

On the Noisy MIMO Interference Channel with Analog Feedback

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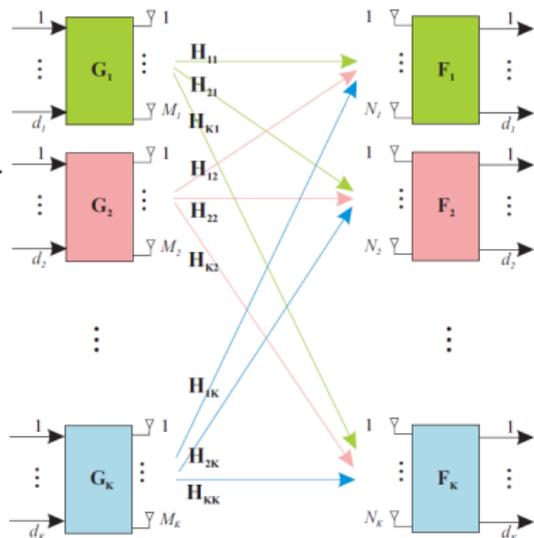
ITA, February 2012

- Introduction to the Noisy MIMO Interference Channel (IFC)
- Signal Space Interference Alignment (IA)
- Centralized CSIT Acquisition
- Distributed CSIT Acquisition
- Output Channel Feedback
- Concluding Remarks

MIMO IFC Introduction

- Interference Alignment (IA) was introduced in [Cadambe, Jafar 2008]
- The objective of IA is to design the Tx beamforming matrices such that the interference at each non intended receiver lies in a common interference subspace
- If alignment is complete at the receiver simple Zero Forcing (ZF) can suppress interference and extract the desired signal
- In [SPAWC2010] we derive a set of interference alignment (IA) feasibility conditions for a K -link frequency-flat MIMO interference channel (IFC)

- $d = \sum_{k=1}^K d_k$

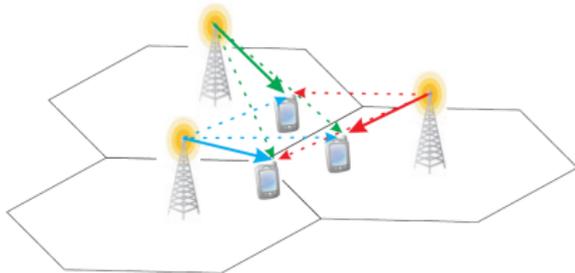


MIMO Interference Channel

Possible Application Scenarios

- Multi-cell cellular systems, modeling intercell interference.

Difference from Network MIMO: no exchange of signals, "only" of channel impulse responses.



- Coexistence of cellular and femto-cells, especially when femtocells are considered part of the cellular solution.



Why IA?

- The number of streams (degrees of freedom (dof)) appearing in a feasible IA scenario correspond to prelogs of feasible multi-user rate tuples in the multi-user rate region.
Max Weighted Sum Rate (WSR) becomes IA at high SNR.
- **Noisy** IFC: interfering signals are not decoded but treated as (Gaussian) noise.
Apparently enough for dof.
- Lots of recent work more generally on rate prelog regions: involves time sharing, use of fractional power.

Noisy MIMO IFC: Some State of the Art

- IA: alternating ZF algorithm [Jafar etal: globecom08],[Heath etal: icassp09].
- IA feasibility: - $K = 2$ MIMO: [JafarFakhereddin:IT07]
- [Yetis,Jafar:T10], [Slock etal:eusipco09,ita10, spawc10]
- $3 \times N \times N$, $3 \times M \times N$: [BreslerTse:arxiv11]
- max WSR: single stream/link
 - approximately: max SINR [Jafar etal: globecom08]
 - eigenvector interpretation of WSR gradient w.r.t. BF: starting [Honig,Utschick:asilo09]
 - added DA-style approach in [Honig,Utschick:allerton10]
- max WSR: multiple streams/link
 - [Slock etal:ita10] application of [Christensen etal:TW08] from MIMO BC
 - further refined in [Negro etal:allerton10], independently suggested use of DA, developed in [Negro etal:ita11]

IA as a Constrained Compressed SVD

- $F_k^H : d_k \times N_k$, $H_{ki} : N_k \times M_i$, $G_i : M_i \times d_i$ $F^H H G =$

$$\begin{bmatrix} F_1^H & 0 & \cdots & 0 \\ 0 & F_2^H & \cdots & \vdots \\ \vdots & \cdots & 0 \\ 0 & \cdots & 0 & F_K^H \end{bmatrix} \begin{bmatrix} H_{11} & H_{12} & \cdots & H_{1K} \\ H_{21} & H_{22} & \cdots & H_{2K} \\ \vdots & \cdots & \ddots & \vdots \\ H_{K1} & H_{K2} & \cdots & H_{KK} \end{bmatrix} \begin{bmatrix} G_1 & 0 & \cdots & 0 \\ 0 & G_2 & \cdots & \vdots \\ \vdots & \cdots & 0 \\ 0 & \cdots & 0 & G_K \end{bmatrix} = \begin{bmatrix} F_1^H H_{11} G_1 & 0 & \cdots & 0 \\ 0 & F_2^H H_{22} G_2 & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \cdots & 0 & F_K^H H_{KK} G_K \end{bmatrix}$$

F^H , G can be chosen to be unitary for IA

- per user vs per stream approaches:

IA: can absorb the $d_k \times d_k$ $F_k^H H_{kk} G_k$ in either F_k^H (per stream LMMSE Rx) or G_k or both.

