Physics 584 Computational Methods The Lorenz Equations and Numerical Simulations of Chaos

Ryan Ogliore

April 18th 2019





Homework Review

Chaotic Dynamics

Lyapunov Exponent

Lyapunov Exponent of the Lorenz Equations

Chaotic motion in the Solar System

Lecture Outline

Lorenz Equations in an Analog Electronic Circuit

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Lorenz Equations

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$$= -\int 10(x-y) = \int 10(y-x)$$

 ...and likewise for dy/dt and dz/dt in the Lorenz equations



Lorenz Equations: Chaotic Waterwheel



The leaky waterwheel follows the Lorenz equations and display analogous behavior.

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Homework (continued)

Edward Lorenz, a meterologist, created a simplified mathematical model for nonlinear atmospheric thermal convection in 1962. Lorenz's model frequently arises in other types of systems, e.g. dynamos and electrical circuits. Now known as the Lorenz equations, this model is a system of three ordinary differential equations:

$$\begin{aligned} \frac{dx}{dt} &= \sigma(y - x),\\ \frac{dy}{dt} &= x(\rho - z) - y\\ \frac{dz}{dt} &= xy - \beta z. \end{aligned}$$

Note the last two equations involve quadratic nonlinearities. The intensity of the fluid motion is parameterized by the variable x; y and z are related to temperature variations in the horizontal and vertical directions.

Homework (continued)

Use Matlab's RK4 solver ode45 to solve this system of ODEs with the following starting points and parameters.

- 1. With $\sigma = 1$, $\beta = 1$, and $\rho = 1$, solve the system of Lorenz Equations for x(t = 0) = 1, y(t = 0) = 1, and z(t = 0) = 1. Plot the orbit of the solution as a three-dimensional plot for times 0–100.
- 2. For the Earth's atmosphere reasonable values are $\sigma = 10$ and $\beta = 8/3$. Also set $\rho = 28$; and using starting values: x(t = 0) = 5, y(t = 0) = 5, and z(t = 0) = 5; solve the system of Lorenz Equations for t = [0, 20]. Plot the orbit of the solution as a three-dimensional plot for t=0-20. Also plot z vs. x. Do any of the orbits that appear to overlap in this plot actually overlap when viewed in the three-dimensional plot?
- 3. Plot *x*, *y*, and *z* vs. time on one graph using Matlab's subplot function.

Homework (continued)

- 4. Use the same parameters as in #2 but add a very small number (e.g. 10^{-6}) to one of the starting values. Plot *x*, *y*, and *z* vs. time for *both* of these curves (one red, one blue). Solve the equation for longer times to see when the two solutions diverge from each other.
- 5. Find a value of ρ (while keeping $\sigma = 10$ and $\beta = 8/3$) such that the solution does not depend sensitively on the initial values. Plot both curves for x, y, and z vs. time as you did in #4.
- 6. For ρ =70, $\sigma = 10$, $\beta = 8/3$, initial starting value (5,5,5), over a time range 0–50, calculate and plot one solution using the default maximum step size for ode45: $0.1 \times (t_{\text{final}} t_{\text{initial}})$, and another solution for $1/1000^{\text{th}}$ of the default. Is this behavior related to the sensitivity on initial starting values you explored in #4?

ode45

- Matlab's ode45 employs the Dormand-Prince method, a type of Runge-Kutta method
- This method computes 4th and 5th order solutions. The difference between these solutions is then taken to be the error of the (fourth-order) solution
- ► If the error is smaller than the tolerance, the step is successful
- If the error is larger than the *tolerance* (see odeset: AbsTol and RelTol), the step is unsuccessful and the step size is decreased by an amount determined by the error/tolerance ratio
- This is an adaptive step-size integration algorithm. The user usually inputs a two-element vector for the time span over which to calculate the solution, the function returns a solution over it.
- ode45 is an extremely accurate solver!

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- Vectorizing code usually makes it easier to read and simpler to understand (but not always)
- Better to do operations explicitly rather than implicitly (e.g. time interpolation for output of ode45)
- > Your code is a reflection of yourself, take pride in it!

