A tutorial for the improved Loewner Framework for modal analysis

The Loewner Framework was extended to modal parameters identification from mechanical systems in [3-5] with its main limitation being a single-input multi-output method. In this tutorial, a computationally improved version of the LF, the iLF, extending the LF capability to multi-input multi-output (MIMO), is presented with the scope of guiding the reader to its use.

The code has been developed in MATLAB R2021b, it is not guaranteed to work in earlier versions of the software. Note that if you are on a MATLAB version lower than R2020a you might have trouble with the tiledlayout legend in this tutorial.

When using this, or part of, tutorial release, please always cite the following:

- [1] G. Dessena and M. Civera, 'Improved Tangential Interpolation-based Multi-input Multi-output Modal Analysis of a Full Aircraft', European Journal of Mechanics A/Solids, vol. 109. Elsevier BV, p. 105495, Jan. 2025. doi: 10.1016/j.euromechsol.2024.105495.
- [2] G. Dessena, M. Civera, L. Zanotti Fragonara, D. I. Ignatyev, and J. F. Whidborne, A Loewner-Based System Identification and Structural Health Monitoring Approach for Mechanical Systems', Structural Control and Health Monitoring, vol. 2023. Hindawi Limited, pp. 1–22, Apr. 18, 2023. doi: 10.1155/2023/1891062.
- [3] G. Dessena, 'A tutorial for the improved Loewner Framework for modal analysis', Software, Universidad Carlos III de Madrid, 2024. doi: 10.5281/zenodo.13863292. Available at: https://zenodo.org/records/zenodo.13863292.

The Model

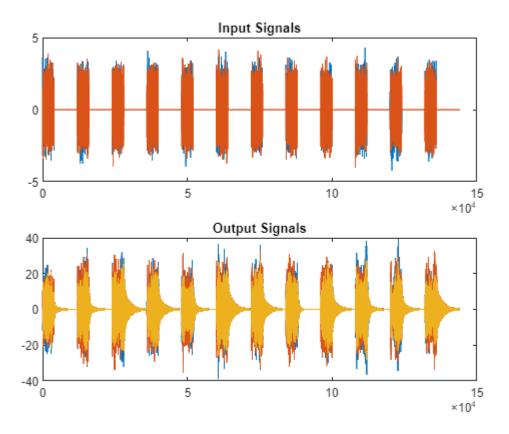
The model is a sample system provided with MATLAB releases and consists of a two-input/three-output system excited by several bursts of random noise (see Figure 1). Each burst lasts for 1 second, and there are 2 seconds between the end of each burst and the start of the next. The data is sampled at 4 kHz.

See below for the time histories and FRFs of the system.

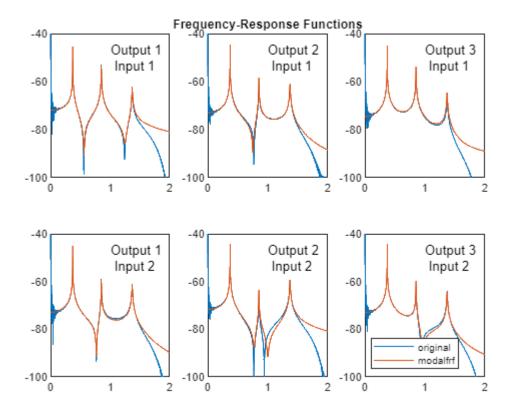
```
clear all
close all

load modaldata

subplot(2,1,1)
plot(Xburst)
title('Input Signals')
subplot(2,1,2)
plot(Yburst)
title('Output Signals')
```



```
burstLen = 12000;
[frf,f] = modalfrf(Xburst,Yburst,fs,burstLen);
phfr = [373 852 1371];
[fn,dr,ms,ofrf] = modalfit(frf,f,fs,6,'PhysFreq',phfr);
for k = 1:2
    for m = 1:3
        subplot(2,3,m+3*(k-1))
        plot(f/1000,10*log10(abs(frf(:,m,k))))
        plot(f/1000,10*log10(abs(ofrf(:,m,k))))
        hold off
        text(1,-50,[['Output ';' Input '] num2str([m k]')])
        ylim([-100 -40])
        if k ==2 & m==3
            legend({'original', 'modalfrf'}, "Location", "southwest")
        end
    end
end
subplot(2,3,2)
title('Frequency-Response Functions')
```



