Lecture 21 Tue. 11.12.2015

Introduction to coupled oscillation and eigenmodes (Ch. 2-4 of Unit 4 11.12.15)

2D harmonic oscillator equations Lagrangian and matrix forms and Reciprocity symmetry
2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase



2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (4 1 3 2) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase









2D HO kinetic energy $T(v_1, v_2)$

$$T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$$



2D HO kinetic energy $T(v_1, v_2)$ $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$

 $2D \text{ HO potential energy } V(x_1, x_2)$ $V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}(x_1 - x_2)^2$ $= \frac{1}{2}(k_1 + k_{12})x_1^2 - k_{12}x_1x_2 + \frac{1}{2}(k_2 + k_{12})x_2^2$

Lagrangian L=*T*-*V*

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (4 1 3 2) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase



2D HO kinetic energy T(v₁, v₂) $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$

$$2D \text{ HO potential energy } V(x_1, x_2)$$

$$V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}(x_1 - x_2)^2$$

$$= \frac{1}{2}(k_1 + k_{12})x_1^2 - k_{12}x_1x_2 + \frac{1}{2}(k_2 + k_{12})x_2^2$$

2D HO Lagrange equations

$$\frac{d}{dt}\left(\frac{\partial T}{\partial \dot{x}_1}\right) = m_1 \ddot{x}_1 = F_1 = -\frac{\partial V}{\partial x_1} = -\left(k_1 + k_{12}\right)x_1 + k_{12}x_2$$
$$\frac{d}{dt}\left(\frac{\partial T}{\partial \dot{x}_2}\right) = m_2 \ddot{x}_2 = F_2 = -\frac{\partial V}{\partial x_2} = k_{12}x_1 - \left(k_2 + k_{12}\right)x_2$$



2D HO kinetic energy $T(v_1, v_2)$ $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$ V:

 $2D \text{ HO potential energy } V(x_1, x_2)$ $V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}(x_1 - x_2)^2$ $= \frac{1}{2}(k_1 + k_{12})x_1^2 - k_{12}x_1x_2 + \frac{1}{2}(k_2 + k_{12})x_2^2$

Lagrangian L=*T*-*V*

2D HO Lagrange equations

$$\frac{d}{dt} \left(\frac{\partial T}{\partial \dot{x}_1} \right) = m_1 \ddot{x}_1 = F_1 = -\frac{\partial V}{\partial x_1} = -\left(k_1 + k_{12}\right) x_1 + k_{12} x_2$$
$$\frac{d}{dt} \left(\frac{\partial T}{\partial \dot{x}_2} \right) = m_2 \ddot{x}_2 = F_2 = -\frac{\partial V}{\partial x_2} = k_{12} x_1 - \left(k_2 + k_{12}\right) x_2$$

2D HO Matrix operator equations

$$\begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix} \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$



2D HO kinetic energy T(v₁, v₂) $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$ $2D \text{ HO potential energy } V(x_1, x_2)$ $V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}(x_1 - x_2)^2$ $= \frac{1}{2}(k_1 + k_{12})x_1^2 - k_{12}x_1x_2 + \frac{1}{2}(k_2 + k_{12})x_2^2$

2D HO Lagrange equations

$$\frac{d}{dt}\left(\frac{\partial T}{\partial \dot{x}_1}\right) = m_1 \ddot{x}_1 = F_1 = -\frac{\partial V}{\partial x_1} = -\left(k_1 + k_{12}\right) x_1 + k_{12} x_2$$
$$\frac{d}{dt}\left(\frac{\partial T}{\partial \dot{x}_2}\right) = m_2 \ddot{x}_2 = F_2 = -\frac{\partial V}{\partial x_2} = k_{12} x_1 - \left(k_2 + k_{12}\right) x_2$$

2D HO Matrix operator equations

$$\begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix} \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Matrix operator notation:

$$\mathbf{M} \cdot |\ddot{\mathbf{x}}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$$



2D HO kinetic energy $T(v_1, v_2)$ $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$ $= \frac{1}{2}\langle \dot{\mathbf{x}} | \mathbf{M} | \dot{\mathbf{x}} \rangle$ 2D HO potential energy $V(x_1, x_2)$ $V = \frac{1}{2}(k_1 + k_{12})x_1^2 - k_{12}x_1x_2 + \frac{1}{2}(k_2 + k_{12})x_2^2$ $= \frac{1}{2}\langle \mathbf{x} | \mathbf{K} | \mathbf{x} \rangle$ where: $\mathbf{K} = \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix}$

2D HO Lagrange equations

$$\frac{d}{dt}\left(\frac{\partial T}{\partial \dot{x}_1}\right) = m_1 \ddot{x}_1 = F_1 = -\frac{\partial V}{\partial x_1} = -\left(k_1 + k_{12}\right)x_1 + k_{12}x_2$$
$$\frac{d}{dt}\left(\frac{\partial T}{\partial \dot{x}_2}\right) = m_2 \ddot{x}_2 = F_2 = -\frac{\partial V}{\partial x_2} = k_{12}x_1 - \left(k_2 + k_{12}\right)x_2$$

2D HO Matrix operator equations

$$\begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix} \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Matrix operator notation:

$$\mathbf{M} \cdot |\ddot{\mathbf{x}}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$$

Lagrangian L=*T*-*V*

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (4 1 3 2) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

Matrix equations and reciprocity symmetry

General form of 2D-HO equation of motion has force matrix components: $\kappa_{11} = k_1 + k_{11}$, $\kappa_{22} = k_2 + k_{22}$ $\begin{pmatrix} F_1 \\ F_2 \end{pmatrix} = \begin{pmatrix} m_1 \ddot{x}_1 \\ m_2 \ddot{x}_2 \end{pmatrix} = -\begin{pmatrix} \kappa_{11} & \kappa_{12} \\ \kappa_{21} & \kappa_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

Off-diagonal force constants satisfy *Reciprocity Relations*:

$$-\kappa_{12} = k_{12} = \frac{\partial F_1}{\partial x_2} = \frac{\partial^2 V}{\partial x_2 \partial x_1} = \frac{\partial^2 V}{\partial x_1 \partial x_2} = \frac{\partial F_2}{\partial x_1} = k_{21} = -\kappa_{21}$$

Rescaling and symmetrization

Each coordinate (x_1, x_2) is rescaled $(q_1 = s_1 x_1, q_2 = s_2 x_2)$ to symmetrize mass factors on \ddot{q}_j -terms.

$$-\frac{m_1}{s_1}\ddot{q}_1 = \kappa_{11}\frac{q_1}{s_1} + \kappa_{12}\frac{q_2}{s_2} \qquad -\ddot{q}_1 = \frac{\kappa_{11}}{m_1}q_1 + \frac{\kappa_{12}s_1}{m_1s_2}q_2 \equiv K_{11}q_1 + K_{12}q_2$$
$$-\frac{m_2}{s_2}\ddot{q}_2 = \kappa_{12}\frac{q_1}{s_1} + \kappa_{22}\frac{q_2}{s_2} \qquad -\ddot{q}_2 = \frac{\kappa_{12}s_2}{m_2s_1}q_1 + \frac{\kappa_{22}}{m_2}q_2 \equiv K_{21}q_1 + K_{22}q_2$$

New constants K_{ij} have pseudo-reciprocity symmetry for a special scale factor ratio: $\frac{s_2}{s_1} = \sqrt{\frac{m_2}{m_1}}$

$$\mathbf{K}_{21} = \frac{\kappa_{12}s_2}{m_2s_1} = \mathbf{K}_{12} = \frac{\kappa_{12}s_1}{m_1s_2} = \frac{-k_{12}}{\sqrt{m_1m_2}}$$

Diagonal constants
$$K_{jj}$$
 are not affected by scaling. To be equal requires: $\frac{\kappa_{11}}{m_1} = \frac{\kappa_{22}}{m_2}$ or: $\frac{\kappa_{11}}{\kappa_{22}} = \frac{m_1}{m_2}$
 $\kappa_{11} = \frac{\kappa_{11}}{m_1} = \frac{k_1 + k_{12}}{m_1}$ $\kappa_{22} = \frac{\kappa_{22}}{m_2} = \frac{k_2 + k_{12}}{m_2}$

Caution is advised since such forced symmetry may give modes with imaginary frequency.

