

# Variational Equation or Continuous Dependence on Initial Condition or Trajectory Sensitivity & Floquet Theory & Poincaré Map

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## 1. General idea of trajectory sensitivity

- Assume a solution  $\phi(t, t_0, x_0)$  to an IVP
- How will the orbit behave if we start very close to it?

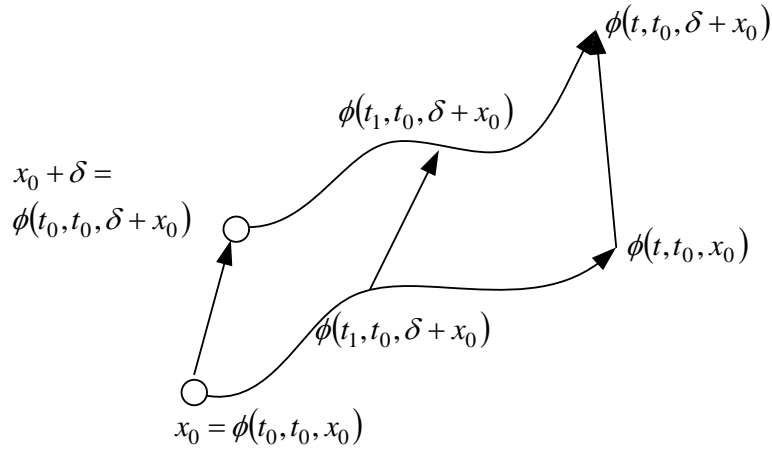


Figure 1

- We define:
  - $\delta = \phi(t_0, t_0, \delta + x_0) - \phi(t_0, t_0, x_0) = \Delta\phi(t_0, t_0, x_0)$
  - $\Delta\phi(t, t_0, x_0) = \phi(t, t_0, \delta + x_0) - \phi(t, t_0, x_0)$
- Now we have to express  $\Delta\phi(t, t_0, x_0)$  as a function of the original perturbation:

$$\begin{aligned} \phi(t, t_0, \delta + x_0) &\stackrel{TS}{=} \phi(t, t_0, x_0) + \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \delta \Leftrightarrow \\ \underbrace{\phi(t, t_0, \delta + x_0) - \phi(t, t_0, x_0)}_{\Delta\phi(t, t_0, x_0)} &= \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \delta \Leftrightarrow \end{aligned}$$

$$\boxed{\Delta\phi(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0)} \quad (1)$$

This is the variational equation.

For  $t = t_0$  we have that

$$\Delta\phi(t_0, t_0, x_0) = \frac{\partial\phi(t_0, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0) \Leftrightarrow \frac{\partial\phi(t_0, t_0, x_0)}{\partial x_0} = I_{n \times n}$$

So if we want to see the sensitivity to the ICs we need to find  $\frac{\partial \phi(t, t_0, x_0)}{\partial x_0}$

## 2. Homogeneous Linear Time Invariant System

If we have a linear system  $\dot{x}(t) = Ax(t)$  then the solution is:

$$\phi(t, t_0, x_0) = e^{A(t-t_0)}x(t_0). \text{ Then of course: } \boxed{\frac{\partial \phi(t, t_0, x_0)}{\partial x_0} = e^{A(t-t_0)}} \quad (2)$$

**Notice that through that last equation we can give another definition for the STM (in this exponential matrix): the partial derivative of the solution wrt the**

**initial condition:**  $\frac{\partial \phi(t, t_0, x_0)}{\partial x_0}$

$$\text{Thus: } \Delta \phi(t, t_0, x_0) = e^{A(t-t_0)} \Delta \phi(t_0, t_0, x_0)$$

## 3. Non - Homogeneous Linear Time Invariant System

If we have a linear system  $\dot{x}(t) = Ax(t) + BU(t)$  then the solution is:

$$\phi(t, t_0, x_0) = e^{A(t-t_0)}x(t_0) + \int_{t_0}^t e^{A(t-\tau)}BU(\tau)d\tau$$

$$\text{Then of course: } \boxed{\frac{\partial \phi(t, t_0, x_0)}{\partial x_0} = e^{A(t-t_0)}} \quad (3)$$

$$\text{Thus: } \Delta \phi(t, t_0, x_0) = e^{A(t-t_0)} \Delta \phi(t_0, t_0, x_0)$$

(see example 1 in the following page)

### Example 1 (LTI):

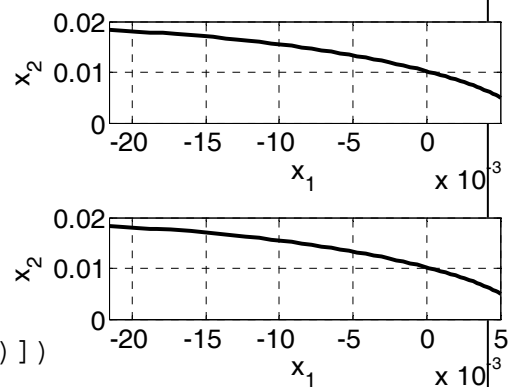
```
% initial
clc, clear, close
syms tau t; t0=0;
A=[1 -4; 2 3]; B=[1;1]; U=sin(t);Utau=sin(tau);

% nominal orbit
x0=[0.01;0.01];
xsol=expm(A*(t-t0))*x0+int(expm(A*(t-
tau))*B*Utau,tau,0,t);
cnt=1;
t2=0:0.01:0.5;
for t1=t2;
xsol1(cnt,1:2)=subs(xsol,t,t1);
cnt=cnt+1;
end

% perturbed orbit
x0=[0.01;0.01]+0.005;
xsol=expm(A*(t-t0))*x0+int(expm(A*(t-
tau))*B*Utau,tau,0,t);
cnt=1;
t2=0:0.01:0.5;
for t1=t2;
xsol2(cnt,1:2)=subs(xsol,t,t1);
cnt=cnt+1;
end

% plot there difference
subplot(2,1,1)
plot(xsol2(:,1)-xsol1(:,1),xsol2(:,2)-xsol1(:,2))
xlim([min(xsol2(:,1)-xsol1(:,1)), max(xsol2(:,1)-
xsol1(:,1))])

% Calculate the diff
Dx0=[0.005; 0.005];
J=expm(A*(t-t0));
cnt=1;
t2=0:0.01:0.5;
for t1=t2;
Dx(cnt,1:2)=subs(J,t,t1)*Dx0;
cnt=cnt+1;
end
subplot(2,1,2)
plot(Dx(:,1),Dx(:,2),'r')
xlim([min(Dx(:,1)), max(Dx(:,1))])
```



#### 4. Homogeneous Linear Time Varying System

If we have a linear system  $\dot{x}(t) = A(t)x(t)$

Then the solution is  $\phi(t, t_0, x_0) = \Phi_{STM}(t, t_0)x_0$  and hence of course

$$\frac{\partial \phi(t, t_0, x_0)}{\partial x_0} = \Phi_{STM}(t, t_0) \quad (4)$$

To find now this STM we have to numerically solve:

$$\frac{d}{dt} \Phi_{STM}(t, t_0) = A(t)\Phi_{STM}(t, t_0), \Phi_{STM}(t_0, t_0) = I_{n \times n}$$

