A MATLAB LIBRARY OF TEMPORAL DISAGGREGATION AND INTERPOLATION METHODS: SUMMARY

Enrique M. Quilis¹

Macroeconomic Research Department
Ministry of Economy and Finance
Paseo de la Castellana, 162
28046 – Madrid (SPAIN)

enrique.quilis@meh.es

June, 2009

¹ I am indebted to A. Abad, J. Bógalo, F. Cardoso, L. Navarro, T. Di Fonzo and S. Relloso for their help. I also thank J.R. Cancelo, A. Cuevas, R. Frutos, M. Jerez, F. Melis, A. Maravall, T. Proietti and E. Salazar for stimulating discussions on temporal disaggregation and benchmarking.

INTRODUCTION

This release² of the Matlab temporal disaggregation and interpolation library includes some new features:

- Univariate interpolation using low-frequency smoothing (via Hodrick-Prescott) and classical benchmarking (via Denton): low_pass_interpolation().
- A new function to perform backtesting (revisions and recursive estimates) of main procedures: backtest().
- Multivariate Denton procedure, proportional variant: denton_prop().
- Several utilities to perform temporal aggregation: moving_acc(), temporal_acc(), ssampler(), copylow().

The library includes a set of functions to perform temporal disaggregation, interpolation, extrapolation, and balancing, according to the following structure:

Univariate bechmarking without indicators:

- Boot-Feibes-Lisman.
- Stram-Wei (an ARIMA model-based method).
 - → Served by tduni_print() and tduni_plot()
 - → ... or by tdprint() and tdplot()
- Low-pass interpolation.

<u>Univariate bechmarking with indicators (QL optimization or model-based, BLUE methods)</u>:

- Denton: additive and proportional variants.
- Chow-Lin: maximum likelihood, weighted least squares, Cochrane-Orcutt.
- Fernández.
- Litterman.
- Santos Silva-Cardoso (dynamic extension of Chow-Lin).
 - → Served by td_print() and td_plot()
 - → ... or by tdprint() and tdplot()
- Guerrero (an ARIMA model-based method).
 - → Served by td print G() and td plot G()

Multivariate bechmarking with indicators and transversal contraint:

- Multivariate Denton: additive and proportional variants.
- Rossi.
- Di Fonzo.

_

 $^{^2}$ Previous versions were released on: December, 2002; November, 2003; August, 2004 and August, 2006.

- → Served by mtd_print() and mtd_plot()
- → ... or by tdprint() and tdplot()

Multivariate balancing:

- Proportional adjustment.
- Van der Ploeg.

The presentation of the functions is self-contained: e.g. help sw shows the detailed structure of the sw() function: purpose, input, output, library and technical references.

A Visual Basic interface has been developed to apply most of the library using Excel spreadsheets. Basically, the interface consists of two main modules: a program which generates and manages a sequence of contextual menus and a linkage function that activates the temporal disaggregation library according to the user's choices, as expressed by means of the corresponding forms. The interface is described in Abad, A. y Quilis, E.M. (2005) "Software to perform temporal disaggregation", Eurostat, Working Paper and Studies, ISSN 1725-4825, p. 2-8.

As in previous releases, the library uses some functions contained in the *Econometric Toolbox* of Professor James LeSage. I strongly recommend the use of this library in combination with his toolbox, in order to achieve important economies of scale in the art and science of quantitative modeling. More information may be obtained in his website:

http://www.spatial-econometrics.com/

UNIVARIATE BENCHMARKING

INFORMATION SET:

LOW-FREQUENCY TIME SERIES (Benchmark)

- BOOT-FEIBES-LISMAN
- STRAM-WEI
- LOW-PASS INTERPOLATION

BOOT-FEIBES-LISMAN: Univariate benchmarking and interpolation

```
function res = bfl(Y,ta,d,sc)
% PURPOSE: Temporal disaggregation using the Boot-Feibes-Lisman method
% SYNTAX: res = bfl(Y,ta,d,sc);
% ------
% OUTPUT: res: a structure
%
       res.meth = 'Boot-Feibes-Lisman'
%
       res.N = Number of low frequency data
%
       res.ta = Type of disaggregation
%
       res.d = Degree of differencing
%
       res.sc = Frequency conversion
%
       res.y = High frequency estimate
       res.et = Elapsed time
%
% INPUT: Y: Nx1 ---> vector of low frequency data
      ta: type of disaggregation
%
%
        ta=1 ---> sum (flow)
%
        ta=2 ---> average (index)
        ta=3 ---> last element (stock) ---> interpolation
%
%
         ta=4 ---> first element (stock) ---> interpolation
%
      d: objective function to be minimized: volatility of ...
%
         d=0 ---> levels
%
         d=1 ---> first differences
%
         d=2 ---> second differences
      sc: number of high frequency data points for each low frequency data point
%
%
        sc= 4 ---> annual to quarterly
%
        sc=12 ---> annual to monthly
        sc= 3 ---> quarterly to monthly
% ------
% LIBRARY: sw
% SEE ALSO: sw, tduni_print, tduni_plot
% ------
% REFERENCE: Boot, J.C.G., Feibes, W. and Lisman, J.H.C. (1967)
% "Further methods of derivation of quarterly figures from annual data",
% Applied Statistics, vol. 16, n. 1, p. 65-75.
```