2D harmonic oscillator equation eigensolutions Geometric method

Matrix-algebraic eigensolutions with example $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues \Rightarrow eigenvectors) Operator orthonormality and Completeness (Idempotent means: $\mathbf{P} \cdot \mathbf{P} = \mathbf{P}$) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

2D harmonic oscillator equation solutions

1. May rewrite equation $\mathbf{M} \cdot |\ddot{\mathbf{x}}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$ in acceleration matrix form: $|\ddot{\mathbf{x}}\rangle = -\mathbf{A}|\mathbf{x}\rangle$ where: $\mathbf{A} = \mathbf{M}^{-1} \cdot \mathbf{K}$

$$\begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix}^{-1} \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \end{pmatrix}$$

2. Need to find *eigenvectors* $|\mathbf{e}_1\rangle$, $|\mathbf{e}_2\rangle$,... of acceleration matrix such that: $\mathbf{A}|\mathbf{e}_n\rangle = \varepsilon_n|\mathbf{e}_n\rangle = \omega_n^2|\mathbf{e}_n\rangle$ Then equations decouple to: $|\mathbf{\ddot{e}}_n\rangle = -\mathbf{A}|\mathbf{e}_n\rangle = -\varepsilon_n|\mathbf{e}_n\rangle = -\omega_n^2|\mathbf{e}_n\rangle$ where ε_n is an *eigenvalue* and ω_n is an *eigenfrequency*

2D harmonic oscillator equation solutions

1. May rewrite equation $\mathbf{M} \cdot |\ddot{\mathbf{x}}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$ in acceleration matrix form: $|\ddot{\mathbf{x}}\rangle = -\mathbf{A}|\mathbf{x}\rangle$ where: $\mathbf{A} = \mathbf{M}^{-1} \cdot \mathbf{K}$

$$\begin{pmatrix} \ddot{x}_{1} \\ \ddot{x}_{2} \end{pmatrix} = - \begin{pmatrix} m_{1} & 0 \\ 0 & m_{2} \end{pmatrix}^{-1} \begin{pmatrix} k_{1} + k_{12} & -k_{12} \\ -k_{12} & k_{2} + k_{12} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} = - \begin{pmatrix} \frac{k_{1} + k_{12}}{m_{1}} & \frac{-k_{12}}{m_{1}} \\ \frac{-k_{12}}{m_{2}} & \frac{k_{2} + k_{12}}{m_{2}} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}$$

2. Need to find *eigenvectors* $|\mathbf{e}_1\rangle$, $|\mathbf{e}_2\rangle$,... of acceleration matrix such that: $\mathbf{A}|\mathbf{e}_n\rangle = \varepsilon_n|\mathbf{e}_n\rangle = \omega_n^2|\mathbf{e}_n\rangle$ Then equations decouple to: $|\mathbf{\ddot{e}}_n\rangle = -\mathbf{A}|\mathbf{e}_n\rangle = -\varepsilon_n|\mathbf{e}_n\rangle = -\omega_n^2|\mathbf{e}_n\rangle$ where ε_n is an *eigenvalue* and ω_n is an *eigenfrequency*

To introduce eigensolutions we take a simple case of unit masses $(m_1=1=m_2)$

So equation of motion is simply: $|\ddot{\mathbf{x}}\rangle = -\mathbf{K}|\mathbf{x}\rangle$

Eigenvectors $|\mathbf{x}\rangle = |\mathbf{e}_n\rangle$ are in special directions where $|\ddot{\mathbf{x}}\rangle = -\mathbf{K}|\mathbf{x}\rangle$ is in same direction as $|\mathbf{x}\rangle$

2D harmonic oscillator equation eigensolutions Geometric method

 ✓ Geometric method
 Matrix-algebraic eigensolutions with example M= (⁴ 1 / ³ 2) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase



Fig. 3.3.4 Plot of potential function $V(x_1, x_2)$ *showing elliptical* $V(x_1, x_2)$ *=const. level curves.*



Fig. 3.3.4 Plot of potential function $V(x_1,x_2)$ *showing elliptical* $V(x_1,x_2)$ *=const. level curves.*



Fig. 3.3.4 Plot of potential function $V(x_1,x_2)$ *showing elliptical* $V(x_1,x_2)$ *=const. level curves.*





Monday, November 16, 2015

 2D harmonic oscillator equation eigensolutions Geometric method
 Matrix-algebraic eigensolutions with example M= (⁴ 1 ₃ 2)
 Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**. $\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \cdots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_1 | \mathbf{M} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{\varepsilon}_1 | \mathbf{M} | \boldsymbol{\varepsilon}_2 \rangle & \cdots & \langle \boldsymbol{\varepsilon}_1 | \mathbf{M} | \boldsymbol{\varepsilon}_n \rangle \\ \langle \boldsymbol{\varepsilon}_2 | \mathbf{M} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{\varepsilon}_2 | \mathbf{M} | \boldsymbol{\varepsilon}_2 \rangle & \cdots & \langle \boldsymbol{\varepsilon}_2 | \mathbf{M} | \boldsymbol{\varepsilon}_n \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_n | \mathbf{M} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{\varepsilon}_n | \mathbf{M} | \boldsymbol{\varepsilon}_2 \rangle & \cdots & \langle \boldsymbol{\varepsilon}_n | \mathbf{M} | \boldsymbol{\varepsilon}_n \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_1 & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_n \end{pmatrix}$

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$



2D harmonic oscillator equation eigensolutions Geometric method Matrix-algebraic eigensolutions with example M= (4 1 3 2)
Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left(\boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

$$\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

Only possible non-zero $\{x, y\}$ if denominator is zero, too!

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

 $0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M})$

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M}|\boldsymbol{\varepsilon}_{k}\rangle = \boldsymbol{\varepsilon}_{k}|\boldsymbol{\varepsilon}_{k}\rangle, \text{ or: } (\mathbf{M}-\boldsymbol{\varepsilon}_{k}\mathbf{1})|\boldsymbol{\varepsilon}_{k}\rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left(\boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

Secular equation has *n*-factors, one for each eigenvalue.

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_1) (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_2) \cdots (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_n)$$

$$\mathbf{M}|\varepsilon\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \varepsilon \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det |\mathbf{M} - \varepsilon \cdot \mathbf{l}| = \det \begin{vmatrix} 4 & 1 \\ 3 & 2 \end{vmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = \det \begin{vmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{vmatrix}$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$
$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M}) = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let: $\varepsilon_1 = 1$ and: $\varepsilon_2 = 5$

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (4 1 3 2)
Secular equation
→ Hamilton-Cayley equation ← d projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left(\boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

Secular equation has *n*-factors, one for each eigenvalue.

det
$$|\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_1) (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_2) \cdots (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_n)$$

Each ε replaced by **M** and each ε_k by $\varepsilon_k \mathbf{1}$ gives *Hamilton-Cayley* matrix equation.

$$\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1}) (\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$$

Obviously true if **M** has diagonal form. (But, that's circular logic. Faith needed!)

$$\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M}) = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let: $\varepsilon_1 = 1$ and: $\varepsilon_2 = 5$

$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1 \cdot \mathbf{1})(\mathbf{M} - 5 \cdot \mathbf{1})$$
$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6 \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (⁴ 1 ₃ ²) Secular equation
→ Hamilton-Cayley equation and projectors
→ Hamilton-Cayley equation and projectors
→ Idempotent projectors (how eigenvalues ⇒ eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left(\boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

Secular equation has *n*-factors, one for each eigenvalue.

det
$$|\mathbf{M} - \varepsilon \mathbf{1}| = 0 = (-1)^n (\varepsilon - \varepsilon_1) (\varepsilon - \varepsilon_2) \cdots (\varepsilon - \varepsilon_n)$$

Each ε replaced by **M** and each ε_k by $\varepsilon_k \mathbf{1}$ gives *Hamilton-Cayley* matrix equation.

 $\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1})(\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$

Obviously true if **M** has diagonal form. (But, that's circular logic. Faith needed!)

Replace *j*th HC-factor by (1) to make *projection operators*

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - \varepsilon_{2}\mathbf{1})\cdots(\mathbf{M} - \varepsilon_{n}\mathbf{1})$$
$$\mathbf{p}_{2} = (\mathbf{M} - \varepsilon_{1}\mathbf{1})(\mathbf{1})\cdots(\mathbf{M} - \varepsilon_{n}\mathbf{1})$$
$$\vdots$$
$$\mathbf{p}_{n} = (\mathbf{M} - \varepsilon_{1}\mathbf{1})(\mathbf{M} - \varepsilon_{2}\mathbf{1})\cdots(\mathbf{1})$$

$$\mathbf{M}|\varepsilon\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \varepsilon \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M}) = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let: $\varepsilon_1 = 1$ and: $\varepsilon_2 = 5$

$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1})$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1}) = \begin{pmatrix} 4-5 & 1 \\ 3 & 2-5 \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix}$$

$$\mathbf{p}_{2} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{1}) = \begin{pmatrix} 4-1 & 1 \\ 3 & 2-1 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix}$$

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left(\boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

Secular equation has *n*-factors, one for each eigenvalue.

det
$$|\mathbf{M} - \varepsilon \mathbf{1}| = 0 = (-1)^n (\varepsilon - \varepsilon_1) (\varepsilon - \varepsilon_2) \cdots (\varepsilon - \varepsilon_n)$$

Each ε replaced by **M** and each ε_k by $\varepsilon_k \mathbf{1}$ gives *Hamilton-Cayley* matrix equation.

 $\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1}) (\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$

Obviously true if M has diagonal form. (But, that's circular logic. Faith needed!)

Replace j^{th} HC-factor by (1) to make projection operators $\mathbf{p}_{k} = \prod_{j \neq k} (\mathbf{M} - \varepsilon_{j} \mathbf{1})$ $\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - \varepsilon_{2} \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_{n} \mathbf{1})$ $\mathbf{p}_{2} = (\mathbf{M} - \varepsilon_{1} \mathbf{1})(-\mathbf{1}) \cdots (\mathbf{M} - \varepsilon_{n} \mathbf{1})$ \vdots $\mathbf{p}_{n} = (\mathbf{M} - \varepsilon_{1} \mathbf{1})(\mathbf{M} - \varepsilon_{2} \mathbf{1}) \cdots (-\mathbf{1})$ (Assume distinct e-values here: Non-degeneracy elause) $\varepsilon_{j} \neq \varepsilon_{k} \neq \dots$

Each \mathbf{p}_k contains *eigen-bra-kets* since: $(\mathbf{M} - \varepsilon_k \mathbf{1})\mathbf{p}_k = 0$ or: $\mathbf{M}\mathbf{p}_k = \varepsilon_k \mathbf{p}_k = \mathbf{p}_k \mathbf{M}$.

