

6.962 Graduate Seminar in Communications

Week 8

Space-Time Modulation for Unknown Fading - Bertrand M. Hochwald and Thomas L. Marzetta

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Motivation

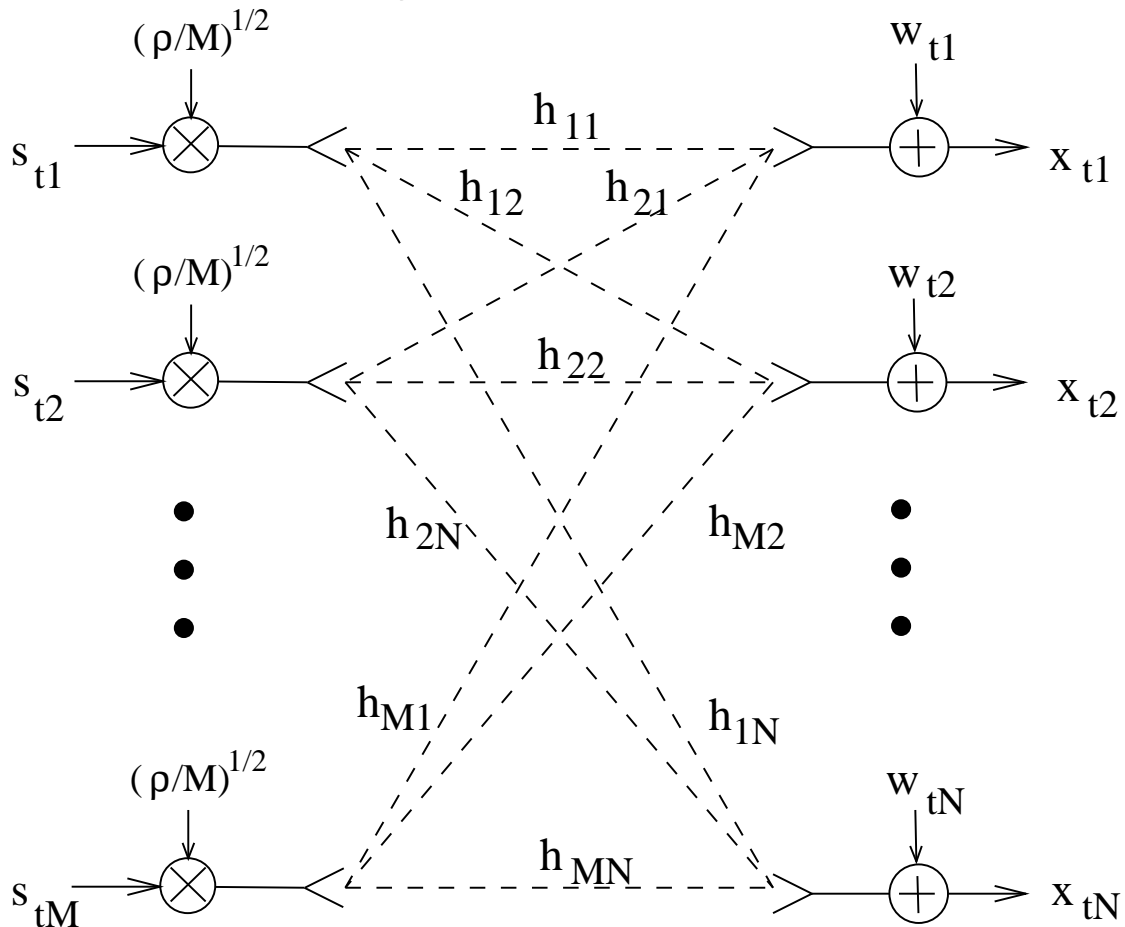
Coherent

- Assume channel is known.
- Need to train / track channel coefficients.
- More antennas, more coefficients.

Non-Coherent

- Channel unknown.
- Non-explicit channel estimation.
- A more general method.

System Model



- M transmitter antennas.
- N receiver antennas.
- Rayleigh flat-fading.
- Piecewise constant for T .
- Additive White Gaussian Noise.