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The *Existence and Uniqueness Theorem* for systems of differential equations guarantees a unique solution for each set of initial conditions:

Consider the initial value problem

$$y'(t) = f(t, y(t)), \qquad y(t_0) = y_0$$

If *f* is uniformly **Lipschitz continuous** in *y* and continuous in *t*, then for some value ε , there exists a unique solution y(t) to the initial value problem on the interval $[t_0 - \varepsilon, t_0 + \varepsilon]$

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We will look into some of the other aspects of the Lorenz equations in today's lecture.

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You explored the behavior of the solution for these equations with various parameters and starting conditions.

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- Such a universe is completely determined by its initial conditions to evolve predictably with time (particularly championed by Laplace).
- Poincaré was the first person to see that Newton's laws of motion, in fact, predicted chaos all along.



Henri Poincaré

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Given a system of arbitrarily many point masses that attract each other according to Newton's laws, under the assumption that no two points ever collide, find a representation of the coordinates of each point as a series solution in a variable that is some known function of time and for all of whose values the series converges uniformly.

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This is the n-body problem:

$$m_i \ddot{\mathbf{q}}_i = \sum_{j \neq i}^n \frac{Gm_i m_j (\mathbf{q}_i - \mathbf{q}_j)}{|\mathbf{q}_i - \mathbf{q}_j|^3}, \ i = 1, \dots, n$$

Poincaré proved that an analytical solution to the three-body problem was not possible.

"...it may happen that small differences in the initial conditions produce very great differences in the final phenomena. A small error in the former will produce an enormous error in the latter. Prediction becomes impossible..." – Henri Poincaré, 1892

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- ...until ~1960, when computer simulations of simple systems of differential equations (e.g. the Lorenz equations) showed that even very simple systems can become chaotic
- Edward Lorenz: "Chaos is when the present determines the future, but the approximate present does not approximately determine the future."

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- For chaotic motion, two trajectories that are initially arbitrarily close in phase space will diverge exponentially in time from each other
- All memory that the two trajectories started out close is lost
- Exponential divergence is key. For nonchaotic motion, nearby trajectories diverge at most *linearly* with time

What is Deterministic Chaos? (Problem #4 in HW)



Exponential divergence

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- Can we understand **exponential divergence**: how two trajectories diverge from a starting point?
- This will tell us:
 - 1. If the system is chaotic
 - 2. How long we are able to accurately predict the evolution of the system.
- This has many applications in real-world problems.
- For example, if the Solar System is chaotic, how long can we predict that the Earth is safe from being hit by a rogue planet/asteroid, or from being ejected from the Solar System?

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- ► We can write b(t) = a(t)+δ(t), where δ(t) is a vector with infinitesimal initial (t = 0) length.
- As the system evolves, we track $\delta(t)$.
- The maximal Lyapunov exponent of the system is the number λ, if it exists, such that:

$$|\delta(t)| \approx |\delta(0)| e^{\lambda t}$$

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- ► Local Lyapunov exponents estimate the local predictability around a point *x*⁰ in phase space. These are easier to calculate than the global Lyapunov exponents.
- ► The local Lyapunov exponents are the eigenvalues of the Jacobian of f (in x' = f(x, t)) at x_0 .

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As you saw in the homework, an arbitrary starting point can diverge greatly from a very close neighboring starting point.

• This is one of the criteria for chaos.

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- Is there any way to predict this, *analytically*, from the equations of the system?
- Let's explore this by first looking at *critical points* (also called *fixed points*).

Critical Points

$$\begin{aligned} x' &= \sigma(y - x), \\ y' &= x(\rho - z) - y, \\ z' &= xy - \beta z. \end{aligned}$$

Critical points are points in phase space that do not evolve. We will investigate the *stability* of fixed points because this will help us understand the general solutions of the system.

First, we find the **critical points** by setting x' = y' = z' = 0(time derivatives are zero \rightarrow no change in x, y, z with time).