 $\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon}\begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M}) = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let: $\varepsilon_1 = 1$ and: $\varepsilon_2 = 5$

$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1})$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1}) = \begin{pmatrix} 4-5 & 1 \\ 3 & 2-5 \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix}$$

$$\mathbf{p}_{2} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{1}) = \begin{pmatrix} 4-1 & 1 \\ 3 & 2-1 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix}$$

$$\mathbf{Mp}_{1} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1 \cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1 \cdot \mathbf{p}_{1}$$

$$\mathbf{Mp}_{2} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5 \cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5 \cdot \mathbf{p}_{2}$$

An *eigenvector* $|\varepsilon_k\rangle$ of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 ε_k is *eigenvalue* associated with eigenvector $|\varepsilon_k\rangle$ direction. A change of basis to $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$ called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left(\boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

Secular equation has *n*-factors, one for each eigenvalue.

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_1) (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_2) \cdots (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_n)$$

Each ε replaced by **M** and each ε_k by $\varepsilon_k \mathbf{1}$ gives *Hamilton-Cayley* matrix equation. $\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1})(\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$

Obviously true if M has diagonal form. (But, that's circular logic. Faith needed!)

Replace *j*th HC-factor by (1) to make *projection operators* $\mathbf{p}_k = \prod_{j \neq k} (\mathbf{M} - \varepsilon_j \mathbf{1})$ $\mathbf{p}_1 = (1)(\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$

 $\mathbf{p}_{2} = (\mathbf{M} - \varepsilon_{1}\mathbf{1})(\mathbf{1})\cdots(\mathbf{M} - \varepsilon_{n}\mathbf{1}) \quad (\text{Assume <u>distinct</u>} e-values here:$ *Non-degeneracy clause*): $\varepsilon_{j} \neq \varepsilon_{k} \neq \dots$

 $\mathbf{p}_n = (\mathbf{M} - \boldsymbol{\varepsilon}_1 \mathbf{1}) (\mathbf{M} - \boldsymbol{\varepsilon}_2 \mathbf{1}) \cdots (\mathbf{1})$

Each \mathbf{p}_k contains *eigen-bra-kets* since: $(\mathbf{M} - \varepsilon_k \mathbf{1})\mathbf{p}_k = 0$ or: $\mathbf{M}\mathbf{p}_k = \varepsilon_k \mathbf{p}_k = \mathbf{p}_k \mathbf{M}$.

since \mathbf{M}^{1} , \mathbf{M}^{2} ,..commute with \mathbf{M} .

Notice \mathbf{p}_k *commutes with* $\mathbf{M}_{,...}$

 $\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M}) = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let: $\varepsilon_1 = 1$ and: $\varepsilon_2 = 5$

$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1 \cdot \mathbf{1})(\mathbf{M} - 5 \cdot \mathbf{1})$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - 5 \cdot \mathbf{1}) = \begin{pmatrix} 4 - 5 & 1 \\ 3 & 2 - 5 \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix}$$

$$\mathbf{p}_{2} = (\mathbf{M} - 1 \cdot \mathbf{1})(\mathbf{1}) = \begin{pmatrix} 4 - 1 & 1 \\ 3 & 2 - 1 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix}$$

$$\mathbf{Mp}_{1} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1 \cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1 \cdot \mathbf{p}_{1}$$

$$\mathbf{Mp}_{2} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5 \cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5 \cdot \mathbf{p}_{2}$$

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (⁴ 1 ₃ ²) Secular equation Hamilton-Cayley equation and projectors
→ Idempotent projectors (w eigenvalues ⇒ eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}} & \text{With example matrix} & \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{j} \mathbf{p}_{k} = \mathbf{p}_{j} \prod_{m \neq k} (\mathbf{M} - \varepsilon_{m} \mathbf{1}) = \prod_{m \neq k} (\mathbf{p}_{j} \mathbf{M} - \varepsilon_{m} \mathbf{p}_{j} \mathbf{1}) & \mathbf{M} \mathbf{p}_{k} = \varepsilon_{k} \mathbf{p}_{k} = \mathbf{p}_{k} \mathbf{M} \\ \text{Multiplication properties of } \mathbf{p}_{j} : \\ \mathbf{p}_{j} \mathbf{p}_{k} = \prod_{m \neq k} (\varepsilon_{j} \mathbf{p}_{j} - \varepsilon_{m} \mathbf{p}_{j}) = \mathbf{p}_{j} \prod_{m \neq k} (\varepsilon_{j} - \varepsilon_{m}) = \begin{cases} \mathbf{0} & \text{if } : j \neq k \\ \mathbf{p}_{k} \prod_{m \neq k} (\varepsilon_{k} - \varepsilon_{m}) & \text{if } : j = k \end{cases} & \mathbf{p}_{k} \mathbf{P}_{k} = k \end{aligned}$$

$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}} & \text{With example matrix} & \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{j}\mathbf{p}_{k} = \mathbf{p}_{j}\prod_{m\neq k} (\mathbf{M} - \varepsilon_{m}\mathbf{1}) = \prod_{m\neq k} (\mathbf{p}_{j}\mathbf{M} - \varepsilon_{m}\mathbf{p}_{j}\mathbf{1}) & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ \text{Multiplication properties of } \mathbf{p}_{j}: \\ \mathbf{p}_{j}\mathbf{p}_{k} = \prod_{m\neq k} (\varepsilon_{j}\mathbf{p}_{j} - \varepsilon_{m}\mathbf{p}_{j}) = \mathbf{p}_{j}\prod_{m\neq k} (\varepsilon_{j} - \varepsilon_{m}) = \begin{cases} \mathbf{0} & \text{if } : j \neq k \\ \mathbf{p}_{k}\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m}) & \text{if } : j = k \end{cases} \\ \mathbf{p}_{k}\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m}) & \text{if } : j = k \end{cases} \\ \text{Last step:} \\ \text{make Idempotent Projectors: } \mathbf{P}_{k} = \frac{\mathbf{P}_{k}}{\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m})} = \frac{\prod_{m\neq k} (\mathbf{M} - \varepsilon_{m}\mathbf{1})}{\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m})} \quad \mathbf{P}_{1} = \frac{(\mathbf{M} - \mathbf{5} \cdot \mathbf{1})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{1})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \end{aligned}$$
$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}} & \text{With example matrix} & \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{j}\mathbf{p}_{k} = \mathbf{p}_{j}\prod_{m\neq k} (\mathbf{M} - \varepsilon_{m}\mathbf{1}) = \prod_{m\neq k} (\mathbf{p}_{j}\mathbf{M} - \varepsilon_{m}\mathbf{p}_{j}\mathbf{1}) & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ \text{Multiplication properties of } \mathbf{p}_{j}: & \mathbf{p}_{j}\mathbf{p}_{k} = \prod_{m\neq k} (\varepsilon_{j}\mathbf{p}_{j} - \varepsilon_{m}\mathbf{p}_{j}) = \mathbf{p}_{j}\prod_{m\neq k} (\varepsilon_{j} - \varepsilon_{m}) = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{p}_{k}\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m}) & if: j = k \end{cases} & \mathbf{p}_{2} = (\mathbf{M} - \mathbf{1} \cdot \mathbf{1}) = \begin{pmatrix} -\mathbf{1} & \mathbf{1} \\ 3 & -3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - \mathbf{1} \cdot \mathbf{1}) = \begin{pmatrix} -\mathbf{1} & \mathbf{1} \\ 3 & -3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - \mathbf{1} \cdot \mathbf{1}) = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - \mathbf{1} \cdot \mathbf{1}) = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - \mathbf{1} \cdot \mathbf{1}) = \begin{pmatrix} -\mathbf{1} & \mathbf{1} \\ 3 & 1 \end{pmatrix} \\ \mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \mathbf{p}_{3} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{4} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{p}_{5} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ -3 & 1 \end{pmatrix} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} \cdot \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} - \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} - \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} - \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} - \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} - \mathbf{1} \\ \mathbf{M} = \begin{pmatrix} \mathbf{M} - \mathbf{1} - \mathbf{1} \\ \mathbf{M}$$

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (4 1 3 2) Secular equation Hamilton-Cayley equation and projectors
→ Idempotent projectors (how eigenvalues⇒eigenvectors)
Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar \partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus Classical 2D-HO: $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ Hamilton-Pauli spinor symmetry (ABCD-Types)

$$\begin{aligned} & \text{Matrix-algebraic method for finding eigenvector and eigenvalues} \\ & \mathbf{p}_{j}\mathbf{p}_{k} = \mathbf{p}_{j}\prod_{m\neq k} (\mathbf{M}-\varepsilon_{m}\mathbf{1}) = \prod_{m\neq k} (\mathbf{p}_{j}\mathbf{M}-\varepsilon_{m}\mathbf{p}_{j}\mathbf{1}) \\ & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ & \mathbf{p}_{1} = (\mathbf{M}-5\cdot\mathbf{1}) = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ 3 & 3 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{$$