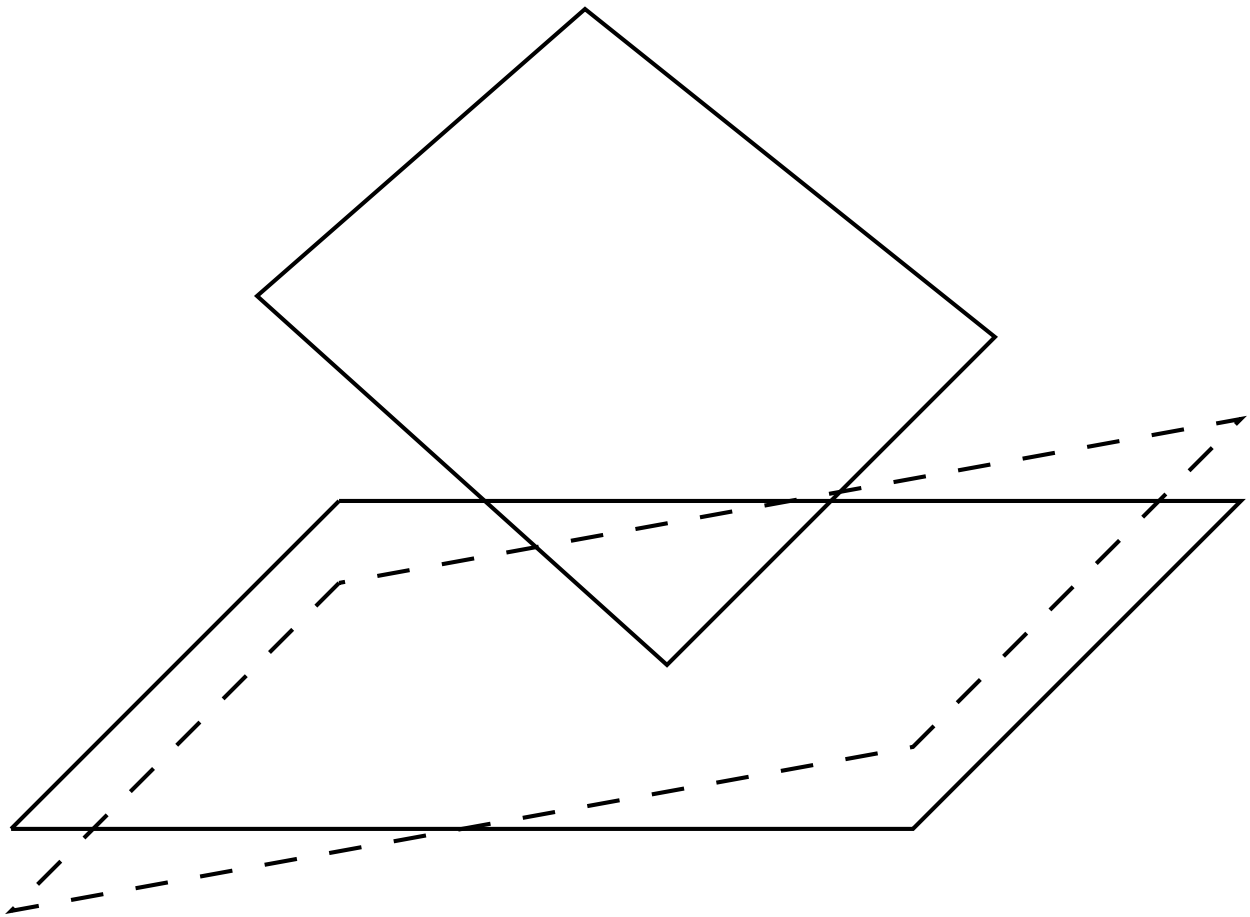
Matrix Representation

$$X = \sqrt{\frac{\rho}{M}}SH + W.$$

$$S = \begin{bmatrix} s_{11} & \cdots & s_{1M} \\ \vdots & \ddots & \vdots \\ s_{T1} & \cdots & s_{TM} \end{bmatrix}, \quad X = \begin{bmatrix} x_{11} & \cdots & x_{1N} \\ \vdots & \ddots & \vdots \\ x_{T1} & \cdots & x_{TN} \end{bmatrix},$$
$$H = \begin{bmatrix} h_{11} & \cdots & h_{1N} \\ \vdots & \ddots & \vdots \\ h_{M1} & \cdots & h_{MN} \end{bmatrix}, \quad W = \begin{bmatrix} w_{11} & \cdots & w_{1N} \\ \vdots & \ddots & \vdots \\ w_{T1} & \cdots & w_{TN} \end{bmatrix},$$

