# **RISE Documentation**

Release 1.0.1

**Junior Maih** 

November 20, 2014

# **CONTENTS**

1	Intro	duction 1
	1.1	RISE at a Glance
	1.2	Capabilities of RISE
	1.3	How RISE works
	1.4	Background and mathematical formulations
	1.5	Using this documentation
	1.6	Citing RISE in your research
	1.7	License and Legal Mumbo-Jumbo
2	Getti	ing started with RISE 5
	2.1	Installation guide
	2.2	Troubleshooting
	2.3	RISE basics/basic principles
	2.4	Tutorial: A toy example
	2.5	How to find help?
	2.6	Where to go from here
3	RISE	E Capabilities 9
	3.1	Overview
	3.2	Markov switching DSGE modeling
	3.3	Markov switching SVAR modeling
	3.4	Markov switching VAR modeling
	3.5	Smooth transition VAR modeling
	3.6	Time-varying parameter modeling
	3.7	Maximum Likelihood and Bayesian Estimation
	3.8	Differentiation
	3.9	Time series
	3.10	Reporting
	3.11	Derivative-free optimization
	3.12	Global sensitivity analysis
4	The I	Markov switching DSGE interface 11
	4.1	The general framework
	4.2	The model file
	4.3	steady state
	4.4	getting information about the model
	4.5	deterministic simulation
	4.6	stochastic solution and simulation
	4.7	Estimation

	4.8 4.9	Forecasting and conditional forecasting	12 12
5	Marl	kov Switching Dynamic Stochastic General Equilibrium Modeling	15
	5.1	methods	15
	5.2	properties	16
	5.3	check_derivatives	17
	5.4	check_optimum	17
	5.5	compute_steady_state	18
	5.6	create_estimation_blocks	18
	5.7	create_state_list	19
	5.8	draw_parameter	19
	5.9	dsge	20
	5.10	estimate	20
	5.11	filter	22
	5.12	forecast	22
	5.13	forecast_real_time	24
	5.14	get	24
	5.15	historical_decomposition	25
	5.16	irf	26
			26
	5.17	is_stable_system	
	5.18	isnan	26
	5.19	load_parameters	27
	5.20	log_marginal_data_density	27
	5.21	log_posterior_kernel	28
	5.22	log_prior_density	28
	5.23	monte_carlo_filtering	29
	5.24	posterior_marginal_and_prior_densities	29
	5.25	posterior_simulator	30
	5.26	print_estimation_results	30
	5.27	print_solution	30
	5.28	prior_plots	31
	5.29	report	32
	5.30	resid	33
	5.31	set	33
	5.32	set_solution_to_companion	34
	5.33	simulate	34
	5.34	simulate_nonlinear	35
	5.35	simulation_diagnostics	36
	5.36	solve	36
	5.37	solve_alternatives	37
	5.38	stoch_simul	37
	5.39	theoretical_autocorrelations	37
	5.40	theoretical_autocovariances	38
	5.41	variance_decomposition	38
_			
6		nced-form VAR modeling	41
	6.1	methods	41
	6.2	properties	42
	6.3	check_identification	42
	6.4	check_optimum	43
	6.5	draw_parameter	43
	6.6	estimate	44
	6.7	forecast	46

	6.8	8	47
	6.9	historical_decomposition	47
	6.10	irf	48
	6.11	isnan	49
	6.12	load_parameters	50
	6.13	log_marginal_data_density	50
	6.14		51
	6.15		51
	6.16		52
	6.17		52
	6.18		53
	6.19		53
	6.20		54
	6.21	<u> </u>	54 54
	6.22	I and the second	
			55 
	6.23		55
	6.24		56
	6.25		56
	6.26	_ &	58
	6.27		58
	6.28	<del>-</del>	58
	6.29	structural_form	59
	6.30	template	60
	6.31	theoretical_autocorrelations	60
	6.32	theoretical_autocovariances	61
	6.33	variance_decomposition	61
7	Struc	ctural VAR modeling	63
	7.1	methods	63
	7.2	properties	64
	7.3	check_optimum	64
	7.4		65
	7.5	<u>-1</u>	65
	7.6		67
	7.7		68
	7.8		69
	7.9	_ 1	69
	7.10		71
		isnan	
	7.11		
	7 10	load_parameters	71
	7.12	load_parameters	72
	7.13	load_parameters   7     log_marginal_data_density   7     log_posterior_kernel   7	72 72
	7.13 7.14	load_parameters       7         log_marginal_data_density       7         log_posterior_kernel       7         log_prior_density       7	72 72 72
	7.13 7.14 7.15	load_parameters       7         log_marginal_data_density       7         log_posterior_kernel       7         log_prior_density       7         msvar_priors       7	72 72 72 73
	7.13 7.14	load_parameters       7         log_marginal_data_density       7         log_posterior_kernel       7         log_prior_density       7         msvar_priors       7         posterior_marginal_and_prior_densities       7	72 72 72 73 73
	7.13 7.14 7.15	load_parameters       7         log_marginal_data_density       7         log_posterior_kernel       7         log_prior_density       7         msvar_priors       7         posterior_marginal_and_prior_densities       7	72 72 72 73
	7.13 7.14 7.15 7.16	load_parameters       7         log_marginal_data_density       7         log_posterior_kernel       7         log_prior_density       7         msvar_priors       7         posterior_marginal_and_prior_densities       7         posterior_simulator       7	72 72 72 73 73
	7.13 7.14 7.15 7.16 7.17	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results	72 72 73 73 74
	7.13 7.14 7.15 7.16 7.17 7.18	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots	72 72 73 73 74
	7.13 7.14 7.15 7.16 7.17 7.18 7.19	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots report	72 72 73 73 74 74
	7.13 7.14 7.15 7.16 7.17 7.18 7.19 7.20	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots report set	72 72 73 73 74 74 74
	7.13 7.14 7.15 7.16 7.17 7.18 7.19 7.20 7.21	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots report set set_solution_to_companion	72 72 73 73 74 74 75
	7.13 7.14 7.15 7.16 7.17 7.18 7.19 7.20 7.21 7.22 7.23	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots report set set_solution_to_companion simulate	72 72 73 73 74 74 75 76 77
	7.13 7.14 7.15 7.16 7.17 7.18 7.19 7.20 7.21 7.22 7.23 7.24	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots report set set_solution_to_companion simulate simulation_diagnostics	72 72 73 73 74 74 75 76 76 77
	7.13 7.14 7.15 7.16 7.17 7.18 7.19 7.20 7.21 7.22 7.23	load_parameters log_marginal_data_density log_posterior_kernel log_prior_density msvar_priors posterior_marginal_and_prior_densities posterior_simulator print_estimation_results prior_plots report set set_solution_to_companion simulate simulation_diagnostics solve	72 72 73 73 74 74 75 76 77

	7.27 7.28 7.29 7.30 7.31	svar	79 79 80
8	Time	series 8	3
9		ov Chain Monte Carlo for Bayesian Estimation 8	
	9.1	Metropolis Hastings	
	9.2	Gibbs sampling	
	9.3	Marginal data density	5
10	Deriv	ative-free optimization 8	7
11		e Carlo Filtering	
	11.1	methods	
	11.2	properties	
	11.3	addlistener	
	11.4	cdf	
	11.5	<u></u> 1	
	11.6	correlation_patterns_plot	
	11.7	delete	
	11.8	eq	
		findobj	
		findprop	
		ge	
		gt	
		isvalid9	
		kolmogorov_smirnov_test	
		le	
		lt	
		ne	
		notify	
		scatter	
12		dimensional model representation 9	
	12.1	methods	
	12.2	properties	
	12.3	estimate	
	12.4	first_order_effect	
	12.5	hdmr	
	12.6	metamodel	
	12.7	plot_fit	
	12.8	polynomial_evaluation	
	12.9	polynomial_integration	
	12.10	polynomial_multiplication	9
13	Cont	ributing to RISE	1
	13.1	contributing new code	1
	13.2	contributing by helping maintain existing code	1
	13.3	other ways to contribute	1
	13.4	recommended development setup	
	13.5	RISE structure	1

	13.6 useful links, FAQ, checklist	101
14	Acknowledgements	103
15	Bibliography	105
16	Indices and tables	107

# INTRODUCTION

#### 1.1 RISE at a Glance

#### 1.1.1 What is RISE?

RISE is the acronym for Rationality In Switching Environments.

It is an object-oriented Matlab toolbox primarily designed for solving and estimating nonlinear dynamic stochastic general equilibirium (**DSGE**) or more generally Rational Expectations(**RE**) models with **switching parameters**.

Leading references in the field include various papers by Roger Farmer, Dan Waggoner and Tao Zha and Eric Leeper among others.

RISE uses perturbation to approximate the nonlinear Markov Switching Rational Expectations (MSRE) model and solves it using efficient algorithms.

RISE also implements special cases of the general Switching MSRE model. This includes

- VARs with and without switching parameters
- SVARs with and without switching paramters
- Time-varying parameter VARs
- etc.

#### 1.1.2 Motivation for RISE development

• The world is not constant, it is switching

# 1.2 Capabilities of RISE

#### 1.2.1 DSGE modeling

- constant parameters
- switching parameters
  - exogenous switching
  - endogenous switching
- optimal policy (with and without switching)

- discretion
- commitment
- loose commitment
- optimized simple rules
- Deterministic simulation
- Stochastic simulation
- higher-order perturbations

#### 1.2.2 VAR modeling

- constant parameters
  - zero restrictions
  - sign restrictions
  - restrictions on lag structure
  - linear restrictions
- switching parameters
  - linear restrictions

#### 1.2.3 SVAR modeling

- · constant parameters
- switching parameters
  - linear restrictions

## 1.2.4 Time-Varying parameter VAR modeling

Under implementation

#### 1.2.5 Smooth transition VAR modeling

Not yet implemented

## 1.2.6 Forecasting and Conditional Forecasting

#### 1.2.7 Global sensitivity analysis

- · Monte carlo filtering
- High dimensional model representation

#### 1.2.8 Maximum Likelihood and Bayesian Estimation

- · linear restrictions
- · nonlinear restrictions
- 1.2.9 Time series
- 1.2.10 Reporting
- 1.3 How RISE works
- 1.3.1 Object orientation
- 1.3.2 Basic principles
  - you can pass different options at any time

# 1.4 Background and mathematical formulations

- 1.5 Using this documentation
- 1.5.1 how to find help
- 1.5.2 Road map
- 1.6 Citing RISE in your research

# 1.7 License and Legal Mumbo-Jumbo

```
Copyright (c) 2009-2014, Junior Maih All rights reserved.
```

Redistribution and use in source and binary forms, with or without modification, are permitted provided that the following conditions are met:

 Redistributions of source code must retain the above copyright notice, this list of conditions and the following disclaimer.

1.3. How RISE works 3

- 2. Redistributions in binary form must reproduce the above copyright notice, this list of conditions and the following disclaimer in the documentation and/or other materials provided with the distribution.
- 3. Neither the name of the RISE Toolbox nor the names of its contributors may be used to endorse or promote products derived from this software without specific prior written permission.

THIS SOFTWARE IS PROVIDED BY THE COPYRIGHT HOLDERS AND CONTRIBUTORS "AS IS" AND ANY EXPRESS OR IMPLIED WARRANTIES, INCLUDING, BUT NOT LIMITED TO, THE IMPLIED WARRANTIES OF MERCHANTABILITY AND FITNESS FOR A PARTICULAR PURPOSE ARE DISCLAIMED. IN NO EVENT SHALL THE COPYRIGHT HOLDER OR CONTRIBUTORS BE LIABLE FOR ANY DIRECT, INDIRECT, INCIDENTAL, SPECIAL, EXEMPLARY, OR CONSEQUENTIAL DAMAGES (INCLUDING, BUT NOT LIMITED TO, PROCUREMENT OF SUBSTITUTE GOODS OR SERVICES; LOSS OF USE, DATA, OR PROFITS; OR BUSINESS INTERRUPTION) HOWEVER CAUSED AND ON ANY THEORY OF LIABILITY, WHETHER IN CONTRACT, STRICT LIABILITY, OR TORT (INCLUDING NEGLIGENCE OR OTHERWISE) ARISING IN ANY WAY OUT OF THE USE OF THIS SOFTWARE, EVEN IF ADVISED OF THE POSSIBILITY OF SUCH DAMAGE.

# **GETTING STARTED WITH RISE**

## 2.1 Installation guide

#### 2.1.1 Software requirements

I order to use RISE, the following software will need to be installed:

- Matlab version? or higher
- MikTex (Windows users) MacTex (mac users)

#### 2.1.2 How to obtain RISE

There are (at least) two ways to acquire RISE:

#### The zip file option

- 1. Go online to https://github.com/jmaih/RISE\_toolbox
- 2. download the zip file and unzip it in some directory on your computer.

This option is not recommended but is convenient for people who are not allowed to install new software on their machines/laptop.

#### Github for the bleeding-edge installation (highly recommended)

- 1. Go to http://windows.github.com if you are a windows user or to http://mac.github.com if you are a mac user
- 2. Create an account online through the website and download the Github program
- 3. Sign in both online and on the github on your machine. It is obvious online, but on your machine, just go to Github>Preference>Account
- 4. Go online to https://github.com/jmaih/RISE\_toolbox
- 5. Look for an icon with title 'Clone in Desktop' (or possibly clone in mac). There are options to locate where the repository will reside

The reason why this option is recommended is that you don't need to re-download the whole toolbox every time a marginal update is made. With one click and within seconds you can have the version of the toolbox on your computer updated.

#### The git option (never tested!!!)

The following has never been tested and so the syntax might be wrong:

```
git clone https://github.com/jmaih/RISE_toolbox.git
```

#### **Testing your installation**

More on this later...

