

پردازش سیگنال گرافی

دکتر آرش امینی



دانشگاه صنعتی شریف

مهندسی برق

برنا خدابنده ۴۰۰۱۰۹۸۹۸

تمرین شماره ۳

تاریخ: ۱۴۰۲/۱۰/۲

$$h(L) = \sum_{i=0}^{m-1} \alpha_i L^i \rightsquigarrow h(\lambda) = \sum_{i=0}^{m-1} \alpha_i \lambda^i$$

(L ∩ M)

$h(\lambda)$: nonzero for k values of $\lambda: \exists \lambda_1, \dots, \lambda_k: h(\lambda_i) \neq 0$.

$$A = W, d_i = n \Rightarrow L = rI - A \Rightarrow \lambda_i(L) = r - \lambda_i(A), \mathbb{F}_r(L) = \mathbb{F}_r(A)$$

minimal polynomial: $Q(A) = \sum_{i=0}^s \beta_i A^i = 0, \forall \lambda \in \text{eig}(A): Q(\lambda_A) = 0$

$$Q(A) = \sum_{i=0}^s \beta_i A^i = \sum_{i=0}^s \beta_i (rI - L)^i = \sum_{i=0}^s q_i L^i = 0 = Q'(L) \cdot Q'(\lambda_i) = 0$$

$$\deg(Q'(\lambda)) = \deg(Q(\lambda)) = s \geq d+1 : \exists (i,j): \forall k < d: A_{(i,j)}^k = 0 \\ A_d^{(i,j)} = 1$$

$$\Rightarrow \deg(Q) \geq d+1, \exists \lambda_1, \dots, \lambda_n: h(\lambda_i) \neq 0, \lambda_{k+1}, \dots, \lambda_n: h(\lambda_i) = 0$$

only distinct terms

$$\text{let } f(\lambda) = h(\lambda) \times (\lambda - \lambda_1) \times (\lambda - \lambda_2) \times \dots \times (\lambda - \lambda_k) : h(\lambda_i) \neq 0$$

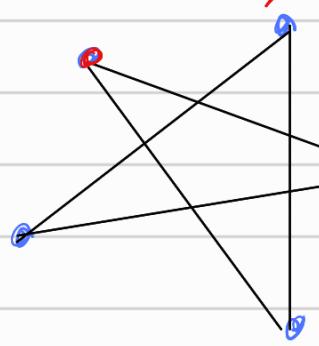
$$\Rightarrow \forall \lambda \in \text{eig}(L): f(\lambda) = 0, \deg(f(\lambda)) = k + \deg(h(\lambda))$$

$$\Rightarrow \deg(f(\lambda)) = m - 1 + k$$

$$Q(\lambda): \text{minimal polynomial} \Rightarrow \deg(f) \geq \deg(Q) \geq d+1$$

$$\Rightarrow m - 1 + k \geq d + 1 \Rightarrow m \geq d - k + 2$$

$k=2, r=0, s=1$



$\forall (i,j) \in E : A_{ij} = 1, A_{ij}^2 = \text{length 2 between neighbor} = r$ (e.g.)

$\forall (i,j) \notin E : A_{ij} = 0, A_{ij}^2 = \dots \text{non-neighbor} = s$ ($i \neq j$)

$\forall i=j : A_{ii} = 0, A_{ii}^2 = k$ (\curvearrowleft)

$A^2 - rA + sI : \forall i=j : (A^2 + (s-r)A)_{(i,j)} = s, (i=j), k$

$\Rightarrow \text{let } P(A) = A^2 + (s-r)A + (s-k)I \Rightarrow P(A)_{(i,j)} = s$

$\Rightarrow P(A) = s\mathbf{1}\mathbf{1}^T, \text{rank}(P(A)) = 1 \Rightarrow 1 \text{ non zero eigenvalue}$

$P(A)\mathbf{1} = s\mathbf{1}\mathbf{1}^T\mathbf{1} = ns\mathbf{1} \Rightarrow \mathbf{1} \in \text{eig}(P(A))$

$A\mathbf{1} = k\mathbf{1} \Rightarrow \mathbf{1} \in \text{eig}(A), \lambda_1 = k, \mathbf{1} = v_1$

$\forall v \in \mathbb{R}^n \perp \mathbf{1} : P(A)v = 0 \Rightarrow \forall v \in \text{eig}(A)/\mathbf{1} : P(A)v = 0$

$\Rightarrow [A^2 + (s-r)A + (s-k)I]v = 0 \Rightarrow \lambda_A^2 + (s-r)\lambda_A + (s-k) = 0$

$\Rightarrow \lambda_A = \frac{r-s \pm \sqrt{(s-r)^2 + 4(s-k)}}{2}, L = D - A = kI - A$

$\Rightarrow \lambda_L = k - \lambda_A, \lambda_A \in \{k, \frac{r-s + \sqrt{(r-s)^2 + 4(s-k)}}{2}, \frac{r-s - \sqrt{(r-s)^2 + 4(s-k)}}{2}\}$

$\Rightarrow \lambda_L = k - \lambda_A \Rightarrow \lambda_L \in \{0, \frac{2k-r+s+\sqrt{(r-s)^2+4(s-k)}}{r}, \frac{2k-r+s-\sqrt{(r-s)^2+4(s-k)}}{r}\}$

$\sum (x_i)_{(i,i)} = 0 \Rightarrow x \in \mathbb{R}^n \perp \mathbf{1} \Rightarrow x = \sum_{v \in \text{eig}(L) \perp \mathbf{1}} \gamma_i v_i = \gamma_2 v_2 + \gamma_3 v_3$

$f(L) = \sum_{i=0}^{m-1} \alpha_i L^i, LV = \theta_{2,3} v_2, v_3$

$$y = f(L)x = \sum_{i=0}^{m-1} \alpha_i L^i x = \sum_{i=0}^{m-1} \alpha_i L^i (v_2 v_2 + v_3 v_3) \\ = r_2 \sum \alpha_i L^i v_2 + r_3 \sum \alpha_i L^i v_3 = r_2 v_2 \sum \alpha_i \theta_2^i + r_3 v_3 \sum \alpha_i \theta_3^i$$

$$\Rightarrow y = r_2 f(\theta_2) v_2 + r_3 f(\theta_3) v_3$$

$$y^\top x = 0 \Rightarrow r_2^2 f(\theta_2) + r_3^2 f(\theta_3) = 0, \|x\|_2 = 1 \Rightarrow r_2^2 + r_3^2 = 1$$

$$\Rightarrow r_2^2 (f(\theta_2) - f(\theta_3)) + f(\theta_3) = 0 \Rightarrow r_2^2 = \frac{f(\theta_3)}{f(\theta_3) - f(\theta_2)}, r_3^2 = \frac{f(\theta_2)}{f(\theta_2) - f(\theta_3)}$$

$$\|y\|_2^2 = r_2^2 f(\theta_2)^2 + r_3^2 f(\theta_3)^2$$

$$\|y\|_2^2 = \frac{1}{f(\theta_2) - f(\theta_3)} \cdot \left[f(\theta_3) f(\theta_2)^2 - f(\theta_2) f(\theta_3)^2 \right]$$

$$\|y\|_2^2 = f(\theta_2) f(\theta_3) \Rightarrow \|y\|_2 = \sqrt{f(\theta_2) f(\theta_3)}$$