Notice that through that last equation we can give another definition for the STM (in this case: the partial derivative of the solution wrt the initial condition:

$$\frac{\partial \phi(t, t_0, x_0)}{\partial x_0}$$

#### Example 2 (LTV):

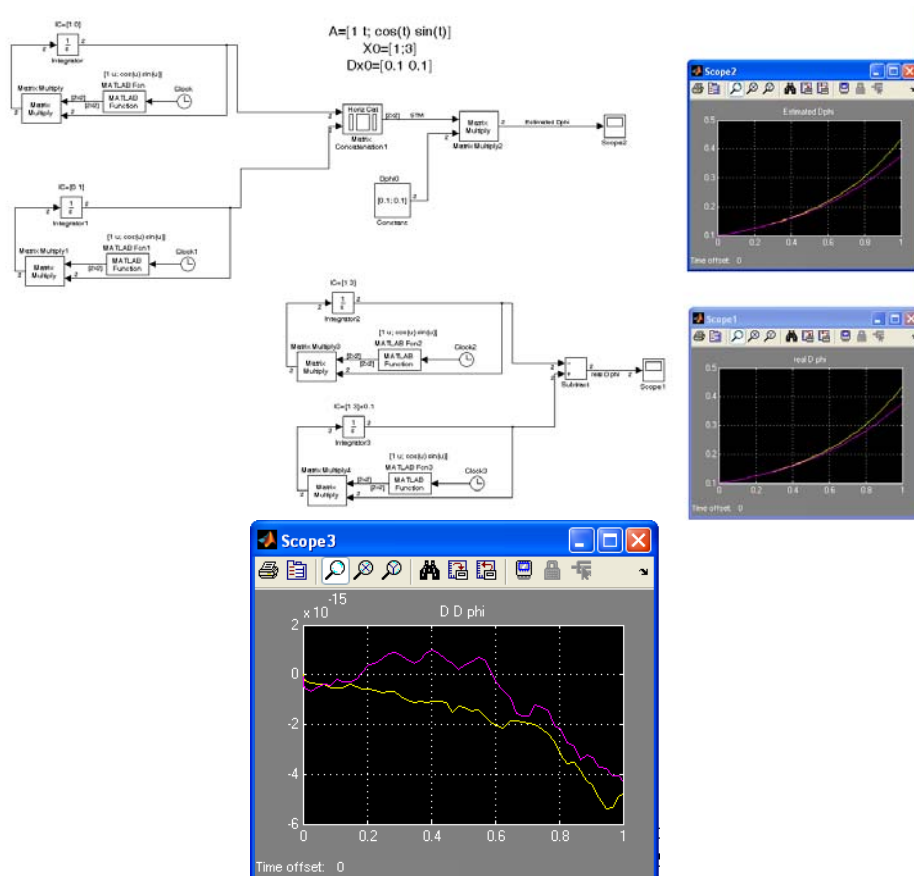


Figure 2

## 5. Non – Homogeneous Linear Time Varying System

The analysis here is exactly the same as before so no examples are given.

## 6. Nonlinear system

If we have a nonlinear system:

$$\begin{aligned} \phi(t, t_0, x_0) &= x_0 + \int_0^t f(\phi(\tau, t_0, x_0), \tau) d\tau \Leftrightarrow \\ \frac{\partial \phi(t, t_0, x_0)}{\partial x_0} &= I + \int_0^t \frac{\partial f(\phi(\tau, t_0, x_0), \tau)}{\partial x_0} d\tau \Leftrightarrow \\ \frac{\partial \phi(t, t_0, x_0)}{\partial x_0} &= I + \int_0^t \frac{\partial f(\phi(\tau, t_0, x_0), \tau)}{\partial \phi(\tau, t_0, x_0)} \frac{\partial \phi(\tau, t_0, x_0)}{\partial x_0} d\tau \Leftrightarrow \\ \frac{d}{dt} \left( \underbrace{\frac{\partial \phi(t, t_0, x_0)}{\partial x_0}}_{Y(t)} \right) &= \underbrace{\frac{\partial f(\phi(t, t_0, x_0), t)}{\partial \phi(t, t_0, x_0)}}_{\text{Jacobian}} \frac{\partial \phi(t, t_0, x_0)}{\partial x_0} \Leftrightarrow \\ \frac{d}{dt} (\phi_{x_0}(t, t_0, x_0)) &= \left. \frac{\partial f(x, t)}{\partial x} \right|_{x=\phi(t, t_0, x_0)} \phi_{x_0}(t, t_0, x_0) \end{aligned}$$

Now, the RHS must be equal to the LHS for  $t = t_0$  which means that

$\phi_{x_0}(t_0, t_0, x_0) = I_{n \times n}$ . Also, in a linear system  $\frac{\partial f(x, t)}{\partial x} = A(t)$  which means again that  $\phi_{x_0}(t, t_0, x_0)$  is nothing more than the STM.

### Example 3 (NL):

Consider the system  $\left. \begin{aligned} \frac{dx}{dt} &= -y + x(1 - x^2 - y^2) \\ \frac{dy}{dt} &= x + y(1 - x^2 - y^2) \end{aligned} \right\}$ . The numerical solution for  $x(0)=1$

and  $y(0)=0$  is:

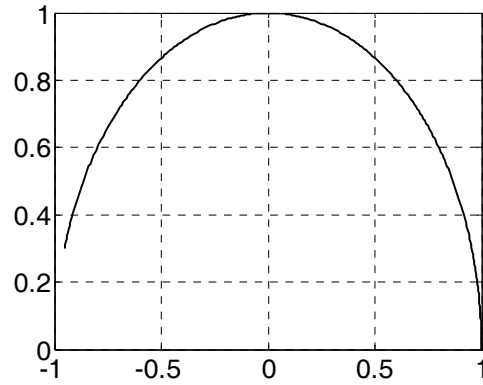


Figure 3

The Jacobian is:  $\frac{\partial f(x)}{\partial x} = \begin{bmatrix} 1-3x^2-y^2 & -1-2xy \\ 1-2xy & 1-x^2-3y^2 \end{bmatrix}$  which needs to be evaluated

along the orbit  $\phi(t, t_0, x_0)$  in order to find the STM:

$$\frac{d}{dt} (\phi_{x_0}(t, t_0, x_0)) = \frac{\partial f(x, t)}{\partial x} \Big|_{x=\phi(t, t_0, x_0)} \phi_{x_0}(t, t_0, x_0), \phi_{x_0}(t_0, t_0, x_0) = I_{n \times n}$$

The Jacobian is:

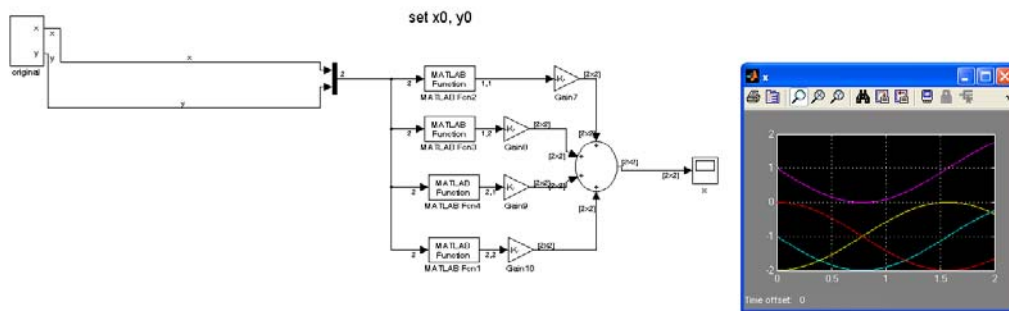


Figure 4

Hence, after 2s the STM has the value:

```
>> [DPA' DPB']
ans =
    -0.00762199518287    -0.90929742682579
     0.01665436331215    -0.41614683654700
```

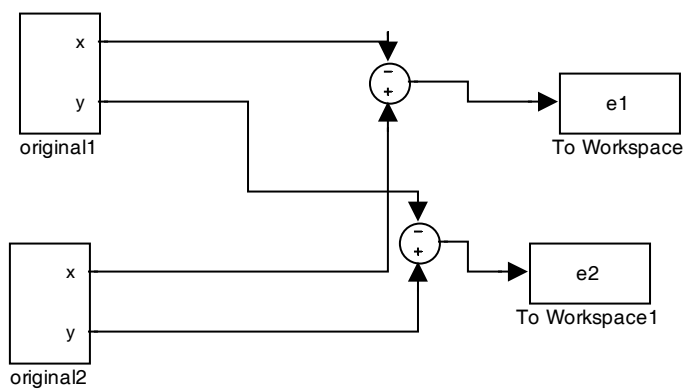
So let's add a small perturbation  $\Delta\phi(t_0, t_0, x_0) = \begin{bmatrix} 0.001 \\ 0.002 \end{bmatrix}$  and I expect the perturbation

after 2s to be:

```
>> [DPA' DPB'] * [0.001; 0.002]
```

ans =

```
-0.00182621684883  
-0.00081563930978
```



```
>> [e1; e2]
```

ans =

```
-0.00182360306785  
-0.00081662773015
```

**Summary**



So to summarise:

- To find how the perturbations evolve we use:

$$\Delta\phi(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0)$$

- Thus we need  $\frac{\partial\phi(t, t_0, x_0)}{\partial x_0}$  which is the STM.

