

A MATLAB LIBRARY OF TEMPORAL DISAGGREGATION METHODS: SUMMARY

Enrique M. Quilis¹

Instituto Nacional de Estadística
Paseo de la Castellana, 183
28046 - Madrid (SPAIN)

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1. INTRODUCTION

This release of the Matlab temporal disaggregation library includes some new features:

- stock first as a temporal disaggregation case (interpolation)
- new graphs for univariate Denton
- univariate Denton, proportional variant
- univariate temporal disaggregation by means of an ARIMA model-based procedure due to Guerrero (1990)
- multivariate temporal disaggregation by means of a two-step method due to Rossi (1982). The first step requires a preliminary univariate disaggregation that may be performed by Fernández, Chow-Lin or Litterman.

The library includes a set of function to perform temporal disaggregation (distribution, averaging and interpolation), according to the following structure:

Adjustment or quadratic programming methods:

- bfl (Boot-Feibes-Lisman)
- denton_uni, denton_uni_prop
- sw (Stram-Wei method)

served by: tduni_print (ASCII output), tduni_plot (graphic output)

Model-based (or BLUE) methods:

- chowlin
- fernandez
- litterman
- ssc (Santos Silva-Cardoso method: a dynamic version of Chow-Lin)

served by: td_print (ASCII output), td_plot (graphic output)

- guerrero

served by: td_print_G (ASCII output), td_plot (graphic output)

Multivariate methods that include a transversal restriction:

- rossi
- denton
- difonzo

served by: mtd_print (ASCII output), mtd_plot (graphic output)

Extrapolation is feasible using chowlin, fernandez, litterman, ssc and difonzo. Constrained extrapolation can be performed also by means of difonzo.

The presentation of the functions is self-contained: help text, script to run the function and output (ASCII file and plots).

This library is rather specific. Combining it with the *Econometrics Toolbox* of Professor James LeSage is a sensible decision. In fact, some procedures require to have access to it, although this dependence may be circumvented by appropriate code modification. For more information, consult his Internet site:

<http://jpl.econ.utoledo.edu/faculty/lesage>

2. BOOT-FEIBES-LISMAN

PURPOSE: Temporal disaggregation using the Boot-Feibes-Lisman method

SYNTAX: res=bfl(Y,ta,d,s);

OUTPUT: res: a structure
res.meth = 'Boot-Feibes-Lisman';
res.N = Number of low frequency data
res.ta = Type of disaggregation
res.s = Frequency conversion
res.d = Degree of differencing
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
s: number of high frequency data points for each low frequency data point
s= 4 ---> annual to quarterly
s=12 ---> annual to monthly
s= 3 ---> quarterly to monthly

LIBRARY: sw

SEE ALSO: tduni_print, tduni_plot

REFERENCE: Boot, J.C.G., Feibes, W. y Lisman, J.H.C. (1967)
"Further methods of derivation of quarterly figures from annual data",
Applied Statistics, vol. 16, n. 1, p. 65-75.

Application:

```
Y=load('c:\x\td\data\Y.anu');  
res=bfl(Y,1,1,12);  
tduni_print(res,'td.sal');  
tduni_plot(res);  
edit td.sal
```

ASCII file containing detailed output:

```
*****  
TEMPORAL DISAGGREGATION METHOD: Boot-Feibes-Lisman  
*****
```

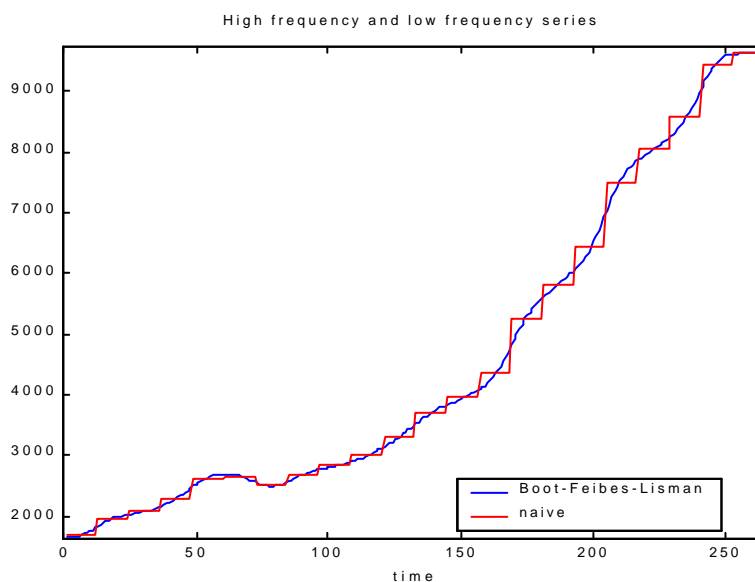
```
-----  
Number of low-frequency observations : 22  
Frequency conversion : 12  
Number of high-frequency observations : 264  
-----
```

```
Degree of differencing : 1  
Type of disaggregation: sum (flow).  
-----
```

High frequency series (columnwise):

```
-----  
4972. 2800  
4971. 1389  
.....  
.....  
.....  
7898. 7692  
7899. 3631  
7899. 6600  
-----
```

```
Elapsed time: 0.3200  
-----
```



3. STRAM-WEI

PURPOSE: Temporal disaggregation using the Stram-Wei method.

SYNTAX: res = sw(Y,ta,d,s,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.s = 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
s: number of high frequency data points for each low frequency data point
 s= 4 ---> annual to quarterly
 s=12 ---> annual to monthly
 s= 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. & 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.