(30) | m

$$y_i = \sqrt{n} y_{i-1} \otimes s_i \Rightarrow \hat{y}_i = \sqrt{n} \hat{y}_{i-1} \otimes \hat{s}_i$$

$$\hat{y}_i(k) = \sqrt{n} \hat{y}_{i-1}(k) \hat{s}_i(k), \quad \hat{s}_i = U^H s_i = U_i^H$$

$$\hat{y}_i(k) = \sqrt{n} \hat{y}_{i-1}(k) U_{ki}^H = \sqrt{n} U_{ki}^H \hat{y}_{i-1}(k) = \sqrt{n} \bar{U}_{ik} \hat{y}_{i-1}(k) = \hat{y}_i(k)$$

$$\langle y_i \rangle = \frac{1}{n} \langle y_i, \mathbb{1} \rangle = \frac{1}{n} \langle \hat{y}_i, \hat{\mathbb{1}} \rangle = \langle \hat{y}_i, \frac{1}{n} \hat{\mathbb{1}} \rangle$$

we assume $\mathbb{1} \in \text{eig}(\text{shift})$, like if we were using L as shift

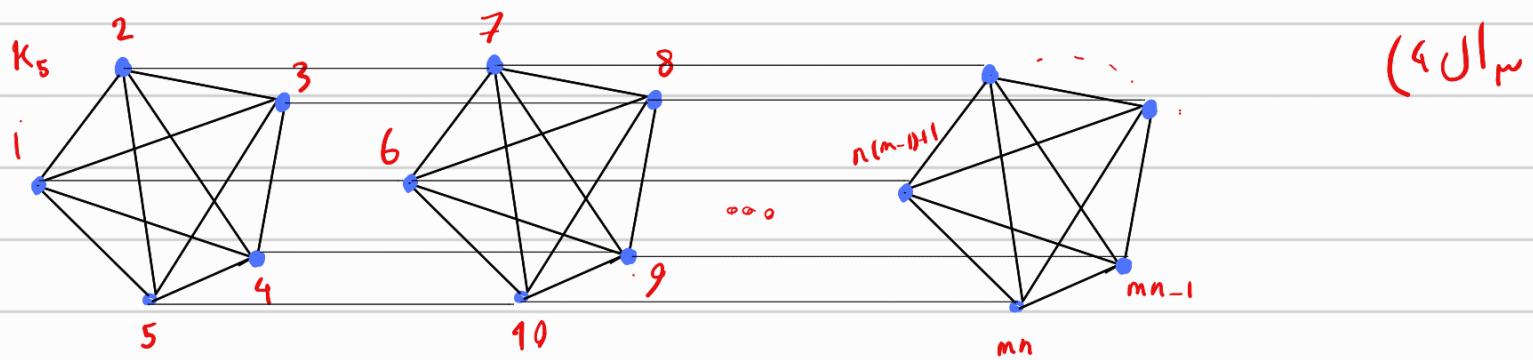
$$\Rightarrow \exists i_0: U_{1i_0} = U_{2i_0} = \dots = U_{ni_0} = \frac{1}{\sqrt{n}} (U^H \mathbb{1}) \approx i_0: \text{index of column of } \mathbb{1} \text{ in } U$$

$$\Rightarrow \hat{\mathbb{1}} = U^H \mathbb{1} = s_{i_0} \times \sum \frac{1}{\sqrt{n}} = \sqrt{n} s_{i_0} \Rightarrow \frac{1}{n} \hat{\mathbb{1}} = \frac{1}{\sqrt{n}} s_{i_0}$$

$$\Rightarrow \langle y_i \rangle = \langle \hat{y}_i, \frac{1}{n} \hat{\mathbb{1}} \rangle = \langle \hat{y}_i, \frac{1}{\sqrt{n}} s_{i_0} \rangle = \frac{1}{\sqrt{n}} \hat{y}_i(i_0) = m_i$$

$$m_i = \frac{1}{\sqrt{n}} \hat{y}_i(i_0) = \frac{1}{\sqrt{n}} \cdot \sqrt{n} \bar{U}_{ii_0} \hat{y}_{i-1}(i_0) = \frac{1}{\sqrt{n}} \hat{y}_{i-1}(i_0) = m_{i-1} = m_i$$

$$\Rightarrow m_n = m_0 \Rightarrow \boxed{\langle y_n \rangle = \langle y_0 \rangle}$$



We can easily see: $H = K_n \times K_m$, \times : cartesian product.

$$L_H = D_H - W_H = I_n \otimes D_{K_m} + D_{K_n} \otimes I_m - W_H = I_n \otimes L_{K_m} + L_{K_n} \otimes I_m$$

$$\rightarrow \text{let } \beta_{ij} = \psi_i \otimes \phi_j, L_{K_m} \phi_j = \gamma_j \phi_j, L_{K_n} \psi_i = \lambda_i \psi_i$$

$$L_H \beta_{ij} = (L_{K_n} \otimes I_m + I_n \otimes L_{K_m}) \psi_i \otimes \phi_j = (L_{K_n} \psi_i) \otimes \phi_j + \psi_i \otimes (L_{K_m} \phi_j)$$

$$(\lambda_i + \gamma_j) \psi_i \otimes \phi_j = (\lambda_i + \gamma_j) \beta_{ij} = L_H \beta_{ij}, \text{ there are } n \times m \text{ distinct } \beta_{ij} \text{ values} \Rightarrow \text{all eigenvectors}$$

$$\lambda_i : \text{eig}(K_n) = \{0, n, 2n, \dots, n\} \rightarrow \lambda_i \in \{0, n\}$$

$$\gamma_j : \text{eig}(K_m) = \{0, m, 2m, \dots, m\} \rightarrow \gamma_j \in \{0, m\}$$

Eigenvalues: $\Lambda_H = \{0, n, n, \dots, n, m, m, \dots, m, mn, mn, \dots, mn\}$

Eigenvectors: $\beta_{ij} = \psi_i \otimes \phi_j, \alpha \rightarrow \alpha_1, \alpha_2 : \alpha_2 = \alpha \bmod n$
 $\forall \alpha \in \mathbb{R} \quad \alpha_1 = \text{floor}(\alpha/n)$

