A Fundamental Tradeoff in Multiple Antenna Channels

Lizhong Zheng and David Tse Department of EECS, U.C. Berkeley

Feb 26, 2002

MSRI Information Theory Workshop

Wireless Fading Channels



Fundamental characteristic of wireless channels: multipath fading due to constructive and destructive interference.

Channel varies over time as well as frequency.

• Multi-antenna communication is a hot field in recent years.

- Multi-antenna communication is a hot field in recent years.
- But the research community has a split personality.

- Multi-antenna communication is a hot field in recent years.
- But the research community has a split personality.
- There are two very different views of how multiple antennas can be used.

-

_





• Additional independent signal paths increase diversity.

-



- Additional independent signal paths increase diversity.
- Diversity: receive



- Additional independent signal paths increase diversity.
- Diversity: receive, transmit



- Additional independent signal paths increase diversity.
- Diversity: receive, transmit or both.



- Additional independent signal paths increase diversity.
- Diversity: receive, transmit or both.
- Compensate against channel unreliability.

-





Another way to view a 2×2 system:

• Increases the degrees of freedom in the system



Another way to view a 2×2 system:

- Increases the degrees of freedom in the system
- Multiple antennas provide parallel spatial channels: spatial multiplexing



Another way to view a 2×2 system:

- Increases the degrees of freedom in the system
- Multiple antennas provide parallel spatial channels: spatial multiplexing
- Fading is exploited as a source of randomness.

Diversity vs. Multiplexing



Multiple antenna channel provides two types of gains:

Diversity Gain vs. Spatial Multiplexing Gain

Diversity vs. Multiplexing



Multiple antenna channel provides two types of gains:

Diversity Gain vs. Spatial Multiplexing Gain

Existing schemes focus on one type of gain.

A Different Point of View

Both types of gains can be achieved simultaneously in a given multiple antenna channel

A Different Point of View

Both types of gains can be achieved simultaneously in a given multiple antenna channel, but there is a fundamental tradeoff.

A Different Point of View

Both types of gains can be achieved simultaneously in a given multiple antenna channel, but there is a fundamental tradeoff.

We propose a unified framework which encompasses both diversity and multiplexing and study the optimal tradeoff.

Outline

- Problem formulation and main result on optimal tradeoff.
- Sketch of proof.
- Comparison of existing schemes.

Channel Model



 $\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w}, \qquad w_i \sim \mathcal{CN}(0, 1)$

- Rayleigh fading i.i.d. across antenna pairs $(h_{ij} \sim C\mathcal{N}(0,1))$.
- Focus on codes of T symbols, where H remains constant (slow, flat fading)
- **H** is known at the receiver but not the transmitter.
- SNR is the average signal-to-noise ratio at each receive antenna.

Motivation: Binary Detection

 $\mathbf{y} = \mathbf{h}\mathbf{x} + \mathbf{w}$ $P_e \approx P(\|\mathbf{h}\| \text{ is small }) \propto SNR^{-1}$

Motivation: Binary Detection

 $\mathbf{y} = \mathbf{h}\mathbf{x} + \mathbf{w}$ $P_e \approx P(\|\mathbf{h}\| \text{ is small }) \propto SNR^{-1}$

$$\begin{array}{l} \mathbf{y}_1 = \mathbf{h}_1 \mathbf{x} + \mathbf{w}_1 \\ \mathbf{y}_2 = \mathbf{h}_2 \mathbf{x} + \mathbf{w}_2 \end{array} \right\} \qquad P_e \quad \approx P(\|\mathbf{h}_1\|, \|\mathbf{h}_2\| \text{ are both small}) \\ \propto \mathsf{SNR}^{-2} \end{array}$$

