

## Contents

<b>1 FIGURE 9 – BACCALA ET AL. (2016) DTF: UNIFIED ASYMPTOTIC THEORY</b>	<b>1</b>
Generating data set for analysis . . . . .	1
Equation (11) . . . . .	2
Connectivity diagram . . . . .	2
DTF estimation . . . . .	4
Figure depicted in the article Baccala et al (2016) . . . . .	5
Some remarks: . . . . .	6

## 1 FIGURE 9 – BACCALA ET AL. (2016) DTF: UNIFIED ASYMP-TOTIC THEORY

DESCRIPTION:

Routine `figure9_example3_idtf_ns2000.m` publish

Linear five-dimensional VAR(2) model

LA Baccala, DY Takahashi, K Sameshima (2016) Directed Transfer Function: Unified Asymptotic Theory and Some of its Implications. *IEEE Transactions on Biomedical Engineering* **PP**.

<http://dx.doi.org/10.1109/TBME.2016.2550199>

Example 3: Asymptotic DTF: Example 3 5-VAR model - open / closed loop.

x1 ==> x2 ==> x3 ==> x4 <=> x5  
^-----/-

\*This is the close loop model.\*

## Contents

- Generating data set for analysis
- Equation (11)
- Connectivity diagram
- DTF estimation
- Figure depicted in the article Baccala et al (2016)
- Some remarks:

### Generating data set for analysis

```
clear; clc; format compact
flgPlotStyle = 'Print'; % or 'Screen' mode
flgRandomize = 0; % Generate the specific data set used in Fig. 9.
ns = 2000; % number of sample points
nDiscard = 20000; % number of points discarded at beginning of simulation
p = 2; % model order

if (exist('figure9_example3_idtf_ns2000.mat') == 2) & is_octave & ~flgRandomize
    ,
    load figure9_example3_idtf_ns2000
else
    [u] = fbaccala2016_example3(ns, nDiscard, flgRandomize);
    if ~is_octave & ~flgRandomize,
```

```

    save figure9_example3_idtf_ns2000 u
end;
end;

chLabels = [];% Using default labeling schema for channel identification

```

```

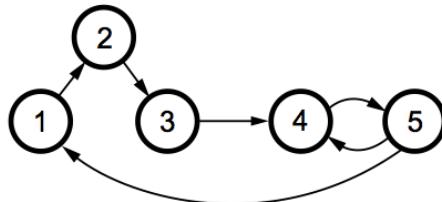
=====
Asymptotic DTF: Example 3 5-VAR model - open / closed loop.
x1 ==> x2 ==> x3 ==> x4 <==> x5
^-----/
This is the close loop model.
=====
```

### Equation (11)

$$\begin{aligned}
 x_1(n) &= 0.25\sqrt{2}x_1(n-1) + 0.25\sqrt{2}x_2(n-1) + w_1(n) \\
 x_2(n) &= -0.25\sqrt{2}x_1(n-1) + 0.25\sqrt{2}x_2(n-1) + w_2(n) \\
 x_3(n) &= 0.5x_2(n-1) + w_3(n)
 \end{aligned} \tag{12}$$

Equation (11) from Baccala et al. *IEEE Trans Biomed Engin.*, 2016.

### Connectivity diagram



Example 3 loop connectivity structure following (11). Signals from any structure reach all other structures. from Baccala et al. *IEEE Trans Biomed Engin.*, 2016.