Application:

```
Y=load('c:\x\td\data\Y.anu');  
N = length(Y); n = s*N;  
% Defining the VCV matrix of stationary high-frequency time series  
% Assumption of the example: IMA(d,2)  
th1 = 0.9552; th2 = -0.0015; va = 0.87242 * ((223.5965)^2);  
acf0 = va * (1+th1^2+th2^2); acf1 = -va * th1 * (1-th2); acf2 = -va * th2;  
a0(1:n-d)=acf0; a1(1:n-d-1)=acf1; a2(1:n-d-2)=acf2;  
v=diag(a0)+diag(a1,-1)+diag(a2,-2); v=v+tril(v);  
res = sw(Y,1,1,4,v);  
tduni_print(res,'sw.sal');  
tduni_plot(res);  
edit sw.sal
```

ASCII file containing detailed output:

TEMPORAL DISAGGREGATION METHOD: Stram-Wei

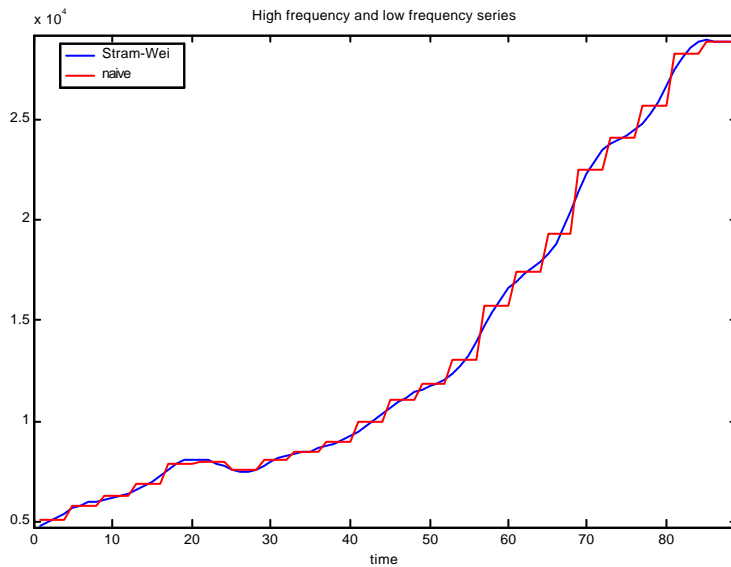
Number of low-frequency observations : 22
Frequency conversion : 4
Number of high-frequency observations : 88

Degree of differencing : 1
Type of disaggregation: sum (flow).

High frequency series (columnwise):

4792.4658
5015.8665
.....
.....
.....
28880.7153
28822.8148

Elapsed time: 0.1100



4. DENTON

PURPOSE: Temporal disaggregation using the Denton method

SYNTAX: res=denton_uni(Y,x,ta,d,s);

OUTPUT: res: a structure
res.meth = 'Denton';
res.N = Number of low frequency data
res.ta = Type of disaggregation
res.s = 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
s: number of high frequency data points for each low frequency data point
s= 4 ---> annual to quarterly
s=12 ---> annual to monthly
s= 3 ---> quarterly to monthly

LIBRARY: aggreg, bfl

SEE ALSO: 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.

Application:

```
Y=load('c:\x\td\data\Y.prn');  
x=load('c:\x\td\data\x.ind');  
res=denton_uni(Y,x,1,1,4);  
tduni_print(res,'td.sal');  
tduni_plot(res);  
edit td.sal
```


ASCII file containing detailed output:

```
*****  
TEMPORAL DISAGGREGATION METHOD: Denton  
*****
```

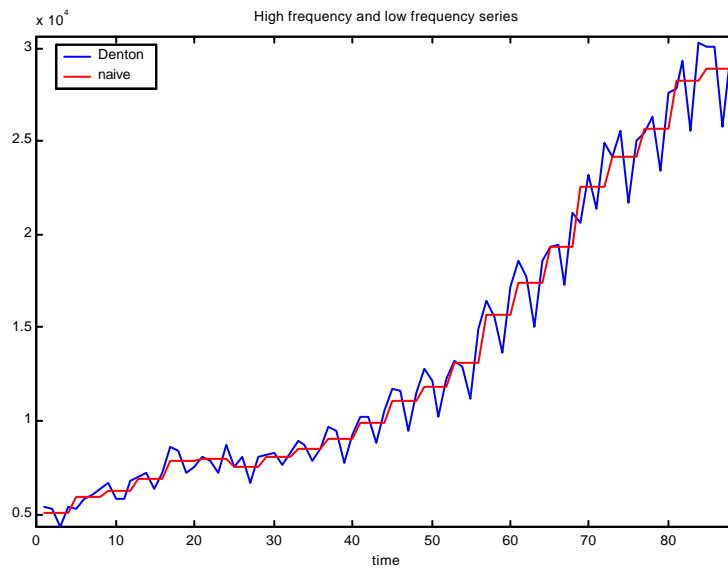
```
-----  
Number of low-frequency observations : 22  
Frequency conversion : 4  
Number of high-frequency observations : 88  
-----
```

