

edX Robo4 Mini MS – Locomotion Engineering
Block 1 – Week 2 – Unit 1
A Linear Time Invariant Mechanical System

Video 2.1
**Segment 1.2.1.1
Prismatic 1 DoF Physics**

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with
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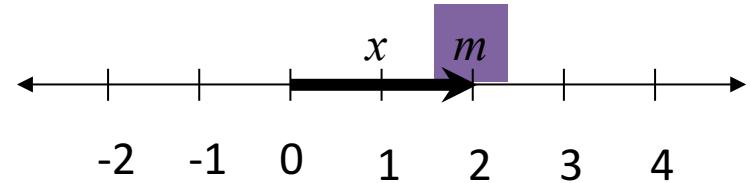
1 DoF (Prismatic) Kinematics

- Moving body with “mass” m
- Frame of Reference
 - coordinates
 - extension, x
- 1 DoF Kinematics
 - moving mass has velocity

(symbol \coloneqq means “defined as”)

- acceleration:
changes in velocity

(“overdots” denote time derivatives)

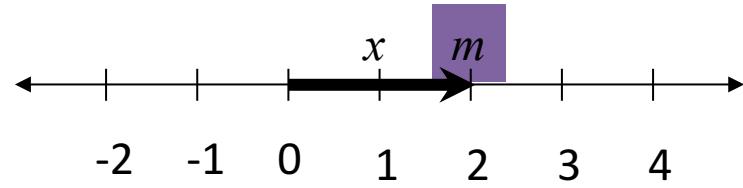


$$v(t) := \dot{x}(t) = \frac{d}{dt}x(t)$$

$$a(t) := \dot{v}(t) = \ddot{x}(t)$$

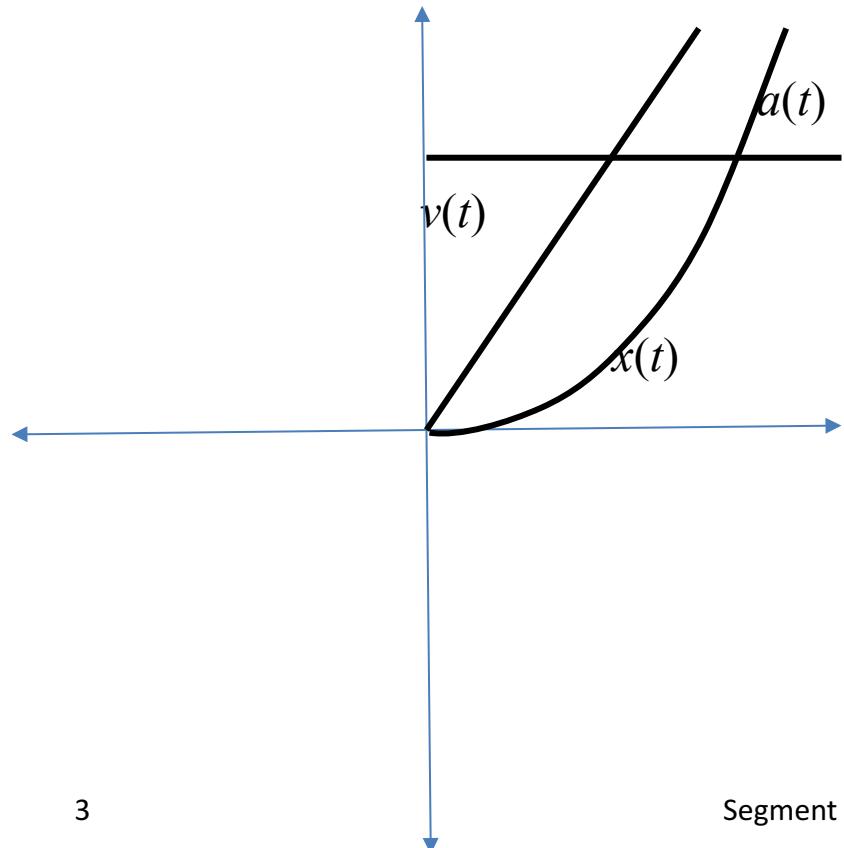
Infinitesimal 1 DoF Kinematics

- Relationships between extension, velocity, acceleration



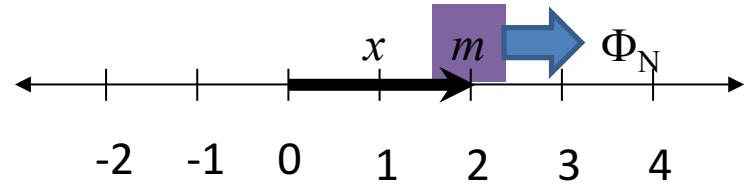
$$v(t) := \dot{x}(t) = \frac{d}{dt}x(t)$$