Matlab code to find critical points (symbolic solution):

```
clearvars
2 close all
4 syms x y z xp yp zp
5 syms sigma positive
6 syms rho positive
7 syms beta positive
8
  [x_c, y_c, z_c, -x, conditions] = solve([0==sigma_*(y-x); 0==x*(rho_-z)-y; 0==x*y-beta_*z], ...
9
      'ReturnConditions', true, 'Real', true);
10
11
  disp([x_c,y_c,z_c,conditions]);
13
14 % >>lyapunov_example
                                 0.
                                                                     0. rho < 1 | 1 <= rho]
15 %
                                                                0.
      beta (1/2) * (rho - 1) (1/2), beta (1/2) * (rho - 1) (1/2), rho - 1,
16 %
                                                                              1 <= rho]
17 \% [-beta^{(1/2)}*(rho - 1)^{(1/2)}, -beta^{(1/2)}*(rho - 1)^{(1/2)}, rho - 1.
                                                                                     1 <= rhol
```
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Critical points (x' = y' = z' = 0): 1. (0, 0, 0)2. $(\sqrt{\beta(\rho - 1)}, \sqrt{\beta(\rho - 1)}, \rho - 1)$ 3. $(-\sqrt{\beta(\rho - 1)}, -\sqrt{\beta(\rho - 1)}, \rho - 1)$... if $\rho \ge 1$

...and only (0, 0, 0) for $\rho < 1$.

Stable: if we start close, but not exactly on, the critical point, the solution will evolve toward the fixed point.

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- We can explore this by calculating the Lyapunov exponents of the system at these critical points.
- (The Lyapunov exponents are the eigenvalues of the Jacobian of the system.)

Stability of Critical Points



- Consider a 1-D system: $\frac{dx}{dt} = f(x)$
- Critical points: f(x) = 0
- Jacobian J = f'(x) (slope of f(x))
- ► Near stable point x₀, if f(x) > 0 for x < x₀: dx/dt is also positive and you move to the right, back toward the critical point.
- ► ..., if *f*(*x*) < 0 for *x* > *x*₀: dx/dt is negative and you move to the left, also back toward the critical point.
- \rightarrow the critical point is **stable** if f'(x) < 0

Frame Title



- Consider a 1-D system: $\frac{dx}{dt} = f(x)$
- Critical points: f(x) = 0
- Jacobian J = f'(x) (slope of f(x))
- Stable critical points $\rightarrow f'(x) < 0$
- Multidimensional case: Negative eigenvalues of Jacobian → time evolution points back to critical point (*stable critical point*)
 Positive eigenvalues → points away from critical point (*unstable*)

critical point)

The Jacobian matrix of the Lorenz system at the critical point (0, 0, 0):

$$\mathbf{J} = \frac{d\mathbf{f}}{d\mathbf{x}} = \begin{bmatrix} \frac{\partial \mathbf{f}}{\partial x_1} & \cdots & \frac{\partial \mathbf{f}}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

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$$\mathbf{J_1} = \begin{bmatrix} -\sigma & \sigma & 0\\ \rho & -1 & 0\\ 0 & 0 & -\beta \end{bmatrix}$$

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Which has the characteristic polynomial (det $(\lambda I - J_1)$):

$$\lambda^{3} + (\sigma + 1 + \beta)\lambda^{2} + (\beta(\sigma + 1) + \sigma(1 - \rho))\lambda + \beta\sigma(1 - \rho)$$

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Eigenvalues (Lyapunov exponents) are the roots of this polynomial:

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Which has the characteristic polynomial (det $(\lambda I - J_1)$):

$$\lambda^{3} + (\sigma + 1 + \beta)\lambda^{2} + (\beta(\sigma + 1) + \sigma(1 - \rho))\lambda + \beta\sigma(1 - \rho)$$

Eigenvalues (Lyapunov exponents) are the roots of this polynomial: $-\beta, -\sigma/2 \pm \left(\sqrt{4\rho\sigma - 2\sigma + \sigma^2 + 1}\right)/2 - 1/2$

