EE 4323 – Nonlinear Control Systems

Module 10: Sinusoidal-Input Describing Function Methods – SIDF Analysis

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Topic Outline

- Introduction
- Basic Concepts
- SIDF Calculations
- Qualitative Behavior of SIDFs; Heuristics
- Classical Harmonic Balance (Limit Cycle Conditions)
- Classical SIDF I/O Analysis ("Transfer Functions")
- Examples
- Systems with Multiple Nonlinearities
- Modern Algebraic SIDF Methods
- Examples

References:

- D. P. Atherton, *Nonlinear Control Engineering*, Van Nostrand, 1975 (Reprinted as Student Edition, 1982).
- A. Gelb & W. Vander Velde, Multiple-Input DF's and Nonlinear System Design, McGraw-Hill, 1968.
- J. H. Taylor, *Describing Functions*, an article in the *Electrical Engineering Encyclopedia*, John Wiley & Sons, Inc., New York, 1999.

Introduction

- Problem to be addressed: Analyzing Periodic Phenomena
 - Nonlinear oscillations (limit cycles)
 - Response of a nonlinear system to sinusoidal inputs
- Considerations:
 - Simulation is often too time-consuming and cumbersome, especially for parametric trade-off studies.
 - There are situations in which simulation is almost useless for studying periodic behavior (e.g., flight control systems)
 - It's the only method that handles high-order systems and/or systems with multiple nonlinearities with ease

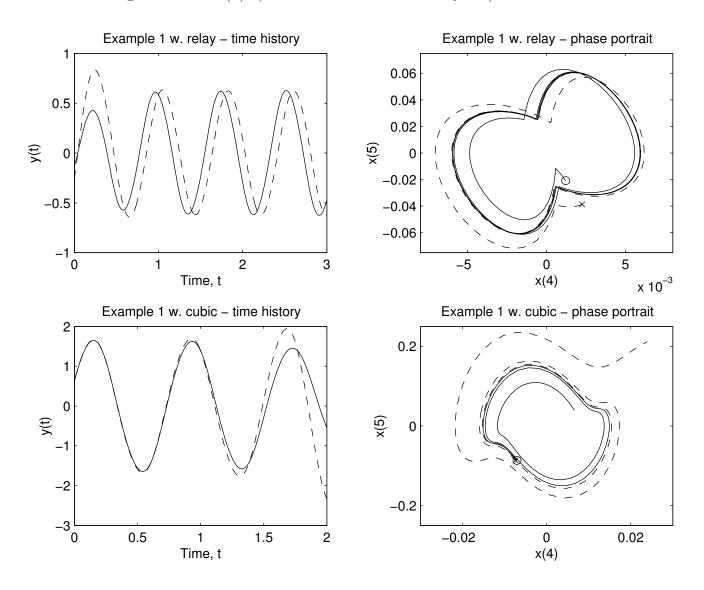
Importance of Periodic Effects

- A limit cycle may be **desired**, with a specified frequency and amplitude can you design the system? This is the only practical way to build an oscillator . . .
- A limit cycle may be **unwanted but unavoidable** is it small enough or slow enough to be acceptable?
- An unstable limit cycle is a **stability boundary** is it <u>large</u> enough?
- A nonlinear system may be driven by **sinusoidal inputs** how will it respond?
 - SIDF I/O models for different amplitudes \rightarrow diagnosis
 - SIDF I/O models exhibit interesting phenomena, e.g.,
 "jump resonance"
 - SIDF I/O models form an excellent basis for control system design – later

Definition of Limit Cycles

A <u>simple limit cycle</u> is a periodic trajectory in the state space, $x^*(t+T) = x^*(t)$, $\forall t$ where T is the period, such that all nearby trajectories:

- asymptotically approach $x^*(t)$ (a **stable** limit cycle) **or**
- diverge from $x^*(t)$ (an **unstable** limit cycle)



Note that **limit cycles** are not generally **ellipses**

Definition of Sinuoidal-Input Response



- Input: $u(t) = u_0 + a\cos(\omega t)$
- Output: **may be** periodic; to be safe, run a simulation:

$$y = \sum_{k=0}^{\infty} b_k \cos(k\omega t + \psi_k)$$

• "Transfer function" for the fundamental component:

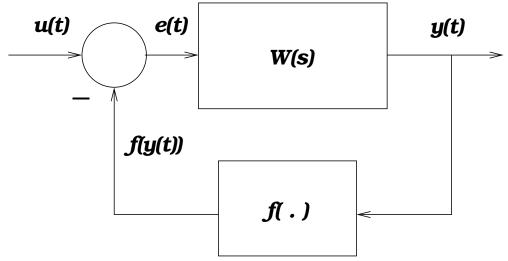
$$G(j\omega; u_0, a) = \frac{b_1}{a} \exp(j\psi_1) \tag{1}$$

• Operating point ("DC level"): $b_0(u_0, a)$

Note that the "transfer function" and operating point are coupled Hereafter we will call $G(j\omega; u_0, a)$ an **SIDF Input/Output Model** (SIDF I/O Model)

Basic System Models

Classical Case:



$$Y(s) \triangleq \mathcal{L}(y(t)) = \frac{p(s)}{q(s)} E(s)$$

$$\triangleq W(s) \mathcal{L}(e(t))$$

$$e(t) = u(t) - f(y(t))$$
(2)

Multivariable Case:

$$\dot{x} = f(x, u(t))$$

$$y(t) = h(x, u(t))$$
(3)

For **limit cycle analysis:** $u(t) = u_0$

For **forced response:** $u(t) = u_0 + \text{Re} \left[a \exp(j\omega t) \right]$

Basic System Model in MATLAB

- Given: $W(s) = 2/(s^2 + 3s + 7)$ and $f(y) = 4y^3$
- In ODE form: $\ddot{y} + 3\dot{y} + 7y + 8y^3 = 2u(t)$
- In state-space form (one realization): $x^T = [y \ \dot{y}]$

$$\dot{x_1} = x_2
\dot{x_2} = -7x_1 - 3x_2 - 8x^3 + 2u(t)$$

• In MATLAB:

```
function xdot = basic(t,x)
% Example in controllable canonical form:
% JH Taylor - 9 July 2002

num = 2; den = [ 1 3 7 ];
u = 3.5 * sin(10*t);
xdot(1) = x(2);
xdot(2) = - den(3)*x(1) - den(2)*x(2) + num*(u - 4*x(1)^3);
xdot = xdot(:);
```

• To run a simulation:

```
tspan = [ 0 6*pi/10 ]; % three cycles of sin(10*t)
x0 = [ 1.2 -3.4 ]; % arbitrary initial condition
[t,x] = ode45('basic',tspan,x0); % model is in basic.m
plot(t,x(:,1)); % plot first state only
```

Basic Idea of the Describing Function Method

- Knowledge of signal **form** and **amplitude** is essential in understanding the behavior of a nonlinear system
- Linear system approaches are the most powerful tools we have for analysis
- Replacing nonlinearities with **signal-dependent linear gains** ("**describing functions**") provides the best way to take advantage of linear system approaches to understand the behavior of a nonlinear system; this process is called "quasilinearization"
- You will see examples that use the *machinery* of Nyquist plots and root locus plots, ... but the **underlying theory** is entirely different. The Routh-Hurwitz technique can also be exploited, but it is difficult compared to the others