$$a(t) := \dot{v}(t) = \ddot{x}(t)$$



Newton's Law

- The mechanics of work
 - forces “act” on masses
 - Galileo: change velocities
 - Newton: change momentum
 - special case: position invariance
 - implies constant of proportionality
 - we call “mass”



$$m \ddot{x} = \Phi_N$$

ODEs from State-dependent forces

- forces often depend on motion

- e.g. viscous damping
- proportional (& opposed) to ν

$$\Phi_D(\dot{x}) := -b\dot{x}$$

- mechanics now gives
- an Ordinary Differential Equation (ODE)

$$m\ddot{x} = \Phi_D(\dot{x}) = -b\dot{x}$$

From ODE to Vector Field

- Aim: view ODEs $m\ddot{x} = -b\dot{x} =: \Phi_D(\dot{x})$ geometrically
 - rethink velocity as a “state”
 - to get a “first order” ODE (i.e., state variables isolated on LHS)
 - defined by a “vector field” (VF)
- $$v := \dot{x}$$
- $$\dot{v} = f_D(v)$$
- $$f_D(v) := \lambda v \quad (1)$$
- $$\lambda := -\frac{b}{m}$$

Solution of the Initial Condition Problem

$$e^{\lambda t} := \sum_{k=0}^{\infty} (\lambda t)^k / k!$$

$$\frac{d}{dt} e^{\lambda t} = \frac{d}{dt} \left(1 + \sum_{k=1}^{\infty} (\lambda t)^k / k! \right)$$

$$= \lambda \sum_{k=1}^{\infty} k (\lambda t)^{(k-1)} / k!$$

$$= \lambda e^{\lambda t} \stackrel{\checkmark}{=} f_D(e^{\lambda t})$$

Properties of the Exponential

- infinite series $e^{\lambda t}$
 - exists (converges) for all arguments
 - identity at $t = 0$
 - never vanishes
- so for each t
 - $e^{\lambda t}v_0$ is an invertible linear function
 - taking any initial velocity, v_0 to a new velocity, v_t
- and we get
 - a t -parametrized family
 - of invertible linear functions

$$f_D^t(v_0) := e^{\lambda t} v_0 \quad (2)$$

The Flow Generated by a Vector Field

- “flow”

- a t -parametrized family
- of invertible (generally nonlinear)
- functions of state

$$f_D^t(v_0) := e^{\lambda t} v_0 \quad (2)$$

- “generated” by VF $\dot{v} = f_D(v) \quad (1)$

- used for all “nice” VF

- reflecting those properties
- even when we don’t know how
- to write down the specific infinite series

Adding Compliance

- Robotics setting usually introduces compliance
 - (position dependent opposing force)
 - e.g., Hooke's law spring force

$$\Phi_{\text{HS}}(x) := -kx$$

$$\begin{aligned}\varphi_{\text{HS}}(x) &:= - \int_0^x \Phi_{\text{HS}}(\chi) d\chi \quad (3) \\ &= \frac{1}{2} kx^2\end{aligned}$$

- with associated potential

Newton Gives A 2nd Order ODE

- sum applied forces to change momentum getting new ODE

expressed as

- change in momentum
- or change in velocity

$$\begin{aligned} m\ddot{x} &= \Phi_D(\dot{x}) + \Phi_{HS}(x) \\ &= -b\dot{x} - kx \\ &=: \Phi_{DHO}(\mathbf{x}) \end{aligned} \tag{4}$$

$$\mathbf{x} := \begin{bmatrix} x \\ v \end{bmatrix} = \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

$$\ddot{\mathbf{x}} = \Xi_{DHO}(\mathbf{x}) := \frac{1}{m} \Phi_{DHO}(\mathbf{x})$$

DHO := “damped harmonic oscillator” (expecting “under-damped” gains)

Moving Ahead

- Seeking a VF

- given scalar
2nd order
ODE

$$\ddot{x} = \Xi_{\text{DHO}}(\mathbf{x})$$

- re-express
via vector, \mathbf{x}
- to get 1st
order VF

$$\mathbf{x} = \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

$$\dot{\mathbf{x}} = f_{\text{DHO}}(\mathbf{x})$$

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Segment 1.2.1.2 LTI Flow

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From Scalar 2nd Order to Vector 1st Order