WSR: can absorb unitary factors of SVD of $F_k^H H_{kk} G_k$ in F_k^H , G_k without loss in rate $\Rightarrow F^H H G = \text{diagonal}$.

Interference Alignment: Feasibility Conditions (1)

- To derive the existence conditions we consider the ZF conditions

$$\underbrace{\mathbf{F}_k^H}_{d_k \times N_k} \underbrace{\mathbf{H}_{kl}}_{N_k \times M_l} \underbrace{\mathbf{G}_l}_{M_l \times d_l} = \mathbf{0}, \quad \forall l \neq k$$

$$\text{rank}(\mathbf{F}_k^H \mathbf{H}_{kk} \mathbf{G}_k) = d_k, \quad \forall k \in \{1, 2, \dots, K\}$$

- rank requirement \Rightarrow SU MIMO condition: $d_k \leq \min(M_k, N_k)$
- The total number of variables in \mathbf{G}_k is $d_k M_k - d_k^2 = d_k(M_k - d_k)$
Only the subspace of \mathbf{G}_k counts, it is determined up to a $d_k \times d_k$ mixture matrix.
- The total number of variables in \mathbf{F}_k^H is $d_k N_k - d_k^2 = d_k(N_k - d_k)$
Only the subspace of \mathbf{F}_k^H counts, it is determined up to a $d_k \times d_k$ mixture matrix.

Interference Alignment: Feasibility Conditions (2)

- A solution for the interference alignment problem can only exist if the **total number of variables is greater than or equal to the total number of constraints** i.e.,

$$\begin{aligned}\sum_{k=1}^K d_k(M_k - d_k) + \sum_{k=1}^K d_k(N_k - d_k) &\geq \sum_{i \neq j=1}^K d_i d_j \\ \Rightarrow \sum_{k=1}^K d_k(M_k + N_k - 2d_k) &\geq (\sum_{k=1}^K d_k)^2 - \sum_{k=1}^K d_k^2 \\ \Rightarrow \sum_{k=1}^K d_k(M_k + N_k) &\geq (\sum_{k=1}^K d_k)^2 + \sum_{k=1}^K d_k^2\end{aligned}$$

- In the symmetric case: $d_k = d$, $M_k = M$, $N_k = N$:

$$d \leq \frac{M+N}{K+1}$$

- For the $K = 3$ user case ($M = N$): $d = \frac{M}{2}$.

With 3 parallel MIMO links, half of the (interference-free) resources are available!

However $d \leq \frac{1}{(K+1)/2} M < \frac{1}{2} M$ for $K > 3$.

MWSR: Maximum Weighted Sum Rate (WSR)

The received signal at the k -th receiver is:

$$\mathbf{y}_k = \mathbf{H}_{kk} \mathbf{G}_k \mathbf{x}_k + \sum_{\substack{l=1 \\ l \neq k}}^K \mathbf{H}_{kl} \mathbf{G}_l \mathbf{x}_l + \mathbf{n}_k$$

Introduce the interference plus noise covariance matrix at receiver

$$k: \mathbf{R}_{\bar{k}} = \mathbf{R}_{nn} + \sum_{l \neq k} \mathbf{H}_{kl} \mathbf{G}_l \mathbf{G}_l^H \mathbf{H}_{kl}^H.$$

The WSR criterion is

$$\mathcal{R} = \sum_{k=1}^K u_k \log \det(\mathbf{I} + \mathbf{G}_k^H \mathbf{H}_{kk}^H \mathbf{R}_{\bar{k}}^{-1} \mathbf{H}_{kk} \mathbf{G}_k) \quad (1)$$

$$\text{s.t. } \text{Tr}\{\mathbf{G}_k^H \mathbf{G}_k\} \leq P_k$$

This criterion is highly non convex in the Tx BFs \mathbf{G}_k .