Classical Definition of a SIDF

- Given: a specific nonlinearity f(v) and an input signal form, $v(t) = v_0 + \text{Re}(a \exp(j\omega t))$
- Find: the quasilinear model

$$f(v) \stackrel{\sim}{=} f_0(v_0, a) + N(v_0, a) \cdot a \exp(j\omega t) \tag{4}$$

such that **mean square approximation error** is minimized

- Method 1: $f_0(v_0, a)$ and $N(v_0, a)$ are determined by **Fourier** analysis (constant plus first harmonic terms)
- Method 2: $f_0(v_0, a)$ and $N(v_0, a)$ are determined by using **trigonometric identities** (for power-law and product-type nonlinearities

Calculating SIDFs – Piece-Wise-Linear Case

Ideal relay: $f(y) = D \cdot \text{sgn}(y)$ where we assume no DC level, $y(t) = a \cos(\omega t)$

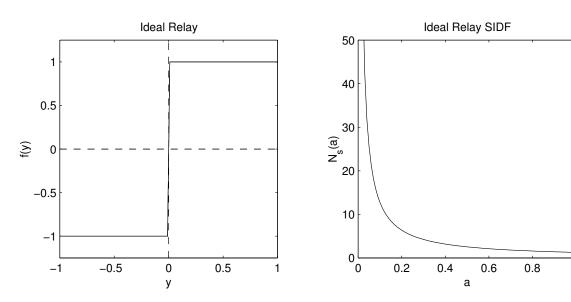
Set up and evaluate the integral for the first Fourier coefficient divided by a as follows:

$$N_{s}(a) = \frac{1}{\pi a} \int_{0}^{2\pi} f(a \cos(x)) \cdot \cos(x) dx$$

$$= \frac{4D}{\pi a} \int_{0}^{\pi/2} \cos(x) dx \quad \text{(by symmetry)}$$

$$= \frac{4D}{\pi a}$$

$$= \frac{4D}{\pi a}$$
(5)



This makes good, intuitive sense; this SIDF was used to generate the stable limit cycle on slide 5

Calculating SIDFs – Trig. Identity Case

Cubic nonlinearity: $f(y) = K y^3(t)$; again, assume $y(t) = a\cos(\omega t)$

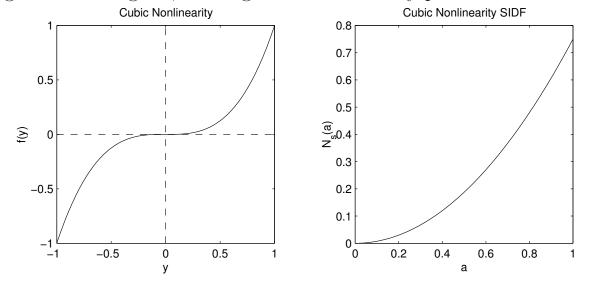
Directly write the Fourier expansion using trigonometric identities:

$$f(a\cos(\omega t)) = K[a\cos(\omega t)]^{3}$$

$$= Ka^{3}[\frac{3}{4}\cos(\omega t) + \frac{1}{4}\cos(3\omega t)]$$

$$\stackrel{\sim}{=} \frac{3Ka^{2}}{4} \cdot a\cos(\omega t)$$
(6)

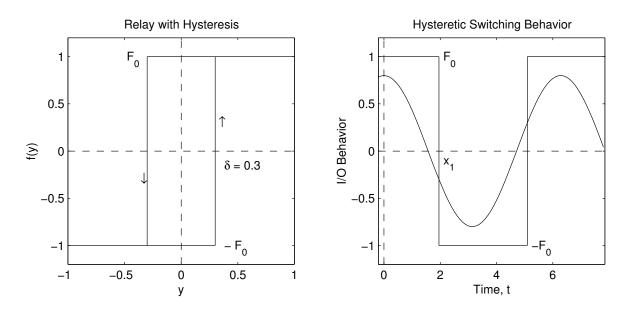
so $N_s(a) = 3 K a^2/4$. Trigonometric identities are a shortcut to using Fourier integrals; use trig. identities for any power-law element.



This also makes good, intuitive sense; this SIDF was used to generate the unstable limit cycle on slide 5

Calculating SIDFs – Multi-valued Case

Setting up the Fourier integrals requires care:



$$N(a) = \frac{1}{\pi a} \int_0^{2\pi} f(a \cos(x)) \cdot \exp(-jx) dx$$

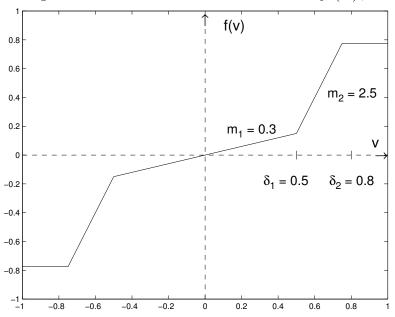
$$= \frac{2F_0}{\pi a} \left\{ \int_0^{x_1} \exp(-jx) dx - \int_{x_1}^{\pi} \exp(-jx) dx \right\}$$
 (by symmetry)
where $x_1 = \cos^{-1}(-\delta/a)$;

$$= \begin{cases} \frac{4F_0}{\pi a} \left\{ \sqrt{1 - (\delta/a)^2} - j \, \delta/a \right\} & a > \delta \\ 0 & a \le \delta \end{cases}$$
 (7)

Note that $N(a) \triangleq 0$ if $a \leq \delta$... the relay **does not switch** \Rightarrow the output is not periodic; more importantly, **this SIDF is complex-valued**, because it causes **lag**

Calculating SIDFs – Details

Given: a typical piecewise linear characteristic f(v),



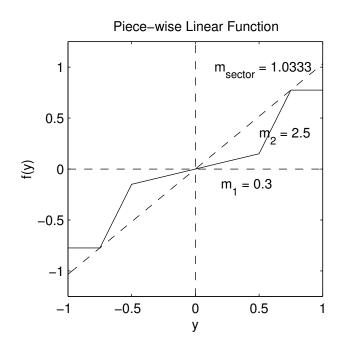
Set up the Fourier integral for this function for $a > \delta_2$:

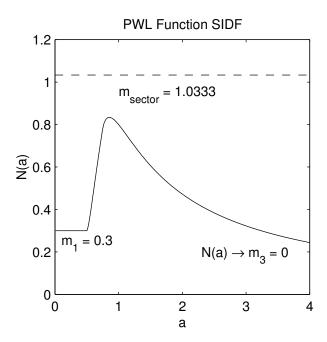
- Integrate from 0 to $\pi/2$, by symmetry, and break the integral into three intervals, $[0, x_1]$, $[x_1, x_2]$, $[x_2, \pi/2]$ where $x_1 = cos^{-1}(\delta_2/a)$ and $x_2 = cos^{-1}(\delta_1/a)$
- Be sure to use the correct function for f(v) in each interval, $f(v) = m_1 v$ in interval 3, $f(v) = m_2 (v \delta_1) + m_1 \delta_1$ in interval 2, $f(v) = m_2 \delta_2 (m_2 m_1) \delta_1 \triangleq D_0$ in interval 1
- Therefore the complete SIDF Fourier integral is:

$$N(a) = \frac{4}{\pi a} \left[\int_0^{x_1} D_0 \cos(x) dx + \int_{x_1}^{x_2} [m_2 a \cos(x) - (m_2 - m_1) \delta_1] \cos(x) dx + \int_{x_2}^{\pi/2} m_1 a \cos^2(x) dx \right]$$