- Recall Newton's 2nd order ODE $\ddot{x} = \Xi_{\text{DHO}}(\mathbf{x})$
- Now recast as 1st order vector ODE
 - using matrix calculus
 - d/dt operating on array
 - yields array of d/dt entries

$$= -\frac{k}{m}x - \frac{b}{m}\dot{x}$$
$$\mathbf{x} := \begin{bmatrix} x \\ \dot{x} \end{bmatrix}$$

$$\begin{aligned}\dot{\mathbf{x}} &= \begin{bmatrix} v \\ \Xi_{\text{DHO}}(\mathbf{x}) \end{bmatrix} \\ &= \begin{bmatrix} \dot{x} \\ -\frac{b}{m}\dot{x} - \frac{k}{m}x \end{bmatrix} =: f_{\text{DHO}}(\mathbf{x}) \quad (5) \\ &= A\mathbf{x}; \quad A := \begin{bmatrix} 0 & 1 \\ -\frac{k}{m} & -\frac{b}{m} \end{bmatrix}\end{aligned}$$

- good time to review calculus, matrix algebra
- soon start working with matrix calculus

Matrix Exponential via Diagonalization

- Notice VF is

LTI

("linear time invariant")

$$f_{\text{DHO}}(\mathbf{x}) = A\mathbf{x}$$

- Diagonalize

$$A = E\Lambda E^{-1}; \quad \Lambda = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

- Exponentiate $e^{At} = \sum_{k=0}^{\infty} \frac{1}{k!} (E\Lambda E^{-1} t)^k$

- exponent of diagonal matrix

$$\begin{aligned} &= E \left(\sum_{k=0}^{\infty} \frac{1}{k!} (\Lambda t)^k \right) E^{-1} \\ &= E \left(\sum_{k=0}^{\infty} \frac{1}{k!} \begin{bmatrix} (\lambda_1 t)^k & 0 \\ 0 & (\lambda_2 t)^k \end{bmatrix} \right) E^{-1} \end{aligned}$$

- is matrix of diagonal exponents

$$= E \begin{bmatrix} e^{t\lambda_1} & 0 \\ 0 & e^{t\lambda_2} \end{bmatrix} E^{-1}$$

Closed Form Solution of LTI VF

- Time derivative

- of matrix exponential

$$\begin{aligned}\frac{d}{dt} e^{At} &= \frac{d}{dt} \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \sum_{k=1}^{\infty} \frac{1}{k!} (E \Lambda E^{-1} t)^k \right) \\ &= \frac{d}{dt} E \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \sum_{k=1}^{\infty} \frac{1}{k!} (\Lambda t)^k \right) E^{-1} \\ &= E \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} E^{-1}.\end{aligned}$$

- satisfies ODE

$$\sum_{k=1}^{\infty} \frac{1}{(k-1)!} E \begin{bmatrix} (\lambda_1 t)^{(k-1)} & 0 \\ 0 & (\lambda_2 t)^{(k-1)} \end{bmatrix} E^{-1}$$

- Yields closed form solution

$$= A e^{At}$$

- for flow $f_{\text{DHO}}^t(\mathbf{x}_0) := e^{At} \mathbf{x}_0$ (6)

- reference:

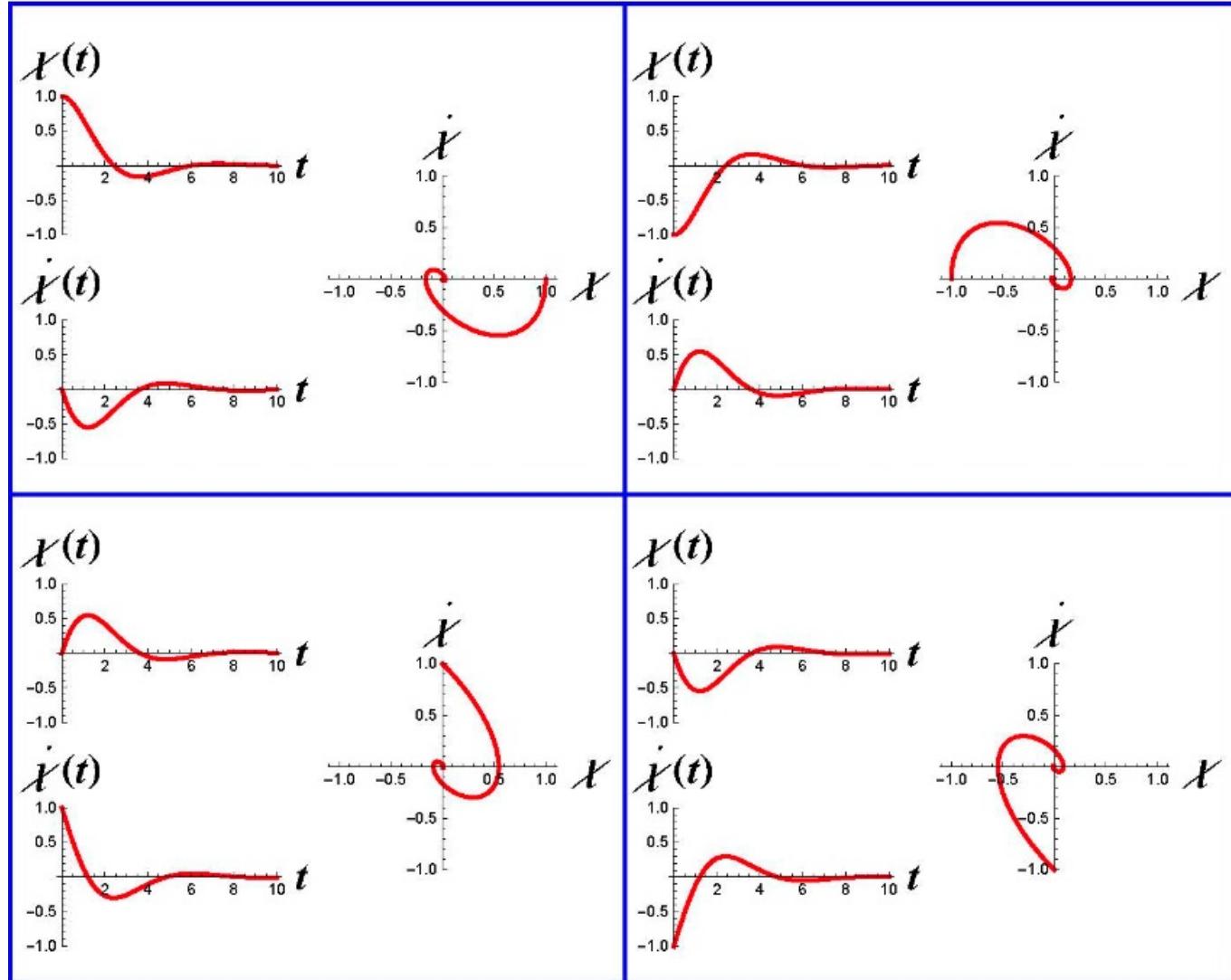
Compare (6) to eqn (2) in segment 1.2.1.1

- a time-parametrized family
 - of invertible linear functions

[M. W. Hirsch, S. Smale, and R. L. Devaney, Differential equations, dynamical systems and an introduction to chaos, vol. 60. Access Online via Elsevier, 2004]

Visualizing ODE Solutions

- 2nd order scalar
 - family of paired scalar-valued functions of time
 - parametrized by two ICs
 - for position & velocity
- 1st order vector
 - family of single vector-valued functions of time
 - parametrized by vector IC