STRAM-WEI: Univariate benchmarking and interpolation based on ARIMA modeling

```
function res = sw(Y,ta,d,sc,v)
% PURPOSE: Temporal disaggregation using the Stram-Wei method.
% -----
% SYNTAX: res = sw(Y,ta,d,sc,v);
% ------
% OUTPUT: res: a structure
       res.meth = 'Stram-Wei';
%
       res.N: = Number of low frequency data
%
       res.ta = Type of disaggregation
       res.d = Degree of differencing
%
%
       res.sc = Frequency conversion
       res.H = nxN temporal disaggregation matrix
%
%
       res.y = High frequency estimate
       res.et = Elapsed time
% ------
% INPUT: Y: Nx1 ---> vector of low frequency data
%
      ta: type of disaggregation
%
        ta=1 ---> sum (flow)
%
        ta=2 ---> average (index)
%
        ta=3 ---> last element (stock) ---> interpolation
%
        ta=4 ---> first element (stock) ---> interpolation
%
      d: number of unit roots
%
      sc: number of high frequency data points for each low frequency data point
%
        sc= 4 ---> annual to quarterly
%
        sc=12 ---> annual to monthly
%
        sc= 3 ---> quarterly to monthly
      v: (n-d)x(n-d) VCV matrix of high frequency stationary series
%
% LIBRARY: aggreg, aggreg_v, dif, movingsum
% SEE ALSO: bfl, tduni print, tduni plot
% REFERENCE: Stram, D.O. and Wei, W.W.S. (1986) "A methodological note on the
% disaggregation of time series totals", Journal of Time Series Analysis,
% vol. 7, n. 4, p. 293-302.
```

LOW-PASS INTERPOLATION: Univariate benchmarking and interpolation based on smoothing (via Hodrick-Prescott) and time-domain benchmarking (Denton)

```
function [y,w,x] = low_pass_interpolation(Y,ta,d,sc,lambda);
% PURPOSE: Low-pass interpolation using Hodrick-Prescott and Denton
% ------
% SYNTAX: [y,w,x] = low pass interpolation(Y,ta,d,sc,lambda);
% -----
% OUTPUT: y: nx1 ---> final interpolation
%
      w: nx1 ---> intermediate interpolation (low-pass filtering of x)
      x: nx1 ---> initial interpolation (padding Y with zeros)
% ------
% INPUT: Y: Nx1 ---> vector of low frequency data
      ta: 1x1 type of disaggregation
%
        ta=1 ---> sum (flow)
%
        ta=2 ---> average (index)
        ta=3 ---> last element (stock) ---> interpolation
%
%
        ta=4 ---> first element (stock) ---> interpolation
%
      d: 1x1 objective function to be minimized: volatility of ...
%
        d=0 ---> levels
        d=1 ---> first differences
%
%
        d=2 ---> second differences
%
      sc: 1x1 number of high frequency data points for each low frequency data
point
%
        sc= 4 ---> annual to quarterly
        sc=12 ---> annual to monthly
%
%
        sc= 3 ---> quarterly to monthly
%
      lambda: 1x1 --> balance between adjustment and smoothness (HP
      low-pass filter)
%
% LIBRARY: copylow, hp, denton_uni
% ------
% SEE ALSO: bfl, sw
% -----
```

UNIVARIATE BENCHMARKING WITH INDICATORS

INFORMATION SET:

LOW-FREQUENCY TIME SERIES (Benchmark)
HIGH-FREQUENCY TIME SERIES (Indicators)

- DENTON
- DENTON PROPORTIONAL
- CHOW-LIN by Maximum Likelihood and Weighted Least Squares
- FERNANDEZ
- LITTERMAN by Maximum Likelihood and Weighted Least Squares
- SANTOS SILVA-CARDOSO by Maximum Likelihood and Weighted Least Squares
- GUERRERO

DENTON_UNI: Univariate benchmarking and interpolation

```
function res = denton_uni(Y,x,ta,d,sc)
% PURPOSE: Temporal disaggregation using the Denton method
% SYNTAX: res = denton_uni(Y,x,ta,d,sc);
% ------
% OUTPUT: res: a structure
       res.meth = 'Denton';
%
             = Number of low frequency data
%
       res.N
%
       res.ta = Type of disaggregation
%
       res.sc = Frequency conversion
%
              = Degree of differencing
       res.d
%
       res.y = High frequency estimate
       res.x = High frequency indicator
%
%
       res.U = Low frequency residuals
%
       res.u = High frequency residuals
       res.et = Elapsed time
%
% -----
% INPUT: Y: Nx1 ---> vector of low frequency data
%
      x: nx1 ---> vector of low frequency data
%
      ta: type of disaggregation
%
         ta=1 ---> sum (flow)
%
         ta=2 ---> average (index)
         ta=3 ---> last element (stock) ---> interpolation
%
         ta=4 ---> first element (stock) ---> interpolation
%
      d: objective function to be minimized: volatility of ...
%
%
         d=0 ---> levels
         d=1 ---> first differences
%
%
         d=2 ---> second differences
      sc: number of high frequency data points for each low frequency data point
%
         sc= 4 ---> annual to quarterly
%
%
         sc=12 ---> annual to monthly
         sc= 3 ---> quarterly to monthly
   -----
% LIBRARY: aggreg, bfl
% SEE ALSO: denton uni prop, tduni plot, tduni print
% REFERENCE: Denton, F.T. (1971) "Adjustment of monthly or quarterly
% series to annual totals: an approach based on quadratic minimization",
% Journal of the American Statistical Society, vol. 66, n. 333, p. 99-102.
```

DENTON_UNI_PROP: Univariate benchmarking and interpolation, proportional variant

```
function res = denton uni prop(Y,x,ta,d,sc)
% PURPOSE: Temporal disaggregation: Denton method, proportional variant
% -----
% SYNTAX: res = denton uni prop(Y,x,ta,d,sc)
% ------
% OUTPUT: res: a structure
      res.meth = 'Proportional Denton';
%
%
      res.N = Number of low frequency data
%
      res.ta = Type of disaggregation
%
      res.sc = Frequency conversion
%
      res.d = Degree of differencing
      res.y = High frequency estimate
%
%
      res.x = High frequency indicator
%
      res.U = Low frequency residuals
      res.u
%
             = High frequency residuals
      res.et = Elapsed time
%
% -----
% INPUT: Y: Nx1 ---> vector of low frequency data
      x: nx1 ---> vector of low frequency data
%
%
      ta: type of disaggregation
%
        ta=1 ---> sum (flow)
%
        ta=2 ---> average (index)
        ta=3 ---> last element (stock) ---> interpolation
%
%
        ta=4 ---> first element (stock) ---> interpolation
%
      d: objective function to be minimized: volatility of ...
%
        d=0 ---> levels
%
        d=1 ---> first differences
%
        d=2 ---> second differences
%
      sc: number of high frequency data points for each low frequency data point
%
        sc= 4 ---> annual to quarterly
%
        sc=12 ---> annual to monthly
        sc= 3 ---> quarterly to monthly
% -----
% LIBRARY: aggreg
% ------
% SEE ALSO: denton_uni, tduni_plot, tduni_print
% ------
% REFERENCE: Denton, F.T. (1971) "Adjustment of monthly or quarterly
% series to annual totals: an approach based on quadratic minimization",
% Journal of the American Statistical Society, vol. 66, n. 333, p. 99-102.
```