Here the iLF with MIMO capability is implemented on the system above. The iLF ability to deal with missing data is used to avoid the drift at lower frequency shown in the FRF plots. For the interpolation f > 100Hz are considered.

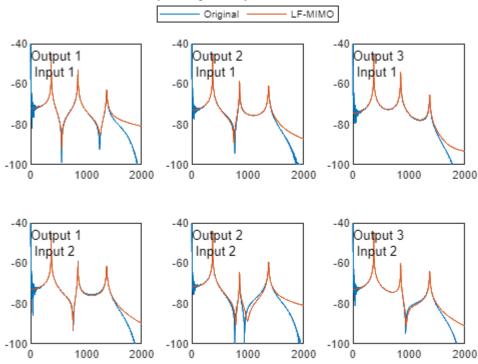
```
si = f*complex(0,1);
for i =1:max(size(ofrf))
    Hi(:,:,i) = squeeze(ofrf(i,:,:));
end
tic
[id,model,fit] = iLF_id(Hi(:,:,301:2:end),si(301:2:end),6);
t=toc;
disp(string(t)+" s on MATLAB R2021b - Windows 10 - 128 GB DDR4 RAM - 16-core
Intel(R) Xeon(R) CPU E5-2698 v3 @ 2.30GHz")
```

3.9684 s on MATLAB R2021b - Windows 10 - 128 GB DDR4 RAM - 16-core Intel(R) Xeon(R) CPU E5-2698 v3 @ 2.30GHz

Let's now plot the iLF-fitted model against the original FRF:

```
for m = 1:3
    nexttile
    plot(abs(si),10*log10(abs(frf(:,m,k))))
    hold on
    plot(abs(si),10*log10(abs(squeeze(FRF_fit(m,k,:)))))
    hold off
    text(1,-50,[['Output ';' Input '] num2str([m k]')])
    ylim([-100 -40])
    end
end
leg = legend({'Original', 'LF-
MIMO'},'Location','northoutside','Orientation','horizontal');
title(tl,'Frequency-Response Functions')
leg.Layout.Tile = 'north';
```

Frequency-Response Functions



Modal parameters Comparison

The expected modal parameters (first row - ω_n , second row - ζ_n , third to last row - ϕ_n) are the following:

```
ans =
```

```
    (372.6801
    852.5041
    1370.57

    (0.000780824
    0.001785605
    0.002870592

    (0.8469745
    1.0
    0.6384277

    1.0
    0.2523416
    -1.0

    (0.9336985
    -0.7849187
    0.3279202
```

The modal parameter identified via modalfit - standard option is least squares complex exponential (LSCE):

```
    372.6797
    852.51
    1370.56

    0.0007898842
    0.001788482
    0.002863741

    0.8539813
    1.0
    0.7205156

    1.0
    0.2996384
    -1.0

    0.9452163
    -0.8566122
    0.4397303
```

Lastly, those identified via iLF are:

```
vpa(id.ident,7)
```

ans =

```
    372.6797
    852.5096
    1370.562

    0.0007898458
    0.001788817
    0.002862365

    0.8502998
    1.0
    0.6535707

    1.0
    0.2924284
    -1.0

    0.9379627
    -0.8496108
    0.3439804
```