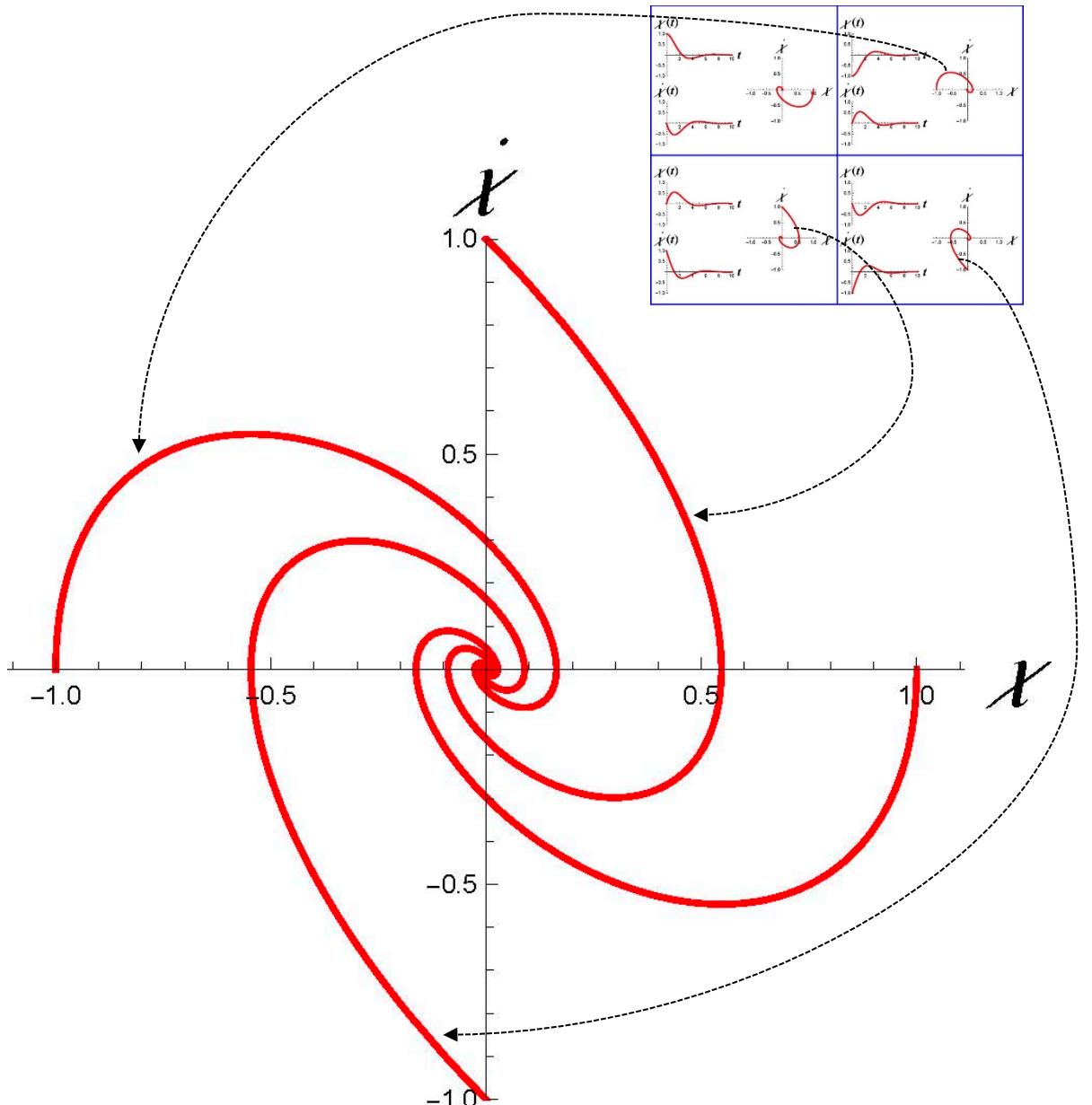


Solutions Through Four Different ICs

- 2nd order scalar view: paired scalar trajectories
- 1st order vector view: single vector trajectory

Introducing the Dynamical Systems View

- 2 dimensional VF
- generates 2 dimensional flow
 - family of vector-valued
 - transformations of vectors
 - parametrized by time



Robotics Virtue of Dynamical Systems View

- Family of (time-parametrized) state space transformations
- Helps in thinking about
 - Periodic Behavior
 - relates states at one phase
 - to states at any other phase
 - Asymptotic Behavior
 - never really know exact IC
 - rarely care about exact time trajectory
 - want “every” IC to “end up” in goal state(s)

Diagonalization as Change of Coordinates

$$\dot{\mathbf{x}} = f_{\text{DHO}}(\mathbf{x}) = A\mathbf{x}$$

- Diagonalization: what does it “mean”?

- it’s a change of coordinates

$$\mathbf{z} := E^{-1}\mathbf{x} =: h_{\text{diag}}(\mathbf{x})$$

$$\begin{aligned} \Rightarrow \dot{\mathbf{z}} &= E^{-1}\dot{\mathbf{x}} \\ &= E^{-1}f_{\text{DHO}}(\mathbf{x}) = E^{-1}f_{\text{DHO}}(E\mathbf{z}) \\ &= E^{-1}AE\mathbf{z} = \Lambda\mathbf{z} =: f_{\text{diag}}(\mathbf{z}) \end{aligned} \tag{7}$$

- yielding a new expression for the “same” VF

$$\dot{\mathbf{z}} = f_{\text{diag}}(\mathbf{z}) = \Lambda\mathbf{z}$$

Change of Coordinates for Conjugacy

- Change of coordinates (CC) to get flow
 - transform from x to z coordinates
 - write out (easy) expression for flow in z coordinates
 - then transform expression back into x coordinates
- Say that the two VFs are “conjugate,” $f_{\text{DHO}} \stackrel{h_{\text{diag}}}{\sim} f_{\text{diag}}$

Need for Different Coordinate Systems

- Value of new z coordinates
 - useful: easy to write down $\mathbf{x} \in \mathbb{R}^2; \mathbf{z} \in \mathbb{C}^2$ flow
 - awkward: introduce complex numbers $A \in \mathbb{R}^{2 \times 2}; E, \Lambda \in \mathbb{C}^{2 \times 2}$
- Consider using different coordinates
 - [Hirsch et al., 2004, Ch. 3.4]
 - substituting column sum and difference vectors
 - in place of original columns $\mathbf{y} := S^{-1}\mathbf{x} =: h_{\text{RC}}(\mathbf{x})$ of E
 - yields “real canonical” form
 - real coordinates and coefficients
 - useful expression for flow

$$S := E \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Real Canonical Coordinates