$$\frac{1}{M} \sum_{m=1}^M E(|s_{tm}|^2) = 1, \text{ for } t = 1, \dots, T.$$

A Subspace Picture



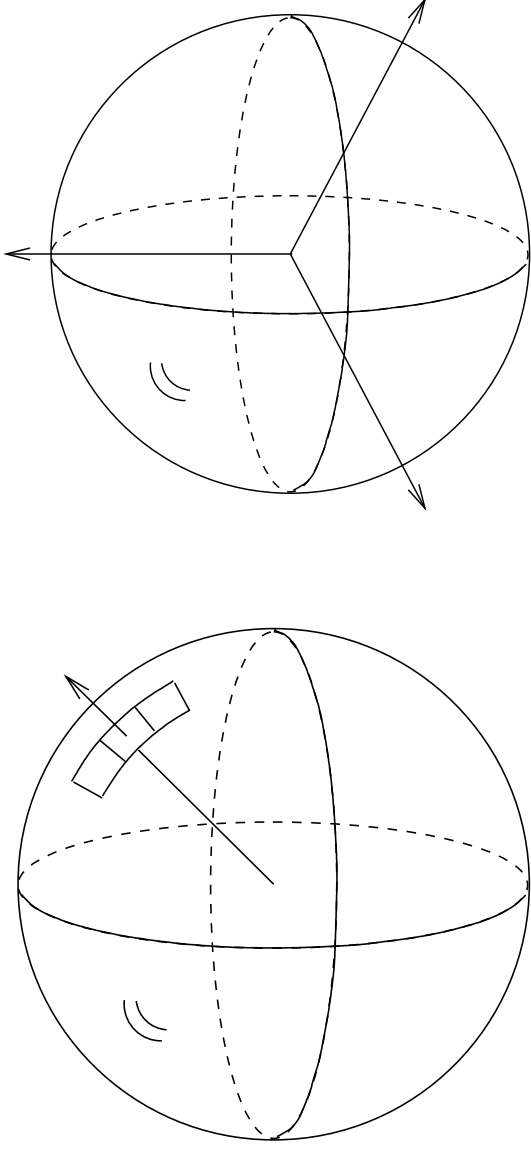
Channel Input/Output Probability

$$p(X|S) = \frac{\exp \left(-\text{tr} \left\{ [I_T + (\rho/M)SS^\dagger]^{-1} XX^\dagger \right\} \right)}{\pi^{TN} \det^N [I_T + (\rho/M)SS^\dagger]}$$

$$\begin{aligned} H &\sim I \\ SH &\sim SS^\dagger \\ X = \sqrt{\frac{\rho}{M}}SH + W &\sim I_T + (\rho/M)SS^\dagger \end{aligned}$$

- It is SS^\dagger that matters in design.
- Don't need $M > T$.
- $p(X|S\Phi^\dagger) = p(X|S) \quad \forall \text{ unitary } \Phi$

Isotropically Distributed (i.d.) unit vectors and unitary matrices



rotational symmetry: $p(\Phi) = p(\Theta^\dagger \Phi), \forall \Theta : \Theta^\dagger \Theta = I$

Capacity Achieving Distribution

$$p(X|S) = \frac{\exp \left(-\text{tr} \left\{ [I_T + (\rho/M)SS^\dagger]^{-1} XX^\dagger \right\} \right)}{\pi^{TN} \det^N [I_T + (\rho/M)SS^\dagger]}$$

$$I(X; S) = E_{X,S} \left\{ \log \left(\frac{p(X|S)}{E_S \{p(X|S)\}} \right) \right\}$$

$$C(X; S) = \sup_{p(S)} I(X; S)$$

Theorem: The signal matrix that achieves capacity can always be factored as $S = \Phi V$, where Φ is an $T \times M$ isotropically distributed unitary matrix, and V is an independent $M \times M$ real, nonnegative, diagonal matrix.

A Quick Proof

First, show $S = \Phi V$ by performing SVD on S .

$$S = \Phi V \Psi^\dagger$$

$$SS^\dagger = (\Phi V \Psi^\dagger)(\Psi V \Phi^\dagger) = (\Phi V)(\Phi V)^\dagger$$

Second, show Φ is i.d. by rotating S .

$$S' = \Theta S = \Theta \Phi V$$

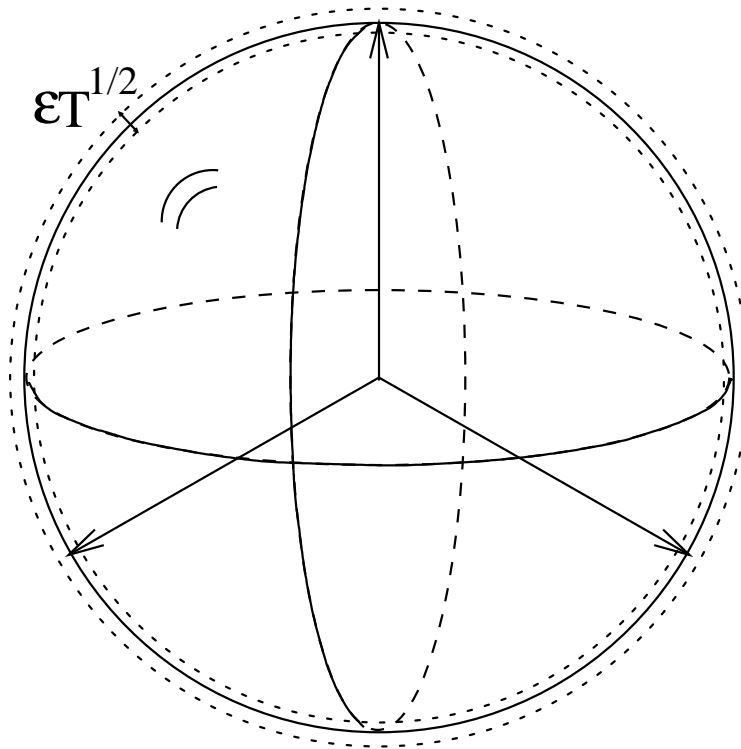
$$\Theta \text{ is i.d.} \implies \Theta \Phi \text{ is i.d.}$$

Capacity Achieving Distribution for \mathbf{V}

for fixed M , as $T \rightarrow \infty$, that $\mathbf{V} \rightarrow \sqrt{T} \mathbf{I}_M$.