Matlab code to calculate eigenvalues of the Jacobian (Lyapunov exponents) at the critical point (0, 0, 0):

```
1 % Clear variables, close plots
2 clearvars
a close all
6 syms x y z xp yp zp lambda
7 syms sigma positive
syms rho_ positive
» syms beta_ positive
n % Calculate the Jacobian
J=jacobian([sigma_*(v-x);x*(rho_-z)-v;x*v-beta_*z],[x,v,z]);
# % Substitute the first critical point (0.0.0) into the Jacobian
IS J_1=subs(J, [x, y, z], [0, 0, 0])
17 % [ -sigma , sigma .
a % [
      rho_, -1,
10 % E
         0.
                   0. -beta 1
24 % Calculate the eigenvalues of this Jacobian
22 eigvals=eig(J 1):
24 %
                                                                    -heta
25 %
     - sigma /2 - (4*rho *sigma - 2*sigma + sigma ^2 + 1)^(1/2)/2 - 1/2
25 %
       (4*rho *sigma - 2*sigma + sigma ^2 + 1)^{(1/2)/2} - sigma /2 - 1/2
28 % Plot the three eigenvalues for sigma=10, beta=8/3, and a range of rho
20 rho vals=0:50:
30 eigvals_=zeros(numel(rho_vals),numel(eigvals));
m for ii=1:numel(rho_vals)
     eigvals_(ii,:)=double(subs(eigvals,[sigma_,beta_,rho_],[10,8./3.,rho_vals(ii)]);
n end
m plot(rho vals.eigvals )
x xlabel('\rho')
w vlabel('Eigenvalue')
a legend('Eigenvalue #1', 'Eigenvalue #2', 'Eigenvalue #3', 'Location', 'NorthWest')
m title('\sigma=10, \beta=8/3')
ø pdfname='eigenvals_lorenz_jacobian.pdf';
a print('-dpdf',pdfname);
a [~,~]=system(['pdfcrop ' pdfname ' ' pdfname]);
```

Calculating Lyapunov Exponent for (0,0,0), ρ =0–5



Calculating Lyapunov Exponent for (0,0,0), ρ =0–50



Does the Lyapunov Exponent predict the time-dependent deviation between an orbit starting at the initial value (0,0,0) and another orbit starting very close to (0,0,0)?

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Let's look at $\rho = 50$

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Let's look at $\rho = 50$

Reminder:

$$\begin{split} \delta(t) &= b(t) - a(t) \\ |\delta(t)| &\approx |\delta(0)| e^{\lambda t} \end{split}$$







Solutions to the Lorenz equations occupy a finite volume of phase space...



Solutions to the Lorenz equations occupy a finite volume of phase space, so the two trajectories cannot deviate arbitrarily far apart. **The Lyapunov exponent does not tell the whole story.**



Analogously, a double pendulum can appear to behave chaotically but its motion is still constrained by its Hamiltonian. It can only occupy a finite region of phase space.

Lyapunov exponents (eigenvalues of Jacobian):

$$\begin{split} \lambda_1 &= -\beta \\ \lambda_2 &= -\frac{\sigma}{2} + \frac{\sqrt{4\rho\sigma - 2\sigma + \sigma^2 + 1}}{2} - \frac{1}{2} \\ \lambda_3 &= -\frac{\sigma}{2} - \frac{\sqrt{4\rho\sigma - 2\sigma + \sigma^2 + 1}}{2} - \frac{1}{2} \end{split}$$

For $\rho < 1$ ($\sigma = 10$, $\beta = 8/3$) at critical point (0, 0, 0), all three Lyapunov exponents are negative: $|\delta(t)| \approx |\delta(0)|e^{\lambda t} \rightarrow$ no chaos. This means (0,0,0) is a stable critical point for $\rho < 1$.



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That is, the origin is a "sink" and all orbits with nearby starting points are drawn to it.

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REMINDER: Critical Points

Lorenz Equations:

$$\begin{aligned} x' &= \sigma(y-x), \\ y' &= x(\rho-z) - y, \\ z' &= xy - \beta z. \end{aligned}$$

Critical points (x' = y' = z' = 0): 1. (0, 0, 0)2. $(\sqrt{\beta(\rho - 1)}, \sqrt{\beta(\rho - 1)}, \rho - 1)$ 3. $(-\sqrt{\beta(\rho - 1)}, -\sqrt{\beta(\rho - 1)}, \rho - 1)$...if $\rho \ge 1$

...and only (0, 0, 0) for $\rho \leq 1$.

$$(\pm\sqrt{\beta(\rho-1)},\pm\sqrt{\beta(\rho-1)},\rho-1)=C^{\pm}?$$

What about the other two critical points:

$$(\pm\sqrt{\beta(\rho-1)},\pm\sqrt{\beta(\rho-1)},\rho-1)=C^{\pm}?$$

Are these stable (non-chaotic) or unstable (chaotic)?