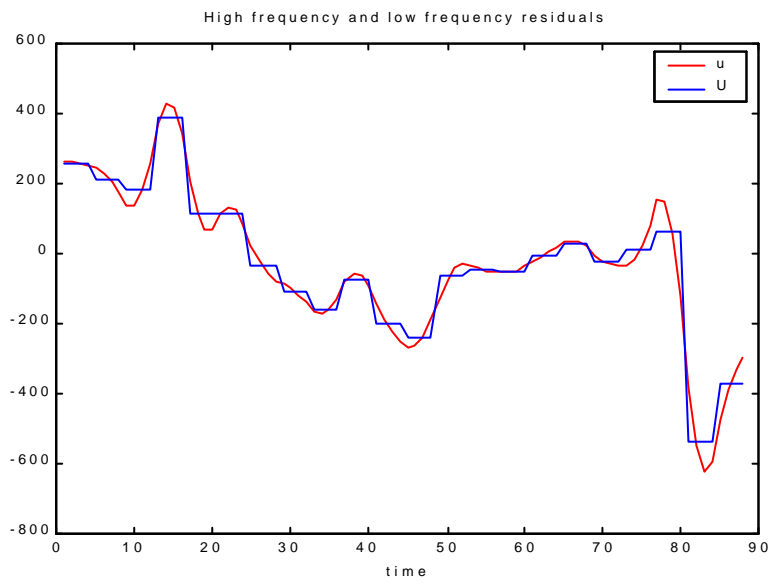
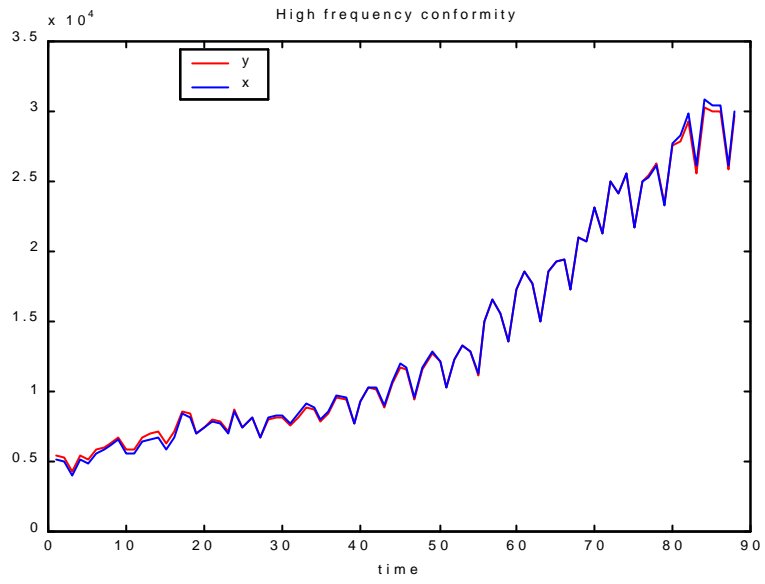
```
Degree of differencing : 1  
Type of disaggregation: sum (flow).  
-----
```

High frequency series (columnwise):

```
-----  
15374. 9285  
15169. 7571  
.....  
.....  
.....  
24883. 3098  
20609. 0705  
24415. 4509  
-----
```

```
Elapsed time: 0.0500  
-----
```





5. CHOW-LIN

PURPOSE: Temporal disaggregation using the Chow-Lin method

SYNTAX: res=chowlin(Y,x,ta,s,type);

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.s = 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
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.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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

type: estimation method:

type=0 ---> weighted least squares

type=1 ---> maximum likelihood

LIBRARY: aggreg

SEE ALSO: litterman, fernandez, td_plot, td_print

REFERENCE: Chow, G. y 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.

Application:

```
Y=load('c:\x\td\data\Y.prn');
x=load('c:\x\td\data\x.ind');
res=chowlin(Y,x,1,4,1);
td_print(res,'td.sal',1); % op1=1: series are printed in ASCII file
td_plot(res);
edit td.sal
```

ASCII file containing detailed output:

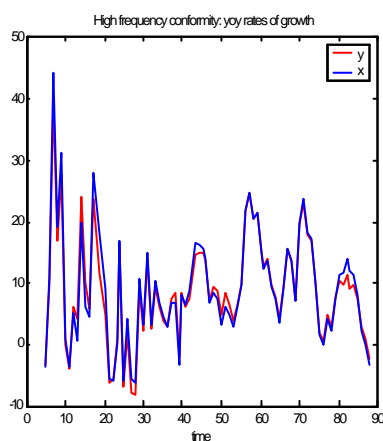
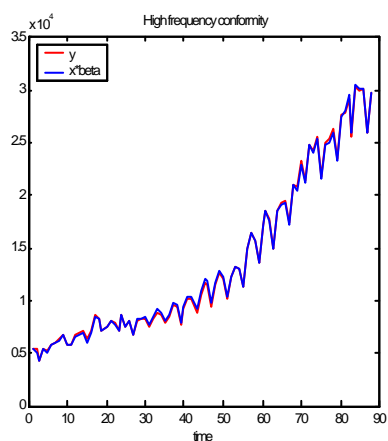
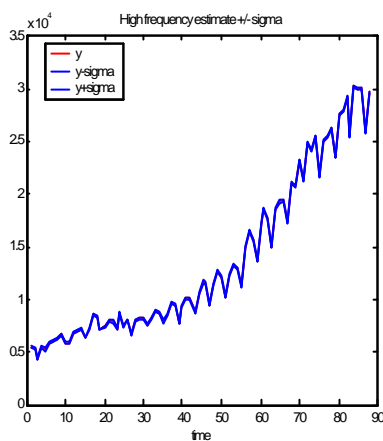
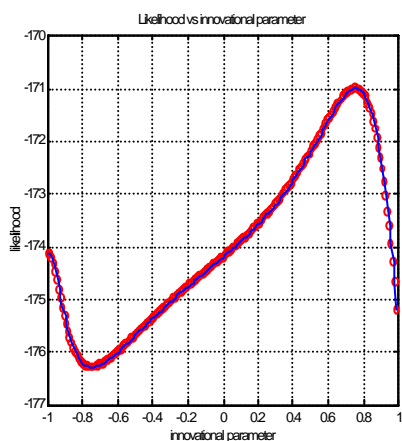
```
*****
TEMPORAL DISAGGREGATION METHOD: Chow-Lin
*****
-----
Number of low-frequency observations : 22
Frequency conversion : 4
Number of high-frequency observations: 88
Number of extrapolations : 0
Number of indicators (+ constant) : 2
-----
Type of disaggregation: sum (flow).
-----
Estimation method: Maximum likelihood.
-----
Beta parameters (columnwise):
* Estimate
* Std. deviation
* t-ratios
-----
215.4518      111.7079      1.9287
 0.9828      0.0069      142.0272
-----
Innovational parameter: 0.7600
-----
AIC: 10.0340
BIC: 10.1828
-----
Low-frequency correlation
- levels : 0.9998
- yoy rates : 0.9617
-----
High-frequency correlation
- levels : 0.9998
- yoy rates : 0.9812
-----
High-frequency volatility of yoy rates
- estimate : 8.4282
- indicator : 9.0226
- ratio : 0.9341
-----
```

