

# Generalized Singular Value Decomposition for Coordinated Beamforming in MIMO systems

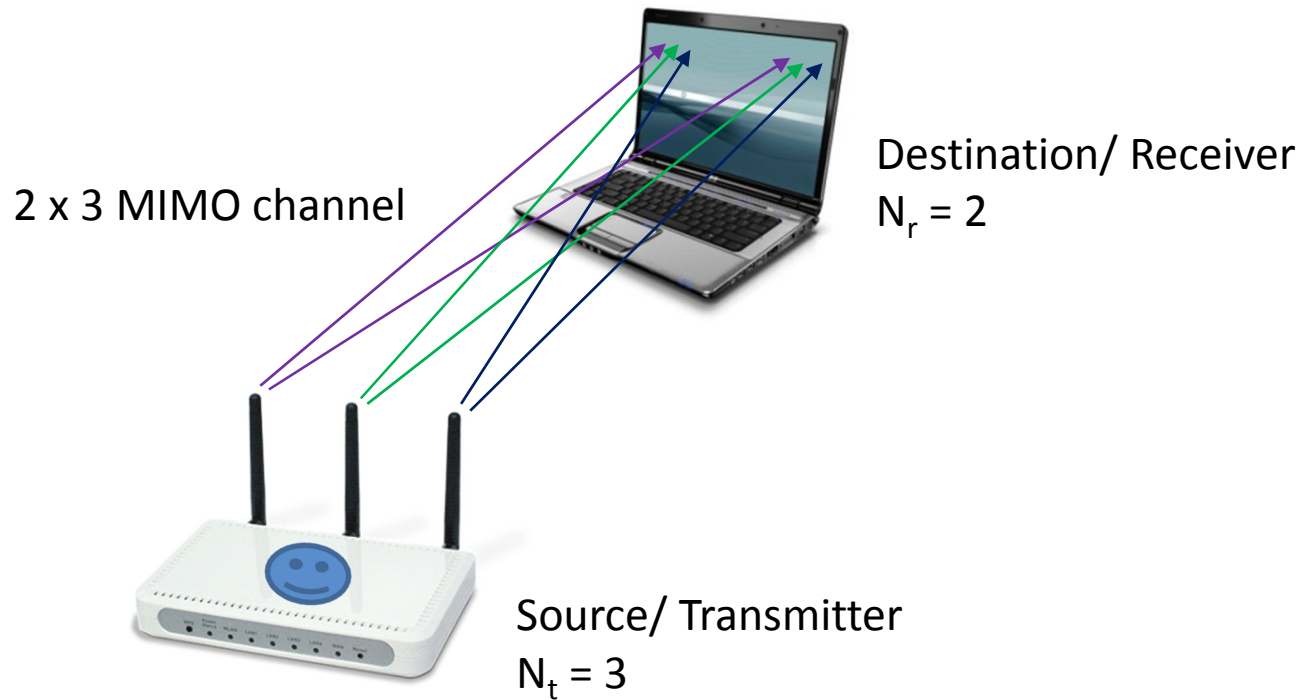
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Globecom 2010

# Multiple Input Multiple Output (MIMO) Channel



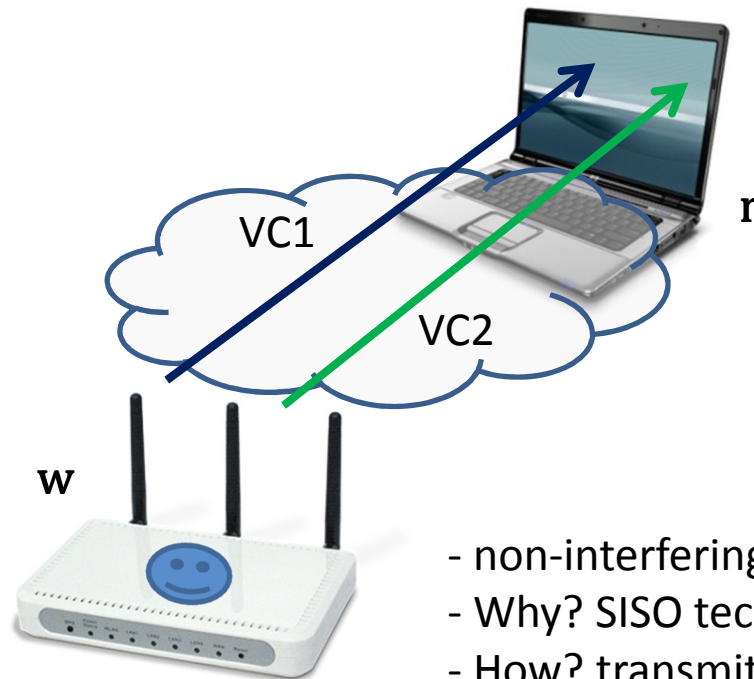
System Model:

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{n}$$

channel matrix

$$\mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \end{bmatrix}_{2 \times 3}$$

# Beamforming (+ Spatial Multiplexing)



- non-interfering, point-to-point virtual channels (VCs)
- Why? SISO techniques for modulation, coding...
- How? transmitter and/or receiver signal processing

transmit precoding matrix  $\mathbf{w}$   
 receiver reconstruction matrix  $\mathbf{r}$

$$\mathbf{y} = \mathbf{r}(\mathbf{H}\mathbf{w}\mathbf{x} + \mathbf{n})$$

$$\mathbf{r}\mathbf{H}\mathbf{w} = \begin{bmatrix} \lambda & \mathbf{0} \\ \mathbf{0} & \mu \end{bmatrix}_{2 \times 2}$$

$\uparrow$              $\uparrow$   
 VC1        VC2

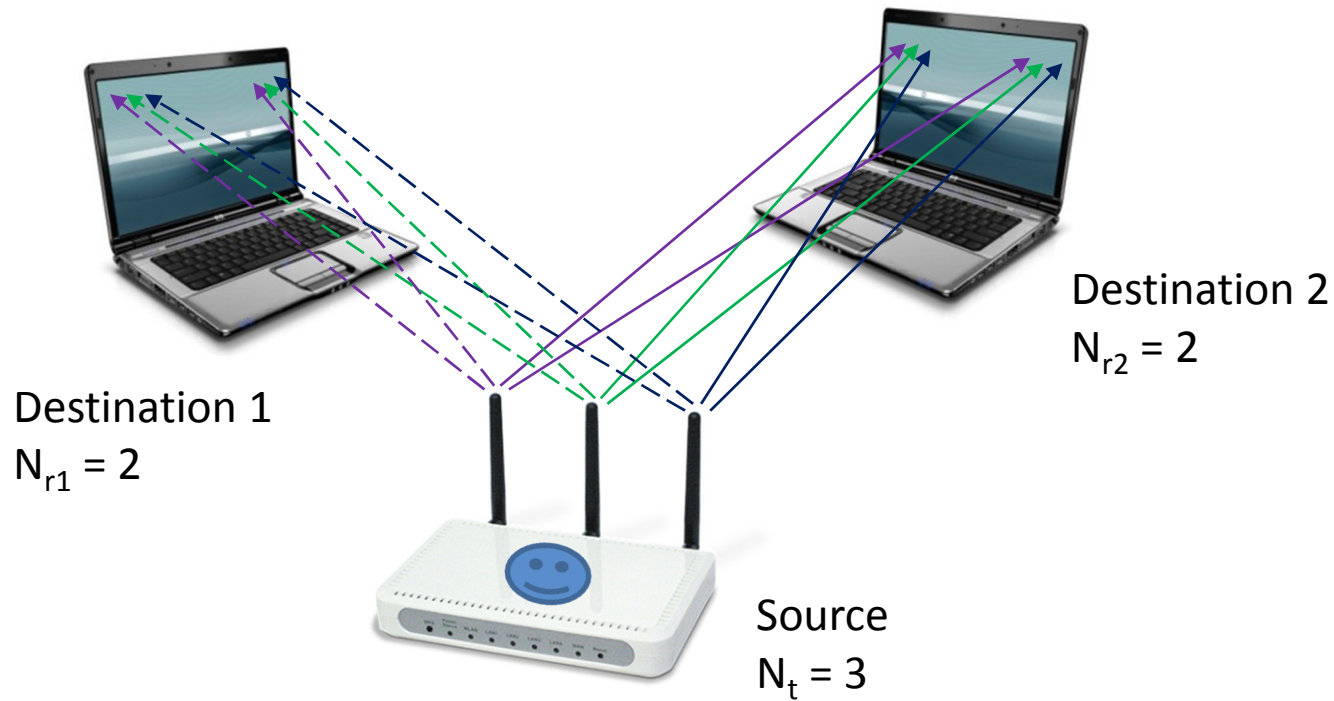
e.g.