- Hence to find the STM:

- For LTI systems  $\frac{\partial\phi(t, t_0, x_0)}{\partial x_0} = e^{A(t-t_0)}$

- For LTV systems  $\frac{d}{dt} \Phi_{STM}(t, t_0) = A(t) \Phi_{STM}(t, t_0), \Phi_{STM}(t_0, t_0) = I_{n \times n}$

- For NL systems again solve the previous equation but now

$$A(t) = \left. \frac{\partial f(x, t)}{\partial x} \right|_{x=\phi(t, t_0, x_0)}$$

## 7. Stability Analysis

Now, let's assume that we have a periodic orbit then we can still apply the previous method and we can try to find the perturbations using:

$$\Delta\phi(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0) \text{ or } \Delta\phi(t, t_0, x_0) = \Phi(t, t_0, x_0) \Delta\phi(t_0, t_0, x_0)$$

Notice that we used  $\Phi(t, t_0, x_0)$  instead of  $\Phi(t, t_0)$  as in the general nonlinear case

$\frac{\partial\phi(t, t_0, x_0)}{\partial x_0}$  may still be a function of  $x_0$ .

This matrix now can be found by solving

$$\frac{d}{dt} \Phi_{STM}(t, t_0) = \left. \frac{\partial f(x, t)}{\partial x} \right|_{x=\phi(t, t_0, x_0)} \Phi_{STM}(t, t_0), \Phi_{STM}(t_0, t_0) = I_{n \times n}$$

Let's evaluate the above perturbations at  $t=t_0+T$ :

$$\Delta\phi(t_0 + T, t_0, x_0) = \Phi(T + t_0, t_0, x_0) \Delta\phi(t_0, t_0, x_0)$$

Then we call the matrix  $\Phi(T + t_0, t_0, x_0)$  the Monodromy Matrix (MM) of the limit cycle.

$$\Delta\phi(2T + t_0, t_0, x_0) = \Phi(T + t_0, t_0, x_0)\Delta\phi(T + t_0, t_0, x_0) \Leftrightarrow$$

$$\text{Now, } \Delta\phi(2T + t_0, t_0, x_0) = \Phi(T + t_0, t_0, x_0)\Phi(T + t_0, t_0, x_0)\Delta\phi(t_0, t_0, x_0) \Leftrightarrow$$

$$\Delta\phi(2T + t_0, t_0, x_0) = \Phi^2(T + t_0, t_0, x_0)\Delta\phi(t_0, t_0, x_0)$$

$$\text{And also, } \Delta\phi(2T + t_0, t_0, x_0) = \Phi(2T + t_0, t_0, x_0)\Delta\phi(t_0, t_0, x_0)$$

Thus we immediately see that  $\Phi(2T + t_0, t_0, x_0) = \Phi^2(T + t_0, t_0, x_0)$  which implies

that:  $\Phi(kT + t_0, t_0, x_0) = \Phi^k(T + t_0, t_0, x_0)$  or that

$$\Delta\phi(kT + t_0, t_0, x_0) = \Phi^k(T + t_0, t_0, x_0)\Delta\phi(t_0, t_0, x_0)$$

Using eigenvalue decomposition:  $\Delta\phi(kT + t_0, t_0, x_0) = U^{-1}\Lambda^k U\Delta\phi(t_0, t_0, x_0)$

Which implies that if the all eigenvalues of the monodromy matrix have absolute value less than 1 the orbit is stable.

### Summary:

So to summarise:

- $\Delta\phi(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0}\Delta\phi(t_0, t_0, x_0)$
- Thus we need  $\frac{\partial\phi(t, t_0, x_0)}{\partial x_0}$  which is the STM.
- Hence to find the STM:
  - For LTI systems  $\frac{\partial\phi(t, t_0, x_0)}{\partial x_0} = e^{A(t-t_0)}$
  - For LTV systems  $\frac{d}{dt}\Phi_{STM}(t, t_0) = A(t)\Phi_{STM}(t, t_0)$ ,  $\Phi_{STM}(t_0, t_0) = I_{n \times n}$
  - For NL systems again solve the previous equation but now
$$A(t) = \left. \frac{\partial f(x, t)}{\partial x} \right|_{x=\phi(t, t_0, x_0)}$$
- Evaluate the STM at  $t = t_0 + T$

## 8. Floquet Theory

Another way to see the above is to linearise around the limit cycle:

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

$$\dot{x}(t) = f(x(t), t) = f(x_p(t) + \Delta x(t), t) = f(x_p(t), t) + \left. \frac{\partial f(x(t), t)}{\partial x(t)} \right|_{x=x_p} \Delta x(t) \Leftrightarrow$$

$$\dot{x}(t) - \dot{x}_p(t) = \left. \frac{\partial f(x(t), t)}{\partial x(t)} \right|_{x=x_p} \Delta x(t) \Leftrightarrow$$

$$\frac{d}{dt} \Delta x(t) = \left. \frac{\partial f(x(t), t)}{\partial x(t)} \right|_{x=x_p} \Delta x(t)$$

Which in the previous notation this can be written as:

$$\frac{d}{dt} \Delta \phi(t, t_0, x_0) = \left. \frac{\partial f(x(t), t)}{\partial x(t)} \right|_{x=\phi(t, t_0, x_0)} \Delta \phi(t, t_0, x_0) \text{ and thus the solution is:}$$

$\Delta \phi(t, t_0, x_0) = \Phi_{STM}(t, t_0, x_0) \times \Delta \phi(t_0, t_0, x_0)$  which again implies that

$$\frac{d}{dt} \Phi_{STM}(t, t_0, x_0) = \left. \frac{\partial f(x(t), t)}{\partial x(t)} \right|_{x=\phi(t, t_0, x_0)} \Phi_{STM}(t, t_0, x_0) \text{ and that } \Phi_{STM}(t, t_0, x_0) = I.$$

Hence we have a similar equation to the one before and hence we get the same result.

Thus to find the stability properties we have to find the monodromy matrix and each eigenvalues.

### Summary:

- Linearise the system around the periodic orbit and hence express the perturbations through a LTV differential equation
- To solve that we need the STM which can be found by solving a matrix differential equation:
- $\frac{d}{dt} \Phi_{STM}(t, t_0, x_0) = \left. \frac{\partial f(x(t), t)}{\partial x(t)} \right|_{x=\phi(t, t_0, x_0)} \Phi_{STM}(t, t_0, x_0), \Phi_{STM}(t, t_0, x_0) = I$
- The stability is found by evaluating  $\Phi_{STM}(t, t_0, x_0)$  after  $t = T + t_0$ .

## 9. Poincaré Map

Going back to the general solution of a nonlinear (it can also be linear) system:

$$\phi(t, t_0, x_0) = x_0 + \int_0^t f(\phi(\tau, t_0, x_0), \tau) d\tau$$

Now, the stroboscopic Poincaré map is nothing more than:

$$\phi(T + t_0, t_0, x_0) = x_0 + \int_0^{T+t_0} f(\phi(\tau, t_0, x_0), \tau) d\tau \Leftrightarrow x_{n+1} = P(x_n) = x_n + \int_0^{T+t_0} f(\phi(\tau, t_0, x_n), \tau) d\tau$$

$$\text{The fixed point is } x_{FP} + \int_0^{T+t_0} f(\phi(\tau, t_0, x_{FP}), \tau) d\tau = x_{FP} \Leftrightarrow \int_0^{T+t_0} f(\phi(\tau, t_0, x_{FP}), \tau) d\tau = 0$$

The stability analysis is similar to the previous case:

$$\frac{\partial P(x_{FP})}{\partial x_{FP}} = I + \int_0^{T+t_0} \frac{\partial f(\phi(\tau, t_0, x_0), \tau)}{\partial \phi(\tau, t_0, x_0)} \frac{\partial \phi(\tau, t_0, x_0)}{\partial x_0} d\tau$$

**Hence the monodromy matrix is nothing more than the Jacobian of the Poincaré map!**

Unfortunately again in the general case we have to numerically solve the matrix differential equation:

$$\frac{d}{dt} (\phi_{x_{FP}}(t, t_0, x_{FP})) = \frac{\partial f(x, t)}{\partial x} \Big|_{x=\phi(t, t_0, x_{FP})} \phi_{x_0}(t, t_0, x_{FP}) \text{ for } t \in [0, T]$$

**Summary:**

- Evaluate the solution after T time:  $\phi(T + t_0, t_0, x_0)$ .
- The stability can be found from  $\frac{\partial \phi(t, t_0, x_0)}{\partial x_0}$ .
- To calculate this, again we have to solve a matrix differential equation.