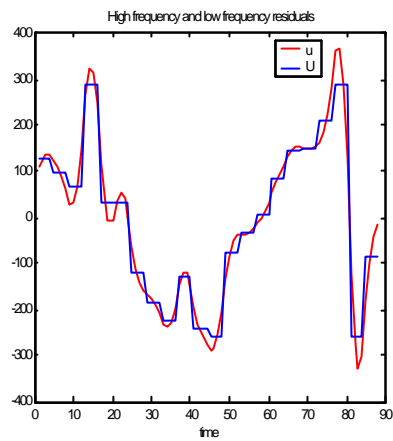
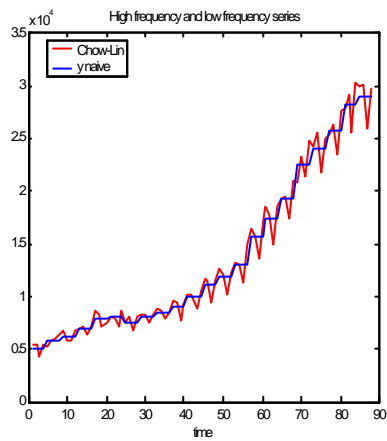
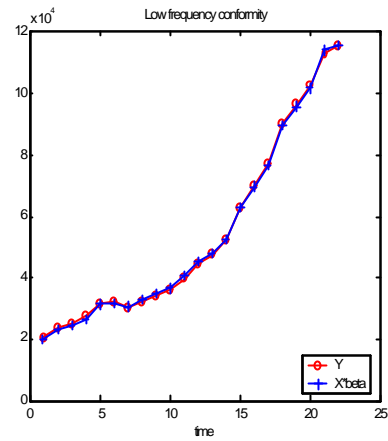
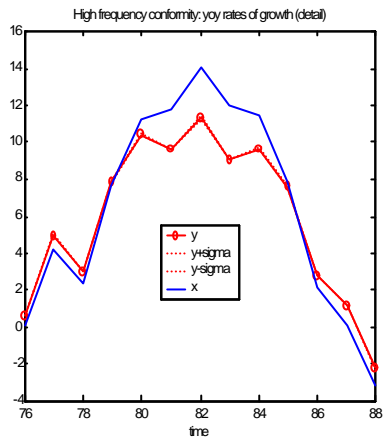
High frequency series (columnwise):

- * Estimate
- * Std. deviation
- * 1 sigma lower limit
- * 1 sigma upper limit
- * Residuals

5400.9896	114.8247	5286.1649	5515.8143	112.3095
5311.2409	83.7296	5227.5112	5394.9705	128.7034
.....
.....
.....
30079.6885	86.7557	29992.9328	30166.4443	-97.4913
25874.7702	86.2867	25788.4835	25961.0569	-43.9249
29614.4998	116.3242	29498.1756	29730.8240	-16.2417

Elapsed time: 1.8100





A variant to be applied with a fixed innovational parameter:

PURPOSE: Temporal disaggregation using the Chow-Lin method
rho parameter is fixed (supplied by the user)

SYNTAX: res=chowlin_fix(Y,x,ta,s,type,rho);

6. FERNÁNDEZ

PURPOSE: Temporal disaggregation using the Fernandez method

SYNTAX: res=fernandez(Y,x,ta,s);

OUTPUT: res: a structure

res.meth = 'Fernandez';
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.s = 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
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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

LIBRARY: aggreg

SEE ALSO: chowlin, litterman, td_plot, td_print

REFERENCE: Fernández, R.B.(1981)"Methodological note on the estimation of time series", Review of Economic and Statistics, vol. 63, n. 3, p. 471-478.

Application:

```
Y=load('c:\x\td\data\Y.prn');
x=load('c:\x\td\data\x.tri');
res=fernandez(Y,x,1,4);
td_print(res,'td.sal',1); % op1=1: series are printed in ASCII file
td_plot(res);
edit td.sal
```

ASCII file containing detailed output:

```
*****
TEMPORAL DISAGGREGATION METHOD: Fernandez
*****
-----
Number of low-frequency observations : 22
Frequency conversion : 4
Number of high-frequency observations: 90
Number of extrapolations : 2
Number of indicators (+ constant) : 2
-----
Type of disaggregation: sum (flow).
-----
Estimation method: Maximum likelihood.
-----
Beta parameters (columnwise):
* Estimate
* Std. deviation
* t-ratios
-----
564.9834      195.9404      2.8834
  0.9360      0.0292      32.0284
-----
Innovational parameter: 1.0000
-----
AIC: 9.6079
BIC: 9.7567
-----
Low-frequency correlation
- levels : 0.9998
- yoy rates : 0.9617
-----
High-frequency correlation
- levels : 0.9997
- yoy rates : 0.9817
-----
High-frequency volatility of yoy rates
- estimate : 8.3477
- indicator : 9.1506
- ratio : 0.9123
-----
```


High frequency series (columnwise):

- * Estimate
- * Std. deviation
- * 1 sigma lower limit
- * 1 sigma upper limit
- * Residuals

5396.6742	91.6250	5305.0492	5488.2992	-0.0000
5297.9198	60.8871	5237.0327	5358.8069	2.3349
.....
.....
.....
30021.1833	73.6977	29947.4856	30094.8810	920.9566
26022.3844	108.3992	25913.9852	26130.7837	977.8951
29586.1687	92.9937	29493.1750	29679.1625	1006.3644
.....
28366.5459	140.8431	28225.7028	28507.3889	1006.3644
29461.6792	176.5235	29285.1557	29638.2027	1006.3644

Elapsed time: 0.0500

Graphs are the same than in the Chow-Lin case, except that the first one (objective function vs innovational parameter) is not generated.

