
38     toler = 10000; % Number of rotation attempt
39 end
40
41 p = lag;
42 k = size(Sigma,1);
43 C1 = chol(Sigma,'lower');
44
45 ir      = nan(k,hor,k);
46 Omeg    = nan(k);
47
48 T = hor; % length of impulse response function.
49
50 shocks = eye(k);
51
52 [Q,index,flag] = findQs(k,f);
53
54 if flag == 1
55     error('Rank_condition_not_satisfied,_the_model_is_overidentified');
56 end
57
58 shock_pos = logical(shocks); % position of shock
59
60 % var_pos = [1,1,4,1,1];
61 % var_pos = [1,2,2,3,3]; % position of corresponding variable to shock
62 % % eg monetary policy shock should result in a positive increase in interest
63 % % rates , aggregate demand shock should result in a positive increase in gdp
64 % % etc.
65
66 R = zeros(k,T,length(shocks),draws); % Contains impulse resonse functions
67 % 1st dimension = variable
68 % 2nd dimension = time
69 % 3th dimension = shock
70 % 4rd dimension = draw
71
72 counter = 1;
73

```

```

74 B      = [Phi(end,:); Phi(1:end-1,:)];
75 Btilde = B(2:end,:)';
76 % Btilde = Phi(1:end-1,:)';
77 alpha = [Btilde; eye(k*(p-1)), zeros(k*(p-1),k)]; % Build companion form matrix
78 if draws > 10
79     wb = waitbar(0, 'Generating_Rotations');
80 end
81
82 tj = 0;
83 while counter < draws+1
84
85     tj = tj + 1;
86
87     C = generateDraw(C1,k);
88
89     P = findP(C,B,Q,p,k,index);
90
91     W = C*P;
92
93     for jj = 1:length(shocks)
94
95         if W(var_pos(jj),jj) < 0
96             shock = -shocks(:,jj);
97         else
98             shock = shocks(:,jj);
99         end
100
101         V = zeros(k*p,T);
102
103         V(1:k,1) = W*shock;
104
105         chk = W*shock;
106         sr_index = ~isnan(sr(:,jj));
107         tmp = sign(chk(sr_index)) - sr(sr_index,jj);
108
109         if any(tmp~=0)
110             jj = 0;
111             break
112         end
113
114         for ii = 2:T
115             V(:,ii) = alpha*V(:,ii-1);
116         end
117
118         R(:, :, jj, counter) = V(1:k,:);
119
120     end

```

```

121
122     if jj == length(shocks)
123         counter = counter + 1;
124     end
125
126     if tj > toler
127         warning('I could not find a rotation')
128         return;
129     end
130     if draws > 10, waitbar(counter/draws, wb); end
131 end
132
133
134 if draws > 10, close(wb); end
135
136 Omeg = W;
137 ir    = R;

```

isOctave.m

```

1 function r = isOctave ()
2     persistent x;
3     if (isempty (x))
4         x = exist ('OCTAVE_VERSION', 'builtin');
5     end
6     r = x;
7 end

```

jacob_bvar.m

```

1 function J=jacob_bvar(param)
2
3 % Filippo Ferroni, 6/1/2015
4 % Revised, 2/15/2017
5 % Revised, 3/21/2018
6
7 J=zeros(length(param),1);
8 for jj = 1 : length(param)
9     J(jj)= bound0prime(param(jj));
10 end
11 J=diag(J);
12
13 %

```

```

14 % function y = bound01prime(x);
15 % y = exp(x)/(1+exp(x))^2;
16 %
17 function y = bound0prime(x);
18 y = exp(x);

```

kf_dk.m

```

1  function [shatnew,signew,lh,yhat,fin,kgpart,yforc]=kf_dk(y,H,shat,sig,G,M)
2
3  % =====
4  % KF-DK
5  % function [shatnew,signew,lh,yhat,fin,kgain,yforc]=kf_dk(y,H,shat,sig,G,M)
6  %
7  % This is Chris Sims's KF with a couple of added outputs
8  % 1) The (Partial) Kalman Gain and  $F(-1)$  matrices obtained using the
9  % Generalized Inverse are part of the output
10 %  $s(t) = G*s(t-1) + R*n(t)$   $V(n(t)) = Q$ 
11 % then  $M=R*Chol(Q)'=R(CQ)'$ 
12 %  $M*M'=R*(CQ'*CQ)*R'$ 
13 %
14 % See KF.MOD. for a related filter
15 % NOTE: KGPART is NOT the appropriate Kalman Gain
16 %  $KG=G*KGPART$  such that
17 %  $KGPART=(P(t)|t-1)*(H')*(F^{-1})$ 
18 % Use this version when the G matix is time varying and adjust to the
19 % timing in DK which have a different timing in the state equation
20 % =====
21 % Revised , 2/15/2017
22 % Revised , 3/21/2018
23
24 lh=zeros(1,2);
25 omega=G*sig*G'+M*M';
26 [uo doo vo]=svd(omega);
27 [u d v]=svd(H*uo*sqrt(doo));
28 first0=min(find(diag(d)<1e-12));
29 if isempty(first0),first0=min(size(H))+1;end
30 u=u(:,1:first0-1);
31 v=v(:,1:first0-1);
32 d=diag(d);d=diag(d(1:first0-1));
33 spred = G*shat;
34 fac=vo*sqrt(doo);
35 yforc = H*spred;
36 yhat=y-yforc;
37 fhalf=(v/d)*u';

```

```

38 fin=fhalf'*fhalf;
39 ferr=fhalf*yhat;
40 lh(1)=-.5*ferr'*ferr;
41 lh(2)=-sum(log(diag(d)));
42 kgpart=fac*fhalf;
43 % Check
44 %comparemat(kgpart,omega*H'*(inv(H*omega*H')));
45 shatnew=fac*ferr+spred;
46 signew=fac*(eye(size(v,1))-v*v')*fac';
47 lh=sum(lh);

```

kfilternan.m

```

1 function [outputkf] = kfilternan(Phi,Sigma,y,options)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % 'kfilternan' runs the Kalman filter and smoother with missing values
5 % References: Durbin and Koopman (2003)
6
7 % Inputs:
8 % - Phi, AR parameters of the VAR
9 % - Sigma, Covariance matrix of the reduced form VAR shocks
10 % - y, data
11 % - options, various options
12 % TVP parameters allowed (3rd dimension of Phi,Sigma and options.tauVec
13 % controls the time of the variation)
14
15 % outputkf : (see below)
16
17 % Filippo Ferroni, 6/1/2015
18 % Revised, 2/15/2017
19 % Revised, 3/21/2018
20 % Revised, 9/11/2019
21 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
22
23 % kfilternan computes the forward Kalman filter with NaN
24
25 [T,var] = size(y);
26 data = y';
27 tauVec = ones(T,1);
28 initialCond = 0;
29 adjustment = 0;
30 index = zeros(var,1);
31 noprint = 0;
32

```

```

33  if nargin > 3
34      if isfield(options,'tauVec') == 1
35          tauVec=optinos.tauVec;
36      end
37      if isfield(options,'initialCond') ==1
38          initialCond = options.initialCond;
39      end
40      if isfield(options,'initialCond')==1 && options.initialCond==2
41          initialCond = options.initialCond;
42          pZero      = options.pZero;
43          aZero      = options.aZero;
44      end
45      if isfield(options,'adjustment')==1
46          adjustment = options.adjustment;
47      end
48      if isfield(options,'index')==1
49          index = options.index;
50      end
51      if isfield(options,'noprint')==1
52          noprint = options.noprint;
53      end
54  end
55
56  if length(tauVec)~=T
57      warning('Tauvec is Longer than T');
58      quer('c');
59  end
60
61  %=====
62  % 1.0 obtain the steady state representation of the VAR
63  for jj = 1 : size(Phi,3)
64      [A(:,:,jj),B(:,:,jj),C(:,:,jj),const(:,jj),Sigma(:,:,jj),~,index_var]=
        var2ss(Phi,Sigma,index);
65  end
66
67  %=====
68  % 1.1. Dimensions and storage
69  ns      = size(B,1);
70  vt      = zeros(var,T);
71  finvt   = zeros(var,var,T);
72  kpartg  = zeros(ns,var,T);
73  logLnc  = zeros(T,1);
74  yfor    = zeros(size(C,1),T);
75  sfor    = zeros(size(A,1),T);
76  % Matrices with one additional entrdataStru.data (initialization)
77  % to recover observables
78  stt     = zeros(ns,T+1);

```

```

79 ptt          = zeros(ns,ns,T+1);
80 W            = eye(var);
81 Zdim          = zeros(T,1);
82 mat_obspos   = zeros(T,var);
83
84 % 1.2 Initialization
85 if initialCond==0
86     stt(:,1) = zeros(ns,1);
87     P0 = lyapunov_symm(A(:, :, tauVec(1)), ...
88         B(:, :, tauVec(1))*(Sigma(:, :, tauVec(1)))' ...
89         *Sigma(:, :, tauVec(1))*(B(:, :, tauVec(1)))');
90     ptt(:, :, 1)=P0;
91 elseif initialCond==1 %non stationary data
92     P0          = 10*eye(size(A,1));
93     stt(:,1)    = zeros(size(A,1),1);
94     ptt(:, :, 1) = P0;
95 elseif initialCond==2
96     P0          = pZero;
97     stt(:,1)    = aZero;
98     ptt(:, :, 1) = P0;
99 end
100
101 nbreak = 0;
102 time   = 0;
103 state  = stt(:,1);
104
105 %=====
106 % 1.3 Start Forward Filter using KF_DK
107 for ii=1:T
108
109     % Handling of missing observations
110     ytt      = data(:,ii);
111
112     % Determine W and position of the NAN
113     ind      = ~isnan(ytt);
114     rowt     = find(~isnan(ytt));
115     ytt      = ytt(ind);
116     Zdim(ii) = length(ytt);
117     Ztt      = W((ind==1),:)*C(:, :, tauVec(ii));
118     mat_obspos(ii,1:Zdim(ii)) = rowt;
119     dimt     =( 1:Zdim(ii) );
120     % Demeaning is done here
121     ytt = ytt - Ztt*const(:,tauVec(ii));
122
123     % Forecast Part
124     % sfor is the state at time t conditional on info at time t-1,  $s(t|t-1)$ 
125     sfor(:,ii) = A(:, :, tauVec(ii))*state;

```



```

126     yfor(:,ii) = C(:, :, tauVec(ii))*(A(:, :, tauVec(ii))*state + const(:,
        tauVec(ii)));
127
128     % % computing the 1, 2,3,4 step ahead forecast
129     % yfrsct(ii, dimt, 1) = yfor(dimt, ii);
130     % yfrsct(ii, dimt, 2) = Ztt*(G(:, :, tauVec(ii))^2*stt(:, ii) + C(:, tauVec(
        ii)));
131     % yfrsct(ii, dimt, 3) = Ztt*(G(:, :, tauVec(ii))^3*stt(:, ii) + C(:, tauVec(
        ii)));
132     % yfrsct(ii, dimt, 4) = Ztt*(G(:, :, tauVec(ii))^4*stt(:, ii) + C(:, tauVec(
        ii)));
133
134     [stt(:, ii+1), ptt(:, :, ii+1), logLnc(ii), vt(dimt, ii), finvt(dimt, dimt, ii), ...
135     kpartg(:, dimt, ii),] = feval(@kf_dk, ytt, Ztt, ... %Z(:, :, tauVec(ii)), ...
136     state, ptt(:, :, ii), A(:, :, tauVec(ii)), ...
137     B(:, :, tauVec(ii))*(Sigma(:, :, tauVec(ii)))');
138
139     % if there is break
140     if ii < T
141         if tauVec(ii+1) - tauVec(ii) > 0
142             nbreak = nbreak + 1;
143             time(nbreak) = ii+1;
144             state = stt(:, ii+1) + adjustment(:, nbreak);
145         else
146             state = stt(:, ii+1);
147         end
148     end
149
150 end
151
152 %=====
153 % 2. Likelihood with Integration Constant
154 outputkf.logLncFull = logLnc;
155 %logLnc = logLnc(dataStru.trainVec(1):dataStru.trainVec(2));
156 logLnc = logLnc(1:end);
157 logL = -0.5*sum(Zdim)*log(2*pi)+sum(logLnc);
158
159 %=====
160 % 3. Truncate filters and obtain initial observations
161 % outSt.yferr=vt';
162 outputkf.yferr = (data- yfor)';
163 yfor = yfor';
164 stt = stt(:, 2:end);
165 ptt = ptt(:, :, 2:end);
166
167 % % add the 1,2,3,4 ste ahead forecasts in the output
168 % for hf = 1

```

```

169 %      tmp = (dataStru.data(:,1+hf:end) - yfrsct(1:end-hf, :, hf) ) * ...
170 %      (dataStru.data(:,1+hf:end) - yfrsct(1:end-hf, :, hf) )';
171 %      outSt.frscterror(:, hf) =
172 % end
173
174 %=====
175 % 4. Disturbance smoother with TV matrices
176 % Obtain the Innovations using a disturbance smoother
177 etamat      = zeros(var,T);
178 smooth_st    = zeros(ns,T);
179 rmat         = zeros(ns,T);
180
181 %=====
182 % 4.1 Initialize RSTAR & start at t=Nobs
183 rstar        = zeros(ns,1);
184 [rstar,etamat(:,end)] = ...
185     smoothdis(rstar,...
186         (Sigma(:,: ,tauVec(end))')*Sigma(:,: ,tauVec(end)),B(:,: ,tauVec(end))',...
187         Ztt', finvt(dimt,dimt,end),zeros(ns),vt(dimt,end));
188 smooth_st(:,end)      = stt(:,end)+ptt(:,end)*rstar;
189 rmat(:,end)           = rstar;
190 Nmat = zeros(ns,ns,T);
191 %=====
192 % 4.2 Begin Backward recursion
193 for ii=(T-1):-1:1
194
195     % Varying dimension in the backward recursions
196     dimt=( 1:Zdim(ii) );
197     % [Ztt]' = [Wtt*Z]' = Z'*Wtt'
198     Ztt = (C(:,: ,tauVec(ii))')*( W( mat_obspos(ii,1:Zdim(ii)),:)');
199
200     [rstar,etamat(:,ii)]=smoothdis(rstar,...
201         (Sigma(:,: ,tauVec(ii))')*Sigma(:,: ,tauVec(ii)),...
202         B(:,: ,tauVec(ii))',Ztt,finvt(dimt,dimt,ii),...
203         ((A(:,: ,tauVec(ii+1))-A(:,: ,tauVec(ii+1)))*...
204         kpartg(:,dimt,ii)*Ztt')'),vt(dimt,ii));
205
206     smooth_st(:,ii)=stt(:,ii)+ptt(:,ii)*rstar;
207
208     rmat(:,ii)=rstar;
209
210 end
211 %=====
212 % 4.3 smoothed initial condition
213 if initialCond~=2
214     a0 = P0*A(:,: ,tauVec(1))'*rstar; % Note: this is only correct in the
        case where a0 = zeros(ns,1)

```

```

215 else
216     iGG = pinv(A(:, :, tauVec(1)));
217     a0 = iGG*(smooth_st(:, 1) - B(:, :, tauVec(1))*etamat(:, 1));
218 end
219
220 etamat = (etamat)';
221 smooth_st = (smooth_st)';
222 stt = stt';
223
224 %=====
225 % % 5. Check Smoother
226 % % Check that Smooth States are identical if using disturbance smoother
227 % % (above) vs. state smoother (below) and if can also recover the observables
228 tol=1e-5;
229
230 Ydem = data - repmat((C(:, :, tauVec(1))*const(:, tauVec(1))), 1, size(data, 2));
231 if max(tauVec) > 1 % only one break
232     indx = min(find(tauVec==2));
233     tmp1 = data(:, 1:indx-1) - repmat((C(:, :, tauVec(1))*const(:, tauVec(1))), 1, indx-1);
234     tmp2 = data(:, indx:end) - repmat((C(:, :, tauVec(2))*const(:, tauVec(2))), 1, size(data, 2)-indx+1);
235     Ydem = [tmp1 tmp2];
236 end
237
238 maxdifYf = max(max(abs(stt(:, index_var) - Ydem')));
239 if noprint,
240 else
241     disp(['Max_Discrepancy_Filtered_vs_Actual_Data:_' num2str(maxdifYf)]);
242 end
243
244 maxdifYs = max(max(abs(smooth_st(:, index_var) - Ydem')));
245 if noprint,
246 else
247     disp(['Max_Discrepancy_Smooth_vs_Actual_Data:_' num2str(maxdifYs)]);
248 end% maxdifY=comparemat(dataStru.data, Ynanfill ');
249
250 if maxdifYf > tol || maxdifYs > tol
251     warning('Smoother_and_Filter_discrepancy_exceeds_tolerance')
252 end
253
254
255 %=====
256 % 6. Store results
257
258 outputkf.index_var = index_var;
259 outputkf.logL = logL;

```