ψ_i : we need ψ_i to be constant and all other to have zero mean.

arbitrarily we choose the columns of the DFT matrix

$$\Rightarrow U_{K_n} = \text{DFT}_n, U_{K_m} = \text{DFT}_m : \text{DFT}_n(l, m) = \frac{1}{\sqrt{n}} e^{\frac{2\pi i}{n} lm}$$

we use $D^n \equiv \text{DFT}_n$

assuming rates start from 0 instead

$$\beta_{ij} = \gamma_i \otimes \phi_j \Rightarrow U_H = U_{K_n} \otimes U_{K_m} = D^n \otimes D^m = U_H$$

$$y = S_K \otimes S_\ell = U(\hat{S}_K \otimes \hat{S}_\ell) = U(\hat{S}_K \circ \hat{S}_\ell) = U(U^\dagger S_n) \odot (U^\dagger S_\ell)$$

$$\Rightarrow y = U(u_\ell \odot u_K), \quad U_K = \gamma_{K_1} \otimes \phi_{K_2} = D_{K_1}^n \otimes D_{K_2}^m$$

$$u_K(a) = D(a_1, K_1) D(a_2, K_2)$$

$$= \frac{1}{\sqrt{mn}} \exp \left(\underbrace{\frac{2\pi i}{n} a_1 K_1}_{\theta_{ak}} + \underbrace{\frac{2\pi i}{m} a_2 K_2}_{\theta_{al}} \right)$$

$$(U_\ell \odot U_K)(a) = u_\ell(a) U_K(a) = \frac{1}{mn} \exp \left(\theta_{ak} + \theta_{al} \right)$$

$$\theta_{xk} : \text{linear with } x_1^{-1} \Rightarrow \theta_{ak} + \theta_{al} = \frac{2\pi i}{n} (a_1^{-1})(K_1 + l_1 - 1) + \frac{2\pi i}{m} (a_2^{-1})(K_2 + l_2 - 1)$$

$$\Rightarrow \begin{cases} K_1 + l_1 \\ K_2 + l_2 \end{cases} \Rightarrow \text{we want to find } t, \text{ such that } \theta_{ak} + \theta_{al} = \theta_{at} + 2\pi i t$$

$$t = nt_1 + t_2 \Rightarrow t = n(K_1 + l_1) \bmod n + \underbrace{(K_2 + l_2) \bmod n}_{(k \bmod n + l \bmod n) \bmod n}$$

$$t_1 \rightarrow t_1 + n \equiv t \rightarrow t + mn, \quad t \in [1, mn] \Rightarrow t = t \bmod mn = (l+k) \bmod n$$

$$\Rightarrow (U_\ell \odot U_K)(a) = \frac{1}{mn} \exp(\theta_{at}) = \frac{1}{\sqrt{mn}} U_t(a) : \theta_{ak} + \theta_{al} = \theta_{at}$$

$$\Rightarrow U_\ell \odot U_K = \frac{1}{\sqrt{mn}} U_t$$

$$y = U(U_\ell \odot U_K) = \frac{1}{\sqrt{mn}} U U_t = \frac{1}{\sqrt{mn}} S_t = S_\ell \otimes_H S_K$$

in the end we fix for mtrs starting from 1, checked with matlab

$$\Rightarrow t = 1 + \left\{ \left(n \left[\left(L \frac{k-1}{n} \right) + \left(L \frac{l-1}{n} \right) \right] \bmod n \right) + (l+k-2) \bmod n \right\} \bmod mn$$

$\omega(x)$: largest eigenvalue of x : $\hat{x}(\omega(x)) \neq 0$ (50) ↗

$PW_\omega(G) = \text{span}(u_i : \lambda_i \leq \omega) \Rightarrow$ all signals with bandwidth ω

V : edges of G , SCV : a "good sampling" if for two signals

$$f, g : f(S) = g(S), S \in V \Rightarrow f(V) = g(V) \quad (I)$$

$\omega_c(S)$: largest ω such that (I) holds for $f, g \in PW_\omega(G)$

$L_2(S)$: if $f \in L_2(S) \Rightarrow \forall i \notin S : f(i) = 0$

(a)

$$\forall f, g \in PW_\omega(G) : f = \sum_{\lambda_i \leq \omega} \alpha_i u_i = U \hat{f} \quad g = \sum_{\lambda_i \leq \omega} \beta_i u_i = U \hat{g}$$

$$\hat{f}(\lambda > \omega) = \hat{g}(\lambda > \omega) = 0, \quad f(S) = I_{S \downarrow} f, \quad g(S) = I_{S \downarrow} g$$

$$\text{let } U_\omega = \begin{bmatrix} u_{1,1} & \dots & u_{1,\omega} \\ u_{2,1} & \dots & u_{2,\omega} \\ \vdots & \ddots & \vdots \\ u_{n,1} & \dots & u_{n,\omega} \end{bmatrix}, \quad \hat{f}_\omega = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_\omega \end{bmatrix} \Rightarrow f = \sum \hat{f}_i u_i = \sum \hat{f}_{\omega(i)} U_{\omega(i)}$$

$$\Rightarrow f = U_\omega \hat{f}_\omega, \quad g = U_\omega \hat{g}_\omega$$

$$f(S) = I_{S \downarrow} U_\omega \hat{f}_\omega, \quad g(S) = I_{S \downarrow} U_\omega \hat{g}_\omega \Rightarrow \text{if this leads to}$$

$$\hat{f}_\omega = \hat{g}_\omega \Rightarrow S \text{ is sufficient}$$

$$f(S) = g(S) \Rightarrow I_{S \downarrow} U_\omega \hat{f}_\omega = I_{S \downarrow} U_\omega \hat{g}_\omega \Rightarrow I_{S \downarrow} U_\omega (\hat{f}_\omega - \hat{g}_\omega) = 0$$

$$\Rightarrow \hat{f}_\omega - \hat{g}_\omega \in N(I_{S \downarrow} U_\omega) \Rightarrow \text{if } N(I_{S \downarrow} U_\omega) = \emptyset \Rightarrow \hat{f}_\omega = \hat{g}_\omega \Rightarrow f(V) = g(V)$$

$$I_{S \downarrow} U_\omega = \left[\begin{array}{c|c|c|c|c} & & & & \\ \hline & & & & \\ \hline & \vdots & & & \\ \hline & \cdots & \cdots & \cdots & \cdots \\ \hline & 1 & & & \\ \hline & \vdots & & & \\ \hline & \cdots & \cdots & \cdots & \cdots \\ \hline & 1 & & & \\ \hline \end{array} \right] \begin{bmatrix} u_{1,1} & \dots & u_{1,\omega} \\ \vdots & \ddots & \vdots \\ u_{n,1} & \dots & u_{n,\omega} \end{bmatrix} = \begin{bmatrix} u_{1,1} & \dots & u_{1,\omega} \\ u_{2,1} & \dots & u_{2,\omega} \\ \vdots & \ddots & \vdots \\ u_{i_1,1} & \dots & u_{i_1,\omega} \\ \vdots & \ddots & \vdots \\ u_{i_n,1} & \dots & u_{i_n,\omega} \end{bmatrix} = U_{S, \omega}$$

we need: $\text{rank}(U_{S, \omega}) = |S|$; if so $\Rightarrow N(U_{S, \omega}) = \emptyset$

$$f(S) = \bigcup_{\omega} \hat{f}_{\omega}, N(\bigcup_{\omega} f_{\omega}) = \emptyset \Rightarrow \exists x \in \mathbb{R}^{\omega}: \bigcup_{\omega} f_{\omega} x = 0$$