CHOW-LIN: Univariate benchmarking and interpolation based on indicators. High-frequency model: linear model + AR(1) disturbances. Estimation is performed by maximum likelihood or weighted least squares.

```
function res = chowlin(Y,x,ta,sc,type,opC,rl)
% PURPOSE: Temporal disaggregation using the Chow-Lin method
% ------
% SYNTAX: res = chowlin(Y,x,ta,sc,type,opC,rl)
% -----
% OUTPUT: res: a structure
%
         res.meth ='Chow-Lin';
%
         res.ta = type of disaggregation
%
         res.type = method of estimation
%
        res.opC = option related to intercept
               = nobs. of low frequency data
%
         res.N
%
         res.n = nobs. of high-frequency data
%
         res.pred = number of extrapolations
         res.sc = frequency conversion between low and high freq.
%
%
                 = number of regressors (including intercept)
         res.p
%
         res.Y = low frequency data
         res.x
%
                 = high frequency indicators
%
         res.y = high frequency estimate
%
         res.y_dt = high frequency estimate: standard deviation
%
         res.y lo = high frequency estimate: sd - sigma
%
         res.y_up = high frequency estimate: sd + sigma
         res.u = high frequency residuals
%
%
                = low frequency residuals
         res.U
%
         res.beta = estimated model parameters
%
         res.beta sd = estimated model parameters: standard deviation
%
         res.beta t = estimated model parameters: t ratios
%
         res.rho = innovational parameter
         res.aic = Information criterion: AIC
%
%
         res.bic = Information criterion: BIC
%
         res.val = Objective function used by the estimation method
%
         res.wls = Weighted least squares as a function of rho
%
         res.loglik = Log likelihood as a function of rho
%
         res.r = grid of innovational parameters used by the estimation method
% INPUT: Y: Nx1 ---> vector of low frequency data
%
       x: nxp ---> matrix of high frequency indicators (without intercept)
%
       ta: type of disaggregation
%
         ta=1 ---> sum (flow)
%
         ta=2 ---> average (index)
%
         ta=3 ---> last element (stock) ---> interpolation
         ta=4 ---> first element (stock) ---> interpolation
%
%
       sc: number of high frequency data points for each low frequency data points
%
         sc= 4 ---> annual to quarterly
         sc=12 ---> annual to monthly
%
%
         sc= 3 ---> quarterly to monthly
```

```
%
      type: estimation method:
%
         type=0 ---> weighted least squares
%
         type=1 ---> maximum likelihood
%
       opC: 1x1 option related to intercept
%
         opc = -1: pretest intercept significance
%
         opc = 0: no intercept in hf model
%
         opc = 1: intercept in hf model
%
      rl: innovational parameter (Optional)
         rl: 0x0 ---> rl=[0.5 \ 0.06 \ 0.99] (e.g. rl=[])
%
%
         rl: 1x1 ---> fixed value of rho parameter
%
         rl: 1x3 ---> [r_min: dr : r_max] search is performed
           on this range, using dr as step.
%
% LIBRARY: chowlin_W
% ------
% SEE ALSO: litterman, fernandez, td_plot, td_print, chowlin_co
% -----
% REFERENCE: Chow, G. and Lin, A.L. (1971) "Best linear unbiased
% distribution and extrapolation of economic time series by related
% series", Review of Economic and Statistics, vol. 53, n. 4, p. 372-375.
% Bournay, J. and Laroque, G. (1979) "Reflexions sur la methode
% d'elaboration des comptes trimestriels", Annales de l'INSEE, n. 36, p. 3-30.
```

FERNANDEZ: Univariate benchmarking and interpolation based on indicators. High-frequency model: linear model + I(1) disturbances (random walk). Estimation is performed by generalized least squares.

```
function res = fernandez(Y,x,ta,sc,opC)
% PURPOSE: Temporal disaggregation using the Fernandez method
% ------
% SYNTAX: res = fernandez(Y,x,ta,sc,opC);
% ------
% OUTPUT: res: a structure
%
        res.meth ='Fernandez';
%
        res.ta = type of disaggregation
%
        res.type = method of estimation
               = nobs. of low frequency data
%
        res.N
%
        res.n = nobs. of high-frequency data
%
        res.pred = number of extrapolations
        res.sc = frequency conversion between low and high freq.
%
%
                 = number of regressors (including intercept)
        res.p
%
        res.Y = low frequency data
        res.x
%
                 = high frequency indicators
        res.y = high frequency estimate
%
%
        res.y_dt = high frequency estimate: standard deviation
%
        res.y lo = high frequency estimate: sd - sigma
%
        res.y_up = high frequency estimate: sd + sigma
        res.u = high frequency residuals
%
%
        res.U = low frequency residuals
%
        res.beta = estimated model parameters
%
        res.beta sd = estimated model parameters: standard deviation
%
        res.beta t = estimated model parameters: t ratios
        res.aic = Information criterion: AIC
%
%
        res.bic
                 = Information criterion: BIC
% INPUT: Y: Nx1 ---> vector of low frequency data
%
      x: nxp ---> matrix of high frequency indicators (without intercept)
%
       ta: type of disaggregation
%
         ta=1 ---> sum (flow)
%
         ta=2 ---> average (index)
         ta=3 ---> last element (stock) ---> interpolation
%
         ta=4 ---> first element (stock) ---> interpolation
%
       sc: number of high frequency data points for each low frequency data points
%
%
         sc= 4 ---> annual to quarterly
%
         sc=12 ---> annual to monthly
%
         sc= 3 ---> quarterly to monthly
       opC: 1x1 option related to intercept
%
%
         opc = -1: pretest intercept significance
%
         opc = 0: no intercept in hf model
         opc = 1: intercept in hf model
%
```

٥/،	
%	LIBRARY: fernandez_W
%	SEE ALSO: chowlin, litterman, td_plot, td_print
, 0	REFERENCE: Fernandez, R.B.(1981)"Methodological note on the
	estimation of time series", Review of Economic and Statistics, vol. 63, n. 3, p. 471-478.