```

260 outputkf.filteredSt      = stt;
261 outputkf.CovSt           = ptt;
262 outputkf.smoothSt        = smooth_st;
263 outputkf.innovations      = etamat;
264 outputkf.forecastObs      = yfor;
265 outputkf.aZero           = a0;
266 outputkf.adjustment      = adjustment;
267 outputkf.nbreak          = nbreak;
268 outputkf.time            = time;
269 outputkf.smoothSt_plus_ss = zeros(size(smooth_st));
270 outputkf.filteredSt_plus_ss = zeros(size(smooth_st));
271 if time>0
272     outputkf.smoothSt_plus_ss(1:time-1,:) = smooth_st(1:time-1,:) + repmat(
        const(:,1)',time-1,1);
273     outputkf.smoothSt_plus_ss(time:end,:) = smooth_st(time:end,:) + repmat(
        const(:,2)',T-time+1,1);
274     outputkf.filteredSt_plus_ss(1:time-1,:) = stt(1:time-1,:) + repmat(const
       (:,1)',time-1,1);
275     outputkf.filteredSt_plus_ss(time:end,:) = stt(time:end,:) + repmat(const
       (:,2)',T-time+1,1);
276
277 else
278     outputkf.smoothSt_plus_ss = smooth_st + repmat(const(:,end)',size(
        smooth_st,1),1);
279     outputkf.filteredSt_plus_ss = stt + repmat(const(:,end)',size(smooth_st,1)
        ,1);
280 end
281
282 Ydem(find(isnan(Ydem))) = 0;
283 tmpf = outputkf.yferr';
284 tmpf(find(isnan(tmpf))) = 0;
285 outputkf.r2 = 1 - diag(tmpf * tmpf') ./ diag(Ydem*Ydem');
286 %=====
287 % % 7. Simulating the state vector
288 etatilde      = zeros(var,T);
289
290 for tt = 1 : T
291     Qt = (Sigma(:,:,tauVec(tt))')*Sigma(:,:,tauVec(tt));
292     Ct = Qt - Qt*B(:,:,tauVec(tt))'*Nmat(:,:,tt)*B(:,:,tauVec(tt))*Qt;
293     [a,b,~] = svd(Ct);
294     iS      = a* sqrt(b);
295     dt = iS*randn(var,1);
296     etatilde(:,tt) = dt + Qt*B(:,:,tauVec(tt))'*rmat(:,tt);
297 end
298 if nbreak == 0
299     [~,smooth_sim]=kfilterRegSplitSimulation(a0,etatilde);
300 else

```

```

301     options.nbreak          = nbreak;
302     options.time            = time;
303     options.adjustment      = adjustment;
304     [~,smooth_sim] = kfilterRegSplitSimulation(a0,etatilde,options);
305 end
306 if time>0
307     outputkf.smoothSt_sim_plus_ss(1:time-1,:) = smooth_sim(1:time-1,:) +
        repmat(const(:,1)',time-1,1);
308     outputkf.smoothSt_sim_plus_ss(time:end,:) = smooth_sim(time:end,:) +
        repmat(const(:,2)',T-time+1,1);
309 else
310     outputkf.smoothSt_sim_plus_ss = smooth_sim + repmat(const(:,end)',size(
        smooth_sim,1),1);
311 end
312
313
314 %% Subroutine kfilterRegSplitSimulation allows to simulate the model
315 % Inputs
316 % sInitial: [ns 1] Initial State
317 % innovMat: [nx T] matrix of innovations
318 % By being a sub-routine it has access to all variables defined above
319 % Be-careful not to use an index (ii,jj) used above or to repeat variable
320 % names
321 function [ySim,sSim]=kfilterRegSplitSimulation(sInitial,innovMat,options)
322     if nargin < 3
323         nbreak = 0;
324         adjustment =0;
325         time =0;
326     else
327         if isfield(options,'time')
328             time = options.time;
329         else
330             error('time_of_the_break_is_missing')
331         end
332         if isfield(options,'adjustment')
333             adjustment = options.adjustment;
334         else
335             error('State_adjustment_is_missing')
336         end
337     end
338
339     if ~isequal(size(innovMat),[var T])
340         error('Input_innovMat_must_be_[nx_T]')
341     end
342     sSim=zeros(ns,T);
343     ySim=zeros(size(data));
344     sSim(:,1)=A(:, :, tauVec(1))*sInitial + ...

```

```

345         B(:, :, tauVec(1))*innovMat(:, 1);
346     ySim(:, 1) = C(:, :, tauVec(1))*sSim(:, 1); % + C(:, tauVec(1));
347     for kk = 2:T;
348         if any(kk == time)
349             % sSim(:, kk-1) = sSim(:, kk-1);
350             sSim(:, kk-1) = sSim(:, kk-1) + adjustment(:, find(kk == time));
351         end
352         sSim(:, kk) = A(:, :, tauVec(kk))*sSim(:, kk-1) + ...
353             B(:, :, tauVec(kk))*innovMat(:, kk);
354         ySim(:, kk) = C(:, :, tauVec(kk))*sSim(:, kk); % + C(:, tauVec(kk));
355     end
356     ySim = ySim';
357     sSim = sSim';
358 end
359
360 function [rzero, what] = smoothdis(rone, Q, Rtr, Ztr, Fnvr, Ltr, v)
361     % function [rzero, what] = smoothdis(rone, Q, Rtr, Ztr, Fnvr, Ltr, v)
362     %
363     % Disturbance smoother recursion for a model with no error in the obs.
364     % equation using the formulas in Durbin and Koopman Ch. 4.2
365     % rzero = Z' * F^(-1) * v + L' * rone
366     % what = Q * R' * rone
367     %
368     % Inputs
369     % Rtr = R'
370     % Ztr = Z'
371     % Ltr = L'
372     % Fnvr = F^(-1)
373     %
374     % Revised, 2/15/2017
375     % Revised, 3/21/2018
376
377     rzero = Ztr * Fnvr * v + Ltr * rone;
378     what = Q * Rtr * rzero;
379 end
380
381
382 %% End of File
383 end

```

lagX.m

```

1 function XLag = lagX(X, lags)
2 %lagX Create matrix of lagged time series
3 %

```

```

4  if size(X,2) > size(X,1)
5      X = X'; % Ensure a column vector
6  end
7
8  missingValue = NaN; % Assign default missing value
9  Lags = length(lags); % Number of lags to apply to each time series
10 [T,ny] = size(X);
11 XLag = missingValue(ones(T,ny*Lags)); % Preallocate
12
13 for c = 1:Lags
14
15     L      = lags(c);
16     columns = (ny*(c-1)+1):c*ny; % Columns to fill , this lag
17
18     if L > 0 % Time delays
19         XLag((L + 1):end,columns) = X(1:(end - L), :);
20
21     elseif L < 0 % Time leads
22         XLag(1:(end + L),columns) = X((1 - L):end, :);
23
24     else % No shifts
25         XLag(:,columns) = X;
26
27     end
28
29 end

```

lag_crit_var.m

```

1  function lag_crit_var(y,maxlag)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % 'lag_crit_var' prints various info crit (not allowed NaN)
5
6  % Filippo Ferroni , 6/1/2015
7  % Revised , 2/15/2017
8  % Revised , 3/21/2018
9  % Revised , 9/11/2019
10 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
11
12 crt_ = nan(4,maxlag);
13
14 opt.K =1;
15 for ilag = 1 : maxlag
16     tmp_ = bvar(y,ilag,opt);

```

```

17     crt_(1,ilag) = tmp_.InfoCrit.AIC;
18     crt_(2,ilag) = tmp_.InfoCrit.SIC;
19     crt_(3,ilag) = tmp_.InfoCrit.HQIC;
20     crt_(4,ilag) = tmp_.InfoCrit.BIC;
21 end
22 rownam = {'AIC','SIC','HQIC','BIC'};
23 for ilag = 1 : maxlag
24     disp('=====')
25     X = sprintf('%s=%0.0g','Number of lags',ilag);
26     disp(X)
27     for jj = 1: size(crt_,1)
28         X = sprintf('%s=%0.5g',rownam{jj},crt_(jj,ilag));
29         disp(X)
30     end
31 end

```

lyapunov_symm.m

```

1 % Copyright (C) 2006–2017 Dynare Team
2 % solves  $x - a*x*a' = b$  for  $b$  (and then  $x$ ) symmetrical
3 function [x,info]=lyapunov_symm(a,b)
4     info = 0;
5     n = size(b,1);
6     if n == 1
7         x=b/(1-a*a);
8         return
9     end
10    x=zeros(n,n);
11    [u,t]=schur(a);
12    b=u'*b*u;
13    for i=n:-1:2
14        if t(i,i-1) == 0
15            if i == n
16                c = zeros(n,1);
17            else
18                c = t(1:i,:)*(x(:,i+1:end)*t(i,i+1:end)') + ...
19                    t(i,i)*t(1:i,i+1:end)*x(i+1:end,i);
20            end
21            q = eye(i)-t(1:i,1:i)*t(i,i);
22            x(1:i,i) = q\b(1:i,i)+c;
23            x(i,1:i-1) = x(1:i-1,i)';
24        else
25            if i == n
26                c = zeros(n,1);
27                c1 = zeros(n,1);

```



```

28     else
29         c = t(1:i,:)*(x(:,i+1:end)*t(i,i+1:end)')+...
30             t(i,i)*t(1:i,i+1:end)*x(i+1:end,i)+...
31             t(i,i-1)*t(1:i,i+1:end)*x(i+1:end,i-1);
32         c1 = t(1:i,:)*(x(:,i+1:end)*t(i-1,i+1:end)')+...
33             t(i-1,i-1)*t(1:i,i+1:end)*x(i+1:end,i-1)+...
34             t(i-1,i)*t(1:i,i+1:end)*x(i+1:end,i);
35     end
36     q = [eye(i)-t(1:i,1:i)*t(i,i) -t(1:i,1:i)*t(i,i-1);...
37          -t(1:i,1:i)*t(i-1,i) eye(i)-t(1:i,1:i)*t(i-1,i-1)];
38     z = q\b(1:i,i)+c;b(1:i,i-1)+c1];
39     x(1:i,i) = z(1:i);
40     x(1:i,i-1) = z(i+1:end);
41     x(i,1:i-1)=x(1:i-1,i)';
42     x(i-1,1:i-2)=x(1:i-2,i-1)';
43     i = i - 1;
44 end
45 end
46 if i == 2
47     c = t(1,:)*(x(:,2:end)*t(1,2:end)')+t(1,1)*t(1,2:end)*x(2:end,1);
48     x(1,1)=(b(1,1)+c)/(1-t(1,1)*t(1,1));
49 end
50 x=u*x*u';

```

matrictint.m

```

1  function w = matrictint(S, df, XXi)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Computes the log of the integral of the kernel of the PDF of a
5  % normal-inverse-Wishart distribution.
6  %
7  % S:    parameter of inverse-Wishart distribution
8  % df:   number of degrees of freedom of inverse-Wishart distribution
9  % XXi:  first component of VCV matrix of matrix-normal distribution
10 %
11 % Computes the integral over (Phi, Sigma) of:
12 %
13 %  $\det(\Sigma)^{-k/2} \exp(-0.5 * \text{Tr}((\text{Phi}-\text{PhiHat})' * (\text{XXi})^{(-1)} * (\text{Phi}-\text{PhiHat}) * \Sigma^{(-1)})) * \\$ 
14 %  $\det(\Sigma)^{((df+ny+1)/2)} \exp(-0.5 * \text{Tr}(\Sigma^{(-1)} * S))$ 
15 %
16 % (where k is the dimension of XXi and ny is the dimension of S and
17 % Sigma)
18

```

```

19 % Original file downloaded from:
20 % http://sims.princeton.edu/yftp/VARtools/matlab/matricint.m
21 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
22 k=size(XXi,1);
23 ny=size(S,1);
24 [cx,p]=chol(XXi);
25 [cs,q]=chol(S);
26
27 if any(diag(cx)<100*eps)
28     error('singular_XXi')
29 end
30 if any(diag(cs)<100*eps))
31     error('singular_S')
32 end
33
34 % Matrix-normal component
35 w1 = 0.5*k*ny*log(2*pi)+ny*sum(log(diag(cx)));
36
37 % Inverse-Wishart component
38 w2 = -df*sum(log(diag(cs))) + 0.5*df*ny*log(2) + ny*(ny-1)*0.25*log(pi) +
    ggamaln(ny, df);
39
40 w = w1 + w2;
41
42 function lgg = ggamaln(m, df)
43 if df <= (m-1)
44     error('Too_few_df_in_ggamaln: _increase_the_#_of_obs_or_decrease_the_#_of_
        lags.')
45 else
46     garg = 0.5*(df+(0:-1:1-m));
47     lgg = sum(gamaln(garg));
48 end

```

max_fevd.m

```

1 function Qbar = max_fevd(i, h, j, Phi, Sigma, Kappa)
2 % finds the rotation where shock j maximizes the fevd of variable i at horizon
   h
3 % see also fevd.m
4
5 if nargin < 6
6     Kappa = 1000;
7 end
8
9 N = size(Sigma,1);

```

```

10 crit = nan(Kappa,1);
11 Q      = nan(N,N,Kappa);
12 for k = 1 : Kappa
13     Q(:, :, k) = generateQ(N);           % generate an orthonormal matrix
14     FEVD      = fevd(h, Phi, Sigma, Q(:, :, k)); % compute the FEVD
15     % Calculate the contribution of shock j, to variable i forecast error
16     % volatility, at horizon h
17     crit(k,1) = FEVD(i, h, j);
18 end
19 % Pick the maximum
20 [~, index] = max(crit);
21 Qbar       = Q(:, :, index);
22 end

```

mniw_log_dnsty.m

```

1 function log_dnsty = mniw_log_dnsty(prior, posterior, var)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % 'logmlike' computes the marginal likelihood for the NM-IW
5 % Inputs:
6 % Output: marginal data density
7 % Filippo Ferroni, 3/21/2020
8 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
9
10 [nobs, ny] = size(var.y);
11 nk         = size(var.X, 2);
12
13 Fv         = chol(prior.Phi.cov)';
14 Fo         = chol(inv(prior.Sigma.scale))';
15 iV         = inv(prior.Phi.cov);
16
17 var.e = var.y - var.X*posterior.PhiHat;
18
19 log_dnsty = - nobs*ny/2 * log(pi);
20 log_dnsty = log_dnsty - nobs / 2 * log(det(prior.Sigma.scale));
21 log_dnsty = log_dnsty - ny/2 * log(det( eye(nk) + Fv'*var.X'*var.X*Fv ));
22 % % Giannone, Lenza Primiceri (2015) Appendix A13-A14
23 FF        = eye(ny) ...
24           + Fo'* ((posterior.PhiHat - prior.Phi.mean)'* iV * (posterior.PhiHat -
25                     prior.Phi.mean) ...
26           + var.e'* var.e) * Fo;
27 log_dnsty = log_dnsty - (nobs + prior.Sigma.df)/2 * log(det( FF ));
28 log_dnsty = log_dnsty + ggammaln(ny, (nobs + prior.Sigma.df)/2) ;

```

```

28 log_dnsty = log_dnsty - ggammaIn(ny,prior.Sigma.df/2);
29
30
31 function lgg = ggammaIn(m, df)
32 if df <= (m-1)
33     error('too_few_df_in_ggammaIn;_increase_the_number_of_observations_or_
        reduce_the_number_of_lags')
34 else
35     garg = 0.5*(df+(0:-1:1-m));
36     lgg = sum(gammaIn(garg));
37 end
38
39 % posterior.S = var.u' * var.u + prior.Sigma.scale + ...
40 %     prior.Phi.mean' * Ai * prior.Phi.mean + ...
41 %     var.B' * (var.X'*var.X) * var.B ...
42 %     - posterior.PhiHat' * (var.X'*var.X + Ai) * posterior.PhiHat;
43 % FF = posterior.S;
44 % log_dnsty1 = - nobs*ny/2 * log(pi);
45 % log_dnsty1 = log_dnsty1 + prior.Sigma.df / 2 * log(det(prior.Sigma.scale));
46 % log_dnsty1 = log_dnsty1 - ny/2 * log(det(prior.Phi.cov));
47 % log_dnsty1 = log_dnsty1 - ny/2 * log(det( var.X'*var.X + iV )); % X'X + inv(
    V)
48 % %log_dnsty1 = log_dnsty1 - ny/2 * log(det( eye(nk) + Fv'*var.X'*var.X*Fv ));
49 % % % Giannone, Lenza Primiceri (2015) Appendix A13-A14
50 % FF = prior.Sigma.scale ...
51 %     + (posterior.PhiHat - prior.Phi.mean)'* iV * (posterior.PhiHat - prior.
    Phi.mean) ...
52 %     + var.e'* var.e;
53 % % FF = eye(ny) ...
54 % %     + Fo'* ((posterior.PhiHat - prior.Phi.mean)'* iV * (posterior.PhiHat -
    prior.Phi.mean) ...
55 % %     + var.e'* var.e) * Fo;
56 % log_dnsty1 = log_dnsty1 - (nobs + prior.Sigma.df)/2 * log(det( FF ));
57 % log_dnsty1 = log_dnsty1 + ggammaIn(ny,(nobs + prior.Sigma.df)/2) - ggammaIn(
    ny, prior.Sigma.df/2);

```

ols_reg.m