Let's take a look at the error, in percentage, for ω_n and ζ_n wrt to the expected data:

```
Dw_n = 100*(id.ident(1,:)-modalParams.hammerMISO.fn0')./modalParams.hammerMISO.fn0';
vpa(Dw_n,5) % Natural Frequency error in %
```

```
ans = (-0.000088517 \ 0.00064676 \ -0.00059298)
```

```
Dz_n = 100*(id.ident(2,:)-modalParams.hammerMISO.dr0')./modalParams.hammerMISO.dr0';
vpa(Dz_n,5) % Modal Damping Ratios error in %
```

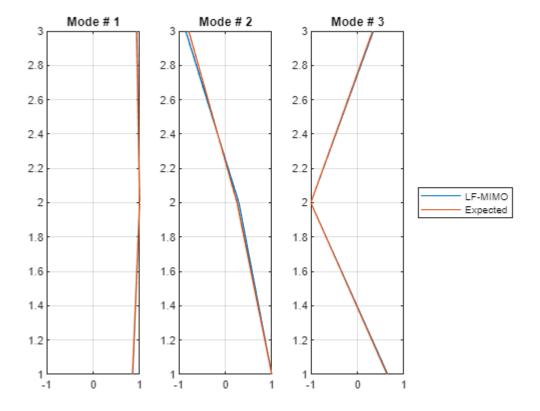
```
ans = (1.1554 \ 0.17987 \ -0.28662)
```

Now lets plot the iLF identified ϕ_n versus those expected:

```
figure
t,tiledlayout(1,3)
```

t = 3.9684

```
for i = 1:3
    nexttile
    lf = plot(id.ident(3:end,i),1:3);
    hold on
    ef = plot(ms0(:,i),1:3);
    hold off
    xlim([-1 1]);
    ylim([1 3]);
    title('Mode # '+string(i));
    grid on
end
    legend([lf,ef],{'LF-MIMO','Expected'},Location="eastoutside")
```



We can graphically see they are very similar. Let's confirm this by calculating the Modal Assurance Criterion (MAC) between the expected ϕ_n and those identified via iLF:

```
vpa(round(compute_mac(ms0,id.ident(3:end,:)),3),3)
```

ans =

$$\begin{pmatrix} 1.0 & 0.026 & 0.004 \\ 0.031 & 0.998 & 0.007 \\ 0.006 & 0.002 & 1.0 \end{pmatrix}$$

Hence, the great coherence between the expected and iLF-identified ϕ_n is confirmed as the values on the MAC matrix diagonal are, or very close to, 1.

Conclusions

This example shows the accuracy of the iLF on a simple MIMO system. For the results obtained for more complex system the reader is reffered to the article accompained by this tutorial [1].

References

- [1] G. Dessena and M. Civera, 'Improved Tangential Interpolation-based Multi-input Multi-output Modal Analysis of a Full Aircraft', European Journal of Mechanics A/Solids, vol. 109. Elsevier BV, p. 105495, Jan. 2025. doi: 10.1016/j.euromechsol.2024.105495.
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- [3] G. Dessena, M. Civera, L. Zanotti Fragonara, D. I. Ignatyev, and J. F. Whidborne, 'A Loewner-Based System Identification and Structural Health Monitoring Approach for Mechanical Systems', Structural Control and Health Monitoring, vol. 2023. Hindawi Limited, pp. 1–22, Apr. 18, 2023. doi: 10.1155/2023/1891062.
- [4] G. Dessena, M. Civera, D. I. Ignatyev, J. F. Whidborne, L. Zanotti Fragonara, and B. Chiaia, 'The Accuracy and Computational Efficiency of the Loewner Framework for the System Identification of Mechanical Systems', Aerospace, vol. 10, no. 6. MDPI AG, p. 571, Jun. 20, 2023. doi: 10.3390/aerospace10060571.
- [5] G. Dessena, M. Civera, A. Pontillo, D. I. Ignatyev, J. F. Whidborne, and L. Zanotti Fragonara, 'Noise-robust Modal Parameter Identification and Damage Assessment for Aero-structures', Aircraft Engineering and Aerospace Technology, Aug. 2024. doi: 10.1108/AEAT-06-2024-0178.