- eigenmode transmission
- zero forcing

from linear algebra

- SVD
- pseudo-inverse

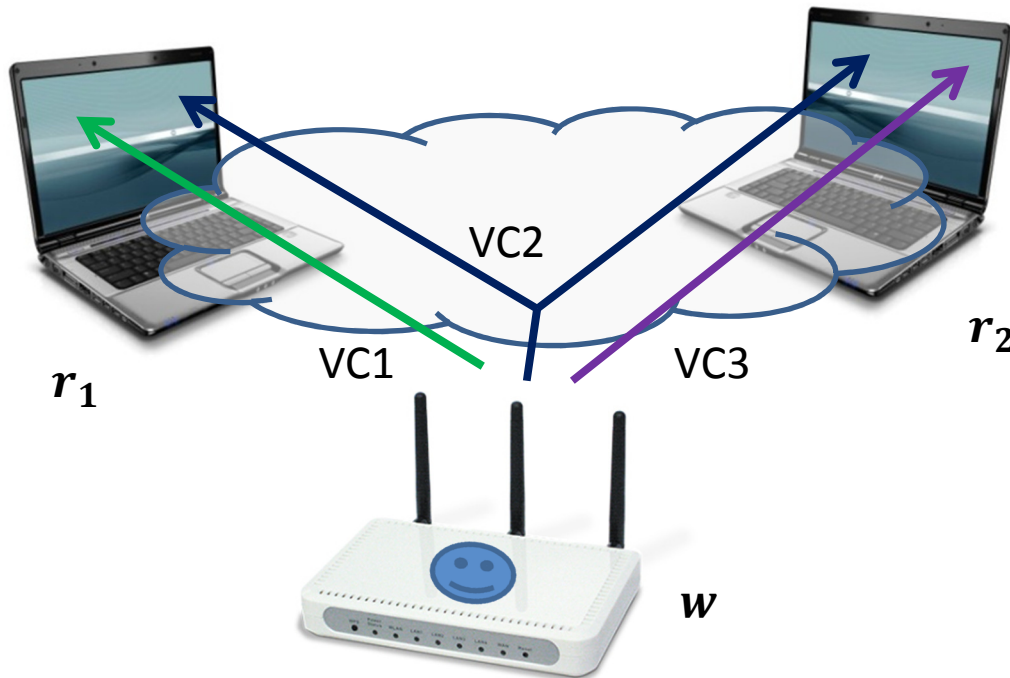
## Two-User MIMO Channel



- Single Source, Two Destinations

$$\begin{aligned} \mathbf{y}_1 &= \mathbf{H}_1 \mathbf{x} + \mathbf{n}_1 \\ \mathbf{y}_2 &= \mathbf{H}_2 \mathbf{x} + \mathbf{n}_2 \end{aligned}$$

# Coordinated Beamforming



$$\begin{aligned}
 \mathbf{y}_1 &= \mathbf{r}_1 \mathbf{H}_1 \mathbf{w} x + \mathbf{n}_1 \\
 \mathbf{y}_2 &= \mathbf{r}_2 \mathbf{H}_2 \mathbf{w} x + \mathbf{n}_2
 \end{aligned}$$

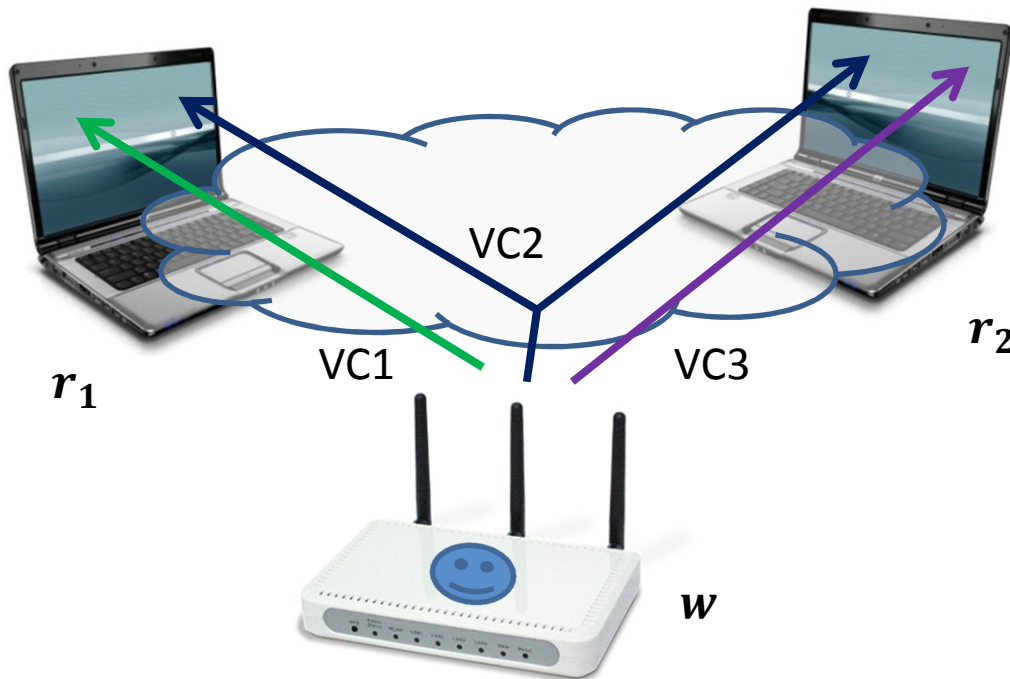
VC1, VC3 : point-to-point : private VCs  
 VC2 : point-to-2 point : common VC

$$\mathbf{r}_1 \mathbf{H}_1 \mathbf{w} = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \mu_1 & 0 \end{bmatrix}_{2 \times 3}$$

$$\mathbf{r}_2 \mathbf{H}_2 \mathbf{w} = \begin{bmatrix} 0 & \mu_2 & 0 \\ 0 & 0 & \lambda_2 \end{bmatrix}_{2 \times 3}$$