Intuition:

- No preference to any antenna.
- Since average energy is T , in the high dimension limit, most vectors lie in a thin shell of thickness $\epsilon\sqrt{T}$ and radius \sqrt{T} .



Deterministic Codebook Design

We want to design

$$S_1 = \Phi_1 V_1, S_2 = \Phi_2 V_2, \dots, S_L = \Phi_L V_L,$$

$$\text{where } L = 2^{RT} \text{ and } \Phi_l^\dagger \Phi_l = I.$$

We can pick $V_1 = V_2 = \dots = V_L = \sqrt{T}$.

We still need to find $\Phi_1, \Phi_2, \dots, \Phi_L$.

Maximum Likelihood Non-Coherent Decoder

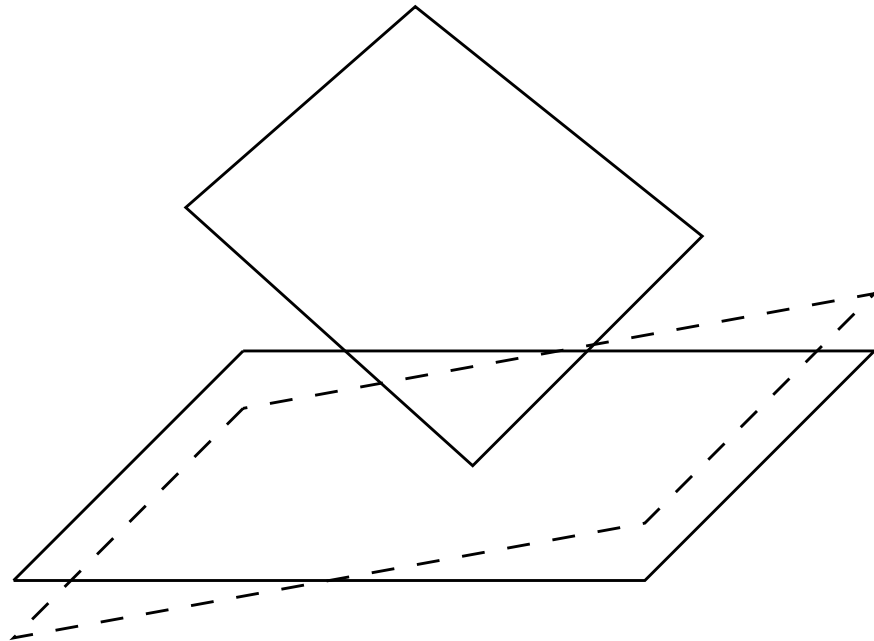
$$\begin{aligned}\Phi_{ml} &= \operatorname{argmax}_{\Phi_l=\Phi_1,\dots,\Phi_L} \operatorname{tr}\{X^\dagger \Phi_l \Phi_l^\dagger X\} \\ &= \operatorname{argmax}_{\Phi_l=\Phi_1,\dots,\Phi_L} \sum_{i,j} |[\Phi_l^\dagger X]_{ij}|^2\end{aligned}$$

Assume l th signal is transmitted.

$$\begin{aligned}X &= \sqrt{\frac{\rho}{M}} S_l H + W \\ &= \sqrt{\frac{\rho T}{M}} \Phi_l H + \Phi_l \Phi_l^\dagger W + (W - \Phi_l \Phi_l^\dagger W) \\ &= \sqrt{\frac{\rho T}{M}} \Phi_l H + W_l^{in} + W_l^{out} \\ \Phi_l^\dagger X &= \sqrt{\frac{\rho T}{M}} H + \Phi_l^\dagger W_l^{in}\end{aligned}$$

The ML decoder $\Phi_{ml} = \operatorname{argmax}_{\Phi_l} \operatorname{tr}\{X^\dagger \Phi_l \Phi_l^\dagger X\}$

- Maximize energy in $\Phi_l^\dagger X$.
- Minimize energy in W_l^{out} .
- Find nearest valid subspace.



Channel Estimation

Channel estimation falls right out of the ML non-coherent decoder.

Assume l th signal is transmitted.

$$\begin{aligned}\hat{H}_l &= \left(\frac{\rho T}{M}\right)^{(-\frac{1}{2})} \Phi_l^\dagger X \\ &= H + \left(\frac{\rho T}{M}\right)^{(-\frac{1}{2})} \Phi_l^\dagger W_l^{in}\end{aligned}$$

This ML non-coherent decoder has an equivalent interpretation as a generalized likelihood ratio test (GLRT).

$$\Phi_{ml} = \operatorname{argmax}_{\Phi_l} \operatorname{argmax}_{H_l} p(X|\Phi_l, H_l)$$

Two-Signal Probability of Error

$$P_e \leq \frac{1}{2} \prod_{m=1}^M \left[\frac{1}{1 + \frac{(\rho T/M)^2 (1-d_m^2)}{4(1+\rho T/M)}} \right]^N$$

where $1 \geq d_1 \geq \dots \geq d_M \geq 0$ are the singular values of the $M \times M$ correlation matrix $\Phi_2^\dagger \Phi_1$.

