

Adaptive Space-Time Shift Keying Systems

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Outline

- 1 Introduction
 - Motivations
- 2 CSTSK MIMO System
 - Transmitter Model
 - Receiver Model
- 3 Adaptive CSTSK MIMO System
 - Training Based Adaptive CSTSK
 - Semi-Blind Iterative Scheme
- 4 Conclusions

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MIMO Landscape

MIMO: **Space** and **Time** dimensions; **Diversity** and **Multiplexing** gains

- **Vertical Bell Lab layered space-time** (V-BLAST)
 - Offers high multiplexing gain at high decoding complexity owing to inter-channel interference (ICI)
- **Space-time block codes** (STBCs)
 - Maximum diversity gain at expense of bandwidth efficiency
- **Linear dispersion codes** (LDCs)
 - Flexible tradeoff between diversity and multiplexing gains
- **Spatial modulation** (SM) and **space-shift keying** (SSK)
 - Mainly multiplexing gain, can achieve receive diversity
 - No ICI \Rightarrow low-complexity single-antenna ML detection

Unified MIMO Architecture

- **Space-time shift keying** (STSK): unified MIMO including V-BLAST, STBCs, LDCs, SM and SSK as special cases
 - Fully exploit both spatial and time dimensions
 - Flexible diversity versus multiplexing gain tradeoff
 - **No ICI** with low-complexity single-antenna ML detection
- **Coherent** STSK (CSTSK):
 - Better performance and flexible design
 - Requires channel state information (CSI)
- **Differential** STSK:
 - Doubling noise power, limited design in modulation scheme and choice of linear dispersion matrices
 - **No need for CSI**

Coherent MIMO

- Ability of an MIMO system to approach its **capacity** heavily relies on accuracy of CSI
- **Training** based schemes: capable of accurately estimating MIMO channel at expense of large training overhead \Rightarrow considerable reduction in system throughput
- **Blind** methods: high complexity and slow convergence, also unavoidable estimation and decision ambiguities
- **Semi-blind** methods offer attractive practical means of implementing adaptive MIMO systems
 - Low-complexity ML data detection in STSK \Rightarrow efficient semi-blind iterative channel estimation and data detection

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Transmitted Signal

- Each block $\mathbf{S}(i) \in \mathbb{C}^{N_T \times T_n}$ is generated from $\log_2(L \cdot Q)$ bits by

$$\mathbf{S}(i) = s(i)\mathbf{A}(i)$$

- $\log_2(L)$ bits decides $s(i)$ from L -PSK/QAM modulation scheme

$$s(i) \in \mathcal{S} = \{s_l \in \mathbb{C}, 1 \leq l \leq L\}$$

- $\log_2(Q)$ bits selects $\mathbf{A}(i)$ from set of Q dispersion matrices

$$\mathbf{A}(i) \in \mathcal{A} = \{\mathbf{A}_q \in \mathbb{C}^{N_T \times T_n}, 1 \leq q \leq Q\}$$

Each dispersion matrix meets power constraint $\text{tr}[\mathbf{A}_q^H \mathbf{A}_q] = T_n$

- Normalised throughput per time-slot of this CSTSK scheme is

$$R = \frac{\log_2(Q \cdot L)}{T_n} \text{ [bits/symbol]}$$

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Received Signal

- Received signal matrix $\mathbf{Y}(i) \in \mathbb{C}^{N_R \times T_n}$ takes MIMO model

$$\mathbf{Y}(i) = \mathbf{H}\mathbf{S}(i) + \mathbf{V}(i)$$

- Channel matrix $\mathbf{H} \in \mathbb{C}^{N_R \times N_T}$: each element obeys $\mathcal{CN}(0, 1)$
- Noise matrix $\mathbf{V}(i) \in \mathbb{C}^{N_R \times T_n}$: each element obeys $\mathcal{CN}(0, N_0)$
- Signal to noise ratio (SNR) is defined as

$$\text{SNR} = E_s/N_0$$

E_s is average symbol energy of L -PSK/QAM modulation scheme

- Let $\text{vec}[\cdot]$ be vector stacking operator, \mathbf{I}_M be $M \times M$ identity matrix and \otimes be Kronecker product