#### 2.1.3 Loading and starting RISE

1. Locate the RISE\_toolbox directory and add its path to matlab in the command window as

```
addpath('C:/Users/JMaih/GithubRepositories/RISE_toolbox')
```

- 2. You will need to adapt this path to conform with the location of the toolbox on your machine.
- 3. run rise\_startup()

#### 2.1.4 Updating RISE

New features are constantly added, efficiency is improved, users sometimes report bugs that are corrected. All this makes it necessary to update RISE every now and then in order to keep abreast of the latest changes and developments.

However, updating RISE depends on precisely how you installed it in the first place:

- If you downloaded a zip file, you will have to redownload a zip file even if the recent change was just an added comma.
- if instead you invested in opening a github account, with one click you will be able to update just the changes you don't have.
- with git, you would just execute the command

```
git pull
```

# 2.2 Troubleshooting

# 2.3 RISE basics/basic principles

1. create an empty RISE object e.g.

```
tao=rise.empty(0);
```

- 2. run methods(rise) or methods(tao) to see the functions/methods that can be applied to a RISE object
- 3. run those methods on r". e.g. "irf(r)", simulate(r)", solve(r)", etc. this will give you the default options of each method and tell you how you can modify the behavior of the method

# 2.4 Tutorial: A toy example

#### 2.4.1 Foerster, Rubio-Ramirez, Waggoner and Zha (2014)

They consider the following model:

$$E_{t} \begin{bmatrix} 1 - \beta \frac{\left(1 - \frac{\kappa}{2} (\Pi_{t} - 1)^{2}\right) Y_{t}}{\left(1 - \frac{\kappa}{2} (\Pi_{t+1} - 1)^{2}\right) Y_{t+1}} \frac{1}{e^{\mu_{t+1}}} \frac{R_{t}}{\Pi_{t+1}} \\ (1 - \eta) + \eta \left(1 - \frac{\kappa}{2} (\Pi_{t} - 1)^{2}\right) Y_{t} + \beta \kappa \frac{\left(1 - \frac{\kappa}{2} (\Pi_{t} - 1)^{2}\right)}{\left(1 - \frac{\kappa}{2} (\Pi_{t+1} - 1)^{2}\right)} (\Pi_{t+1} - 1) \Pi_{t+1} - \kappa (\Pi_{t} - 1) \Pi_{t} \\ \left(\frac{R_{t-1}}{R_{ss}}\right)^{\rho} \Pi_{t}^{(1-\rho)\psi} \exp\left(\sigma \varepsilon_{t}\right) - \frac{R_{t}}{R_{ss}} \\ with \\ \mu_{t+1} = \bar{\mu} + \sigma \hat{\mu}_{t+1}. \end{bmatrix}$$

The first equation is an Euler equation, the second equation a Phillips curve and the third equation a nonlinear Taylor rule.

The switching parameters are  $\mu$  and  $\psi$ .

#### 2.4.2 The RISE code

The RISE code with parameterization is given by

```
endogenous PAI, Y, R
exogenous EPS_R
parameters a_tp_1_2, a_tp_2_1, betta, eta, kappa, mu, mu_bar, psi, rhor, sigr
parameters(a,2) mu, psi
model
        1-betta*(1-.5*kappa*(PAI-1)^2)*Y*R/((1-.5*kappa*(PAI(+1)-1)^2)*Y(+1)*exp(mu)*PAI(+1));
        1-eta+eta*(1-.5*kappa*(PAI-1)^2)*Y+betta*kappa*(1-.5*kappa*(PAI-1)^2)*(PAI(+1)-1)*PAI(+1)/(1-.5*kappa*(PAI-1)^2)
        -kappa*(PAI-1)*PAI;
        (R(-1)/steady_state(R))^rhor*(PAI/steady_state(PAI))^((1-rhor)*psi)*exp(sigr*EPS_R)-R/steady_state(PAI))
steady_state_model(unique,imposed)
    PAI=1;
    Y=(eta-1)/eta;
    R=exp(mu_bar)/betta*PAI;
parameterization
        a_tp_1_2,1-.9;
        a_tp_2_1,1-.9;
        betta, .99;
        kappa, 161;
        eta, 10;
        rhor, .8;
        sigr, 0.0025;
        mu_bar, 0.02;
        mu(a,1), 0.03;
```

```
mu(a,2), 0.01;
psi(a,1), 3.1;
psi(a,2), 0.9;
```

#### 2.4.3 Running the example

Assume this example is saved in a file named frwz\_nk.rs . The to run this example in Matlab, we run the following commands:

```
frwz=rise('frwz_nk'); % load the model and its parameterization
frwz=solve(frwz); % Solving the model
print_solution(frwz) % print the solution
```

# 2.5 How to find help?

# 2.6 Where to go from here

**CHAPTER** 

THREE

# RISE CAPABILITIES

O 4				
.7 _1	<i>(</i> )	VA		$\alpha$
3.1	V	VC	IVI	CAA

- 3.2 Markov switching DSGE modeling
- 3.3 Markov switching SVAR modeling
- 3.4 Markov switching VAR modeling
- 3.5 Smooth transition VAR modeling
- 3.6 Time-varying parameter modeling
- 3.7 Maximum Likelihood and Bayesian Estimation
- 3.8 Differentiation
- 3.8.1 numerical differentiation
- 3.8.2 Symbolic differentiation
- 3.8.3 Automatic/Algorithmic differentiation
- 3.9 Time series
- 3.10 Reporting
- 3.11 Derivative-free optimization
- 3.12 Global sensitivity analysis
- 3.12.1 Monte Carlo filtering

# THE MARKOV SWITCHING DSGE INTERFACE

# 4.1 The general framework

The general form of the models is:

$$E_{t} \sum_{r_{t+1}=1}^{h} \pi_{r_{t}, r_{t+1}} (I_{t}) \, \tilde{d}_{r_{t}} \left( b_{t+1} \left( r_{t+1} \right), b_{t} \left( r_{t} \right), b_{t-1}, \varepsilon_{t}, \theta_{r_{t+1}} \right) = 0$$

- The switching of the parameters is governed by Markov processes and can be endogenous.
- Agents can have information about future events

#### 4.2 The model file

#### 4.2.1 Conventions

#### 4.2.2 Variable declarations

#### 4.2.3 Expressions

- parameters and variables
  - inside the model
  - outside the model
- · operators
- functions
  - built-in functions
  - external/user-defined functions

#### 4.2.4 model declaration

· model equations

- · endogenous transition probabilities
- auxiliary parameters/variables
- · inequality restrictions

#### 4.2.5 auxiliary variables

#### 4.2.6 initial and terminal conditions

#### 4.2.7 shocks on exogenous variables

#### 4.2.8 other general declarations

# 4.3 steady state

- finding the steady state with the RISE nonlinear solver
- · using a steady state file
- using the steady state model

# 4.4 getting information about the model

#### 4.5 deterministic simulation

#### 4.6 stochastic solution and simulation

- · computing the stochastic solution
- · typology and ordering of variables
- first-order approximation
- second-order approximation
- third-order approximation
- fourth-order approximation
- fifth-order approximation

#### 4.7 Estimation

# 4.8 Forecasting and conditional forecasting

# 4.9 Optimal policy

• optimal simple rules

• Commitment, discretion and loose commitment

4.9. Optimal policy

# MARKOV SWITCHING DYNAMIC STOCHASTIC GENERAL EQUILIBRIUM MODELING

#### 5.1 methods

- [ check\_derivatives ](dsge/check\_derivatives)
- [ check\_optimum ](dsge/check\_optimum)
- [compute\_steady\_state](dsge/compute\_steady\_state)
- [create\_estimation\_blocks](dsge/create\_estimation\_blocks)
- [create\_state\_list](dsge/create\_state\_list)
- [ draw\_parameter ](dsge/draw\_parameter)
- [dsge](dsge/dsge)
- [ estimate ](dsge/estimate)
- [ filter ](dsge/filter)
- [ forecast ](dsge/forecast)
- [forecast\_real\_time](dsge/forecast\_real\_time)
- [ get ](dsge/get)
- [ historical\_decomposition ](dsge/historical\_decomposition)
- [ irf ](dsge/irf)
- [ is\_stable\_system ](dsge/is\_stable\_system)
- [isnan](dsge/isnan)
- [load\_parameters](dsge/load\_parameters)
- [log\_marginal\_data\_density](dsge/log\_marginal\_data\_density)
- [log posterior kernel](dsge/log posterior kernel)
- [ log\_prior\_density ](dsge/log\_prior\_density)
- [ monte\_carlo\_filtering ](dsge/monte\_carlo\_filtering)
- [ posterior\_marginal\_and\_prior\_densities ](dsge/posterior\_marginal\_and\_prior\_densities)

- [posterior\_simulator](dsge/posterior\_simulator)
- [ print\_estimation\_results ](dsge/print\_estimation\_results)
- [ print\_solution ](dsge/print\_solution)
- [ prior\_plots ](dsge/prior\_plots)
- [report](dsge/report)
- [resid](dsge/resid)
- [ set ](dsge/set)
- [set\_solution\_to\_companion](dsge/set\_solution\_to\_companion)
- [ simulate ](dsge/simulate)
- [ simulate\_nonlinear ](dsge/simulate\_nonlinear)
- [ simulation\_diagnostics ](dsge/simulation\_diagnostics)
- [ solve ](dsge/solve)
- [ solve\_alternatives ](dsge/solve\_alternatives)
- [stoch\_simul](dsge/stoch\_simul)
- [theoretical\_autocorrelations](dsge/theoretical\_autocorrelations)
- [theoretical\_autocovariances](dsge/theoretical\_autocovariances)
- [ variance\_decomposition ](dsge/variance\_decomposition)

# 5.2 properties

- [definitions] -
- [equations] -
- [folders\_paths] -
- [dsge\_var] -
- [filename] -
- [legend] -
- [endogenous] -
- [exogenous] -
- [parameters] -
- [observables] -
- [markov\_chains] -
- [options] -
- [estimation] -
- [solution] -
- [filtering] -

# 5.3 check\_derivatives

check\_derivatives - compares the derivatives and the solutions from various differentiation techniques

#### **5.3.1 Syntax**

```
check_derivatives(obj)
retcode=check_derivatives(obj)
```

#### **5.3.2 Inputs**

• obj [riseldsge]: model object or vectors of model objects

#### 5.3.3 Outputs

• **retcode** [numeric]: 0 if no problem is encountered during the comparisons. Else the meaning of recode can be found by running decipher(retcode)

#### 5.3.4 More About

- The derivatives computed are 'automatic', 'symbolic' or 'numerical'
- The comparisons are done relative to automatic derivatives, which are assumed to be the most accurate.

#### 5.3.5 Examples

See also:

# 5.4 check\_optimum

- **5.4.1 Syntax**
- **5.4.2 Inputs**
- 5.4.3 Outputs
- 5.4.4 More About
- 5.4.5 Examples

See also:

Help for dsge/check\_optimum is inherited from superclass RISE\_GENERIC

# 5.5 compute\_steady\_state

H1 line

- **5.5.1 Syntax**
- **5.5.2 Inputs**
- 5.5.3 Outputs
- 5.5.4 More About
- 5.5.5 Examples

See also:

# 5.6 create\_estimation\_blocks

H1 line

- **5.6.1 Syntax**
- **5.6.2 Inputs**
- 5.6.3 Outputs
- 5.6.4 More About
- 5.6.5 Examples

See also:

# 5.7 create\_state\_list

create\_state\_list creates the list of the state variables in the solution

#### **5.7.1 Syntax**

```
final_list=create_state_list(m)
final_list=create_state_list(m,orders)
```

#### **5.7.2 Inputs**

- m [dsgelrise]: model object
- orders [integer arrayl{1:m.options.solve\_order}] : approximation orders
- **compact\_form** [truel{false}]: if true, only unique combinations will be returned. Else, all combinations will be returned.

#### 5.7.3 Outputs

- final\_list [cellstr] : list of the state variables
- kept [vector] : location of kept state variables (computed only if compact\_form is set to true)

#### 5.7.4 More About

## 5.7.5 Examples

See also:

# 5.8 draw\_parameter

- **5.8.1 Syntax**
- **5.8.2 Inputs**
- 5.8.3 Outputs
- 5.8.4 More About

#### 5.8.5 Examples

See also:

Help for dsge/draw\_parameter is inherited from superclass RISE\_GENERIC

## 5.9 dsge

~~ no help found

#### 5.10 estimate

estimate - estimates the parameters of a RISE model

#### 5.10.1 Syntax

```
obj=estimate(obj)
obj=estimate(obj, varargin)
```

#### 5.10.2 Inputs

- obj [riseldsgelrfvarlsvar]: model object
- varargin additional optional inputs among which the most relevant for estimation are:
- estim\_parallel [integer|{1}]: Number of starting values
- estim\_start\_from\_mode [truelfalsel{[]}]: when empty, the user is prompted to answer the question as to whether to start estimation from a previously found mode or not. If true or false, no question is asked.
- estim\_start\_date [numericlcharlserial date]: date of the first observation to use in the dataset provided for estimation
- estim\_end\_date [numericlcharlserial date]: date of the last observation to use in the dataset provided for estimation
- estim\_max\_trials [integer|{500}]: When the initial value of the log-likelihood is too low, RISE uniformly draws from the prior support in search for a better starting point. It will try this for a maximum number of estim\_max\_trials times before squeaking with an error.