- P_e is the lowest when $d_1 = \dots = d_M = 0$.
- P_e is the highest when $d_1 = \dots = d_M = 1$.

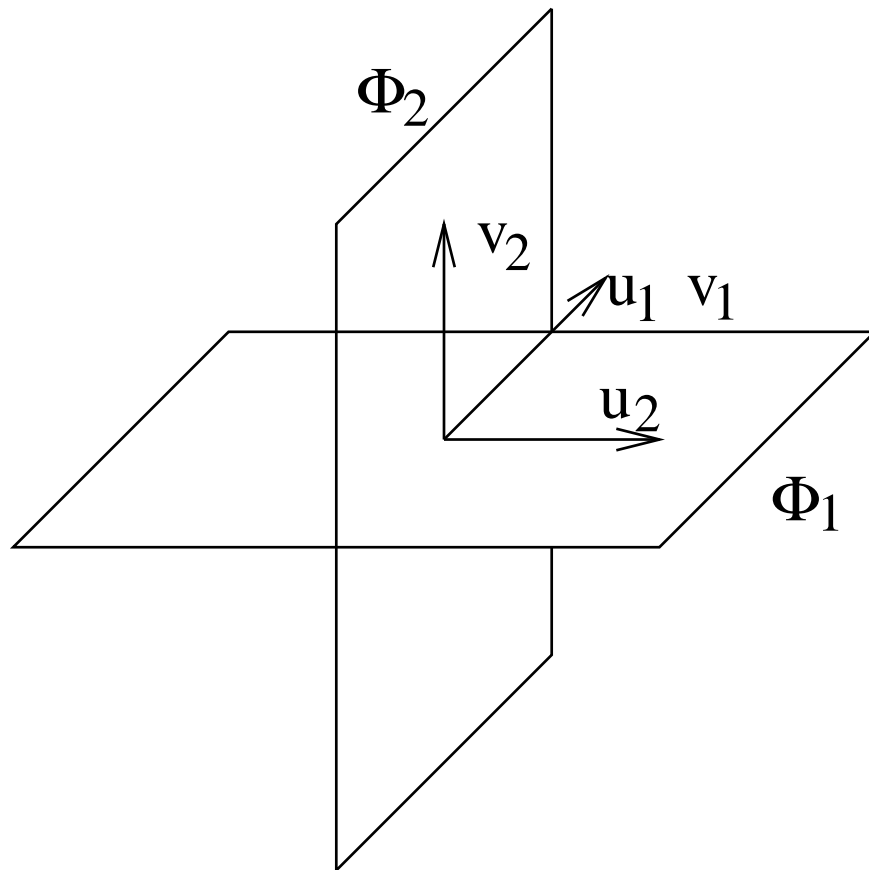
The d'_i s corresponds to the cosines of a set of *principle angles* defined between the two subspaces Φ_1 and Φ_2 .

Principle Angles

A set of *principle angles*, $\theta_1, \dots, \theta_M$, can be defined between two M -dimensional subspaces Φ_1 and Φ_2 by

- Pick vectors $u_1 \in \Phi_1$, $v_1 \in \Phi_2$, s.t., $u_1^\dagger v_1$ is maximized, i.e., having the smallest angle in between. $u_1^\dagger v_1 = d_1 = \cos \theta_1$.
- Pick vectors $u_2 \in \Phi_1$, $v_2 \in \Phi_2$, s.t., $u_2 \perp u_1$ and $v_2 \perp v_1$, and $u_2^\dagger v_2$ is maximized.
 $u_2^\dagger v_2 = d_2 = \cos \theta_2$.
- \vdots
- Pick vectors $u_k \in \Phi_1$, $v_k \in \Phi_2$, s.t., $u_k \perp u_j$ and $v_k \perp v_j$ for $j = 1, \dots, k-1$, and $u_k^\dagger v_k$ is maximized.
 $u_k^\dagger v_k = d_k = \cos \theta_k$.
- Continue until $k = M$.

Principle Angles Example



$$d_1 = 1 \quad d_2 = 0$$

$$\theta_1 = 0^\circ \quad \theta_2 = 90^\circ$$

Iterative Design Method for $M = 1$

We want to place L vectors in a T -dimensional space with maximum separation between the nearest pair.

1. Compute d_{max} , the maximum of the magnitudes of all $L(L - 1)/2$ distinct inner products, and choose a pair of vectors whose inner product is d_{max} .
2. “Separate” the pair by moving each vector a small amount in opposite directions along the difference vector between the pair.
3. Renormalize the pair, if needed.
4. Repeat Steps 1-3 until d_{max} stops decreasing.