$\uparrow$     $\uparrow$     $\uparrow$   
 VC1   VC2   VC3

# Coordinated Beamforming



## GSVD-based Beamforming

- non-iterative
- coordinated b/f scheme
- for 2 users
- possibility hinted in [Khisti2007]

$$\mathbf{r}_1 \mathbf{H}_1 \mathbf{w} = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \mu_1 & 0 \end{bmatrix}_{2 \times 3}$$

$$\mathbf{r}_2 \mathbf{H}_2 \mathbf{w} = \begin{bmatrix} 0 & \mu_2 & 0 \\ 0 & 0 & \lambda_2 \end{bmatrix}_{2 \times 3}$$

$\uparrow$        $\uparrow$        $\uparrow$   
 VC1    VC2    VC3

# Generalized Singular Value Decomposition (GSVD)

- joint matrix decomposition technique

- **Definition 1:** Van Loan form [Loan1976]

*a restricted definition*

Consider two matrices,  $\mathbf{H} \in \mathcal{C}^{m \times n}$  with  $m \geq n$ , and  $\mathbf{G} \in \mathcal{C}^{p \times n}$  having the same number  $n$  of columns. Let  $q = \min(p, n)$ .  $\mathbf{H}$  and  $\mathbf{G}$  can be jointly decomposed as

$$\mathbf{H} = \mathbf{U}\mathbf{\Sigma}\mathbf{Q} \text{ and } \mathbf{G} = \mathbf{V}\mathbf{\Lambda}\mathbf{Q}$$

where

- (i)  $\mathbf{U} \in \mathcal{C}^{m \times m}$ ,  $\mathbf{V} \in \mathcal{C}^{p \times p}$  are unitary,
- (ii)  $\mathbf{Q} \in \mathcal{C}^{n \times n}$  is non-singular,
- (iii)  $\mathbf{\Sigma} = \text{diag}(\sigma_1, \dots, \sigma_n) \in \mathcal{C}^{m \times n}$ ,  $\sigma_i \geq 0$ , and
- (iv)  $\mathbf{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_q) \in \mathcal{C}^{p \times n}$ ,  $\lambda_i \geq 0$ .

- use for b/f:

$$\mathbf{w} = \alpha \mathbf{Q}^{-1}, \mathbf{r}_1 = \mathbf{U}^H, \mathbf{r}_2 = \mathbf{V}^H \quad (\mathbf{r}_1 \mathbf{H} \mathbf{w} = \alpha \mathbf{\Sigma}, \text{ and } \mathbf{r}_2 \mathbf{G} \mathbf{w} = \alpha \mathbf{\Lambda})$$

*normalization coefficient*

- **Definition 2:** Paige & Saunders form [**Paige1981**]

Consider two matrices,  $\mathbf{H} \in \mathcal{C}^{m \times n}$  and  $\mathbf{G} \in \mathcal{C}^{p \times n}$  having the same number  $n$  of columns. Let  $\mathbf{C}^H = [\mathbf{H}^H \quad \mathbf{G}^H] \in \mathcal{C}^{n \times (m+p)}$ ,  $t = \text{rank}(\mathbf{C})$ ,  $r = t - \text{rank}(\mathbf{G})$ , and  $s = \text{rank}(\mathbf{H}) + \text{rank}(\mathbf{G}) - t$ .

$\mathbf{H}$  and  $\mathbf{G}$  can be jointly decomposed as

$$\begin{aligned} \mathbf{H} &= \mathbf{U}[\mathbf{\Sigma} \quad \mathbf{0}_1]\mathbf{Q} = \mathbf{U}\mathbf{\Sigma}\widehat{\mathbf{Q}}, \\ \mathbf{G} &= \mathbf{V}[\mathbf{\Lambda} \quad \mathbf{0}_2]\mathbf{Q} = \mathbf{V}\mathbf{\Lambda}\widehat{\mathbf{Q}}, \end{aligned}$$