- estim\_start\_vals [{[]}|struct]: when not empty, the parameters whose names are fields of the structure will see their start values updated or overriden by the information in estim\_start\_vals. There is no need to provide values to update the start values for the estimated parameters.
- estim\_general\_restrictions [{[]}|function handle]: when not empty, the argument should be a function handle that takes as input a parameterized RISE object and returns a scalar or vector of numbers representing the strength of the violation of the nonlinear constraints. Those constraints will be added to the constraints already included in a rise/dsge file before being presented to the optimization function.
- estim\_linear\_restrictions [{[]}|cell]: This is most often used in the estimation of rfvar or svar models either to impose block exogeneity or to impose other forms of linear restrictions. When not empty, estim\_linear\_restrictions must be a 2-column cell: Each row of the first column represents a particular linear combination of the estimated parameters. Those linear combinations are constructed using the coef class. Check help for coef.coef for more details. Each row of the second column holds the value of the linear combination.
- estim\_blocks [{[]}|cell]: When not empty, this triggers blockwise optimization. For further information on how to set blocks, see help for dsge.create\_estimation\_blocks
- estim\_priors [{[]}|struct]: This provides an alternative to setting priors inside the rise/dsge model file. Each field of the structure must be the name of an estimated parameter. Each field will hold a cell array whose structure is described in help rise\_generic.setup\_priors.
- estim\_penalty [numericl{1e+8}]: value of the objective function when a problem occurs. Possible problems include: no solution found very low likelihood stochastic singularity problems computing the initial covariance matrix non-positive definite covariance matrices etc.
- estim\_cond\_vars [charlcellstr]: list of the variables to condition on during estimation of a dsge model. This then assumes that the data provided for estimation have several pages. The first page is the actual data, while the subsequent pages are the expectations data.
- optimset [struct]: identical to matlab's optimset
- optimizer [charlfunction handlelcell]: This can be the name of a standard matlab optimizer or RISE optimization routine or a user-defined optimization procedure available of the matlab search path. If the optimizer is provided as a cell, then the first element of the cell is the name of the optimizer or its handle and the remaining entries in the cell are additional input arguments to the user-defined optimization routine. A user-defined optimization function should have the following syntax

[xfinal, ffinal, exitflag] = optimizer (fh, x0, lb, ub, options, varargin);

#### That is, it accepts as inputs:

- **fh**: the function to optimize
- x0: a vector column of initial values of the parameters
- lb: a vector column of lower bounds
- ub: a vector column of upper bounds
- options: a structure of options whose fields will be similar to matlab's optimset
- varargin: additional arguments to the user-defined optimization procedure

#### That is, it provides as outputs:

- xfinal: the vector of final values
- ffinal: the value of fh at xfinal
- exitflag: a flag similar to the ones provided by matlab's

optimization functions.

5.10. estimate 21

- hessian\_type [{'fd'}l'opg']: The hessian is either computed by finite differences (fd) or by outer-product-gradient (opg)
- hessian\_repair [{false}ltrue]: If the Hessian is not positive definite, it nevertheless can be repaired and prepared for a potential mcmc simulation.

#### **5.10.3 Outputs**

• **obj** [riseldsgelrfvarlsvar]: model object parameterized with the mode found and holding additional estimation results and statistics that can be found under obj.estimation

#### 5.10.4 More About

recursive estimation may be done easily by passing a different estim\_end\_date at the beginning of each estimation run.

#### 5.10.5 Examples

See also:

Help for dsge/estimate is inherited from superclass RISE\_GENERIC

#### 5.11 filter

H1 line

5.11.1 Syntax

5.11.2 Inputs

**5.11.3 Outputs** 

5.11.4 More About

#### 5.11.5 Examples

See also:

#### 5.12 forecast

forecast - computes forecasts for riseldsgelsvarlrfvar models

#### 5.12.1 Syntax

cond\_fkst\_db=forecast (obj, varargin)

#### **5.12.2 Inputs**

- obj [riseldsgelsvarlrfvar]: model object
- varargin : additional inputs coming in pairs. These include but are not restricted to: forecast\_to\_time\_series [{true}|false]: sets the output to time

series format or not

- forecast\_nsteps [integerl{12}]: number of forecasting steps
- forecast\_start\_date [char|numeric|serial date]: date when the forecasts start (end of history + 1)
- forecast\_conditional\_hypothesis [{jma}|ncp|nas]: in dsge models in which agents have information beyond the current period, this option determines the number of periods of shocks need to match the restrictions: Hypothesis jma assumes that irrespective of how

many periods of conditioning information are remaining, agents always receive information on the same number of shocks.

- Hypothesis ncp assumes there are as many shocks periods as the number of the number of conditioning periods
- \* Hypothesis nas assumes there are as many shocks periods as the number of anticipated steps

#### **5.12.3 Outputs**

• **cond\_fkst\_db** [structlmatrix]: depending on the value of **forecast\_to\_time\_series** the returned output is a structure with time series or a cell containing a matrix and the information to reconstruct the time series.

#### 5.12.4 More About

- the historical information as well as the conditioning information come from the same database. The time series must be organized such that for each series, the first page represents the actual data and all subsequent pages represent conditional information. If a particular condition is "nan", that location is not constrained
- Conditional forecasting for nonlinear models is also supported. However, the solving of the implied nonlinear problem may fail if the model displays instability
- Both HARD CONDITIONS and SOFT CONDITIONS are implemented but the latter are currently disabled in expectation of a better user interface.
- The data may also contain time series for a variable with name **regime** in that case, the forecast/simulation paths are computed following the information therein. **regime** must be a member of 1:h, where h is the maximum number of regimes.

#### 5.12.5 Examples

See also: simulate

Help for dsge/forecast is inherited from superclass RISE\_GENERIC

5.12. forecast 23

## 5.13 forecast\_real\_time

forecast\_real\_time - forecast from each point in time

#### 5.13.1 Syntax

```
- [ts_fkst,ts_rmse,rmse,Updates]=forecast_real_time(obj)
- [ts_fkst,ts_rmse,rmse,Updates]=forecast_real_time(obj,varargin)
```

#### 5.13.2 Inputs

- obj [dsgelsvarlrfvar] : model object
- varargin: valid optional inputs coming in pairs. The main inputs of interest for changing the default behavior are: fkst\_rt\_nahead [integer]: number of periods ahead

#### **5.13.3 Outputs**

- ts\_fkst [struct]: fields are forecasts in the form of ts objects for the different endogenous variables
- ts\_rmse [struct] : fields are RMSEs in the form of ts objects for the different endogenous variables
- rmse [matrix] : RMSEs for the different endogenous variables
- **Updates** [struct]: fields are the updated (in a filtering sense) in the form of ts objects for the different endogenous variables

#### 5.13.4 More About

#### 5.13.5 Examples

See also: plot\_real\_time

# 5.14 get

- 5.14.1 Syntax
- 5.14.2 Inputs
- **5.14.3 Outputs**
- 5.14.4 More About
- 5.14.5 Examples

See also:

Help for dsge/get is inherited from superclass RISE\_GENERIC

# 5.15 historical\_decomposition

historical\_decomposition Computes historical decompositions of a DSGE model

#### 5.15.1 Syntax

```
[Histdec,obj]=history_dec(obj)
[Histdec,obj]=history_dec(obj,varargin)
```

#### 5.15.2 Inputs

- obj : [riseldsgelrfvarlsvar] model(s) for which to compute the decomposition. obj could be a vector of models
- varargin : standard optional inputs **coming in pairs**. Among which: **histdec\_start\_date** : [charlnumericl{''}] : date at which the

decomposition starts. If empty, the decomposition starts at he beginning of the history of the dataset

#### **5.15.3 Outputs**

Histdec: [structcell array] structure or cell array of structures with the decompositions in each model. The decompositions are given in terms of: - the exogenous variables - InitialConditions: the effect of initial conditions - risk: measure of the effect of non-certainty equivalence - switch: the effect of switching (which is also a shock!!!) - steady\_state: the contribution of the steady state

#### 5.15.4 Remarks

- the elements that do not contribute to any of the variables are automatically discarded.
- **N.B**: a switching model is inherently nonlinear and so, strictly speaking, the type of decomposition we do for linear/linearized constant-parameter models is not feasible. RISE takes an approximation in which the variables, shocks and states matrices across states are averaged. The averaging weights are the smoothed probabilities.

#### 5.15.5 Examples

See also:

Help for dsge/historical\_decomposition is inherited from superclass RISE\_GENERIC

#### 5.16 irf

H1 line

5.16.1 Syntax

5.16.2 Inputs

**5.16.3 Outputs** 

5.16.4 More About

5.16.5 Examples

See also:

# 5.17 is\_stable\_system

H1 line

5.17.1 Syntax

5.17.2 Inputs

**5.17.3 Outputs** 

5.17.4 More About

5.17.5 Examples

See also:

#### **5.18** isnan

- 5.18.1 Syntax
- 5.18.2 Inputs
- **5.18.3 Outputs**
- 5.18.4 More About
- 5.18.5 Examples

See also:

Help for dsge/isnan is inherited from superclass RISE\_GENERIC

# 5.19 load\_parameters

H1 line

- 5.19.1 Syntax
- 5.19.2 Inputs
- **5.19.3 Outputs**
- 5.19.4 More About
- 5.19.5 Examples

See also:

Help for dsge/load\_parameters is inherited from superclass RISE\_GENERIC

# 5.20 log\_marginal\_data\_density

- 5.20.1 Syntax
- 5.20.2 Inputs
- **5.20.3 Outputs**
- 5.20.4 More About
- 5.20.5 Examples

See also:

Help for dsge/log\_marginal\_data\_density is inherited from superclass RISE\_GENERIC

# 5.21 log posterior kernel

H1 line

- 5.21.1 Syntax
- 5.21.2 Inputs
- **5.21.3 Outputs**
- 5.21.4 More About
- 5.21.5 Examples

See also:

Help for dsge/log\_posterior\_kernel is inherited from superclass RISE\_GENERIC

# 5.22 log\_prior\_density

- 5.22.1 Syntax
- **5.22.2 Inputs**
- **5.22.3 Outputs**
- 5.22.4 More About
- 5.22.5 Examples

Help for dsge/log\_prior\_density is inherited from superclass RISE\_GENERIC

## 5.23 monte\_carlo\_filtering

H1 line

- 5.23.1 Syntax
- 5.23.2 Inputs
- **5.23.3 Outputs**
- 5.23.4 More About
- 5.23.5 Examples

See also:

## 5.24 posterior marginal and prior densities

H1 line

- 5.24.1 Syntax
- 5.24.2 Inputs
- **5.24.3 Outputs**
- 5.24.4 More About
- 5.24.5 Examples

See also:

Help for dsge/posterior\_marginal\_and\_prior\_densities is inherited from superclass RISE\_GENERIC

## 5.25 posterior\_simulator

H1 line

5.25.1 Syntax

5.25.2 Inputs

**5.25.3 Outputs** 

5.25.4 More About

5.25.5 Examples

See also:

Help for dsge/posterior\_simulator is inherited from superclass RISE\_GENERIC

## 5.26 print\_estimation\_results

H1 line

5.26.1 Syntax

5.26.2 Inputs

**5.26.3 Outputs** 

5.26.4 More About

5.26.5 Examples

See also:

Help for dsge/print\_estimation\_results is inherited from superclass RISE\_GENERIC

## 5.27 print\_solution

print\_solution - print the solution of a model or vector of models

## 5.27.1 Syntax

```
print_solution(obj)
print_solution(obj,varlist)
print_solution(obj,varlist,orders)
print_solution(obj,varlist,orders,compact_form)
print_solution(obj,varlist,orders,compact_form,precision)
print_solution(obj,varlist,orders,compact_form,precision,equation_format)
print_solution(obj,varlist,orders,compact_form,precision,equation_format,file2save2)
outcell=print_solution(obj,...)
```

### 5.27.2 Inputs

- **obj** [riseldsge] : model object or vector of model objects
- varlist [charlcellstrl{[]}]: list of variables of interest
- orders [numericl{[1:solve\_order]}]: orders for which we want to see the solution
- **compact\_form** [{true}|false]: if true, only the solution of unique tuples (i,j,k) such that  $i \le j \le k$  is presented. If false, the solution of all combinations is presented. i.e. (i,j,k)(i,k,j)(j,i,k)(j,k,i)(k,i,j)(k,j,i)
- **precision** [charl{'%8.6f'}]: precision of the numbers printed
- equation\_format [truel{false}]: if true, the solution is presented in the form of equations for each endogenous variable (not recommended)
- file2save2 [charl{''}]: if not empty, the solution is written to a file rather than printed on screen. For this to happen, print\_solution has to be called without ouput arguments

## **5.27.3 Outputs**

• outcell [cellstr]: If an output is requested, the solution is not printed on screen or to a file.

#### 5.27.4 More About

If a model is solved, say, up to 3rd order, one may still want to see the first-order solution or the solution up to second-order only or any combination of orders.

#### 5.27.5 Examples

See also:

# 5.28 prior\_plots

H1 line

5.28. prior\_plots 31

5.28.1 Syntax

5.28.2 Inputs

**5.28.3 Outputs** 

5.28.4 More About

### 5.28.5 Examples

See also:

Help for dsge/prior\_plots is inherited from superclass RISE\_GENERIC

## 5.29 report

**REPORT** assigns the elements of interest to a rise\_report.report object

### 5.29.1 Syntax

::

- REPORT(rise.empty(0)) : displays the default inputs
- REPORT(obj,destination\_root,rep\_items): assigns the reported elements in rep\_items to destination\_root
- REPORT(obj,destination\_root,rep\_items,varargin): assigns varargin to obj before doing the rest

#### 5.29.2 Inputs

- obj : [riseldsge]
- destination root: [rise report.report]: handle for the actual report
- rep\_items: [charlcellstr]: list of desired items to report. This list can only include: 'endogenous', 'exogenous', 'observables', 'parameters', 'solution', 'estimation', 'estimation\_statistics', 'equations', 'code'

## **5.29.3 Outputs**

none

#### 5.29.4 More About

### 5.29.5 Examples

See also:

Help for dsge/report is inherited from superclass RISE\_GENERIC

### 5.30 resid

H1 line

5.30.1 Syntax

5.30.2 Inputs

**5.30.3 Outputs** 

5.30.4 More About

5.30.5 Examples

See also:

### 5.31 set

set - sets options for dsgelrise models

## 5.31.1 Syntax

obj=set(obj,varargin)

### 5.31.2 Inputs

- **obj** [riseldsge]: model object
- varargin: valid input arguments coming in pairs. Notable fields to that can be set include and are not restricted to: solve\_shock\_horizon [integer|struct|cell]
  - for the integer case, all shocks are set to the same integer
  - for the struct case, the input must be a structure with shock names as fields. Only the shock names whose value is to change have to be listed. In this case, different shocks can have different horizons k. The default is k=0 i.e. agents don't see into the future
  - for the cell case, the cell should have two columns. The first column includes the names of the shocks
    whose horizon is to change. The second column includes the horizon for each shock name on the left.
  - solve\_function\_mode [{explicit/amateur}|vectorized/professional|disc]
    - \* in the amateur or explicit mode the functions are kept in cell arrays of anonymous functions and evaluated using for loops
    - \* in the vectorized or professional mode the functions are compacted into one long and unreadable function.
    - \* in the disc mode the functions are written to disc in a subdirectory called routines.