- Augmented WSR cost function: BFs \mathbf{g}_{kn} plus Rx filters \mathbf{f}_{kn} and weights w_{kn} :

$$\begin{aligned} \mathcal{O} = & - \sum_{k=1}^K u_k \sum_{n=1}^{d_k} (-\ln(w_{kn}) + w_{kn}(1 - \mathbf{f}_{kn}^H \mathbf{H}_{kk} \mathbf{g}_{kn})(1 - \mathbf{f}_{kn}^H \mathbf{H}_{kk} \mathbf{g}_{kn})^H \\ & + \underbrace{\mathbf{f}_{kn}^H (\mathbf{R}_{v_k} + \sum_{(im) \neq (kn)} \mathbf{H}_{ki} \mathbf{g}_{im} \mathbf{g}_{im}^H \mathbf{H}_{ki}^H)}_{\mathbf{R}_{\bar{kn}}} \mathbf{f}_{kn}) + \sum_{k=1}^K \lambda (P_k - \sum_{n=1}^{d_k} \mathbf{g}_{kn}^H \mathbf{g}_{kn}) \end{aligned} \quad (2)$$

Alternating optimization \Rightarrow quadratic or convex subproblems:

- Opt w_{kn} given $\mathbf{g}_{kn}, \mathbf{f}_{kn}$: $-\ln(w_{kn}) + w_{kn}e_{kn} \Rightarrow w_{kn} = e_{kn}^{-1}$
- Opt \mathbf{f}_{kn} given w_{kn}, \mathbf{g}_{kn} : MMSE solution
- Opt \mathbf{f}_{kn} given w_{kn}, \mathbf{g}_{kn} : MMSE-style solution (UL-DL duality)

- MISO BC (MU-MISO DL) w CSIT acquisition:
[KobayashiCaireJindal:IT10]
- TDD MISO BC w CSIT acquisition: [SalimSlock:JWCN11]
- Space-Time Coding for Analog Channel Feedback:
[ChenSlock:isit08]
- [NegroShenoySlockGhauri: eusipco09]: TDD MIMO IFC IA iterative design via UL/DL duality and TDD reciprocity
- Interference Alignment with Analog CSI Feedback:
[ElAyachHeath:Milcom10]
Centralized approach: BS's are connected to a central unit gathering all CSI, performing BF computations and redistributing BF's.
- [Jafar:GLOBECOM10] Blind IA
- [VazeVaranasi:submit] DoF region for MIMO IFC with feedback
- [SuhTse:IT11] GDoF for IFC with feedback

- Distributed approach: no other connectivity assumed than the UL/DL IFC. **FB over reversed IFC**
- "distributed" = "duplicated" (decentralized)
- A distributed approach does not have to be iterative. It can be done with a finite overhead (finite prelog loss) and finite SNR loss compared to full CSI, even as $\text{SNR} \rightarrow \infty$. Hence **of interest compared to non-coherent (no/outdated CSIT) IFC approaches.**
- Distributed ($\mathcal{O}(K^2)$) requires more FB than Centralized ($\mathcal{O}(K)$).
- centralized/decentralized IFC CSIT estimation (only exchange of data at temporal coherence variation rate), vs NW-MIMO/CoMP (exchange of data at symbol/sample rate)
- Multiple Rx antennas \Rightarrow Rx training also crucial!
- TDD vs FDD, depends on distributed/centralized.
- Channel FB vs Output feedback (OFB)
- "Practical" scheme far from unique

- Perfect CSI:

Rx signal at the k -th receiver :

$$\mathbf{y}_k = \sum_{i=1}^K \sum_{m=1}^{d_i} \mathbf{H}_{ki} \mathbf{g}_{i,m} x_{i,m} + \mathbf{v}_k$$

Estimate stream (k, n) :

$$\hat{x}_{k,n} = \mathbf{f}_{k,n} \mathbf{H}_{kk} \mathbf{g}_{k,n} x_{k,n} + \sum_{i=1}^K \sum_{m \neq n} \mathbf{f}_{k,n} \mathbf{H}_{ki} \mathbf{g}_{i,m} x_{i,m} + \mathbf{f}_{k,n} \mathbf{v}_k$$

- Imperfect CSI:

$$\underbrace{\hat{\mathbf{f}}_{k,n}}_{\text{est. at Rx } k} \quad \underbrace{\mathbf{H}_{ki}}_{\text{true}} \quad \underbrace{\hat{\mathbf{g}}_{i,m}}_{\text{est. at Tx } i}$$

- signal of interest in direct link:

$$\widehat{\mathbf{f}}_{k,n} \mathbf{H}_{kk} \widehat{\mathbf{g}}_{k,n} = \underbrace{\widehat{\mathbf{f}}_{k,n} \mathbf{H}_{kk} \widehat{\mathbf{g}}_{k,n}}_{\text{known to Rx}} + \underbrace{\widehat{\mathbf{f}}_{k,n} \mathbf{H}_{kk} \widehat{\mathbf{g}}_{k,n}}_{\text{put in interf.}}$$

3 Partial CSI Rate Analysis Approaches (1)

- 1 Bound loss of partial CSI ergodic rate to full CSI ergodic rate.

$$\text{e.g. } \mathcal{R}_k^{PCSI}(\rho) \leq \left(1 - \frac{T_{\text{overhead}}}{T}\right) \mathcal{R}_k^{FCSI}(\rho/\alpha_k)$$