LITTERMAN: Univariate benchmarking and interpolation based on indicators. High-frequency model: linear model + ARI(1,1) disturbances (markovian random walk). Estimation is performed by maximum likelihood or weighted least squares.

```
function res = litterman(Y,x,ta,sc,type,opC,rl)
% PURPOSE: Temporal disaggregation using the Litterman method
% SYNTAX: res = litterman(Y,x,ta,sc,type,opC,rl)
% -----
% OUTPUT: res: a structure
%
         res.meth ='Litterman';
%
         res.ta = type of disaggregation
         res.type = method of estimation
%
%
         res.opC = option related to intercept
         res.N = nobs. of low frequency data
res.n = nobs. of high-frequency data
%
%
%
         res.pred = number of extrapolations
%
         res.sc = frequency conversion between low and high freq.
         res.p
%
                  = number of regressors (including intercept)
%
         res.Y = low frequency data
         res.x = high frequency indicators
res.y = high frequency estimate
%
%
%
         res.y_dt = high frequency estimate: standard deviation
         res.y_lo = high frequency estimate: sd - sigma
%
%
         res.y_up = high frequency estimate: sd + sigma
         res.u = high frequency residuals
res.U = low frequency residuals
%
%
                   = low frequency residuals
         res.U
%
         res.beta = estimated model parameters
%
         res.beta sd = estimated model parameters: standard deviation
%
         res.beta_t = estimated model parameters: t ratios
%
         res.rho = innovational parameter
%
         res.aic = Information criterion: AIC
%
         res.bic = Information criterion: BIC
%
         res.val = Objective function used by the estimation method
%
         res.wls = Weighted least squares as a function of rho
%
         res.loglik = Log likelihood as a function of rho
         res.r = grid of innovational parameters used by the estimation method
% -----
% INPUT: Y: Nx1 ---> vector of low frequency data
%
       x: nxp ---> matrix of high frequency indicators (without intercept)
%
       ta: type of disaggregation
%
          ta=1 ---> sum (flow)
%
          ta=2 ---> average (index)
%
          ta=3 ---> last element (stock) ---> interpolation
%
          ta=4 ---> first element (stock) ---> interpolation
       sc: number of high frequency data points for each low frequency data points
%
%
         sc= 4 ---> annual to quarterly
%
          sc=12 ---> annual to monthly
%
          sc= 3 ---> quarterly to monthly
```

```
%
      type: estimation method:
%
         type=0 ---> weighted least squares
%
         type=1 ---> maximum likelihood
%
      opC: 1x1 option related to intercept
%
         opc = -1: pretest intercept significance
%
         opc = 0 : no intercept in hf model
%
         opc = 1: intercept in hf model
%
      rl: innovationl parameter
         rl: 0x0 ---> rl=[0.5 \ 0.06 \ 0.99] (e.g. rl=[])
%
%
         rl: 1x1 ---> fixed value of rho parameter
%
         rl: 1x3 ---> [r_min: dr : r_max] search is performed
           on this range, using dr as step.
%
% LIBRARY: litterman_W
% ------
% SEE ALSO: chowlin, fernandez, td_plot, td_print
% -----
% REFERENCE: Litterman, R.B. (1983a) "A random walk, Markov model
% for the distribution of time series", Journal of Business and
% Economic Statistics, vol. 1, n. 2, p. 169-173.
```