Qualitative Behavior of SIDFs

Considering the same nonlinearity f(v) –

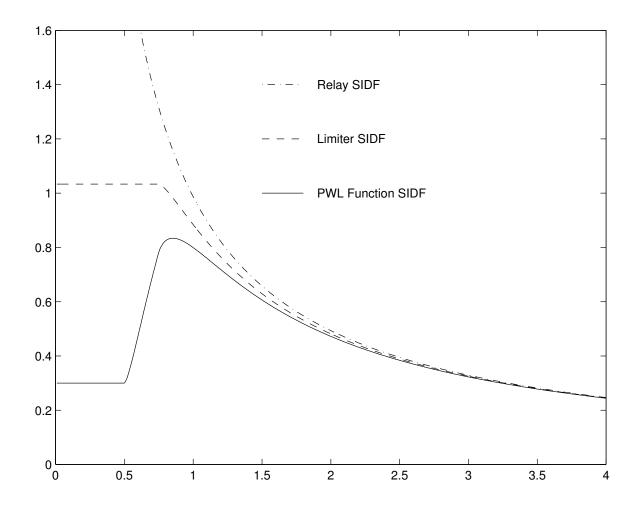




- For small signals $N(a) \stackrel{\sim}{=} N(0) = [df/dv]_{v=0} = m_1$ (if the derivative exists)
- The SIDF cannot lie outside the slopes of the enclosing sector
- The SIDF is always continuous, even though the nonlinearity derivative is discontinuous
- The SIDF always approaches the ultimate slope of the nonlinearity as $a \to \infty$ (zero for this example)

Qualitative Behavior of SIDFs (Cont'd)

ullet For large signals (a>2) the "details" near the origin do not make much difference



Conclusion: In many cases we don't need to evaluate N(a) exactly for a qualitative analysis . . .

Calculating SIDFs in MATLAB

• First, define the basic "saturation function" used in calculating SIDFs for piece-wise-linear functions:

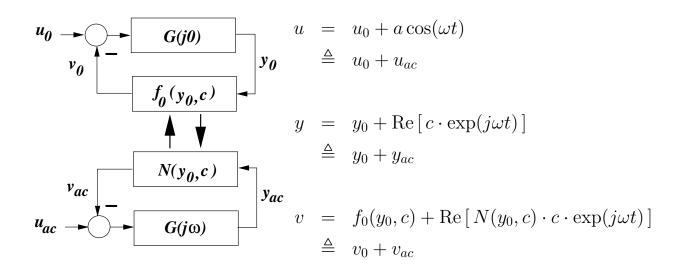
$$f_{\text{sat}} = \begin{cases} \operatorname{sign}(x), & |x| \ge 1\\ 2\left[\sin^{-1}(x) + x\sqrt{1 - x^2}\right]/\pi, & |x| < 1 \end{cases}$$
 (8)

- The SIDF for a general limiter is $N_{LIM}(a) = m f_{\rm Sat}(\delta/a)$
- The SIDF for the piece-wise-linear example is $N_{PWL}(a) = m_1 f_{\text{Sat}}(\delta_1/a) + m_2 \left[f_{\text{Sat}}(\delta_2/a) f_{\text{Sat}}(\delta_1/a) \right]$
- Therefore the previous plots are obtained as follows:

```
D = m1*d1 + m2*(d2 - d1); m_sect = D/d2;
  av = 0.01:0.01:4.0;
  for i = 1:length(av);
    DFqual(i) = m1*f_sat(d1/av(i)) + m2*(f_sat(d2/av(i))-f_sat(d1/av(i)));
    DFlim(i) = m_sect*f_sat(d2/av(i)); % limiter
    DFrel(i) = 4*D / (pi*av(i)); % relay
  end
  plot(av,DFqual,av,DFlim,'--',av,DFrel,'-.');
  axis([0 4 0 1.6]);
where:
  function f_sat = f_gvdv(x)
  % saturation function "f" for calculating SIDFs for PWL functions
  % Gelb & Vander Velde, Appendix B, p. 519
  % JH Taylor - 18 June 2002
  if abs(x) >= 1,
    fdf = sign(x);
  else
    fdf = 2*(asin(x) + x*sqrt(1 - x*x))/pi;
  end
```

Harmonic Balance – Limit Cycle Conditions

1. Classical Case:



DC Harmonic Balance: $y_0 = G(j0)[u_0 - f_0(y_0, c)]$... must be solved simultaneously with an AC harmonic balance relation (below) to obtain y_0 and c

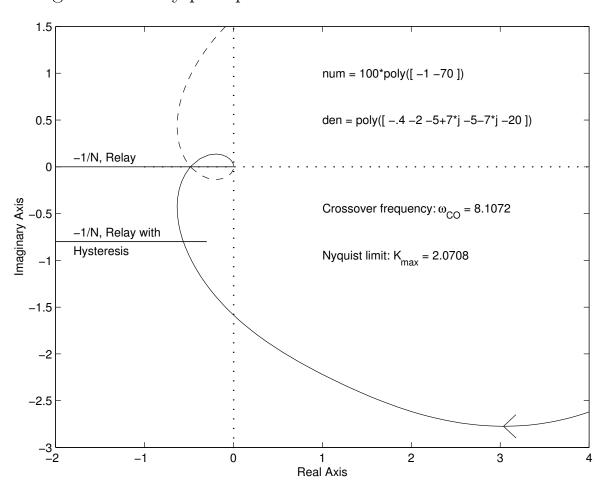
AC Harmonic Balance:

- Limit Cycles: a = 0; $G(j\omega) \cdot N(y_0, c) = -1$ must be satisfied for some $\{y_0, c, \omega\}$ for limit cycle prediction
- Forced Response: $a \neq 0$; $c = \frac{G(j\omega)}{1+N(y_0,c)\cdot G(j\omega)} \cdot a$; solve for $c(j\omega; u_0, a)$ then $N(y_0, c)$ to obtain the overall "transfer function" $W(j\omega; u_0, a) = \frac{G(j\omega)}{1+N(y_0,c)\cdot G(j\omega)}$ (the frequency response from u_{ac} to y_{ac})

Classical Limit Cycle Analysis Using a Nyquist-plot Approach

Given
$$G(s) = \frac{s^2 + 71s + 70}{10s^5 + 32.4s^4 + 3468s^3 + 21616s^2 + 37712s + 11840}$$
:

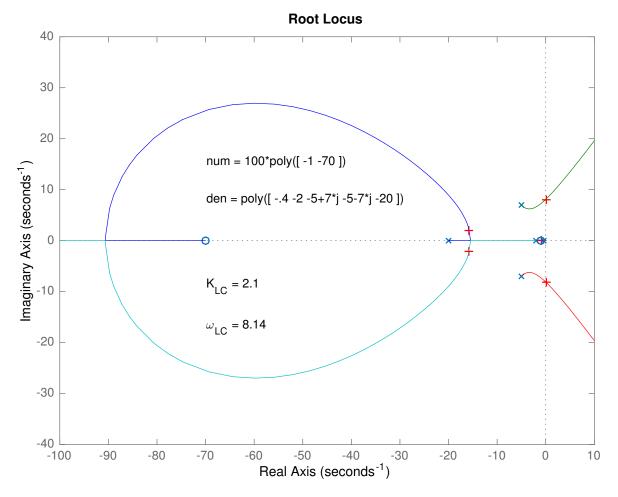
The condition $G(j\omega) \cdot N = -1$ (or $G(j\omega) = -1/N$) is easily investigated on a Nyquist plot:



Note: This is **not** the Nyquist test for stability!