$\rightarrow f(S) \neq 0 \Rightarrow f \notin L_2(S^c)$ * same for g , except for
 $f=0$ on $g=0$

$$\rightarrow \forall f, g \in PW_{\omega}(G) \wedge f(S) = g(S) \Leftrightarrow f(v) = g(v) \Rightarrow f, g \notin L_2(S^c)$$

$$\Rightarrow PW_{\omega}(G) \cap L_2(S^c) = \{0\}$$

$$x \in PW_{\omega}(G) \Leftrightarrow x = \sum_{\lambda_i < \omega} c_i u_i = \bigcup_{\omega} \hat{f}_{\omega} : PW_{\omega}(G) = C(U)$$

(same as before)

$$L_2(S^c) \cap PW_{\omega}(G) = \{0\} \Rightarrow \forall \tilde{\phi} \in L_2(S^c) : \tilde{\phi} \notin PW_{\omega}(G) = C(U_{\omega})$$

$$\rightarrow \tilde{\phi} = \bigcup_{\omega} \tilde{\phi}_{\omega} + e : e \in (C(U_{\omega}))^{\perp} = R^{\perp} - C(U_{\omega}) = C(U) - C(U_{\omega}) = \text{span}\{u_i : \lambda_i > \omega\}$$

$$\rightarrow \tilde{\phi} \in C(U_{\omega}) \cup \text{span}\{u_i : \lambda_i > \omega\} = \tilde{\phi} : \left[\begin{array}{c} \vdots \\ \hline \tilde{\phi} \\ \vdots \end{array} \right]_{\omega} \rightarrow \text{has at least one non-zero term}$$

$$\rightarrow \omega(\tilde{\phi}) > \omega \Leftrightarrow \omega < \inf_{\substack{\tilde{\phi} \in L_2(S^c) \\ \tilde{\phi} \neq 0}} \omega(\tilde{\phi})$$

in short, $\forall \tilde{\phi} \in L_2(S^c) : \tilde{\phi} \notin PW_{\omega}(G)$
 so it must have higher frequencies than ω

$$\inf_{\substack{\tilde{\phi} \in L_2(S^c) \\ \tilde{\phi} \neq 0}} \omega(\tilde{\phi}) \triangleq \omega_c(S) \Leftrightarrow$$

$$\omega < \inf_{\substack{\tilde{\phi} \in L_2(S^c) \\ \tilde{\phi} \neq 0}} \omega(\tilde{\phi}) \triangleq \omega_c(S)$$

$\inf_{\substack{\tilde{\phi} \in L_2(S^c) \\ \tilde{\phi} \neq 0}} \omega(\tilde{\phi}) = \omega_c(S)$: since choosing any ω higher than this would result in S not being a good choice.

$D = \underbrace{[I_{NM} \ 0_{NM} \ \dots \ 0_{NM}]}_m \in \mathbb{C}^{(NM) \times N}$, Dx : sampled (6J/m)

$$B^M = \begin{bmatrix} (A^M)_{1,1} & 0 \\ (A^M)_{2,1} & (A^M)_{2,2} \end{bmatrix}, (A^M)_{1,1} \in \mathbb{C}^{(NM) \times (NM)}: \text{first } (NM) \text{ rows/cols}$$

$$\bar{A} = DB^MD^T$$

$$H(A) = \sum_{k=1}^L h_k A^k, \quad D^T D = \begin{bmatrix} I_{NM} \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} [I_{NM} \ 0 \ \dots \ 0] = \begin{bmatrix} I_{NM} \ 0 \ \dots \ 0 \\ 0 \ \ddots \ \vdots \ \vdots \\ \vdots \ \ddots \ \vdots \\ 0 \ \dots \ \dots \ 0 \end{bmatrix} = \begin{bmatrix} I_{NM} & 0 \\ 0 & 0 \end{bmatrix} \quad (a)$$

$$\bar{A} = DB^MD^T = [I_{NM} \ 0] \begin{bmatrix} (A^M)_{1,1} & 0 \\ (A^M)_{2,1} & (A^M)_{2,2} \end{bmatrix} \begin{bmatrix} I_{NM} \\ 0 \end{bmatrix} = (A^M)_{1,1}$$

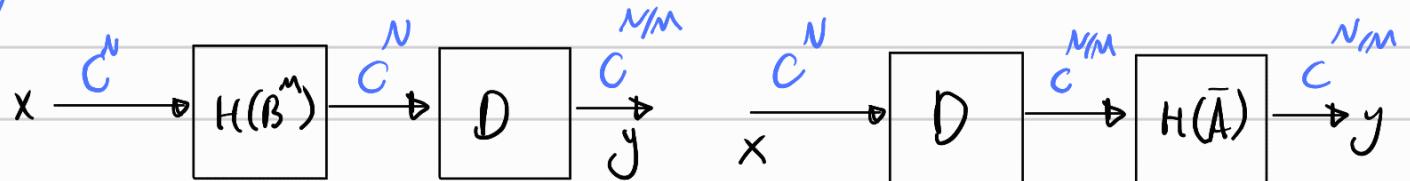
$$DB^M = [I_{NM} \ 0] \begin{bmatrix} (A^M)_{1,1} & 0 \\ (A^M)_{2,1} & (A^M)_{2,2} \end{bmatrix} = \cancel{(A^M)_{1,1}} \ 0 = \bar{A}D$$

$$\Rightarrow DB^M = \bar{A}D \Rightarrow D(B^M)^k = (DB^M)(B^M)^{k-1} = \bar{A}D(B^M)^{k-1} \underset{\text{same pattern}}{=} (\bar{A})^k D$$