Motivation: Binary Detection

 $\mathbf{y} = \mathbf{h}\mathbf{x} + \mathbf{w}$ $P_e \approx P(\|\mathbf{h}\| \text{ is small }) \propto SNR^{-1}$

$$\begin{array}{l} \mathbf{y}_1 = \mathbf{h}_1 \mathbf{x} + \mathbf{w}_1 \\ \mathbf{y}_2 = \mathbf{h}_2 \mathbf{x} + \mathbf{w}_2 \end{array} \right\} \qquad P_e \quad \approx P(\|\mathbf{h}_1\|, \|\mathbf{h}_2\| \text{ are both small}) \\ \propto \mathsf{SNR}^{-2} \end{array}$$

Definition A scheme achieves diversity gain d, if

 $P_e \sim \mathrm{SNR}^{-d}$

Motivation: Binary Detection

 $\mathbf{y} = \mathbf{h}\mathbf{x} + \mathbf{w}$ $P_e \approx P(\|\mathbf{h}\| \text{ is small }) \propto SNR^{-1}$

$$\begin{array}{l} \mathbf{y}_1 = \mathbf{h}_1 \mathbf{x} + \mathbf{w}_1 \\ \mathbf{y}_2 = \mathbf{h}_2 \mathbf{x} + \mathbf{w}_2 \end{array} \right\} \qquad P_e \quad \approx P(\|\mathbf{h}_1\|, \|\mathbf{h}_2\| \text{ are both small}) \\ \propto \mathsf{SNR}^{-2} \end{array}$$

Definition A scheme achieves diversity gain d, if

$$P_e \sim \mathrm{SNR}^{-d}$$

Acutal error probability instead of pairwise error probability. (eg. Tarokh et al 98, Guey et al 99)

How to Define Spatial Multiplexing Gain

Motivation: (Telatar'95, Foschini'96)

Ergodic capacity:

 $C(SNR) \approx \min\{M, N\} \log SNR \quad (bps/Hz),$

Equivalent to $\min\{M, N\}$ parallel spatial channels.

How to Define Spatial Multiplexing Gain

Motivation: (Telatar'95, Foschini'96)

Ergodic capacity:

 $C(SNR) \approx \min\{M, N\} \log SNR \quad (bps/Hz),$

Equivalent to $\min\{M, N\}$ parallel spatial channels.

A scheme is a sequence of codes, one at each SNR level.

How to Define Spatial Multiplexing Gain

Motivation: (Telatar'95, Foschini'96)

Ergodic capacity:

 $C(SNR) \approx \min\{M, N\} \log SNR \quad (bps/Hz),$

Equivalent to $\min\{M, N\}$ parallel spatial channels.

A scheme is a sequence of codes, one at each SNR level.

Definition A scheme achieves spatial multiplexing gain r, if

 $R = r \log \mathsf{SNR} \quad (bps/Hz)$

Increasing data rates instead of fixed data rate. (cf. Tarokh et al 98)

Fundamental Tradeoff

A scheme achieves

Fundamental Tradeoff

A scheme achieves

Fundamental tradeoff: for any r, the maximum diversity gain achievable: $d^*(r)$.

 $r \to d^*(r)$



As long as $T \ge M + N - 1$: (0,MN) (1,(M-1)(N-1))Diversity Gain: d^{*}(r) (min{M,N},0) Spatial Multiplexing Gain: r=R/log SNR

As long as $T \ge M + N - 1$: (0,MN) (1,(M-1)(N-1))Diversity Gain: d^{*}(r) (2, (M-2)(N-2)) $(min{M,N},0)$

Spatial Multiplexing Gain: r=R/log SNR

As long as $T \ge M + N - 1$: (0,MN) (1,(M-1)(N-1))Diversity Gain: d^{*}(r) (2, (M-2)(N-2)) (r, (M-r)(N-r)) (min{M,N},0) Spatial Multiplexing Gain: r=R/log SNR
Main Result: Optimal Tradeoff

As long as $T \ge M + N - 1$: (0,MN) (1,(M-1)(N-1)) d ^{*} (r) Diversity Gain: (2, (M-2)(N-2)) (r, (M-r)(N-r)) $(\min\{M,N\},0)$ Spatial Multiplexing Gain: r=R/log SNR

For integer r, it is as though r transmit and r receive antennas were dedicated for multiplexing and the rest provide diversity.