7. LITTERMAN

PURPOSE: Temporal disaggregation using the Litterman method

SYNTAX: res=litterman(Y,x,ta,s,type);

OUTPUT: res: a structure

res.meth = 'Litterman';
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.s = 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
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.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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

type: estimation method:

type=0 ---> weighted least squares

type=1 ---> maximum likelihood

LIBRARY: aggreg

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

Application:

```
Y=load('c:\x\td\data\Y.prn');
x=load('c:\x\td\data\x.tri');
res=litterman(Y,x,1,4,0);
td_print(res,'td.sal',0); % op1=0: series are not printed in ASCII file
td_plot(res);
edit td.sal
```

ASCII file containing detailed output:

```
*****
TEMPORAL DISAGGREGATION METHOD: Litterman
*****
-----
Number of low-frequency observations : 22
Frequency conversion                 : 4
Number of high-frequency observations: 90
Number of extrapolations             : 2
Number of indicators (+ constant)    : 2
-----
Type of disaggregation: sum (flow).
-----
Estimation method: Weighted least squares.
-----
Beta parameters (columnwise):
* Estimate
* Std. deviation
* t-ratios
-----
1205.4851      233.5241      5.1621
  0.7910        0.0480      16.4821
-----
Innovational parameter: 0.9700
-----
AIC: 7.9478
BIC: 8.0966
-----
Low-frequency correlation
- levels      : 0.9998
- yoy rates   : 0.9617
-----
High-frequency correlation
- levels      : 0.9994
- yoy rates   : 0.9735
-----
High-frequency volatility of yoy rates
- estimate    : 7.6249
- indicator   : 9.1506
- ratio       : 0.8333
-----
Elapsed time: 2.5300
```

A variant to be applied with a fixed innovational parameter:

PURPOSE: Temporal disaggregation using the Litterman method
mu parameter is fixed (supplied by the user)

SYNTAX: res=litterman_fix(Y,x,ta,s,type,mu);

Graphical output contains the same information than in the Chow-Lin case.

8. SANTOS SILVA-CARDOSO (ssc)

function res=ssc(Y,x,ta,s,type)

PURPOSE: Temporal disaggregation using the dynamic Chow-Lin method proposed by Santos Silva-Cardoso (2001).

SYNTAX: res=ssc(Y,x,ta,s,type);

OUTPUT: res: a structure

res.meth	= 'Santos Silva-Cardoso';
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.s	= frequency conversion between low and high freq.
res.p	= number of regressors (+ 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
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
res.beta	= estimated model parameters (excluding y(0))
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
res.val	= Objective function used by the estimation method
res.r	= grid of dynamic parameters used by the estimation method
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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

type: estimation method:

type=0 ---> weighted least squares

type=1 ---> maximum likelihood

LIBRARY: aggreg

SEE ALSO: chowlin, litterman, fernandez, td_plot, td_print

REFERENCE: Santos, J.M.C. y Cardoso, F.(2001) "The Chow-Lin method using dynamic models", Economic Modelling, vol. 18, p. 269-280.

Application:

```
Y=load('c:\x\td\data\Y.prn');
x=load('c:\x\td\data\x.tri');
res=ssc(Y,x,1,4,1);
td_print(res,'td.sal',0);
edit td.sal;
% Calling graph function
td_plot(res);
```

ASCII file containing detailed output:

TEMPORAL DISAGGREGATION METHOD: Santos Silva-Cardoso

Number of low-frequency observations : 32
Frequency conversion : 4
Number of high-frequency observations: 128
Number of extrapolations : 0
Number of indicators (+ constant) : 2

Type of disaggregation: sum (flow).

Estimation method: Maximum likelihood.

Beta parameters (columnwise):

* Estimate		
* Std. deviation		
* t-ratios		

1.0946	3.7817	0.2895
0.6718	0.0049	136.9983

Dynamic parameter: 0.2600

Long-run beta parameters (columnwise):

1.4792
0.9078

Truncation remainder: expected y(0):

* Estimate		
* Std. deviation		
* t-ratios		

310.3328	90.5351	3.4278

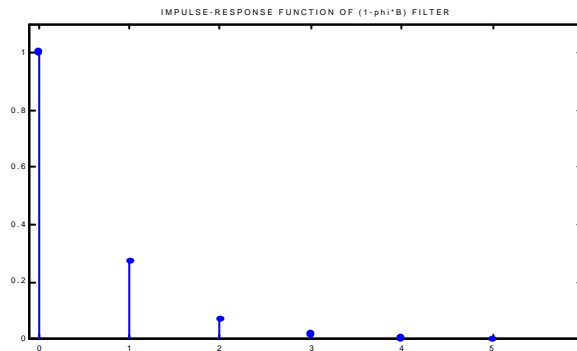
AIC: 5.2524
BIC: 5.3898

Low-frequency correlation
- levels : 0.9994
- yoy rates : 0.8561

High-frequency correlation
- levels : 0.9993
- yoy rates : 0.8881

High-frequency volatility of yoy rates
- estimate : 2.0592
- indicator : 2.3430
- ratio : 0.8789

Graphical output contains the same information than in the Chow-Lin case and includes a plot of the implied impulse-response function:



A variant to be applied with a fixed innovational parameter:

PURPOSE: Temporal disaggregation using the Santos Silva-Cardoso method
Phi parameter is fixed (supplied by the user)

SYNTAX: res=ssc_fix(Y,x,ta,s,type,phi);

9. GUERRERO

function res=guerrero(Y,x,ta,s,rexw,rex);

PURPOSE: ARIMA-based temporal disaggregation: Guerrero method

SYNTAX: res=guerrero(Y,x,ta,s,rexw,rex);

OUTPUT: res: a structure

res.meth = 'Guerrero';
res.ta = type of disaggregation
res.N = nobs. of low frequency data
res.n = nobs. of high-frequency data
res.pred = number of extrapolations
res.s = frequency conversion between low and high freq.
res.p = number of regressors (+ intercept)
res.Y = low frequency data
res.x = high frequency indicators
res.w = scaled indicator (preliminary hf estimate)
res.y1 = first stage high frequency estimate
res.y = final 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.delta = high frequency discrepancy (y1-w)
res.u = high frequency residuals (y-w)
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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

rexw, rexd ---> a structure containing the parameters of ARIMA model
for indicator and discrepancy, respectively (see calT function)

LIBRARY: aggreg, calT, numpar, ols

SEE ALSO: chowlin, litterman, fernandez, td_print, td_plot

REFERENCE: Guerrero, V. (1990) "Temporal disaggregation of time series: an ARIMA-based approach", International Statistical Review, vol. 58, p. 29-46.