### Overall Summary:

1. In all cases we start from assuming that we have the solution  $\phi(t, t_0, x_0)$ .
2. Express the perturbations as:  $\Delta\phi(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0)$
3. To find  $\Phi_{STM}(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0}$  solve the matrix DE:  $\frac{d}{dt}\Phi_{STM}(t, t_0, x_0) = A(t)\Phi_{STM}(t, t_0, x_0)$ ,  $\Phi_{STM}(t_0, t_0, x_0) = I_{n \times n}$
4. The solution of the above equation will give  $\Phi_{STM}(t, t_0, x_0)$  and hence we can get the MM:  $\Phi_{STM}(T + t_0, t_0, x_0)$ . The eigenvalues of this will determine the stability as  $\Delta\phi(t_0 + kT, t_0, x_0) = \left(\frac{\partial\phi(t, t_0, x_0)}{\partial x_0}\right)^k \Delta\phi(t_0, t_0, x_0)$
5. For the Floquet Theory we must first derive  $\frac{d}{dt}\Delta\phi(t, t_0, x_0) = \frac{\partial f(x(t), t)}{\partial x(t)} \Big|_{x=\phi(t, t_0, x_0)} \Delta\phi(t, t_0, x_0)$  which is LTV and hence we need the STM to express  $\Delta\phi(t, t_0, x_0) = \frac{\partial\phi(t, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0)$ .
6. For the Poincaré map it is exactly the same as before but now we write directly:  $\Delta\phi(t_0 + T, t_0, x_0) = \frac{\partial\phi(t_0 + T, t_0, x_0)}{\partial x_0} \Delta\phi(t_0, t_0, x_0)$ .

### Thus

- The Floquet Theory needs an extra step than the Trajectory Sensitivity
- The Poincaré map is just a special case of the Trajectory Sensitivity

## Example: Poincaré Map vs Floquet Theory

### 1. Initial numerical analysis

A system is given by

$$\left. \begin{aligned} \frac{dx}{dt} &= -y + x(1 - x^2 - y^2) \\ \frac{dy}{dt} &= x + y(1 - x^2 - y^2) \end{aligned} \right\}$$

The EPs are:

$$\left. \begin{aligned} 0 &= -y + x(1 - x^2 - y^2) \\ 0 &= x + y(1 - x^2 - y^2) \end{aligned} \right\}$$

From the 1<sup>st</sup> eqn:  $-y + x(1 - x^2 - y^2) = 0 \Leftrightarrow y = x(1 - x^2 - y^2)$

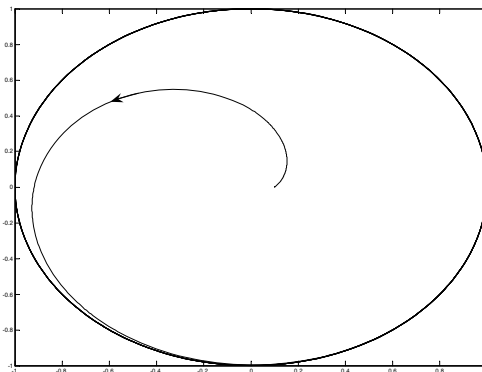
And by replacing that at the 2<sup>nd</sup> eqn:

$$0 = x + x(1 - x^2 - y^2)(1 - x^2 - y^2)$$

$$0 = x + x(1 - x^2 - y^2)^2 \Rightarrow \begin{cases} x = 0 \Rightarrow y = 0 \\ 1 + (1 - x^2 - y^2)^2 = 0 \text{ which is inconsistent} \end{cases}$$

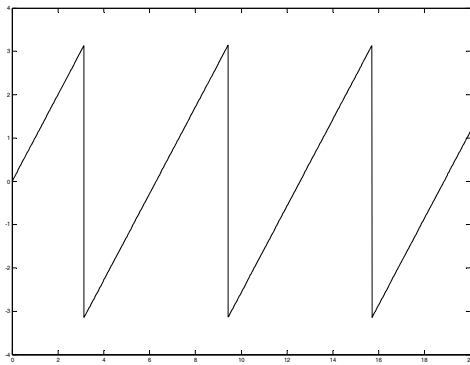
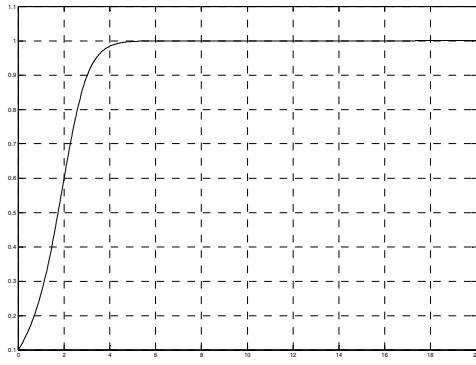
Hence there is only one equilibrium point at the origin.

The system has a stable limit cycle as it can be seen by its numerical simulation.



To study the system we can use polar coordinates:

$$\begin{aligned} x &= r \cos(\theta) \\ y &= r \sin(\theta) \end{aligned} \text{ which will give: } r^2 = x^2 + y^2 \text{ and } \theta = \tan^{-1}\left(\frac{y}{x}\right)$$



## 2. Polar Coordinates

Lets calculate the model in polar coordinates:

$$\frac{dr^2}{dt} = \frac{d}{dt}(x^2 + y^2)$$

$$2r \frac{dr}{dt} = 2x \frac{dx}{dt} + 2y \frac{dy}{dt}$$

$$\frac{dr}{dt} = \frac{1}{r} \left( x \frac{dx}{dt} + y \frac{dy}{dt} \right)$$

$$\frac{dr}{dt} = \frac{1}{r} \left( x(-y + x(1 - x^2 - y^2)) + y(x + y(1 - x^2 - y^2)) \right)$$

Also,  $x = r \cos(\theta)$ ,  $y = r \sin(\theta)$ , so  $x^2 + y^2 = r^2$ , Hence  $\frac{dr}{dt} = \frac{1}{r} \left( x(-y + x(1 - r^2)) + y(x + y(1 - r^2)) \right)$

$$\frac{dr}{dt} = \frac{1}{r} \left( -xy + x^2(1 - r^2) + yx + y^2(1 - r^2) \right)$$

$$\frac{dr}{dt} = \frac{1}{r} \left( -xy + x^2(1 - r^2) + yx + y^2(1 - r^2) \right)$$

$$\frac{dr}{dt} = \frac{1}{r} \left( x^2(1 - r^2) + y^2(1 - r^2) \right)$$

$$\frac{dr}{dt} = \frac{1}{r} \left( (x^2 + y^2)(1 - r^2) \right)$$

$$\frac{dr}{dt} = \frac{1}{r} \left( r^2(1 - r^2) \right)$$

$$\frac{dr}{dt} = r(1 - r^2)$$

And

$$\theta = \tan^{-1}\left(\frac{y}{x}\right) \Rightarrow \frac{d\theta}{dt} = \frac{d}{dt}\left(\frac{y}{x}\right) \frac{1}{1 + \left(\frac{y}{x}\right)^2} = \frac{\frac{dy}{dt}x - \frac{dx}{dt}y}{x^2} \frac{x^2}{x^2 + y^2} =$$

$$\left(\frac{dy}{dt}x - \frac{dx}{dt}y\right) \frac{1}{x^2 + y^2} =$$

$$\left((x + y(1 - x^2 - y^2))x - (-y + x(1 - x^2 - y^2))y\right) \frac{1}{x^2 + y^2} =$$