Main Result: Optimal Tradeoff

As long as $T \ge M + N - 1$:



For integer r, it is as though r transmit and r receive antennas were dedicated for multiplexing and the rest provide diversity.

Revisit the 2×2 Example



Revisit the 2×2 Example (ctd.)



Revisit the 2×2 Example (ctd.)



• Tradeoff bridges the gap between the two types of approaches.

Adding More Antennas



Adding More Antennas



• Capacity result : increasing min{*M*, *N*} by 1 adds 1 more degree of freedom.

Adding More Antennas



- Capacity result: increasing min{*M*, *N*} by 1 adds 1 more degree of freedom.
- Tradeoff curve: increasing both M and N by 1 yields multiplexing gain +1 for any diversity requirement d.

-



-

-



-

-



-





Outline

- Problem formulation and main result on optimal tradeoff.
- Sketch of proof.
- Comparison of existing schemes.

Converse: Outage Bound

- Outage formulation for quasi-static scenarios. (Ozarow et al 94, Telatar 95)
- Look at the mutual information per symbol $I(Q, \mathbf{H})$ as a function of the input distribution and channel realization.

Converse: Outage Bound

- Outage formulation for quasi-static scenarios. (Ozarow et al 94, Telatar 95)
- Look at the mutual information per symbol $I(Q, \mathbf{H})$ as a function of the input distribution and channel realization.
- Error probability for finite block length T is asymptotically lower bounded by the outage probability:

 $\inf_{Q} P_{\mathbf{H}} \left[I(Q, \mathbf{H}) < R \right].$

Converse: Outage Bound

- Outage formulation for quasi-static scenarios. (Ozarow et al 94, Telatar 95)
- Look at the mutual information per symbol $I(Q, \mathbf{H})$ as a function of the input distribution and channel realization.
- Error probability for finite block length T is asymptotically lower bounded by the outage probability:

$$\inf_{Q} P_{\mathbf{H}} \left[I(Q, \mathbf{H}) < R \right].$$

• At high SNR, i.i.d. Gaussian input Q^{\ast} is asymptotically optimal, and

 $I(Q^*, \mathbf{H}) = \log \det \left[I + \mathsf{SNRHH}^* \right].$

Outage Analysis

• Scalar 1×1 channel:

For target rate $R = r \log \text{SNR}$, r < 1,

$$P\{\log(1 + \mathsf{SNR} \|\mathbf{h}\|^2) < r \log \mathsf{SNR}\}$$

$$\sim P\left\{\|\mathbf{h}\|^2 < \mathsf{SNR}^{-(1-r)}\right\}$$

$$\sim \mathsf{SNR}^{-(1-r)}$$

$$\implies d_{out}(r) = 1 - r$$

• Outage occurs when the channel gain $\|\mathbf{h}\|^2$ is small.

Outage Analysis

• Scalar 1×1 channel:

For target rate $R = r \log SNR$, r < 1,

$$P\{\log(1 + \mathsf{SNR} \|\mathbf{h}\|^2) < r \log \mathsf{SNR}\}$$

$$\sim P\left\{\|\mathbf{h}\|^2 < \mathsf{SNR}^{-(1-r)}\right\}$$

$$\sim \mathsf{SNR}^{-(1-r)}$$

$$\implies d_{out}(r) = 1 - r$$

- Outage occurs when the channel gain $\|\mathbf{h}\|^2$ is small.
- More generally, outage occurs for the multi-antenna channel when some or all of the singular values of **H** are small.
- But unlike the scalar channel, there are many ways for this to happen in a vector channel.

Typical Outage Behavior

 $\mathbf{v} =$ vector of singular values of \mathbf{H} .