$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}}_{\mathbf{p}_{1}\mathbf{p}_{n}} & \text{With example matrix} & \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{1}\mathbf{p}_{n} = \mathbf{p}_{1}\prod_{max}(\mathbf{M} - \varepsilon_{m}\mathbf{1}) = \prod_{max}(\mathbf{p}_{1}\mathbf{M} - \varepsilon_{m}\mathbf{p}_{1}\mathbf{1}) & \mathbf{M}\mathbf{p}_{1} = \varepsilon_{n}\mathbf{p}_{1} = \mathbf{p}_{n}\mathbf{M} \\ \text{Multiplication properties of } \mathbf{p}_{1}: \\ \mathbf{p}_{1}\mathbf{p}_{n} = \prod_{max}(\varepsilon_{n}\mathbf{p}_{n} - \varepsilon_{m}\mathbf{p}_{n}) = \mathbf{p}_{1}\prod_{max}(\varepsilon_{1} - \varepsilon_{m}) & \text{if } : j \neq k \\ \text{make Idempotent Projectors: } \mathbf{p}_{n} = \prod_{max}(\varepsilon_{n} - \varepsilon_{m}) & \text{if } : j = k \\ \mathbf{p}_{1}\prod_{max}(\varepsilon_{n} - \varepsilon_{m}) & \text{if } : j \neq k \\ \mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} \mathbf{0} & \text{if } : j \neq k \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \mathbf{p}_{1} = (\varepsilon_{n} - \varepsilon_{m}) & \text{if } : j = k \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) \\ \text{idempotent means: } \mathbf{P} \cdot \mathbf{P} = \mathbf{P} \end{pmatrix} \\ \mathbf{p}_{1}\mathbf{p}_{1} = \varepsilon_{n}\mathbf{p}_{n} = \begin{bmatrix} \mathbf{P}_{n} \\ \mathbf{P}_{1} = (\mathbf{P}_{n} - \varepsilon_{m}) \\ \mathbf{P}_{1}\mathbf{p}_{2} = \begin{pmatrix} \mathbf{0} & \text{if } : j \neq k \\ \mathbf{p}_{2} & \text{if } : j = k \\ \mathbf{P}_{1}\mathbf{p}_{2} = \varepsilon_{n}\mathbf{p}_{n} = \mathbf{P}_{n}\mathbf{P}_{n} \\ \text{implies: } \\ \mathbf{P}_{1}\mathbf{p}_{1} = \varepsilon_{n}\mathbf{p}_{n} = \mathbf{P}_{n}\mathbf{P}_{n} \\ \mathbf{P}_{1}\mathbf{p}_{2} = \mathbf{P}_{n}\mathbf{P}_{n} = (\mathbf{P}_{n} - \varepsilon_{n}) \\ \mathbf{P}_{1}\mathbf{p}_{2}\mathbf{p}$$

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (⁴ 1 ₃ 2) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors)
→ Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar \partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus Classical 2D-HO: $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ Hamilton-Pauli spinor symmetry (ABCD-Types)





2D harmonic oscillator equation eigensolutions Geometric method Matrix-algebraic eigensolutions with example M= $\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒ eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) ✓ Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors 2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed π/2 phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus Classical 2D-HO: $\partial_t^2 \mathbf{x} = -\mathbf{K} \cdot \mathbf{x}$ Hamilton-Pauli spinor symmetry (ABCD-Types)





2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M= (4 1 3 2) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions
✓ Functional spectral decomposition
✓ Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)*







2D harmonic oscillator equation eigensolutions *Geometric method Matrix-algebraic eigensolutions with example* $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors *Idempotent projectors (how eigenvalues* \Rightarrow *eigenvectors) Operator orthonormality and Completeness (Idempotent means:* **P**·**P**=**P**) Spectral Decompositions Functional spectral decomposition - Orthonormality vs. Completeness vis-a'-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors 2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase 2D-HO eigensolution example with asymmetric (A-Type) symmetry

Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)*



Orthonormality vs. Completeness vis-a`-vis Operator vs. State Operator expressions for orthonormality appear quite different from expressions for completeness. $\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n}$ Orthonormality vs. Completeness vis-a`-vis Operator vs. State Operator expressions for orthonormality appear quite different from expressions for completeness. $\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n}$

 $|\varepsilon_{j}\rangle\langle\varepsilon_{j}|\varepsilon_{k}\rangle\langle\varepsilon_{k}|=\delta_{jk}|\varepsilon_{k}\rangle\langle\varepsilon_{k}| \text{ or: } \langle\varepsilon_{j}|\varepsilon_{k}\rangle=\delta_{jk} \qquad \mathbf{1}=|\varepsilon_{1}\rangle\langle\varepsilon_{1}|+|\varepsilon_{2}\rangle\langle\varepsilon_{2}|+...+|\varepsilon_{n}\rangle\langle\varepsilon_{n}|$

Orthonormality vs. Completeness vis-a`-vis Operator vs. State

Operator expressions for orthonormality appear quite different from expressions for completeness.

$$\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} \\ \epsilon_{j} \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | = \delta_{jk} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \text{or:} \quad \langle \epsilon_{j} | \epsilon_{k} \rangle = \delta_{jk} \qquad \mathbf{1} = |\epsilon_{1} \rangle \langle \epsilon_{1} | + |\epsilon_{2} \rangle \langle \epsilon_{2} | + \dots + |\epsilon_{n} \rangle \langle \epsilon_{n} \rangle \\ \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | = \delta_{jk} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \text{or:} \quad \langle \epsilon_{j} | \epsilon_{k} \rangle = \delta_{jk} \qquad \mathbf{1} = |\epsilon_{1} \rangle \langle \epsilon_{1} | + |\epsilon_{2} \rangle \langle \epsilon_{2} | + \dots + |\epsilon_{n} \rangle \langle \epsilon_{n} \rangle \\ \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \text{or:} \quad \langle \epsilon_{j} | \epsilon_{k} \rangle = \delta_{jk} \qquad \mathbf{1} = |\epsilon_{j} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{n} \rangle \\ \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \text{or:} \quad \langle \epsilon_{j} | \epsilon_{k} \rangle = \delta_{jk} \qquad \mathbf{1} = |\epsilon_{j} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{n} \rangle \\ \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \text{or:} \quad \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \mathbf{1} = |\epsilon_{k} \rangle \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \mathbf{1} = |\epsilon_{k} \rangle \langle \epsilon_{k} | \epsilon_{k} \rangle \langle \epsilon_{k} | \epsilon_{$$

State vector representations of orthonormality are quite similar to representations of completeness. Like 2-sides of the same coin.

 $\{|x\rangle, |y\rangle\} \text{-orthonormality with } \{|\varepsilon_1\rangle, |\varepsilon_2\rangle\} \text{-completeness}$ $\langle x|y\rangle = \delta_{x,y} = \langle x|\mathbf{1}|y\rangle = \langle x|\varepsilon_1\rangle \langle \varepsilon_1|y\rangle + \langle x|\varepsilon_2\rangle \langle \varepsilon_2|y\rangle.$

 $\{|\varepsilon_{I}\rangle, |\varepsilon_{2}\rangle\} \text{-orthonormality with } \{|x\rangle, |y\rangle\} \text{-completeness}$ $\langle \varepsilon_{i}|\varepsilon_{j}\rangle = \delta_{i,j} = \langle \varepsilon_{i}|\mathbf{1}|\varepsilon_{j}\rangle = \langle \varepsilon_{i}|x\rangle\langle x|\varepsilon_{j}\rangle + \langle \varepsilon_{i}|y\rangle\langle y|\varepsilon_{j}\rangle$

Orthonormality vs. Completeness vis-a`-vis Operator vs. State

Operator expressions for orthonormality appear quite different from expressions for completeness.

$$\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} \\ \epsilon_{j} \langle \epsilon_{j} | \epsilon_{k} \rangle \langle \epsilon_{k} | = \delta_{jk} | \epsilon_{k} \rangle \langle \epsilon_{k} | \quad \text{or:} \quad \langle \epsilon_{j} | \epsilon_{k} \rangle = \delta_{jk} \qquad \mathbf{1} = |\epsilon_{1} \rangle \langle \epsilon_{1} | + |\epsilon_{2} \rangle \langle \epsilon_{2} | + \dots + |\epsilon_{n} \rangle \langle \epsilon_{n} \rangle$$

State vector representations of orthonormality are quite **similar** to representations of completeness. Like 2-sides of the same coin.

 $\{|x\rangle, |y\rangle\} \text{-orthonormality with } \{|\varepsilon_{I}\rangle, |\varepsilon_{2}\rangle\} \text{-completeness} \\ \langle x|y\rangle = \delta_{x,y} = \langle x|\mathbf{1}|y\rangle = \langle x|\varepsilon_{1}\rangle\langle \varepsilon_{1}|y\rangle + \langle x|\varepsilon_{2}\rangle\langle \varepsilon_{2}|y\rangle. \\ \langle x|y\rangle = \delta(x,y) = \psi_{I}(x)\psi_{I}^{*}(y) + \psi_{2}(x)\psi_{2}^{*}(y) + ... \\ \text{Dirac } \delta\text{-function} \\ \{|\varepsilon_{I}\rangle, |\varepsilon_{2}\rangle\} \text{-orthonormality with } \{|x\rangle, |y\rangle\} \text{-completeness} \\ \langle \varepsilon_{i}|\varepsilon_{j}\rangle = \delta_{i,j} = \langle \varepsilon_{i}|\mathbf{1}|\varepsilon_{j}\rangle = \langle \varepsilon_{i}|x\rangle\langle x|\varepsilon_{j}\rangle + \langle \varepsilon_{i}|y\rangle\langle y|\varepsilon_{j}\rangle$

However Schrodinger wavefunction notation $\psi(x) = \langle x | \psi \rangle$ shows quite a difference...