$$\left((x + y(1 - r^2))x - (-y + x(1 - r^2))y\right) \frac{1}{r^2} =$$

$$\left((x^2 + yx(1 - r^2)) - (-y^2 + xy(1 - r^2))\right) \frac{1}{r^2} =$$

$$\left(x^2 + yx(1 - r^2) + y^2 - xy(1 - r^2)\right) \frac{1}{r^2} =$$

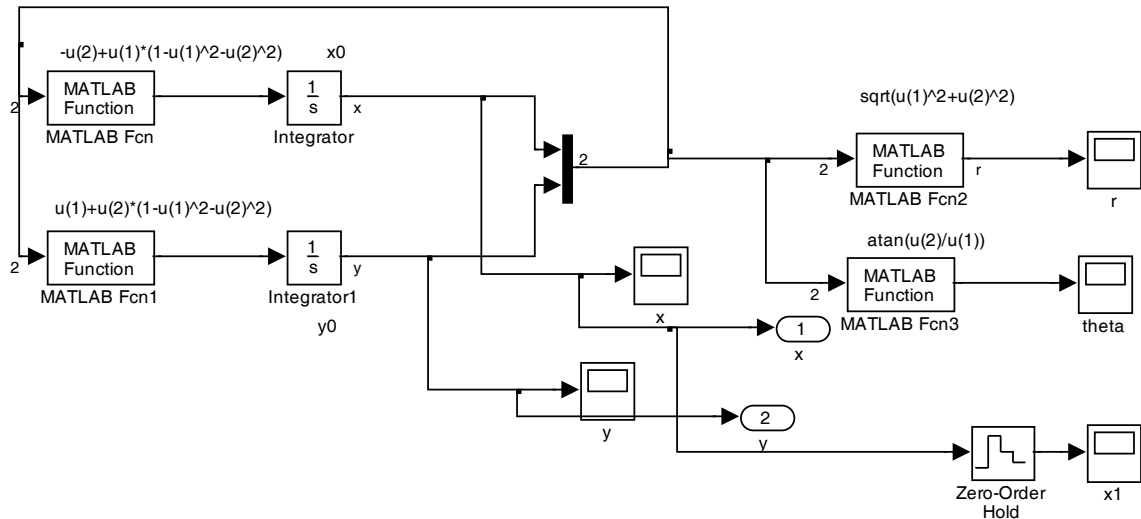
$$\left(x^2 + y^2\right) \frac{1}{r^2} = r^2 \frac{1}{r^2} = 1$$



So the system in polar coordinates is:

$$\begin{cases} \frac{dr}{dt} = r(1-r^2) \\ \frac{d\theta}{dt} = 1 \end{cases}$$

, Simulink validated this as well.



## 2.1 Equilibria and solution of system in polar coordinates

The 2<sup>nd</sup> equation has no EP but the first has at:

$$r(1-r^2) = 0 \Rightarrow \begin{cases} r = 0 \\ r = 1 \end{cases}$$

The Jacobian is  $\frac{\partial}{\partial r}(r-r^3) = 1-3r^2 \Rightarrow \begin{cases} r = 0 : 1 \\ r = 1 : -2 \end{cases}$ . So the origin is unstable and  $r=1$  is stable.

The ODE in polar coordinates can be solved and we have:

$$r(t, r_0) = \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-2t} \right]^{-1/2}$$

$$\theta(t, \theta_0) = \theta_0 + t$$

So we can see that the angle  $\theta$  will continuously increase while the amplitude  $r$  will converge to 1.

## 2.2 Poincare map and Jacobian from polar coordinates

But regarding the original system the cosine/sine of the angle will be periodic of period  $2\pi$  and as the amplitude will converge to a steady state value then this implies that the system will have a limit cycle which will be a perfect circle of radius 1, centre at the origin and a period of  $2\pi$ .

So if we start at  $r_0$  then the next point after one period (Poincare map) is:

$$r(2\pi, r_0) = P(r_0) = \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-4\pi} \right]^{-1/2}, \text{ which can be written as:}$$

$$r_{n+1} = P(r_n) = \left[ 1 + \left( \frac{1}{r_n^2} - 1 \right) e^{-4\pi} \right]^{-1/2}$$

To find the fixed point:

$$\begin{aligned} r_n &= \left[ 1 + \left( \frac{1}{r_n^2} - 1 \right) e^{-4\pi} \right]^{-1/2} \Leftrightarrow r_n \left[ 1 + \left( \frac{1}{r_n^2} - 1 \right) e^{-4\pi} \right]^{1/2} = 1 \Leftrightarrow r_n^2 \left[ 1 + \left( \frac{1}{r_n^2} - 1 \right) e^{-4\pi} \right] = 1 \Leftrightarrow \\ r_n^2 + (1 - r_n^2) e^{-4\pi} &= 1 \Leftrightarrow r_n^2 + e^{-4\pi} - r_n^2 e^{-4\pi} = 1 \Leftrightarrow r_n^2 (1 - e^{-4\pi}) = 1 - e^{-4\pi} \Leftrightarrow r_n^2 = 1 \Rightarrow r_n = 1 \end{aligned}$$

As the FP is 1 then the Jacobian at that point is:

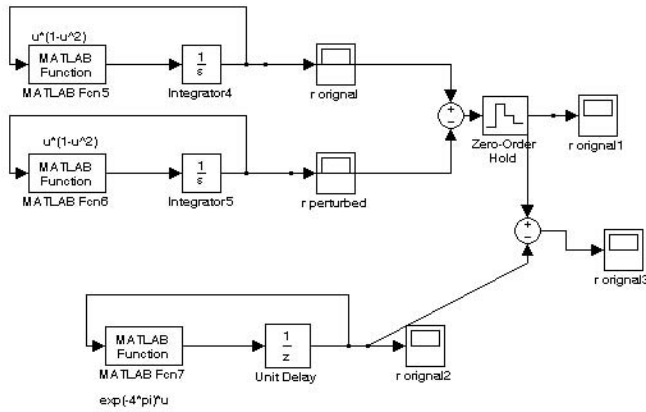
$$\frac{\partial P(r_n)}{\partial r_n^2} = e^{-4\pi} r_n^3 \left[ 1 + \left( \frac{1}{r_n^2} - 1 \right) e^{-4\pi} \right]^{-3/2} \Rightarrow \left. \frac{\partial P(r_n)}{\partial r_n^2} \right|_{r_n=r_0} = e^{-4\pi} r_0^3 \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-4\pi} \right]^{-3/2}$$

And hence:  $\left. \frac{\partial P(r_n)}{\partial r_n^2} \right|_{r_n=1} = e^{-4\pi} = 3.48 \times 10^{-6} < 1$  and hence it is stable.

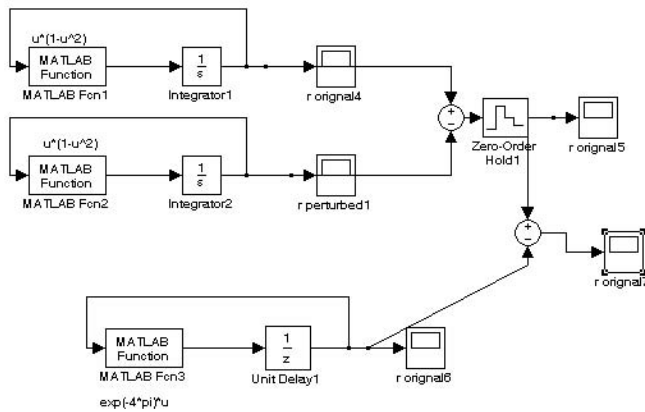
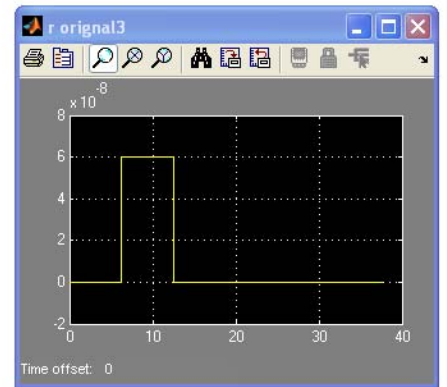
Or that simply the perturbations around the fixed point are given by:

$$\Delta r(n+1) = e^{-4\pi} \Delta r(n)$$

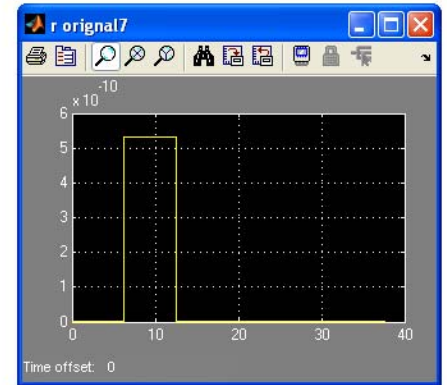
which has been crosschecked using Simulink. To do that I simulated the original system at  $r_0=1$  and the same system with  $r_0=0.1$  and I sampled their difference with  $T = 2\pi$ . Then I simulated the  $\Delta r(n+1) = e^{-4\pi} \Delta r(n)$  with initial condition  $1 - 0.9 = 0.1$  and I compared the previous difference and that output. The smaller the IC the smaller their difference:



D=0.1



D=0.01



### 3. Cartesian coordinates

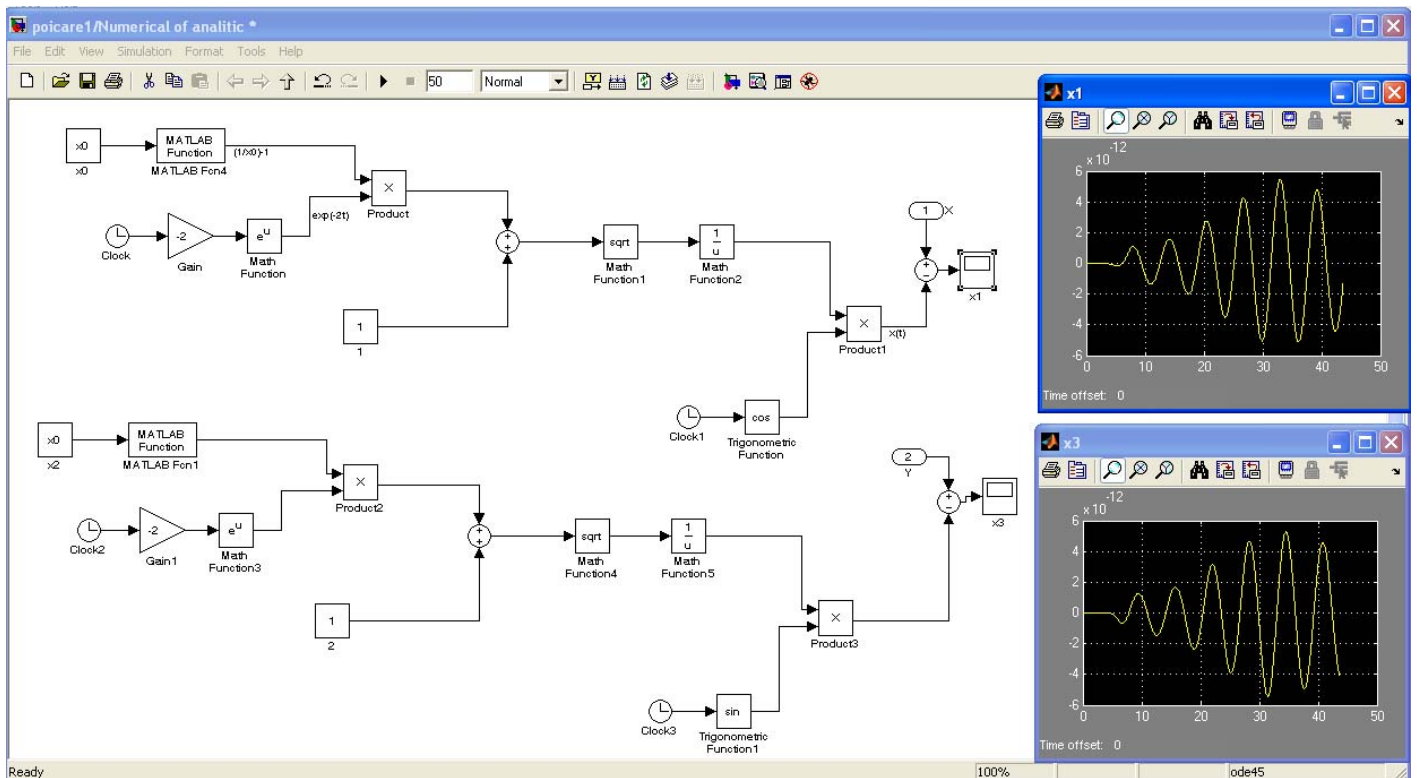
#### 3.1 Solution in Cartesian coordinates

Now let's find the solution in the x-y plane:

$$\left. \begin{aligned}
 x = r \cos(\theta) \\
 y = r \sin(\theta)
 \end{aligned} \right\} \theta_0=0 \Rightarrow \left. \begin{aligned}
 x &= \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \cos(t) \\
 y &= \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \sin(t)
 \end{aligned} \right\} \Rightarrow \left. \begin{aligned}
 x &= \left[ 1 + \left( \frac{1}{x_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \cos(t) \\
 y &= \left[ 1 + \left( \frac{1}{x_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \sin(t)
 \end{aligned} \right\}$$

$$\left. \begin{aligned}
 x(0) = r(0) \cos(0) \\
 y(0) = r(0) \sin(0)
 \end{aligned} \right\} \Rightarrow \left. \begin{aligned}
 x(0) = r(0) \\
 y(0) = 0
 \end{aligned} \right\}$$

This has been validated by Simulink. I.e. I simulated the previous set of equations that give analytical solution and compare that with the numerical that I get from Simulink. The error was  $10^{-14}$ .



If I do not assume that  $\theta=0$ :

$$\left. \begin{aligned} x = r \cos(\theta) \\ y = r \sin(\theta) \end{aligned} \right\} \theta_0=0 \Rightarrow \left. \begin{aligned} x &= \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \cos(\theta_0 + t) \\ y &= \left[ 1 + \left( \frac{1}{r_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \sin(\theta_0 + t) \end{aligned} \right\} \Rightarrow \left. \begin{aligned} x &= \left[ 1 + \left( \frac{1}{x_0^2 + y_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \cos \left( \tan^{-1} \left( \frac{y_0}{x_0} \right) + t \right) \\ y &= \left[ 1 + \left( \frac{1}{x_0^2 + y_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \sin \left( \tan^{-1} \left( \frac{y_0}{x_0} \right) + t \right) \end{aligned} \right\}$$

$$\left. \begin{aligned} r(0) &= \sqrt{x^2(0) + y^2(0)} \\ \theta(0) &= \tan^{-1} \left( \frac{y(0)}{x(0)} \right) \end{aligned} \right\}$$

### 3.2 Jacobian in Cartesian coordinates

Now let's find the Poincare map in the x-y domain:

$$\mathbf{P}(x_0, y_0) \Big|_{t=2\pi} = \begin{bmatrix} \left[ 1 + \left( \frac{1}{x_0^2 + y_0^2} - 1 \right) e^{-4\pi} \right]^{-1/2} \cos \left( \tan^{-1} \left( \frac{y_0}{x_0} \right) + 2\pi \right) \\ \left[ 1 + \left( \frac{1}{x_0^2 + y_0^2} - 1 \right) e^{-4\pi} \right]^{-1/2} \sin \left( \tan^{-1} \left( \frac{y_0}{x_0} \right) + 2\pi \right) \end{bmatrix}$$

The Jacobian of that is rather cumbersome so I will use Matlab

```

clc, clear
syms x0 y0 T

r2=x0^2+y0^2; ra=(1/r2)-1;
rb=ra*exp(2*T); rb=ra*exp(-2*T);
rc=1/sqrt(1+rb); rat=y0/x0;
th1=atan(rat)+T;

f1=rc*cos(th1); f2=rc*sin(th1);
Df11=diff(f1,x0); Df12=diff(f1,y0);
Df21=diff(f2,x0); Df22=diff(f2,y0);
Df11=subs(Df11,{T,x0,y0},{2*pi,1,0});
Df12=subs(Df12,{T,x0,y0},{2*pi,1,0});
Df21=subs(Df21,{T,x0,y0},{2*pi,1,0});
Df22=subs(Df22,{T,x0,y0},{2*pi,1,0});
DF=[Df11 Df12; Df21 Df22]
eig(DF)

```

DF =

```

0.00000348734236 0.000000000000000
-0.000000000000000 1.000000000000000

```

We get 2 eigenvalues:

```

0.00000348734236
1.000000000000000

```

The 1<sup>st</sup> one is the same as the one that we got from the polar coordinates and 1 is from the fact that the system is autonomous.