Data pre-processing: detrending and normalization options

```

flgDetrend = 1;% Detrending the data set
flgStandardize = 0;% No standardization
[nChannels,nSegLength] = size(u);
if nChannels > nSegLength, u = u.';
[nChannels,nSegLength] = size(u);
end;
if flgDetrend,
for i=1:nChannels, u(i,:) = detrend(u(i,:)); end;
disp('Time series were detrended.');
end;
if flgStandardize,
for i=1:nChannels, u(i,:) = u(i,:)/std(u(i,:)); end;
disp('Time series were scale-standardized.');
end;

```

```
Time series were detrended.
```

## MVAR model estimation

```
maxIP = 30; % maximum model order to consider.  
alg = 1; % 1: Nutall-Strand MVAR estimation algorithm  
criterion = 1; % 1: AIC, Akaike Information Criteria  
disp('Running MVAR estimation and GCT analysis routines.')  
[IP, pf, A, pb, B, ef, eb, vaic, Vaicv] = mvar(u, maxIP, alg, criterion);  
disp(['Number of channels = ' int2str(nChannels) ' with ' ...  
    int2str(nSegLength) ' data points; MAR model order = ' int2str(IP) '.']);
```

```
Running MVAR estimation and GCT analysis routines.  
maxOrder limited to 30  
IP=1 vaic=79027.676838  
IP=2 vaic=76136.019225  
IP=3 vaic=76162.273276
```

```
Number of channels = 5 with 2000 data points; MAR model order = 2.
```

## Testing for adequacy of MAR model fitting through Portmanteau test

```
h = 20; % testing lag  
MVARadequacy_signif = 0.05; % VAR model estimation adequacy significance  
% level  
aValueMVAR = 1 - MVARadequacy_signif; % Confidence value for the testing  
flgPrintResults = 1;
```

Granger causality test (GCT) and instantaneous GCT

```
gct_signif = 0.01; % Granger causality test significance level  
igct_signif = 0.01; % Instantaneous GCT significance level  
flgPrintResults = 1;  
[Tr_gct, pValue_gct, Tr_igct, pValue_igct] = gct_alg(u, A, pf, gct_signif, ...  
    igct_signif, flgPrintResults);
```

```
-----  
GRANGER CAUSALITY TEST  
=====  
Connectivity matrix:  


|      |      |      |      |      |
|------|------|------|------|------|
| NaN  | 0    | 0    | 0    | 1.00 |
| 1.00 | NaN  | 0    | 0    | 0    |
| 0    | 1.00 | NaN  | 0    | 0    |
| 0    | 0    | 1.00 | NaN  | 1.00 |
| 0    | 0    | 0    | 1.00 | NaN  |

  
Granger causality test p-values:  


|      |      |      |      |      |
|------|------|------|------|------|
| NaN  | 0.54 | 0.11 | 0.47 | 0    |
| 0    | NaN  | 0.27 | 0.59 | 0.97 |
| 0.45 | 0    | NaN  | 0.23 | 0.88 |
| 0.13 | 0.89 | 0    | NaN  | 0    |
| 0.36 | 0.87 | 0.49 | 0    | NaN  |

  
-----  
INSTANTANEOUS GRANGER CAUSALITY TEST  
=====  
Instantaneous connectivity matrix:  


|     |   |   |   |   |
|-----|---|---|---|---|
| NaN | 0 | 0 | 0 | 0 |
|-----|---|---|---|---|


```

```

          0      NaN      0      0      0
          0      0      NaN      0      0
          0      0      0      NaN      0
          0      0      0      0      NaN
Instantaneous Granger causality test p-values:
      NaN      0.56      0.64      0.08      0.41
      0.56      NaN      0.94      0.33      0.28
      0.64      0.94      NaN      0.74      0.03
      0.08      0.33      0.74      NaN      0.71
      0.41      0.28      0.03      0.71      NaN
>>> Instantaneous Granger causality NOT detected.