Classical Limit Cycle Analysis Using a Root Locus-plot Approach

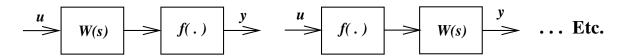
The condition that the root locus of $G(j\omega)$ has poles on the imaginary axis is easily investigated on a root locus plot:



```
clear; close all
den = poly([ -.4 -2 -5+7*j -5-7*j -20 ]);
num = 100*poly([ -1 -70 ]);
rlocus(num,den)
[K,poles] = rlocfind(num,den);
```

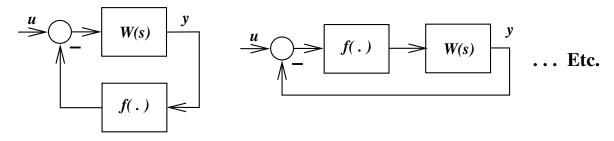
Limitations of SIDF Analysis

• Situations when SIDFs are exact:



The first harmonic of y is correct

• Situations when SIDFs are **not** exact:



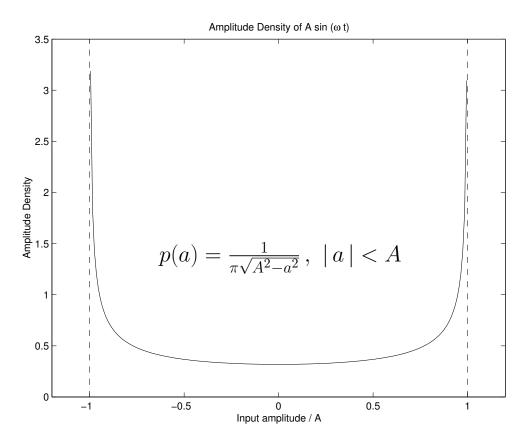
The first harmonic of y is approximate

- How to deal with inexact situations:
 - Consider the validity of the "low-pass filter hypothesis": The nonlinearity input is essentially sinusoidal due to the filtering of higher harmonics by $W(j\omega)$ if true, SIDF results should be accurate.
 - Consider how well-behaved the system nonlinearity is
 - Look at simulation results, assess the importance of higher harmonics (distortion)

Limitations of SIDF Analysis (Cont'd)

Except for multi-valued nonlinearities (hysteresis, backlash etc.) the DF is not dependent on the assumption of periodicity – only the **amplitude distribution** matters

- For a triangular ("saw-tooth") wave the DF is the same as that for a uniformly distributed random variable
- In many control applications the sine-wave distribution is a good approximation:



Limitations of SIDF Analysis (Cont'd)

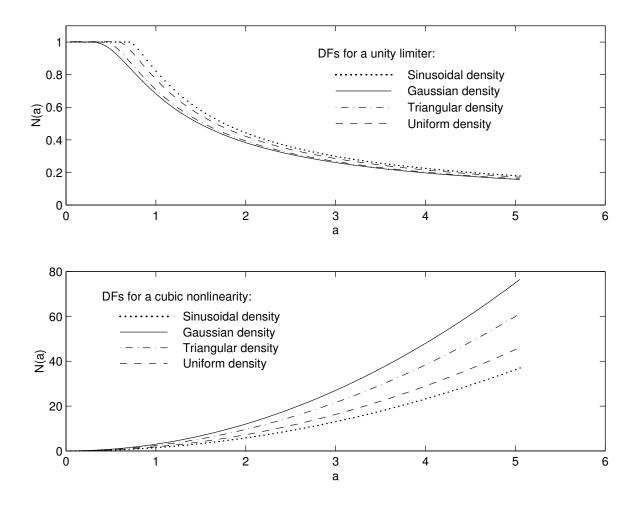
• The basic DF definition for f(v) where v has a bias b is:

$$F_0(\sigma, b) = \int_{-\infty}^{\infty} f(b+z) p(z) dz$$
 (9)

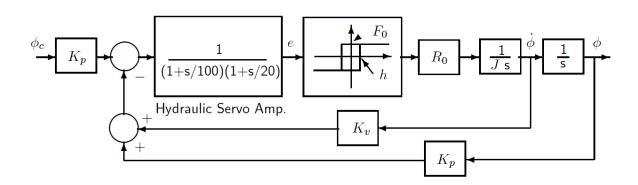
$$N_z(\sigma, b) = \frac{1}{\sigma^2} \int_{-\infty}^{\infty} z f(b+z) p(z) dz \qquad (10)$$

where σ is the generalized input signal amplitude, $\sigma = \sqrt{\mathrm{E}[(v-b)^2]}$ which for a sinusoidal signal is $\sigma_s = a/\sqrt{2}$

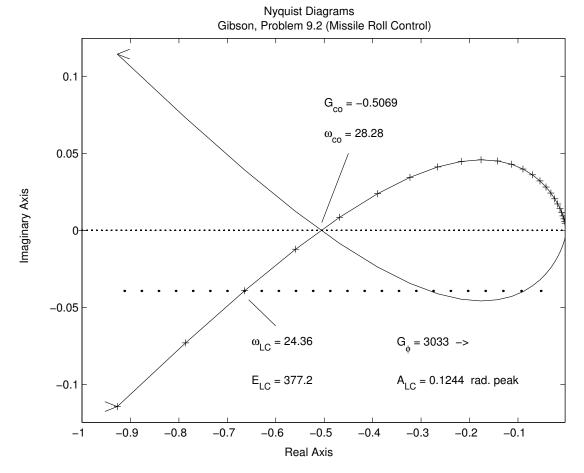
• For many nonlinearities the DF is not particularly sensitive to the amplitude distribution:



Example: Limit Cycle Analysis, Missile Roll-Control Loop



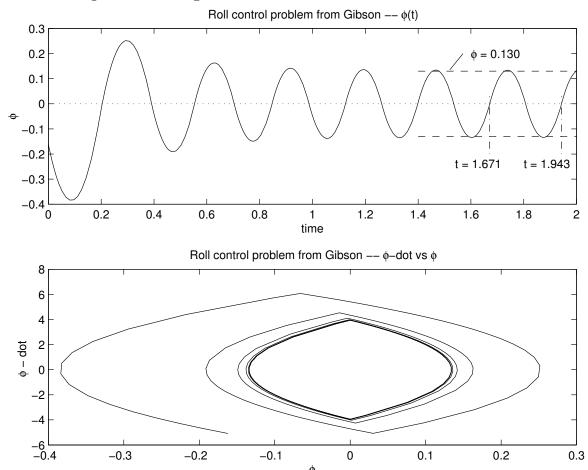
The relay-with-hysteresis implements "bang-bang" control, a simple robust method that is easy to implement



Note: $G_{\phi} = \text{gain from } e \text{ to } \phi$

Limit Cycle Verification

Simulation provides a good verification:



$$T = 0.272 \text{ sec} \rightarrow \omega_{LC} = 23.1 \text{ rad/sec}$$

Compare with the SIDf result on the previus slide:

$$T = 0.258 \text{ sec} \rightarrow \omega_{LC} = 24.36 \text{ rad/sec}$$

The SIDF spproach is usually this accurate – excellent!