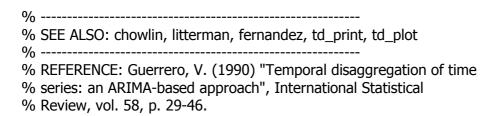
SSC: Univariate benchmarking and interpolation based on indicators. High-frequency model: dynamic linear model. Estimation is performed by maximum likelihood or weighted least squares.

```
function res = ssc(Y,x,ta,sc,type,opC,rl)
% PURPOSE: Temporal disaggregation using the dynamic Chow-Lin method
        proposed by Santos Silva-Cardoso (2001).
% SYNTAX: res = ssc(Y,x,ta,sc,type,opC,rl);
% -----
% OUTPUT: res: a structure
%
         res.meth ='Santos Silva-Cardoso';
%
         res.ta = type of disaggregation
%
         res.type = method of estimation
%
         res.opC = option related to intercept
%
                  = nobs. of low frequency data
         res.N
%
         res.n
                  = nobs. of high-frequency data
         res.pred = number of extrapolations
%
%
         res.sc = frequency conversion between low and high freq.
%
                  = number of regressors (including intercept)
         res.p
%
         res.Y
                  = low frequency data
                  = high frequency indicators
%
         res.x
%
                  = high frequency estimate
         res.y
%
         res.y dt = high frequency estimate: standard deviation
%
         res.y lo = high frequency estimate: sd - sigma
%
         res.y_up = high frequency estimate: sd + sigma
%
                  = high frequency residuals
         res.u
%
         res.U
                  = low frequency residuals
         res.gamma = estimated model parameters (including y(0))
%
%
         res.gamma sd = estimated model parameters: standard deviation
%
         res.gamma t = estimated model parameters: t ratios
%
         res.rho
                   = dynamic parameter phi
%
                    = estimated model parameters (excluding y(0))
         res.beta
         res.beta sd = estimated model parameters: standard deviation
%
%
         res.beta t = estimated model parameters: t ratios
%
                 = innovational parameter
         res.phi
                  = Information criterion: AIC
%
         res.aic
%
         res.bic
                  = Information criterion: BIC
                = Objective function used by the estimation method
%
         res.val
         res.wls = Weighted least squares as a function of phi
%
%
         res.loglik = Log likelihood as a function of phi
                 = grid of innovational parameters used by the estimation method
%
         res.r
%
                   = elapsed time
         res.et
% INPUT: Y: Nx1 ---> vector of low frequency data
%
       x: nxp ---> matrix of high frequency indicators (without intercept)
%
       ta: type of disaggregation
%
          ta=1 ---> sum (flow)
%
         ta=2 ---> average (index)
         ta=3 ---> last element (stock) ---> interpolation
%
%
         ta=4 ---> first element (stock) ---> interpolation
```

```
%
       sc: number of high frequency data points for each low frequency data points
%
          sc= 4 ---> annual to quarterly
%
          sc=12 ---> annual to monthly
%
          sc= 3 ---> quarterly to monthly
%
       type: estimation method:
%
          type=0 ---> weighted least squares
%
          type=1 ---> maximum likelihood
%
       opC: 1x1 option related to intercept
%
          opc = -1: pretest intercept significance
%
          opc = 0: no intercept in hf model
%
          opc = 1: intercept in hf model
%
       rl: innovational parameter
%
          rl: 0x0 ---> rl=[0.5 \ 0.06 \ 0.99] (e.g. rl=[])
%
          rl: 1x1 ---> fixed value of phi parameter
         rl: 1x3 ---> [r_min: dr : r_max] search is performed
%
%
            on this range, using dr as step.
% ------
% LIBRARY: ssc W
% SEE ALSO: chowlin, litterman, fernandez, td_plot, td_print
% REFERENCE: Santos, J.M.C. and Cardoso, F.(2001) "The Chow-Lin method
% using dynamic models", Economic Modelling, vol. 18, p. 269-280.
% Di Fonzo, T. (2002) "Temporal disaggregation of economic time series:
% towards a dynamic extension", Dipartimento di Scienze Statistiche,
% Universita di Padova, Working Paper n. 2002-17.
```

GUERRERO: Univariate benchmarking and interpolation based on indicators. High-frequency model: ARIMA-based benchmarking.

```
function res = querrero(Y,x,ta,sc,rexw,rexd,opC)
% PURPOSE: ARIMA-based temporal disaggregation: Guerrero method
% SYNTAX: res = guerrero(Y,x,ta,sc,rexw,rexd,opC);
% -----
% OUTPUT: res: a structure
%
       res.meth ='Guerrero';
%
       res.ta = type of disaggregation
       res.opC = option related to intercept
%
%
                 = nobs. of low frequency data
       res.N
%
                = nobs. of high-frequency data
       res.n
%
       res.pred = number of extrapolations
%
                = frequency conversion between low and high freq.
       res.sc
%
                 = number of regressors (+ intercept)
       res.p
%
                 = low frequency data
       res.Y
%
                 = high frequency indicators
       res.x
%
                 = scaled indicator (preliminary hf estimate)
       res.w
%
       res.y1
                 = first stage high frequency estimate
%
                 = final high frequency estimate
       res.y
%
       res.y_dt = high frequency estimate: standard deviation
%
       res.y lo = high frequency estimate: sd - sigma
%
       res.y up = high frequency estimate: sd + sigma
%
       res.delta = high frequency discrepancy (y1-w)
%
             = high frequency residuals (y-w)
       res.u
%
       res.U = low frequency residuals (Cu)
%
       res.beta = estimated parameters for scaling x
%
       res.k = statistic to test compatibility
%
       res.et = elapsed time
% -----
% INPUT: Y: Nx1 ---> vector of low frequency data
%
       x: nxp ---> matrix of high frequency indicators (without intercept)
%
       ta: type of disaggregation
%
         ta=1 ---> sum (flow)
         ta=2 ---> average (index)
%
%
         ta=3 ---> last element (stock) ---> interpolation
%
         ta=4 ---> first element (stock) ---> interpolation
%
       sc: number of high frequency data points for each low frequency data points
%
         sc= 4 ---> annual to quarterly
%
         sc=12 ---> annual to monthly
%
         sc= 3 ---> quarterly to monthly
%
       rexw, rexd ---> a structure containing the parameters of ARIMA model
         for indicator and discrepancy, respectively (see calT function)
%
%
       opC: 1x1 option related to intercept
%
         opc = -1: pretest intercept significance
%
         opc = 0: no intercept in hf model
         opc = 1: intercept in hf model
% LIBRARY: guerrero_W
```



MULTIVARIATE BENCHMARKING WITH INDICATORS AND TRANSVERSAL CONSTRAINTS

INFORMATION SET:

LOW-FREQUENCY VECTOR TIME SERIES (Benchmarks)
HIGH-FREQUENCY VECTOR TIME SERIES (Indicators)
HIGH-FREQUENCY TIME SERIES (Transversal constraint)