where  $\widehat{\mathbf{Q}} = \{\mathbf{Q}\}_{R(t)}$  and

- (i)  $\mathbf{U} \in \mathcal{C}^{m \times m}$ ,  $\mathbf{V} \in \mathcal{C}^{p \times p}$  are unitary,
- (ii)  $\mathbf{Q} \in \mathcal{C}^{n \times n}$  is non-singular,
- (iii)  $\mathbf{0}_1 \in \mathcal{C}^{m \times (n-t)}$ ,  $\mathbf{0}_2 \in \mathcal{C}^{p \times (n-t)}$  are zero matrices
- (iv)  $\mathbf{\Sigma} \in \mathcal{C}^{m \times t}$ ,  $\mathbf{\Lambda} \in \mathcal{C}^{p \times t}$  having structures

$$\mathbf{\Sigma} = \begin{bmatrix} \mathbf{I}_H & & \\ & \widetilde{\mathbf{\Sigma}} & \\ & & \mathbf{0}_H \end{bmatrix}, \text{ and } \mathbf{\Lambda} = \begin{bmatrix} \mathbf{0}_G & & \\ & \widetilde{\mathbf{\Lambda}} & \\ & & \mathbf{I}_G \end{bmatrix}.$$

$\mathbf{I}_H \in \mathcal{C}^{r \times r}$ ,  $\mathbf{I}_G \in \mathcal{C}^{(t-r-s) \times (t-r-s)}$  are unitary.  $\mathbf{0}_H \in \mathcal{C}^{(m-r-s) \times (t-r-s)}$ ,  $\mathbf{0}_G \in \mathcal{C}^{(p-t+r) \times r}$  are zero matrices possibly having no rows or no columns.

$$\widetilde{\mathbf{\Sigma}} = \text{diag}(\sigma_1, \dots, \sigma_s), \widetilde{\mathbf{\Lambda}} = \text{diag}(\lambda_1, \dots, \lambda_s) \in \mathcal{C}^{s \times s}$$

such that  $1 \geq \sigma_1 \geq \dots \geq \sigma_s \geq 0$  and  $\sigma_i^2 + \lambda_i^2 = 1$ .



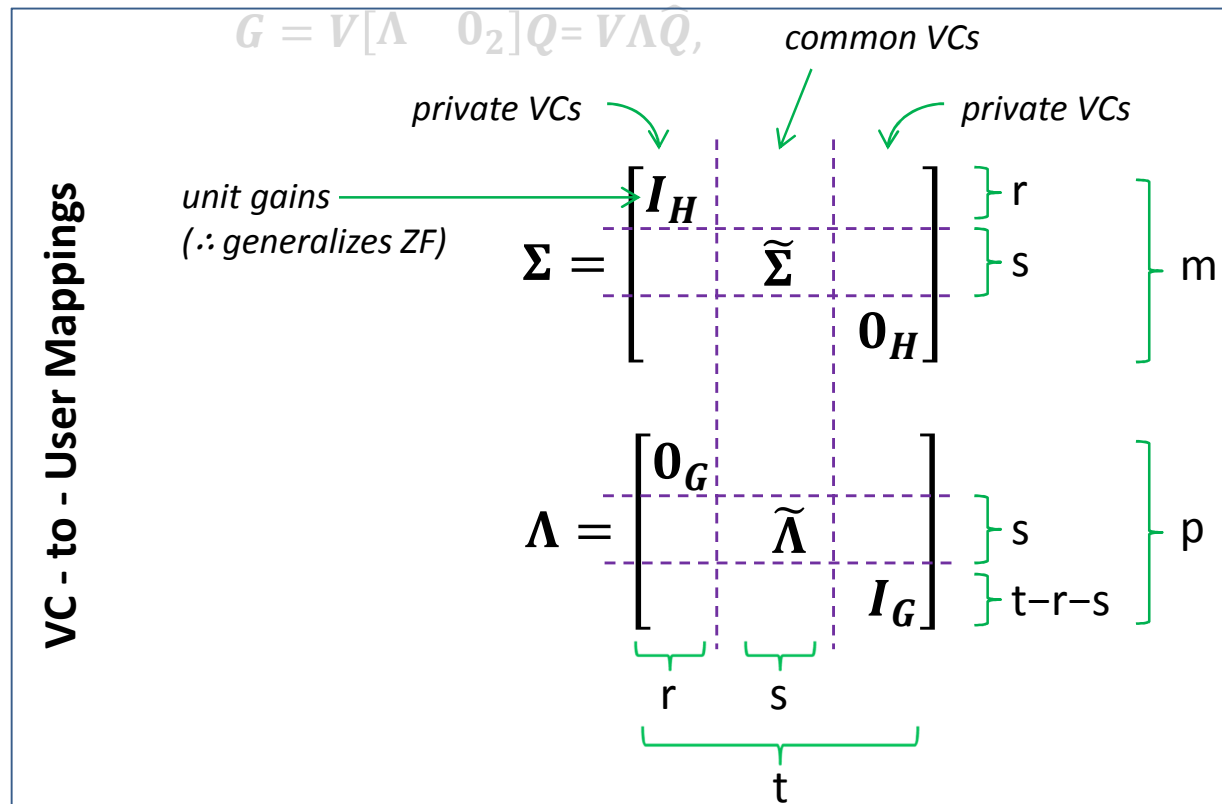
- **Definition 2:** Paige & Saunders form [Paige1981]