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- When we do this, we find that for some values of *ρ*, the eigenvalues have an imaginary component.

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- Are these stable (non-chaotic) or unstable (chaotic)?
- ▶ We'll need to calculate their Lyapunov exponents...
- When we do this, we find that for some values of *ρ*, the eigenvalues have an imaginary component.
- There are also bifurcations and other interesting behavior.

Lyapunov Exponent of Lorenz Equations: C^{\pm}





What does a complex eigenvalue mean for the orbits?
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Recall (the deviation between two orbits):

 $|\delta(t)| \approx |\delta(0)| e^{\lambda t}$

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oscillatory orbits (they begin at ρ =1.346) with a frequency given by the imaginary component.

Types of Critical Points & Lyapunov Exponent

$$\lambda = A + Bi$$

| А | В | Туре |
|----------|----------|-------------------|
| Negative | Zero | Stable |
| Negative | Non-zero | Attractor |
| Positive | Zero | Chaotic |
| Positive | Non-zero | Strange Attractor |

- Stable: if we start close, but not exactly on, the critical point, the solution will evolve toward the fixed point.
- Attractor: A region in space that is invariant under the evolution of time and attracts most, if not all, nearby trajectories.
- Strange attractor: An attractor that displays chaotic behavior (i.e., high sensitivity to initial conditions).



 $\sigma=10,\,\beta=8/3,\,\rho=10.$ Orbits spiral towards either C^+ or $C^-.$



Plotting the eigenvalues for higher ρ , we see that at $\rho \approx 13.4$ two eigenvalues switch to each other's eigenvectors.

```
1 % Clear variables, close plots
2 clearvars:
3 close all
s % Declare symbolic variables
6 syms x y z xp yp zp lambda
7 syms sigma_ positive
s syms rho_ positive
syms beta_ positive
11 % Calculate the Jacobian
12 J=jacobian([sigma_*(y-x);x*(rho_-z)-y;x*y-beta_*z],[x,y,z]);
    Substitute the critical point C- into the Jacobian
14 %
_{15} J 1=subs(J,[x,y,z],[-sqrt(beta *(rho -1)),-sqrt(beta *(rho -1)), rho -1]);
17 % [ -sigma_, sigma_,
      rho_, -1,
18 %
19 % [
         0.
                  0. -beta 1
    Calculate the eigenvalues of this Jacobian
21 %
_{22} [eigvecsm, eigvalsm]=eig(J_1);
23 eigvals=diag(eigvalsm);
24 %
                                                                   -beta
     - sigma /2 - (4*rho *sigma - 2*sigma + sigma ^2 + 1)^(1/2)/2 - 1/2
25 %
26 %
       (4*rho_*sigma_ - 2*sigma_ + sigma_2 + 1)^{(1/2)/2} - sigma_2 - 1/2
    Plot the three eigenvalues for sigma=10, beta=8/3, and a range of rho
28 %
20 rho vals=linspace(0.15.100);
meigvals_=zeros(numel(rho_vals),numel(eigvals));
31 eigvecs_=zeros(numel(rho_vals),numel(eigvals),3);
x for ii=1:numel(rho vals)
     eigvals (ii.:)=double(subs(eigvals.[sigma .beta .rho ].[10.8./3..rho vals(ii)]):
     eigvecs_(ii,:::)=double(subs(eigvecsm,[sigma_,beta_,rho_],[10,8./3.,rho_vals(ii)]));
35 end
subplot(2.2.1)
model: max(rho vals)].[0.01, '--');
set(pp(4), 'Color', [0.5, 0.5, 0.5])
w xlabel('\rho')
4 vlabel('Eigenvalue')
g legend('Eigenvalue #1', 'Eigenvalue #2', 'Eigenvalue #3', 'Location', 'West')
s title('\sigma=10, \beta=8/3')
44
subplot(2.2.2)
# plot(rho vals.squeeze(eigvecs (:.1.[1.2.3])))

    title('Eigenvector #1')

# xlabel('\rho')
vlabel('Eigenvector Component')
```



Each switched eigenvector also changes direction, and instead, a point starting closer to C^+ will eventually settle at C^- (and vice-versa).