5.30. resid 33

#### **5.31.3 Outputs**

• obj [riseldsge]: model object

#### 5.31.4 More About

### 5.31.5 Examples

obj=set(obj,'solve\_shock\_horizon',struct('shock1',2,'shock3',4)) obj=set(obj,'solve\_shock\_horizon',5) See also: rise\_generic.set

## 5.32 set\_solution\_to\_companion

H1 line

5.32.1 Syntax

5.32.2 Inputs

**5.32.3 Outputs** 

5.32.4 More About

5.32.5 Examples

See also:

### 5.33 simulate

simulate - simulates a RISE model

#### 5.33.1 Syntax

[db, states, retcode] = simulate(obj, varargin)

### 5.33.2 Inputs

- obj [rfvarldsgelriselsvar]: model object
- varargin: additional arguments including but not restricted to
  - simul\_periods [integer|{100}]: number of simulation periods
  - simul\_burn [integer|{100}]: number of burn-in periods

- simul\_algo [[{mt19937ar}| mcg16807|mlfg6331\_64|mrg32k3a| shr3conglswb2712]]: matlab's seeding algorithms
- **simul\_seed** [numericl{0}]: seed of the computations
- simul\_historical\_data [tslstructl{''}]: historical data from which the simulations are based. If empty, the simulations start at the steady state.
- simul\_history\_end\_date [charlinteger|serial date]: last date of history
- simul regime [integer|vector|{[]}]: regimes for which the model is simulated
- simul\_update\_shocks\_handle [function handle]: we may want to update the shocks if some condition on the state of the economy is satisfied. For instance, shock monetary policy to keep the interest rate at the floor for an extented period of time if we already are at the ZLB/ZIF. simul\_update\_shocks\_handle takes as inputs the current shocks and the state vector (all the endogenous variables) and returns the updated shocks. But for all this to be put into motion, the user also has to turn on simul\_do\_update\_shocks by setting it to true.
- simul\_do\_update\_shocks [true|{false}]: update the shocks based on simul\_update\_shocks\_handle or not.
- simul\_to\_time\_series [{true}|false]: if true, the output is a time series, else a cell array with a matrix and information on elements that help reconstruct the time series.

### **5.33.3 Outputs**

- **db** [struct|cell array]: if **simul\_to\_time\_series** is true, the output is a time series, else a cell array with a matrix and information on elements that help reconstruct the time series.
- states [vector]: history of the regimes over the forecast horizon
- **retcode** [integer]: if 0, the simulation went fine. Else something got wrong. In that case one can understand the problem by running decipher(retcode)

#### 5.33.4 More About

• **simul\_historical\_data** contains the historical data as well as conditional information over the forecast horizon. It may also include as an alternative to **simul\_regime**, a time series with name **regime**, which indicates the regimes over the forecast horizon.

#### 5.33.5 Examples

See also:

Help for dsge/simulate is inherited from superclass RISE\_GENERIC

## 5.34 simulate\_nonlinear

- 5.34.1 Syntax
- 5.34.2 Inputs
- **5.34.3 Outputs**
- 5.34.4 More About
- 5.34.5 Examples

## 5.35 simulation\_diagnostics

H1 line

- 5.35.1 Syntax
- 5.35.2 Inputs
- **5.35.3 Outputs**
- 5.35.4 More About
- **5.35.5 Examples**

See also:

Help for dsge/simulation\_diagnostics is inherited from superclass RISE\_GENERIC

## **5.36** solve

H1 line

- 5.36.1 Syntax
- 5.36.2 Inputs
- **5.36.3 Outputs**
- 5.36.4 More About
- 5.36.5 Examples

obj=solve(obj,'solve\_shock\_horizon',struct('shock1',2,'shock3',4)) obj=solve(obj,'solve\_shock\_horizon',5)

## 5.37 solve\_alternatives

H1 line

- 5.37.1 Syntax
- 5.37.2 Inputs
- **5.37.3 Outputs**
- 5.37.4 More About
- 5.37.5 Examples

See also:

## 5.38 stoch\_simul

H1 line

- 5.38.1 Syntax
- 5.38.2 Inputs
- **5.38.3 Outputs**
- 5.38.4 More About
- 5.38.5 Examples

See also:

Help for dsge/stoch\_simul is inherited from superclass RISE\_GENERIC

## 5.39 theoretical\_autocorrelations

- 5.39.1 Syntax
- 5.39.2 Inputs
- **5.39.3 Outputs**
- 5.39.4 More About
- 5.39.5 Examples

Help for dsge/theoretical\_autocorrelations is inherited from superclass RISE\_GENERIC

## 5.40 theoretical\_autocovariances

H1 line

- 5.40.1 Syntax
- 5.40.2 Inputs
- **5.40.3 Outputs**
- 5.40.4 More About
- 5.40.5 Examples

See also:

Help for dsge/theoretical\_autocovariances is inherited from superclass RISE\_GENERIC

## 5.41 variance\_decomposition

- 5.41.1 Syntax
- 5.41.2 Inputs
- **5.41.3 Outputs**
- 5.41.4 More About
- 5.41.5 Examples

Help for dsge/variance\_decomposition is inherited from superclass RISE\_GENERIC

# REDUCED-FORM VAR MODELING

### 6.1 methods

- [ check\_identification ](rfvar/check\_identification)
- [ check\_optimum ](rfvar/check\_optimum)
- [ draw\_parameter ](rfvar/draw\_parameter)
- [ estimate ](rfvar/estimate)
- [ forecast ](rfvar/forecast)
- [ get ](rfvar/get)
- [historical decomposition] (rfvar/historical decomposition)
- [ irf ](rfvar/irf)
- [isnan](rfvar/isnan)
- [ load\_parameters ](rfvar/load\_parameters)
- [log\_marginal\_data\_density](rfvar/log\_marginal\_data\_density)
- [log\_posterior\_kernel](rfvar/log\_posterior\_kernel)
- [log\_prior\_density](rfvar/log\_prior\_density)
- [ msvar\_priors ](rfvar/msvar\_priors)
- [ posterior\_marginal\_and\_prior\_densities ](rfvar/posterior\_marginal\_and\_prior\_densities)
- [ posterior\_simulator ](rfvar/posterior\_simulator)
- [print\_estimation\_results](rfvar/print\_estimation\_results)
- [ prior\_plots ](rfvar/prior\_plots)
- [report](rfvar/report)
- [ rfvar ](rfvar/rfvar)
- [ set ](rfvar/set)
- [ set solution to companion ](rfvar/set solution to companion)
- [ simulate ](rfvar/simulate)
- [ simulation\_diagnostics ](rfvar/simulation\_diagnostics)
- [solve](rfvar/solve)

- [ stoch\_simul ](rfvar/stoch\_simul)
- [ structural\_form ](rfvar/structural\_form)
- [ template ](rfvar/template)
- [theoretical\_autocorrelations](rfvar/theoretical\_autocorrelations)
- [theoretical\_autocovariances](rfvar/theoretical\_autocovariances)
- [variance\_decomposition](rfvar/variance\_decomposition)

## 6.2 properties

- [identification] -
- [structural\_shocks] -
- [nonlinear\_restrictions] -
- [constant] -
- [nlags] -
- [legend] -
- [endogenous] -
- [exogenous] -
- [parameters] -
- [observables] -
- [markov\_chains] -
- [options] -
- [estimation] -
- [solution] -
- [filtering] -

## 6.3 check\_identification

- **6.3.1 Syntax**
- **6.3.2 Inputs**
- 6.3.3 Outputs
- 6.3.4 More About
- 6.3.5 Examples

## 6.4 check\_optimum

H1 line

- 6.4.1 Syntax
- **6.4.2 Inputs**
- 6.4.3 Outputs
- 6.4.4 More About
- 6.4.5 Examples

See also:

Help for rfvar/check\_optimum is inherited from superclass RISE\_GENERIC

## 6.5 draw\_parameter

H1 line

- **6.5.1 Syntax**
- **6.5.2 Inputs**
- 6.5.3 Outputs
- 6.5.4 More About
- 6.5.5 Examples

See also:

Help for rfvar/draw\_parameter is inherited from superclass RISE\_GENERIC

#### 6.6 estimate

estimate - estimates the parameters of a RISE model

## **6.6.1 Syntax**

```
obj=estimate(obj)
obj=estimate(obj, varargin)
```

### **6.6.2 Inputs**

- **obj** [riseldsgelrfvarlsvar]: model object
- varargin additional optional inputs among which the most relevant for estimation are:
- estim\_parallel [integer|{1}]: Number of starting values
- **estim\_start\_from\_mode** [truelfalsel{[]}]: when empty, the user is prompted to answer the question as to whether to start estimation from a previously found mode or not. If true or false, no question is asked.
- estim\_start\_date [numericlcharlserial date]: date of the first observation to use in the dataset provided for estimation
- estim\_end\_date [numericlcharlserial date]: date of the last observation to use in the dataset provided for estimation
- estim\_max\_trials [integer|{500}]: When the initial value of the log-likelihood is too low, RISE uniformly draws from the prior support in search for a better starting point. It will try this for a maximum number of estim\_max\_trials times before squeaking with an error.
- estim\_start\_vals [{[]}|struct]: when not empty, the parameters whose names are fields of the structure will see their start values updated or overriden by the information in estim\_start\_vals. There is no need to provide values to update the start values for the estimated parameters.
- estim\_general\_restrictions [{[]}|function handle]: when not empty, the argument should be a function handle that takes as input a parameterized RISE object and returns a scalar or vector of numbers representing the strength of the violation of the nonlinear constraints. Those constraints will be added to the constraints already included in a rise/dsge file before being presented to the optimization function.
- estim\_linear\_restrictions [{[]}|cell]: This is most often used in the estimation of rfvar or svar models either to impose block exogeneity or to impose other forms of linear restrictions. When not empty, estim\_linear\_restrictions must be a 2-column cell: Each row of the first column represents a particular linear combination of the estimated parameters. Those linear combinations are constructed using the coef class. Check help for coef.coef for more details. Each row of the second column holds the value of the linear combination.
- estim\_blocks [{[]}|cell]: When not empty, this triggers blockwise optimization. For further information on how to set blocks, see help for dsge.create\_estimation\_blocks
- estim\_priors [{[]}|struct]: This provides an alternative to setting priors inside the rise/dsge model file. Each field of the structure must be the name of an estimated parameter. Each field will hold a cell array whose structure is described in help rise\_generic.setup\_priors.

- estim\_penalty [numericl{1e+8}]: value of the objective function when a problem occurs. Possible problems include: no solution found very low likelihood stochastic singularity problems computing the initial covariance matrix non-positive definite covariance matrices etc.
- estim\_cond\_vars [charlcellstr]: list of the variables to condition on during estimation of a dsge model. This then assumes that the data provided for estimation have several pages. The first page is the actual data, while the subsequent pages are the expectations data.
- optimset [struct]: identical to matlab's optimset
- optimizer [charlfunction handlelcell]: This can be the name of a standard matlab optimizer or RISE optimization routine or a user-defined optimization procedure available of the matlab search path. If the optimizer is provided as a cell, then the first element of the cell is the name of the optimizer or its handle and the remaining entries in the cell are additional input arguments to the user-defined optimization routine. A user-defined optimization function should have the following syntax

```
[xfinal, ffinal, exitflag] = optimizer (fh, x0, lb, ub, options, varargin);
```

#### That is, it accepts as inputs:

- **fh**: the function to optimize
- x0: a vector column of initial values of the parameters
- **lb**: a vector column of lower bounds
- ub: a vector column of upper bounds
- options: a structure of options whose fields will be similar to matlab's optimset
- varargin: additional arguments to the user-defined optimization procedure

#### That is, it provides as outputs:

- xfinal: the vector of final values
- ffinal: the value of fh at xfinal
- exitflag: a flag similar to the ones provided by matlab's

optimization functions.

- hessian\_type [{'fd'}l'opg']: The hessian is either computed by finite differences (fd) or by outer-product-gradient (opg)
- hessian\_repair [{false}|true]: If the Hessian is not positive definite, it nevertheless can be repaired and prepared for a potential mcmc simulation.

#### 6.6.3 Outputs

• **obj** [riseldsgelrfvarlsvar]: model object parameterized with the mode found and holding additional estimation results and statistics that can be found under obj.estimation

#### 6.6.4 More About

recursive estimation may be done easily by passing a different estim\_end\_date at the beginning of each estimation run.