Application:

```
Y=load('c:\x\td\data\Y.prn');
x=load('c:\x\td\data\x.tri');
% -----
% Inputs for td library
% Type of aggregation
ta=1;
% Frequency conversion
s=12;
% Model for w: (0,1,1)(1,0,1)
rexw.ar_reg = [1];
rexw.d = 1;
rexw.ma_reg = [1 -0.40];
rexw.ar_sea = [1 0 0 0 0 0 0 0 0 0 0 -0.85];
rexw.bd = 0;
rexw.ma_sea = [1 0 0 0 0 0 0 0 0 0 0 -0.79];
rexw.sigma = 4968.716^2;
% Model for the discrepancy: (1,2,0)(1,0,0)
% See: Martinez and Guerrero, 1995, Test, 4(2), 359-76.
rexd.ar_reg = [1 -0.43];
rexd.d = 2;
rexd.ma_reg = [1];
rexd.ar_sea = [1 0 0 0 0 0 0 0 0 0 0 0.62];
rexd.bd = 0;
rexd.ma_sea = [1];
rexd.sigma = 76.95^2;
% Calling the function: output is loaded in structure res
res=guerrero(Y,x,ta,s,rexw,rex);
% Calling printing function
% Name of ASCII file for output
file_sal='guerrero.sal';
output=0; % Do not include series
td_print_G(res,file_sal,output);
edit guerrero.sal;
% Calling graph function
td_plot(res);
```

ASCII file containing detailed output:

```
*****
TEMPORAL DISAGGREGATION METHOD: Guerrero
*****

-----
Number of low-frequency observations :    5
Frequency conversion                 :   12
Number of high-frequency observations:   60
Number of extrapolations             :    0
Number of indicators (+ constant)    :    2
-----

Type of disaggregation: sum (flow).
-----

Estimation method: BLUE.
-----

Beta parameters (columnwise):
```

```

* Estimate
* Std. deviation
* t-ratios
-----
219988.6766      974531.6756      4.4299
1723.8723       6174.6540        3.5819
-----
AIC:    7.5245
BIC:    7.3683
-----
Low-frequency correlation (Y,X)
- levels      : 0.9003
- yoy rates   : 0.9973
-----
High-frequency correlation (y,x)
- levels      : 0.9289
- yoy rates   : 0.9835
-----
High-frequency volatility of yoy rates
- estimate    : 3.6623
- indicator   : 6.2899
- ratio       : 0.5823
-----
High-frequency correlation (y,x*beta)
- levels      : 0.9289
- yoy rates   : 0.9832
-----
Compatibility test:
- k :    0.9526
-----

ARIMA model for scaled indicator:

( 0 1 1 ) ( 1 0 1 )

- Regular AR operator:

1.0000

- Regular MA operator:

1.0000 -0.4000

- Seasonal AR operator:

1.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 -0.8500

- Seasonal MA operator:

1.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 -0.7900
-----

ARIMA model for discrepancy :

( 1 2 0 ) ( 1 0 0 )

- Regular AR operator:

1.0000 -0.4300

- Regular MA operator:

1.0000

```

- Seasonal AR operator:

1.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.6200

- Seasonal MA operator:

1.0000

Elapsed time: 0.4400



Graphical output contains the same information than in the Chow-Lin case.

10. MULTIVARIATE ROSSI

function res = rossi(Y,x,z,ta,s,type);

PURPOSE: Multivariate temporal disaggregation with transversal constraint

SYNTAX: res = rossi(Y,x,z,ta,s,type);

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.s = Frequency conversion
res.y = High frequency estimate
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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

type: univariate temporal disaggregation procedure used to compute preliminary estimates

type = 1 ---> Fernandez

type = 2 ---> Chow-Lin

type = 3 ---> Litterman

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.

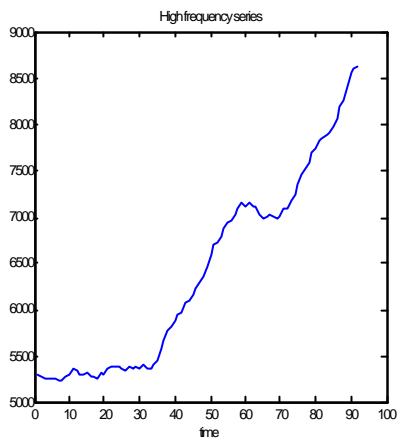
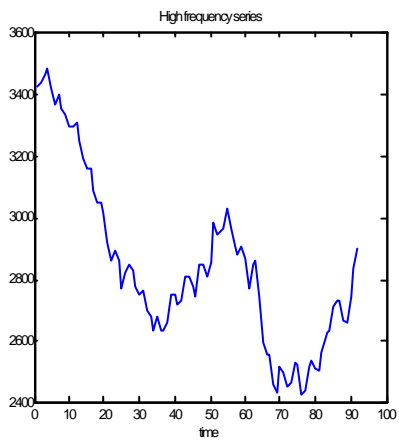
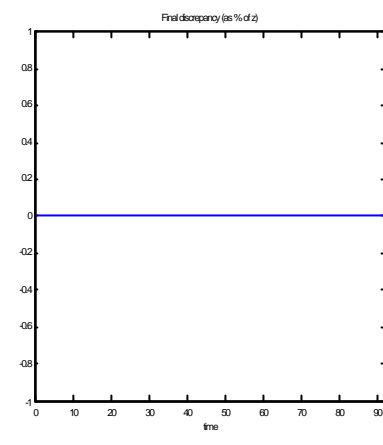
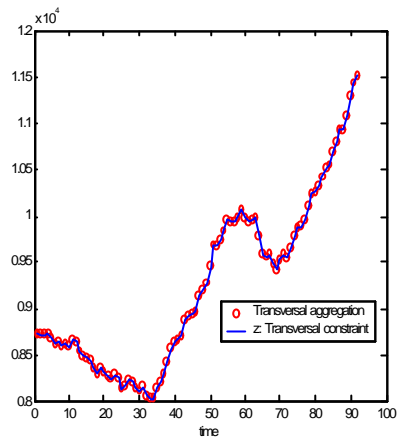
di Fonzo, T. (1994) "Temporal disaggregation of a system of time series when the aggregate is known: optimal vs. adjustment methods", INSEE-Eurostat Workshop on Quarterly National Accounts, Paris, december.