Equivalent Signal Model

- Introduce notations

$$\begin{aligned}\bar{\mathbf{y}}(i) &= \text{vec}[\mathbf{Y}(i)] \in \mathbb{C}^{N_R T_n \times 1} & \bar{\mathbf{H}} &= \mathbf{I}_{T_n} \otimes \mathbf{H} \in \mathbb{C}^{N_R T_n \times N_T T_n} \\ \bar{\mathbf{v}}(i) &= \text{vec}[\mathbf{V}(i)] \in \mathbb{C}^{N_R T_n \times 1} & \Theta &= [\text{vec}[\mathbf{A}_1] \cdots \text{vec}[\mathbf{A}_Q]] \in \mathbb{C}^{N_T T_n \times Q}\end{aligned}$$

$$\mathbf{k}(i) = \underbrace{[0 \cdots 0]_{q-1}}_{q-1} s(i) \underbrace{[0 \cdots 0]_{Q-q}}_{Q-q}^T \in \mathbb{C}^{Q \times 1}$$

where q is index of dispersion matrix \mathbf{A}_q activated

- Equivalent transmitted signal vector $\mathbf{k}(i)$ takes value from set

$$\mathcal{K} = \{\mathbf{k}_{q,l} \in \mathbb{C}^{Q \times 1}, 1 \leq q \leq Q, 1 \leq l \leq L\}$$

which contains $Q \cdot L$ legitimate transmitted signal vectors

$$\mathbf{k}_{q,l} = \underbrace{[0 \cdots 0]_{q-1}}_{q-1} s_l \underbrace{[0 \cdots 0]_{Q-q}}_{Q-q}^T, 1 \leq q \leq Q, 1 \leq l \leq L$$

where s_l is the l th symbol in the L -point constellation \mathcal{S}

- Equivalent received signal model: $\bar{\mathbf{y}}(i) = \bar{\mathbf{H}} \Theta \mathbf{k}(i) + \bar{\mathbf{v}}(i)$

Maximum Likelihood Detection

- Free from ICI \Rightarrow low-complexity single-antenna ML detector, only searching $L \cdot Q$ points !
- Let (q, l) correspond to specific input bits of i th STSK block, which are mapped to s_l and \mathbf{A}_q
- Then ML estimates (\hat{q}, \hat{l}) are given by

$$(\hat{q}, \hat{l}) = \arg \min_{\substack{1 \leq q \leq Q \\ 1 \leq l \leq L}} \|\bar{\mathbf{y}}(i) - \bar{\mathbf{H}} \Theta \mathbf{k}_{q,l}\|^2 = \arg \min_{\substack{1 \leq q \leq Q \\ 1 \leq l \leq L}} \|\bar{\mathbf{y}}(i) - s_l (\bar{\mathbf{H}} \Theta)_q\|^2$$

where $(\bar{\mathbf{H}} \Theta)_q$ denotes q th column of the matrix $\bar{\mathbf{H}} \Theta$

- Assume channel's coherence time lasts the duration of τ STSK blocks. Then complexity of detecting $\tau \log_2(Q \cdot L)$ bits is

$$C_{\text{ML}} \approx 4QT_n N_R (3\tau L + 2N_T) \text{ [Flops]}$$

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Least Square Channel Estimate

- Assume number of available training blocks is M and training data are arranged as

$$\begin{aligned}\mathbf{Y}_{tM} &= [\mathbf{Y}(1) \mathbf{Y}(2) \cdots \mathbf{Y}(M)] \\ \mathbf{S}_{tM} &= [\mathbf{S}(1) \mathbf{S}(2) \cdots \mathbf{S}(M)]\end{aligned}$$

- Then LSCE based on $(\mathbf{Y}_{tM}, \mathbf{S}_{tM})$ is given by

$$\hat{\mathbf{H}}_{\text{LSCE}} = \mathbf{Y}_{tM} \mathbf{S}_{tM}^H (\mathbf{S}_{tM} \mathbf{S}_{tM}^H)^{-1}$$