- DENTON: additive and proportional variants.
- ROSSI
- DI FONZO

DENTON: Multivariate benchmarking and interpolation, with transversal constraint. Additive variant.

```
function res = denton(Y,x,z,ta,sc,d)
% PURPOSE: Multivariate temporal disaggregation with transversal constraint
0/_ -----
% SYNTAX: res = denton(Y,x,z,ta,sc,d);
% ------
% OUTPUT: res: a structure
       res.meth = 'Multivariate Denton';
%
%
       res.N = Number of low frequency data
%
       res.n = Number of high frequency data
%
       res.pred = Number of extrapolations (=0 in this case)
%
       res.ta = Type of disaggregation
       res.sc = Frequency conversion
%
%
       res.d = Degree of differencing
%
       res.y = High frequency estimate
%
       res.z
             = High frequency constraint
       res.et = Elapsed time
%
% -----
% INPUT: Y: NxM ---> M series of low frequency data with N observations
      x: nxM ---> M series of high frequency data with n observations
%
      z: nx1 ---> high frequency transversal constraint
%
      ta: type of disaggregation
%
        ta=1 ---> sum (flow)
         ta=2 ---> average (index)
%
%
        ta=3 ---> last element (stock) ---> interpolation
%
         ta=4 ---> first element (stock) ---> interpolation
%
      sc: number of high frequency data points for each low frequency data points
%
         sc= 4 ---> annual to quarterly
%
         sc=12 ---> annual to monthly
%
         sc= 3 ---> quarterly to monthly
%
      d: objective function to be minimized: volatility of ...
%
        d=0 ---> levels
%
        d=1 ---> first differences
        d=2 ---> second differences
% LIBRARY: aggreg, aggreg v, dif, vec, desvec
% ------
% SEE ALSO: denton_prop, difonzo, mtd_print, mtd_plot
% -----
% REFERENCE: Di Fonzo, T. (1994) "Temporal disaggregation of a system of
% time series when the aggregate is known: optimal vs. adjustment methods",
% INSEE-Eurostat Workshop on Quarterly National Accounts, Paris, December
```

DENTON_PROP: Multivariate benchmarking and interpolation, with transversal constraint. Proportional variant.

```
function res = denton prop(Y,x,z,ta,sc,d)
% PURPOSE: Multivariate temporal disaggregation, with transversal constraint.
       Proportional variant.
% ------
% SYNTAX: res = denton prop(Y,x,z,ta,sc,d)
% -----
% OUTPUT: res: a structure
%
       res.meth = 'Proportional Multivariate Denton';
%
       res.N
            = Number of low frequency data
%
              = Number of high frequency data
       res.n
%
       res.pred = Number of extrapolations (=0 in this case)
       res.ta = Type of disaggregation
%
%
       res.sc = Frequency conversion
       res.d = Degree of differencing
%
%
       res.y
             = High frequency estimate
%
       res.z = High frequency constraint
%
       res.et = Elapsed time
      ·
% INPUT: Y: NxM ---> M series of low frequency data with N observations
      x: nxM ---> M series of high frequency data with n observations
%
      z: nx1 ---> high frequency transversal constraint
%
      ta: type of disaggregation
        ta=1 ---> sum (flow)
%
%
        ta=2 ---> average (index)
%
        ta=3 ---> last element (stock) ---> interpolation
        ta=4 ---> first element (stock) ---> interpolation
%
      sc: number of high frequency data points for each low frequency data points
%
%
        sc= 4 ---> annual to quarterly
        sc=12 ---> annual to monthly
%
%
        sc= 3 ---> quarterly to monthly
%
      d: objective function to be minimized: volatility of ...
%
        d=0 ---> levels
%
        d=1 ---> first differences
        d=2 ---> second differences
% -----
% LIBRARY: aggreg, aggreg_v, dif, vec, desvec
% ------
% SEE ALSO: denton, difonzo, mtd print, mtd plot
% REFERENCE: Di Fonzo, T. (1994) "Temporal disaggregation of a system of
% time series when the aggregate is known: optimal vs. adjustment methods",
% INSEE-Eurostat Workshop on Quarterly National Accounts, Paris, december
```

ROSSI: Multivariate benchmarking and interpolation with indicators, with transversal constraint.

```
function res = rossi(Y,x,z,ta,sc,opMethod,type)
% PURPOSE: Multivariate temporal disaggregation with transversal constraint
% SYNTAX: res = rossi(Y,x_ini,z,ta,sc,opMethod);
% ------
% OUTPUT: res: a structure
       res.meth = 'Multivariate Rossi';
%
%
       res.N = Number of low frequency data
%
       res.n = Number of high frequency data
%
       res.pred = Number of extrapolations (=0 in this case)
%
       res.ta = Type of disaggregation
       res.sc = Frequency conversion
%
%
       res.y = High frequency estimate
%
       res.z = High frequency constraint
       res.et = Elapsed time
% INPUT: Y: NxM ---> M series of low frequency data with N observations
%
       x:
            nxM ---> M series of high frequency data with n observations
            nx1 ---> high frequency transversal constraint
%
%
       ta: type of disaggregation
%
         ta=1 ---> sum (flow)
%
         ta=2 ---> average (index)
          ta=3 ---> last element (stock) ---> interpolation
%
%
         ta=4 ---> first element (stock) ---> interpolation
%
       sc: number of high frequency data points for each low frequency data points
%
          sc= 4 ---> annual to quarterly
%
          sc=12 ---> annual to monthly
%
          sc= 3 ---> quarterly to monthly
       opMethod: univariate temporal disaggregation procedure used to compute
%
%
       preliminary estimates
%
          opMethod = 1 -> Fernandez
%
          opMethod = 2 \rightarrow Chow-Lin (optimized for rl=[], see chowlin)
          opMethod = 3 -> Litterman (optimized for rl=[], see litterman)
%
%
          Pretesting for intercept is made: opC = -1
%
       type: estimation method:
%
         type=0 ---> weighted least squares
          type=1 ---> maximum likelihood
% LIBRARY: aggreg, vec, desvec, fernandez, chowlin, litterman
% SEE ALSO: denton, difonzo, mtd_print, mtd_plot
% REFERENCE: Rossi, N. (1982)"A note on the estimation of disaggregate
% time series when the aggregate is known", Review of Economics and Statistics,
% vol. 64, n. 4, p. 695-696.
% Di Fonzo, T. (1994) "Temporal disaggregation of a system of
% time series when the aggregate is known: optimal vs. adjustment methods",
% INSEE-Eurostat Workshop on Quarterly National Accounts, Paris, december.
```