6.6. estimate 45

#### 6.6.5 Examples

See also:

Help for rfvar/estimate is inherited from superclass RISE\_GENERIC

### 6.7 forecast

forecast - computes forecasts for riseldsgelsvarlrfvar models

#### **6.7.1** Syntax

cond\_fkst\_db=forecast(obj,varargin)

### **6.7.2 Inputs**

- **obj** [riseldsgelsvarlrfvar]: model object
- varargin: additional inputs coming in pairs. These include but are not restricted to: forecast\_to\_time\_series [{true}|false]: sets the output to time

series format or not

- **forecast\_nsteps** [integer|{12}]: number of forecasting steps
- forecast start date [char|numeric|serial date]: date when the forecasts start (end of history + 1)
- forecast\_conditional\_hypothesis [{jma}|ncplnas]: in dsge models in which agents have information beyond the current period, this option determines the number of periods of shocks need to match the restrictions: Hypothesis jma assumes that irrespective of how

many periods of conditioning information are remaining, agents always receive information on the same number of shocks.

- \* Hypothesis ncp assumes there are as many shocks periods as the number of the number of conditioning periods
- \* Hypothesis nas assumes there are as many shocks periods as the number of anticipated steps

## 6.7.3 Outputs

• **cond\_fkst\_db** [structlmatrix]: depending on the value of **forecast\_to\_time\_series** the returned output is a structure with time series or a cell containing a matrix and the information to reconstruct the time series.

#### 6.7.4 More About

- the historical information as well as the conditioning information come from the same database. The time series must be organized such that for each series, the first page represents the actual data and all subsequent pages represent conditional information. If a particular condition is "nan", that location is not constrained
- Conditional forecasting for nonlinear models is also supported. However, the solving of the implied nonlinear problem may fail if the model displays instability

- Both HARD CONDITIONS and SOFT CONDITIONS are implemented but the latter are currently disabled in expectation of a better user interface.
- The data may also contain time series for a variable with name **regime** in that case, the forecast/simulation paths are computed following the information therein. **regime** must be a member of 1:h, where h is the maximum number of regimes.

## 6.7.5 Examples

See also: simulate

Help for rfvar/forecast is inherited from superclass RISE\_GENERIC

## 6.8 get

H1 line

6.8.1 Syntax

**6.8.2 Inputs** 

6.8.3 Outputs

6.8.4 More About

### 6.8.5 Examples

See also:

Help for rfvar/get is inherited from superclass RISE\_GENERIC

## 6.9 historical\_decomposition

historical\_decomposition Computes historical decompositions of a DSGE model

## 6.9.1 Syntax

```
[Histdec,obj]=history_dec(obj)
[Histdec,obj]=history_dec(obj,varargin)
```

6.8. get 47

### **6.9.2 Inputs**

- obj : [riseldsgelrfvarlsvar] model(s) for which to compute the decomposition. obj could be a vector of models
- varargin : standard optional inputs **coming in pairs**. Among which: **histdec\_start\_date** : [charlnumericl{''}] : date at which the

decomposition starts. If empty, the decomposition starts at he beginning of the history of the dataset

### 6.9.3 Outputs

• Histdec: [structlcell array] structure or cell array of structures with the decompositions in each model. The decompositions are given in terms of: - the exogenous variables - **InitialConditions**: the effect of initial conditions - **risk**: measure of the effect of non-certainty equivalence - **switch**: the effect of switching (which is also a shock!!!) - **steady\_state**: the contribution of the steady state

#### 6.9.4 Remarks

- the elements that do not contribute to any of the variables are automatically discarded.
- **N.B**: a switching model is inherently nonlinear and so, strictly speaking, the type of decomposition we do for linear/linearized constant-parameter models is not feasible. RISE takes an approximation in which the variables, shocks and states matrices across states are averaged. The averaging weights are the smoothed probabilities.

### 6.9.5 Examples

See also:

Help for rfvar/historical\_decomposition is inherited from superclass RISE\_GENERIC

#### 6.10 irf

irf - computes impulse responses for a RISE model

## 6.10.1 Syntax

```
myirfs=irf(obj)
myirfs=irf(obj,varargin)
```

## 6.10.2 Inputs

- **obj** [riseldsgelrfvarlsvar]: single or vector of RISE models
- varargin: optional options coming in pairs. The notable ones that will influence the behavior of the impulse responses are:
- irf\_shock\_list [charlcellstrl{''}]: list of shocks for which we want to compute impulse responses
- **irf\_var\_list** [char|cellstr|{''}]: list of the endogenous variables we want to report

- **irf\_periods** [integer|{40}]: length of the irfs
- **irf\_shock\_sign** [numericl-1|{1}]: sign or scale of the original impulse. If **irf\_shock\_sign** >0, we get impulse responses to a positive shock. If **irf\_shock\_sign** <0, the responses are negative. If **irf\_shock\_sign** =0, all the responses are 0.
- **irf\_draws** [integerl{50}]: number of draws used in the simulation impulse responses in a nonlinear model. A nonlinear model is defined as a model that satisfies at least one of the following criteria solved at an order >1 has more than one regime and option **irf\_regime\_specific** below is

set to false

- irf\_type [{irf}|girf]: type of irfs. If the type is irf, the impulse responses are computed directly exploiting the fact that the model is linear. If the type is girf, the formula for the generalized impulse responses is used: the irf is defined as the expectation of the difference of two simulation paths. In the first path the initial impulse for the shock of interest is nonzero while it is zero for the second path. All other shocks are the same for both paths in a given simulation.
- irf\_regime\_specific [{true}|false]: In a switching model, we may or may not want to compute impulse responses specific to each regime.
- irf\_use\_historical\_data [{false}|true]: if true, the data stored in option simul\_historical\_data are used as initial conditions. But the model has to be nonlinear otherwise the initial conditions are set to zero. This option gives the flexibility to set the initial conditions for the impulse responses.
- **irf\_to\_time\_series** [{true}|false]: If true, the output is in the form of time series. Else it is in the form of a cell containing the information needed to reconstruct the time series.

### 6.10.3 Outputs

• myirfs [{struct}|cell]: Impulse response data

#### 6.10.4 More About

- for linear models or models solved up to first order, the initial conditions as well as the steady states are set to 0 in the computation of the impulse responses.
- for nonlinear models, the initial conditions is the ergodic mean

### 6.10.5 Examples

See also:

Help for rfvar/irf is inherited from superclass RISE\_GENERIC

#### 6.11 isnan

H1 line

6.11. isnan 49

- 6.11.1 Syntax
- **6.11.2 Inputs**
- **6.11.3 Outputs**
- 6.11.4 More About
- 6.11.5 Examples

Help for rfvar/isnan is inherited from superclass RISE\_GENERIC

# 6.12 load\_parameters

H1 line

- 6.12.1 Syntax
- **6.12.2 Inputs**
- **6.12.3 Outputs**
- 6.12.4 More About
- 6.12.5 Examples

See also:

Help for rfvar/load\_parameters is inherited from superclass RISE\_GENERIC

## 6.13 log\_marginal\_data\_density

- 6.13.1 Syntax
- **6.13.2 Inputs**
- **6.13.3 Outputs**
- 6.13.4 More About
- 6.13.5 Examples

Help for rfvar/log\_marginal\_data\_density is inherited from superclass RISE\_GENERIC

## 6.14 log\_posterior\_kernel

H1 line

- 6.14.1 Syntax
- 6.14.2 Inputs
- **6.14.3 Outputs**
- 6.14.4 More About
- 6.14.5 Examples

See also:

Help for rfvar/log\_posterior\_kernel is inherited from superclass RISE\_GENERIC

# 6.15 log\_prior\_density

6.15.1 Syntax

**6.15.2 Inputs** 

**6.15.3 Outputs** 

6.15.4 More About

6.15.5 Examples

See also:

Help for rfvar/log\_prior\_density is inherited from superclass RISE\_GENERIC

## 6.16 msvar\_priors

H1 line

6.16.1 Syntax

6.16.2 Inputs

**6.16.3 Outputs** 

6.16.4 More About

6.16.5 Examples

See also:

Help for rfvar/msvar\_priors is inherited from superclass SVAR

# 6.17 posterior\_marginal\_and\_prior\_densities

- 6.17.1 Syntax
- **6.17.2 Inputs**
- **6.17.3 Outputs**
- 6.17.4 More About
- 6.17.5 Examples

Help for rfvar/posterior\_marginal\_and\_prior\_densities is inherited from superclass RISE\_GENERIC

## 6.18 posterior\_simulator

H1 line

- 6.18.1 Syntax
- 6.18.2 Inputs
- **6.18.3 Outputs**
- 6.18.4 More About
- 6.18.5 Examples

See also:

Help for rfvar/posterior\_simulator is inherited from superclass RISE\_GENERIC

# 6.19 print\_estimation\_results

6.19.1 Syntax

6.19.2 Inputs

**6.19.3 Outputs** 

6.19.4 More About

### 6.19.5 Examples

See also:

Help for rfvar/print\_estimation\_results is inherited from superclass RISE\_GENERIC

## 6.20 prior\_plots

H1 line

6.20.1 Syntax

6.20.2 Inputs

**6.20.3 Outputs** 

6.20.4 More About

## 6.20.5 Examples

See also:

Help for rfvar/prior\_plots is inherited from superclass RISE\_GENERIC

## 6.21 report

**REPORT** assigns the elements of interest to a rise\_report.report object

### 6.21.1 Syntax

::

- REPORT(rise.empty(0)) : displays the default inputs
- REPORT(obj,destination\_root,rep\_items): assigns the reported elements in rep\_items to destination\_root
- REPORT(obj,destination\_root,rep\_items,varargin) : assigns varargin to obj before doing the rest

## **6.21.2 Inputs**

- obj : [riseldsge]
- destination\_root : [rise\_report.report] : handle for the actual report
- rep\_items: [charlcellstr]: list of desired items to report. This list can only include: 'endogenous', 'exogenous', 'observables', 'parameters', 'solution', 'estimation', 'estimation\_statistics', 'equations', 'code'

### **6.21.3 Outputs**

none

#### 6.21.4 More About

### 6.21.5 Examples

See also:

Help for rfvar/report is inherited from superclass RISE\_GENERIC

## 6.22 rfvar

~~ no help found

### 6.23 set

set - sets options for RISE models

### 6.23.1 Syntax

obj=set(obj,varargin)

### 6.23.2 Inputs

- obj [riseldsgelsvarlrfvar]: model object
- varargin: valid input arguments coming in pairs.

### **6.23.3 Outputs**

• obj [riseldsgelsvarlrfvar]: model object

6.22. rfvar 55

#### 6.23.4 More About

• one can force a new field into the options by prefixing it with a '+' sign. Let's say yourfield is not part of the options and you would like to force it to be in the options because it is going to be used in some function or algorithm down the road. Then you can run m=set(m,'+yourfield',value). then m will be part of the new options.

## 6.23.5 Examples

See also:

Help for rfvar/set is inherited from superclass RISE\_GENERIC

## 6.24 set\_solution\_to\_companion

H1 line

6.24.1 Syntax

6.24.2 Inputs

**6.24.3 Outputs** 

6.24.4 More About

#### 6.24.5 Examples

See also:

Help for rfvar/set\_solution\_to\_companion is inherited from superclass SVAR

## 6.25 simulate

simulate - simulates a RISE model

## 6.25.1 Syntax

[db, states, retcode] = simulate(obj, varargin)

### 6.25.2 Inputs

- **obj** [rfvarldsgelriselsvar]: model object
- varargin: additional arguments including but not restricted to
  - simul\_periods [integer|{100}]: number of simulation periods
  - simul\_burn [integer|{100}]: number of burn-in periods
  - simul\_algo [[{mt19937ar}| mcg16807|mlfg6331\_64|mrg32k3a| shr3cong|swb2712]]: matlab's seeding algorithms
  - **simul\_seed** [numeric|{0}]: seed of the computations
  - simul\_historical\_data [tslstructl{''}]: historical data from which the simulations are based. If empty, the simulations start at the steady state.
  - simul\_history\_end\_date [charlinteger|serial date]: last date of history
  - simul\_regime [integer|vector|{[]}]: regimes for which the model is simulated
  - simul\_update\_shocks\_handle [function handle]: we may want to update the shocks if some condition on the state of the economy is satisfied. For instance, shock monetary policy to keep the interest rate at the floor for an extented period of time if we already are at the ZLB/ZIF. simul\_update\_shocks\_handle takes as inputs the current shocks and the state vector (all the endogenous variables) and returns the updated shocks. But for all this to be put into motion, the user also has to turn on simul\_do\_update\_shocks by setting it to true.
  - simul\_do\_update\_shocks [true|{false}]: update the shocks based on simul\_update\_shocks\_handle or not.
  - simul\_to\_time\_series [{true}|false]: if true, the output is a time series, else a cell array with a matrix and information on elements that help reconstruct the time series.

## **6.25.3 Outputs**

- **db** [structcell array]: if **simul\_to\_time\_series** is true, the output is a time series, else a cell array with a matrix and information on elements that help reconstruct the time series.
- states [vector]: history of the regimes over the forecast horizon
- **retcode** [integer]: if 0, the simulation went fine. Else something got wrong. In that case one can understand the problem by running decipher(retcode)

#### 6.25.4 More About

• **simul\_historical\_data** contains the historical data as well as conditional information over the forecast horizon. It may also include as an alternative to **simul\_regime**, a time series with name **regime**, which indicates the regimes over the forecast horizon.