-

$$\begin{aligned} \widehat{\mathbf{f}}_{k,n} \mathbf{H}_{ki} \widehat{\mathbf{g}}_{i,m} &= (\mathbf{f}_{k,n} + \widetilde{\mathbf{f}}_{k,n}) \mathbf{H}_{ki} (\mathbf{g}_{i,m} + \widetilde{\mathbf{g}}_{i,m}) \\ &= \mathbf{f}_{k,n} \mathbf{H}_{k,i} \mathbf{g}_{i,m} + 3 \text{ error terms} \end{aligned}$$

3 Partial CSI Rate Analysis Approaches (2)

- 2 Bound loss of partial CSI ergodic rate to full CSI ergodic rate for case of channel pdf = that of the estimated channel: provides closer bounds, but requires ergodic rate expressions with different channel statistics.

-

$$\begin{aligned}\widehat{\mathbf{f}}_{k,n} \mathbf{H}_{ki} \widehat{\mathbf{g}}_{i,m} &= (\widehat{\mathbf{f}}_{k,n}^{(i)} + \widetilde{\mathbf{f}}_{k,n}^{(i)}) (\widehat{\mathbf{H}}_{ki}^{(i)} + \widetilde{\mathbf{H}}_{ki}^{(i)}) \widehat{\mathbf{g}}_{i,m} \\ &= \widehat{\mathbf{f}}_{k,n}^{(i)} \widehat{\mathbf{H}}_{ki}^{(i)} \widehat{\mathbf{g}}_{i,m} + 3 \text{ error terms}\end{aligned}$$

3 Partial CSI Rate Analysis Approaches (3)

- ③ High SNR ρ rate asymptote: $\mathcal{R} = a \log(\rho) + b + \mathcal{O}(1/\rho)$
 a : multiplexing gain (prelog, dof), b : rate offset a, b
independent of:
- MMSE regularization (MMSE-ZF filters suffice)
 - optimized WF (uniform WF suffices)
 - LMMSE channel estimation (becomes deterministic estimation)
 -

$$\begin{aligned}\widehat{\mathbf{f}}_{k,n} \mathbf{H}_{ki} \widehat{\mathbf{g}}_{i,m} &= (\widehat{\mathbf{f}}_{k,n}^{(i)} + \widetilde{\mathbf{f}}_{k,n}^{(i)}) (\widehat{\mathbf{H}}_{ki}^{(i)} + \widetilde{\mathbf{H}}_{ki}^{(i)}) \widehat{\mathbf{g}}_{i,m} \\ &= \underbrace{\widehat{\mathbf{f}}_{k,n}^{(i)} \widehat{\mathbf{H}}_{ki}^{(i)} \widehat{\mathbf{g}}_{i,m}}_{= 0} + \widehat{\mathbf{f}}_{k,n} \widetilde{\mathbf{H}}_{ki}^{(i)} \widehat{\mathbf{g}}_{i,m} + \widetilde{\mathbf{f}}_{k,n}^{(i)} \mathbf{H}_{ki} \mathbf{g}_{i,m}\end{aligned}$$

High SNR Rate Analysis

- Asymptote $\mathcal{R} = a \log(\rho) + b$ permits meaningful optimization for finite (but high) SNR, and may lead to more than minimal FB.
- At very high SNR ρ , only rate prelog a (dof) counts. Its maximization requires FB to be minimal (channel just identifiable).
- At moderate SNR, finding an optimal compromise between estimation overhead and channel quality will involve a properly adjusted overhead. However, the overhead issue is not the only reason for a possibly diminishing multiplexing gain a as SNR decreases, also reducing the number of streams $\{d_k\}$ may lead to a better compromise (as for full CSI).
- The rate offset b is already a non-trivial rate characteristic even in the full CSI case. b may increase as the number of streams decreases, due to reduced noise enhancement.

Unification Stationary & Block Fading

- Doppler Spectrum is bandlimited to $1/T$ ($1/D$ in figure)
- Nyquist's Theorem : downsampling possible with factor T
- Vectorize channel coefficients over T , matrix spectrum of rank 1, MIMO prediction error of rank 1.
- Hence channel coefficient evolution during current "coherence period" T is along a single basis vector, plus prediction from past.
- Block fading: basis vector = rectangular window and prediction from the past = 0