Consider two matrices,  $H \in \mathbb{C}^{m \times n}$  and  $G \in \mathbb{C}^{p \times n}$  having the same number  $n$  of columns. Let  $C^H = [H^H \ G^H] \in \mathbb{C}^{n \times (m+p)}$ ,  $t = \text{rank}(C)$ ,  $r = t - \text{rank}(G)$ , and  $s = \text{rank}(H) + \text{rank}(G) - t$ .

$H$  and  $G$  can be jointly decomposed as *holds for arbitrary ranks*

$$H = U[\Sigma \ 0_1]Q = U\Sigma\hat{Q},$$

$$G = V[\Lambda \ 0_2]Q = V\Lambda\hat{Q},$$



$$\tilde{\Sigma} = \text{diag}(\sigma_1, \dots, \sigma_s), \tilde{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_s) \in \mathbb{C}^{s \times s}$$

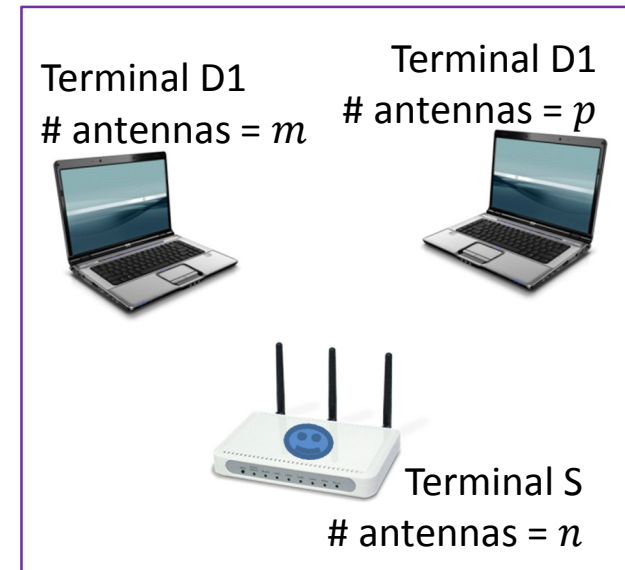
such that  $1 \geq \sigma_1 \geq \dots \geq \sigma_s \geq 0$  and  $\sigma_i^2 + \lambda_i^2 = 1$

*common VC gains*  $\in (0,1)$

# GSVD-based Beamforming

- b/f matrices using GSVD of channel matrices
- number of 'common' vs. 'private' VCs (assuming: no rank deficiency)

Configuration	# common channels $S \rightarrow \{D1, D2\}$	# private channels	
		$S \rightarrow D1$	$S \rightarrow D2$
$m > n, p \leq n$	$p$	$n - p$	0
$m \leq n, p > n$	$m$	0	$n - m$
$m \geq n, p \geq n$	$n$	0	0
$m < n, p < n$ $(m + p) > n$	$m + p - n$	$n - p$	$n - m$
$n \geq (m + p)$	0	$m$	$p$



- MATLAB implementation

```

% channel matrices
H = (randn(m,n)+i*randn(m,n))/sqrt(2);
G = (randn(p,n)+i*randn(p,n))/sqrt(2);
% D1, D2: diagonalized channels
[V,U,X,Lambda,Sigma] = gsvd(G,H);
w = X*inv(X'*X); C = [H' G']'; t = rank(C);
r = t - rank(G); s = rank(H)+rank(G)-t;
D1 = U(:,r+1:r+s)'*H*w(:,r+1:r+s);
D2 = V(:,1:s)'*G*w(:,r+1:r+s);
    
```

## Applications

- MIMO subsystems
- single-source transmitting private/ common messages to 2 destinations
- each destination possibly requiring to process some of the messages

e.g.