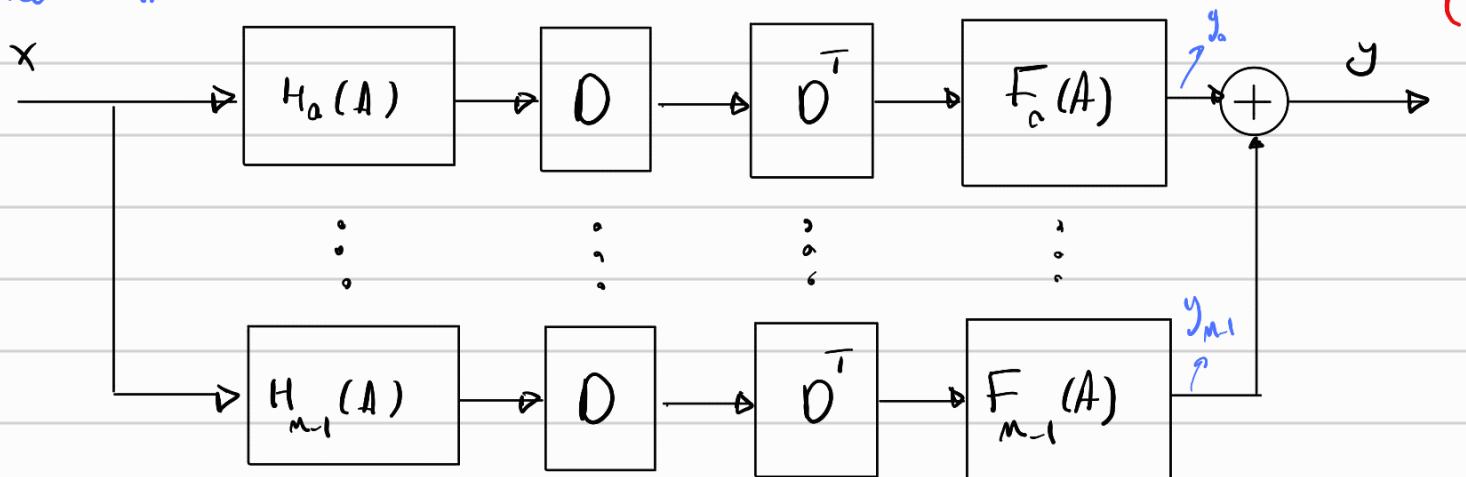
$$\Rightarrow D(B^M)^k = (\bar{A})^k D \Rightarrow D H(B^M) = D \left(\sum_{k=1}^L h_k (B^M)^k \right) = \sum_{k=1}^L h_k D(B^M)^k$$

$$= \sum_{k=1}^L h_k \bar{A}^k D = \left(\sum_{k=1}^L h_k \bar{A}^2 \right) D = \boxed{H(\bar{A})D = D H(\bar{A})}$$

system sketch:



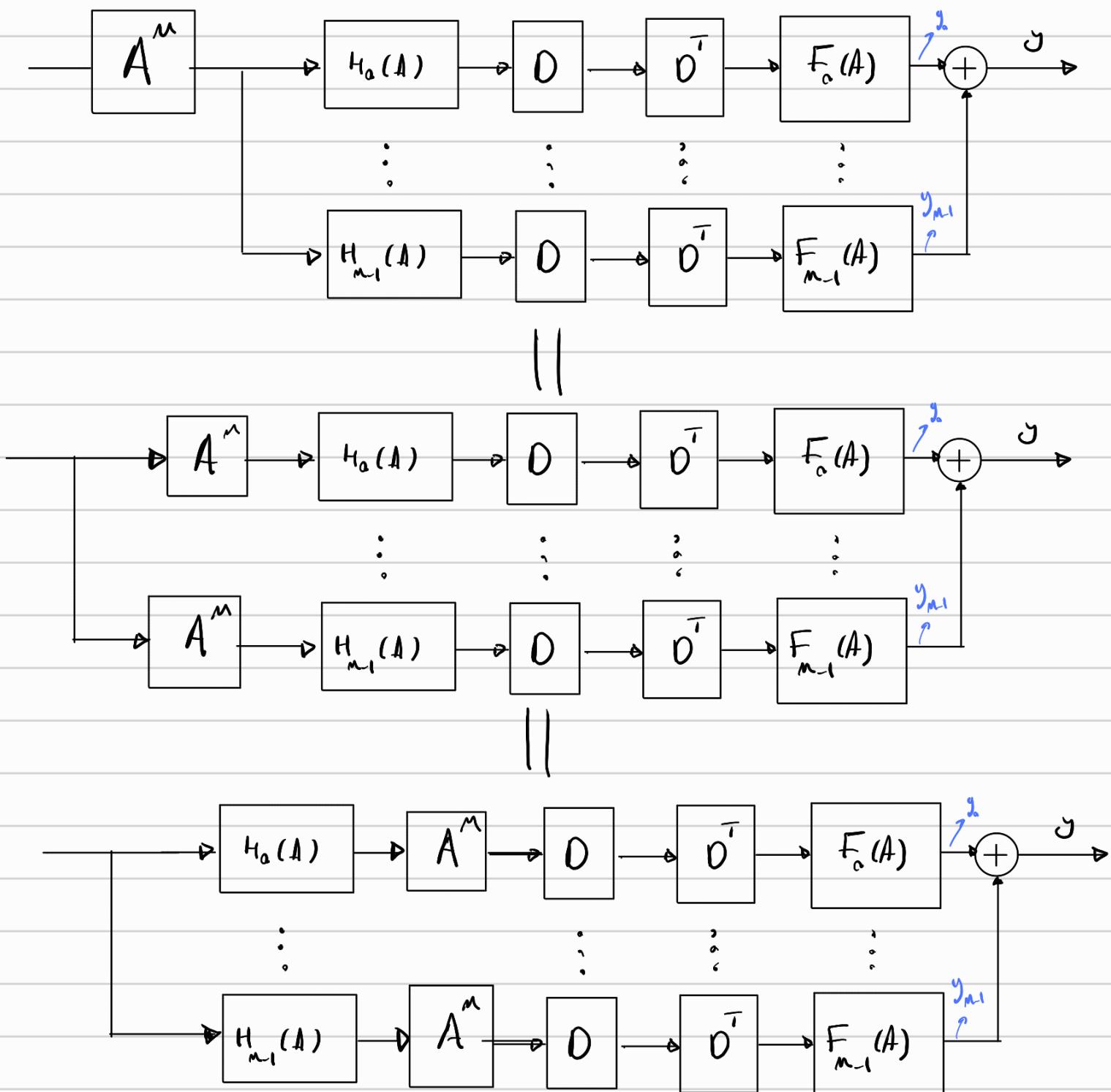
filter bank:



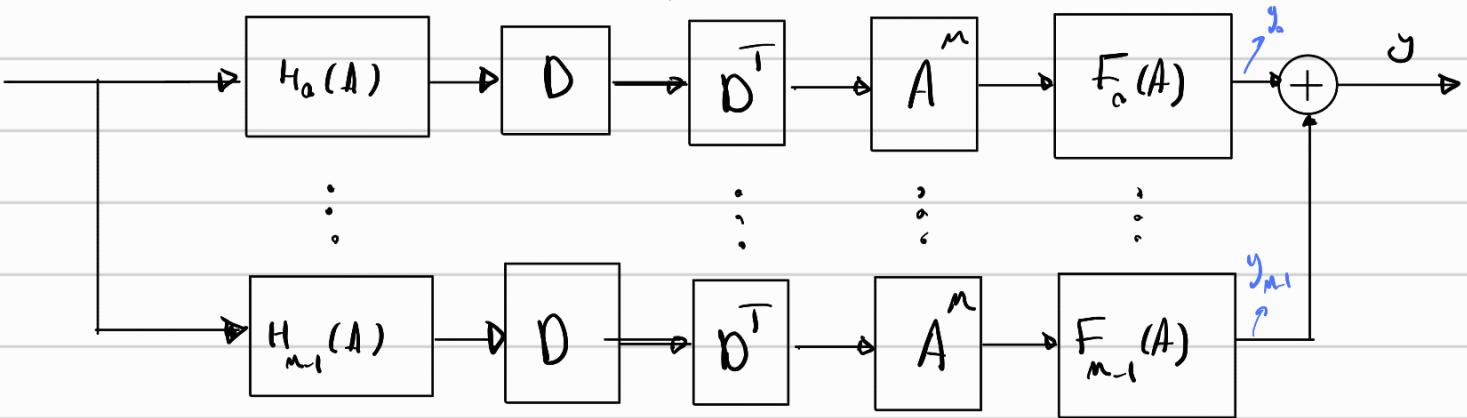
$$D^T D A^m = \begin{bmatrix} I_{N_m} \\ 0 \\ \vdots \\ 0 \end{bmatrix} [I_{N_m} \ 0 \ \dots \ 0] \begin{bmatrix} (A^m)_{1,1} & (A^m)_{1,2} \\ (A^m)_{2,1} & (A^m)_{2,2} \end{bmatrix} = \begin{bmatrix} (A^m)_{1,1} & (A^m)_{1,2} \\ 0 & 0 \end{bmatrix}$$

$$A^m D^T D = \begin{bmatrix} (A^m)_{1,1} & (A^m)_{1,2} \\ (A^m)_{2,1} & (A^m)_{2,2} \end{bmatrix} \begin{bmatrix} I_{N_m} \\ 0 \\ \vdots \\ 0 \end{bmatrix} [I_{N_m} \ 0 \ \dots \ 0] = \begin{bmatrix} (A^m)_{1,1} & 0 \\ (A^m)_{2,1} & 0 \end{bmatrix}$$

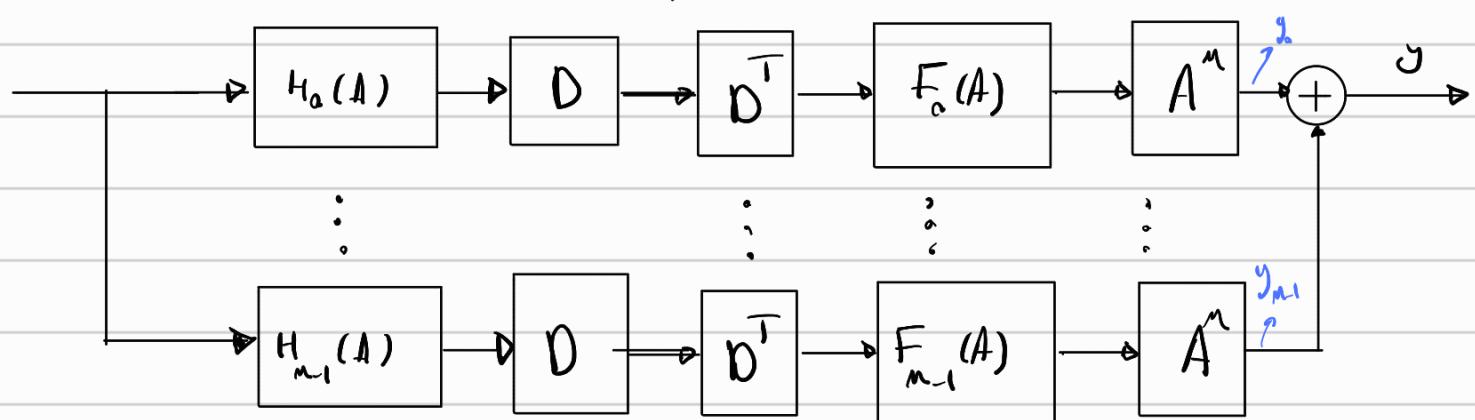
$$\Rightarrow A^m D^T D = D^T D A^m \text{ iff } (A^m)_{2,1} = 0, (A^m)_{1,2} = 0$$



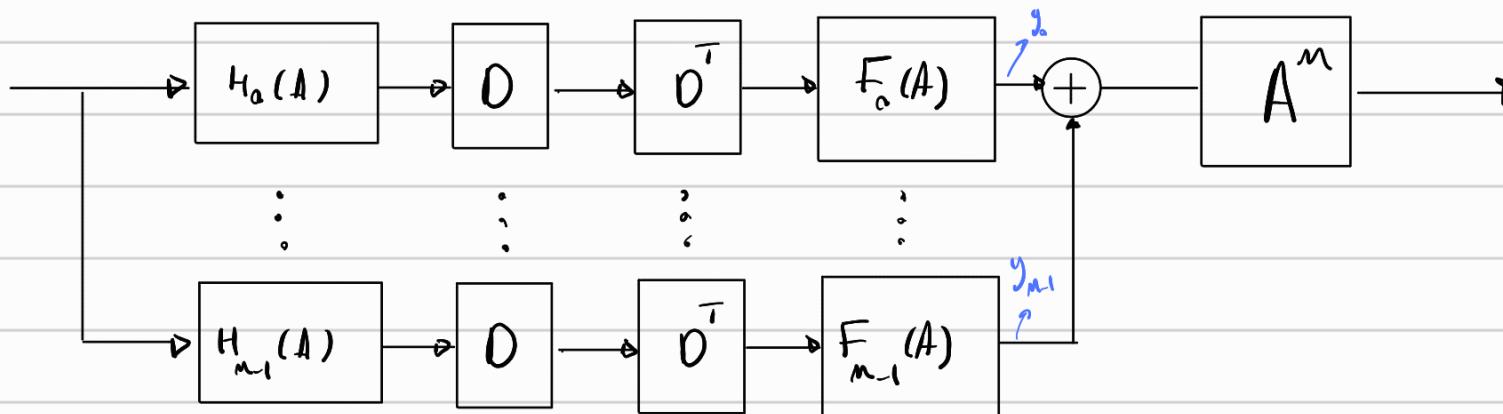
$$\text{X} \quad A^m D^T D \neq D^T D A^m$$