```

## DTF estimation

DTF analysis results are saved in **c** structure. See `asymp_dtf.m` or issue `>> help asymp_dtf` command for more detail.

```

metric = 'info'; % euc = original PDC or DTF;
                  % diag = generalized PDC (gPDC) or directed coherence (DC);
                  % info = information PDC (iPDC) or iDTF.

nFreqs = 128;
alpha = 0.01;

c = asymp_dtf(u,A,pf,nFreqs,metric,alpha);

* Information DTF and asymptotic statistics

```

$|_iDTF(\lambda)|^2$  Matrix Layout Plotting

```

switch lower(flgPlotStyle)
  case 'print'
    flgColor =[0];       % white background
    flgMax = 'TCI';
    flgSignifColor = 1; % black + gray
    flgScale = 3;       % [0 max(flgMax)]
  otherwise % 'screen'
    flgColor =[1];       % Colored background
    flgMax = 'TCI';
    flgSignifColor = 3; % red + green
    flgScale = 2;       % [0 1]/[0 .1]/[0 .01]
end;

% -----Plotting options flag setting-----
%      [1 2 3 4 5 6 7]
flgPrinting=[1 1 1 2 2 0 1];
%      | | | | | 7 Spectra(0: w/o SS; 1: Linear; 2: log-scale)
%      | | | | | 6 Coherence
%      | | | | 5 Plot lower confidence limit (legacy)
%      | | | | 4 Plot upper confidence limit
%      | | | 3 Significant DTF(w) in red line (legacy)
%      | | 2 Patnaik threshold level in black dashed-line
%      1 plot DTF
%-----


fs = 1;           % sampling frequency

```

```

w = fs*(0:(nFreqs-1))/2/nFreqs;
w_max = fs/2;

h=figure;
set(h, 'NumberTitle', 'off', 'MenuBar', 'none', ...
    'Name', '[Asymptotic DTF] Fig 9. Example 3 - iDTF, ns = 2000')

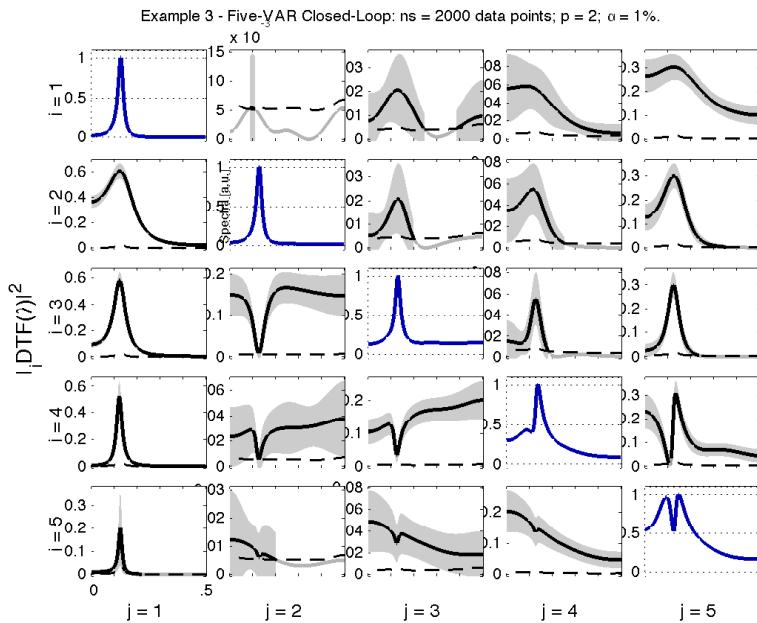
[h xlabel ylabel] = xplot(c, ...
    flgPrinting, fs, w_max, chLabels, flgColor, flgScale, flgMax, flgSignifColor);

% xplot_title(alpha, metric, measure(c));

[ax, hT] = suplabel(['Example 3 - Five-VAR Closed-Loop: ns = ' ...
    int2str(ns) ' data points; p = ' int2str(c.p) ' \alpha = ' ...
    int2str(100*alpha) '%.'], 't');
set(hT, 'FontSize', 10); % Subtitle font size

drawnow

```



Uncomment the command line bellow to generate an eps output file

```
% print -depsc2 -painters Fig9_example3_idtf_ns2000.eps
```

### Figure depicted in the article Baccala et al (2016)

Figure 9, reproduced from article.

### Some remarks:

1. As usual, figure 9 underwent some cosmetic edit and addition of a inlet graph

This completes the **Figure 9** generation (Baccala et al, 2016)'