Orthonormality vs. Completeness vis-a`-vis Operator vs. State

Operator expressions for orthonormality appear quite different from expressions for completeness.

$$\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} \\ \varepsilon_{j} \rangle \langle \varepsilon_{j} | \varepsilon_{k} \rangle \langle \varepsilon_{k} | = \delta_{jk} | \varepsilon_{k} \rangle \langle \varepsilon_{k} | \quad \text{or:} \quad \langle \varepsilon_{j} | \varepsilon_{k} \rangle = \delta_{jk} \qquad \mathbf{1} = |\varepsilon_{1} \rangle \langle \varepsilon_{1} | + |\varepsilon_{2} \rangle \langle \varepsilon_{2} | + \dots + |\varepsilon_{n} \rangle \langle \varepsilon_{n} \rangle \langle \varepsilon_{n} \rangle$$

State vector representations of orthonormality are quite **similar** to representations of completeness. Like 2-sides of the same coin.

> $\{|x\rangle, |y\rangle\} \text{-orthonormality with } \{|\varepsilon_{I}\rangle, |\varepsilon_{2}\rangle\} \text{-completeness}$ $\langle x|y\rangle = \delta_{x,y} = \langle x|\mathbf{1}|y\rangle = \langle x|\varepsilon_{1}\rangle\langle\varepsilon_{1}|y\rangle + \langle x|\varepsilon_{2}\rangle\langle\varepsilon_{2}|y\rangle.$ $\langle x|y\rangle = \delta(x,y) = \psi_{I}(x)\psi_{I}^{*}(y) + \psi_{2}(x)\psi_{2}^{*}(y) + ...$ $Dirac \delta\text{-function}$ $\{|\varepsilon_{I}\rangle, |\varepsilon_{2}\rangle\} \text{-orthonormality with } \{|x\rangle, |y\rangle\} \text{-completeness}$ $\langle \varepsilon_{i}|\varepsilon_{j}\rangle = \delta_{i,j} = \langle \varepsilon_{i}|\mathbf{1}|\varepsilon_{j}\rangle = \langle \varepsilon_{i}|x\rangle\langle x|\varepsilon_{j}\rangle + \langle \varepsilon_{i}|y\rangle\langle y|\varepsilon_{j}\rangle$ $\langle \varepsilon_{i}|\varepsilon_{j}\rangle = \delta_{i,j} = ... + \psi_{i}^{*}(x)\psi_{j}(x) + \psi_{2}(y)\psi_{2}^{*}(y) + ... \rightarrow \int dx\psi_{i}^{*}(x)\psi_{j}(x)$

However Schrodinger wavefunction notation $\psi(x) = \langle x | \psi \rangle$ shows quite a difference... ...particularly in the orthonormality integral.

2D harmonic oscillator equation eigensolutions *Geometric method Matrix-algebraic eigensolutions with example* $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors *Idempotent projectors (how eigenvalues* \Rightarrow *eigenvectors) Operator orthonormality and Completeness (Idempotent means:* **P**·**P**=**P**) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors 2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase 2D-HO eigensolution example with asymmetric (A-Type) symmetry

Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)* A Proof of Projector Completeness (Truer-than-true by Lagrange interpolation) Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points x_1, x_2, \ldots, x_N .

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

A Proof of Projector Completeness (Truer-than-true) Compare matrix completeness relation and functional spectral decompositions

 $\mathbf{1} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{\substack{j \neq k}}^{N} (x - x_j)}{\prod_{\substack{j \neq k}}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$.

A Proof of Projector Completeness (Truer-than-true) Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$. A Proof of Projector Completeness (Truer-than-true) Compare matrix completeness relation and functional spectral decompositions

 $\mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{\substack{j \neq k}} (x - x_j)}{\prod_{\substack{j \neq k}} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\varepsilon_{k}} \mathbf{P}_{k} = \sum_{\varepsilon_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \varepsilon_{m} \mathbf{1})}{\prod_{m \neq k} (\varepsilon_{k} - \varepsilon_{m})} \qquad f(\mathbf{M}) = f(\varepsilon_{1})\mathbf{P}_{1} + f(\varepsilon_{2})\mathbf{P}_{2} + \dots + f(\varepsilon_{n})\mathbf{P}_{n} = \sum_{\varepsilon_{k}} f(\varepsilon_{k})\mathbf{P}_{k} = \sum_{\varepsilon_{k}} f(\varepsilon_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \varepsilon_{m} \mathbf{1})}{\prod_{m \neq k} (\varepsilon_{k} - \varepsilon_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line,



Compare matrix completeness relation and functional spectral decompositions $\mathbf{1} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1}) \mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2}) \mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n}) \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line, two separate points uniquely determine a sloping line,



Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{I} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line, two separate points uniquely determine a sloping line,

three separate points iniquely determine a parabola, etc.



Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{\substack{j \neq k}} (x - x_j)}{\prod_{\substack{j \neq k}} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line, two separate points uniquely determine a sloping line, three separate points uniquely determine a parabola, etc.

Lagrange interpolation formula \rightarrow *Completeness formula* as $x \rightarrow M$ and as $x_k \rightarrow \varepsilon_k$ and as $P_k(x_k) \rightarrow P_k$

Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line, two separate points uniquely determine a sloping line, three separate points uniquely determine a parabola, etc.

Lagrange interpolation formula \rightarrow *Completeness formula* as $x \rightarrow M$ and as $x_k \rightarrow \varepsilon_k$ and as $P_k(x_k) \rightarrow P_k$

All distinct values $\varepsilon_1 \neq \varepsilon_2 \neq ... \neq \varepsilon_N$ satisfy $\Sigma \mathbf{P}_k = \mathbf{1}$.

Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{I} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1}) \mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2}) \mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n}) \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line, two separate points uniquely determine a sloping line, three separate points uniquely determine a parabola, etc.

Lagrange interpolation formula \rightarrow *Completeness formula* as $x \rightarrow M$ and as $x_k \rightarrow \varepsilon_k$ and as $P_k(x_k) \rightarrow P_k$

All distinct values $\varepsilon_1 \neq \varepsilon_2 \neq \dots \neq \varepsilon_N$ satisfy $\Sigma \mathbf{P}_k = \mathbf{1}$. Completeness is *truer than true* as is seen for N = 2.

$$\mathbf{P}_{1} + \mathbf{P}_{2} = \frac{\prod_{j \neq 1} \left(\mathbf{M} - \varepsilon_{j}\mathbf{1}\right)}{\prod_{j \neq 1} \left(\varepsilon_{1} - \varepsilon_{j}\right)} + \frac{\prod_{j \neq 1} \left(\mathbf{M} - \varepsilon_{j}\mathbf{1}\right)}{\prod_{j \neq 1} \left(\varepsilon_{2} - \varepsilon_{j}\right)} = \frac{\left(\mathbf{M} - \varepsilon_{2}\mathbf{1}\right)}{\left(\varepsilon_{1} - \varepsilon_{2}\right)} + \frac{\left(\mathbf{M} - \varepsilon_{1}\mathbf{1}\right)}{\left(\varepsilon_{2} - \varepsilon_{1}\right)} = \frac{\left(\mathbf{M} - \varepsilon_{2}\mathbf{1}\right) - \left(\mathbf{M} - \varepsilon_{1}\mathbf{1}\right)}{\left(\varepsilon_{1} - \varepsilon_{2}\right)} = \frac{-\varepsilon_{2}\mathbf{1} + \varepsilon_{1}\mathbf{1}}{\left(\varepsilon_{1} - \varepsilon_{2}\right)} = \mathbf{1} \text{ (for all } \varepsilon_{j})$$

Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{I} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k}) \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

with *Lagrange interpolation formula* of function f(x) approximated by its value at N points $x_1, x_2, ..., x_N$.

$$L(f(x)) = \sum_{k=1}^{N} f(x_k) P_k(x) \quad \text{where: } P_k(x) = \frac{\prod_{j \neq k}^{N} (x - x_j)}{\prod_{j \neq k}^{N} (x_k - x_j)}$$

Each polynomial term $P_m(x)$ has zeros at each point $x=x_j$ except where $x=x_m$. Then $P_m(x_m)=1$. So at each of these points this L-approximation becomes exact: $L(f(x_j))=f(x_j)$.

If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

One point determines a constant level line, two separate points uniquely determine a sloping line, three separate points uniquely determine a parabola, etc.