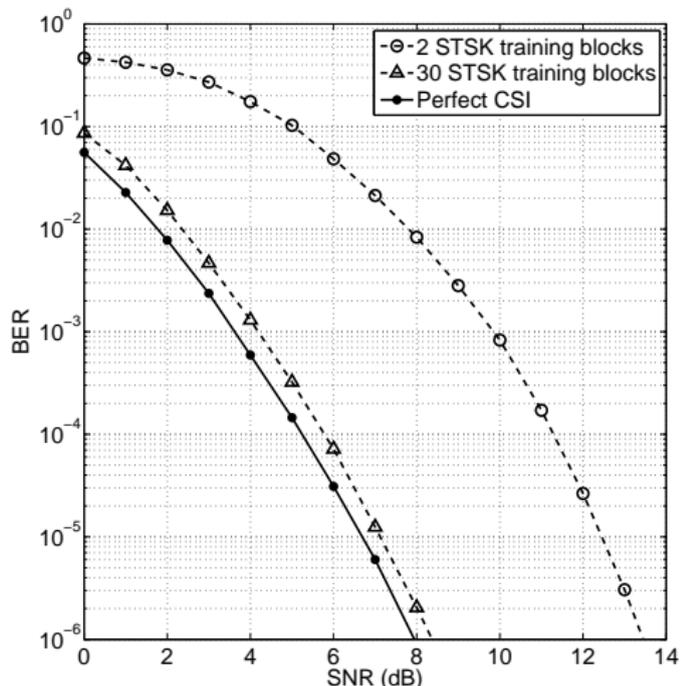
- In order for $\mathbf{S}_{tM} \mathbf{S}_{tM}^H$ to have full rank of N_T , it is necessary that $M \cdot T_n \geq N_T$ and this requires a minimum of

$$M = \left\lceil \frac{N_T}{T_n} \right\rceil \text{ training blocks}$$

- However, to achieve an accurate channel estimate, large training overhead is required

(4, 4, 2, 4) QPSK Example

- Convolution code with code rate 2/3, octally represented generator polynomials of $G_1 = [23, 35]_8$ and $G_2 = [5, 13]_8$
- Hard-input hard-output Viterbi algorithm decoding
- ($N_T = 4, N_R = 4, T_n = 2, Q = 4$) with $L = 4$ QPSK modulation
- Frame of 800 information source bits, after channel coding, are mapped to $\tau = 300$ STSK blocks
- Average over 100 channel realisations



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Semi-Blind Iterative algorithm

Use minimum $M = \left\lceil \frac{N_T}{T_n} \right\rceil$ training blocks to obtain initial $\hat{\mathbf{H}}_{\text{LSCE}}$, and let observation data for ML detector be $\mathbf{Y}_{d\tau} = [\mathbf{Y}(1) \mathbf{Y}(2) \cdots \mathbf{Y}(\tau)]$

- 1 Set iteration index $t = 0$ and channel estimate $\tilde{\mathbf{H}}^{(t)} = \hat{\mathbf{H}}_{\text{LSCE}}$;
- 2 Given $\tilde{\mathbf{H}}^{(t)}$, perform ML detection on $\mathbf{Y}_{d\tau}$ and carry out channel decoding on detected bits. Corresponding detected information bits, after passing through channel coder again, are re-modulated to yield

$$\hat{\mathbf{S}}_{e\tau}^{(t)} = [\hat{\mathbf{S}}^{(t)}(1) \hat{\mathbf{S}}^{(t)}(2) \cdots \hat{\mathbf{S}}^{(t)}(\tau)];$$

- 3 Update channel estimate with decision-directed LSCE

$$\tilde{\mathbf{H}}^{(t+1)} = \mathbf{Y}_{d\tau} (\hat{\mathbf{S}}_{e\tau}^{(t)})^H \left(\hat{\mathbf{S}}_{e\tau}^{(t)} (\hat{\mathbf{S}}_{e\tau}^{(t)})^H \right)^{-1};$$

- 4 Set $t = t + 1$: If $t < l_{\max}$, go to Step 2); otherwise, stop.