### 6.25.5 Examples

See also:

Help for rfvar/simulate is inherited from superclass RISE\_GENERIC

6.25. simulate 57

# 6.26 simulation\_diagnostics

H1 line

6.26.1 Syntax

6.26.2 Inputs

**6.26.3 Outputs** 

6.26.4 More About

6.26.5 Examples

See also:

Help for rfvar/simulation\_diagnostics is inherited from superclass RISE\_GENERIC

## **6.27** solve

H1 line

6.27.1 Syntax

6.27.2 Inputs

**6.27.3 Outputs** 

6.27.4 More About

6.27.5 Examples

See also:

## 6.28 stoch\_simul

- 6.28.1 Syntax
- 6.28.2 Inputs
- **6.28.3 Outputs**
- 6.28.4 More About
- 6.28.5 Examples

Help for rfvar/stoch\_simul is inherited from superclass RISE\_GENERIC

## 6.29 structural\_form

structural\_form finds A structural form given the imposed restrictions

### 6.29.1 Syntax

```
newobj=structural_form(obj)
newobj=structural_form(obj, varargin)
```

## 6.29.2 Inputs

- **obj**: [rfvar]: reduced form VAR object
- varargin : standard optional inputs **coming in pairs**. Among which:
  - restrict\_lags : [cell arrayl{''}] : restrictions on the lag structure. There are two equivalent syntaxes for this:
    - \* {'var name1@var name2{lag}'}
    - \* {'alag(var\_name1,var\_name2)'}: here alag should be understood as a-lag, where lag is the "lag" e.g. a1(infl,unemp) means unemp does not enter the infl equation at lag 1.
  - restrict\_irf\_sign: [cell arrayl{''}]: sign restrictions on the impulse responses. The general syntax is {'var\_name{period}@shock\_name','sign'} and the default period is "0" (for contemporaneous). That means {'var\_name{0}@shock\_name','+'} and {'var\_name@shock\_name','+'} are equivalent
  - restrict\_irf\_zero: [cell arrayl{''}]: zero restrictions on the impulse responses. The general syntax is {'var\_name{period}@shock\_name'} and the default period is "0" (for contemporaneous). That means {'var\_name{0}@shock\_name'} and {'var\_name@shock\_name'} are equivalent
  - structural\_shocks: [cell arrayl{''}]: List of structural shocks. The shock names can be entered with or without their description. For instance: {'E\_PAI', 'E\_U', 'E\_MP'} {'E\_PAI', "inflation shock"', 'E\_U', "unempl shock"', 'E\_MP'}
  - irf\_sample\_max [[numericl{10000}]][maximum number of trials in] the drawing of rotation matrices

## **6.29.3 Outputs**

• newobj : [rfvar]: new rfvar object with the drawn structural form

#### 6.29.4 More About

- RISE automatically orders the endogenous variables alphabetically and tags each equation with one of the endogenous variables. This may be useful for understanding the behavior of **restrict\_lags** above.
- The Choleski identification scheme is not implemented per se. The user has to explicitly enter the zeros in the right places. This gives the flexibility in implementing the restrictions. For instance, one could imagine a scheme in which choleski restrictions hold only in the long run.
- With only zero restrictions, one cannot expect the impulse responses to automatically have the correct sign. The
  rotation imposes zero restrictions but not the sign. If you would like to have correctly-signed impulse responses
  there are two choices: explicitly add sign restrictions multiply the impulse responses for the wrongly-signed
  shock with minus.
- If the signs are not explicitly enforced under zeros restrictions, in an exercise in which one draws many rotations, some will have one sign and some others a diffferent sign. Here, perhaps more than elsewhere, it is important to add some sign restrictions to have consistent results throughout.
- Many periods can be entered simultaneously. For instance 'var\_name{0,3,5,10:20,inf}@shock\_name'
- long-run restrictions are denoted by "inf". For instance 'var\_name{inf}@shock\_name'
- Identification for Markov switching VARs is not implemented/supported.

## 6.29.5 Examples

See also:

## 6.30 template

~~ no help found

## 6.31 theoretical\_autocorrelations

- 6.31.1 Syntax
- 6.31.2 Inputs
- **6.31.3 Outputs**
- 6.31.4 More About
- 6.31.5 Examples

Help for rfvar/theoretical\_autocorrelations is inherited from superclass RISE\_GENERIC

## 6.32 theoretical\_autocovariances

H1 line

- 6.32.1 Syntax
- 6.32.2 Inputs
- **6.32.3 Outputs**
- 6.32.4 More About
- 6.32.5 Examples

See also:

Help for rfvar/theoretical\_autocovariances is inherited from superclass RISE\_GENERIC

# 6.33 variance\_decomposition

- 6.33.1 Syntax
- 6.33.2 Inputs
- **6.33.3 Outputs**
- 6.33.4 More About
- 6.33.5 Examples

Help for rfvar/variance\_decomposition is inherited from superclass RISE\_GENERIC

# STRUCTURAL VAR MODELING

### 7.1 methods

- [ check\_optimum ](svar/check\_optimum)
- [ draw\_parameter ](svar/draw\_parameter)
- [estimate](svar/estimate)
- [ forecast ](svar/forecast)
- [ get ](svar/get)
- [historical\_decomposition](svar/historical\_decomposition)
- [ irf ](svar/irf)
- [isnan](svar/isnan)
- [load\_parameters](svar/load\_parameters)
- [log\_marginal\_data\_density](svar/log\_marginal\_data\_density)
- [ log\_posterior\_kernel ](svar/log\_posterior\_kernel)
- [log\_prior\_density](svar/log\_prior\_density)
- [ msvar\_priors ](svar/msvar\_priors)
- [posterior\_marginal\_and\_prior\_densities](svar/posterior\_marginal\_and\_prior\_densities)
- [ posterior\_simulator ](svar/posterior\_simulator)
- [ print\_estimation\_results ](svar/print\_estimation\_results)
- [ prior\_plots ](svar/prior\_plots)
- [report](svar/report)
- [ set ](svar/set)
- [ set\_solution\_to\_companion ](svar/set\_solution\_to\_companion)
- [ simulate ](svar/simulate)
- [ simulation diagnostics ](svar/simulation diagnostics)
- [solve](svar/solve)
- [ stoch\_simul ](svar/stoch\_simul)
- [svar](svar/svar)

- [template](svar/template)
- [theoretical\_autocorrelations](svar/theoretical\_autocorrelations)
- [ theoretical\_autocovariances ](svar/theoretical\_autocovariances)
- [variance\_decomposition](svar/variance\_decomposition)

## 7.2 properties

- [constant] -
- [nlags] -
- [legend] -
- [endogenous] -
- [exogenous] -
- [parameters] -
- [observables] -
- [markov\_chains] -
- [options] -
- [estimation] -
- [solution] -
- [filtering] -

## 7.3 check\_optimum

H1 line

- **7.3.1 Syntax**
- **7.3.2 Inputs**
- 7.3.3 Outputs
- 7.3.4 More About

## 7.3.5 Examples

See also:

Help for svar/check\_optimum is inherited from superclass RISE\_GENERIC

#### 7.4 draw parameter

H1 line

- **7.4.1 Syntax**
- **7.4.2 Inputs**
- 7.4.3 Outputs
- 7.4.4 More About

#### 7.4.5 Examples

See also:

Help for svar/draw\_parameter is inherited from superclass RISE\_GENERIC

#### 7.5 estimate

estimate - estimates the parameters of a RISE model

#### **7.5.1 Syntax**

```
obj=estimate(obj)
obj=estimate(obj,varargin)
```

#### **7.5.2 Inputs**

- obj [riseldsgelrfvarlsvar]: model object
- varargin additional optional inputs among which the most relevant for estimation are:
- estim\_parallel [integer|{1}]: Number of starting values
- estim\_start\_from\_mode [true|false|{[]}]: when empty, the user is prompted to answer the question as to whether to start estimation from a previously found mode or not. If true or false, no question is asked.
- estim\_start\_date [numericlcharlserial date]: date of the first observation to use in the dataset provided for estimation
- **estim\_end\_date** [numericlcharlserial date]: date of the last observation to use in the dataset provided for estimation
- estim\_max\_trials [integer|{500}]: When the initial value of the log-likelihood is too low, RISE uniformly draws from the prior support in search for a better starting point. It will try this for a maximum number of estim\_max\_trials times before squeaking with an error.
- estim\_start\_vals [{[]}|struct]: when not empty, the parameters whose names are fields of the structure will see their start values updated or overriden by the information in estim\_start\_vals. There is no need to provide values to update the start values for the estimated parameters.

- estim\_general\_restrictions [{[]}|function handle]: when not empty, the argument should be a function handle that takes as input a parameterized RISE object and returns a scalar or vector of numbers representing the strength of the violation of the nonlinear constraints. Those constraints will be added to the constraints already included in a rise/dsge file before being presented to the optimization function.
- estim\_linear\_restrictions [{[]}|cell]: This is most often used in the estimation of rfvar or svar models either to impose block exogeneity or to impose other forms of linear restrictions. When not empty, estim\_linear\_restrictions must be a 2-column cell: Each row of the first column represents a particular linear combination of the estimated parameters. Those linear combinations are constructed using the coef class. Check help for coef.coef for more details. Each row of the second column holds the value of the linear combination.
- estim\_blocks [{[]}|cell]: When not empty, this triggers blockwise optimization. For further information on how to set blocks, see help for dsge.create\_estimation\_blocks
- estim\_priors [{[]}|struct]: This provides an alternative to setting priors inside the rise/dsge model file. Each field of the structure must be the name of an estimated parameter. Each field will hold a cell array whose structure is described in help rise\_generic.setup\_priors.
- estim\_penalty [numericl{1e+8}]: value of the objective function when a problem occurs. Possible problems include: no solution found very low likelihood stochastic singularity problems computing the initial covariance matrix non-positive definite covariance matrices etc.
- estim\_cond\_vars [charlcellstr]: list of the variables to condition on during estimation of a dsge model. This then assumes that the data provided for estimation have several pages. The first page is the actual data, while the subsequent pages are the expectations data.
- optimset [struct]: identical to matlab's optimset
- optimizer [charlfunction handlelcell]: This can be the name of a standard matlab optimizer or RISE optimization routine or a user-defined optimization procedure available of the matlab search path. If the optimizer is provided as a cell, then the first element of the cell is the name of the optimizer or its handle and the remaining entries in the cell are additional input arguments to the user-defined optimization routine. A user-defined optimization function should have the following syntax

```
[xfinal, ffinal, exitflag] = optimizer (fh, x0, lb, ub, options, varargin);\\
```

#### That is, it accepts as inputs:

- **fh**: the function to optimize
- x0: a vector column of initial values of the parameters
- **lb**: a vector column of lower bounds
- ub: a vector column of upper bounds
- options: a structure of options whose fields will be similar to matlab's optimset
- varargin: additional arguments to the user-defined optimization procedure

#### That is, it provides as outputs:

- xfinal: the vector of final values
- ffinal: the value of fh at xfinal
- exitflag: a flag similar to the ones provided by matlab's

optimization functions.

- hessian\_type [{'fd'}l'opg']: The hessian is either computed by finite differences (fd) or by outer-product-gradient (opg)
- hessian\_repair [{false}ltrue]: If the Hessian is not positive definite, it nevertheless can be repaired and prepared for a potential mcmc simulation.

#### 7.5.3 Outputs

• **obj** [riseldsgelrfvarlsvar]: model object parameterized with the mode found and holding additional estimation results and statistics that can be found under obj.estimation

#### 7.5.4 More About

recursive estimation may be done easily by passing a different estim\_end\_date at the beginning of each estimation run.

#### 7.5.5 Examples

See also:

Help for svar/estimate is inherited from superclass RISE\_GENERIC

#### 7.6 forecast

forecast - computes forecasts for riseldsgelsvarlrfvar models

#### **7.6.1 Syntax**

```
cond_fkst_db=forecast(obj,varargin)
```

#### **7.6.2 Inputs**

- obj [riseldsgelsvarlrfvar]: model object
- varargin : additional inputs coming in pairs. These include but are not restricted to: forecast\_to\_time\_series [{true}|false]: sets the output to time

series format or not

- forecast\_nsteps [integer|{12}]: number of forecasting steps
- forecast\_start\_date [charlnumericlserial date]: date when the forecasts start (end of history + 1)
- forecast\_conditional\_hypothesis [{jma}|ncp|nas]: in dsge models in which agents have information beyond the current period, this option determines the number of periods of shocks need to match the restrictions: Hypothesis jma assumes that irrespective of how

many periods of conditioning information are remaining, agents always receive information on the same number of shocks.

- \* Hypothesis ncp assumes there are as many shocks periods as the number of the number of conditioning periods
- \* Hypothesis nas assumes there are as many shocks periods as the number of anticipated steps

7.6. forecast 67

#### 7.6.3 Outputs

• **cond\_fkst\_db** [structlmatrix]: depending on the value of **forecast\_to\_time\_series** the returned output is a structure with time series or a cell containing a matrix and the information to reconstruct the time series.

#### 7.6.4 More About

- the historical information as well as the conditioning information come from the same database. The time series must be organized such that for each series, the first page represents the actual data and all subsequent pages represent conditional information. If a particular condition is "nan", that location is not constrained
- Conditional forecasting for nonlinear models is also supported. However, the solving of the implied nonlinear problem may fail if the model displays instability
- Both HARD CONDITIONS and SOFT CONDITIONS are implemented but the latter are currently disabled in expectation of a better user interface.
- The data may also contain time series for a variable with name **regime** in that case, the forecast/simulation paths are computed following the information therein. **regime** must be a member of 1:h, where h is the maximum number of regimes.