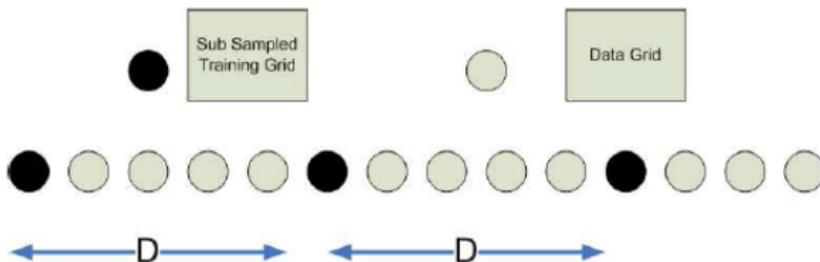
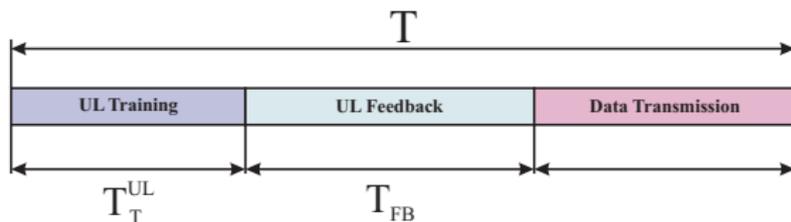


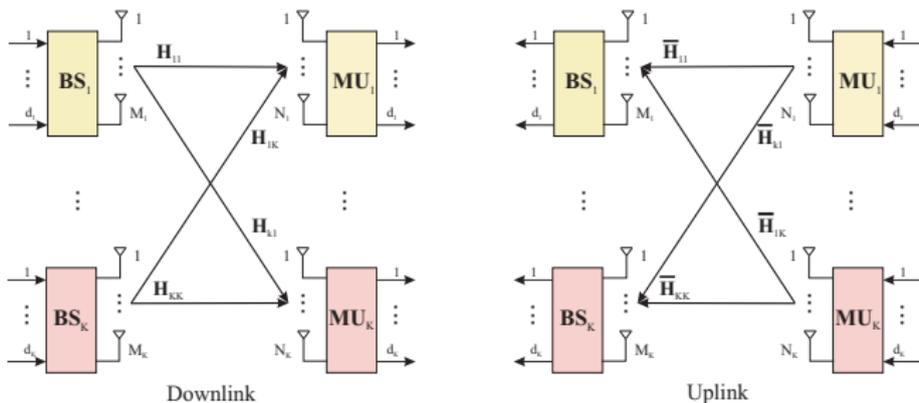
Figure 1: Subsampling Grid.

Centralized Approach



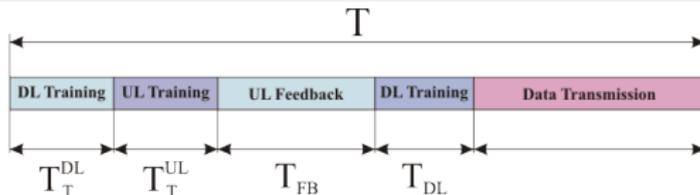
- Proposed by Heath [Milcom10,arxiv]
- The authors extrapolate the single antenna case, where only the estimate of the overall ch-BF gain and associated SINR is required
- In the MIMO IFC Rx not only needs to estimate the ch-BF cascade but also the I+N covariance matrix
- Not trivial. Training length similar as for the BF determination (order K) is required.
- Rate analysis of type 1.

FDD Communication

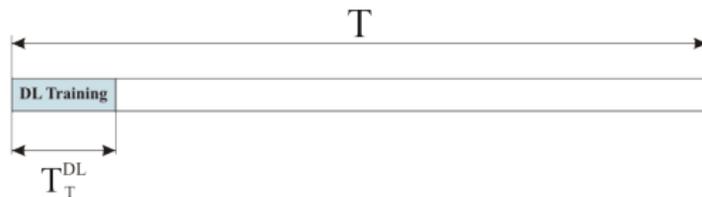


- We Assume FDD transmission scheme
- Downlink channel matrix \mathbf{H}_{ki} from BS_i to MU_k
- Uplink channel matrix $\bar{\mathbf{H}}_{ik}$ from MU_k to BS_i
- Analyze both centralized and distributed approaches.

Transmission Phases



- We consider a block fading channel model with Coherence time interval T
- The general channel matrix $\mathbf{H}_{ik} \sim \mathcal{N}(0, \mathbf{I})$
- To acquire the necessary CSI at BS and MU side several training and feedback phases are necessary
- Hence a total overhead of T_{ovrhd} channel usage is dedicated to BS-MU signaling
- Only part of the time $T_{data} = T - T_{ovrhd}$ is dedicated to real data transmission