```

1 function [output] = ols_reg(Y,X,options)
2
3 % this function computes the ols coefficients of Y on X
4 % Y TxN
5 % X Txk (interepct at the end)
6 % Y = XB + E
7

```

```

8  nindex      = find(sum(isnan([Y,X]),2)==0);
9  index       = find(sum(isnan([Y,X]),2)>0);
10 Y(index,:)  = [];
11 X(index,:)  = [];
12 [N,K]       = size(X);
13 robust_se_  = 0;
14 L           = round(N/4);
15
16 if nargin > 2
17     if isfield(options,'robust_se_') == 1
18         robust_se_ = options.robust_se_;
19     end
20     if isfield(options,'L') == 1
21         L = options.L;
22     end
23 end
24
25
26 ny = size(Y,2);
27
28 XX    = (X'*X);
29 iXX   = XX\eye(K);
30 Bols  = X\Y;           % Bols    = (X'*X)\(X'*Y);
31 err   = Y - X*Bols;
32 se    = nan(K,ny);
33 Sols  = zeros(K*ny);
34
35 if robust_se_==2 % NW Robust SE
36     %-----%
37     Serror    = (err'*err);
38     nwWeights = (L+1-(1:L))./(L+1);
39     for j= 1 : L
40         G      = (err(j+1 : N, :)'*err(1 : N-j , :));
41         Serror = Serror + nwWeights(j) * (G + G');
42     end
43     Serror = Serror/(N-K);
44     Sols   = kron(diag(diag(Serror)),iXX);
45     se     = reshape(sqrt(diag(Sols)), K, ny);
46
47 elseif robust_se_ == 1 % Hamilton (1994), Ch 10 pag 282, eq (10.5.20)
48     %-----%
49     for vv = 1: ny % equation by equation
50         u= err(:,vv);
51         errs=X.*u;
52         V0 = [errs'*errs] / N ; %regular weighting matrix
53         for ind_i = (1:L)
54             S = errs(1:N-ind_i,:)'*errs(1+ind_i:N,:) / N;

```

```

55         V0 = V0 + (1 - ind_i/(L+1))*(S + S');
56     end
57     %      D      =      inv((X'*X)/N);
58     %      varb   = 1/N*D*V1*D;
59     Solsj = N * iXX * V0 * iXX;
60     Sols((vv-1)*K+1 : vv*K, (vv-1)*K+1 : K*vv) = Solsj;
61 end
62 Serror = (err'*err)/(N-K); % not sure this is the correct Coavariance of
    the shocks.
63 se      = reshape(sqrt(diag(Sols)), K, ny);
64
65
66 elseif robust_se_ == 5 % Matlab HAC function
67 %-----%
68 if exist('hac') ==2
69     % find constant
70     index = find(sum(diff(X,2),1)~=0);
71     indexC = find(sum(diff(X,2),1)==0);
72     for vv = 1: ny
73 %         [EstCoeffCov0, se0, ~] = hac(X(:, index), Y(:, ny), 'display', 'off', '
type', 'HC');
74         [EstCoeffCov0, se0, ~] = hac(X(:, index), Y(:, ny), 'display', 'off', '
bandwidth', 'AR1OLS'); % remove the intercept
75         se(index, vv) = se0(2:end);
76         se(indexC, vv) = se0(1); %intercept
77         Sols((vv-1)*K+1 : vv*K, (vv-1)*K+1 : K*vv) = EstCoeffCov0; %
order is not correct
78     end
79     Serror = 1/(N-K)*(err'*err);
80 %         output.TtestRobust = coeff./se;
81 %         output.pvalueRobust = tpdf(output.TtestRobust, N-K);
82 else
83     error('Matlab_Econ_Toolbox_missing')
84 end
85 else
86     Serror = 1/(N-K)*(err'*err);
87     Sols = kron(diag(diag(Serror)), iXX);
88     se = reshape(sqrt(diag(Sols)), K, ny);
89 end
90
91 Ttest = Bols./ reshape(sqrt(diag(Sols)), K, ny);
92 ESS = diag((X*Bols - mean(Y))'*(X*Bols - mean(Y)));
93 RSS = diag(err' * err);
94 TSS = diag((Y - mean(Y))' * (Y - mean(Y)));
95 R2 = ones(length(ESS),1) - RSS ./ TSS;
96 adjR2 = ones(length(ESS),1) - (ones(length(ESS),1) - R2)*(N-1)/(N-K);
97 Ftest = ESS/(K-1) ./ diag(Serror);

```

```

98
99 for v = 1 : ny
100     output.logl(v,1) = -N/2*log(2*pi*Error(v,v)) - RSS(v,1)/(2*Error(v,v))
101     ;
102     [output.AIC(v,1), output.SIC(v,1), output.HQIC(v,1)] = IC(output.logl(v,1)
103         , N, K);
104 end
105
106 output.beta = Bols; % OLS estimator
107 output.error = err; % (TxN) matrix of Residuals
108 output.Serror = Serror; % Covariance matrix of Residuals
109 output.Sols = Sols; % Covariance matrix of Bols
110 output.Ttest = Ttest; % t-statistics
111 output.pvalue = tpdf(Ttest,N-K); % p-value
112 output.Ftest = Ftest; % F-test
113 output.R2 = R2; % R2
114 output.adjR2 = adjR2; % Adjusted R2
115 output.yfit = X*Bols; % Fitted Values
116 output.N = N; % # of observation used
117 output.K = K; % # of regressors
118 output.nindex = nindex; % index of missing observation
119 output.index = index; % index of observation used
120 output.se = se;
121 output.XX = XX;
122 output.X = X;
123 output.Y = Y;
124
125 % if nargin > 2
126 % [EstCov,se,coeff] = hac(X(:,1:end-1),Y,'display','off'); % remove the
127 % intercept
128 % output.serob = se(2:end);
129 % output.serob(end+1) = se(1); %intercept
130 % output.TtestRobust = coeff./se;
131 % output.pvalueRobust = tpdf(output.TtestRobust,N-K);
132 % end

```

p2p.m

```

1 function [posterior,prior] = p2p(hh, shrinkage,prior0,olsreg,F,G,Fo,
2     positions_nylags,position_constant)
3 %
4 %*****
5 % Conjugate Prior: MN-IW for Direct Methods
6 %*****

```

```

6
7 % settings
8 if isempty(position_constant) == 0
9     nx = 1;
10 else
11     nx = 0;
12 end
13
14 ny      = size(G,2);
15 nylags  = size(F,1);
16 iIminusF = inv(eye(nylags) - F);
17
18 % constructing the prior mean
19 Fhh      = F^hh;
20 Fohh     = (iIminusF * (eye(nylags) - Fhh)) * Fo;
21 if hh < 40
22     priorSigmaScale = zeros(ny);
23     for j = 0 : hh
24         priorSigmaScale = priorSigmaScale + G' * F^(hh-j) * G * prior0.Sigma.
                scale * G' * F^(hh-j)' * G;
25     end
26 else
27     % solves  $x - a * x * a' = b$  for  $b$  (and then  $x$ ) symmetrical
28     % function [x,info]=lyapunov_symm(a,b)
29     [priorSigmaScale0,~] = lyapunov_symm(F,G*prior0.Sigma.scale*G' - F^(hh+1)*
                G * prior0.Sigma.scale * G' * F^(hh+1)');
30     % max(max(abs(priorSigmaScale - G'*priorSigmaScale0*G)))
31     priorSigmaScale      = G'*priorSigmaScale0*G;
32 end
33
34 prior.BetaMean = [Fhh(1:ny,:)'; Fohh(1 : ny,1)'];
35 prior.BetaVar  = prior0.Phi.cov * 1/ shrinkage;
36
37 prior.df = prior0.Sigma.df;          % usually number of regressors minus
    the 2
38 prior.XXi = inv( prior.BetaVar );    %  $V\{-1\}$ 
39 prior.S    = priorSigmaScale;        %  $\Sigma_0$ 
40
41 % retrieve the OLS
42 B_ = olsreg.beta([positions_nylags position_constant], :);
43 XX_ = olsreg.X(:, [positions_nylags position_constant])' * olsreg.X(:, [
    positions_nylags position_constant]);
44 E_ = olsreg.error;
45 XXp_ = XX_ + prior.XXi;
46
47 % construct the posterior
48 posterior.df = olsreg.N - nylags - nx + prior0.Sigma.df;

```



```

49 posterior.XXi      = inv( XXp_ );
50 posterior.PhiHat    = posterior.XXi * (XX_ * B_ + prior.XXi * prior.BetaMean);
51 posterior.S         = ...
52     E_ ' * E_ + priorSigmaScale + prior.BetaMean' * prior.XXi * prior.BetaMean +
    ...
53     B_ ' * XX_ * B_ - posterior.PhiHat' * XXp_ * posterior.PhiHat;
54 % FF      = prior.Sigma.scale + (posterior1.PhiHat - prior1.BetaMean)' * prior1 .
    XXi * (posterior1.PhiHat - prior1.BetaMean) ...
55 %      + posterior1.E_ ' * posterior1.E_ ;
56
57 posterior.B_      = B_ ;
58 posterior.XX_     = XX_ ;           %olsreg.X(:,[positions_nylags position_constant]) '
    * olsreg.X(:,[positions_nylags position_constant]);
59 posterior.E_      = E_ ;           %olsreg.error ' ;
60 posterior.XXp_    = XXp_ ;         %XX_ + prior.XXi ;
61 posterior.U_      = olsreg.Y - olsreg.X(:,[positions_nylags position_constant]) *
    posterior.PhiHat;% posterior errors ;

```

pc_T.m

```

1  % principal components with normalization F'F/T=I
2  % X is observed
3  % r is the true number of true factors
4  % F is T by r matrix of true factors
5  % Lambda N by r is the true loading matrix
6  % C=F*Lambda' T by N is the true common component
7  % chat is the estimated common component
8
9  function [ehat,fhat,lambda,ss,Scale]=pc_T(yy,nfac,DEMEAN)
10
11  Scale = ones(size(yy,2),1);
12
13  if DEMEAN == 2
14      [y,Scale]=standard(yy);
15  elseif DEMEAN ==1
16      y=demean(yy);
17  else
18      y=yy;
19  end
20
21
22  [bigt,bign]=size(y);
23  yy=y*y';
24  [Fhat0,eigval,Fhat1]=svd(yy);
25  fhat=Fhat0(:,1:nfac)*sqrt(bigt);

```

```

26  lambda=y'*fhat/bignt;
27
28  %chi2=fhat*lambda';
29  %diag(lambda'*lambda)
30  %diag(fhat'*fhat) % this should equal the largest eigenvalues
31  %sum(diag(eigval(1:nfac,1:nfac)))/sum(diag(eigval))
32  %mean(var(chi2)) % this should equal decomposition of variance
33
34  ehat=y-fhat*lambda';
35
36  ve2=sum(ehat'.*ehat')'/bign;
37  ss=diag(eigval);

```

plot_all_irfs.m

```

1  function plot_all_irfs_(irfs,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Filippo Ferroni, 6/1/2015
5  % Revised, 2/15/2017
6  % Revised, 3/21/2018
7
8  % input : irfs
9  % 1st dimension: variable
10 % 2nd dimension: horizon
11 % 3rd dimension: shock
12 % 4th dimension: draws
13 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
14
15 nvar = size(irfs,1);
16 hor = size(irfs,2);
17 nshocks = size(irfs,3);
18 ndraws = size(irfs,4);
19 nplots = [nvar nshocks];
20 savefig_yes = 0;
21 conf_sig = 0.68;
22 normz = 1;
23 add_irfs_yes = 0;
24 normz_yes = 0;
25 add_multiple_bands_yes = 0;
26
27 if nargin < 2
28     disp('You did not provided names for shocks nor variables.')
29     disp('I call them Var_1, Var_2, ... and Shck_1, ...')
30     for v = 1 : nvar

```

```

31         eval(['varnames{',    num2str(v)  '}=_=_','Var_=_', num2str(v)  '','',';'])
32     end
33     for v = 1 : nshocks
34         eval(['shocksnames{', num2str(v)  '}=_=_','Shck_=_', num2str(v)  '','',';'])
35     end
36 else
37     if isfield(options,'varnames')==1
38         varnames = options.varnames;
39     end
40     if isfield(options,'shocksnames')== 1
41         shocksnames = options.shocksnames;
42         if length(shocksnames) ~= nshocks
43             error('There is a mismatch between the number of shocks and the
44                 names')
45         end
46     else
47         for v = 1 : nshocks
48             eval(['shocksnames{', num2str(v)  '}=_=_','Shck_=_', varnames{v}  '','',';'])
49         end
50     end
51     if isfield(options,'normz')==1 && options.normz==1,
52         normz_yes = 1;
53     end
54     if isfield(options,'conf_sig')==1;
55         conf_sig = options.conf_sig;
56     end
57     if isfield(options,'nplots')==1;
58         nplots = options.nplots;
59     end
60     if isfield(options,'saveas_strng')==1;
61         savefig_yes = 1;
62         % setting the names of the figure to save
63         fnam_suffix = [ 'irfs_' options.saveas_strng ];
64         fnam_dir     = '.';
65     end
66     if isfield(options,'saveas_dir')==1;
67         % setting the folder where to save the figure
68         fnam_dir = options.saveas_dir;
69         if exist(fnam_dir,'dir')== 0
70             mkdir(fnam_dir)
71         end
72     end
73     if isfield(options,'add_irfs')==1
74         % setting the folder where to save the figure
75         add_irfs = options.add_irfs;
76         add_irfs_yes = 1;
77     end

```

```

77     if isfield(options,'conf_sig_2')==1
78         add_multiple_bands_yes = 1;
79         sort_idx_2 = round((0.5 + [-options.conf_sig_2, options.conf_sig_2,
80             0]/2) * ndraws);
81     end
82 end
83
84
85
86 % nfigs = ceil(length(varnames)/(nplots(1)*nplots(2)));
87 % nplots = repmat(nplots,nfigs,1);
88 % for j=1:size(nplots,1),
89 %     nbofplots(j)=nplots(j,1)*nplots(j,2);
90 % end
91 %
92 % ntotplots = sum(nbofplots);
93 % if ntotplots<length(varnames),
94 %     nfigplus = ceil((length(pplotvar)-ntotplots)/nbofplots(end));
95 %     lastrow=nplots(end,:);
96 %     lastrow=repmat(lastrow,nfigplus,1);
97 %     nplots = [nplots;lastrow];
98 %     nbofplots = [nbofplots repmat(nbofplots(end),1,nfigplus)];
99 % end
100 %
101 % conf_sig = 0.68;
102 % sort_idx = round((0.5 + [-conf_sig, conf_sig, 0]/2) * options.K);
103 %
104 % sims_shock_down_conf = normz * sims_shock_sort(:, :, :, sort_idx(1));
105 % sims_shock_up_conf = normz * sims_shock_sort(:, :, :, sort_idx(2));
106 % sims_shock_median = normz * sims_shock_sort(:, :, :, sort_idx(3));
107
108 if ndraws > 1
109     sort_idx = round((0.5 + [-conf_sig, conf_sig, 0]/2) * ndraws);
110     irf_sort = sort(irfs,4);
111     if normz_yes == 1
112         % normalize the IRF relative to a 100 bpt increase in the first
113         % variable, first horizon, first shock
114         normz = 1/ irf_sort(1, 1, 1, sort_idx(3));
115     end
116     if sort_idx(1) == 0,
117         sort_idx(1) = 1;
118         warning('IRF_Bands_not_reliable...You_have_too_few_draws.')
119     end
120     irf_Median = normz * squeeze(irf_sort(:, :, :, sort_idx(3)));
121     irf_low = normz * squeeze(irf_sort(:, :, :, sort_idx(1)));
122     irf_up = normz * squeeze(irf_sort(:, :, :, sort_idx(2)));

```

```

123     if add_multiple_bands_yes == 1
124         irf_low_low = normz * squeeze(irf_sort(:, :, :, sort_idx_2(1) ));
125         irf_up_up   = normz * squeeze(irf_sort(:, :, :, sort_idx_2(2) ));
126     end
127
128     else
129         irf_Median = normz * irfs;
130         irf_low    = normz * irfs;
131         irf_up     = normz * irfs;
132     end
133
134     jplot = 0;
135     % jfig = 0;
136     figure('name', ['All_IRFs' ] );
137     for sho = 1 : nshocks
138         for var= 1: size(varnames,2)
139
140
141             jplot=jplot+1;
142             subplot(nplots(1),nplots(2),jplot)
143
144             if add_multiple_bands_yes == 1
145                 h = area([irf_low_low(var,:,sho)',...
146                     irf_low(var,:,sho)' - irf_low_low(var,:,sho)',...
147                     irf_up(var,:,sho)' - irf_low(var,:,sho)',...
148                     irf_up_up(var,:,sho)' - irf_up(var,:,sho)'] );%, 'FaceColor
149                     ', [.85 .85 .85] );
150                 set(h(4), 'FaceColor', [.95 .95 .95])
151                 set(h(3), 'FaceColor', [.85 .85 .85])
152                 set(h(2), 'FaceColor', [.95 .95 .95])
153                 set(h(1), 'FaceColor', [1 1 1])
154                 set(h, 'linestyle', 'none')
155                 hold on
156
157             else
158                 h = area([irf_low(var,:,sho)',...
159                     irf_up(var,:,sho)' - irf_low(var,:,sho)'] );%, 'FaceColor ', [.85
160                     .85 .85] );
161                 set(h(2), 'FaceColor', [.85 .85 .85])
162                 set(h(1), 'FaceColor', [1 1 1])
163                 set(h, 'linestyle', 'none')
164                 hold on;
165
166             end
167
168             plot(irf_Median(var,:,sho), 'k');
169             if add_irfs_yes == 1
170                 plot(add_irfs(var,:,sho), 'b', 'LineWidth', 2);

```