Lagrange interpolation formula \rightarrow *Completeness formula* as $x \rightarrow M$ and as $x_k \rightarrow \varepsilon_k$ and as $P_k(x_k) \rightarrow P_k$

All distinct values $\varepsilon_1 \neq \varepsilon_2 \neq \dots \neq \varepsilon_N$ satisfy $\Sigma \mathbf{P}_k = \mathbf{1}$. Completeness is *truer than true* as is seen for N = 2. $\mathbf{P}_1 + \mathbf{P}_2 = \frac{\prod_{j \neq 1} (\mathbf{M} - \varepsilon_j \mathbf{1})}{\prod_{j \neq 1} (\varepsilon_1 - \varepsilon_j)} + \frac{\prod_{j \neq 1} (\mathbf{M} - \varepsilon_j \mathbf{1})}{(\varepsilon_1 - \varepsilon_2)} + \frac{(\mathbf{M} - \varepsilon_1 \mathbf{1})}{(\varepsilon_2 - \varepsilon_1)} = \frac{(\mathbf{M} - \varepsilon_2 \mathbf{1}) - (\mathbf{M} - \varepsilon_1 \mathbf{1})}{(\varepsilon_1 - \varepsilon_2)} = \frac{-\varepsilon_2 \mathbf{1} + \varepsilon_1 \mathbf{1}}{(\varepsilon_1 - \varepsilon_2)} = \mathbf{1} \text{ (for all } \varepsilon_j)$

However, only *select* values ε_k work for eigen-forms $\mathbf{MP}_k = \varepsilon_k \mathbf{P}_k$ or orthonormality $\mathbf{P}_j \mathbf{P}_k = \delta_{jk} \mathbf{P}_k$.

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M=(4 1) Secular equation (3 2)
Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors)
Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition
Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula
Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)* Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors. $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{l})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\left(\frac{1}{2} & -\frac{1}{2}\right)}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$ $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{l})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\left(\frac{3}{2} & \frac{1}{2}\right)}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle\langle\boldsymbol{\varepsilon}_{2}|$ Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors. $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{i})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\left(\frac{1}{2} & -\frac{1}{2}\right)}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$ $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{i})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\left(\frac{3}{2} & \frac{1}{2}\right)}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle\langle\boldsymbol{\varepsilon}_{2}|$

Load distinct bras $\langle \varepsilon_1 |$ and $\langle \varepsilon_2 |$ into d-tran rows, kets $|\varepsilon_1 \rangle$ and $|\varepsilon_2 \rangle$ into <u>inverse</u> d-tran columns.
Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors. $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{l})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\left(\frac{1}{2} & -\frac{1}{2}\right)}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$ $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{l})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\left(\frac{3}{2} - \frac{1}{2}\right)}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle\langle\boldsymbol{\varepsilon}_{2}|$

Load distinct bras $\langle \varepsilon_1 |$ and $\langle \varepsilon_2 |$ into d-tran rows, kets $|\varepsilon_1 \rangle$ and $|\varepsilon_2 \rangle$ into inverse d-tran columns.

$$\left\{ \left\langle \boldsymbol{\varepsilon}_{1} \right| = \left(\begin{array}{cc} \frac{1}{2} & -\frac{1}{2} \end{array} \right), \left\langle \boldsymbol{\varepsilon}_{2} \right| = \left(\begin{array}{cc} \frac{3}{2} & \frac{1}{2} \end{array} \right) \right\} , \quad \left\{ \left| \boldsymbol{\varepsilon}_{1} \right\rangle = \left(\begin{array}{cc} \frac{1}{2} \\ -\frac{3}{2} \end{array} \right), \left| \boldsymbol{\varepsilon}_{2} \right\rangle = \left(\begin{array}{cc} \frac{1}{2} \\ \frac{1}{2} \end{array} \right) \right\} \right\}$$

 $\begin{array}{c} (\boldsymbol{\varepsilon}_{1},\boldsymbol{\varepsilon}_{2}) \leftarrow (1,2) \ d\text{-Tran matrix} \\ \begin{pmatrix} \left\langle \boldsymbol{\varepsilon}_{1} \middle| x \right\rangle & \left\langle \boldsymbol{\varepsilon}_{1} \middle| y \right\rangle \\ \left\langle \boldsymbol{\varepsilon}_{2} \middle| x \right\rangle & \left\langle \boldsymbol{\varepsilon}_{2} \middle| y \right\rangle \end{array} \end{pmatrix} = \left(\begin{array}{c} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{array} \right) \\ \begin{pmatrix} \left\langle x \middle| \boldsymbol{\varepsilon}_{1} \right\rangle & \left\langle x \middle| \boldsymbol{\varepsilon}_{2} \right\rangle \\ \left\langle y \middle| \boldsymbol{\varepsilon}_{1} \right\rangle & \left\langle y \middle| \boldsymbol{\varepsilon}_{2} \right\rangle \end{array} \right) = \left(\begin{array}{c} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{array} \right) \\ \end{array}$

 $(\varepsilon_{1}, \varepsilon_{2}) \leftarrow (1, 2) \ d-Tran \ matrix$ $(1, 2) \leftarrow (\varepsilon_{1}, \varepsilon_{2}) \ \text{INVERSE} \ d-Tran \ matrix$ $\begin{pmatrix} \langle \varepsilon_{1} | x \rangle & \langle \varepsilon_{1} | y \rangle \\ \langle \varepsilon_{2} | x \rangle & \langle \varepsilon_{2} | y \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} , \quad \begin{pmatrix} \langle x | \varepsilon_{1} \rangle & \langle x | \varepsilon_{2} \rangle \\ \langle y | \varepsilon_{1} \rangle & \langle y | \varepsilon_{2} \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix}$ Use Dirac labeling for all components so transformation is OK $\begin{pmatrix} \langle \varepsilon_{1} | x \rangle & \langle \varepsilon_{1} | y \rangle \\ \langle \varepsilon_{2} | x \rangle & \langle \varepsilon_{2} | y \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle x | \mathbf{K} | x \rangle & \langle x | \mathbf{K} | y \rangle \\ \langle y | \mathbf{K} | x \rangle & \langle y | \mathbf{K} | y \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle x | \varepsilon_{1} \rangle & \langle x | \varepsilon_{2} \rangle \\ \langle y | \varepsilon_{1} \rangle & \langle y | \varepsilon_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \varepsilon_{1} | \mathbf{K} | \varepsilon_{1} \rangle & \langle \varepsilon_{1} | \mathbf{K} | \varepsilon_{2} \rangle \\ \langle \varepsilon_{2} | \mathbf{K} | \varepsilon_{1} \rangle & \langle \varepsilon_{2} | \mathbf{K} | \varepsilon_{2} \rangle \end{pmatrix}$ $\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 5 \end{pmatrix}$

Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors. $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{I})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \end{pmatrix}}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$ $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{1})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{3}{2} & \frac{1}{2} \end{pmatrix}}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle \langle \boldsymbol{\varepsilon}_{2}|$ Load distinct bras $\langle \varepsilon_1 |$ and $\langle \varepsilon_2 |$ into d-tran rows, kets $|\varepsilon_1 \rangle$ and $|\varepsilon_2 \rangle$ into <u>inverse</u> d-tran columns. $\left\{ \left\langle \boldsymbol{\varepsilon}_{1} \right| = \left(\begin{array}{cc} \frac{1}{2} & -\frac{1}{2} \end{array} \right), \left\langle \boldsymbol{\varepsilon}_{2} \right| = \left(\begin{array}{cc} \frac{3}{2} & \frac{1}{2} \end{array} \right) \right\}, \quad \left\{ \left| \boldsymbol{\varepsilon}_{1} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ -\frac{3}{2} \end{array} \right|, \left| \boldsymbol{\varepsilon}_{2} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ \frac{1}{2} \end{array} \right| \right\}$ $(\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2) \leftarrow (1, 2) d$ -Tran matrix $(1,2) \leftarrow (\varepsilon_1, \varepsilon_2)$ INVERSE *d*-Tran matrix $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} , \quad \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_2 \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_2 \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix}$ Use Dirac labeling for all components so transformation is OK $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}$ $\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} \qquad \cdot \qquad \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \qquad \cdot \qquad \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix} \qquad = \qquad \begin{pmatrix} 1 & 0 \\ 0 & 5 \end{pmatrix}$ Check inverse-d-tran is really inverse of your d-tran. $\begin{array}{c|c} \langle \boldsymbol{\varepsilon}_1 | 1 \rangle & \langle \boldsymbol{\varepsilon}_1 | 2 \rangle \\ \langle \boldsymbol{\varepsilon}_2 | 1 \rangle & \langle \boldsymbol{\varepsilon}_2 | 2 \rangle \end{array} \end{array} \right) \cdot \left(\begin{array}{c} \langle 1 | \boldsymbol{\varepsilon}_1 \rangle & \langle 1 | \boldsymbol{\varepsilon}_2 \rangle \\ \langle 2 | \boldsymbol{\varepsilon}_1 \rangle & \langle 2 | \boldsymbol{\varepsilon}_2 \rangle \end{array} \right) = \left(\begin{array}{c} \langle \boldsymbol{\varepsilon}_1 | 1 | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{\varepsilon}_1 | 1 | \boldsymbol{\varepsilon}_2 \rangle \\ \langle \boldsymbol{\varepsilon}_2 | 1 | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{\varepsilon}_2 | 1 | \boldsymbol{\varepsilon}_2 \rangle \end{array} \right)$