$$\dot{\mathbf{y}} = S^{-1} \dot{\mathbf{x}}$$

$$= S^{-1} f_{\text{DHO}}(\mathbf{x}) = S^{-1} f_{\text{DHO}}(S\mathbf{y})$$

$$= S^{-1} A S \mathbf{y} = \tilde{A} \mathbf{y} =: f_{\text{RC}}(\mathbf{y})$$

$$\tilde{A} := (\sigma I_2 + \omega J_2) \quad I_2 := \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}; \quad J_2 := \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

$$\sigma := \frac{1}{2} (\lambda_1 + \lambda_2) =: \Re(\lambda_i)$$

$$\omega := \frac{1}{2} i (\lambda_1 - \lambda_2) =: \Im(\lambda_i)$$

$$e^{t\tilde{A}} = e^{t(\sigma I_2 + \omega J_2)}$$

$$= e^{t\sigma} e^{t\omega J_2} = e^{t\sigma} \begin{bmatrix} \cos \omega t & -\sin \omega t \\ \sin \omega t & \cos \omega t \end{bmatrix}$$

Moving ahead

- need different coordinates
 - for different purposes
 - e.g. soon interpret flow as CC
- these examples were linear
 - have matrix representations
 - easy to compute in closed form
- soon need nonlinear CC
 - no matrix representation
 - get used to using functional names
 - e.g. $f_{\text{DHO}}(x)$ rather than Ax
 - e.g. $h_{\text{diag}}(x)$ rather than $E^{-1}x$

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Segment 1.2.1.3

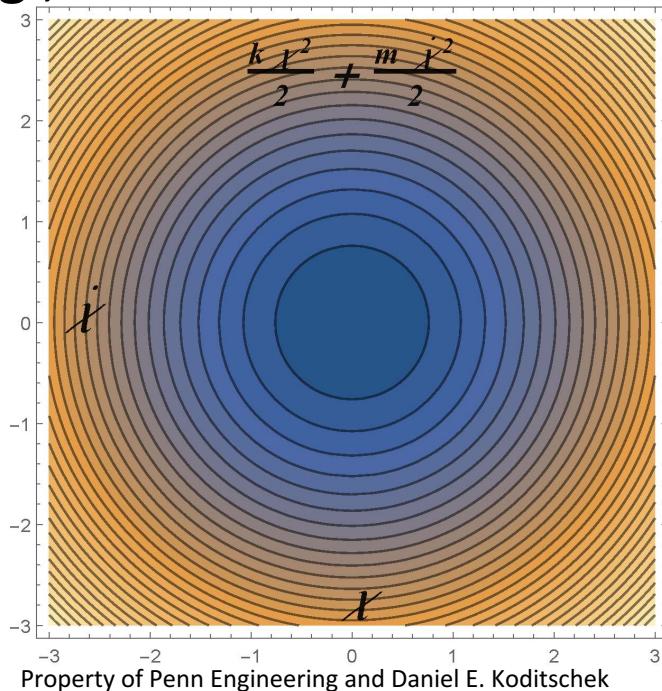
Energy, Power, Basins

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Total Energy as a Norm

- Recall energy components
 - kinetic energy
 - Hooke's law spring potential
 - Comprise **total energy**
- Now re-interpret

$$\eta_{\text{HO}}(\mathbf{x}) := \kappa(\dot{\mathbf{x}}) + \varphi_S(x)$$
 - as a norm
 - on the “phase plane”
 - e.g., for $k = m = 1$