```

168         end
169 %         if add_multiple_bands_yes == 1
170 %             plot(irf-up-up(var,:),sho),'k:','LineWidth',1.2);
171 %             plot(irf-low-low(var,:),sho),'k:','LineWidth',1.2);
172 %         end
173         hold on;
174         plot(zeros(1,hor),'k')
175         hold on;
176         hold on
177         axis tight
178         if jplot <= nvar
179             title(varnames{var})
180         end
181         if jplot == nvar*(sho-1) + 1
182             ylabel(shocksnames{sho})
183         end
184         set(gcf,'position',[50 50 800 650])
185     end
186 end
187 if savefig_yes == 1,
188     STR_RECAP = [ fnam_dir '/' fnam_suffix ];
189     saveas(gcf,STR_RECAP,'fig');
190     saveas(gcf,STR_RECAP,'eps');
191     savefigure_pdf([STR_RECAP '.pdf']);
192 end

```

plot_frctst.m

```

1  function plot_frctst_(frctsts,y,time,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Filippo Ferroni, 6/1/2015
5  % Revised, 2/15/2017
6  % Revised, 3/21/2018
7  % Revised, 8/08/2019
8
9  % input : frctsts
10 % 1st dimension: horizon
11 % 2nd dimension: variable
12 % 3rdth dimension: draws
13 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
14
15 if length(time) ~= length(y)
16     error('Mismatch between time and in-sample data (y)');
17 end

```

```

18  if size(time,2)>size(time,1)
19      time=time';
20  end
21
22  hor      = size(frcsts,1);
23  nvar     = size(frcsts,2);
24  ndraws   = size(frcsts,3);
25  nplots   = [ceil(sqrt(nvar)) ceil(sqrt(nvar))];
26  savefig_yes = 0;
27  conf_sig  = 0.68;
28  add_frcst_yes = 0;
29  add_multiple_bands_yes = 0;
30  fnam_dir  = '.';
31  fnam_suffix = 'frcsts';
32  trasf_yes = 0;
33
34  % retrieve the frequency
35  integerTest = ~mod(time,1);
36  indx        = find(integerTest==1);
37  frq         = indx(2)-indx(1);
38
39  if frq == 12 % monthly
40      timefor = time(end) + 1/12 : 1/12 : time(end) + hor/12;
41      time = [time; timefor'];
42  elseif frq == 4 % quarterly
43      timefor = time(end) + 1/4 : 1/4 : time(end) + hor/4;
44      time = [time; timefor'];
45  elseif frq == 1 % annual
46      timefor = time(end) + 1 : 1 : time(end) + hor;
47      time = [time; timefor'];
48  elseif frq == 48 % weekly
49      timefor = time(end) + 1/48 : 1/48 : time(end) + hor/48;
50      time = [time; timefor'];
51  else
52      error('Frequency_not_defined.')
53  end
54  %
55  % if strmatch(freq,'m') == 1 %#ok< *MATCH2>
56  %     timefor = time(end) + 1/12 : 1/12 : time(end) + hor/12;
57  %     time = [time; timefor'];
58  % elseif strmatch(freq,'q') == 1
59  %     timefor = time(end) + 1/4 : 1/4 : time(end) + hor/4;
60  %     time = [time; timefor'];
61  % elseif strmatch(freq,'a') == 1
62  %     timefor = time(end) + 1 : 1 : time(end) + hor;
63  %     time = [time; timefor'];
64  % else

```

```

65 %      error('You need to provide a frequency: ''m'', ''q'' or ''a''.')
66 % end
67 time_start = 1;
68 time_end = length(time);
69
70 if nargin < 4
71     disp('You did not provided names for variables.')
72     disp('I call them Var_1, Var_2, ...')
73     for v = 1 : nvar
74         eval(['varnames{', num2str(v), '} = ''Var_', num2str(v), ''';'])
75     end
76 else
77     if isfield(options, 'time_start') == 1
78         time_start = find(options.time_start == time);
79         if isempty(time_start) == 1
80             error('''time_start'' is not included in ''T''.')
81         end
82     end
83     if isfield(options, 'order_transform') == 1
84         trasf_yes = 1;
85         order_trasform = options.order_transform;
86         if length(order_trasform) ~= nvar
87             error('Mismatch between the ''order_transform'' size and # of
              variables.')
88         end
89     end
90     if isfield(options, 'varnames') == 1
91         varnames = options.varnames;
92         if length(varnames) ~= nvar
93             error('Mismatch between the # varnames and # of variables to plot'
              )
94         end
95     else
96         disp('You did not provided names for the endogenous variables.')
97         disp('I call them Var_1, Var_2, ...')
98         for v = 1 : nvar
99             eval(['varnames{', num2str(v), '} = ''Var_', num2str(v), ''';'])
100         end
101     end
102     if isfield(options, 'nplots') == 1
103         nplots = options.nplots;
104     end
105     if isfield(options, 'saveas_strng') == 1
106         savefig_yes = 1;
107         % setting the names of the figure to save
108         fnam_suffix = [ fnam_suffix options.saveas_strng ];
109     end

```



```

110     if isfield(options,'saveas_dir') == 1
111         savefig_yes = 1;
112         % setting the folder where to save the figure
113         fnam_dir = options.saveas_dir;
114         if exist(fnam_dir,'dir') == 0
115             mkdir(fnam_dir)
116         end
117     end
118     if isfield(options,'add_frcst') ==1
119         add_frcst = options.add_frcst;
120         if size(add_frcst) ~= [length(time), nvar]
121             error('The''add_frcst'' dimensions must be in-sample + output-of-
                sample length and the # of variables to plot')
122         end
123         add_frcst_yes = 1;
124     end
125     if isfield(options,'conf_sig') ==1
126         conf_sig = options.conf_sig;
127     end
128     if isfield(options,'conf_sig_2') ==1
129         if options.conf_sig_2 < conf_sig
130             error('Additional confidence bands should be larger than'' options
                .conf_sig''.')
131         end
132         add_multiple_bands_yes = 1;
133         sort_idx_2 = round((0.5 + [-options.conf_sig_2, options.conf_sig_2,
                0]/2) * ndraws);
134     end
135 end
136
137 nfigs = ceil(length(varnames)/( nplots(1)*nplots(2)) );
138 nplots = repmat(nplots,nfigs,1);
139 for j=1:size(nplots,1),
140     nbofplots(j)=nplots(j,1)*nplots(j,2);
141 end
142
143 ntotplots = sum(nbofplots);
144 if ntotplots<length(varnames),
145     nfigplus = ceil((length(pplotvar)-ntotplots)/nbofplots(end));
146     lastrow=nplots(end,:);
147     lastrow=repmat(lastrow,nfigplus,1);
148     nplots = [nplots;lastrow];
149     nbofplots = [nbofplots repmat(nbofplots(end),1,nfigplus)];
150 end
151
152 if ndraws > 1
153     frcsts_ = nan(length(time),nvar,ndraws);

```

```

154     for kk = 1 : ndraws
155         frcsts_(:, :, kk) = [y; frcsts(:, :, kk)];
156     end
157     frcsts = frcsts_;
158
159     sort_idx = round((0.5 + [-conf_sig, conf_sig, 0]/2) * ndraws);
160
161     if trasf_yes == 1
162         frcsts_ = nan(size(frcsts));
163         for var = 1 : nvar
164             if order_trasform(var) == 1 % period over period
165                 frcsts_(2:end, var, :) = (frcsts(2:end, var, :) - frcsts(1:end-1,
166                                     var, :)) ;
167
168                 elseif order_trasform(var) == 100 % percentage period over period
169                     frcsts_(2:end, var, :) = 100*(frcsts(2:end, var, :) - frcsts(1:end
170                                     -1, var, :));
171
172                 elseif order_trasform(var) == 12 % percentage 12 periord over 12
173                     period (year over year % change f or monthly data)
174                     frcsts_(13:end, var, :) = 100*(frcsts(13:end, var, :) - frcsts(1:
175                                     end-12, var, :));
176
177                 elseif order_trasform(var) == 4 % percentage 4 periord over 4
178                     period (year over year % change for quarterly data)
179                     frcsts_(5:end, var, :) = 100*(frcsts(5:end, var, :) - frcsts(1:end
180                                     -4, var, :));
181
182                 elseif order_trasform(var) == 400
183                     frcsts_(2:end, var, :) = 400*(frcsts(2:end, var, :) - frcsts(1:end
184                                     -1, var, :));
185
186                 elseif order_trasform(var) == 1200
187                     frcsts_(2:end, var, :) = 1200*(frcsts(2:end, var, :) - frcsts(1:
188                                     end-1, var, :));
189
190             else
191                 frcsts_( :, var, :) = frcsts(:, var, :);
192             end
193         end
194     end
195     frcsts = frcsts_;
196
197     frcsts_sort = sort(frcsts, 3);
198     if sort_idx(1) == 0
199         sort_idx(1) = 1;
200         warning('Bands not reliable...You have too few draws.')

```

```

193     end
194     irf_Median = squeeze(frcsts_sort(:, :, sort_idx(3) ));
195     irf_low    = squeeze(frcsts_sort(:, :, sort_idx(1) ));
196     irf_up     = squeeze(frcsts_sort(:, :, sort_idx(2) ));
197     if add_multiple_bands_yes == 1
198         irf_low_low = squeeze(frcsts_sort(:, :, sort_idx_2(1) ));
199         irf_up_up   = squeeze(frcsts_sort(:, :, sort_idx_2(2) ));
200     end
201
202 else
203     irf_Median = [y; frcsts];
204
205     if trasf_yes == 1
206         irf_Median_ = nan(size(irf_Median));
207         for var = 1 : nvar
208             if order_trasform(var) == 1
209                 irf_Median_(2:end,var) = diff(irf_Median(:,var)) ;
210
211                 elseif order_trasform(var) == 12 % percentage 12 periord over 12
212                     period (year over year % change f or monthly data)
213                     irf_Median_(13:end,var) = 100*(irf_Median(13:end,var) —
214                         irf_Median(1:end—12,var));
215
216                     elseif order_trasform(var) == 4 % percentage 4 periord over 4
217                         period (year over year % change for quarterly data)
218                         irf_Median_(5:end,var) = 100*(irf_Median(5:end,var) —
219                             irf_Median(1:end—4,var));
220
221                         elseif order_trasform(var) == 100
222                             irf_Median_(2:end,var) = 100*diff(irf_Median(:,var));
223
224                             elseif order_trasform(var) == 400
225                                 irf_Median_(2:end,var) = 400*diff(irf_Median(:,var));
226
227                                 elseif order_trasform(var) == 1200
228                                     irf_Median_(2:end,var) = 1200*diff(irf_Median(:,var));
229
230                                 else
231                                     irf_Median_(:,var) = irf_Median(:,var);
232                                 end
233                             end
234                         end
235                     irf_Median = irf_Median_;
236
237     end
238
239     irf_low    = irf_Median;
240     irf_up     = irf_Median;
241
242
243

```

```

236 end
237
238 if trasf_yes ==1 && add_frcst_yes ==1
239     add_frcst_ = nan(size(add_frcst));
240     for var = 1 : nvar
241         if order_trasform(var) == 1
242             add_frcst_(2:end,var) = diff(add_frcst(:,var)) ;
243
244             elseif order_trasform(var) == 12 % percentage 12 periord over 12
                period (year over year % change for monthly data)
245                 add_frcst_(13:end,var) = 100*(add_frcst(13:end,var) - add_frcst(1:
                    end-12,var));
246
247 %             elseif order_trasform(var) == 4 % percentage 4 periord over 4 period
                (year over year % change for quarterly data)
248 %                 add_frcst_(4:end,var) = 100*(add_frcst(4:end,var) - add_frcst(1:
                    end-3,var));
249
250                 elseif order_trasform(var) == 4 % percentage 4 periord over 4 period (
                    m over m % change for monthly data or YoY for Q data)
251                     add_frcst_(5:end,var) = 100*(add_frcst(5:end,var) - add_frcst(1:
                        end-4,var));
252
253                     elseif order_trasform(var) == 100
254                         add_frcst_(2:end,var) = 100*diff(add_frcst(:,var));
255
256                     elseif order_trasform(var) == 400
257                         add_frcst_(2:end,var) = 400*diff(add_frcst(:,var));
258
259                     elseif order_trasform(var) == 1200
260                         add_frcst_(2:end,var) = 1200*diff(add_frcst(:,var));
261
262                     else
263                         add_frcst_(:,var) = add_frcst(:,var);
264                     end
265                 end
266             add_frcst = add_frcst_;
267         end
268
269
270 jplot = 0;
271 jfig = 0;
272
273 index_time = time_start:time_end;
274
275 for var= 1: nvar
276

```

```

277     if jplot==0,
278         figure('name',['Forecasts'] );
279         jfig=jfig+1;
280     end
281
282     jplot=jplot+1;
283     subplot(nplots(1),nplots(2),jplot)
284
285
286     if add_multiple_bands_yes == 1
287
288         h = area([time(index_time)],[irf_low_low(index_time,var),...
289             irf_low(index_time,var) - irf_low_low(index_time,var),...
290             irf_up(index_time,var) - irf_low(index_time,var),...
291             irf_up_up(index_time,var) - irf_up(index_time,var)]];%, 'FaceColor
292             ', [.85 .85 .85]);
293         set(h(4),'FaceColor',[.95 .95 .95])
294         set(h(3),'FaceColor',[.85 .85 .85])
295         set(h(2),'FaceColor',[.95 .95 .95])
296         set(h(1),'FaceColor',[1 1 1])
297         set(h,'linestyle','none')
298         hold on
299         min_ = min(irf_low_low(index_time,var));
300         max_ = max(irf_up_up(index_time,var));
301     else
302         h = area([time(index_time)],[irf_low(index_time,var),...
303             irf_up(index_time,var) - irf_low(index_time,var)]];%, 'FaceColor
304             ', [.85 .85 .85]);
305         set(h(2),'FaceColor',[.85 .85 .85])
306         set(h(1),'FaceColor',[1 1 1])
307         set(h,'linestyle','none')
308         min_ = min(irf_low(index_time,var));
309         max_ = max(irf_up(index_time,var));
310     end
311     hold on;
312     plot(time(index_time),irf_Median(index_time,var),'k');
313     hold on;
314     plot(time(index_time),zeros(length(index_time),1),'k');
315     hold on;
316     axis tight
317     ylim([min_,max_]);
318     if add_frcst_yes == 1
319         plot(time(index_time),add_frcst(index_time,var),'b','LineWidth',2);
320         ylim([min([min_,min(add_frcst(index_time,var))]) , max([max_,max(
321             add_frcst(index_time,var))]) ] ));
322     end
323     title(varnames{var})

```

```

321     set(gcf,'position',[50 50 800 650])
322     if jplot==nbofplots(jfig) || var==length(varnames)
323         %legend(legenda);
324         if savefig_yes == 1
325             STR_RECAP = [ fnam_dir '/' fnam_suffix '_' int2str(jfig)];
326             saveas(gcf,STR_RECAP,'fig');
327             saveas(gcf,STR_RECAP,'eps');
328             % savefigure_pdf(STR_RECAP);
329             savefigure_pdf([STR_RECAP '.pdf']);
330         end
331         jplot=0;
332     end
333 end

```

plot_irfs.m

```

1  function plot_irfs_(irfs,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Filippo Ferroni, 6/1/2015
5  % Revised, 2/15/2017
6  % Revised, 3/21/2018
7  % Revised, 8/08/2019
8
9  % input : irfs
10 % 1st dimension: variable
11 % 2nd dimension: horizon
12 % 3rd dimension: shock
13 % 4th dimension: draws
14 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
15
16 nvar      = size(irfs,1);
17 hor       = size(irfs,2);
18 nshocks   = size(irfs,3);
19 ndraws    = size(irfs,4);
20 nplots    = [ceil(sqrt(nvar)) ceil(sqrt(nvar))];
21 savefig_yes = 0;
22 conf_sig   = 0.68;
23 normz      = 1;
24 add_irfs_yes = 0;
25 normz_yes  = 0;
26 add_multiple_bands_yes = 0;
27 fnam_dir   = '.';
28 fnam_suffix = 'irfs_';
29 fontsize   = 12;

```