- relay networks (AF, DF, CF)
- broadcasting
- physical layer multicasting
- two-way-relay networks

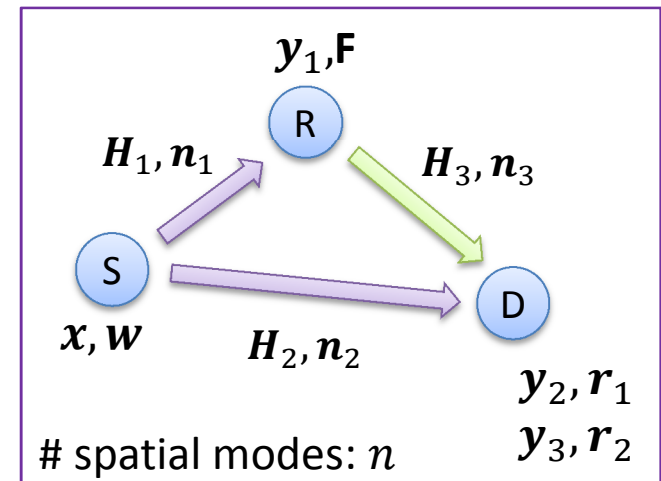
# Example (AF Relaying)

Why?

SVD-based b/f is also possible if effective channel is point-to-point

$$\underline{\text{TS1}}: \mathbf{y}_1 = \mathbf{H}_1 \mathbf{w} \mathbf{x} + \mathbf{n}_1, \mathbf{y}_2 = \mathbf{H}_2 \mathbf{w} \mathbf{x} + \mathbf{n}_2$$

$$\underline{\text{TS2}}: \mathbf{y}_3 = \mathbf{H}_3 \mathbf{F} \mathbf{y}_1 + \mathbf{n}_3$$



GSVD-based b/f (in TS1)

$$\text{GSVD} : \mathbf{H}_1 = \mathbf{U}_1 \mathbf{\Sigma}_1 \mathbf{Q}, \mathbf{H}_2 = \mathbf{U}_2 \mathbf{\Sigma}_2 \mathbf{Q}$$

$$\text{SVD} : \mathbf{H}_3 = \mathbf{V} \mathbf{\Lambda} \mathbf{R}^H$$

$$\Rightarrow \mathbf{w} = \alpha \{ \mathbf{Q}^{-1} \}_{C(n)}, \mathbf{F} = \mathbf{R} \mathbf{U}_1^H$$

$$\mathbf{r}_1 = \{ \mathbf{U}_2^H \}_{R(n)}, \mathbf{r}_2 = \{ \mathbf{V}^H \}_{R(n)}$$

$$\hat{\mathbf{y}} = \mathbf{r}_1 \mathbf{y}_2 + \mathbf{r}_2 \mathbf{y}_3,$$

Assumption: Van Loan form of GSVD holds

Note: an expression for  $\alpha$  not known

SVD-based b/f (for comparison)

$$\hat{\mathbf{y}} = \mathbf{r} \begin{bmatrix} \mathbf{y}_3 \\ \mathbf{y}_2 \end{bmatrix} = \mathbf{r} \left( \underbrace{\begin{bmatrix} \mathbf{H}_3 \mathbf{F} \mathbf{H}_1 \\ \mathbf{H}_2 \end{bmatrix}}_{\hat{\mathbf{H}}} \mathbf{w} \mathbf{x} + \begin{bmatrix} \mathbf{H}_3 \mathbf{F} \mathbf{n}_1 \\ \mathbf{0} \end{bmatrix} + \begin{bmatrix} \mathbf{n}_3 \\ \mathbf{n}_2 \end{bmatrix} \right)$$

$$\text{SVD} : \hat{\mathbf{H}} = \hat{\mathbf{U}} \hat{\mathbf{\Sigma}} \hat{\mathbf{V}}^H,$$

$$\mathbf{H}_1 = \mathbf{U}_1 \mathbf{\Sigma}_1 \mathbf{V}_1^H, \mathbf{H}_3 = \mathbf{U}_3 \mathbf{\Sigma}_3 \mathbf{V}_3^H$$

$$\Rightarrow \mathbf{w} = \{ \hat{\mathbf{V}} \}_{C(n)}, \mathbf{F} = \mathbf{V}_3 \mathbf{U}_1^H,$$