Laplace Principle:

 $p_{out} = \min_{\mathbf{v} \in \mathsf{Out}} \mathsf{SNR}^{-f(\mathbf{v})}$

Typical Outage Behavior

 $\mathbf{v} =$ vector of singular values of \mathbf{H} .

Laplace Principle:

$$p_{out} = \min_{\mathbf{v} \in \mathsf{Out}} \mathsf{SNR}^{-f(\mathbf{v})}$$

Result:

At target rate $R = r \log SNR$, outage typically occurs when **H** is near a rank $\lfloor r \rfloor$ matrix, i.e. out of the min $\{M, N\}$ non-zero squared singular values:

- $\lfloor r \rfloor$ of them are order 1;
- $\min\{M, N\} \lfloor r \rfloor + 1$ of them are are order SNR⁻¹;
- 1 of them is order SNR^{-(r-⌊r⌋)} (just small enough to cause outage)

When r is integer, exactly r squared singular values are order 1 and $\min\{M, N\}$ are order SNR⁻¹.

Scalar Channel



Scalar Channel









$$p_{out} \sim \mathsf{SNR}^{-(M-r)(N-r)},$$

(M-r)(N-r) is the dimension of the normal space to the sub-manifold of rank r matrices within the set of all $M \times N$ matrices.

Piecewise Linearity of Tradeoff Curve



Scalar channel: qualitatively same outage behavior for all r.

Vector channel: qualitatively different outage behavior for different r.

Achievability: Random Codes

- Outage performance achievable as codeword length $T \rightarrow \infty$.
- But what about for finite *T*?
- Look at the performance of i.i.d Gaussian random codes.
- Can the outage behavior be achieved?

Errors can occur due to three events:

Errors can occur due to three events:

• Channel **H** is aytically bad (outage)

Errors can occur due to three events:

- Channel **H** is aytically bad (outage)
- Additive Gaussian noise atypically large.

Errors can occur due to three events:

- Channel **H** is aytically bad (outage)
- Additive Gaussian noise atypically large.
- Random codewords are atypically close together.

Errors can occur due to three events:

- Channel **H** is aytically bad (outage)
- Additive Gaussian noise atypically large.
- Random codewords are atypically close together.

Outage analysis only needs to focus on the first event, but for finite T all three effects come into play.

Multiplicative Fading vs Additive Noise

Look at two codewords at Euclidean distance x.

Multiplicative Fading vs Additive Noise

Look at two codewords at Euclidean distance x.

• Error Event A: H typical, AWGN large

 $P(A) \sim \exp(-x).$

• Error Event B: H near singular, AWGN typical

 $P(B) \sim x^{-\alpha}.$

Multiplicative Fading vs Additive Noise

Look at two codewords at Euclidean distance x.

• Error Event A: H typical, AWGN large

 $P(A) \sim \exp(-x).$

• Error Event B: H near singular, AWGN typical

 $P(B) \sim x^{-\alpha}.$

At high SNR, $x \to \infty$

 $\implies P(B) \gg P(B).$
Multiplicative Fading vs Code Randomness

- Distance between random codewords may deviate from typical distance *x*.
- Error Event C: codewords atypically close

$$P(C) \sim x^{-\beta},$$

Also polynomial in x, just like the effect due to channel fading.

- As long as $T \ge M + N 1$, the typical error event is due to bad channel rather than bad codewords.
- For T < M + N 1, random codes are not good enough. (ISIT 02)

To Fade or Not to Fade?

Line-of-Sight vs Fading Channel



• In a scalar 1×1 system, line-of-sight AWGN is better.

Line-of-Sight vs Fading Channel



- In a scalar 1×1 system, line-of-sight AWGN is better.
- In a vector $M \times N$ system, it depends.

Outline

- Problem formulation and main result on optimal tradeoff.
- Sketch of proof.
- Comparison of existing schemes.