```

30
31 if nargin <2
32     disp('You did not provided names for shocks nor variables.')
33     disp('I call them Var_1, Var_2, ... and Shck_1, ...')
34     for v = 1 : nvar
35         eval(['varnames{', num2str(v) '} = ', 'Var_', num2str(v) ', '];')
36     end
37     for v = 1 : nshocks
38         eval(['shocksnames{', num2str(v) '} = ', 'Shck_', num2str(v) ', '];')
39     end
40 else
41     if isfield(options, 'varnames') ==1
42         varnames = options.varnames;
43         if length(varnames) ~= nvar
44             error('There is a mismatch between the number of var and the names')
45         end
46     else
47         disp('You did not provided names for the endogenous variables.')
48         disp('I call them Var_1, Var_2, ...')
49         for v = 1 : nvar
50             eval(['varnames{', num2str(v) '} = ', 'Var_', num2str(v) ', '];')
51         end
52     end
53     if isfield(options, 'shocksnames') == 1
54         shocksnames = options.shocksnames;
55         if length(shocksnames) ~= nshocks
56             error('There is a mismatch between the number of shocks and the names')
57         end
58     else
59         disp('You did not provided names for the shocks.')
60         disp('I call them Shck_1, Shck_2, ...')
61         for v = 1 : nshocks
62             eval(['shocksnames{', num2str(v) '} = ', 'Shck_', num2str(v) ', '];')
63         end
64     end
65     if isfield(options, 'normz') ==1 && options.normz==1,
66         normz_yes = 1;
67     end
68     if isfield(options, 'conf_sig') ==1;
69         conf_sig = options.conf_sig;
70     end
71     if isfield(options, 'nplots') ==1;
72         nplots = options.nplots;
73     end
74     if isfield(options, 'saveas_strng') ==1;

```

```

75         savefig_yes = 1;
76         % setting the names of the figure to save
77         fnam_suffix = [ fnam_suffix options.saveas_strng ];
78     end
79     if isfield(options,'saveas_dir')==1;
80         savefig_yes = 1;
81         % setting the folder where to save the figure
82         fnam_dir = options.saveas_dir;
83         if exist(fnam_dir,'dir')==0
84             mkdir(fnam_dir)
85         end
86     end
87     if isfield(options,'add_irfs')==1
88         % setting the folder where to save the figure
89         add_irfs = options.add_irfs;
90         add_irfs_yes = 1;
91     end
92     if isfield(options,'conf_sig_2')==1
93         add_multiple_bands_yes = 1;
94         sort_idx_2 = round((0.5 + [-options.conf_sig_2, options.conf_sig_2,
95                                     0]/2) * ndraws);
96     end
97     if isfield(options,'fontsize')==1
98         % title font size
99         fontsize = options.fontsize;
100     end
101 end
102
103
104
105 nfigs = ceil(length(varnames)/(nplots(1)*nplots(2)));
106 nplots = repmat(nplots,nfigs,1);
107 for j=1:size(nplots,1),
108     nbofplots(j)=nplots(j,1)*nplots(j,2);
109 end
110
111 ntotplots = sum(nbofplots);
112 if ntotplots<length(varnames),
113     nfigplus = ceil((length(pplotvar)-ntotplots)/nbofplots(end));
114     lastrow=nplots(end,:);
115     lastrow=repmat(lastrow,nfigplus,1);
116     nplots = [nplots;lastrow];
117     nbofplots = [nbofplots repmat(nbofplots(end),1,nfigplus)];
118 end
119
120 % conf_sig = 0.68;

```



```

121 % sort_idx = round((0.5 + [-conf_sig, conf_sig, 0]/2) * options.K);
122 %
123 % sims_shock_down_conf = normz * sims_shock_sort(:, :, :, sort_idx(1));
124 % sims_shock_up_conf   = normz * sims_shock_sort(:, :, :, sort_idx(2));
125 % sims_shock_median    = normz * sims_shock_sort(:, :, :, sort_idx(3));
126
127 if ndraws > 1
128     sort_idx = round((0.5 + [-conf_sig, conf_sig, 0]/2) * ndraws);
129     irf_sort = sort(irfs,4);
130     if normz_yes == 1
131         % normalize the IRF relative to a 100 bpt increase in the first
132         % variable, first horizon, first shock
133         normz = 1/ irf_sort(1, 1, 1, sort_idx(3) );
134     end
135     if sort_idx(1) == 0,
136         sort_idx(1) = 1;
137     end
138     irf_Median = normz * squeeze(irf_sort(:, :, :, sort_idx(3) ));
139     irf_low    = normz * squeeze(irf_sort(:, :, :, sort_idx(1) ));
140     irf_up     = normz * squeeze(irf_sort(:, :, :, sort_idx(2) ));
141     if add_multiple_bands_yes == 1
142         irf_low_low = normz * squeeze(irf_sort(:, :, :, sort_idx_2(1) ));
143         irf_up_up   = normz * squeeze(irf_sort(:, :, :, sort_idx_2(2) ));
144     end
145
146 else
147     irf_Median = normz * irfs;
148     irf_low    = normz * irfs;
149     irf_up     = normz * irfs;
150 end
151
152 jplot = 0;
153 jfig = 0;
154 for sho = 1 : nshocks
155     % figm = figure('Name', ['IRFs of ' deblank(varnames{sho})] );
156     for var= 1: nvar %size(varnames,2)
157
158         if jplot==0,
159             figure('name', ['IRFs_of_' shocksnames{sho}] );
160             jfig=jfig+1;
161         end
162
163         jplot=jplot+1;
164         subplot(nplots(1),nplots(2),jplot)
165
166         if add_multiple_bands_yes == 1
167

```

```

168         h = area([irf_low_low(var,:,sho)',...
169                 irf_low(var,:,sho)' - irf_low_low(var,:,sho)',...
170                 irf_up(var,:,sho)' - irf_low(var,:,sho)',...
171                 irf_up_up(var,:,sho)' - irf_up(var,:,sho)']);%, 'FaceColor
                ', [.85 .85 .85]);
172         set(h(4), 'FaceColor', [.95 .95 .95])
173         set(h(3), 'FaceColor', [.85 .85 .85])
174         set(h(2), 'FaceColor', [.95 .95 .95])
175         set(h(1), 'FaceColor', [1 1 1])
176         set(h, 'linestyle', 'none')
177         hold on
178
179     else
180         h = area([irf_low(var,:,sho)',...
181                 irf_up(var,:,sho)' - irf_low(var,:,sho)']);%, 'FaceColor ', [.85
                .85 .85]);
182         set(h(2), 'FaceColor', [.85 .85 .85])
183         set(h(1), 'FaceColor', [1 1 1])
184         set(h, 'linestyle', 'none')
185     end
186     %         hold on
187     %         plot(irf.cholesky.up(var,:,sho), 'b:');
188     %         hold on;
189     %         plot(irf.cholesky.low(var,:,sho), 'b:');
190     hold on;
191     plot(irf.Median(var,:,sho), 'k');
192     hold on;
193     if add_irfs_yes == 1
194         for hh = 1: size(add_irfs,4)
195             plot(add_irfs(var,:,sho,hh), 'b', 'LineWidth', 2);
196         end
197     end
198     hold on;
199     plot(zeros(1,hor), 'k')
200     hold on;
201     hold on
202     axis tight
203     title(varnames{var}, 'FontSize', fontsize)
204     set(gcf, 'position', [50 50 800 650])
205     if jplot==nbofplots(jfig) || var==length(varnames)
206         %legend(legenda);
207         if savefig_yes == 1
208             STR_RECAP = [ fnam_dir '/' fnam_suffix '-' shocksnames{sho} '-'
                ' int2str(jfig)];
209             saveas(gcf, STR_RECAP, 'fig');
210             saveas(gcf, STR_RECAP, 'eps');
211             %savefigure_pdf(STR_RECAP);

```

```

212             savefigure_pdf([STR_RECAP '.pdf']);
213         end
214         jplot=0;
215     end
216 end
217 jfig = 0;
218 end

```

plot_sdcmp.m

```

1  function plot_sdcmp(input,BVAR,options)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Filippo Ferroni , 6/1/2015
5  % Revised , 2/15/2017
6  % Revised , 3/21/2018
7
8  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
9
10 nshocks = BVAR.N;
11 for v = 1 : nshocks
12     eval(['nameshock{' num2str(v) '}_=_','Shck' num2str(v) ',';'])
13 end
14 for v = 1 : BVAR.N
15     eval(['pplotvar{' num2str(v) '}_=_','Var' num2str(v) ',';'])
16 end
17 for v = 1 : BVAR.N
18     eval(['ex_names_{' num2str(v) '}_=_{'','Shck' num2str(v) ',';'])
19 end
20 ex_names_ = ex_names_';
21 for v = 1 : BVAR.N
22     eval(['leg{' num2str(v) '}_=_','Shck' num2str(v) ',';'])
23 end
24 leg{end+1} = 'initial_condition';
25
26 TT = 1:1:size(input,1);
27 Tlim = [TT(1) TT(end)];
28
29 [~,positions] = ismember(pplotvar,BVAR.varnames);
30 dcmp_group_yes = 0;
31 dmcp_type = 'stacked';
32 tags = [];
33 colors_decomp_yes = 0;
34 savefig_yes = 0;
35 % initial_state_dcmp = 0;

```

```

36 % addplot_ = 0;
37 % addplot0_ = 0;
38
39 if nargin > 2
40     if isfield(options,'plotvar_') == 1
41         clear plotvar_
42         pplotvar          = options.plotvar_;
43     end
44     if isfield(options,'snames_') == 1
45         clear ex_names_
46         ex_names_          = options.snames_;
47     end
48     if isfield(options,'stag_') == 1
49         clear leg
50         leg                 = options.stag_;
51     end
52     if isfield(options,'snames_') == 1 && isfield(options,'stag_') == 0
53         error('You need to provide also ''stag_''')
54     end
55 %     if isfield(options,'snames_') == 0 && isfield(options,'stag_') == 1
56 %         error('You need to provide also ''snames_''')
57 %     end
58     if size(ex_names_,1)+1 ~= length(leg)
59         error('Mismatch between shock aggregations and shocks names')
60     end
61     if isfield(options,'time') == 1
62         TT = options.time;
63         if isfield(options,'Tlim') == 1;
64             Tlim=options.Tlim;
65             if Tlim(1) < TT(1)
66                 warning('You have set a initial date that starts earlier than
67                     _the first obs');
68                 warning('I change it with the frist obs');
69                 Tlim(1) = TT(1);
70             end
71             if Tlim(2) > TT(end)
72                 warning('You have set a final date that exceeds the forecast _
73                     horizon');
74                 warning('I change it with the endo of forecast');
75                 Tlim(2) = TT(end);
76             end
77         end
78     end
79     if isfield(options,'save_strng') == 1
80         tags = options.save_strng;

```

```

81     end
82     if isfield(options,'dcmp_grouped')==1 && options.dcmp_grouped ==1
83         dcmp_group_yes = 1;
84         dmcpl_type      = 'grouped';
85     end
86     if isfield(options,'plotvarnames')==1
87         pplotvarname = options.plotvarnames;
88         if length(pplotvarname)~=length(pplotvar)
89             error('The number of plot titles (pplotvarname) and of plot
          variables needs to coincide')
90         end
91     end
92     if isfield(options,'colors_decomp')==1 && options.colors_decomp==1
93         colors_decomp_yes =1;
94     end
95     if isfield(options,'colors_decomp')==1 && options.colors_decomp==2
96         colors_decomp_yes =2;
97     end
98     if isfield(options,'colors_decomp')==1 && options.colors_decomp==3
99         colors_decomp_yes =3;
100    end
101    if isfield(options,'saveas_dir')==1
102        savefig_yes = 1;
103        % setting the folder where to save the figure
104        fnam_dir = options.saveas_dir;
105        if exist(fnam_dir,'dir')== 0
106            mkdir(fnam_dir)
107        end
108    end
109
110    % if isfield(options,'addplots_yes')==1 && options.addplots_yes==1,
111    %     tags = [ tags 'memo'];
112    % if isfield(input,'frcsts')==1
113    %     addplot_=1;
114    %     frcsts = input.frcsts.states.Mean(:,positions);
115    %     s_s = input.frcsts.states.steady(positions);
116    % end
117    % if isfield(input,'frcsts0')==1
118    %     addplot0_=1;
119    %     frcsts0 = input.frcsts0.states.Mean(:,positions);
120    %
121    % end
122    % end
123 end
124
125 [~,positions] = ismember(pplotvar,BVAR.varnames);
126 pplotvarname = pplotvar;

```

```

127
128 deco = input;
129
130 % setting the names of the figure to save
131 fnam_suffix = [tags '_shcks_dcmp'];
132
133 ngroups0 = size(ex_names_,1);
134 % ngroups = ngroups0+1+no_initial_effect;
135 if colors_decomp_yes == 0
136     func = @(x) colorspace('RGB->Lab',x);
137     MAP = distinguishable_colors(ngroups0+1,'w',func);
138     % MAP = CreateColorMap(ngroups0+1);
139     MAP(end,:) = [0.7 0.7 0.7];
140 elseif colors_decomp_yes == 1
141     MAP = zeros(4,3);
142     MAP(end,:) = [0.7 0.7 0.7]; % gray
143     MAP(end-1,:) = [0.2 0.5 0.99]; % blue
144     MAP(end-2,:) = [1 1 0]; % yellow
145     MAP(end-3,:) = [0.8 0 0.8]; % purple
146     % MAP(end-4,:) = [0 0.7 0]; % green
147     % MAP(end-5,:) = [1 0 0]; % red
148     % MAP(end-6,:) = [0 0 1]; % blue
149     % MAP(end-7,:) = [.5 .5 0]; % light green
150 elseif colors_decomp_yes == 2
151     MAP = zeros(10,3);
152     MAP(end,:) = [0.7 0.7 0.7]; % gray
153     MAP(end-1,:) = [0.2 0.5 0.99]; % blue
154     MAP(end-2,:) = [0.8 0 0.8]; % purple
155     MAP(end-3,:) = [1 1 0]; % yellow
156     MAP(end-4,:) = [0 0.7 0]; % green
157     MAP(end-5,:) = [1 0 0]; % red
158     MAP(end-6,:) = [0 0 1]; % blue
159     MAP(end-7,:) = [.5 .5 0]; % light green
160     MAP(end-8,:) = [0 0.25 0]; %brown
161     MAP(end-9,:) = [0 0.9 0.9]; % violet
162 elseif colors_decomp_yes == 3
163     MAP = zeros(11,3);
164     MAP(end,:) = [0.7 0.7 0.7]; % gray
165     MAP(end-1,:) = [0.2 0.5 0.99]; % blue
166     MAP(end-2,:) = [0.8 0 0.8]; % purple
167     MAP(end-3,:) = [1 1 0]; % yellow
168     MAP(end-4,:) = [0 0.7 0]; % green
169     MAP(end-5,:) = [1 0 0]; % red
170     MAP(end-6,:) = [0 0 1]; % blue
171     MAP(end-7,:) = [.5 .5 0]; % light green
172     MAP(end-8,:) = [0 0.25 0]; %brown
173     MAP(end-9,:) = [0 0.9 0.9]; % violet

```

```

174     MAP(end-10,:) = [0.5 0.1 0.1]; % violet
175 end
176
177 st = find(Tlim(1)==TT);
178 en = find(Tlim(2)==TT);
179 if savefig_yes == 1
180     fidTxt = fopen([fnam_dir '\legenda_' fnam_suffix '_plots.txt'],'w');
181     fprintf(fidTxt,['LEGENDA_' tags '_SHOCK_DECOMPOSITION_PLOTS\n']);
182     fprintf(fidTxt,['\n']);
183 end
184
185 for j = 1 : size(pplotvar,2)
186     clear sdec sdec_tot,
187     indx = positions(j);
188     sdec0 = squeeze(deco(:,indx,:));
189     for i=1:ngroups0
190         clear index,
191         for ii=1:size(ex_names_{i},2)
192             indbuf = strmatch(ex_names_{i}{ii},namesshock,'exact');
193             if ~isempty(indbuf)
194                 index(ii) = indbuf;
195             elseif ~isempty(ex_names_{i}{ii})
196                 error(['Shock_name_',ex_names_{i}{ii}, '_not_found.']);
197             end
198         end
199         sdec(:,i)=sum(sdec0(:,index),2);
200         sdec0(:,index)=0;
201     end
202
203 h= figure('Name',['Shocks_Decomposition_for_' pplotvarname{j}]);
204 sdec_tot=[sdec, sum(sdec0,2)];
205 if dcmp-group_yes == 0
206     ind_pos = (sdec_tot>0);
207     ind_neg = (sdec_tot<0);
208     temp_neg = cumsum(sdec_tot.*ind_neg,2).*ind_neg;
209     temp_pos = cumsum(sdec_tot.*ind_pos,2).*ind_pos;
210     temp = temp_neg + temp_pos;
211     for kk = size(temp,2) : -1 : 1
212         hold on
213         bbar = bar(TT(st:en),temp(st:en,kk),dmcp_type,'EdgeColor',[0 0 0])
214             ;
215         set(bbar, 'FaceColor', MAP(kk,:))
216         shading faceted; hold on;
217     end
218     leg0 = leg(end:-1:1);
219 else

```