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in general, the described filter-bank is not MGSI

and depends on $D^T D A^m \neq A^m D^T D$, which would be true

iff $(A^m)_{1,2} = 0$, $(A^m)_{2,1} = 0$. here we checked

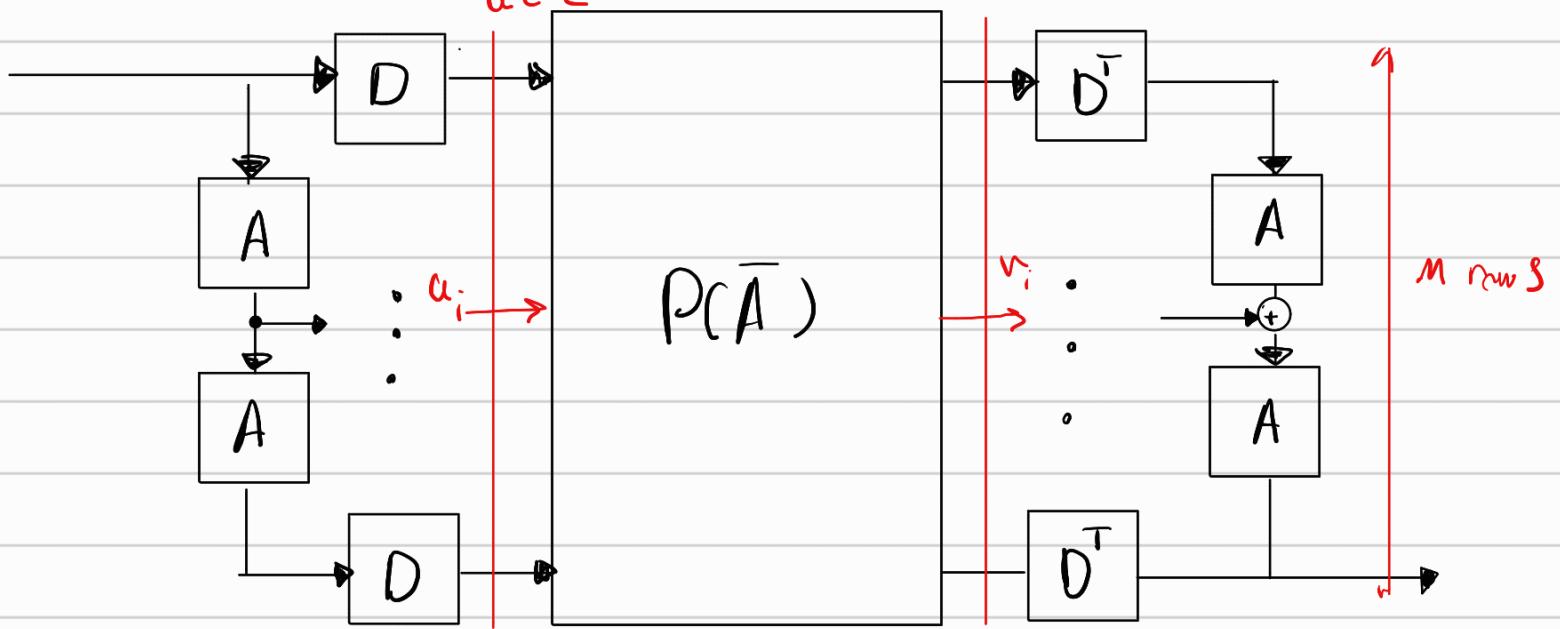
$S A^m = A^m S$ in the general form, for standard

GSI we can simply use $M=1$.

(b)

we proved this in the last part, $(A^m)_{1,2} = 0, (A^m)_{2,1} = 0$

$$H_i(A) = \sum_{r=0}^{M-1} A^r E_{i,r}(A^m), F_i(A) = \sum_{r=0}^{M-1} A^{M-r-1} R_{i,r}(A^m) \quad (c)$$



$$u_i = D A x \in C^{NM}, u = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_{M-1} \end{bmatrix} = \begin{bmatrix} D A^0 \\ D A^1 \\ \vdots \\ D A^{M-1} \end{bmatrix} x = \text{diag}_m(D) \begin{bmatrix} A^0 \\ A^1 \\ \vdots \\ A^{M-1} \end{bmatrix} x \in C^N$$

MIMO system:

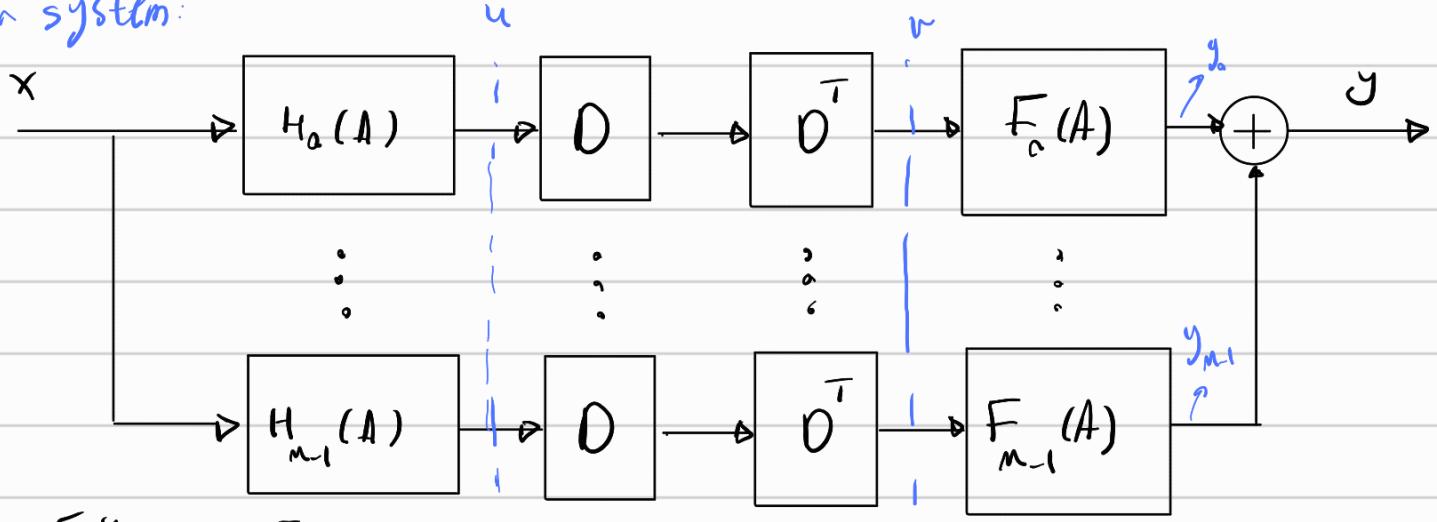
$$v_i = \sum_{j=0}^{M-1} P(\bar{A}) u_j \Rightarrow v = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_{M-1} \end{bmatrix} = P(\bar{A}) u, P(\bar{A}) = \begin{bmatrix} P(\bar{A})_{i,j} \end{bmatrix}_{N \times N}^{NM \times NM}$$