Downlink Training Phase

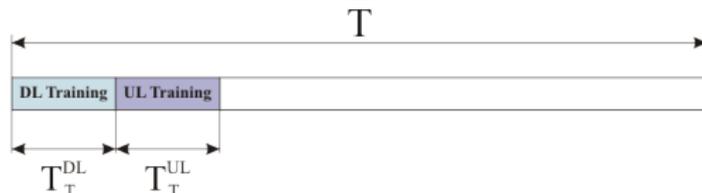
- Each BS_i Tx (\perp) pilot sequences with power P_T^{DL}
- MU_k estimates all DL channels connected to it:
 $\mathbf{H}_k = [\mathbf{H}_{k1}, \dots, \mathbf{H}_{kK}]$
- The duration of the DL training phase is

$$T_T^{DL} \geq \sum_{k=1}^K M_k$$

- Using MMSE estimation we get $\mathbf{H}_k = \hat{\mathbf{H}}_k + \tilde{\mathbf{H}}_k$

$$\hat{\mathbf{H}}_k \sim \mathcal{N}\left(0, \frac{P_T^{DL}}{\sigma^2 + P_T^{DL}} \mathbf{I}\right), \quad \tilde{\mathbf{H}}_k \sim \mathcal{N}\left(0, \frac{\sigma^2}{\sigma^2 + P_T^{DL}} \mathbf{I}\right)$$

we call $\sigma_{\tilde{H}}^2$ and $\sigma_{\hat{H}}^2$ the variance of the error and estimate



Uplink Training Phase (dual of DL Training Phase)

- Each MU_i sends a set of pilots symbols with power P_T^{UL}
- BS_k estimate the compound UL channel matrix:
 $\bar{\mathbf{H}}_k = [\bar{\mathbf{H}}_{k1}, \dots, \bar{\mathbf{H}}_{kK}]$
- The duration of the UL training phase is

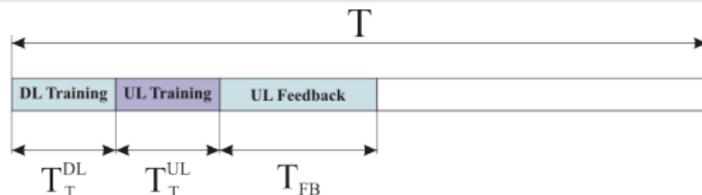
$$T_T^{UL} \geq \sum_{k=1}^K N_k$$

- Using MMSE estimation we get $\bar{\mathbf{H}}_k = \hat{\mathbf{H}}_k + \tilde{\mathbf{H}}_k$

$$\hat{\mathbf{H}}_k \sim \mathcal{N}\left(0, \frac{P_T^{UL}}{\sigma^2 + P_T^{UL}} \mathbf{I}\right), \quad \tilde{\mathbf{H}}_k \sim \mathcal{N}\left(0, \frac{\sigma^2}{\sigma^2 + P_T^{UL}} \mathbf{I}\right)$$

we call $\sigma_{\tilde{H}}^2$ and $\sigma_{\hat{H}}^2$ the variance of the error and estimate

Uplink Feedback Phase



- After the UL and DL training phases each device knows all channels directly connected to it
 - To compute the Tx beamformers, complete IFC channel knowledge is required
 - Each MU feeds back its channel knowledge (CFB) using *Analog Feedback*
 - Two different approaches are possible:
 - (a) Centralized Processing
 - (b) Distributed Computation
- (a) A Central Controller acquires complete CSI and computes all the BF, and disseminates this information.
- (b) Each BS acquires complete CSI to compute all the BF, then uses only its own BF.

Uplink Feedback Phase: Centralized Processing

- The received symbol vector received at each BS is sent to the Central Controller for the estimation of DL channels. Staking all the received symbols together we get:

$$\bar{\mathbf{Y}} = \sqrt{P_{FB}} \underbrace{\begin{bmatrix} \bar{\mathbf{H}}_{11} & \dots & \bar{\mathbf{H}}_{1K} \\ \vdots & \ddots & \vdots \\ \bar{\mathbf{H}}_{K1} & \dots & \bar{\mathbf{H}}_{KK} \end{bmatrix}}_{M \times N} \underbrace{\begin{bmatrix} \hat{\mathbf{H}}_1 & \mathbf{0} & \dots & \mathbf{0} \\ \mathbf{0} & \hat{\mathbf{H}}_2 & \dots & \mathbf{0} \\ \vdots & \vdots & \ddots & \mathbf{0} \\ \mathbf{0} & \dots & \dots & \hat{\mathbf{H}}_K \end{bmatrix}}_{N \times KM} \underbrace{\begin{bmatrix} \Phi_1 \\ \vdots \\ \Phi_K \end{bmatrix}}_{KM \times T_{FB}} + \underbrace{\begin{bmatrix} \mathbf{V}_1 \\ \vdots \\ \mathbf{V}_K \end{bmatrix}}_{\bar{\mathbf{V}}}$$

where $N = \sum_i N_i$ and $M = \sum_i M_i$

- To satisfy the identifiability condition the minimum CFB length is

$$T_{FB} \geq \frac{N \times M}{\sum_i \min\{N_i, M_i\}} \propto K$$

- To extract the i -th feedback contribution we use LS estimate based on the UL channel estimate $\hat{\mathbf{H}}_{ik}$

$$\bar{\mathbf{Y}}\boldsymbol{\Phi}_i = \sqrt{P_{FB}} \underbrace{\begin{bmatrix} \bar{\mathbf{H}}_{i1} \\ \vdots \\ \bar{\mathbf{H}}_{iK} \end{bmatrix}}_{\bar{\mathbf{H}}_i} \hat{\mathbf{H}}_i + \bar{\mathbf{V}}\boldsymbol{\Phi}_i$$