Harmonic Balance "Transfer Functions"

Two methods for generating the SIDF I/O model $G(j\omega; u_0, a)$:

- 1. Analytic approach: solve the AC Harmonic Balance equation for $c(j\omega; u_0, a)$, divide by a
 - (a) Advantage: you can tell, for example, when solutions do not exist
 - (b) <u>Disadvantage</u>: it's difficult to carry out if the nonlinear system is at all complicated
- 2. <u>Simulation approach</u>: develop a simulation model for the non-linear dynamic system with a sinusoidal input, simulate to obtain the steady-state response, perform Fourier analysis of the result
 - (a) Advantages: No need to assume that the input to each nonlinearity is sinusoidal, the number of system states and nonlinearities is relatively unimportant
 - (b) <u>Disadvantages</u>: May be quite time consuming, may be difficult to interpret the results

Harmonic Balance "Transfer Function" – Classical Duffing's Equation

Duffing's Equation: $\ddot{x} + 2\zeta \dot{x} + x + x^3 = a\cos(\omega t)$

This represents, for example, a normalized mass-spring-damper system with a hardening spring; in the standard form $W(s) = 1/(s^2 + 2\zeta s + 1)$, $u(t) = a\cos(\omega t)$ and $f(\cdot) = x^3$

Let b be the <u>amplitude</u> of the fundamental component of x; then quasilinearize Duffing's equation to obtain:

$$b^{2}\left[(1+\frac{3}{4}b^{2}-\omega^{2})^{2}+(2\zeta\omega)^{2}\right]=a^{2}$$

or, if we let $B = b^2$,

$$B\left[(1 + \frac{3}{4}B - \omega^2)^2 + (2\zeta\omega)^2 \right] = a^2$$

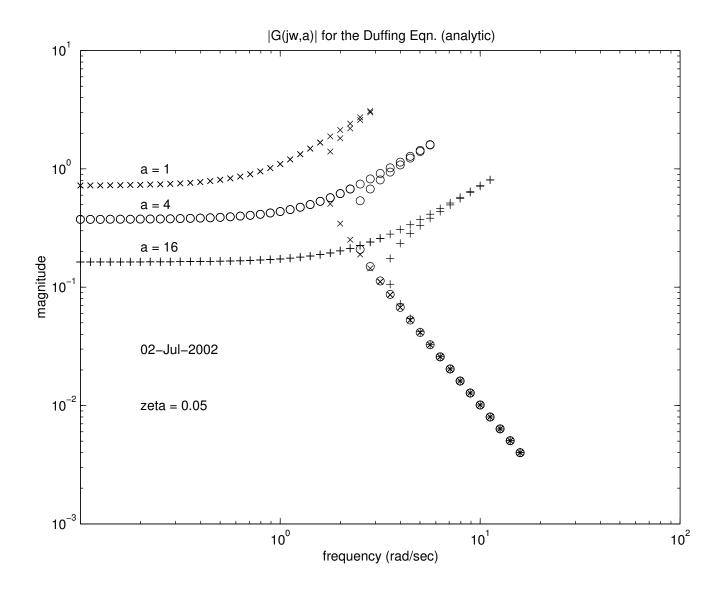
finally,

$$\frac{9}{16}B^3 + \frac{3}{2}(1-\omega^2)B^2 + \left[(1-\omega^2)^2 + (2\zeta\omega)^2\right]B - a^2 = 0 \quad (11)$$

The last simple polynomial equation may have 1 or 3 real roots, depending on a and ω :

Duffing's Equation "Transfer Function"

The results for several values of a are as follows:



Here we see a **jump resonance phenomenon**, which can be a real surprise to an experimentalist!

Solving the Duffing Problem in MATLAB

• First, define the polynomial in Eqn. 11 multiplied by 16/3:

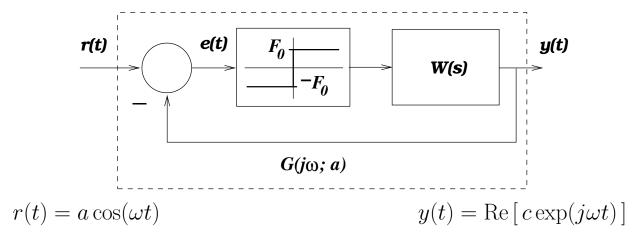
```
function soln = duff_poly(a,w,zeta)
% polynomial to be solved for Duffing's Equation
beta = 1 - w*w; gamma = 2*zeta*w; K = 16/3;
C(1) = 3; C(2) = 8*beta; C(3) = K*(beta^2 + gamma^2);
C(4) = - K*a*a;
soln = sqrt(roots(C)./(a*a));
```

• Now, set up loops for 3 amplitudes and 45 frequencies:

```
zeta = 0.050;
for jj=1:3 %% amplitude loop
  a = 4^{(jj-1)} %% a = 1, 4, 16
  av(jj) = a;
  for ii=1:45 %% frequency loop
    w = 10^{((ii-21)/20)} %% w_min = 0.1, w_max = 10
    wv(ii) = w;
    G = duff_poly(a,w,zeta);
    % discard any complex conjugate
    if imag(G(1)) = 0 \mid imag(G(2)) = 0,
      for iii=1:3
        if imag(G(iii)) == 0, RG = G(iii); end
      G(1) = RG; G(2) = RG; G(3) = RG; % only one real root exists
    GM(ii,3*jj-2) = G(1); GM(ii,3*jj-1) = G(2); GM(ii,3*jj) = G(3);
  end % frequency loop
end % amplitude loop
%% plotting
loglog(wv,GM(:,1),'x',wv,GM(:,2),'x',wv,GM(:,3),'x', ...
      wv,GM(:,4),'o',wv,GM(:,5),'o',wv,GM(:,6),'o', ...
      wv,GM(:,7),'+',wv,GM(:,8),'+',wv,GM(:,9),'+');
title('|G(jw,a)| for the Duffing Eqn. (analytic)')
xlabel('frequency (rad/sec)');
ylabel('magnitude');
```

Harmonic Balance "Transfer Functions" (Cont'd)

Closed-loop system with relay:



Harmonic Balance Relation:

$$c = (a - c) \cdot \frac{4F_0}{\pi |a - c|} W(j\omega)$$

• Magnitude part:

$$M(j\omega) \triangleq |W(j\omega)|;$$

 $|G(j\omega;a)| \triangleq \frac{|c|}{a} = \frac{4F_0}{\pi a} M(j\omega)$

• Phase part:

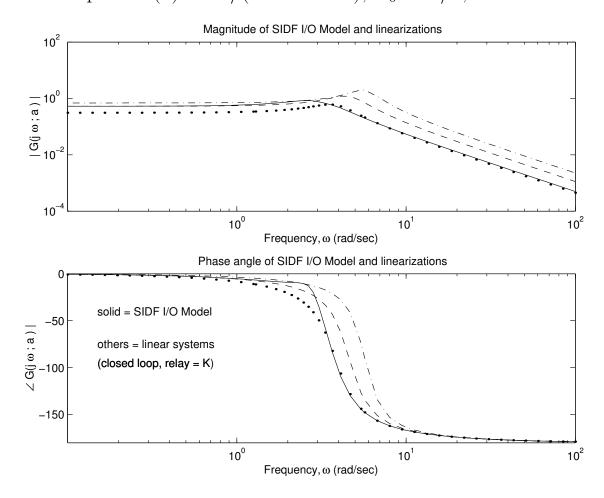
$$\psi \triangleq \angle W(j\omega) ;$$

$$\angle G(j\omega; a) = \psi - \sin^{-1} \left(\frac{4F_0}{\pi a} M(j\omega) \sin(\psi) \right)$$