Simulation Settings

- Performance was assessed using estimated mean square error

$$J_{\text{MSE}}(\tilde{\mathbf{H}}) = \frac{1}{\tau \cdot N_R \cdot T_n} \sum_{i=1}^{\tau} \|\mathbf{Y}(i) - \tilde{\mathbf{H}}\hat{\mathbf{S}}(i)\|^2$$

mean channel estimation error

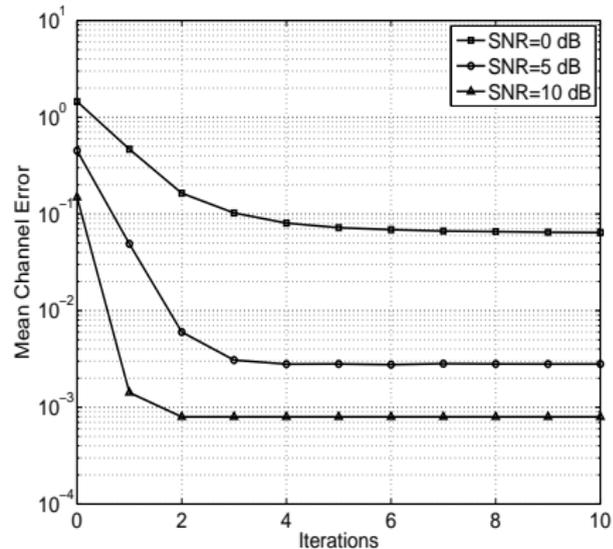
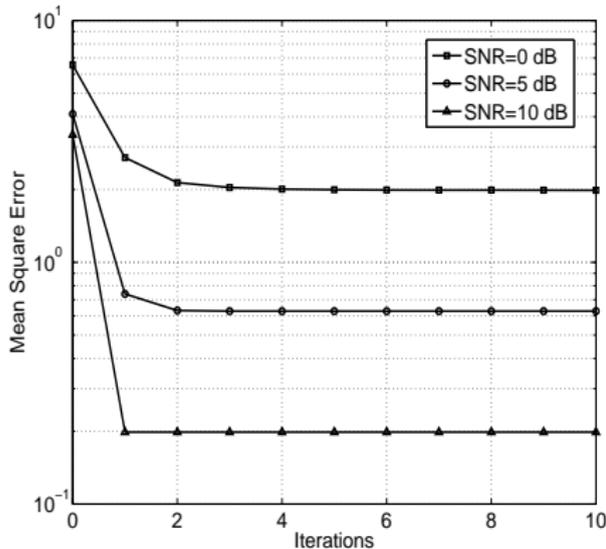
$$J_{\text{MCE}}(\tilde{\mathbf{H}}) = \frac{1}{N_R \cdot N_T} \|\mathbf{H} - \tilde{\mathbf{H}}\|^2$$

and BER, where $\tilde{\mathbf{H}}$ is channel estimate, $\hat{\mathbf{S}}(i)$ are ML-detected and re-modulated data, and \mathbf{H} is true MIMO channel matrix

- Performance averaged over 100 channel realisations
- Convolution code with code rate 2/3, octally represented generator polynomials of $G_1 = [23, 35]_8$ and $G_2 = [5, 13]_8$
- Hard-input hard-output Viterbi algorithm for channel decoding

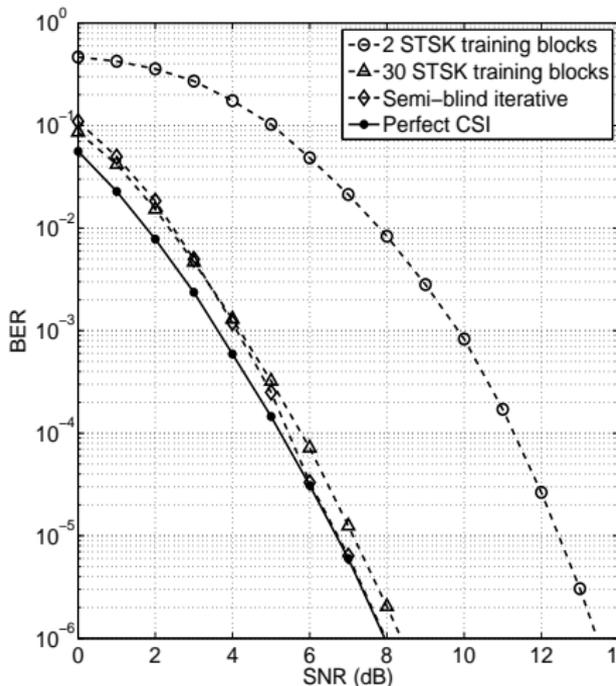
(4, 4, 2, 4) QPSK (Convergence)