- Using the UL channel estimate the LS estimator is: $\bar{\mathbf{H}}_i^{LS} = P_{FB}^{-\frac{1}{2}} (\hat{\mathbf{H}}_i^H \hat{\mathbf{H}}_i)^{-1} \hat{\mathbf{H}}_i^H$

$$\hat{\hat{\mathbf{H}}}_i = \hat{\mathbf{H}}_i + P_{FB}^{\frac{1}{2}} \bar{\mathbf{H}}_i^{LS} \tilde{\tilde{\mathbf{H}}}_i \hat{\mathbf{H}}_i + \bar{\mathbf{H}}_i^{LS} \bar{\mathbf{V}}\boldsymbol{\Phi}_i = \mathbf{H}_i - \underbrace{\tilde{\tilde{\mathbf{H}}}_i + P_{FB}^{\frac{1}{2}} \bar{\mathbf{H}}_i^{LS} \tilde{\tilde{\mathbf{H}}}_i \hat{\mathbf{H}}_i + \bar{\mathbf{H}}_i^{LS} \bar{\mathbf{V}}\boldsymbol{\Phi}_i}_{\tilde{\tilde{\mathbf{H}}}_i}$$

- The estimate of the CFB can be written in function of the true DL channel \mathbf{H}_i plus the estimation error $\tilde{\tilde{\mathbf{H}}}_i$

$$\text{Cov}(\tilde{\tilde{\mathbf{H}}}_i | \hat{\mathbf{H}}_i) = \sigma_{\tilde{\tilde{\mathbf{H}}}_i}^2 \mathbf{I} + [(\sigma_{\tilde{\tilde{\mathbf{H}}}_i}^2 \sigma_{\tilde{\tilde{\mathbf{H}}}_i}^2) + \frac{\sigma^2}{P_{FB}}] (\hat{\mathbf{H}}_i^H \hat{\mathbf{H}}_i)^{-1}$$

- The estimation error is then distributed as $\mathcal{N}(0, \sigma_{\tilde{\tilde{\mathbf{H}}}_i}^2)$

Uplink Feedback Phase: Distributed Processing

- The received symbols at BS_k can be described as follows

$$\bar{\mathbf{Y}}_k = \sqrt{P_{FB}} \underbrace{\begin{bmatrix} \bar{\mathbf{H}}_{k1} & \dots & \bar{\mathbf{H}}_{kK} \end{bmatrix}}_{M_k \times N} \underbrace{\begin{bmatrix} \hat{\mathbf{H}}_1 & \mathbf{0} & \dots & \mathbf{0} \\ \mathbf{0} & \hat{\mathbf{H}}_2 & \dots & \mathbf{0} \\ \vdots & & \ddots & \mathbf{0} \\ \mathbf{0} & \dots & \dots & \hat{\mathbf{H}}_K \end{bmatrix}}_{N \times KM} \underbrace{\begin{bmatrix} \Phi_1 \\ \vdots \\ \Phi_K \end{bmatrix}}_{KM \times T_{FB}} + \mathbf{V}_k$$

- To satisfy the identifiability condition the minimum CFB length is

$$T_{FB} \geq \frac{N \times M}{\min_i \{M_i, N_i\}} \propto K^2$$

- To extract the i -th feedback contribution at BS_k we use LS estimate based on the UL channel estimate $\hat{\mathbf{H}}_{ki}$

$$\bar{\mathbf{Y}}_k \boldsymbol{\Phi}_i = \sqrt{P_{FB}} \bar{\mathbf{H}}_{ki} \hat{\mathbf{H}}_i + \mathbf{V}_k \boldsymbol{\Phi}_i$$

- Using the UL channel estimate the LS estimator is: $\bar{\mathbf{H}}_{ki}^{LS} = P_{FB}^{-\frac{1}{2}} (\hat{\mathbf{H}}_{ki}^H \hat{\mathbf{H}}_{ki})^{-1} \hat{\mathbf{H}}_{ki}^H$

$$\hat{\hat{\mathbf{H}}}_i = \hat{\mathbf{H}}_i + P_{FB}^{\frac{1}{2}} \bar{\mathbf{H}}_{ki}^{LS} \tilde{\tilde{\mathbf{H}}}_{ki} \hat{\mathbf{H}}_i + \bar{\mathbf{H}}_{ki}^{LS} \mathbf{V}_k \boldsymbol{\Phi}_i = \mathbf{H}_i - \underbrace{\tilde{\mathbf{H}}_i + P_{FB}^{\frac{1}{2}} \bar{\mathbf{H}}_{ki}^{LS} \tilde{\tilde{\mathbf{H}}}_{ki} \hat{\mathbf{H}}_i + \bar{\mathbf{H}}_{ki}^{LS} \mathbf{V}_k \boldsymbol{\Phi}_i}_{\tilde{\hat{\mathbf{H}}}_i}$$