```

220         temp          = sdec_tot;
221         bbar = bar(TT(st:en),temp(st:en,:),dmcp_type,'EdgeColor',[0 0 0]);
                colormap(MAP);
222         %             hleg=legend(leg(1:1:end),'interpreter','none','location','
                Best ');
223         %             shading faceted;
224         %
225     end
226     %         set(hleg,'position',[0.5 0.15 0.4 0.2],'units','normalized')
227     hold on
228     axis tight
229     fillips = sum(sdec_tot,2);
230     hold on, h1=plot(TT(st:en),fillips(st:en),'k-d');
231     set(h1,'MarkerFaceColor','k')
232
233     %         if addplot_ ==1
234     %             hold on, h1=plot(TT(st:en),frcsts(st:en,j),'k-*','LineWidth',2);
235     %             hold on, h1=plot(TT(st:en),s_s(j)*ones(length(TT(st:en)),1),'k
    -.','LineWidth',2);
236     %             leg0{end+1} = 'Current Forecast';
237     %             leg0{end+1} = 'steady state';
238     %         end
239     %
240     %         if addplot0_ ==1
241     %             hold on, h1=plot(TT(st:en),frcsts0(st:en,j),'r-o','LineWidth',2)
    ;
242     %             leg0{end+1} = 'Previous Forecast';
243     %         end
244
245     set(gcf,'position',[50 50 800 650])
246     hleg=legend(leg0,'interpreter','none','location','Best');
247     shading faceted;
248
249     %         define_.timestart + (define_.nobs-1)/4
250     %plot the last in+sample obs
251     %         hold on; plot([TT(define_.nobs) TT(define_.nobs)],[low up],'color
    ',[1 0 0],'LineWidth',2)
252     %         h=vline(define_.timestart + (define_.nobs-1)/4, 'k', 'Last Obs');
253     %         set(h,'Linewidth',2)
254
255     title(pplotvarname{j})
256     %         set(gca,'Xtick',TT(st:6:en))
257     %         tmp_str= sample2date(TT(st:6:en));
258     set(gca,'Xtick',TT(st:6:en))
259     %         tmp_str= sample2date(TT(st:2:en));
260     %         set(gca,'Xticklabel',tmp_str)
261     %         tmp_str= sample2date(TT);

```



```

262      %      STR_RECAP = [ 'model_' fname_suffix '_' tmp_str{st} '_' tmp_str{end}
                '_' int2str(j) ];
263
264      if savefig_yes == 1
265          STR_RECAP = [ fname_dir '/svar_' fname_suffix '_' int2str(j) ];
266          saveas(gcf,STR_RECAP,'fig');
267          saveas(gcf,STR_RECAP,'pdf');
268
269          fprintf(fidTxt,['The figure _sdcmp_' fname_suffix '_' int2str(j) ' _
                contains the following variable:\n' ]);
270          tmp = strrep(char(pplotvarname{j} ),'\','_');
271          fprintf(fidTxt,[tmp ';' _]);
272
273          fprintf(fidTxt,['\n' ]);
274          fprintf(fidTxt,['\n' ]);
275          close all
276      end
277
278  end
279
280  if savefig_yes == 1
281      fclose(fidTxt);
282  end
283
284  end

```

quer.m

```

1  function quer(str);
2  % function quer(str);
3  % Query user
4  % if str=='c' Continue or not
5  % if str=='g' close Graphs or not
6  str=upper(str);
7  switch str
8  case 'C'
9
10     button=questdlg('Continue?_Y/N',...
11         'Continue_Program',...
12         'Yes','No','Yes');
13     switch button
14     case 'No'
15         error('Program_execution_stopped');
16     case 'Yes'
17         fprintf('%-30s\n','_');

```

```

18     end
19 case 'G'
20     button=questdlg('Close_All_Graphs_Y/N?', 'Graphs', 'No', 'Yes', 'Yes');
21     switch button
22     case 'No'
23         fprintf('%-30s\n', ' ');
24     case 'Yes'
25         close all;
26     end
27 end

```

rand_inverse_wishart.m

```

1  function G = rand_inverse_wishart(m, v, H_inv_upper_chol)
2
3  % function G = rand_inverse_wishart(m, v, H_inv_upper_chol)
4  % rand_inverse_wishart Pseudo random matrices drawn from an
5  % inverse Wishart distribution
6  % G = rand_inverse_wishart(m, v, H_inv_upper_chol)
7  % Returns an m-by-m matrix drawn from an inverse-Wishart distribution.
8  %
9  % INPUTS:
10 %     m:          dimension of G and H_inv_upper_chol.
11 %     v:          degrees of freedom, greater or equal than m.
12 %     H_inv_chol: upper cholesky decomposition of the inverse of the
13 %                  matrix parameter.
14 %                  The upper cholesky of the inverse is requested here
15 %                  in order to avoid to recompute it at every random draw.
16 %                  H_inv_upper_chol = chol(inv(H))
17 % OUTPUTS:
18 %     G:           $G \sim IW(m, v, H)$  where  $H = inv(H\_inv\_upper\_chol)' * H\_inv\_upper\_chol$ 
19 %                  or, equivalently, using the correspondence between Wishart
20 %                  and
21 %                  inverse-Wishart:  $inv(G) \sim W(m, v, S)$  where
22 %                   $S = H\_inv\_upper\_chol' * H\_inv\_upper\_chol = inv(H)$ 
23 % SPECIAL REQUIREMENT
24 %     none
25 %
26
27 % Copyright (C) 2003–2009 Dynare Team
28 %
29 % This file is part of Dynare.
30 %

```

```

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43
44 X = randn(v, m) * H_inv_upper_chol;
45
46
47 % At this point, X'*X is Wishart distributed
48 % G = inv(X'*X);
49
50 % Rather compute inv(X'*X) using the SVD
51 [U,S,V] = svd(X, 0);
52 SSi = 1 ./ (diag(S) .^ 2);
53 G = (V .* repmat(SSi', m, 1)) * V';

```

reorderVAR.m

```

1 function [Phip,Sigmap] = reorderVAR(Phi,Sigma,reordering)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % Filippo Ferroni, 6/1/2015
5 % Revised, 2/15/2017
6 % Revised, 3/21/2018
7
8 % Permute the autoregressive matrix and the variance covariance of the
9 % shocks with the new ordering of the variables.
10 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
11
12 N = size(Sigma,1);
13
14 if length(reordering) ~= N,
15     error('Not enough permutations: the vector with new ordering of variables _
16         must have size N.')
17 end
18 Phip = Phi(:,reordering);

```

```

19 Sigmap = nan(N);
20
21 for jj = 1 : length(reordering)
22     for hh = 1 : length(reordering)
23
24         tmp = Sigma(reordering(jj),reordering(hh));
25         Sigmap(jj, hh) = tmp;
26     end
27
28 end
29
30 end

```

rescaleFAVAR.m

```

1 function [ReScale] = rescaleFAVAR(STD,Lambda,n_1,order_pc)
2
3 if nargin < 4
4     % principal component ordered first
5     order_pc = 1;
6 end
7
8 n_w = size(Lambda,2);
9 n_2 = size(Lambda,1);
10
11 ReScale= NaN;
12
13 if size(STD,2)>size(STD,1)
14     STD =STD';
15 end
16
17
18 switch order_pc
19     case 1 % factor first
20
21         Scale_ = repmat([STD; ones(n_1,1)], 1, n_w+n_1);
22         Lambda_ = [Lambda zeros(n_2 , n_1); zeros(n_1,n_w+n_1)];
23         Lambda_(n_2+1 : n_2+n_1 , n_w + 1 : n_w+n_1) = eye(n_1);
24         ReScale = Scale_ .* Lambda_ ;
25
26     case 2 % factor second
27
28         Scale_ = repmat([ones(n_1,1); STD], 1, n_w +n_1);
29         Lambda_ = [zeros(n_1,n_1+n_w); zeros(n_2 , n_1) Lambda];
30         Lambda_(1:n_1, 1:n_1) = eye(n_1);

```

```

31         ReScale = Scale_ .* Lambda_ ;
32
33 end

```

rfvar3.m

```

1  function var=rfvar3(ydata,lags,xdata,breaks,lambda,mu)
2  %function var=rfvar3(ydata,lags,xdata,breaks,lambda,mu)
3  % This algorithm goes for accuracy without worrying about memory requirements.
4  % ydata:    dependent variable data matrix
5  % xdata:    exogenous variable data matrix
6  % lags:     number of lags
7  % breaks:   rows in ydata and xdata after which there is a break. This allows
            for
8  %           discontinuities in the data (e.g. war years) and for the
            possibility of
9  %           adding dummy observations to implement a prior. This must be a
            column vector.
10 %           Note that a single dummy observation becomes lags+1 rows of the
            data matrix,
11 %           with a break separating it from the rest of the data. The function
            treats the
12 %           first lags observations at the top and after each "break" in ydata
            and xdata as
13 %           initial conditions.
14 % lambda:   weight on "co-persistence" prior dummy observations. This
            expresses
15 %           belief that when data on *all* y's are stable at their initial
            levels, they will
16 %           tend to persist at that level. lambda=5 is a reasonable first try.
            With lambda<0,
17 %           constant term is not included in the dummy observation, so that
            stationary models
18 %           with means equal to initial ybar do not fit the prior mean. With
            lambda>0, the prior
19 %           implies that large constants are unlikely if unit roots are present
            .
20 % mu:       weight on "own persistence" prior dummy observation. Expresses
            belief
21 %           that when y_i has been stable at its initial level, it will tend to
            persist
22 %           at that level, regardless of the values of other variables. There
            is
23 %           one of these for each variable. A reasonable first guess is mu=2.
24 %           The program assumes that the first lags rows of ydata and xdata are

```

```

    real data , not dummies.
25 %      Dummy observations should go at the end , if any . If pre-sample x's are
    not available ,
26 %      repeating the initial xdata(lags+1,:) row or copying xdata(lags+1:2*
    lags,:) into
27 %      xdata(1:lags,:) are reasonable substitutes . These values are used in
    forming the
28 %      persistence priors .
29
30 % Original file downloaded from :
31 % http://sims.princeton.edu/yftp/VARtools/matlab/rfvar3.m
32
33 [T,nvar] = size(ydata);
34 nox = isempty(xdata);
35 if ~nox
36     [T2,nx] = size(xdata);
37 else
38     T2 = T;
39     nx = 0;
40     xdata = zeros(T2,0);
41 end
42 % note that x must be same length as y, even though first part of x will not
    be used .
43 % This is so that the lags parameter can be changed without reshaping the
    xdata matrix .
44 if T2 ~= T, error('Mismatch_of_x_and_y_data_lengths'),end
45 if nargin < 4
46     nbreaks = 0;
47     breaks = [];
48 else
49     nbreaks = length(breaks);
50 end
51 breaks = [0;breaks;T];
52 smpl = [];
53 for nb = 1:nbreaks+1
54     smpl = [smpl;breaks(nb)+lags+1:breaks(nb+1)]';
55 end
56 Tsmpl = size(smpl,1);
57 X = zeros(Tsmpl,nvar,lags);
58 for is = 1:length(smpl)
59     X(is, :, :) = ydata(smpl(is)-(1:lags),:);
60 end
61 X = [X(:, :) xdata(smpl, :)];
62 y = ydata(smpl, :);
63 % Everything now set up with input data for y=Xb+e
64
65 % Add persistence dummies

```

```

66 if lambda ~= 0 || mu > 0
67     ybar = mean(ydata(1:lags,:),1);
68     if ~nox
69         xbar = mean(xdata(1:lags,:),1);
70     else
71         xbar = [];
72     end
73     if lambda ~= 0
74         if lambda>0
75             xdum = lambda*[ repmat(ybar,1,lags) xbar];
76         else
77             lambda = -lambda;
78             xdum = lambda*[ repmat(ybar,1,lags) zeros(size(xbar))];
79         end
80         ydum = zeros(1,nvar);
81         ydum(1,:) = lambda*ybar;
82         y = [y;ydum];
83         X = [X;xdum];
84     end
85     if mu>0
86         xdum = [ repmat(diag(ybar),1,lags) zeros(nvar,nx)]*mu;
87         ydum = mu*diag(ybar);
88         X = [X;xdum];
89         y = [y;ydum];
90     end
91 end
92
93 % Compute OLS regression and residuals
94 [v1,d,vr] = svd(X,0);
95 di = 1./diag(d);
96 B = (vr.*repmat(di',nvar*lags+nx,1))*v1'*y;
97 u = y-X*B;
98 xxi = vr.*repmat(di',nvar*lags+nx,1);
99 xxi = xxi*xxi';
100
101 var.B = B;
102 var.u = u;
103 var.xxi = xxi;
104 var.y = y;
105 var.X = X;

```

savefigure_pdf.m

```

1 function [ figformat ] = savefigure_pdf( name,varargin )
2

```

```

3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Filippo Ferroni , 6/1/2015
5  % Revised , 2/15/2017
6  % Revised , 3/21/2018
7
8  % SAVEFIGURE.PDF: this function saves figures in .pdf format
9  % name: Name of the figure , example: 'figure1 '
10 % type: Format of the figure , example: 'fig ', 'espc '
11 % If no input is provided , figure saved twice with fig and espc
12 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
13
14 set(gcf,'Units','Inches');
15 pos = get(gcf,'Position');
16 set(gcf,'PaperPositionMode','Auto','PaperUnits','Inches','PaperSize',[pos(3),
    pos(4)])
17 print(gcf,name,'-dpdf','-r0');
18 figformat=1;
19 end

```

shade.m

```

1  function []=shade(start,finish,colorstr,up,low);
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % Filippo Ferroni , 6/1/2015
5  % Revised , 2/15/2017
6  % Revised , 3/21/2018
7
8  % function []=shade(start , finish , colorstr );
9  %
10 % start and finish are Nx1 vectors of starting and ending years .
11 % The function shades between the start and finish pairs using colorstr
12 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
13
14 if ~exist('colorstr'); colorstr='y'; end; % default is yellow
15 curax=axis;
16 y=[curax(3)+low up*curax(4) up*curax(4) curax(3)+low];
17 hold on;
18 for i=1:length(start);
19     x=[start(i) start(i) finish(i) finish(i)];
20     fill(x,y,colorstr);
21 end;
22
23 % Now, prevent the shading from covering up the lines in the plot .
24 h = findobj(gca,'Type','line');

```



```

25 set(h,'EraseMode','xor');
26
27 h = findobj(gca,'Type','patch');
28 set(h,'EdgeColor','none');
29
30 % This last one makes the tick marks visible
31 set(gca, 'Layer', 'top')

```

sign2matrix.m

```

1 function [f,sr] = sign2matrix(signs,ny)
2
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 % Filippo Ferroni , 6/1/2015
5 % Revised , 2/15/2017
6 % Revised , 3/21/2018
7
8 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
9
10 yr = ones(ny);
11 ys = ones(ny);
12 y = nan(ny);
13 for ell = 1 : length(signs)
14     eval(signs{ell})
15 end
16
17 f(1:ny , 1:ny) = ys;
18 f(1+ny:2*ny , 1:ny) = yr;
19 sr = y;

```

standard.m

```

1 function [x,Scale]=standard(y)
2 T=size(y,1);
3 my= repmat(nanmean(y),T,1);
4 sy= repmat(nanstd(y),T,1);
5 x=(y-my)./sy;
6 x(find(isnan(x)))=randn;
7 Scale = nanstd(y)';
8 %x=(y-kron(mean(y),ones(rows(y),1)))./kron(std(y),ones(rows(y),1));

```

var2ss.m

```

1  function [A,B,C,const,Sigma,lags,index_var]=var2ss(Phi,Sigma,index)
2
3  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4  % var2ss converts the VAR into a state space system of this form
5  %  $x(t) = A x(t-1) + B \Sigma' u(t) \sim N(0, I)$ 
6  %  $y(t) = C*(cons + x(t-1))$ 
7  % where A is the companion form of the lag structure
8
9  % Filippo Ferroni, 6/1/2015
10 % Revised, 2/15/2017
11 % Revised, 3/21/2018
12 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
13
14 if nargin < 3
15     %% all stocks
16     index = zeros(size(Sigma,1),1);
17 end
18
19 Sigma = chol(Sigma);
20
21 N      = size(Sigma,1);
22 [m , n] = size(Phi);
23 lags    = (m-1)/n;
24
25 % companion form
26 A      = [Phi(1 : N * lags, :)'; eye(N*(lags-1), N*lags)];
27 B      = eye(N * lags, N);
28 C      = eye(N, N * lags);
29
30 IminusAlags = eye(N);
31 for ell = 1 : lags
32     IminusAlags = IminusAlags - Phi(N * (ell - 1) + 1 : N * ell, :)';
33 end
34 iIminusAlags = inv(IminusAlags);
35
36 const    = [iIminusAlags*Phi(end, :)'; zeros(N*(lags-1), 1)];
37
38 % index =0 % stock:       $xq(t) = xm(t)$ 
39 % index =1 % TBA
40 % index =2 % deflator/real flow:  $xq(t) = 1/3( xm(t) + xm(t-1) + xm(t-2))$ 
41
42 index_var = 1 : N;
43
44 for vv = 1 : N

```

