

基于投影与斜投影

① 投影矩阵 $\pi^2 = \pi$

$$\text{range } \pi = \{ \pi y \}$$

$$\text{ker } \pi = \{ x : \pi x = 0 \}$$

$$x = \pi x + (x - \pi x)$$

$$x \in \text{Range } \pi \cap \text{ker } \pi$$

$$\pi x = 0 \quad x = \pi y = \pi^2 y = \pi x = 0$$

$$\text{Range}(\pi) \oplus \text{ker}(\pi) = \mathbb{R}^N$$

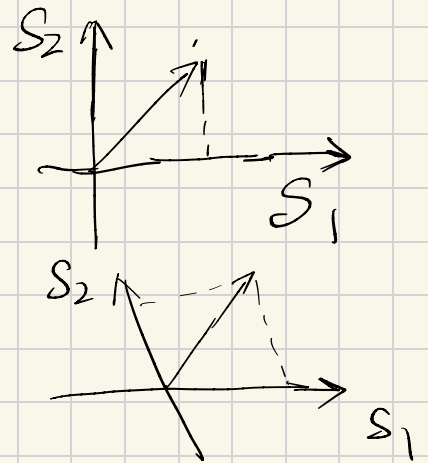
↓
特征值为 1 的特征向量
 $\pi(\pi y) = \pi(y)$

↓
特征值为 0 的特征向量

$$\pi = P \begin{bmatrix} I_m & \\ & 0_{N-m} \end{bmatrix} P^{-1}$$

$$\pi P = P \begin{bmatrix} I_m & \\ & 0_{N-m} \end{bmatrix}$$

$$\begin{aligned} \pi P_1 &= P_1 \Rightarrow S_1 \\ \pi P_2 &= 0 \Rightarrow S_2 \end{aligned}$$



② 三维螺旋

$$V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$\pi_{VVT} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

斜投影

$$V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \quad W = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$\pi_{V,W} = V(W^T V)^{-1} W^T$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

$$S_2 = \left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \right\}$$

$$= \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\pi_{V,W} \begin{pmatrix} \cos t \\ \sin t \\ t \end{pmatrix} = \begin{bmatrix} \cos t + t \\ \sin t + t \\ 0 \end{bmatrix}$$

③ 不变性

$$W, V \in \mathbb{R}^{N \times k}$$

变化基底

$$\tilde{W} = WQ_1 \quad \tilde{V} = VQ_2 \quad Q_i \in \mathbb{R}^{k \times k} \text{ 可逆矩阵}$$

$$\begin{aligned} (\tilde{W}^T \tilde{V})^{-1} \tilde{W}^T &= (Q_1^T W^T V Q_2)^{-1} Q_1^T W^T \\ &= Q_2^{-1} (W^T V)^{-1} W^T \end{aligned}$$

$$\tilde{q}(0) = Q_2^{-1} (W^T V)^{-1} W^T u_0 = Q_2^{-1} q(0)$$

$$\begin{aligned} \frac{d\tilde{q}(t)}{dt} &= (\tilde{W}^T \tilde{V})^{-1} \tilde{W}^T f(\tilde{V} \tilde{q}(t), t, x) \\ &= Q_2^{-1} (W^T V)^{-1} W^T f(V Q_2 \tilde{q}(t), t, x) \end{aligned}$$

$$\frac{dQ_2 \tilde{q}(t)}{dt} = (W^T V)^{-1} W^T f(V Q_2 \tilde{q}(t), t, x)$$

$$\Rightarrow \tilde{q}(t) = Q_2^{-1} q(t)$$

$$\Rightarrow \tilde{u}(t) = \tilde{V} \tilde{q}(t) = V Q_2 \tilde{q}(t) = V q(t) = u(t)$$

④ 误差分析

HDM:

$$\frac{d u(t)}{d t} = A u(t) \quad u(t) \in \mathbb{R}^N$$

PROM:

$$u(t) = V q(t) \quad V \in \mathbb{R}^{N \times k} \quad V^T V = I$$

$$\frac{d q(t)}{d t} = V^T A V q(t) \quad q(0) = V^T u(0)$$

误差:

$$\varepsilon_v = V V^T u(t) - V q(t) \quad \varepsilon_{v \perp} = u - V V^T u$$

$$[V \ V_c] \Rightarrow I = [V \ V_c] [V \ V_c]^T = V V^T + V_c V_c^T$$

$$V V^T \varepsilon_v = \varepsilon_v \Rightarrow \|\varepsilon_v\|_2 = \|V^T \varepsilon_v\|_2$$

$$V_c V_c^T \varepsilon_{v \perp} = V_c V_c^T u = \varepsilon_{v \perp} \quad \|\varepsilon_{v \perp}\|_2 = \|V_c^T \varepsilon_{v \perp}\|_2$$

$$\frac{d \varepsilon_v}{d t} = V V^T A u - V V^T A V q(t)$$

$$= V V^T A (u - V q(t))$$

$$= V V^T A \varepsilon_v + \underline{V V^T A \varepsilon_{v \perp}} \quad \text{外力}$$

ODE:

$$\dot{x}(t) = Ax + u \quad x(0) = 0$$

$$\Rightarrow x(t) = \int_0^t e^{A(t-\tau)} u(\tau) d\tau$$

$$\|x\|_2 \leq \|F(T; A)\| \|u\|_2$$

$$\frac{d V^T \varepsilon_v}{dt} = (V^T A V) V^T \varepsilon_v + (V^T A V_c) V_c^T \varepsilon_{v_\perp}$$

$$\|\varepsilon_v\|_2 \leq \|F(T, V^T A V)\|_2 \|V^T A V_c\|_2 \|\varepsilon_{v_\perp}\|_2$$

⑤ 稳定性

$$\frac{du}{dt} = Au \quad u = e^{At} u_0$$

$$A = \begin{bmatrix} 1 & -3.5 \\ 0.6 & -2 \end{bmatrix}$$

$$\lambda(A) = \{-0.1127 \quad -0.8873\}$$

$$V = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$V^T A V = [1]$$

对于 A 是对称矩阵的情形

A 可对角化, 特征值为实数, 满足 $\lambda(A) \leq 0$

$$\lambda(A) \leq 0$$

$$\Rightarrow v^T A v \leq 0 \quad \forall v$$

$$\text{因为 } A = Q^T \Sigma Q$$

$$v^T Q^T \Sigma Q v = \sum \lambda_i u_i^2 \leq 0$$

$$\text{其中 } u = Qv$$

$$\Rightarrow u^T V^T A V u \leq 0 \quad \forall u$$

$$\Rightarrow V^T A V u = \lambda u \quad \text{那么 } \lambda \leq 0$$

$$\text{因为 } u^T V^T A V u = \lambda u^T u \leq 0$$