$$\mathbf{r} = \{ \hat{\mathbf{U}}^H \}_{R(n)} = \mathbf{r}_1 = \mathbf{r}_2$$

Note: spatial modes not easily resolved at **R**

- not usable with DF

optimal  $\mathbf{F}$  not always known

# Example (AF Relaying)...

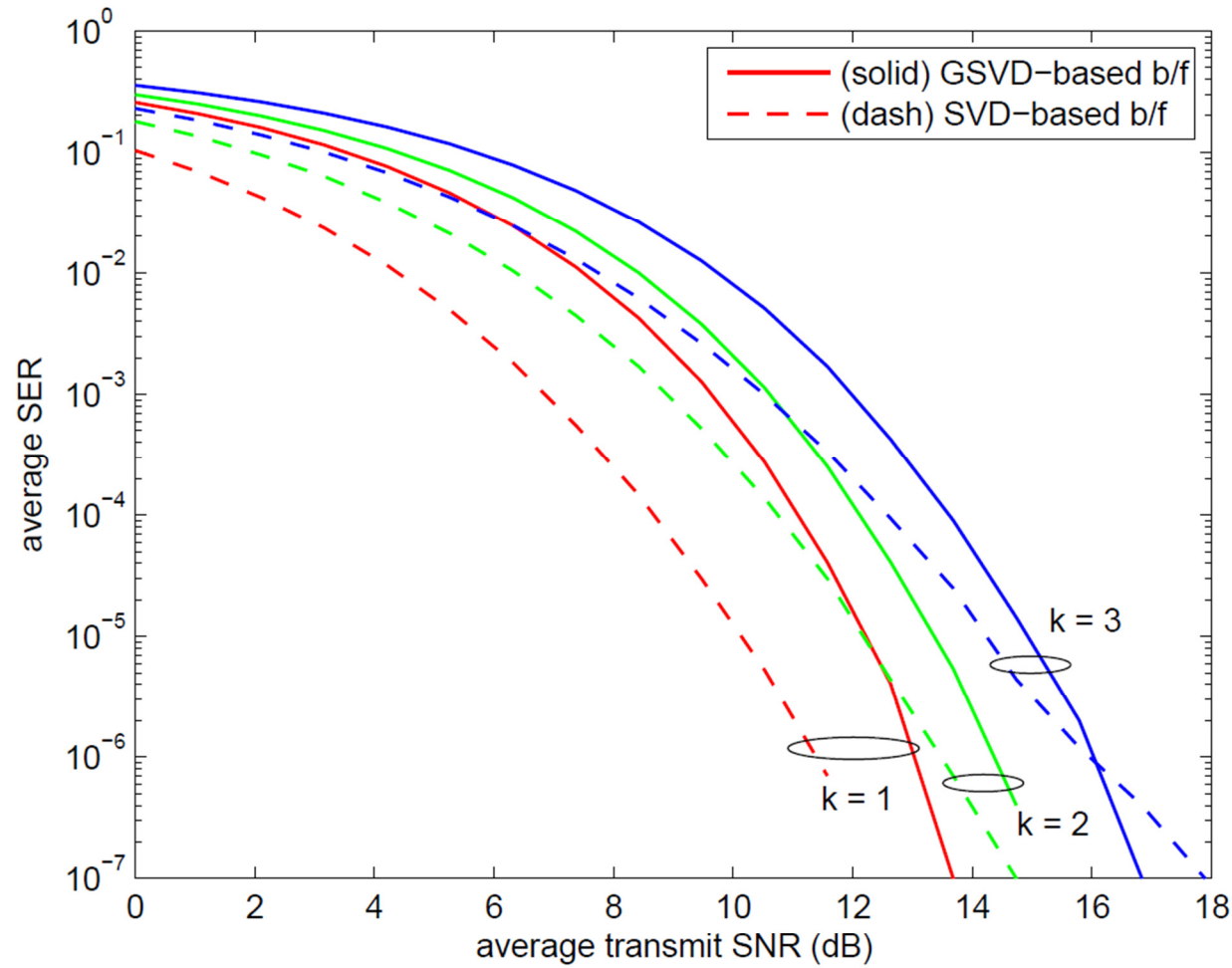


Fig. 6. The average SER of GSVD-based beamforming vs. SVD-based beamforming, for MIMO AF relaying with  $N_s = 4, N_r = 3, N_d = 5$ , for  $n = 3$  common channels.

## References

- **[Khisti2007]** A. Khisti, G. Wornell, A. Wiesel, and Y. Eldar, “On the Gaussian MIMO wiretap channel,” in *Proc. Information Theory, IEEE International Symposium on*, Nice, France, Jun. 2007, pp. 2471–2475.
- **[Loan1976]** C. F. V. Loan, “Generalizing the singular value decomposition,” *SIAM Journal on Numerical Analysis*, vol. 13, no. 1, pp. 76–83, Mar. 1976.
- **[Paige1981]** C. C. Paige and M. A. Saunders, “Towards a generalized singular value decomposition,” *SIAM Journal on Numerical Analysis*, vol. 18, no. 3, pp. 398–405, Jun. 1981.

## Conclusion

- the use of GSVD for coordinated b/f examined
  - a generalization of ZF (for single-source, two-destinations)
  - similarity with SVD-based b/f
- possible applications highlighted [see: paper]
- AF relaying example
  - a comparison with SVD-based b/f
  - SVD-based b/f possible only if the effective channel is point-to-point  
e.g. single-source, single-destination, AF relaying
- improvements
  - robustness
    - perfect CSI requirement
    - matrix inversion ( $\mathbf{Q}^{-1}$ )
  - extending random matrix theory for analysis

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GSVD-based b/f : another 'tool' like SVD-based b/f