```

45     if index(vv) == 1
46 %         C(vv,vv : N : N * 3) = 1;
47     elseif index(vv) == 2 %
48         if lags < 2
49             error('When you specify the flow/deflation aggregation, you need
                    at least 2 lags')
50         end
51         A = [A zeros(size(A,1),1)];
52         A = [A; zeros(1,size(A,2))];
53         A(end,vv : N : N * 2) = 1/3;
54         Ao = eye(length(A));
55         Ao(end,vv) = -1/3;
56         iAo = inv(Ao);
57         A = iAo * A;
58
59         B = [B; zeros(1,size(B,2))];
60         B = iAo * B;
61
62         C = [C zeros(N,1)];
63         C(vv,vv) = 0;
64         C(vv,end) = 1;
65
66         const(size(A,1)) = const(vv);
67
68         index_var(vv) = size(A,1);
69
70     elseif index(vv) == 4 %
71         if lags < 3
72             error('When you specify the flow/deflation aggregation (weekly-
                    monthly or quarterly-annual), you need at least 3 lags')
73         end
74         A = [A zeros(size(A,1),1)];
75         A = [A; zeros(1,size(A,2))];
76         A(end,vv : N : N * 3) = 1/4;
77         Ao = eye(length(A));
78         Ao(end,vv) = -1/4;
79         iAo = inv(Ao);
80         A = iAo * A;
81
82         B = [B; zeros(1,size(B,2))];
83         B = iAo * B;
84
85         C = [C zeros(N,1)];
86         C(vv,vv) = 0;
87         C(vv,end) = 1;
88
89         const(size(A,1)) = const(vv);

```

```

90
91         index_var(vv) = size(A,1);
92
93     end
94
95 end

```

varprior.m

```

1  function [ydum,xdum,breaks]=varprior(nv,nx,lags,mnprior,vprior)
2  %function [ydum,xdum,breaks]=varprior(nv,nx,lags,mnprior,vprior)
3  % ydum, xdum:    dummy observation data that implement the prior
4  % breaks:       vector of points in the dummy data after which new dummy obs's
                    start
5  %
                    Set breaks=T+[0;breaks], ydata=[ydata;ydum], xdum=[xdata;
                    xdum], where
6  %
                    actual data matrix has T rows, in preparing input for
                    rfvar3
7  % nv,nx,lags: VAR dimensions
8  % mnprior.tight: Overall tightness of Minnesota prior
9  % mnprior.decay: Standard deviations of lags shrink as lag^(-decay)
10 % vprior.sig:    Vector of prior modes for diagonal elements of r.f. covariance
                    matrix
11 % vprior.w:      Weight on prior on vcv. 1 corresponds to "one dummy
                    observation" weight
12 %
                    Should be an integer, and will be rounded if not. vprior.
                    sig is needed
13 %
                    to scale the Minnesota prior, even if the prior on sigma
                    is not used itself.
14 %
                    Set vprior.w=0 to achieve this.
15 % Note:         The original Minnesota prior treats own lags asymmetrically,
                    and therefore
16 %
                    cannot be implemented entirely with dummy observations.
                    It is also usually
17 %
                    taken to include the sum-of-coefficients and co-
                    persistence components
18 %
                    that are implemented directly in rfvar3.m. The diagonal
                    prior on v, combined
19 %
                    with sum-of-coefficients and co-persistence components and
                    with the unit own-first-lag
20 %
                    prior mean generates larger prior variances for own than
                    for cross-effects even in
21 %
                    this formulation, but here there is no way to shrink
                    toward a set of unconstrained
22 %
                    univariate AR's.

```

```

23
24 % Original file downloaded from:
25 % http://sims.princeton.edu/yftp/VARtools/matlab/varprior.m
26
27 if ~isempty(mnprior)
28     xdum = zeros(lags+1,nx,lags,nv);
29     ydum = zeros(lags+1,nv,lags,nv);
30     for il = 1:lags
31         ydum(il+1,:,il,:) = il^mnprior.decay*diag(vprior.sig);
32     end
33     ydum(1,:,1,:) = diag(vprior.sig);
34     ydum = mnprior.tight*reshape(ydum,[lags+1,nv,lags*nv]);
35     ydum = flipdim(ydum,1);
36     xdum = mnprior.tight*reshape(xdum,[lags+1,nx,lags*nv]);
37     xdum = flipdim(xdum,1);
38     breaks = (lags+1)*[1:(nv*lags)]';
39     lbreak = breaks(end);
40 else
41     ydum = [];
42     xdum = [];
43     breaks = [];
44     lbreak = 0;
45 end
46 % elle = 0;
47 % for j1 = 1 : nv
48 %     for j2 = 1 : lags
49 %         elle = 1 + elle;
50 %         ydum(:,j1,j2,elle) = ydum(:,j1,j2,elle) * mnprior.unit_root_(j1);
51 %     end
52 % end
53 if ~isempty(vprior) && vprior.w>0
54     ydum2 = zeros(lags+1,nv,nv);
55     xdum2 = zeros(lags+1,nx,nv);
56     ydum2(end,:,:) = diag(vprior.sig);
57     for i = 1:vprior.w
58         ydum = cat(3,ydum,ydum2);
59         xdum = cat(3,xdum,xdum2);
60         breaks = [breaks;(lags+1)*[1:nv]'+lbreak];
61         lbreak = breaks(end);
62     end
63 end
64 dimy = size(ydum);
65 ydum = reshape(permute(ydum,[1 3 2]),dimy(1)*dimy(3),nv);
66 xdum = reshape(permute(xdum,[1 3 2]),dimy(1)*dimy(3),nx);
67 breaks = breaks(1:(end-1));

```

cmintools/bfgsi.m

```

1  function H = bfgsi(H0,dg,dx)
2  % H = bfgsi(H0,dg,dx)
3  % dg is previous change in gradient; dx is previous change in x;
4  % 6/8/93 version that updates inverse hessian instead of hessian
5  % itself.
6  % Copyright by Christopher Sims 1996. This material may be freely
7  % reproduced and modified.
8  if size(dg,2)>1
9      dg=dg';
10 end
11 if size(dx,2)>1
12     dx=dx';
13 end
14 Hdg = H0*dg;
15 dgdx = dg'*dx;
16 if (abs(dgdx) > 1e-12)
17     H = H0 + (1+(dg'*Hdg)/dgdx)*(dx*dx')/dgdx - (dx*Hdg'+Hdg*dx')/dgdx;
18 else
19     disp('bfgs_update_failed.')
20     disp(['|dg|_=' num2str(sqrt(dg'*dg)) ' |dx|_=' num2str(sqrt(dx'*dx))]);
21     disp(['dg'*dx_=' num2str(dgdx)])
22     disp(['H*dg|_=' num2str(Hdg'*Hdg)])
23     H=H0;
24 end
25 save H.dat H

```

cmintools/csminit.m

```

1  function [fhat,xhat,fcount,retcode] = csminit(fcn,x0,f0,g0,badg,H0,varargin)
2  % [fhat,xhat,fcount,retcode] = csminit(fcn,x0,f0,g0,badg,H0,...
3  %                                     P1,P2,P3,P4,P5,P6,P7,P8)
4  % retcodes: 0, normal step. 5, largest step still improves too fast.
5  % 4,2 back and forth adjustment of stepsize didn't finish. 3, smallest
6  % stepsize still improves too slow. 6, no improvement found. 1, zero
7  % gradient.
8  %-----
9  % Modified 7/22/96 to omit variable-length P list, for efficiency and
10 % compilation.
11 % Places where the number of P's need to be altered or the code could be
12 % returned to
13 % its old form are marked with ARGLIST comments.
14 %
15 % Fixed 7/17/93 to use inverse-hessian instead of hessian itself in bfgs

```

```

14 % update .
15 %
16 % Fixed 7/19/93 to flip eigenvalues of H to get better performance when
17 % it 's not psd .
18 %
19 %tailstr = ') ' ;
20 %for i=nargin-6:-1:1
21 %   tailstr=[ ' ,P' num2str(i)   tailstr ] ;
22 %end
23 %ANGLE = .03 ;
24 ANGLE = .005 ;
25 %THETA = .03 ;
26 THETA = .3 ; % (0<THETA<.5) THETA near .5 makes long line searches , possibly
    fewer iterations .
27 FCHANGE = 1000 ;
28 MINLAMB = 1e-9 ;
29 % fixed 7/15/94
30 % MINDX = .0001 ;
31 % MINDX = 1e-6 ;
32 MINDFAC = .01 ;
33 fcount=0 ;
34 lambda=1 ;
35 xhat=x0 ;
36 f=f0 ;
37 fhat=f0 ;
38 g = g0 ;
39 gnorm = norm(g) ;
40 %
41 if (gnorm < 1.e-12) & ~badg % put ~badg 8/4/94
42     retcode =1 ;
43     dxnorm=0 ;
44     % gradient convergence
45 else
46     % with badg true , we don't try to match rate of improvement to directional
47     % derivative . We're satisfied just to get some improvement in f .
48     %
49     %if ( badg )
50     %   dx = -g*FCHANGE/(gnorm*gnorm) ;
51     %   dxnorm = norm(dx) ;
52     %   if dxnorm > 1e12
53     %       disp('Bad , small gradient problem .')
54     %       dx = dx*FCHANGE/dxnorm ;
55     %   end
56 %else
57 % Gauss-Newton step ;
58 %----- Start of 7/19/93 mod -----
59 % [v d] = eig(H0) ;

```

```

60      %toc
61      %d=max(1e-10,abs(diag(d)));
62      %d=abs(diag(d));
63      %dx = -(v.*(ones(size(v,1),1)*d'))*(v'*g);
64      %      toc
65      dx = -H0*g;
66      %      toc
67      dxnorm = norm(dx);
68      if dxnorm > 1e12
69          disp('Near-singular H-problem.')
70          dx = dx*FCHANGE/dxnrm;
71      end
72      dfhat = dx'*g0;
73      %end
74      %
75      %
76      if ~badg
77          % test for alignment of dx with gradient and fix if necessary
78          a = -dfhat/(gnorm*dxnorm);
79          if a<ANGLE
80              dx = dx - (ANGLE*dxnorm/gnorm+dfhat/(gnorm*gnorm))*g;
81              % suggested alternate code:
82              dx = dx*dxnorm/norm(dx)      % This keeps scale invariant to the angle
              correction
83              %
84              dfhat = dx'*g;
85              % dxnorm = norm(dx); % this line unnecessary with modification that
              keeps scale invariant
86              disp(sprintf('Correct_for_low_angle:_%g',a))
87          end
88      end
89      disp(sprintf('Predicted_improvement:_%18.9f',-dfhat/2))
90      %
91      % Have OK dx, now adjust length of step (lambda) until min and
92      % max improvement rate criteria are met.
93      done=0;
94      factor=3;
95      shrink=1;
96      lambdaMin=0;
97      lambdaMax=inf;
98      lambdaPeak=0;
99      fPeak=f0;
100     lambdahat=0;
101     while ~done
102         if size(x0,2)>1
103             dxtest=x0+dx'*lambda;
104         else

```



```

105         dxtest=x0+dx*lambda;
106     end
107     % home
108     f = feval(fcn,dxtest,varargin{:});
109     %ARGLIST
110     %f = feval(fcn,dxtest,P1,P2,P3,P4,P5,P6,P7,P8,P9,P10,P11,P12,P13);
111     % f = feval(fcn,x0+dx*lambda,P1,P2,P3,P4,P5,P6,P7,P8);
112     disp(sprintf('lambda=%10.5g; f=%20.7f',lambda,f))
113     %debug
114     %disp(sprintf('Improvement too great? f0-f: %g, criterion: %g',f0-f, -(1-
115         THETA)*dfhat*lambda))
116     if f<fhat
117         fhat=f;
118         xhat=dxtest;
119         lambdahat = lambda;
120     end
121     fcount=fcount+1;
122     shrinkSignal = (~badg & (f0-f < max([-THETA*dfhat*lambda 0]))) | (badg &
123         (f0-f) < 0) ;
124     growSignal = ~badg & ((lambda > 0) & (f0-f > -(1-THETA)*dfhat*lambda)
125         );
126     if shrinkSignal & ((lambda>lambdaPeak) | (lambda<0) )
127         if (lambda>0) & ((~shrink) | (lambda/factor <= lambdaPeak))
128             shrink=1;
129             factor=factor^.6;
130             while lambda/factor <= lambdaPeak
131                 factor=factor^.6;
132             end
133             %if (abs(lambda)*(factor-1)*dxnorm < MINDX) | (abs(lambda)*(factor
134                 -1) < MINLAMB)
135             if abs(factor-1)<MINDFAC
136                 if abs(lambda)<4
137                     retcode=2;
138                 else
139                     retcode=7;
140                 end
141                 done=1;
142             end
143         end
144     end
145     if (lambda<lambdaMax) & (lambda>lambdaPeak)
146         lambdaMax=lambda;
147     end
148     lambda=lambda/factor;
149     if abs(lambda) < MINLAMB
150         if (lambda > 0) & (f0 <= fhat)
151             % try going against gradient, which may be inaccurate
152             lambda = -lambda*factor^6

```

```

148         else
149             if lambda < 0
150                 retcode = 6;
151             else
152                 retcode = 3;
153             end
154             done = 1;
155         end
156     end
157 elseif (growSignal & lambda>0) | (shrinkSignal & ((lambda <=
    lambdaPeak) & (lambda>0)))
158     if shrink
159         shrink=0;
160         factor = factor^.6;
161         %if ( abs(lambda)*(factor-1)*dxnorm< MINDX ) | ( abs(lambda)*(
            factor-1)< MINLAMB)
162         if abs(factor-1)<MINDFAC
163             if abs(lambda)<4
164                 retcode=4;
165             else
166                 retcode=7;
167             end
168             done=1;
169         end
170     end
171     if ( f<fPeak ) & (lambda>0)
172         fPeak=f;
173         lambdaPeak=lambda;
174         if lambdaMax<=lambdaPeak
175             lambdaMax=lambdaPeak*factor*factor;
176         end
177     end
178     lambda=lambda*factor;
179     if abs(lambda) > 1e20;
180         retcode = 5;
181         done =1;
182     end
183 else
184     done=1;
185     if factor < 1.2
186         retcode=7;
187     else
188         retcode=0;
189     end
190 end
191 end
192 end

```

```
193 disp(sprintf('Norm of dx_%10.5g', dxnorm))
```

cmintools/csminwel.m

```
1 function [fh,xh,gh,H,itct,fcount,retcodeh] = csminwel(fcn,x0,H0,grad,crit,nit,
    varargin)
2 %[fhat,xhat,ghat,Hhat,itct,fcount,retcodehat] = csminwel(fcn,x0,H0,grad,crit,
    nit,varargin)
3 % fcn: string naming the objective function to be minimized
4 % x0: initial value of the parameter vector
5 % H0: initial value for the inverse Hessian. Must be positive definite.
6 % grad: Either a string naming a function that calculates the gradient, or
    the null matrix.
7 % If it's null, the program calculates a numerical gradient. In this
    case fcn must
8 % be written so that it can take a matrix argument and produce a row
    vector of values.
9 % crit: Convergence criterion. Iteration will cease when it proves
    impossible to improve the
10 % function value by more than crit.
11 % nit: Maximum number of iterations.
12 % varargin: A list of optional length of additional parameters that get handed
    off to fcn each
13 % time it is called.
14 % Note that if the program ends abnormally, it is possible to retrieve
    the current x,
15 % f, and H from the files g1.mat and H.mat that are written at each
    iteration and at each
16 % hessian update, respectively. (When the routine hits certain kinds
    of difficulty, it
17 % write g2.mat and g3.mat as well. If all were written at about the
    same time, any of them
18 % may be a decent starting point. One can also start from the one with
    best function value.)
19 [nx,no]=size(x0);
20 nx=max(nx,no);
21 Verbose=1;
22 NumGrad= isempty(grad);
23 done=0;
24 itct=0;
25 fcount=0;
26 snit=100;
27 %tailstr = ')';
28 %stailstr = [];
29 % Lines below make the number of Pi's optional. This is inefficient, though,
```

```

        and precludes
30 % use of the matlab compiler. Without them, we use feval and the number of Pi
    's must be
31 % changed with the editor for each application. Places where this is required
    are marked
32 % with ARGLIST comments
33 %for i=nargin-6:-1:1
34 %     tailstr=[ ' ,P' num2str(i)   tailstr ];
35 %     stailstr=[' P' num2str(i) stailstr ];
36 %end
37 f0 = feval(fcn,x0,varargin{:});
38 %ARGLIST
39 %f0 = feval(fcn,x0,P1,P2,P3,P4,P5,P6,P7,P8,P9,P10,P11,P12,P13);
40 % disp('first fcn in csminwel.m -----') % Jinill on 9/5/95
41 if f0 > 1e50, disp('Bad_initial_parameter.'), return, end
42 if NumGrad
43     if length(grad)==0
44         [g badg] = numgrad(fcn,x0, varargin{:});
45         %ARGLIST
46         %[g badg] = numgrad(fcn,x0,P1,P2,P3,P4,P5,P6,P7,P8,P9,P10,P11,P12,P13);
47     else
48         badg=any(find(grad==0));
49         g=grad;
50     end
51     %numgrad(fcn,x0,P1,P2,P3,P4);
52 else
53     [g badg] = feval(grad,x0,varargin{:});
54     %ARGLIST
55     %[g badg] = feval(grad,x0,P1,P2,P3,P4,P5,P6,P7,P8,P9,P10,P11,P12,P13);
56 end
57 retcode3=101;
58 x=x0;
59 f=f0;
60 H=H0;
61 cliff=0;
62 while ~done
63     g1=[]; g2=[]; g3=[];
64     %addition fj. 7/6/94 for control
65     disp('-----')
66     disp('-----')
67     %disp('f and x at the beginning of new iteration ')
68     disp(sprintf('f at the beginning of new iteration , %20.10f',f))
69     %-----Comment out this line if the x vector is long-----
70     disp([sprintf('x = ') sprintf('%15.8g_%15.8g_%15.8g_%15.8g\n',x)]);
71     %-----
72     itct=itct+1;
73     [f1 x1 fc retcode1] = csminit(fcn,x,f,g,badg,H,varargin{:});

```