A point starting closer to C^+ will eventually settle at C^- (and vice-versa).



 ρ =13–24: the imaginary part of the eigenvalues increase, meaning the frequency of orbits around C^{\pm} increase. Solutions oscillate between C^+ and C^- many times before finally spiraling into them.



 ρ =13–24: the real part of the eigenvalues for C^{\pm} are still negative, so the solution will settle into either C^{\pm} eventually, but the time it takes to do so can vary sensitively on the initial conditions.

Problem #5

Problem #5: Find a value of ρ (while keeping $\sigma = 10$ and $\beta = 8/3$) such that the solution does not depend sensitively on the initial values.



 $\sigma = 10$, $\beta = 2.6667$

Solutions spirals into $C^{-} = (-7.1, -7.3, 19.0)$



Some of you found interesting behavior in this region. Setting ρ =23.001 and ρ =23.002 and integrating t = 0-30 appeared to straddle a transition into chaos.

Problem #5



But if we integrate out to longer times, we see that each of these ρ values *eventually* spirals into C^{\pm} (as they should because the real-part of the eigenvalues are still negative).

Problem #5



But if we integrate out to longer times, we see that each of these ρ values *eventually* spirals into C^{\pm} (as they should because the real-part of the eigenvalues are still negative).

But the time a solution takes to do so (and whether it goes to C^+ or C^-) can vary *hugely* with small changes in ρ , initial conditions, or integration time step! *This is a fascinating region!*

Problem #5: *ρ*=23.001100 & 23.001101



Problem #5: *ρ*=23.001100 & 23.001101





Around 24.74, one eigenvalue of C^{\pm} turns positive again (Hopf bifurcation), so these attractors become chaotic, and because they have nonzero imaginary components, the orbits are spirals.

σ=10.00, β=2.67, γ=28.00



In this regime C^{\pm} are strange attractors (displaying chaotic behavior). (This is the regime of the classic "Lorenz Attractor").

A limit cycle is an isolated closed trajectory. Limit cycles can be stable (nearby orbits spiral into it as $t \to \infty$)



...unstable (nearby orbits spiral into it as $t \to -\infty$)



...or semi-stable, for example if the limit cycle is stable for trajectories approaching from inside, but unstable for trajectories approaching from outside.



There are many surprises in the parameter space, and many cannot be easily predicted:

- ▶ Point and chaotic attractors for $24.06 < \rho < 24.74$
- Strange limit cycle behavior at Hopf bifurcation
- The return of a global attracting limit cycle for intervals at large ρ
- T(3,2) torus knot at $\rho = 99.96$
- Lots more...

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- Lots more...

Exploring the phase space of the Lorenz equations is like a walk in the jungle.

Attractor is a stable limit cycle for ρ =100



Lecture Outline

Lorenz Equations in an Analog Electronic Circuit

Homework Review

Chaotic Dynamics

Lyapunov Exponent

Lyapunov Exponent of the Lorenz Equations

One of the reasons King Oscar wanted to know the solution to the n-body problem was to determine if the Solar System is stable.



FIGURE 5. Eccentricity of a typical chaotic trajectory over a longer time interval. The time is now measured in millions of years. Bursts of high-eccentricity behaviour are interspersed with intervals of irregular low-eccentricity behaviour, proken by occasional spikes.

 Asteroids in orbital resonances with Jupiter can be sent into Earth-crossing orbits.

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- Pluto has a Lyapunov exponent of 1/20 Myr⁻¹. This implies a Lyapunov time scale of ~20 Myr before neighboring orbits diverge significantly (i.e. it's not possible to determine Pluto's orbit precisely ~100 Myr from now).

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- ► A rocket launch will change the Earth's position by ~0.5° after 100 Myr.
- But the Solar Sytem's dynamics are governed by a Hamiltonian, so chaos can only be local (a specific area of phase space).
- Arnold diffusion (nonconservation of invariant quantities) would allow for local chaos to diffuse throughout the phase space.
- Arnold diffusion would cause more catastrophic chaos, like ejection of planets, but the timescales for this *appear* to be large.