Application:

```
Y=load('YY.anu'); % Loading low frequency data
x=load('x.tri'); % Loading high frequency data
z=load('z.prn'); % Loading high frequency transversal restriction
res=rossi(Y,x,z,2,4,1);
mtd_print(res,'mtd.sal');
edit mtd.sal;
mtd_plot(res,z);
```

ASCII file containing detailed output:

```
*****
TEMPORAL DISAGGREGATION METHOD: Multivariate Rossi
*****
-----
Number of low-frequency observations   :   23
Frequency conversion                   :    4
Number of high-frequency observations  :   92
Number of extrapolations               :    0
-----
Type of disaggregation: average (index).
-----
Preliminary univariate disaggregation:  Fernandez
-----
High frequency series (columnwise):
  * Point estimate
-----
3424.2881  5311.2720
3436.0588  5280.4786
.....
.....
.....
2835.1833  8614.4139
2899.5740  8625.9809
-----
Elapsed time:   1.2600
```



11. MULTIVARIATE DENTON

function res = denton(Y,x,z,ta,s,d);

PURPOSE: Multivariate temporal disaggregation with transversal constraint

SYNTAX: res = denton(Y,x,z,ta,s,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.s = Frequency conversion
res.d = Degree of differencing
res.y = High frequency estimate
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: nzx1 ---> 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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 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: 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

Application:

```
Y=load('YY.anu'); % Loading low frequency data
x=load('x.tri'); % Loading high frequency data
z=load('z.prn'); % Loading high frequency transversal restriction
res=denton(Y,x,z,2,4,1);
mtd_print(res,'mtd.sal');
edit mtd.sal;
mtd_plot(res,z);
```

ASCII file containing detailed output:

```
*****
TEMPORAL DISAGGREGATION METHOD: Multivariate Denton
*****
```

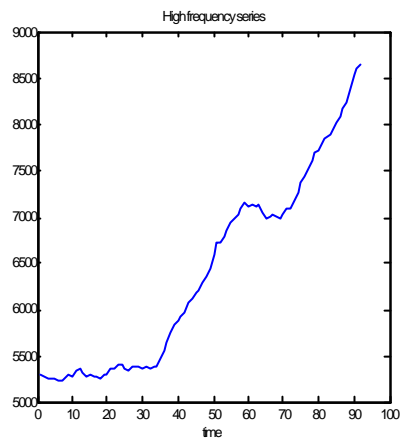
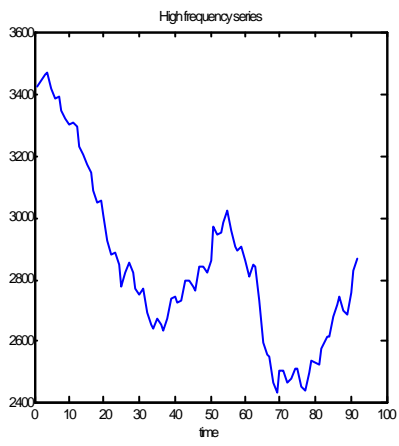
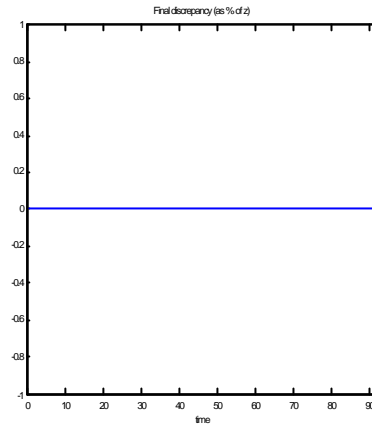
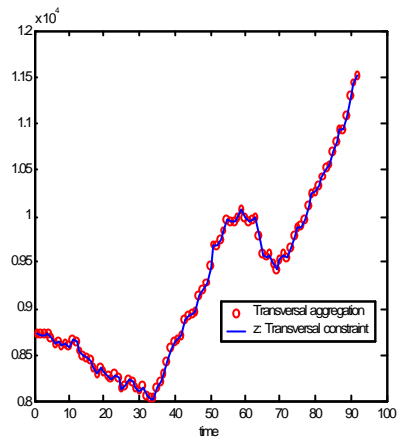
```
-----
Number of low-frequency observations : 23
Frequency conversion                  : 4
Number of high-frequency observations : 92
Number of extrapolations              : 0
-----
```

```
Degree of differencing                : 1
Type of disaggregation: average (index).
-----
```

```
High frequency series (columnwise):
* Point estimate
```

```
-----
3752. 9096  4982. 6505
3459. 3681  5257. 1693
.....
.....
.....
2757. 8458  8545. 8074
2825. 1411  8624. 4561
2867. 5816  8657. 9733
-----
```

```
Elapsed time: 0.2800
```