Monday, November 16, 2015

 $\begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{bmatrix} \quad \cdot \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{bmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors. $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{I})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \end{pmatrix}}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$ $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{1})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{3}{2} & \frac{1}{2} \end{pmatrix}}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle\langle\boldsymbol{\varepsilon}_{2}|$ Load distinct bras $\langle \varepsilon_1 |$ and $\langle \varepsilon_2 |$ into d-tran rows, kets $|\varepsilon_1 \rangle$ and $|\varepsilon_2 \rangle$ into <u>inverse</u> d-tran columns. $\left\{ \left\langle \boldsymbol{\varepsilon}_{1} \right| = \left(\begin{array}{cc} \frac{1}{2} & -\frac{1}{2} \end{array} \right), \left\langle \boldsymbol{\varepsilon}_{2} \right| = \left(\begin{array}{cc} \frac{3}{2} & \frac{1}{2} \end{array} \right) \right\}, \quad \left\{ \left| \boldsymbol{\varepsilon}_{1} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ -\frac{3}{2} \end{array} \right|, \left| \boldsymbol{\varepsilon}_{2} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ \frac{1}{2} \end{array} \right| \right\}$ $(\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2) \leftarrow (1, 2) d$ -Tran matrix $(1,2) \leftarrow (\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2)$ INVERSE *d*-Tran matrix $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} , \quad \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_2 \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_2 \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix}$ Use Dirac labeling for all components so transformation is OK $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}$ $\left(\begin{array}{ccc} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{array}\right) \qquad \cdot \qquad \left(\begin{array}{ccc} 4 & 1 \\ 3 & 2 \end{array}\right) \qquad \cdot \qquad \left(\begin{array}{ccc} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{array}\right) \qquad = \qquad \left(\begin{array}{ccc} 1 & 0 \\ 0 & 5 \end{array}\right)$ Check inverse-d-tran is really inverse of your d-tran. In standard quantum matrices inverses are "easy" $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{1} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{1} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{z}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{z} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \\ \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix} \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} | \boldsymbol{z} \rangle & \langle \boldsymbol{z} | \boldsymbol{z} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{z} |$

Monday, November 16, 2015

2D harmonic oscillator equations

Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method Matrix-algebraic eigensolutions with example M=(4 1 Secular equation (3 2) Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a'-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry \checkmark Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)*



Eigenvalues K_k and squared eigenfrequencies $\omega_0(\varepsilon_k)^2$ $K_1 = \omega_0^2(\varepsilon_1) = 9$, $K_2 = \omega_0^2(\varepsilon_2) = 11$,



$$\mathbf{P}_{1} = \frac{\begin{pmatrix} K_{11} - K_{2} & K_{12} \\ K_{12} & K_{22} - K_{2} \end{pmatrix}}{K_{1} - K_{2}} = \frac{\begin{pmatrix} 10 - 11 & -1 \\ -1 & 10 - 11 \end{pmatrix}}{9 - 11} = \frac{\begin{pmatrix} 1 & +1 \\ +1 & 1 \end{pmatrix}}{2} \qquad \qquad \mathbf{P}_{2} = \frac{\begin{pmatrix} K_{11} - K_{1} & K_{12} \\ K_{12} & K_{22} - K_{1} \end{pmatrix}}{K_{2} - K_{1}} = \frac{\begin{pmatrix} 10 - 9 & -1 \\ -1 & 10 - 9 \end{pmatrix}}{11 - 9} = \frac{\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}}{2}$$



Eigenbra vectors: $\langle \varepsilon_1 | = (1/\sqrt{2} + 1/\sqrt{2}), \langle \varepsilon_2 | = (1/\sqrt{2} - 1/\sqrt{2})$



Eigenbra vectors:
$$\langle \varepsilon_1 | = (1/\sqrt{2} + 1/\sqrt{2}), \langle \varepsilon_2 | = (1/\sqrt{2} - 1/\sqrt{2})$$

Mixed mode dynamics

$$\begin{aligned} |x(t)\rangle &= |\varepsilon_1\rangle \quad \langle \varepsilon_1 | x(0) \rangle e^{-i\omega_1 t} + |\varepsilon_2\rangle \quad \langle \varepsilon_2 | x(0) \rangle e^{-i\omega_2 t} \\ \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} &= \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \langle \varepsilon_1 | x(0) \rangle e^{-i\omega_1 t} + \begin{pmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \langle \varepsilon_2 | x(0) \rangle e^{-i\omega_2 t} \end{aligned}$$





Monday, November 16, 2015

Videos of Coupled Pendula aided by Overhead Projector





View on YouTube YouTube

Launch embedded videos using your browser/App or ⇐ view on YouTube ⇒



Stronger coupling on the right, illustrated indirectly by a darker looking spring on screen



BoxIt (Beating) Web Simulation



BoxIt (Beating) Web Simulation



BoxIt (Beating) Web Simulation

2D harmonic oscillator equations

Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M=(4 1) Secular equation (3 2)
Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors)
Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition
Orthonormality vs. Completeness vis-a`-vis Operator vs. State
Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)*

2D-HO beats and mixed mode geometry



2D-HO beats and mixed mode geometry



2D-HO beats and mixed mode geometry



2D-HO beats and mixed mode geometry



2D harmonic oscillator equations

Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M=(4 1 Secular equation (3 2) Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)*



 $Det(\mathbf{K}) = 7.13 - 27 = 91 - 27 = 64$ $Trace(\mathbf{K}) = 7 + 13 = 20$









 $\mathbf{P}_{1} = \frac{\begin{pmatrix} K_{11} - K_{2} & K_{12} \\ K_{12} & K_{22} - K_{2} \end{pmatrix}}{K_{1} - K_{2}} = \frac{\begin{pmatrix} 7 - 16 & -3\sqrt{3} \\ -3\sqrt{3} & 13 - 16 \end{pmatrix}}{4 - 16} = \frac{\begin{pmatrix} 9 & +3\sqrt{3} \\ +3\sqrt{3} & 3 \end{pmatrix}}{12}$ $= \frac{\begin{pmatrix} 3 & \sqrt{3} \\ \sqrt{3} & 1 \end{pmatrix}}{4} = \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix} (\sqrt{3}/2 & 1/2) = |\varepsilon_{1}\rangle\langle\varepsilon_{1}|$



$$= \frac{\begin{pmatrix} 3 & \sqrt{3} \\ \sqrt{3} & 1 \\ 4 & = \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix} \begin{pmatrix} \sqrt{3}/2 & 1/2 \\ 1/2 \end{pmatrix} (\sqrt{3}/2 & 1/2 \end{pmatrix} = |\varepsilon_1\rangle\langle\varepsilon_1|$$

$$= \frac{\begin{pmatrix} 1 & -\sqrt{3} \\ -\sqrt{3} & 3 \\ 4 & = \begin{pmatrix} -1/2 \\ \sqrt{3}/2 \\ \sqrt{3}/2 \end{pmatrix} (-1/2 & \sqrt{3}/2 \end{pmatrix} = |\varepsilon_2\rangle\langle\varepsilon_2|$$

$$= \frac{\begin{pmatrix} 1 & -\sqrt{3} \\ -\sqrt{3} & 3 \\ 4 & = \begin{pmatrix} -1/2 \\ \sqrt{3}/2 \\ \sqrt{3}/2 \end{pmatrix} (-1/2 & \sqrt{3}/2 \end{pmatrix} = |\varepsilon_2\rangle\langle\varepsilon_2|$$



2D harmonic oscillator equations

Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method Matrix-algebraic eigensolutions with example M=(4 1 Secular equation (3 2) Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus *Classical 2D-HO:* $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)*













BoxIt Web Simulation


2D harmonic oscillator equations

Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M=(4 1 Secular equation (3 2) Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed $\pi/2$ phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

→ ANALOGY: 2-State Schrodinger: $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ versus Classical 2D-HO: $\partial_t^2 \mathbf{x} = -\mathbf{K} \cdot \mathbf{x}$ Hamilton-Pauli spinor symmetry (ABCD-Types)

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \left(\begin{array}{cc} A & B - iC \\ B + iC & D \end{array}\right) = \mathbf{H}^{\dagger}$$

 H_{jk} matrix must obey: $(H_{jk})^* = H_{kj}$





to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of <u>real</u> 1st-order differential equations.







First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $\left|\Psi\right\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of <u>real</u> 1st-order differential equations.

 $\dot{x}_1 = Ap_1 + Bp_2 - Cx_2 \qquad \dot{p}_1 = -Ax_1 - Bx_2 - Cp_2 \\ \dot{x}_2 = Bp_1 + Dp_2 + Cx_1 \qquad \dot{p}_2 = -Bx_1 - Dx_2 + Cp_1$

Then start with classical Hamiltonian. (Designed to give same result.)

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1}x_{2} + p_{1}p_{2} \right) + C \left(x_{1}p_{2} - x_{2}p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $\left|\Psi\right\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of real real 1st-order differential equations.

$$\dot{x}_1 = Ap_1 + Bp_2 - Cx_2 \qquad \dot{p}_1 = -Ax_1 - Bx_2 - Cp_2 \qquad \underbrace{QM \ vs. \ Classical}_{Equations \ are}$$

$$\dot{x}_2 = Bp_1 + Dp_2 + Cx_1 \qquad \dot{p}_2 = -Bx_1 - Dx_2 + Cp_1 \qquad \underbrace{QM \ vs. \ Classical}_{Equations \ are}$$

$$identical$$

Then start with classical Hamiltonian. (Designed to give same result.)