Notice that we could have calculated the Jacobian if we had solved

$$\frac{d}{dt} \left( \phi_{x_{FP}}(t, t_0, x_{FP}) \right) = \frac{\partial f(x, t)}{\partial x} \Big|_{x=\phi(t, t_0, x_{FP})} \phi_{x_0}(t, t_0, x_{FP}) \text{ for } t \in [0, T]$$

### 3.3 Perturbation calculation using the Jacobian

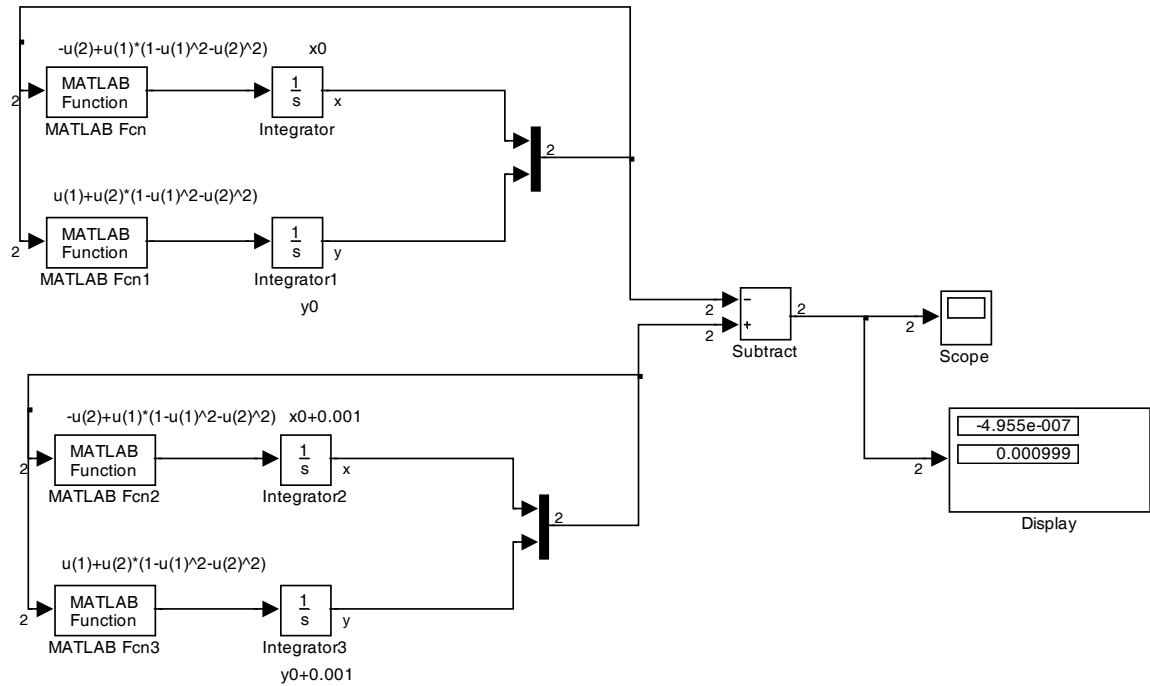
```
>> DF*[0.001;0.001]
```

```
ans =
```

```
1.0e-003 *
```

```
0.00000348734236
```

```
1.000000000000000
```



Note: If I did not know the IC for the Poincare map, i.e. its fixed point I have to use a Newton Rapshon method.

## Floquet Theory

### 4. Jacobian around the periodic orbit

So now I have a nonlinear system with a stable periodic orbit with  $T=6.28\dots$  and also I know its solution. Now I will try to use the Floquet Theory.

The nonlinear vector field is:

$$\mathbf{f}(x, y) = \begin{bmatrix} -y + x(1 - x^2 - y^2) \\ x + y(1 - x^2 - y^2) \end{bmatrix} \text{ and hence } \frac{d\mathbf{X}}{dt} = \mathbf{f}(\mathbf{X}) = \mathbf{f}(x, y)$$

So the Jacobian around the orbit is:

$$\frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} = \begin{bmatrix} \frac{\partial(-y + x(1 - x^2 - y^2))}{\partial x} & \frac{\partial(-y + x(1 - x^2 - y^2))}{\partial y} \\ \frac{\partial(x + y(1 - x^2 - y^2))}{\partial x} & \frac{\partial(x + y(1 - x^2 - y^2))}{\partial y} \end{bmatrix} =$$

$$\frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} = \begin{bmatrix} \frac{\partial(-y + (x - x^3 - xy^2))}{\partial x} & \frac{\partial(-y + (x - x^3 - xy^2))}{\partial y} \\ \frac{\partial(x + (y - yx^2 - y^3))}{\partial x} & \frac{\partial(x + (y - yx^2 - y^3))}{\partial y} \end{bmatrix} \text{ which will give:}$$

$$\frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} = \begin{bmatrix} 1 - 3x^2 - y^2 & -1 - 2xy \\ 1 - 2xy & 1 - x^2 - 3y^2 \end{bmatrix}$$

Now this must be evaluated around the orbit:

$$\left. \begin{aligned} x &= \left[ 1 + \left( \frac{1}{x_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \cos(t) \\ y &= \left[ 1 + \left( \frac{1}{x_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \sin(t) \end{aligned} \right\} \text{ starting}$$

on the point  $x_0=1, y_0=0$  as this is a point on the cycle.

$$\left. \begin{aligned} x_p(t) &= \cos(t) \\ y_p(t) &= \sin(t) \end{aligned} \right\} \text{ so:}$$

$$\frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \Big|_{\mathbf{X}=\mathbf{X}_p} = \begin{bmatrix} 1 - 3\cos(t)^2 - \sin(t)^2 & -1 - 2\cos(t)\sin(t) \\ 1 - 2\cos(t)\sin(t) & 1 - \cos(t)^2 - 3\sin(t)^2 \end{bmatrix}$$

Or:

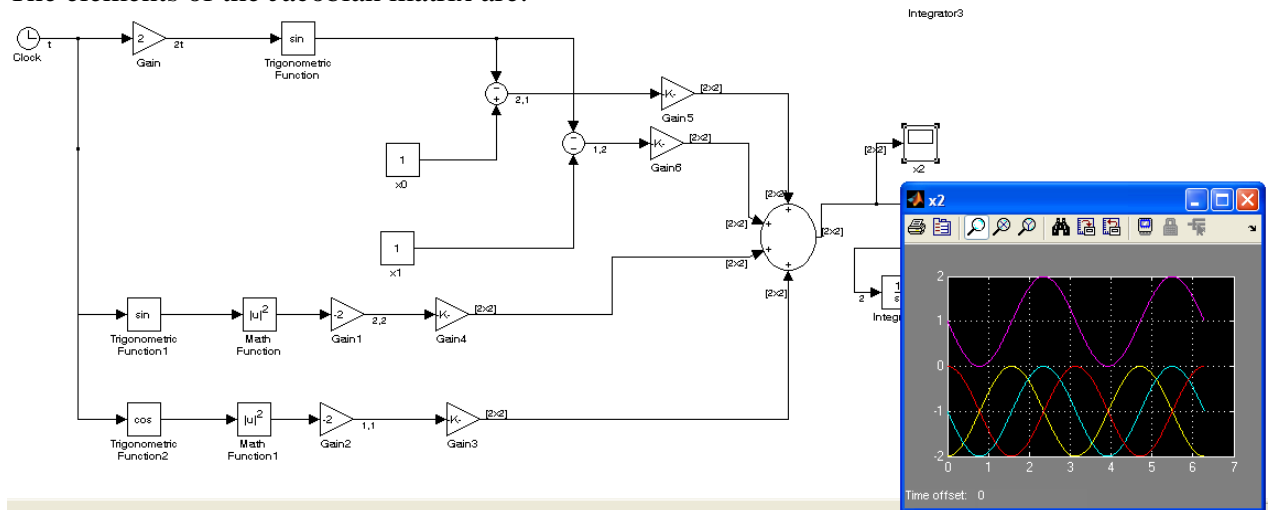
$$\left. \frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \right|_{\mathbf{X}=\mathbf{X}_p} = \begin{bmatrix} 1 - 2\cos(t)^2 - \cos(t)^2 - \sin(t)^2 & -1 - 2\cos(t)\sin(t) \\ 1 - 2\cos(t)\sin(t) & 1 - \cos(t)^2 - 2\sin(t)^2 - \sin(t)^2 \end{bmatrix}$$

$$\left. \frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \right|_{\mathbf{X}=\mathbf{X}_p} = \begin{bmatrix} 1 - 2\cos(t)^2 - 1 & -1 - 2\cos(t)\sin(t) \\ 1 - 2\cos(t)\sin(t) & 1 - 1 - 2\sin(t)^2 \end{bmatrix}$$

$$\left. \frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \right|_{\mathbf{X}=\mathbf{X}_p} = \begin{bmatrix} -2\cos(t)^2 & -1 - \sin(2t) \\ 1 - \sin(2t) & -2\sin(t)^2 \end{bmatrix}$$