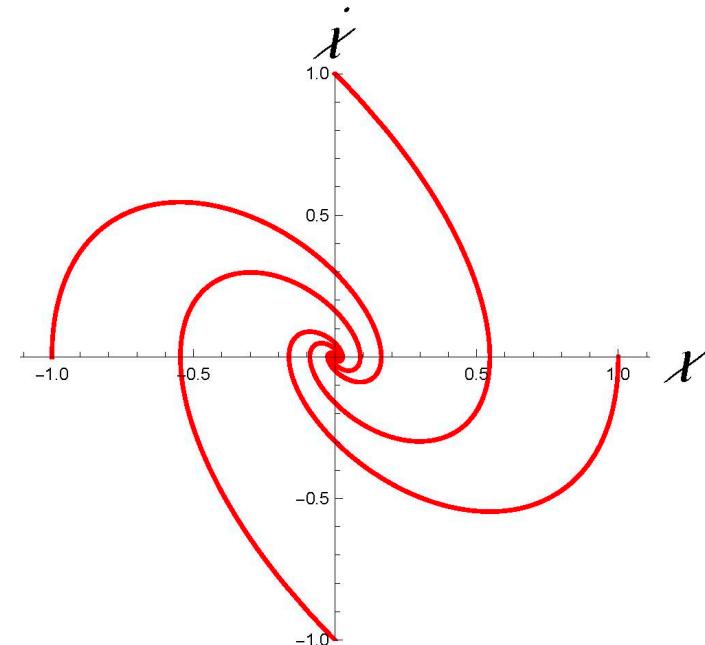
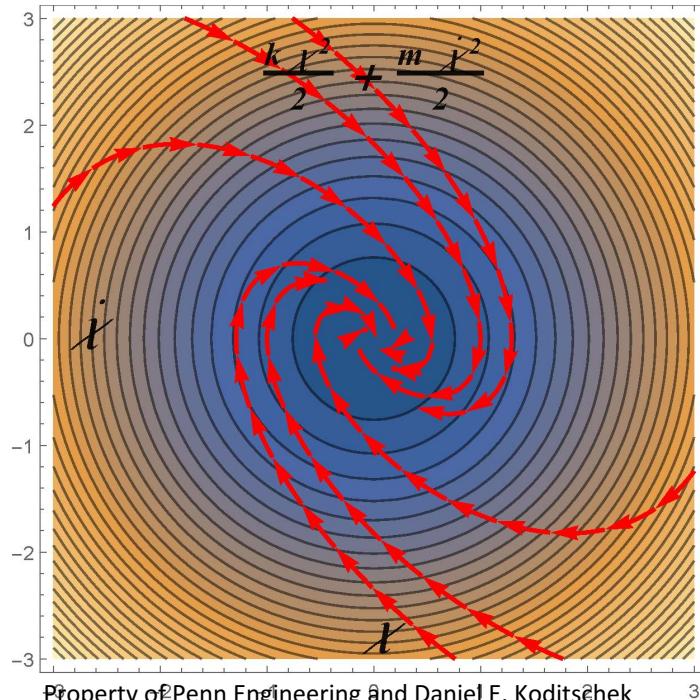


$$\begin{aligned}
 \kappa(\dot{\mathbf{x}}) &:= \frac{1}{2}m\dot{x}^2 \\
 \varphi_S(x) &:= \frac{1}{2}kx^2 \\
 \eta_{\text{HO}}(\mathbf{x}) &:= \kappa(\dot{\mathbf{x}}) + \varphi_S(x) \\
 &= \frac{1}{2} \left[(\sqrt{m}\dot{x})^2 + (\sqrt{k}x)^2 \right] \\
 &= \frac{1}{2} \left\| \begin{bmatrix} \sqrt{k}x \\ \sqrt{m}\dot{x} \end{bmatrix} \right\|^2 \\
 &= \frac{1}{2} \| P^{\frac{1}{2}} \mathbf{x} \|^2 \\
 P &:= \begin{bmatrix} k & 0 \\ 0 & m \end{bmatrix}
 \end{aligned}$$

Total Energy Along the Flow

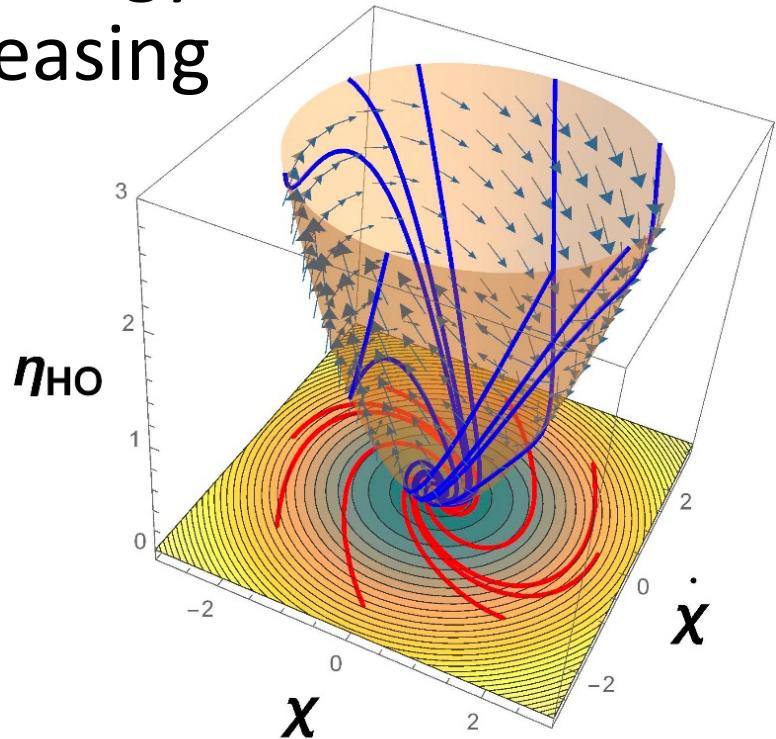
- A state trajectory
 - tracks the flow
 - from specific IC
- Visits energy levels
(function composition symbol)
- Yields energy trajectory

$$\begin{aligned} f_{\text{DHO}}^t(\mathbf{x}_0) &= e^{tA}\mathbf{x}_0 \\ \eta_{\text{HO}} \circ f_{\text{DHO}}^t(\mathbf{x}_0) &= \|P^{\frac{1}{2}}e^{tA}\mathbf{x}_0\|^2 \end{aligned}$$