Harmonic Balance "Transfer Functions" (Cont'd)

Closed-loop system with relay (cont'd)

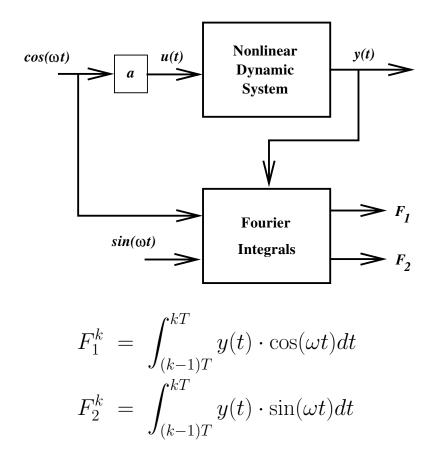
- The magnitude relation is quite straightforward (but it appears that the feedback disappears!)
- The phase relation can only be met if the input amplitude a is large enough that the argument of \sin^{-1} is less than one at all frequencies
- Example: $W(s) = 45/(s^2 + 2s + 9), F_0 = \pi/2, a = 18 \rightarrow$



Note that the SIDF transfer functions has characteristics that cannot be duplicated by any linear $G(j\omega)$

SIDF I/O Models by Simulation

The most efficient approach is to simulate and perform Fourier analysis simultaneously:



from which we obtain:

$$\operatorname{Re} G(j\omega; u_0, a) = \frac{\omega}{\pi a} F_1^k$$

$$\operatorname{Im} G(j\omega; u_0, a) = -\frac{\omega}{\pi a} F_2^k$$

Integrate for k cycles where k is sufficiently large that the magnitude and phase of $G(j\omega; u_0, a)$ have converged to your satisfaction.

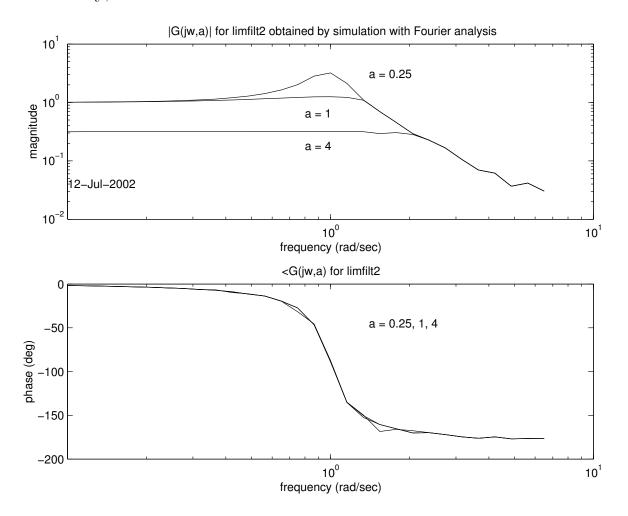
SIDF I/O Model by Simulation in MATLAB

1. Add the Fourier integral states to your model:

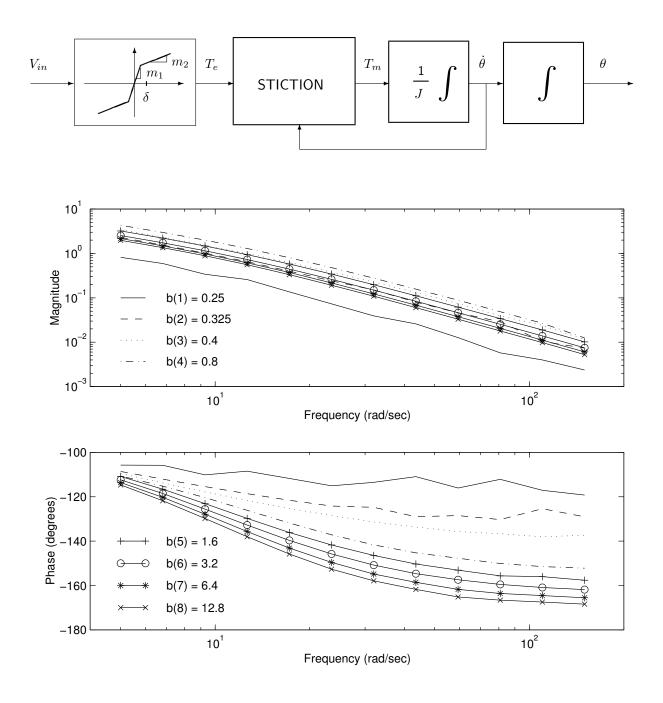
```
function xdot = lim_filt2(t,x)
     Second-order linear model with limiter; model is
  % augmented with Fourier integrals, to obtain G(jw,a)
  % JH Taylor, 10 July 2002
  %
  zeta = 0.15; global Ampl Freq
  u = Ampl*sin(Freq*t);
  xdot(1) = x(2);
  xdot(2) = u - x(1) - 2*zeta*x(2);
  %% define Y and set up the Fourier integrals:
  if abs(x(1)) < 1
     y = x(1);
  else
     y = sign(x(1));
  end
  xdot(3) = y*sin(Freq*t);
  xdot(4) = y*cos(Freq*t);
  xdot = xdot(:); %% end of model lim_filt2
2. Run a simulation to steady state and extract G(j\omega):
  function [mag,phase] = ggen(Model,MAGTOL,PHASETOL)
  %% ggen(model, MAGTOL, PHASETOL) returns the magnitude
  \%\% and phase of an ODE model defined in the file 'Model'.m
  %% JH Taylor - University of New Brunswick - 7 July 2002
  % Initialize:
  global Ampl Freq Xdim;
  k = 0; T = 2*pi/Freq; tspan = [ 0 T ]; x0 = zeros(Xdim, 1);
  [t,x] = ode45(Model,tspan,x0);
  [nrows, ncols] = size(x);
  xf = x(nrows,:);
  mag0 = Freq/(pi*Ampl)*abs(xf(ncols-1)+j*xf(ncols));
```

```
phase0 = atan2(xf(ncols),xf(ncols-1));
  % Simulate cycle-by-cycle until convergence obtained:
  while (k \ge 0)
     k = k+1:
     x0 = xf; % initial condition from last cycle
     x0(ncols-1) = 0; % reset the Fourier states
     x0(ncols) = 0;
     [t,x] = ode45(Model,tspan,x0);
     [nrows,ncols] = size(x);
     xf = x(nrows,:);
     mag = Freq/(pi*Ampl)*abs(xf(ncols-1)+j*xf(ncols));
     phase = atan2(xf(ncols),xf(ncols-1));
     magdiff = abs(20*log10(mag/mag0));
     phasediff = (180/pi)*abs(phase-phase0);
     if ((magdiff >= MAGTOL) | (phasediff >= PHASETOL))
        mag0 = mag;
        phase0 = phase;
     else
        k = -1;
     end
  end;
3. Here is the main executive:
  %% script for generating a set of G(jw,a) for model "mdl"
  %%
      JH Taylor 5 July 2002
  global Ampl Freq Xdim; dpr = 180/pi; % degrees/radian
  mtol = 1; % magnitude tolerance (dB)
  ptol = 5; % phase tolerance (deg)
  mdl = 'lim_filt2' % model = lim_filt2.m (2nd order filter + limiter)
  Xdim = 4; % # states, **including Fourier integrals**
  %
  % amplitude loop
  for jj=1:3
     Ampl = 4^{(jj-2)} \% Ampl = .25, 1, 4
     av(jj) = Ampl;
```

4. Finally, here is the main result:



Example: SIDF I/O Model, Electromechanical System, by Simulation



Power of Classical SIDF Approach

When will SIDF limit cycle predictions be "good"?