DIFONZO: Multivariate benchmarking and interpolation with indicators, with transversal constraint.

```
function res = difonzo(Y,x,z,ta,sc,type,f)
% PURPOSE: Multivariate temporal disaggregation with transversal constraint
% SYNTAX: res = difonzo(Y,x,z,ta,sc,type,f);
% ------
% OUTPUT: res: a structure
        res.meth = 'Multivariate Di Fonzo';
%
%
              = Number of low frequency data
%
        res.n = Number of high frequency data
        res.pred = Number of extrapolations
%
%
        res.ta = Type of disaggregation
        res.sc = Frequency conversion
%
%
        res.type = Model for high frequency innovations
%
        res.beta = Model parameters
%
              = High frequency estimate
        res.y
%
        res.d_y = High frequency estimate: std. deviation
%
        res.z = High frequency constraint
%
        res.et = Elapsed time
% --
% INPUT: Y: NxM ---> M series of low frequency data with N observations
       x: nxm ---> m series of high frequency data with n observations, m>=M see
%
(*)
%
       z: nzx1 ---> high frequency transversal constraint with nz obs.
%
       ta: type of disaggregation
%
          ta=1 ---> sum (flow)
%
          ta=2 ---> average (index)
          ta=3 ---> last element (stock) ---> interpolation
%
          ta=4 ---> first element (stock) ---> interpolation
%
       sc: number of high frequency data points for each low frequency data points
%
%
          sc= 4 ---> annual to quarterly
%
          sc=12 ---> annual to monthly
%
          sc= 3 ---> quarterly to monthly
%
       type: model for the high frequency innvations
%
          type=0 ---> multivariate white noise
%
          type=1 ---> multivariate random walk
% (*) Optional:
       f: 1xM ---> Set the number of high frequency indicators linked to
%
               each low frequency variable. If f is explicitly included,
%
%
               the high frequency indicators should be placed in
%
               consecutive columns
```

%	
% NOTE: Extrapolation is a	automatically performed when n>sN.
% If n=nz>sN restricted	ed extrapolation is applied.
% Finally, if n>nz>sN	extrapolation is perfomed in constrained
% form in the first nz-	sN observatons and in free form in
% the last n-nz observ	ations.
%	
% LIBRARY: aggreg, dif, v	
%	
% SEE ALSO: denton, mtd	
%	
•	T.(1990)"The estimation of M disaggregate time
% series when contempora	aneous and temporal aggregates are known", Review
% of Economics and Statis	tics, vol. 72, n. 1, p. 178-182.

MULTIVARIATE BALANCING

INFORMATION SET:

INITIAL ESTIMATES

- PROPORTIONAL BALANCING
- VAN DER PLOEG

BAL: Transversal balancing by proportional adjustment

```
function yb = bal(y,z)
% PURPOSE: Proportional adjustment of y to a given total z
% ------
% SYNTAX: yb = bal(y,z);
% ------
% OUTPUT: yb : nxM --> balanced series
% -------
% INPUT: y: nxM --> unbalanced series
% z: nx1 --> transversal constraint
% --------
% LIBRARY: vec, desvec
% ------
% SEE ALSO: denton
% -------
% REFERENCE: di Fonzo, T. (1994) "Temporal disaggregation of a system of
% time series when the aggregate is known: optimal vs. adjustment methods",
% INSEE-Eurostat Workshop on Quarterly National Accounts, Paris, december
```

van der PLOEG: Transversal balancing by means of QL optimization

```
function res = vdp(y,S,A,a)
% PURPOSE: Balancing by means of QL optimization (LS estimation)
% SYNTAX: res = vdp(y,S,A,a);
% ------
% OUTPUT: res: a structure with ...
%
           : kx1 vector of balanced variables
             : kxk VCV of final (balanced) estimates
%
%
        lambda: mx1 Lagrange multipliers
% INPUT: y : kx1 vector of unbalanced variables (initial estimates)
%
     S : kxk VCV of initial estimates
%
        A : kxm matrix of linear constraints
            : 1xm vector of autonomous terms related to linear constraints
% Note: a is optional. If it is not explicitly included, the function assumes a=0
% ------
% LIBRARY:
% ------
% REFERENCE: Van der Ploeg, F.(1982)"Reliability and the adjustment
% of sequences of large economic accounting matrices", Journal of
% the Royal Statistical Society, series A, vol. 145, n. 2, p. 169-194.
```

UTILITIES

- **Backtesting**

- Systematic sampler
 Temporal aggregation
 Temporal accumulation
- Moving sum (average)
 Cast low-frequency data into high-frequency format

Backtest: recursive estimates and revisions for some procedures

```
function res = backtest(rex,Nh);
% PURPOSE: Backtesting of temporal disaggregation.
% SYNTAX: res = backtest(rex,Nh);
% ------
% OUTPUT: res: a structure with...
%
       rex: reference structure
       yh : recursive estimations from FROM T=N+Nh T=N
%
       yrev: revision as %, compared with final (full info) estimation
% INPUT: rex: a structure generated by chowlin(), fernandez(), litterman()
%
%
      Nh: 1x1 --> number of low-freq. to compute revisions
% LIBRARY: chowlin, fernandez, litterman, ssc % -------
% NOTE: Revision of estimates when adding low-frequency data from T=N-Nh to
% T=N
```

Systematic sampler

```
function [zs] = ssampler(z,op1,sc)
% PURPOSE: Systematic sampling of a high-frequency time series
% ------
% SYNTAX: zs = ssampler(z,op1,sc)
% ------
% OUTPUT: zs: nx1 sampled time series
% -----
% INPUT: z: nx1 ---> vector of high frequency data
%
      op1: type of temporal aggregation
%
      op1=1 ---> sum (flow)
%
      op1=2 ---> average (index)
      op1=3 ---> last element (stock) ---> interpolation
%
%
      op1=4 ---> first element (stock) ---> interpolation
%
      sc: number of high frequency data points
       for each low frequency data points
% -----
% LIBRARY: copylow, temporal_agg
```

Temporal aggregation

```
function [y] = temporal_agg(z,op1,sc)
% PURPOSE: Temporal aggregation of a time series
% ------
% SYNTAX: y = temporal_agg(z,op1,sc);
% ------
% OUTPUT: y: Nx1 temporally aggregated series
% -----
% INPUT: z: nx1 ---> vector of high frequency data
%
      op1: type of temporal aggregation
%
      op1=1 ---> sum (flow)
%
      op1=2 ---> average (index)
      op1=3 ---> last element (stock) ---> interpolation
%
%
      op1=4 ---> first element (stock) ---> interpolation
%
      sc: number of high frequency data points
%
        for each low frequency data points
      Note: n = sc \times N
% LIBRARY: aggreg
```