- ($N_T = 4, N_R = 4, T_n = 2, Q = 4$) with $L = 4$ QPSK modulation
- Frame of 800 information source bits, after channel coding, are mapped to $\tau = 300$ STSK blocks
- Semi-blind with $M = 2$ training STSK blocks

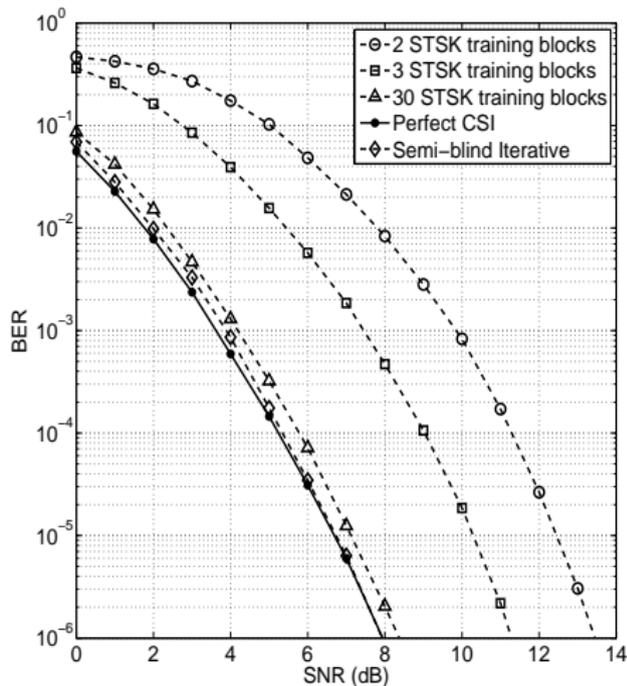


(4, 4, 2, 4) QPSK (Bit Error Rate)

(a) semi-blind with $M = 2$ training

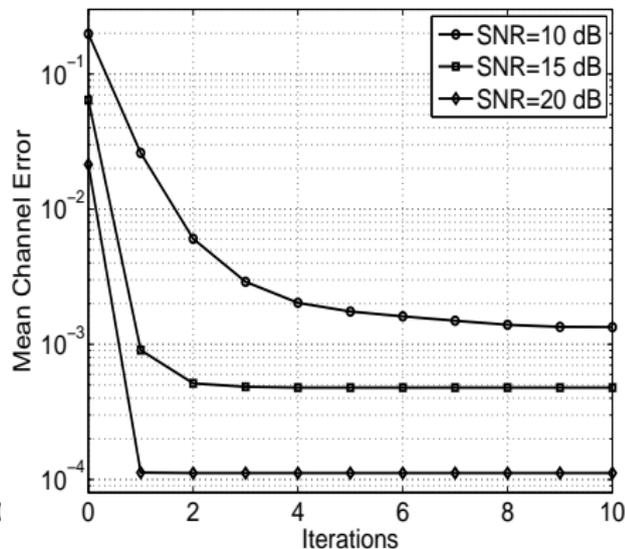
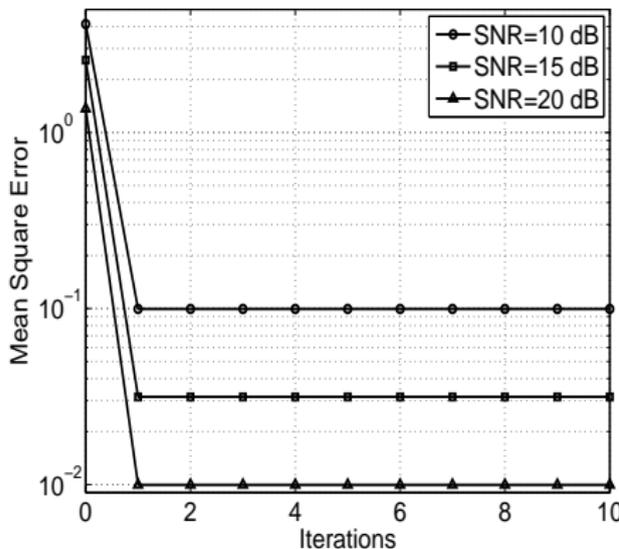