- When -1/N(a) definitely cuts $G(j\omega)$ (not a near miss or near hit)
- When only one limit cycle is predicted (no "nesting")
- When $G(3j\omega_{LC})$ is far from -1/N(a) where ω_{LC} is the predicted limit cycle frequency

When will SIDF I/O models be "good"?

- When the nonlinear system is not highly resonant
- When higher harmonics are not dominant predicted limit cycle frequency

Modern SIDF Analysis

- Given: $\dot{x} = f(x, u)$ with $u(t) = u_0 + \text{Re} \left[a \exp(j\omega t) \right]$
- Assume: $x(t) \stackrel{\sim}{=} x_c + \text{Re} [b \exp(j\omega t)]$
- Quasilinearize the entire state-space system:

$$f(x, u) = f_B(u_0, a, x_c, b)$$

$$+ \operatorname{Re} \left[A_{DF}(u_0, a, x_c, b) \cdot b \exp(j\omega t) \right]$$

$$+ \operatorname{Re} \left[B_{DF}(u_0, a, x_c, b) \cdot a \exp(j\omega t) \right] \quad (12)$$

- Therefore DC harmonic balance is given by $0 = f_B(u_0, a, x_c, b)$
- ... and AC harmonic balance is given by:
 - Nonlinear Oscillations: a = 0, find $b \neq 0$ such that $[j\omega_{LC}I A_{DF}]^{-1}b = 0$ (" A_{DF} has pure imaginary eigenvalues and b is the corresponding eigenvector"), i.e., limit cycles are predicted if solutions b, ω_{LC} exist
 - Forced Response: $b = [j\omega I A_{DF}]^{-1}B_{DF} \cdot a$

SIDFs for Multivariable Functions

- Single-input nonlinearities f(v) are quasilinearized as before
- Multi-variable nonlinearities $f(v_1, v_2, ...)$ are more complicated; products and powers of states are easiest to do:

Given:
$$f(x) = x_1 x_2^2$$

$$= (x_{10} + \text{Re}[a_1 \exp(j\omega t)])(x_{20} + \text{Re}[a_2 \exp(j\omega t)])^2$$

$$= \dots$$

$$\stackrel{\sim}{=} [x_{10} x_{20}^2 + \frac{1}{2} x_{10} | a_2 |^2 + x_{20} a_1 \bullet a_2$$

$$+ [x_{20}^2 + \frac{1}{4} | a_2 |^2] \cdot x_{1,AC}$$

$$+ [2x_{10} x_{20} + \frac{1}{2} a_1 \bullet a_2] \cdot x_{2,AC}$$
(13)

(via trigonometric identities and eliminating higher harmonic terms), where \bullet denotes dot product, $a_1 \bullet a_2 = \operatorname{Re} a_1 \cdot \operatorname{Re} a_2 + \operatorname{Im} a_1 \cdot \operatorname{Im} a_2$

Handling multivariable functions represents a **significant generalization** over the classical approach

Multivariable Limit Cycle Example

The following second-order differential equation has been derived to describe the local behavior of solutions to a two-mode panel flutter model:

$$\ddot{\chi} + (\alpha + \chi^2)\dot{\chi} + (\beta + \chi^2)\chi = 0 \tag{14}$$

Heuristically, it is reasonable to predict that limit cycles may occur for negative α (so the second term provides damping that is negative for small values of χ but positive for large values). Observe also that there are three singularities if β is negative: $\chi_0 = 0, \pm \sqrt{-\beta}$. Making the usual choice of state vector, $x = [\chi \ \dot{\chi}]^T$, the corresponding state vector differential equation is

$$\dot{x} = \begin{bmatrix} \dot{\chi} \\ \ddot{\chi} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\beta & -\alpha \end{bmatrix} x - \begin{bmatrix} 0 \\ x_1^2 (x_1 + x_2) \end{bmatrix}$$
 (15)

The SIDF assumption for this system of equations is that

$$x_1 = \chi \stackrel{\sim}{=} \chi_c + a_1 \cos(\omega t)$$

 $x_2 = \dot{\chi} \stackrel{\sim}{=} -a_1 \omega \sin(\omega t)$

($x_2 = \dot{\chi}$ so x_2 has a center value of 0 and $a_2 = -j \omega a_1$)

Therefore, the combined nonlinearity in Eqn. 15 may be quasilinearized to obtain

$$x_{1}^{2}(x_{1} + x_{2}) = (\chi_{c} + a_{1}\cos(\omega t))^{2}(\chi_{c} + a_{1}\cos(\omega t) - a_{1}\omega\sin(\omega t))$$

$$\stackrel{\sim}{=} (\chi_{c}^{3} + \frac{3}{2}\chi_{c}a_{1}^{2}) + (3\chi_{c}^{2} + \frac{3}{4}a_{1}^{2})a_{1}\cos(\omega t)$$

$$+(\chi_{c}^{2} + \frac{1}{4}a_{1}^{2})(-a_{1}\omega\sin(\omega t))$$

$$\stackrel{\triangleq}{=} f_{B} + N_{1} \cdot a_{1}\cos(\omega t) + N_{2} \cdot (-a_{1}\omega\sin(\omega t))$$
(16)

This result is obtained by expanding the first expression using trigonometric identities and discarding all terms except the fundamental ones (k = 0, 1).

Therefore, we require that

$$F_{DF} = \begin{bmatrix} 0 \\ -\chi_c \left(\beta + \chi_c^2 + \frac{3}{2}a_1^2\right) \end{bmatrix} = 0$$
 (17)

$$A_{DF} = \begin{bmatrix} 0 & 1 \\ -(\beta + 3\chi_c^2 + \frac{3}{4}a_1^2) & -(\alpha + \chi_c^2 + \frac{1}{4}a_1^2) \end{bmatrix} \triangleq \begin{bmatrix} 0 & 1 \\ -\omega_{LC}^2 & 0 \end{bmatrix}$$
 (18)

(so A_{DF} has imaginary eigenvalues $\pm j \omega_{LC}$; again, the canonical form of A_{DF} ensures harmonic balance, not "pure imaginary eigenvalues").

Multivariable Limit Cycle Example (Cont'd)

Relation 17 shows two possibilities:

• Case 1: $\chi_c = 0$, in which case Eqn. 18 yields

$$a_1 = 2\sqrt{-\alpha}$$
, $\omega_{LC} = \sqrt{\beta - 3\alpha}$ (19)

The amplitude a_1 and frequency ω_{LC} must be real for limit cycles to exist. Thus, as conjectured, $\alpha < 0$ is required. for a LC to exist centered about the origin, and β must satisfy $\beta > 3 \alpha$, so β can take on any positive value but cannot be more negative than 3α .