```

74  %ARGLIST
75  %[ f1 x1 fc retcode1] = csminit(fcn,x,f,g,badg,H,P1,P2,P3,P4,P5,P6,P7,...
76  %      P8,P9,P10,P11,P12,P13);
77  % itct=itct+1;
78  fcount = fcount+fc;
79  % erased on 8/4/94
80  % if (retcode == 1) | (abs(f1-f) < crit)
81  %     done=1;
82  % end
83  % if itct > nit
84  %     done = 1;
85  %     retcode = -retcode;
86  % end
87  if retcode1 ~= 1
88      if retcode1==2 | retcode1==4
89          wall1=1; badg1=1;
90      else
91          if NumGrad
92              [g1 badg1] = numgrad(fcn, x1,varargin{:});
93              %ARGLIST
94              %[ g1 badg1] = numgrad(fcn, x1,P1,P2,P3,P4,P5,P6,P7,P8,P9,...
95              %      P10,P11,P12,P13);
96          else
97              [g1 badg1] = feval(grad,x1,varargin{:});
98              %ARGLIST
99              %[ g1 badg1] = feval(grad, x1,P1,P2,P3,P4,P5,P6,P7,P8,P9,...
100             %      P10,P11,P12,P13);
101          end
102          wall1=badg1;
103          % g1
104          save g1 g1 x1 f1 varargin;
105          %ARGLIST
106          %save g1 g1 x1 f1 P1 P2 P3 P4 P5 P6 P7 P8 P9 P10 P11 P12 P13;
107      end
108      if wall1 & (length(H) > 1)%
109          % Bad gradient or back and forth on step length. Possibly at
110          % cliff edge. Try perturbing search direction if problem not 1D
111          %
112          %fcliff=fh; xcliff=xh;
113          Hcliff=H+diag(diag(H).*rand(nx,1));
114          disp('Cliff...Perturbing search direction.')
115          [f2 x2 fc retcode2] = csminit(fcn,x,f,g,badg,Hcliff,varargin{:});
116          %ARGLIST
117          %[ f2 x2 fc retcode2] = csminit(fcn,x,f,g,badg,Hcliff,P1,P2,P3,P4,...
118          %      P5,P6,P7,P8,P9,P10,P11,P12,P13);
119          fcount = fcount+fc; % put by Jinill
120          if f2 < f

```

```

121     if retcode2==2 | retcode2==4
122         wall2=1; badg2=1;
123     else
124         if NumGrad
125             [g2 badg2] = numgrad(fcn, x2,varargin{:});
126             %ARGLIST
127             %[g2 badg2] = numgrad(fcn, x2,P1,P2,P3,P4,P5,P6,P7,P8,...
128             %           P9,P10,P11,P12,P13);
129         else
130             [g2 badg2] = feval(grad,x2,varargin{:});
131             %ARGLIST
132             %[g2 badg2] = feval(grad,x2,P1,P2,P3,P4,P5,P6,P7,P8,...
133             %           P9,P10,P11,P12,P13);
134         end
135         wall2=badg2;
136         % g2
137         badg2
138         save g2 g2 x2 f2 varargin
139         %ARGLIST
140         %save g2 g2 x2 f2 P1 P2 P3 P4 P5 P6 P7 P8 P9 P10 P11 P12 P13;
141     end
142     if wall2
143         disp('Cliff_lagain...Try_traversing')
144         if norm(x2-x1) < 1e-13
145             f3=f; x3=x; badg3=1;retcode3=101;
146         else
147             gcliff=((f2-f1)/((norm(x2-x1))^2))*(x2-x1);
148             if(size(x0,2)>1), gcliff=gcliff', end
149             [f3 x3 fc retcode3] = csminit(fcn,x,f,gcliff,0,eye(nx),
150             varargin{:});
151             %ARGLIST
152             %[f3 x3 fc retcode3] = csminit(fcn,x,f,gcliff,0,eye(nx),P1,
153             %           P2,P3,...
154             %           P4,P5,P6,P7,P8,...
155             %           P9,P10,P11,P12,P13);
156             fcount = fcount+fc; % put by Jinill
157             if retcode3==2 | retcode3==4
158                 wall3=1; badg3=1;
159             else
160                 if NumGrad
161                     [g3 badg3] = numgrad(fcn, x3,varargin{:});
162                     %ARGLIST
163                     %[g3 badg3] = numgrad(fcn, x3,P1,P2,P3,P4,P5,P6,P7,P8
164                     %           ,...
165                     %           P9,P10,P11,P12,P13);
166                 else
167                     [g3 badg3] = feval(grad,x3,varargin{:});

```

```

165             %ARGLIST
166             % [ g3 badg3 ] = feval ( grad , x3 , P1 , P2 , P3 , P4 , P5 , P6 , P7 , P8
               , ...
167             %                                     P9 , P10 , P11 , P12 , P13 ) ;
168             end
169             wall3=badg3;
170             % g3
171             badg3
172             save g3 g3 x3 f3 varargin;
173             %ARGLIST
174             %save g3 g3 x3 f3 P1 P2 P3 P4 P5 P6 P7 P8 P9 P10 P11 P12
               P13;
175             end
176             end
177             else
178                 f3=f; x3=x; badg3=1; retcode3=101;
179             end
180             else
181                 f3=f; x3=x; badg3=1;retcode3=101;
182             end
183             else
184                 % normal iteration , no walls , or else 1D, or else we're finished here
               .
185                 f2=f; f3=f; badg2=1; badg3=1; retcode2=101; retcode3=101;
186             end
187             else
188                 f2=f; f3=f; f1=f; retcode2=retcode1; retcode3=retcode1;
189             end
190             %how to pick gh and xh
191             if f3 < f - crit & badg3==0
192                 ih=3
193                 fh=f3; xh=x3; gh=g3; badgh=badg3; retcodeh=retcode3;
194             elseif f2 < f - crit & badg2==0
195                 ih=2
196                 fh=f2; xh=x2; gh=g2; badgh=badg2; retcodeh=retcode2;
197             elseif f1 < f - crit & badg1==0
198                 ih=1
199                 fh=f1; xh=x1; gh=g1; badgh=badg1; retcodeh=retcode1;
200             else
201                 [fh,ih] = min([f1,f2,f3]);
202                 disp(sprintf(' ih=%d',ih))
203                 %eval ([ ' xh=x ' num2str ( ih ) ' ; ' ])
204                 switch ih
205                     case 1
206                         xh=x1;
207                     case 2
208                         xh=x2;

```

```

209         case 3
210             xh=x3;
211         end %case
212         %eval(['gh=g' num2str(ih) ''])
213         %eval(['retcodeh=retcode' num2str(ih) ''])
214         retcodei=[retcode1,retcode2,retcode3];
215         retcodeh=retcodei(ih);
216         if exist('gh')
217             nogh isempty(gh);
218         else
219             nogh=1;
220         end
221         if nogh
222             if NumGrad
223                 [gh badgh] = numgrad(fcn,xh,varargin{:});
224             else
225                 [gh badgh] = feval(grad, xh,varargin{:});
226             end
227         end
228         badgh=1;
229     end
230     %end of picking
231     %ih
232     %fh
233     %xh
234     %gh
235     %badgh
236     stuck = (abs(fh-f) < crit);
237     if (~badg)&(~badgh)&(~stuck)
238         H = bfgsi(H,gh-g,xh-x);
239     end
240     if Verbose
241         disp('———')
242         disp(sprintf('Improvement_on_iteration_%d=_%18.9f',itct,f-fh))
243     end
244     % if Verbose
245     if itct > nit
246         disp('iteration_count_termination')
247         done = 1;
248     elseif stuck
249         disp('improvement_<_crit_termination')
250         done = 1;
251     end
252     rc=retcodeh;
253     if rc == 1
254         disp('zero_gradient')
255     elseif rc == 6

```



```

256         disp('smallest_step_still_improving_too_slow,_reversed_gradient')
257     elseif rc == 5
258         disp('largest_step_still_improving_too_fast')
259     elseif (rc == 4) | (rc==2)
260         disp('back_and_forth_on_step_length_never_finished')
261     elseif rc == 3
262         disp('smallest_step_still_improving_too_slow')
263     elseif rc == 7
264         disp('warning:_possible_inaccuracy_in_H_matrix')
265     end
266 % end
267 f=fh;
268 x=xh;
269 g=gh;
270 badg=badgh;
271 end
272 % what about making an m-file of 10 lines including numgrad.m
273 % since it appears three times in csminwel.m

```

cmintools/csolve.m

```

1  function [x,rc] = csolve(FUN,x,gradfun,crit,itmax,varargin)
2  %function [x,rc] = csolve(FUN,x,gradfun,crit,itmax,varargin)
3  % FUN should be written so that any parametric arguments are packed in to x,
4  % and so that if presented with a matrix x, it produces a return value of
5  % same dimension of x. The number of rows in x and FUN(x) are always the
6  % same. The number of columns is the number of different input arguments
7  % at which FUN is to be evaluated.
8  %
9  % gradfun: string naming the function called to evaluate the gradient matrix.
10 % If this
11 % is null (i.e. just ""), a numerical gradient is used instead.
12 % crit: if the sum of absolute values that FUN returns is less than this,
13 % the equation is solved.
14 % itmax: the solver stops when this number of iterations is reached, with
15 % rc=4
16 % varargin: in this position the user can place any number of additional
17 % arguments, all
18 % of which are passed on to FUN and gradfun (when it is non-empty)
19 % as a list of
20 % arguments following x.
21 % rc: 0 means normal solution, 1 and 3 mean no solution despite
22 % extremely fine adjustments
23 % in step length (very likely a numerical problem, or a
24 % discontinuity). 4 means itmax

```

```

19 %          termination .
20 %----- delta -----
21 % differencing interval for numerical gradient
22 delta = 1e-6;
23 %-----
24 %----- alpha -----
25 % tolerance on rate of descent
26 alpha=1e-3;
27 %-----
28 %----- verbose -----
29 verbose=1;% if this is set to zero , all screen output is suppressed
30 %-----
31 %----- analyticg -----
32 analyticg=1-isempty(gradfun); %if the grad argument is [], numerical
    derivatives are used .
33 %-----
34 nv=length(x);
35 tvec=delta*eye(nv);
36 done=0;
37 if isempty(varargin)
38     f0=feval(FUN,x);
39 else
40     f0=feval(FUN,x,varargin{:});
41 end
42 af0=sum(abs(f0));
43 af00=af0;
44 itct=0;
45 while ~done
46     if itct>3 & af00-af0<crit*max(1,af0) & rem(itct,2)==1
47         randomize=1;
48     else
49         if ~analyticg
50             if isempty(varargin)
51                 grad = (feval(FUN,x*ones(1,nv)+tvec)-f0*ones(1,nv))/delta;
52             else
53                 grad = (feval(FUN,x*ones(1,nv)+tvec,varargin{:})-f0*ones(1,nv))/
                    delta;
54             end
55         else % use analytic gradient
56             grad=feval(gradfun,x,varargin{:});
57         end
58         if isreal(grad)
59             if rcond(grad)<1e-12
60                 grad=grad+tvec;
61             end
62             dx0=-grad\f0;
63             randomize=0;

```

```

64         else
65             if(verbose), disp('gradient_imaginary'), end
66             randomize=1;
67         end
68     end
69     if randomize
70         if(verbose), fprintf(1, '\n_Random_Search'), end
71         dx0=norm(x)./randn(size(x));
72     end
73     lambda=1;
74     lambdamin=1;
75     fmin=f0;
76     xmin=x;
77     afmin=af0;
78     dxSize=norm(dx0);
79     factor=.6;
80     shrink=1;
81     subDone=0;
82     while ~subDone
83         dx=lambda*dx0;
84         f=feval(FUN,x+dx,varargin{:});
85         af=sum(abs(f));
86         if af<afmin
87             afmin=af;
88             fmin=f;
89             lambdamin=lambda;
90             xmin=x+dx;
91         end
92         if ((lambda > 0) & (af0-af < alpha*lambda*af0)) | ((lambda<0) & (af0-af <
           0) )
93             if ~shrink
94                 factor=factor^.6;
95                 shrink=1;
96             end
97             if abs(lambda*(1-factor))*dxSize > .1*delta;
98                 lambda = factor*lambda;
99             elseif (lambda > 0) & (factor==.6) %i.e., we've only been shrinking
100                 lambda=-.3;
101             else %
102                 subDone=1;
103                 if lambda > 0
104                     if factor==.6
105                         rc = 2;
106                     else
107                         rc = 1;
108                     end
109                 else

```

```

110         rc=3;
111     end
112 end
113 elseif (lambda >0) & (af-af0 > (1-alpha)*lambda*af0)
114     if shrink
115         factor=factor^.6;
116         shrink=0;
117     end
118     lambda=lambda/factor;
119 else % good value found
120     subDone=1;
121     rc=0;
122 end
123 end % while ~subDone
124 itct=itct+1;
125 if(verbose)
126     fprintf(1,'\nitct_%d,af_%g,lambda_%g,rc_%g',itct,afmin,lambdamin,rc)
127     fprintf(1,'\nxxxx_%10g_%10g_%10g_%10g',xmin);
128     fprintf(1,'\nfff_%10g_%10g_%10g_%10g',fmin);
129 end
130 x=xmin;
131 f0=fmin;
132 af00=af0;
133 af0=afmin;
134 if itct >= itmax
135     done=1;
136     rc=4;
137 elseif af0<crit;
138     done=1;
139     rc=0;
140 end
141 end

```

cmintools/initialize_mh.m

```

1 % initialize the MH by minimizing the - logposterior kernel
2 T = length(y);
3 vSwitch = ones(2, 1);
4 % % The parameters might be bounded, but the minimization routine will not
5 % % care, thus we need to make sure that the routine can go anywhere and
6 % % that we are still able to stay within the bounds.
7 % x0 = boundsINV(guess);
8 x0 = guess;
9 % posterior maximization
10 [fh, xh, gh, H, itct, fcount, retcodeh] = csminwel('logpostkernel',x0,.01*eye(

```

```

    length(x0)), [], 10e-5, 1000, centered, y, vSwitch, iPriorScale);
11 % processing the output of the maximization
12 postmode = xh;
13 JJ       = jacob(xh);
14 HH       = JJ * H * JJ';

```

cmintools/numgrad.m

```

1  function [g, badg] = numgrad(fcn,x,varargin)
2  % function [g badg] = numgrad(fcn,x,varargin)
3  %
4  %delta = 1e-6;
5  delta=1e-6;
6  n=length(x);
7  tvec=delta*eye(n);
8  g=zeros(n,1);
9  %-----old way to deal with variable # of P's-----
10 %tailstr = ')';
11 %stailstr = [];
12 %for i=nargin-2:-1:1
13 %   tailstr=[ ' P' num2str(i)   tailstr ];
14 %   stailstr=[ ' P' num2str(i) stailstr ];
15 %end
16 %f0 = eval([fcn '(x' tailstr)]; % Is there a way not to do this?
17 %-----^yes
18 f0 = feval(fcn,x,varargin{:});
19 % disp(' first fcn in numgrad.m -----')
20 %home
21 % disp(' numgrad.m is working. -----') % Jiinil on 9/5/95
22 % sizex=size(x), sizetvec=size(tvec), x,      % Jinill on 9/6/95
23 badg=0;
24 for i=1:n
25     scale=1; % originally 1
26     % i, tveci=tvec(:,i) % , plus=x+scale*tvec(:,i) % Jinill Kim on 9/6/95
27     if size(x,1)>size(x,2)
28         tvecv=tvec(i,:);
29     else
30         tvecv=tvec(:,i);
31     end
32     g0 = (feval(fcn,x+scale*tvecv', varargin{:}) - f0) ...
33         /(scale*delta);
34     % disp(' fcn in the i=1:n loop of numgrad.m -----') % Jinill
35     % disp('          and i is ') % Jinill
36     % i % Jinill

```



```

84 %                disp( ['Banging against wall, parameter ' int2str(i)] );
85 %                else
86 %                    g(i)=g1;
87 %                end
88 %            else
89 %                g(i)=0;
90 %                badg=1;
91 %                disp(['Valley around parameter ' int2str(i)])
92 %            end
93 %        end
94 %    end
95 %end
96 %save g.dat g x f0
97 %eval(['save g g x f0 ' stailstr]);

```