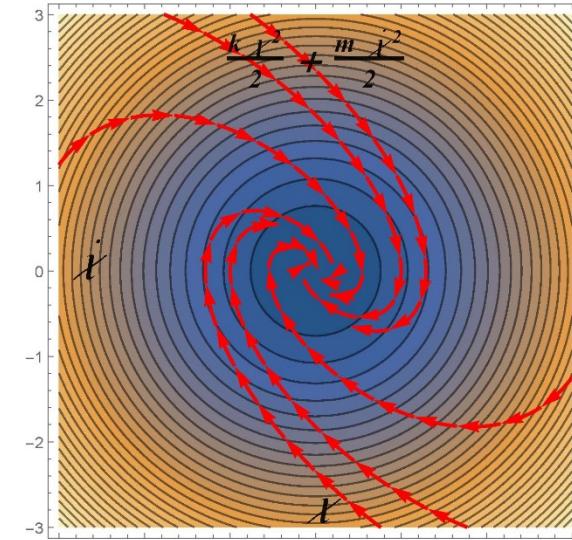


Total Energy is Non-increasing

- Energy trajectory
- Yields **power** trajectory
- Which is non-positive
(dissipated by damping)
- So energy is non-increasing

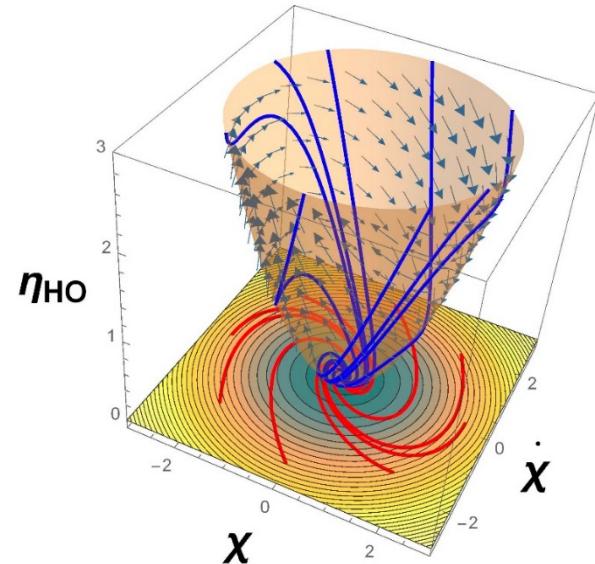


$$\begin{aligned}\dot{\eta}_{\text{HO}} &= \frac{d}{dt} \eta_{\text{HO}} \circ f_{\text{DHO}}^t(\mathbf{x}_0) \\ &= D_{\mathbf{x}} \eta_{\text{HO}} \cdot \frac{d}{dt} f_{\text{DHO}}^t(\mathbf{x}_0) \\ &= D_{\mathbf{x}} \eta_{\text{HO}} \cdot f_{\text{DHO}} \circ f_{\text{DHO}}^t(\mathbf{x}_0) \\ &= [\partial \varphi_S / \partial x, \partial \kappa / \partial \dot{x}] \cdot A f_{\text{DHO}}^t(\mathbf{x}_0) \\ &= [kx(t), m\dot{x}(t)] \left[-\frac{b}{m} \dot{x}(t) - \frac{k}{m} x(t) \right] \\ &= -b\dot{x}(t)^2\end{aligned}$$



The Basin of Energy Decay

- Conclude [Thomson & Tait, 1888]
 - since energy is norm-like
 - and is almost always decaying
 - trajectories must settle down to lowest energy state
- Given VF, see that power
 - is a function of state $\dot{\eta}_{\text{HO}}(\mathbf{x}) := D_{\mathbf{x}}\eta_{\text{HO}}(\mathbf{x}) \cdot f_{\text{DHO}}(\mathbf{x})$ (1)
 - just like energy
- Yields concept of basin
 - ICs whose motions decay
 - to low energy state
 - as predicted by negative power



solid mid-line dot

- denotes array multiplication
- here: $(1 \times 2) \otimes (2 \times 1)$