$$y = \sum_{i=0}^{M-1} A^{M-i-1} D^T v_i = [A^0 D^T \ A^1 D^T \ \dots \ A^{M-2} D^T \ A^{M-1} D^T] \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_{M-1} \end{bmatrix}$$

$$\Rightarrow y = [A^0 \ A^1 \ A^2 \ \dots \ A^{M-1}] \text{diag}_m(D^T) v = [A^0 \ A^1 \ A^2 \ \dots \ A^{M-1}] \text{diag}_m(D) v$$

$$\Rightarrow y = [A^0 \ A^1 \ A^2 \ \dots \ A^{M-1}] \text{diag}(D)^T P(\bar{A}) \text{diag}_m(D) \begin{bmatrix} A^0 \\ A^1 \\ \vdots \\ A^{M-1} \end{bmatrix} x$$

other system:



$$u = \begin{bmatrix} u_0 \\ u_1 \\ \vdots \\ u_{n-1} \end{bmatrix} = \begin{bmatrix} H_0(A) \\ \vdots \\ H_{n-1}(A) \end{bmatrix} x, \quad v = \text{diag}(D^T D) u, \quad y = [F_0(A) \dots F_{n-1}(A)] v$$

$$\Rightarrow y = [F_0(A) \dots F_{n-1}(A)] \text{diag}(D^T D) \begin{bmatrix} H_0(A) \\ \vdots \\ H_{n-1}(A) \end{bmatrix} x$$

$$H_i(A) = \sum_{r=0}^{m-1} A^r E_{i,r}(A^m), \quad F_i(A) = \sum_{r=0}^{m-1} A^{m-r-1} R_{i,r}(A^m)$$

$$\Rightarrow F_i(A) = [A^m \dots A^1 A^0] \begin{bmatrix} R_{i,0}(A^m) \\ R_{i,1}(A^m) \\ \vdots \\ R_{i,m-1}(A^m) \end{bmatrix} = [A^m \dots A^1 A^0] R_i(A^m)$$

$$H_i(A) = \underbrace{[E_{i,0}(A^m) \ E_{i,1}(A^m) \ \dots \ E_{i,m-1}(A^m)]}_{E_i(A^m)} \begin{bmatrix} A^m \\ A^1 \\ \vdots \\ A^0 \end{bmatrix} = E_i(A^m) \begin{bmatrix} A^m \\ A^1 \\ \vdots \\ A^0 \end{bmatrix}$$

$$\begin{bmatrix} H_0(A) \\ H_1(A) \\ \vdots \\ H_{n-1}(A) \end{bmatrix} = \underbrace{\begin{bmatrix} E_0(A^m) \\ E_1(A^m) \\ \vdots \\ E_{n-1}(A^m) \end{bmatrix}}_{E(A^m)} \begin{bmatrix} A^m \\ A^1 \\ \vdots \\ A^0 \end{bmatrix} = E(A^m) \begin{bmatrix} A^m \\ A^1 \\ \vdots \\ A^0 \end{bmatrix}, \quad E(A^m) = \begin{bmatrix} E_0(A^m) & E_1(A^m) & \dots & E_{n-1}(A^m) \\ \vdots & \vdots & \ddots & \vdots \\ E_{n-1}(A^m) & E_{n-2}(A^m) & \dots & E_0(A^m) \end{bmatrix}$$

$$[F_0(A) \ F_1(A) \ \dots \ F_{n-1}(A)] = [A^m \ \dots \ A^1 A^0] \underbrace{[R_0(A^m) \ R_1(A^m) \ \dots \ R_{n-1}(A^m)]}_{R(A^m)}$$

$$\therefore [E(A)F_1(A) \dots F_{m-1}(A)] = [A^T \dots A^T A^T] R((A^T)^T)$$

$$R(A^T) = \begin{bmatrix} R_{0,0}(A^T) & R_{0,1}(A^T) & \dots & R_{0,m-1}(A^T) \\ R_{1,0}(A^T) & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ R_{m-1,0}(A^T) & \dots & \dots & R_{m-1,m-1}(A^T) \end{bmatrix}, \quad A = A^T \Rightarrow A^T = (A^T)^T$$

needed constraint
in noble, only \bar{A} matters

$$\approx y = [A^{m-1} \dots A^T A^T] R^T(A^T) \text{diag}_n(\mathbf{0}) \text{diag}_m(\mathbf{D}) E(A^T) \begin{bmatrix} A \\ A^T \\ \vdots \\ A^{m-1} \end{bmatrix} x$$

$$B^m = \begin{pmatrix} (A^T)_{1,1} & \dots \\ \cancel{(A^T)_{2,1}} & (A^T)_{2,2} \end{pmatrix} = A^T \Rightarrow D \text{H}(A^T) = H(\bar{A}) D$$

$$D E_{in}(A^T) = E_{in}(\bar{A}) D \Rightarrow \text{diag}(D) E(A^T) = E(\bar{A}) \text{diag}(D)$$

$$D R_{in}(A^T) = R_{in}(\bar{A}) D \Rightarrow \text{diag}(D) R(A^T) = R(\bar{A}) \text{diag}(D)$$

$$\therefore R^T(A^T) \text{diag}(D^T) = \text{diag}(D^T) R^T(\bar{A})$$

$$\therefore y = [A^{m-1} \dots A^T] \text{diag}(D^T) R^T(\bar{A}) E(\bar{A}) \text{diag}(D) \begin{bmatrix} A \\ \vdots \\ A^{m-1} \end{bmatrix} x$$

equivalence $\Rightarrow y = [A^{m-1} \dots A^T] \text{diag}(D^T) P(\bar{A}) \text{diag}(D) \begin{bmatrix} A \\ \vdots \\ A^{m-1} \end{bmatrix} x$

$\Rightarrow \boxed{\bar{A} = \bar{A}^T, \quad P(\bar{A}) = R^T(\bar{A}) E(\bar{A})}$

$\Rightarrow \boxed{P_{ij}(\bar{A}) = \sum_{n=0}^{m-1} R_{ni}(\bar{A}) E_{nj}(\bar{A})}$