Deterministic Codebook Design Example : $M = 1$

$$T = 2 \quad L = 2^2 = 4$$

$$d_{max} = 0.707 \quad \theta_{min} = 45^\circ$$

$$T = 5 \quad L = 2^5 = 32$$

$$d_{max} = 0.515 \quad \theta_{min} = 59^\circ$$

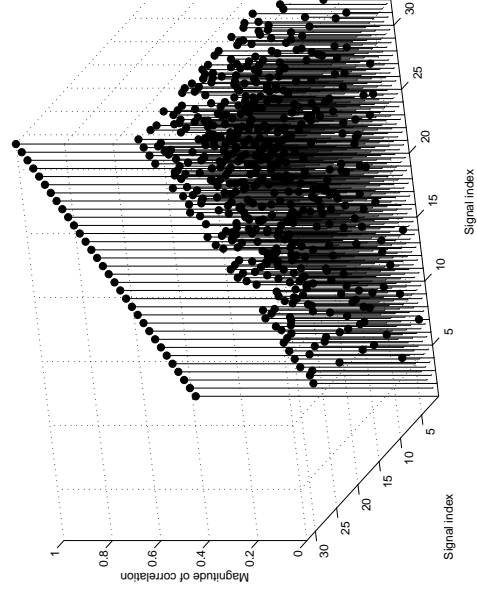
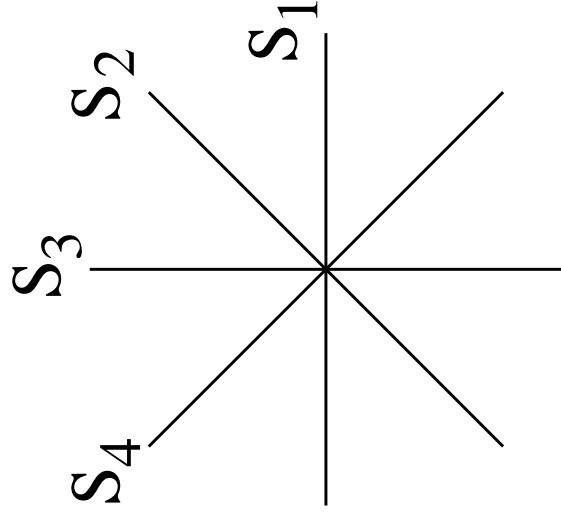


Figure : Magnitudes of correlations between Φ_1, \dots, Φ_{32} for $T = 5$. The diagonal entries with value 1.0 represent each signal correlated with itself.

Deterministic Codebook Design for $M > 1$

For each pair of Φ_l and $\Phi_{l'}$, there is a whole set of (d_1, \dots, d_M) . To determine which pair is the “closest”, we can use

- The Chernoff Bound :

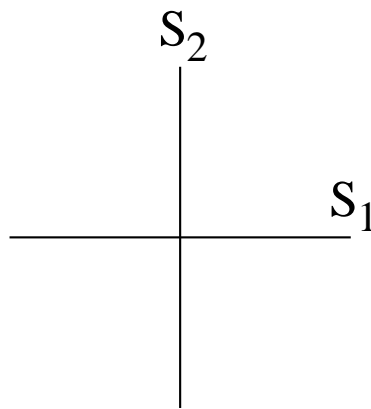
$$P_e \leq \frac{1}{2} \prod_{m=1}^M \left[\frac{1}{1 + \frac{(\rho T/M)^2(1-d_m^2)}{4(1+\rho T/M)}} \right]^N$$

- The Frobenius Norm:

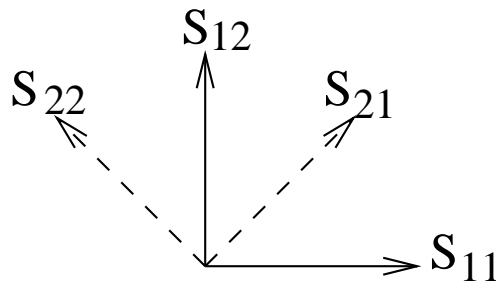
$$\|\Phi_l^\dagger \Phi_{l'}\| = \sqrt{\frac{d_1^2 + \dots + d_M^2}{M}}$$

Example: $T = 2, L = 2, M = 1$ vs. $M = 2$

$M = 1$: Two orthogonal 1-D subspaces,
 $d_1 = 0, \theta_1 = 90^\circ$



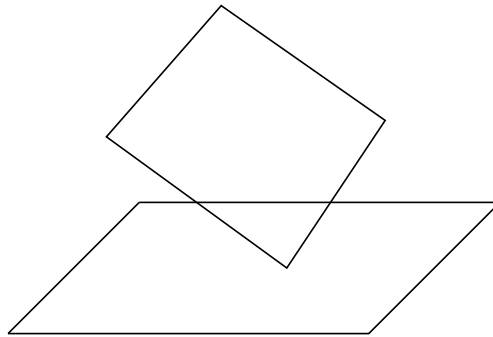
$M = 2$: Two 2-D subspaces entirely overlapping,
 $d_1 = d_2 = 1, \theta_1 = \theta_2 = 0^\circ$



We don't want to use $M = T$.