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1}x_{2} + p_{1}p_{2} \right) + C \left(x_{1}p_{2} - x_{2}p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

Then Hamilton's equations of motion are the following.

$$\dot{x}_1 = \frac{\partial H_c}{\partial p_1} = Ap_1 + Bp_2 - Cx_2 \qquad \dot{p}_1 = -\frac{\partial H_c}{\partial x_1} = -(Ax_1 + Bx_2 + Cp_2)$$
$$\dot{x}_2 = \frac{\partial H_c}{\partial p_2} = Bp_1 + Dp_2 + Cx_1 \qquad \dot{p}_2 = -\frac{\partial H_c}{\partial x_2} = -(Bx_1 + Dx_2 - Cp_1)$$

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $\left|\Psi\right\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of real real 1st-order differential equations. Then start with classical Hamiltonian.

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1}x_{2} + p_{1}p_{2} \right) + C \left(x_{1}p_{2} - x_{2}p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

Then Hamilton's equations of motion are the following.

$$\dot{x}_{1} = Ap_{1} + Bp_{2} - Cx_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Bx_{2} - Cp_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Bx_{2} - Cp_{2}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} + Cp_{1}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} - Cp_{1}$$

$$\dot{p}_{3} = -K \cdot |\mathbf{x}\rangle$$

$$\begin{aligned} \ddot{x}_{1} &= A\dot{p}_{1} + B\dot{p}_{2} - C\dot{x}_{2} \\ &= -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ &= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2} \end{aligned}$$

$$\begin{aligned} \ddot{x}_{2} &= B\dot{p}_{1} + D\dot{p}_{2} + C\dot{x}_{1} \\ &= -B(Ax_{1} + Bx_{2} + Cp_{2}) - D(Bx_{1} + Dx_{2} - Cp_{1}) + C(Ap_{1} + Bp_{2} - Cx_{2}) \\ &= -(AB + BD)x_{1} - (B^{2} + D^{2} + C^{2})x_{2} + C(A + D)p_{1} \end{aligned}$$

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $|\Psi\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of real real 1st-order differential equations. Then start with classical Hamiltonian.

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1} x_{2} + p_{1} p_{2} \right) + C \left(x_{1} p_{2} - x_{2} p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

Then Hamilton's equations of motion are the following.

into pairs of real real 1st-order differential equations.

$$\dot{x}_{1} = Ap_{1} + Bp_{2} - Cx_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Bx_{2} - Cp_{2}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} + Cp_{1}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} + Cx_{1}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} - Cx_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Ax_{1} + Bx_{2} + Cp_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Ax_{1} + Bx_{2} + Cp_{2}$$

$$\dot{p}_{2} = -Bx_{1} - Bx_{2} - Cx_{2}$$

$$\dot{p}_{2} = -B(Ax_{1} + Bx_{2} + Cp_{2}) - D(Bx_{1} + Dx_{2} - Cp_{1}) + C(Ap_{1} + Bp_{2} - Cx_{2})$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2}$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{1}$$

$$\begin{pmatrix} \ddot{x}_{1} \\ \ddot{x}_{2} \end{pmatrix} = -\left(Ax_{1} - Ax_{2} - Ax_$$

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $|\Psi\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of real real 1st-order differential equations. Then start with classical Hamiltonian.

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1} x_{2} + p_{1} p_{2} \right) + C \left(x_{1} p_{2} - x_{2} p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

Then Hamilton's equations of motion are the following.

into pairs of real real 1st-order differential equations.

$$\dot{x}_{1} = Ap_{1} + Bp_{2} - Cx_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Bx_{2} - Cp_{2}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} + Cp_{1}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} + Cx_{1}$$

$$\dot{p}_{2} = -B(Ax_{1} + Bx_{2} + Cp_{2}) - D(Bx_{1} + Dx_{2} - Cp_{1}) + C(Ap_{1} + Bp_{2} - Cx_{2})$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2}$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2}$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2}$$

$$= -(A^{2} + B^{2} + B^{2} - Bx_{1} + Dx_{2} - Cx_{2})x_{2} + C(A + D)p_{1}$$

$$\begin{pmatrix} \ddot{x}_{1} \\ \ddot{x}_{2} \end{pmatrix} = -\left(A^{2} + B^{2} - AB + BD - B^{2} + D^{2} + C^{2}\right)x_{2} + C(A + D)p_{1}$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (A^{2} + B^{2} - B^{2})x_{2} + C^{2} +$$

Here is an operator view of the QM-Classical connection: Take Schrodinger operator $i\partial_t = \mathbf{H}$ (with C=0) and square it!

$$i\frac{\partial}{\partial t} = \begin{pmatrix} A & B \\ B & D \end{pmatrix} \Longrightarrow \left(i\frac{\partial}{\partial t}\right)^2 = \begin{pmatrix} A & B \\ B & D \end{pmatrix}^2 \Longrightarrow -\frac{\partial^2}{\partial t^2} = \begin{pmatrix} A^2 + B^2 & AB + BD \\ AB + BD & B^2 + D^2 \end{pmatrix}$$

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $\left|\Psi\right\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of real real 1st-order differential equations. Then start with classical Hamiltonian.

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1} x_{2} + p_{1} p_{2} \right) + C \left(x_{1} p_{2} - x_{2} p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

Then Hamilton's equations of motion are the following.

$$\begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} A^2 + B^2 & AB + BD \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} Ior C - O \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} Ior C - O \\ Ior C - O \\ Ior C - O \end{pmatrix} = \begin{pmatrix} Ior C - O \\ Ior C -$$

Here is an operator view of the QM-Classical connection: Take Schrodinger operator $i\partial_t = \mathbf{H}$ (with C=0) and square it!

$$i\frac{\partial}{\partial t} = \begin{pmatrix} A & B \\ B & D \end{pmatrix} \Longrightarrow \left(i\frac{\partial}{\partial t}\right)^2 = \begin{pmatrix} A & B \\ B & D \end{pmatrix}^2 \Longrightarrow -\frac{\partial^2}{\partial t^2} = \begin{pmatrix} A^2 + B^2 & AB + BD \\ AB + BD & B^2 + D^2 \end{pmatrix}$$

Conclusion: 2-state Schro-equation $i\hbar \frac{\partial}{\partial t} |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ is like "square-root" of Newton-Hooke. $\sqrt{|\mathbf{x}\rangle} = -\mathbf{K} \cdot |\mathbf{x}\rangle$

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \begin{pmatrix} A & B - iC \\ B + iC & D \end{pmatrix} = \mathbf{H}^{\dagger}$$

that operates on 2-D complex Dirac ket vector $|\Psi\rangle$.

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real x_k and imaginary p_k parts of Ψ_k amplitudes to convert the complex 1st-order equation $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of real real 1st-order differential equations. Then start with classical Hamiltonian.

$$H_{c} = \frac{A}{2} \left(p_{1}^{2} + x_{1}^{2} \right) + B \left(x_{1} x_{2} + p_{1} p_{2} \right) + C \left(x_{1} p_{2} - x_{2} p_{1} \right) + \frac{D}{2} \left(p_{2}^{2} + x_{2}^{2} \right)$$

Then Hamilton's equations of motion are the following.

$$\dot{x}_{1} = Ap_{1} + Bp_{2} - Cx_{2}$$

$$\dot{p}_{1} = -Ax_{1} - Bx_{2} - Cp_{2}$$

$$\dot{p}_{2} = -Bx_{1} - Dx_{2} + Cp_{1}$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{1}} = Ap_{1} + Bp_{2} - Cx_{2}$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} - Cp_{1})$$

$$\dot{p}_{2} = -\frac{\partial H_{c}}{\partial x_{2}} = -(Bx_{1} + Bx_{2} + Cp_{2}) - D(Bx_{1} + Bx_{2} - Cp_{1}) + C(Ap_{1} + Bp_{2} - Cx_{2})$$

$$\dot{p}_{2} = -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2}$$

$$\dot{p}_{2} = -(A^{2} + B^{2} + C^{2})x_{2} + C(A + D)p_{1}$$

$$\begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} A^2 + B^2 & AB + BD \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} For \ C = 0 \\ Is \ form \ of \ 2D \ Hooke \\ harmonic \ oscillator \end{pmatrix} = \begin{pmatrix} \ddot{x}_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{2$$

Here is an operator view of the QM-Classical connection: Take Schrodinger operator $i\partial_t = \mathbf{H}$ (with $C \neq 0$) and square it!

$$i\frac{\partial}{\partial t} = \begin{pmatrix} A & B-iC \\ B+iC & D \end{pmatrix} \Rightarrow \left(i\frac{\partial}{\partial t}\right)^2 = \begin{pmatrix} A & B-iC \\ B+iC & D \end{pmatrix}^2 \Rightarrow -\frac{\partial^2}{\partial t^2} = \begin{pmatrix} A^2+B^2+C^2 & AB+BD-i(AC+CD) \\ AB+BD+i(AC+CD) & B^2+D^2+C^2 \end{pmatrix}$$

Conclusion: 2-state Schro-equation $i\hbar \frac{\partial}{\partial t} |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$ is like "square-root" of Newton-Hooke. $\sqrt{|\mathbf{x}\rangle} = -\mathbf{K} \cdot |\mathbf{x}\rangle$