#### 7.6.5 Examples

See also: simulate

Help for svar/forecast is inherited from superclass RISE\_GENERIC

## 7.7 get

H1 line

**7.7.1 Syntax** 

**7.7.2 Inputs** 

7.7.3 Outputs

7.7.4 More About

#### 7.7.5 Examples

See also:

Help for svar/get is inherited from superclass RISE\_GENERIC

# 7.8 historical\_decomposition

historical\_decomposition Computes historical decompositions of a DSGE model

#### **7.8.1 Syntax**

```
[Histdec,obj]=history_dec(obj)
[Histdec,obj]=history_dec(obj,varargin)
```

#### **7.8.2 Inputs**

- obj : [riseldsgelrfvarlsvar] model(s) for which to compute the decomposition. obj could be a vector of models
- varargin: standard optional inputs coming in pairs. Among which: histdec\_start\_date: [charlnumericl{''}]
   date at which the

decomposition starts. If empty, the decomposition starts at he beginning of the history of the dataset

#### 7.8.3 Outputs

Histdec: [structlcell array] structure or cell array of structures with the decompositions in each model. The decompositions are given in terms of: - the exogenous variables - InitialConditions: the effect of initial conditions - risk: measure of the effect of non-certainty equivalence - switch: the effect of switching (which is also a shock!!!) - steady\_state: the contribution of the steady state

#### 7.8.4 Remarks

- the elements that do not contribute to any of the variables are automatically discarded.
- **N.B**: a switching model is inherently nonlinear and so, strictly speaking, the type of decomposition we do for linear/linearized constant-parameter models is not feasible. RISE takes an approximation in which the variables, shocks and states matrices across states are averaged. The averaging weights are the smoothed probabilities.

#### 7.8.5 Examples

See also:

Help for svar/historical\_decomposition is inherited from superclass RISE\_GENERIC

#### 7.9 irf

irf - computes impulse responses for a RISE model

#### **7.9.1 Syntax**

```
myirfs=irf(obj)
myirfs=irf(obj,varargin)
```

#### **7.9.2 Inputs**

- **obj** [riseldsgelrfvarlsvar]: single or vector of RISE models
- varargin : optional options coming in pairs. The notable ones that will influence the behavior of the impulse responses are:
- irf\_shock\_list [charlcellstrl{''}]: list of shocks for which we want to compute impulse responses
- irf\_var\_list [charlcellstrl{''}]: list of the endogenous variables we want to report
- irf\_periods [integer|{40}]: length of the irfs
- **irf\_shock\_sign** [numericl-1|{1}]: sign or scale of the original impulse. If **irf\_shock\_sign** >0, we get impulse responses to a positive shock. If **irf\_shock\_sign** <0, the responses are negative. If **irf\_shock\_sign** =0, all the responses are 0.
- **irf\_draws** [integerl{50}]: number of draws used in the simulation impulse responses in a nonlinear model. A nonlinear model is defined as a model that satisfies at least one of the following criteria solved at an order >1 has more than one regime and option **irf\_regime\_specific** below is

set to false

- irf\_type [{irf}|girf]: type of irfs. If the type is irf, the impulse responses are computed directly exploiting the fact that the model is linear. If the type is girf, the formula for the generalized impulse responses is used: the irf is defined as the expectation of the difference of two simulation paths. In the first path the initial impulse for the shock of interest is nonzero while it is zero for the second path. All other shocks are the same for both paths in a given simulation.
- irf\_regime\_specific [{true}|false]: In a switching model, we may or may not want to compute impulse responses specific to each regime.
- irf\_use\_historical\_data [{false}|true]: if true, the data stored in option simul\_historical\_data are used as initial conditions. But the model has to be nonlinear otherwise the initial conditions are set to zero. This option gives the flexibility to set the initial conditions for the impulse responses.
- **irf\_to\_time\_series** [{true}|false]: If true, the output is in the form of time series. Else it is in the form of a cell containing the information needed to reconstruct the time series.

#### 7.9.3 Outputs

• myirfs [{struct}|cell]: Impulse response data

#### 7.9.4 More About

- for linear models or models solved up to first order, the initial conditions as well as the steady states are set to 0 in the computation of the impulse responses.
- for nonlinear models, the initial conditions is the ergodic mean

#### 7.9.5 Examples

See also:

Help for svar/irf is inherited from superclass RISE\_GENERIC

#### **7.10** isnan

H1 line

- 7.10.1 Syntax
- 7.10.2 Inputs
- **7.10.3 Outputs**
- 7.10.4 More About
- 7.10.5 Examples

See also:

Help for svar/isnan is inherited from superclass RISE\_GENERIC

# 7.11 load\_parameters

H1 line

- 7.11.1 Syntax
- 7.11.2 Inputs
- **7.11.3 Outputs**
- 7.11.4 More About
- 7.11.5 Examples

See also:

Help for svar/load\_parameters is inherited from superclass RISE\_GENERIC

7.10. isnan 71

# 7.12 log\_marginal\_data\_density

H1 line

- 7.12.1 Syntax
- **7.12.2 Inputs**
- **7.12.3 Outputs**
- 7.12.4 More About
- 7.12.5 Examples

See also:

Help for svar/log\_marginal\_data\_density is inherited from superclass RISE\_GENERIC

# 7.13 log\_posterior\_kernel

H1 line

- 7.13.1 Syntax
- 7.13.2 Inputs
- **7.13.3 Outputs**
- 7.13.4 More About
- 7.13.5 Examples

See also:

Help for svar/log\_posterior\_kernel is inherited from superclass RISE\_GENERIC

# 7.14 log\_prior\_density

H1 line

- 7.14.1 Syntax
- **7.14.2 Inputs**
- **7.14.3 Outputs**
- 7.14.4 More About
- 7.14.5 Examples

See also:

Help for svar/log\_prior\_density is inherited from superclass RISE\_GENERIC

# 7.15 msvar\_priors

H1 line

- 7.15.1 Syntax
- 7.15.2 Inputs
- **7.15.3 Outputs**
- 7.15.4 More About
- 7.15.5 Examples

See also:

## 7.16 posterior\_marginal\_and\_prior\_densities

H1 line

- 7.16.1 Syntax
- 7.16.2 Inputs
- **7.16.3 Outputs**
- 7.16.4 More About
- 7.16.5 Examples

See also:

7.15. msvar\_priors 73

Help for svar/posterior\_marginal\_and\_prior\_densities is inherited from superclass RISE\_GENERIC

# 7.17 posterior\_simulator

H1 line

- 7.17.1 Syntax
- 7.17.2 Inputs
- **7.17.3 Outputs**
- 7.17.4 More About
- 7.17.5 Examples

See also:

Help for svar/posterior\_simulator is inherited from superclass RISE\_GENERIC

# 7.18 print\_estimation\_results

H1 line

- 7.18.1 Syntax
- 7.18.2 Inputs
- **7.18.3 Outputs**
- 7.18.4 More About
- 7.18.5 Examples

See also:

Help for svar/print\_estimation\_results is inherited from superclass RISE\_GENERIC

# 7.19 prior\_plots

H1 line

7.19.1 Syntax

7.19.2 Inputs

**7.19.3 Outputs** 

7.19.4 More About

#### 7.19.5 Examples

See also:

Help for svar/prior\_plots is inherited from superclass RISE\_GENERIC

# 7.20 report

**REPORT** assigns the elements of interest to a rise\_report.report object

#### 7.20.1 Syntax

::

- REPORT(rise.empty(0)) : displays the default inputs
- REPORT(obj,destination\_root,rep\_items): assigns the reported elements in rep\_items to destination\_root
- REPORT(obj,destination\_root,rep\_items,varargin): assigns varargin to obj before doing the rest

#### **7.20.2 Inputs**

- obj : [riseldsge]
- destination\_root : [rise\_report.report] : handle for the actual report
- rep\_items: [charlcellstr]: list of desired items to report. This list can only include: 'endogenous', 'exogenous', 'observables', 'parameters', 'solution', 'estimation', 'estimation\_statistics', 'equations', 'code'

#### **7.20.3 Outputs**

none

#### 7.20.4 More About

#### 7.20.5 Examples

See also:

Help for svar/report is inherited from superclass RISE\_GENERIC

7.20. report 75

#### 7.21 set

set - sets options for RISE models

#### 7.21.1 Syntax

obj=set(obj,varargin)

#### **7.21.2 Inputs**

- obj [riseldsgelsvarlrfvar]: model object
- varargin: valid input arguments coming in pairs.

#### **7.21.3 Outputs**

• obj [riseldsgelsvarlrfvar]: model object

#### 7.21.4 More About

• one can force a new field into the options by prefixing it with a '+' sign. Let's say yourfield is not part of the options and you would like to force it to be in the options because it is going to be used in some function or algorithm down the road. Then you can run m=set(m,'+yourfield',value). then m will be part of the new options.

#### 7.21.5 Examples

See also:

Help for svar/set is inherited from superclass RISE\_GENERIC

# 7.22 set\_solution\_to\_companion

H1 line

- 7.22.1 Syntax
- **7.22.2 Inputs**
- **7.22.3 Outputs**
- 7.22.4 More About
- 7.22.5 Examples

See also:

#### 7.23 simulate

simulate - simulates a RISE model

#### 7.23.1 Syntax

[db, states, retcode] = simulate(obj, varargin)

#### **7.23.2 Inputs**

- obj [rfvarldsgelriselsvar]: model object
- varargin: additional arguments including but not restricted to
  - simul\_periods [integer|{100}]: number of simulation periods
  - simul\_burn [integer|{100}]: number of burn-in periods
  - simul\_algo [[{mt19937ar}| mcg16807|mlfg6331\_64|mrg32k3a| shr3conglswb2712]]: matlab's seeding algorithms
  - $simul\_seed$  [numeric|{0}]: seed of the computations
  - simul\_historical\_data [tslstructl{''}]: historical data from which the simulations are based. If empty, the simulations start at the steady state.
  - simul\_history\_end\_date [charlinteger|serial date]: last date of history
  - simul\_regime [integer|vector|{[]}]: regimes for which the model is simulated
  - simul\_update\_shocks\_handle [function handle]: we may want to update the shocks if some condition on the state of the economy is satisfied. For instance, shock monetary policy to keep the interest rate at the floor for an extented period of time if we already are at the ZLB/ZIF. simul\_update\_shocks\_handle takes as inputs the current shocks and the state vector (all the endogenous variables) and returns the updated shocks. But for all this to be put into motion, the user also has to turn on simul do update shocks by setting it to true.
  - simul\_do\_update\_shocks [true|{false}]: update the shocks based on simul\_update\_shocks\_handle
    or not.
  - simul\_to\_time\_series [{true}|false]: if true, the output is a time series, else a cell array with a matrix and information on elements that help reconstruct the time series.

#### **7.23.3 Outputs**

- **db** [structlcell array]: if **simul\_to\_time\_series** is true, the output is a time series, else a cell array with a matrix and information on elements that help reconstruct the time series.
- states [vector]: history of the regimes over the forecast horizon
- **retcode** [integer]: if 0, the simulation went fine. Else something got wrong. In that case one can understand the problem by running decipher(retcode)

7.23. simulate 77

#### 7.23.4 More About

• **simul\_historical\_data** contains the historical data as well as conditional information over the forecast horizon. It may also include as an alternative to **simul\_regime**, a time series with name **regime**, which indicates the regimes over the forecast horizon.

#### 7.23.5 Examples

See also:

Help for svar/simulate is inherited from superclass RISE\_GENERIC

## 7.24 simulation\_diagnostics

H1 line

7.24.1 Syntax

7.24.2 Inputs

**7.24.3 Outputs** 

7.24.4 More About

#### 7.24.5 Examples

See also:

Help for svar/simulation\_diagnostics is inherited from superclass RISE\_GENERIC

#### **7.25** solve

H1 line

7.25.1 Syntax

7.25.2 Inputs

**7.25.3 Outputs** 

7.25.4 More About

7.25.5 Examples

See also:

# 7.26 stoch\_simul

H1 line

7.26.1 Syntax

7.26.2 Inputs

**7.26.3 Outputs** 

7.26.4 More About

#### 7.26.5 Examples

See also:

Help for svar/stoch\_simul is inherited from superclass RISE\_GENERIC

#### 7.27 svar

~~ no help found

# 7.28 template

~~ no help found

# 7.29 theoretical\_autocorrelations

H1 line

7.26. stoch\_simul 79

7.29.1 Syntax

7.29.2 Inputs

**7.29.3 Outputs** 

7.29.4 More About

7.29.5 Examples

See also:

Help for svar/theoretical\_autocorrelations is inherited from superclass RISE\_GENERIC

# 7.30 theoretical\_autocovariances

H1 line

7.30.1 Syntax

7.30.2 Inputs

**7.30.3 Outputs** 

7.30.4 More About

7.30.5 Examples

See also:

Help for svar/theoretical\_autocovariances is inherited from superclass RISE\_GENERIC

# 7.31 variance\_decomposition

H1 line

- 7.31.1 Syntax
- 7.31.2 Inputs
- **7.31.3 Outputs**
- 7.31.4 More About
- 7.31.5 Examples

See also:

Help for svar/variance\_decomposition is inherited from superclass RISE\_GENERIC

**CHAPTER** 

**EIGHT** 

# TIME SERIES

#### ts Methods:

acos - H1 line acosh - H1 line acot - H1 line acoth - H1 line aggregate - H1 line allmean - H1 line and - H1 line apply - H1 line asin - H1 line asinh - H1 line atanh - H1 line atanh - H1 line automatic\_model\_selection - H1 line bar - H1 line barh - H1 line boxplot - H1 line bsxfun - H1 line cat - concatenates time series along the specified dimension collect - H1 line corr - H1 line corrcoef - H1 line cos - H1 line cosh - H1 line cot - H1 line coth - H1 line cov - H1 line ctranspose - H1 line cumprod - H1 line cumsum - H1 line decompose series - H1 line describe - H1 line display -H1 line double - H1 line drop - H1 line dummy - H1 line eq - H1 line exp - H1 line expanding - H1 line fanchart - H1 line ge - H1 line get - H1 line gt - H1 line head - H1 line hist - H1 line horzcat - H1 line hpfilter - H1 line index - H1 line interpolate - H1 line intersect - H1 line isfinite - H1 line isinf - H1 line isnan - H1 line jbtest - H1 line kurtosis -H1 line le - H1 line log - H1 line lt - H1 line max - H1 line mean - H1 line median - H1 line min - H1 line minus - H1 line mode - H1 line mpower - H1 line mrdivide - H1 line mtimes - H1 line nan - H1 line ne - H1 line numel - H1 line ones - overloads ones for ts objects pages2struct - H1 line plot - H1 line plotyy - H1 line plus - H1 line power - H1 line prctile - Percentiles of a time series (ts) quantile - H1 line rand - H1 line rand - H1 line range - H1 line rdivide - H1 line regress - H1 line reset\_start\_date - H1 line rolling - H1 line sin - H1 line sinh - H1 line skewness - H1 line sort -H1 line spectrum - H1 line std - H1 line step\_dummy - H1 line subsasgn - H1 line subsref - H1 line sum - H1 line tail - H1 line times - H1 line transform - H1 line transpose - H1 line ts - Methods: uminus - H1 line values - H1 line var -H1 line zeros - H1 line