Example: $T = 4$, $L = 2$, $M = 2$ vs. $M = 3$

$M = 2$: $d_1 = d_2 = 0$, $\theta_1 = \theta_2 = 90^\circ$.



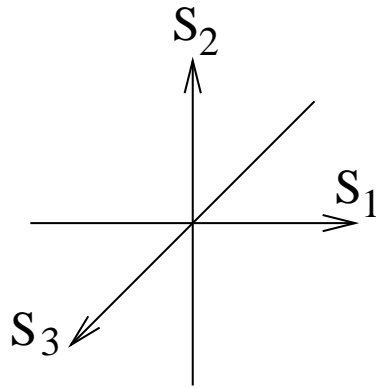
$M = 3$: The intersection of two 3-D subspaces in 4-D space is a 2-D plane. $3 + 3 - 4 = 2$

$$\begin{aligned} d_1 = d_2 = 1 & \quad d_3 = 0 \\ \theta_1 = \theta_2 = 0^\circ & \quad \theta_3 = 90^\circ \end{aligned}$$

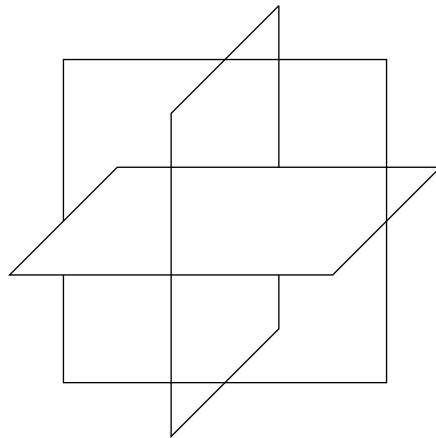
We don't want to use $M > \frac{T}{2}$.

Example: $T = 3$, $L = 3$, $M = 1$ vs. $M = 2$

$M = 1$: Between every pair, $d_1 = 0$ and $\theta_1 = 90^\circ$.



$M = 2$: $d_1 = 1$ and $d_2 = 0$, $\theta_1 = 0^\circ$ and $\theta_2 = 90^\circ$.



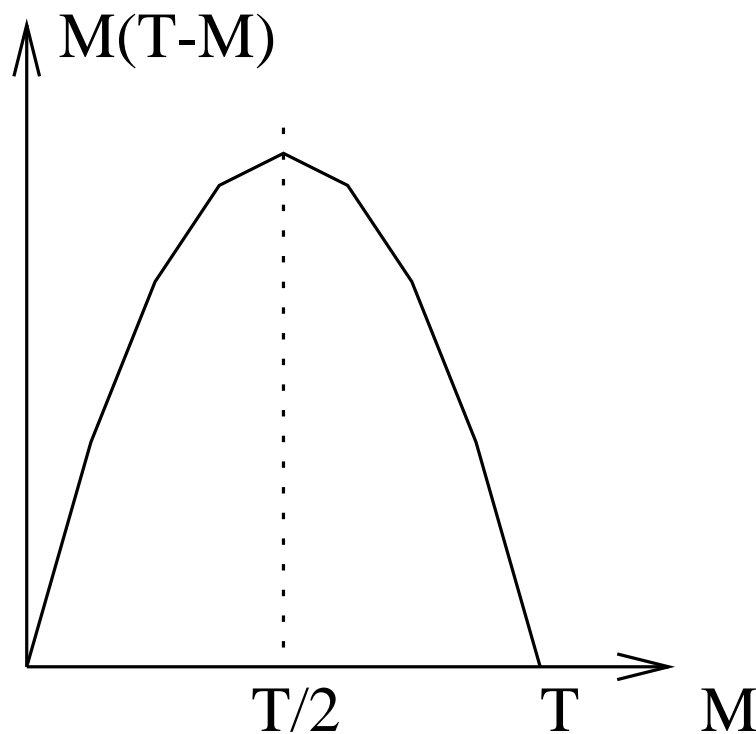
They have the same Chernoff's bound, equally good.

What M is Optimal?

We want to maximize $M(T-M)$. $M = \lfloor \frac{T}{2} \rfloor$ is optimal.

Interpretation: (training view)

- Need time M to solve for channel coefficients.
- Has time $(T - M)$ for data transmission.
- Capacity $\propto M$ when channel is known.



Systematic Design of Unitary Space-Time Codebook

How do we design good codebook efficiently when $L = 2^{RT}$ is large?

$$\Phi_l = \Theta^{l-1} \Phi_1, \quad l = 1, \dots, L$$

where Θ is a $T \times T$ diagonal unitary matrix such that $\Theta^L = I_T$, and Φ_1 is a unitary matrix, $\Phi_1^\dagger \Phi_1 = I_T$.