• Case 2: $\chi_c = \pm \sqrt{(\beta - 6\alpha)/5} \triangleq \pm \chi_{c0}$, yielding

$$a_1 = 2\sqrt{(\alpha - \beta)/5}$$
, $\omega_{LC} = \sqrt{\beta - 3\alpha}$ (20)

For the two limit cycles in Case 2 to exist, centered at $\pm \chi_{c0}$, it is necessary that $3 \alpha < \beta < \alpha$, so again limit cycles cannot exist unless $\alpha < 0$. One additional constraint must be imposed: $|\chi_c| > a_1$ must hold, or the two limit cycles will "overlap"; this condition reduces the permitted range of β to $2 \alpha < \beta < \alpha$.

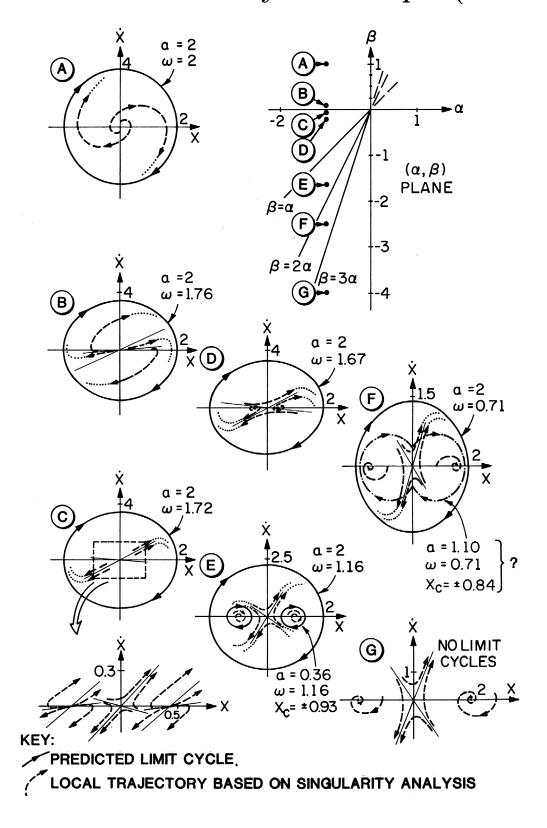
The stability of the Case 1 LC can be determined as follows: Take any $\epsilon > 0$ and perturb the LC amplitude to a slightly larger value, e.g., $a_1^2 = -4\alpha + \epsilon$. Substituting into Eqn. 18 yields

$$A_{DF} = \begin{bmatrix} 0 & 1\\ -(\beta - 3\alpha + \frac{3}{4}\epsilon) & -\frac{1}{4}\epsilon \end{bmatrix}$$
 (21)

which for $\epsilon > 0$ has "slightly stable eigenvalues". Thus a trajectory perturbed just outside the LC will decay, indicating that the Case 1 LC is stable. A similar analysis of the Case 2 LC is more complicated (since a perturbation in a_1 produces a shift in χ_c that must be considered), and thus is omitted.

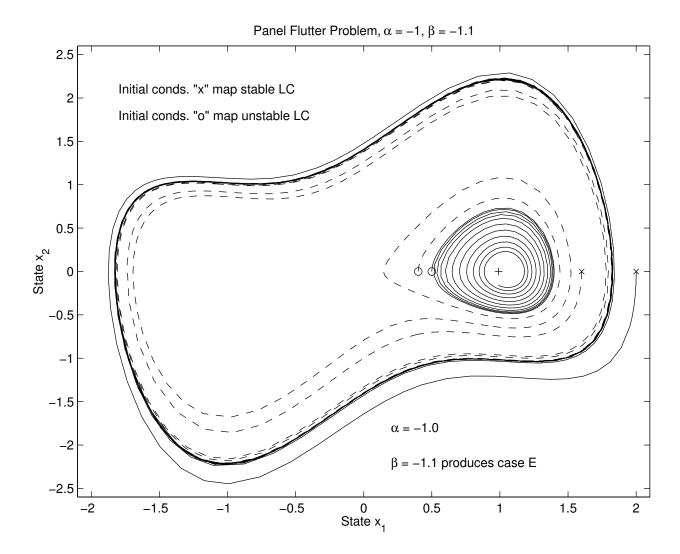
Based on the SIDF-based LC analysis outlined above, the behavior of the original system Eqn. 15 is portrayed for $\alpha = -1$ and seven values of β below:

Multivariable Limit Cycle Example (Cont'd)



Multivariable Limit Cycle Example (Cont'd)

Here is one simulation confirmation of these SIDF results, corresponding to Case E:



The outer limit cycle is clearly stable, while the inner limit cycle is unstable; the equilibrium inside the inner limit cycle is stable. There is another unstable limit cycle centered (approximately) at $x_{10} = -1$, $x_{20} = 0$

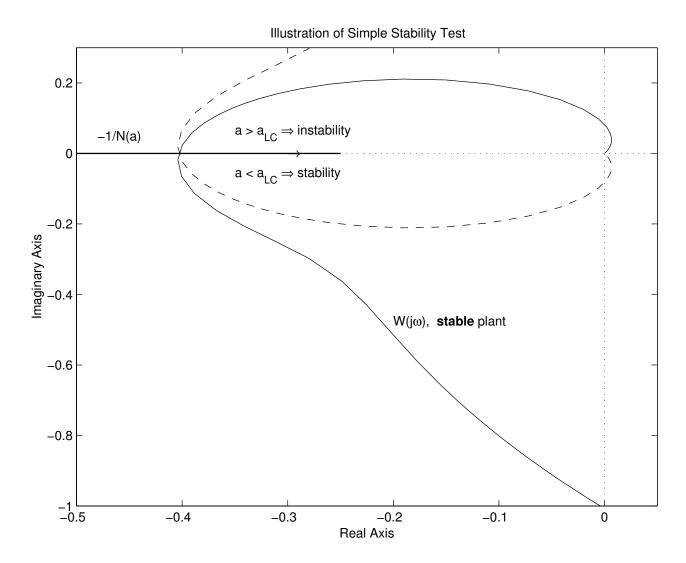
Limit Cycle Stability

The SIDF approach can also yield information on **limit cycle** stability – assume a limit cycle is predicted with amplitude a_{LC} , then:

- The limit cycle is **stable** if $a > a_{LC}$ moves the pure imaginary eigenvalues into the *left* half plane and $a < a_{LC}$ moves the pure imaginary eigenvalues into the *right* half plane
- The limit cycle is **unstable** if the converse is true
- Otherwise the limit cycle is **structurally unstable** (this is an uncommon "borderline" case)
- These conditions are easy to check in cases where there is no bias (DC level), otherwise the coupling between the center value and amplitude (y_c, a) must be taken into account

Limit Cycle Stability (Cont'd)

Here is a limit cycle stability test in the no bias case:



In other words: if -1/N(a) for $a > a_{LC}$ moves inside the RHP map of W(s) the limit cycle is unstable, and conversely.

Another test works if there is no bias and there is only one limit cycle predicted: The limit cycle is stable if the enclosed equilibrium is unstable, and conversely.

SIDF Methods: Conclusions

- SIDF techniques are very powerful for studying periodic behavior (nonlinear oscillations, forced response), even in high order and highly nonlinear dynamic system models, even where discontinuous and multi-valued functions exist
- One of the key uses of this approach is exploration:
 - Finding areas in parameter space where limit cycles exist and boundaries where bifurcations occur
 - Determining how a nonlinear system's response to sinusoidal inputs changes as model parameters change
- SIDF analysis and simulation are highly complementary; both have important roles to play