Temporal accumulation

```
function [za] = temporal_acc(z,op1,sc);
% PURPOSE: Accumulate within a period of sc observations
% ------
% SYNTAX: za = temporal_acc(z,op1,sc);
% ------
% OUTPUT: za: nx1 accumulated vector
% ------
% INPUT: op1: type of temporal aggregation
%
     op1=1 ---> sum (flow)
%
     op1=2 ---> average (index)
%
     sc: number of high frequency data points
%
      for each low frequency data points (freq. conversion)
% ------
% LIBRARY: acc
% SEE ALSO: temporal_agg
% ------
```

Moving sum or average

```
function [zs] = moving_acc(z,op1,sc);
% PURPOSE: Moving sum (average) with accumulation=sc
% SYNTAX: zs = moving_acc(z,op1,sc);
% ------
% OUTPUT: zs: nx1 moving sum (average) vector with (sc-1) initial zeros
% -----
% INPUT: op1: type of temporal aggregation
%
      op1=1 ---> sum (flow)
%
      op1=2 ---> average (index)
%
      sc: number of high frequency data points
%
       for each low frequency data points (freq. conversion) = size
%
       of sum filter
% LIBRARY: movingsum
% ------
% SEE ALSO: temporal_acc, temporal_agg
% ------
```

Cast low-frequency data into high-frequency format

```
function [z] = copylow(Z,op1,sc)
% PURPOSE: Generates a high-frequency time series from a low-frequency one
% SYNTAX: z = copylow(Z,op1,sc);
% ------
% OUTPUT: z : nxk high frequency time series
% -----
% INPUT: Z : an Nxk matrix of low frequency series, columnwise
     op1 : type of temporal aggregation
%
     op1=1 ---> copy sc times the If data
%
%
     op1=2 ---> copy sc times the mean If data
%
     op1=3 ---> last element (stock) ---> interpolation
%
     op1=4 ---> first element (stock) ---> interpolation
%
     sc: number of high frequency data points
     for each low frequency data points (quarterly: sc=4, monthly: sc=12)
% ------
% LIBRARY:
% ------
% SEE ALSO: temporal_agg
```

EXTENSIONS

- CHOW-LIN by Cochrane-Orcutt Extended interpolation

CHOW-LIN _CO: Univariate benchmarking and interpolation based on indicators. High-frequency model: linear model + AR(1) disturbances. Estimation is performed by preliminary Cochrane-Orcutt estimation of the implied low-frequency model.

```
function res = chowlin co(Y,x,ta,sc,opC)
% PURPOSE: Temporal disaggregation using the Chow-Lin method
       (quarterly rho derived from Cochrane-Orcutt annual rho)
% ------
% SYNTAX: res = chowlin_co(Y,x,ta,sc,opC);
% -----
% OUTPUT: res: a structure
%
         res.meth ='Chow-Lin';
%
         res.ta = type of disaggregation
         res.type = method of estimation
%
         res.N = nobs. of low frequency data
res.n = nobs. of high-frequency data
%
%
%
         res.pred = number of extrapolations
         res.sc = frequency conversion between low and high freq.
%
         res.p = number of regressores.Y = low frequency data
                  = number of regressors (including intercept)
%
%
%
         res.x = high frequency indicators
%
         res.y
                  = high frequency estimate
%
         res.y_dt = high frequency estimate: standard deviation
%
         res.y_lo = high frequency estimate: sd - sigma
%
         res.y_up = high frequency estimate: sd + sigma
         res.u = high frequency residuals
%
%
                  = low frequency residuals
         res.U
%
         res.beta = estimated model parameters
%
         res.beta_sd = estimated model parameters: standard deviation
%
         res.beta_t = estimated model parameters: t ratios
%
         res.rho = innovational parameter
%
         res.aic = Information criterion: AIC
%
         res.bic = Information criterion: BIC
%
                  = Cochrane-Orcutt regression (see LeSage
         res.co
%
         Econometric Toolbox)
%
% INPUT: Y: Nx1 ---> vector of low frequency data
%
       x: nxp ---> matrix of high frequency indicators (without intercept)
%
       ta: type of disaggregation
%
          ta=1 ---> sum (flow)
%
          ta=2 ---> average (index)
       sc: number of high frequency data points for each low frequency data points
%
%
          sc= 3 ---> quarterly to monthly
%
          sc= 4 ---> annual to quarterly
%
       opC: 1x1 option related to intercept
%
         opc = -1: pretest intercept significance
%
          opc = 0: no intercept in hf model
%
         opc = 1: intercept in hf model
```

%	
	LIBRARY: chowlin_co
%	SEE ALSO: chowlin, litterman, fernandez, td_plot, td_print
%	REFERENCE: Chow, G. and Lin, A.L. (1971) "Best linear unbiased distribution and extrapolation of economic time series by related
	series", Review of Economic and Statistics, vol. 53, n. 4, p. 372-375.
	Bournay, J. y Laroque, G. (1979) "Reflexions sur la methode d'elaboration des comptes trimestriels". Annales de l'INSEE, n. 36, p. 3-30.

Extended interpolation: generation of basic vector

```
function [c] = aggreg_v_X(op1,sc)
% PURPOSE: Generate a temporal aggregation vector (Extended version)
% SYNTAX: c=aggreg_v_X(op1,sc);
% ------
% OUTPUT: c: 1 x sc temporal aggregation vector
% -----
% INPUT: op1: type of temporal aggregation
      op1 = -1 ---> average (index)
%
      op1 = 0 ---> sum (flow)
%
%
      op1 = h ---> interpolates at h element
%
         e.g. h=sc is stock at end of low-freq. period
%
      sc: number of high frequency data points
%
       for each low frequency data points (freq. conversion)
% ------
% LIBRARY:
% ------
% SEE ALSO: aggreg_v
```