Using the Optimal Tradeoff Curve

Provide a unified framework to compare different schemes.

Use the Optimal Tradeoff Curve

Provide a unified framework to compare different schemes.

For a given scheme, compute

$$r \to d(r)$$

Compare with $d^*(r)$

Focus on two transmit antennas.

 $\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{W}$

Repetition Scheme:

$$\mathbf{X} = \begin{bmatrix} \mathbf{x}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{x}_1 \end{bmatrix}$$

Focus on two transmit antennas.

 $\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{W}$

Repetition Scheme:

$$\mathbf{X} = \begin{bmatrix} \mathbf{x}_1 & 0 \\ 0 & \mathbf{x}_1 \end{bmatrix} \qquad \qquad \mathbf{r} = \|\mathbf{H}\|\mathbf{x}_1 + \mathbf{w}\|$$

Focus on two transmit antennas.

$$\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{W}$$

Repetition Scheme:

$$\mathbf{X} = \left[\begin{array}{cc} \mathbf{x}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{x}_1 \end{array} \right]$$

 $\mathbf{r} = \|\mathbf{H}\|\mathbf{x}_1 + \mathbf{w}$

Alamouti Scheme

$$\mathbf{X} = \left[egin{array}{ccc} \mathbf{x}_1 & -\mathbf{x}_2^* \ \mathbf{x}_2 & \mathbf{x}_1^* \end{array}
ight]$$

Focus on two transmit antennas.

$$\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{W}$$

Repetition Scheme:

$$\mathbf{X} = \begin{bmatrix} \mathbf{x}_1 & 0 \\ 0 & \mathbf{x}_1 \end{bmatrix} \qquad \qquad \mathbf{r} = \|\mathbf{H}\|\mathbf{x}_1 + \mathbf{w}\|\mathbf{x}_1 + \mathbf{w}\|\mathbf{x}\|\|\mathbf{x}_1 + \mathbf{w}\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x}\|\|\mathbf{x$$

Alamouti Scheme

$$\mathbf{X} = \begin{bmatrix} \mathbf{x}_1 & -\mathbf{x}_2^* \\ \mathbf{x}_2 & \mathbf{x}_1^* \end{bmatrix} \qquad \qquad \begin{bmatrix} \mathbf{r}_1 \mathbf{r}_2 \end{bmatrix} = \|\mathbf{H}\| [\mathbf{x}_1 \mathbf{x}_2] + [\mathbf{w}_1 \mathbf{w}_2]$$

Comparison: 2×1 System



Comparison: 2×1 System



Comparison: 2×1 System



Comparison: 2×2 System



Comparison: 2×2 System



Comparison: 2×2 System









Antenna 1:

Antenna 2:



_

_





- Nulling and Canceling
- Independent data streams transmitted over antennas

Original V-BLAST



V-BLAST with optimal rate allocation



Compare to the optimal tradeoff



Compare to the optimal tradeoff



Low diversity due to lack of coding over space



_





_

Antenna 1:		\checkmark	
Antenna 2:	$\boldsymbol{\boldsymbol{<}}$		

 \mathbf{i}





Antenna 1:

_

_

Antenna 2:





Ignore the overhead for now.

D-BLAST: Square System



D-BLAST: Square System



- Can achieve full multiplexing gain
- Maximum diversity gain $d = \frac{N(N+1)}{2}$.

Replace Nulling by MMSE?

D-BLAST+MMSE



D-BLAST+MMSE



 Achieve the optimal: successive cancellation + MMSE has the optimal outage performance.

D-BLAST+MMSE



- Achieve the optimal: successive cancellation + MMSE has the optimal outage performance.
- Difference between MMSE and Nulling.

Penalty due to Overhead



Conclusion

- The diversity-multiplexing tradeoff is a fundamental way of looking at fading channels.
- Same framework can be applied to other scenarios: multiuser, non-coherent, more complex channel models, etc.