#### ts Properties:

varnames - names of the variables in the database start - time of the time series finish - end time of the time series frequency - of the time series NumberOfObservations - number of observations in the time series NumberOfPages - number of pages (third dimension) of the time series NumberOfVariables - number of variables in the time series

# MARKOV CHAIN MONTE CARLO FOR BAYESIAN ESTIMATION

- 9.1 Metropolis Hastings
- 9.2 Gibbs sampling
- 9.3 Marginal data density
- 9.3.1 Laplace approximation
- 9.3.2 Modified harmonic mean
- 9.3.3 Waggoner and Zha (2008)
- 9.3.4 Mueller
- 9.3.5 Chib and Jeliazkov

**CHAPTER** 

**TEN** 

# **DERIVATIVE-FREE OPTIMIZATION**

- differential evolution
- bee algorithm
- biogeography
- studga
- ants

# **MONTE CARLO FILTERING**

#### 11.1 methods

- [ addlistener ](mcf/addlistener)
- [cdf](mcf/cdf)
- [ cdf\_plot ](mcf/cdf\_plot)
- [correlation\_patterns\_plot](mcf/correlation\_patterns\_plot)
- [ delete ](mcf/delete)
- [ eq ](mcf/eq)
- [ findobj ](mcf/findobj)
- [ findprop ](mcf/findprop)
- [ ge ](mcf/ge)
- [ gt ](mcf/gt)
- [ isvalid ](mcf/isvalid)
- [kolmogorov\_smirnov\_test](mcf/kolmogorov\_smirnov\_test)
- [ le ](mcf/le)
- [ lt ](mcf/lt)
- [ mcf ](mcf/mcf)
- [ ne ](mcf/ne)
- [ notify ](mcf/notify)
- [ scatter ](mcf/scatter)

# 11.2 properties

- [lb] -
- [ub] -
- [nsim] -
- [procedure] -

- [parameter\_names] -
- [samples] -
- [is\_behaved] -
- [nparam] -
- [is sampled] -
- [check behavior] -
- [number\_of\_outputs] -
- [user\_outputs] -
- [known\_procedures] -

#### 11.3 addlistener

**ADDLISTENER Add listener for event.** el = ADDLISTENER(hSource, 'Eventname', Callback) creates a listener for the event named Eventname, the source of which is handle object hSource. If hSource is an array of source handles, the listener responds to the named event on any handle in the array. The Callback is a function handle that is invoked when the event is triggered.

el = ADDLISTENER(hSource, PropName, 'Eventname', Callback) adds a listener for a property event. Eventname must be one of the strings 'PreGet', 'PostGet', 'PreSet', and 'PostSet'. PropName must be either a single property name or cell array of property names, or a single meta.property or array of meta.property objects. The properties must belong to the class of hSource. If hSource is scalar, PropName can include dynamic properties.

For all forms, addlistener returns an event.listener. To remove a listener, delete the object returned by addlistener. For example, delete(el) calls the handle class delete method to remove the listener and delete it from the workspace.

See also MCF, NOTIFY, DELETE, EVENT.LISTENER, META.PROPERTY, EVENTS, DYNAM-ICPROPS

Help for mcf/addlistener is inherited from superclass HANDLE

Reference page in Help browser doc mcf/addlistener

#### 11.4 cdf

~~ no help found

# 11.5 cdf\_plot

~~ no help found

## 11.6 correlation\_patterns\_plot

~~ no help found

#### 11.7 delete

**DELETE Delete a handle object.** The DELETE method deletes a handle object but does not clear the handle from the workspace. A deleted handle is no longer valid.

DELETE(H) deletes the handle object H, where H is a scalar handle.

See also MCF, MCF/ISVALID, CLEAR

Help for mcf/delete is inherited from superclass HANDLE

Reference page in Help browser doc mcf/delete

## 11.8 eq

== (EQ) Test handle equality. Handles are equal if they are handles for the same object.

H1 == H2 performs element-wise comparisons between handle arrays H1 and H2. H1 and H2 must be of the same dimensions unless one is a scalar. The result is a logical array of the same dimensions, where each element is an element-wise equality result.

If one of H1 or H2 is scalar, scalar expansion is performed and the result will match the dimensions of the array that is not scalar.

TF = EQ(H1, H2) stores the result in a logical array of the same dimensions.

See also MCF, MCF/GE, MCF/GT, MCF/LE, MCF/LT, MCF/NE

Help for mcf/eq is inherited from superclass HANDLE

11.5. cdf\_plot 91

## 11.9 findobj

**FINDOBJ Find objects matching specified conditions.** The FINDOBJ method of the HANDLE class follows the same syntax as the MATLAB FINDOBJ command, except that the first argument must be an array of handles to objects.

HM = FINDOBJ(H, <conditions>) searches the handle object array H and returns an array of handle objects matching the specified conditions. Only the public members of the objects of H are considered when evaluating the conditions.

See also FINDOBJ, MCF

Help for mcf/findobj is inherited from superclass HANDLE

Reference page in Help browser doc mcf/findobj

# 11.10 findprop

**FINDPROP Find property of MATLAB handle object.** p = FINDPROP(H,'PROPNAME') finds and returns the META.PROPERTY object associated with property name PROPNAME of scalar handle object H. PROPNAME must be a string. It can be the name of a property defined by the class of H or a dynamic property added to scalar object H.

If no property named PROPNAME exists for object H, an empty META.PROPERTY array is returned.

See also MCF, MCF/FINDOBJ, DYNAMICPROPS, META.PROPERTY

Help for mcf/findprop is inherited from superclass HANDLE

Reference page in Help browser doc mcf/findprop

# 11.11 ge

>= (GE) Greater than or equal relation for handles. H1 >= H2 performs element-wise comparisons between handle arrays H1 and H2. H1 and H2 must be of the same dimensions unless one is a scalar. The result is a logical array of the same dimensions, where each element is an element-wise >= result.

If one of H1 or H2 is scalar, scalar expansion is performed and the result will match the dimensions of the array that is not scalar.

TF = GE(H1, H2) stores the result in a logical array of the same dimensions.

See also MCF, MCF/EQ, MCF/GT, MCF/LE, MCF/LT, MCF/NE

Help for mcf/ge is inherited from superclass HANDLE

## 11.12 gt

> (GT) Greater than relation for handles. H1 > H2 performs element-wise comparisons between handle arrays H1 and H2. H1 and H2 must be of the same dimensions unless one is a scalar. The result is a logical array of the same dimensions, where each element is an element-wise > result.

If one of H1 or H2 is scalar, scalar expansion is performed and the result will match the dimensions of the array that is not scalar.

TF = GT(H1, H2) stores the result in a logical array of the same dimensions.

See also MCF, MCF/EQ, MCF/GE, MCF/LE, MCF/LT, MCF/NE

Help for mcf/gt is inherited from superclass HANDLE

#### 11.13 isvalid

**ISVALID Test handle validity.** TF = ISVALID(H) performs an element-wise check for validity on the handle elements of H. The result is a logical array of the same dimensions as H, where each element is the element-wise validity result.

A handle is invalid if it has been deleted or if it is an element of a handle array and has not yet been initialized.

See also MCF, MCF/DELETE

Help for mcf/isvalid is inherited from superclass HANDLE

Reference page in Help browser doc mcf/isvalid

# 11.14 kolmogorov\_smirnov\_test

tests the equality of two distributions using their CDFs

#### 11.15 le

<= (LE) Less than or equal relation for handles. Handles are equal if they are handles for the same object. All comparisons use a number associated with each handle object. Nothing can be assumed about the result of a handle comparison except that the repeated comparison of two handles in the same MATLAB session will yield the same result. The order of handle values is purely arbitrary and has no connection to the state of the handle objects being compared.</p>

H1 <= H2 performs element-wise comparisons between handle arrays H1 and H2. H1 and H2 must be of the same dimensions unless one is a scalar. The result is a logical array of the same dimensions, where each element is an element-wise >= result.

If one of H1 or H2 is scalar, scalar expansion is performed and the result will match the dimensions of the array that is not scalar.

11.12. gt 93

TF = LE(H1, H2) stores the result in a logical array of the same dimensions.

See also MCF, MCF/EQ, MCF/GE, MCF/GT, MCF/LT, MCF/NE

Help for mcf/le is inherited from superclass HANDLE

#### 11.16 lt

< (LT) Less than relation for handles. H1 < H2 performs element-wise comparisons between handle arrays H1 and H2. H1 and H2 must be of the same dimensions unless one is a scalar. The result is a logical array of the same dimensions, where each element is an element-wise < result.

If one of H1 or H2 is scalar, scalar expansion is performed and the result will match the dimensions of the array that is not scalar.

TF = LT(H1, H2) stores the result in a logical array of the same dimensions.

See also MCF, MCF/EQ, MCF/GE, MCF/GT, MCF/LE, MCF/NE

Help for mcf/lt is inherited from superclass HANDLE

#### 11.17 mcf

~~ no help found

#### 11.18 ne

~= (NE) Not equal relation for handles. Handles are equal if they are handles for the same object and are unequal otherwise.

 $H1 \sim H2$  performs element-wise comparisons between handle arrays H1 and H2. H1 and H2 must be of the same dimensions unless one is a scalar. The result is a logical array of the same dimensions, where each element is an element-wise equality result.

If one of H1 or H2 is scalar, scalar expansion is performed and the result will match the dimensions of the array that is not scalar.

TF = NE(H1, H2) stores the result in a logical array of the same dimensions.

See also MCF, MCF/EQ, MCF/GE, MCF/GT, MCF/LE, MCF/LT

Help for mcf/ne is inherited from superclass HANDLE

# 11.19 notify

**NOTIFY Notify listeners of event.** NOTIFY(H, EVENTNAME') notifies listeners added to the event named EVENTNAME on handle object array H that the event is taking place. H is the array of handles to objects triggering the event, and EVENTNAME must be a string.

NOTIFY(H,'EVENTNAME',DATA) provides a way of encapsulating information about an event which can then be accessed by each registered listener. DATA must belong to the EVENT.EVENTDATA class.

See also MCF, MCF/ADDLISTENER, EVENT.EVENTDATA, EVENTS

Help for mcf/notify is inherited from superclass HANDLE

Reference page in Help browser doc mcf/notify

#### 11.20 scatter

~~ no help found

11.19. notify 95

# HIGH DIMENSIONAL MODEL REPRESENTATION

#### 12.1 methods

- [ estimate ](hdmr/estimate)
- [ first\_order\_effect ](hdmr/first\_order\_effect)
- [ hdmr ](hdmr/hdmr)
- [ metamodel ](hdmr/metamodel)
- [ plot\_fit ](hdmr/plot\_fit)
- [polynomial\_evaluation](hdmr/polynomial\_evaluation)
- [ polynomial\_integration ](hdmr/polynomial\_integration)
- [polynomial\_multiplication](hdmr/polynomial\_multiplication)

# 12.2 properties

- [N] -
- [Nobs] -
- [n] -
- [output\_nbr] -
- [theta] -
- [theta\_low] -
- [theta\_high] -
- [g] -
- [x] -
- [expansion\_order] -
- [pol\_max\_order] -
- [poly\_coefs] -

• [Indices] -		
• [coefficients] -		
• [aggregate] -		
• [f0] -		
• [D] -		
• [sample_percentage] -		
• [optimal] -		
• [param_names] -		
12.3 estimate		
~~ no help found		
12.4 first_order_effect		
~~ no help found		
12.5 hdmr		
~~ no help found		
12.6 metamodel		
~~ no help found		
12.7 plot_fit		
~~ no help found		

# 12.8 polynomial\_evaluation

later on, the function that normalizes could come in here so that the normalization is done according to the hdmr\_type of polynomial chosen.

# 12.9 polynomial\_integration

polynomial is of the form  $a0+a1*x+...+ar*x^r$  the integral is then  $a0*x+a1/2*x^2+...+ar/(r+1)*x^(r+1)$ 

# 12.10 polynomial\_multiplication

each polynomial is of the form a0+a1\*x+...+ar\*x^r

# **CONTRIBUTING TO RISE**

- 13.1 contributing new code
- 13.2 contributing by helping maintain existing code
- 13.3 other ways to contribute
- 13.4 recommended development setup
- 13.5 RISE structure
- 13.6 useful links, FAQ, checklist

# **ACKNOWLEDGEMENTS**

Many people have, oftentimes unknowingly, provided help in the form of reporting bugs, making suggestions, asking challenging questions, etc. I would like to single out a few of them but the list is far from exhaustive:

- Dan Waggoner
- Doug Laxton
- Eric Leeper
- Jesper Linde
- Jim Nason
- Kjetil Olsen
- Kostas Theodoridis
- Leif Brubakk
- Marco Ratto
- · Michel Juillard
- Pablo Winnant (dolo)
- Pelin Ilbas
- Raf Wouters
- Tao Zha

**CHAPTER** 

# **FIFTEEN**

# **BIBLIOGRAPHY**

**CHAPTER** 

# **SIXTEEN**

# **INDICES AND TABLES**

- genindex
- modindex
- search