- Any Φ_l is unitary.

$$\Phi_l^\dagger \Phi_l = \Phi_1^\dagger \Theta^{-(l-1)} \Theta^{l-1} \Phi_1 = \Phi_1^\dagger \Phi_1 = I_T$$

- The correlation matrix

$$\Phi_i^\dagger \Phi_j = \Phi_1^\dagger \Theta^{-(i-1)} \Theta^{j-1} \Phi_1 = \Phi_1^\dagger \Theta^{(j-i) \bmod L} \Phi_1$$

- For every matrix Φ_l , the set of correlation matrices it forms with the others are

$$\{\Phi_1^\dagger \Theta \Phi_1, \Phi_1^\dagger \Theta^2 \Phi_1, \dots, \Phi_1^\dagger \Theta^{L-1} \Phi_1\}$$

Design of Θ

$$\Theta^L = I_T \implies [\Theta]_{tt} = e^{i\frac{2\pi}{L}u_t}, \quad t = 1, \dots, T,$$

where the u_t are integers, $0 \leq u_1, \dots, u_T \leq L - 1$.

We only need to search over $\binom{T+L-1}{T}$ sets of $\{u_t\}$ to minimize a cost criterion, such as Chernoff's bound or Frobenius norm.

example : For $T = 4$ and $L = 4$

$$\binom{T+L-1}{T} = \binom{7}{4} = \frac{7 \cdot 6 \cdot 5 \cdot 4}{4 \cdot 3 \cdot 2 \cdot 1} = 35$$

- Choice of Φ_1 is less critical.

Systematic Design Example

$$T = 4, M = 2, L = 4, \text{ and } \Phi_1 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 1 \\ 1 & -1 \end{bmatrix}.$$

Criterion: Minimize the worst case Frobenius norm

$$\delta = \max_{1 \leq l < l' \leq L} \left\| \Phi_l^\dagger \Phi_{l'} \right\| = \max_{1 \leq l < l' \leq L} \sqrt{\frac{d_1^2 + \dots + d_M^2}{M}}$$

Result: $\{u_t\} = 0, 0, 1, 2$, $\Theta = \text{diag}(1, 1, i, -1)$.

	Φ_2	Φ_3	Φ_4
	$\begin{bmatrix} 1 & 1 \\ 1 & -1 \\ i & i \\ -1 & 1 \end{bmatrix}$	$\begin{bmatrix} 1 & 1 \\ 1 & -1 \\ -1 & -1 \\ 1 & -1 \end{bmatrix}$	$\begin{bmatrix} 1 & 1 \\ 1 & -1 \\ -i & -i \\ -1 & 1 \end{bmatrix}$
$d_1(\theta_1)$	$\frac{1}{\sqrt{2}}(45^\circ)$	$1(0^\circ)$	$\frac{1}{\sqrt{2}}(45^\circ)$
$d_2(\theta_2)$	$0(90^\circ)$	$0(90^\circ)$	$0(90^\circ)$
$\left\ \Phi_l^\dagger \Phi_1 \right\ $	$1/2$	$\frac{1}{\sqrt{2}}$	$1/2$

If I use a different $\Phi_1 \dots$

$$T = 4, M = 2, L = 4, \text{ and } \Phi_1 = \begin{bmatrix} 1 & 1 \\ 1 & i \\ 1 & -1 \\ 1 & -i \end{bmatrix}.$$

Result: $\{u_t\} = 0, 1, 2, 3, \Theta = \text{diag}(1, i, -1, -i).$

$$\begin{array}{ccc} \Phi_2 & \Phi_3 & \Phi_4 \\ \begin{bmatrix} 1 & 1 \\ i & -1 \\ -1 & 1 \\ -i & -1 \end{bmatrix} & \begin{bmatrix} 1 & 1 \\ -1 & -i \\ 1 & -1 \\ -1 & i \end{bmatrix} & \begin{bmatrix} 1 & 1 \\ -i & 1 \\ -1 & 1 \\ i & 1 \end{bmatrix} \end{array}$$

$$\begin{array}{cccc} d_1(\theta_1) & 1(0^\circ) & 0(90^\circ) & 1(0^\circ) \\ d_2(\theta_2) & 0(90^\circ) & 0(90^\circ) & 0(90^\circ) \\ \left\| \Phi_l^\dagger \Phi_1 \right\| & \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{array}$$

$$4 \times 4 \text{ FFT matrix } \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{bmatrix}$$

Summary

- Described a non-coherent model for a multiple antenna system, assuming flat-fading and piecewise constant channel, $X = \sqrt{\frac{\rho}{M}}SH + W$.
- Structure of capacity achieving signal, $S = \Phi V$.
- Maximum likelihood decoder,
 $\Phi_{ml} = \operatorname{argmax}_{\Phi_l} \operatorname{tr}\{X^\dagger \Phi_l \Phi_l^\dagger X\}$.
- Two signal probability of error,
 P_e related to $d_m = \cos \theta_m$.
- Design examples with $M = 1$ and $M > 1$.
- $M = \lfloor \frac{T}{2} \rfloor$ is optimal.
- Systematic design method, $\Phi_l = \Theta^{l-1} \Phi_1$.