For  $t=0$  to  $t=T=6.28\dots$

The elements of the Jacobian matrix are:



## 5. Solution of Matrix Differential Equation

### 5.1 Calculation of Monodromy matrix

In order to find the monodromy matrix I need to solve the MDE:

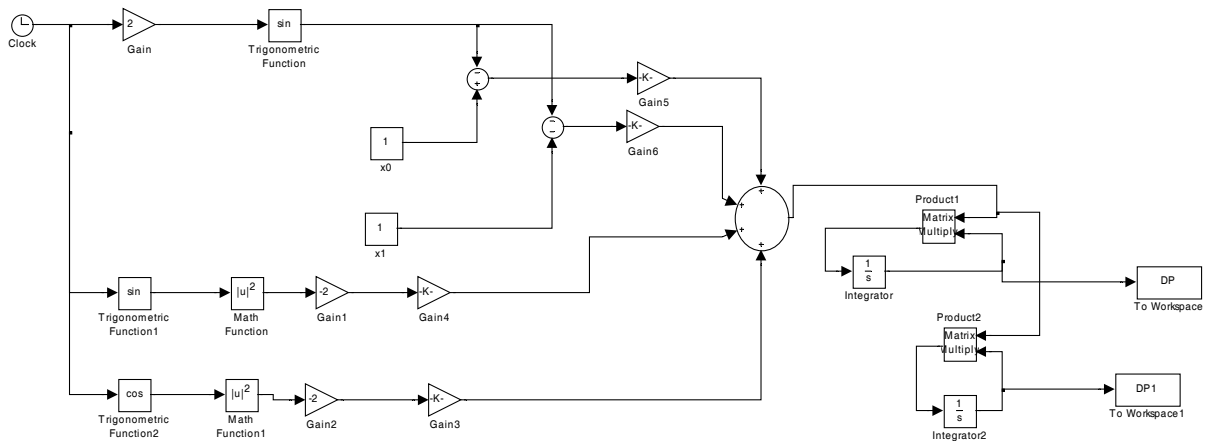
$$\frac{d\Phi(t, t_0)}{dt} = \left. \frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \right|_{\mathbf{X}=\mathbf{X}_p} \Phi(t, t_0)$$

$$\frac{d\Phi(t, t_0)}{dt} = \mathbf{A}(t)\Phi(t, t_0)$$

With the IC =  $\mathbf{I}_2$

The following model will calculate the matrix A from  $t=0$  to  $t=T$ .





```
>> DP=[DP1' DP2']
DP =
    0.00000348734236 -0.0000000000000004
    0.0000000000000004  0.999999999999983
```

Notice that this result is very close to the one that we got from the Poincare Map.

Now, that the monodromy matrix is found we can find the Floquet multipliers:

```
>> eig(DP)
ans =
    0.00000348734236
    0.999999999999983
```

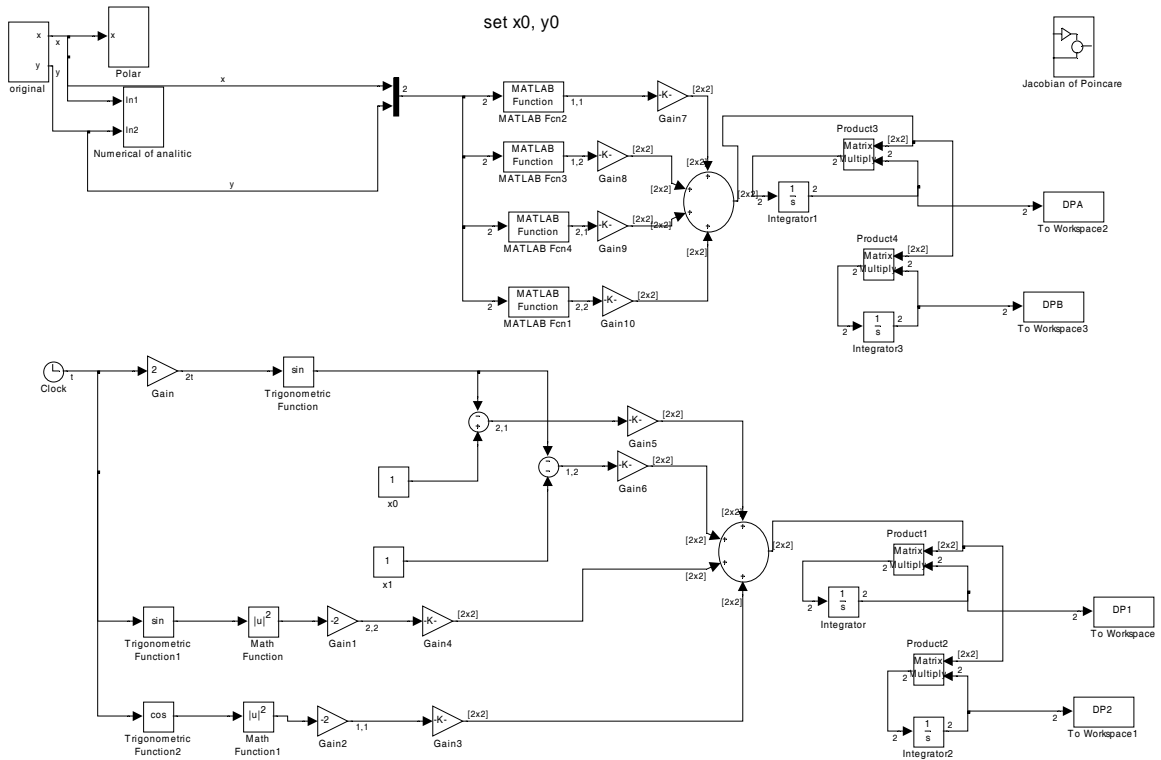
Which are very close to the ones from the PM.

Another way to do that is to numerically solve the original system and then to use the

numerical solution to  $\frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} = \begin{bmatrix} 1-3x^2-y^2 & -1-2xy \\ 1-2xy & 1-x^2-3y^2 \end{bmatrix}$  with

$$\left. \begin{aligned} x &= \left[ 1 + \left( \frac{1}{x_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \cos(t) \\ y &= \left[ 1 + \left( \frac{1}{x_0^2} - 1 \right) e^{-2t} \right]^{-1/2} \sin(t) \end{aligned} \right\} .$$

It is important though to remember that the system must start from  $x_0=1, y_0=0$ .



```
>> eig(DPa)
ans =
    0.00000348734236
    1.00000000000004
```

Hence we get the same results.

### 5.2 Perturbation calculation using the Monodromy matrix

So I have found  $\Phi(t,0)$ , now I will use this matrix to prove that  $\Delta X(t) = \Phi(t,0)\Delta X(0)$  by knowing that  $\Delta X(0) = [0.1 \ 0]$ . So for  $t=0.1T$  I have got from the NL system:

```
0.02572402284312
0.01868959658690
```

```
And from the product
>> E1=[DP1' DP2']*[0.1;0]
E1 =
    0.02302539573201
    0.01672892922346
```

These are in a very good agreement.