12. DI FONZO

function res = difonzo(Y,x,z,ta,s,type,f);

PURPOSE: Multivariate temporal disaggregation with transversal constraint

SYNTAX: res = difonzo(Y,x,z,ta,s,type,f);

OUTPUT: res: a structure

res.meth = 'Multivariate di Fonzo';
res.N = Number of low frequency data
res.n = Number of high frequency data
res.pred = Number of extrapolations
res.ta = Type of disaggregation
res.s = Frequency conversion
res.type = Model for high frequency innovations
res.beta = Model parameters
res.y = High frequency estimate
res.d_y = High frequency estimate: std. deviation
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, $m \geq M$ see (*)

z: nx1 ---> 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

s: number of high frequency data points for each low frequency data points

s= 4 ---> annual to quarterly

s=12 ---> annual to monthly

s= 3 ---> quarterly to monthly

type: model for the high frequency innovations

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 automatically performed when $n > sN$.

If $n = nz > sN$ restricted extrapolation is applied.

Finally, if $n > nz > sN$ extrapolation is performed in constrained form in the first $nz - sN$ observations and in free form in the last $n - nz$ observations.

LIBRARY: aggreg, dif, vec, desvec

SEE ALSO: denton, mtd_print, mtd_plot

REFERENCE: di Fonzo, T. (1990) "The estimation of M disaggregate time series when contemporaneous and temporal aggregates are known", Review of Economics and Statistics, vol. 72, n. 1, p. 178-182.

Application:

```
Y=load('YY.anu'); % Loading low frequency data
x=load('x.tri'); % Loading high frequency data
z=load('z.prn'); % Loading high frequency transversal restriction
res = difonzo(Y,x,z,2,4,1);
mtd_print(res,'mtd.sal');
edit mtd.sal;
mtd_plot(res,z);
```

ASCII file containing detailed output:

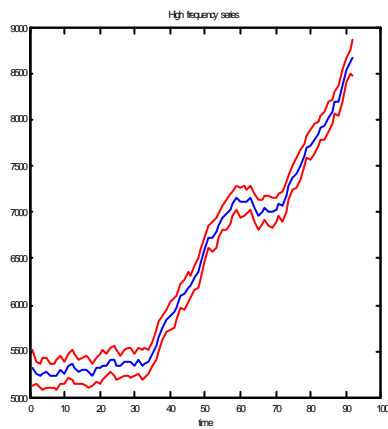
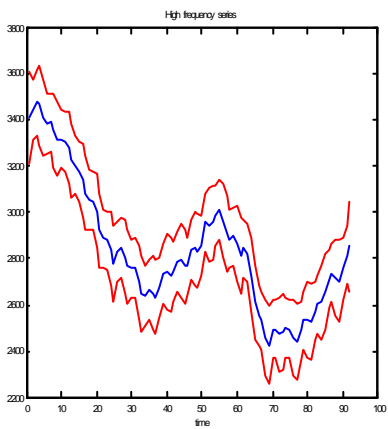
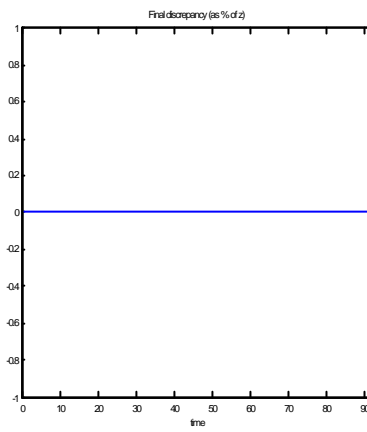
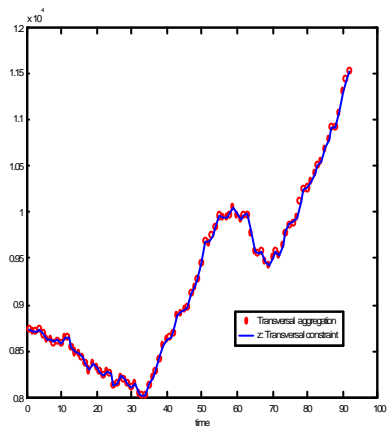
```
*****
TEMPORAL DISAGGREGATION METHOD: Multivariate di Fonzo
*****
-----
Number of low-frequency observations : 23
Frequency conversion : 4
Number of high-frequency observations : 92
Number of extrapolations : 0
-----
Model for the innovations: random walk.
Type of disaggregation: average (index).
-----
High frequency series (columnwise):
  * Point estimate
-----
3413. 3839 5322. 1762
3447. 4092 5269. 1282
.....
.....
.....
2758. 4657 8545. 1875
2817. 9882 8631. 6090
2856. 1605 8669. 3944
```

High frequency series (columnwise):
 * Std. deviation

197. 8732 197. 8732
127. 3900 127. 3900

137. 9397 137. 9397
128. 1006 128. 1006
194. 9112 194. 9112

Elapsed time: 0. 3300



APPENDIX: RELATIONSHIPS AMONG FUNCTIONS IN THE LIBRARY

The “X → Y” notation means “X function calls Y function”.

- bfl → sw
- denton_uni → aggreg, bfl
- sw → aggreg, aggreg_v, dif, movingsum

- chowlin → aggreg
- fernandez → aggreg
- litterman → aggreg
- ssc → aggreg
- guerrero → aggreg, calT, numpar, **ols**^(*)

- rossi → aggreg, vec, desvec, fernandez, chowlin, litterman
- denton → aggreg, aggreg_v, dif, vec, desvec
- difonzo → aggreg, dif, vec, desvec

- bal → vec, desvec
- td_print → tasa, aggreg
- td_print_G → tasa, aggreg, **mprint**^(*)
-
- td_plot → tasa
- tduni_plot → temporal_agg

^(*) From James Lesage's *Econometric Toolbox*

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