(b) semi-blind with $M = 3$ training



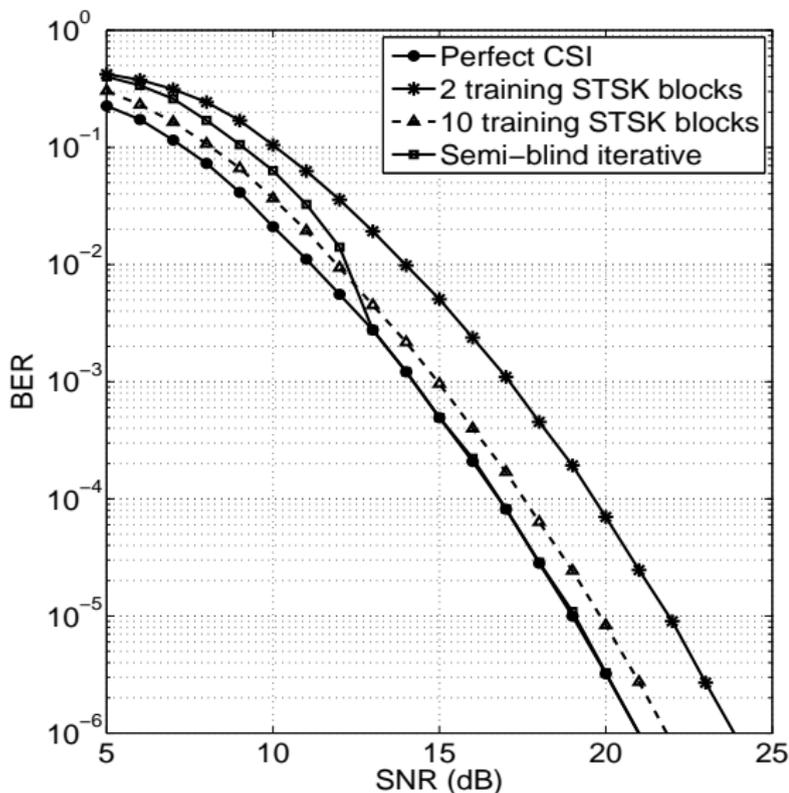
(4, 2, 2, 4) 16QAM (Convergence)

- ($N_T = 4, N_R = 2, T_n = 2, Q = 4$) with $L = 16$ QAM modulation
- Frame of 800 information source bits, after channel coding, are mapped to $\tau = 200$ STSK blocks
- Semi-blind with $M = 2$ training STSK blocks



$(4, 2, 2, 4)$ 16QAM (Bit Error Rate)

Semi-blind with $M = 2$ training



Summary

- A semi-blind iterative channel estimation and data detection scheme for coherent STSK systems
- Use minimum number of training STSK blocks to provide initial LSCE for aiding the iterative procedure
- Proposed semi-blind iterative channel estimation and ML data detection scheme is inherently low-complexity
- Typically no more than five iterations to converge to optimal ML detection performance obtained with perfect CSI

References

- 1 S. Sugiura, S. Chen and L. Hanzo, "A unified MIMO architecture subsuming space shift keying, OSTBC, BLAST and LDC," to be presented at *VTC 2010-Fall* (Ottawa, Canada), Sept. 6-9, 2010, 5 pages
- 2 S. Sugiura, S. Chen and L. Hanzo, "Space-time shift keying: A unified MIMO architecture," to be presented at *Globecom 2010* (Miami, Florida, USA), Dec.6-10, 2010, 5 pages
- 3 S. Sugiura, S. Chen and L. Hanzo, "Coherent and differential space-time shift keying: A dispersion matrix approach," accepted for publication, *IEEE Trans. Communications*, 2010, 12 pages