- The estimate of the CFB can be written in function of the true DL channel \mathbf{H}_i plus the estimation error $\tilde{\hat{\mathbf{H}}}_i$

$$\text{Cov}(\tilde{\hat{\mathbf{H}}}_i | \hat{\mathbf{H}}_{ki}) = \sigma_{\tilde{\mathbf{H}}_i}^2 \mathbf{I} + [(\sigma_{\tilde{\mathbf{H}}_i}^2 \sigma_{\tilde{\tilde{\mathbf{H}}}_{ki}}^2) + \frac{\sigma^2}{P_{FB}}] (\hat{\mathbf{H}}_{ki}^H \hat{\mathbf{H}}_{ki})^{-1}$$

- The estimation error is then distributed as $\mathcal{N}(0, \sigma_{\tilde{\hat{\mathbf{H}}}_i}^2)$

- Simplest precoding: time multiplexing and repetition coding. To allow finer granularity of FB overhead: uses constant amplitude unitary spreading matrices (eg DFT).
- (linear) Rx strategies for analog FB: many variants possible
 - MMSE (Bayesian \mathbf{H}), MMSE-ZF (deterministic \mathbf{H})
 - assuming $\hat{\mathbf{H}}$ correct, accounting for $\tilde{\mathbf{H}}$
- analog STC: more spatial multiplexing leads to less overhead but less noise enhancement (especially crucial in distributed approach)

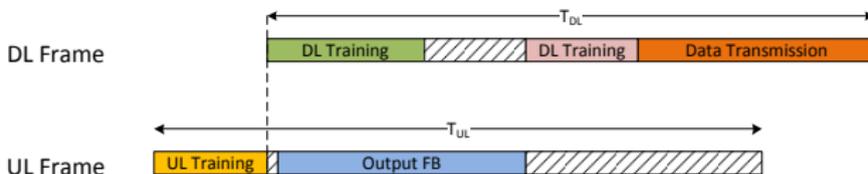
- Once DL channel estimates available, need to perform BF design, e.g. according to Maximum Weighted Sum Rate.
- Can ignore channel estimation errors (full CSIT type design) or acknowledge them (partial CSIT type design).
- Similar considerations for centralized or distributed approaches.

- Model the information of the channel at the transmit side in terms of a Gaussian prior representing mean and covariance information

$$\mathbf{H}_{ij} = \hat{\mathbf{H}}_{ij} + (\mathbf{R}_{ij}^t)^{\frac{1}{2}} \tilde{\mathbf{H}}_{ij} (\mathbf{R}_{ij}^r)^{\frac{H}{2}} \quad (3)$$

- $\mathbf{R}_{ij}^t = \mathbf{I}$ is the Tx side covariance matrix, $\mathbf{R}_{ij}^r = \tilde{\sigma}^2 \mathbf{I}$ is the covariance matrix at the Rx side
- $\tilde{\mathbf{H}}_{ij}$ is a matrix with iid Gaussian, zero mean and unit variance, entries
- The relation between the WSR and the weighted mean squared error (WMSE) to approximate the maximization of the expected WSR with the expectation of the WMSE can be exploited
- This leads to an approximate solution but easy to be handled

Output Feedback



- Each MU feeds back to all BS the noiseless version of its received signal using un-quantized feedback: **Output FB** (OFB).
- In FDD systems UL and DL transmission can take place at the same time
- T_{UL} represents the UL coherence Time
- T_{DL} represents the DL coherence Time
- OFB phase can start one time instant after the beginning of the DL training phase

Output Feedback



- Output FB allows us to reduce the overhead due to CSI exchange
- In channel FB each MU has to wait the end of the DL training phase before being able to FB DL channel estimates
- For easy of exposition we consider $M_i = N_t \forall i$, $N_i = N_r \forall i$ where $N_t \geq N_r$

- The Rx signal at MU_k at time $[t]$ during the DL training phase is

$$\mathbf{y}_k[t] = \sum_{i=1}^K \mathbf{H}_{ki} \mathbf{s}_i[t] + \mathbf{n}_k[t]$$

- At $[t+1]$ MU_k feeds back to all BSs the noiseless version of the RX signal at time instant $[t]$ (OFB).
- So BS_l receives:

$$\bar{\mathbf{y}}_l[t+1] = \sum_{j=1}^K \bar{\mathbf{H}}_{lj} \bar{\mathbf{x}}_j[t+1] + \bar{\mathbf{n}}_k[t+1] = \sum_{j=1}^K \bar{\mathbf{H}}_{lj} \alpha_j \sum_{i=1}^K \mathbf{H}_{ji} \mathbf{s}_i[t] + \bar{\mathbf{n}}_k[t+1]$$

α_j takes into account the TX power constraint at MU_j

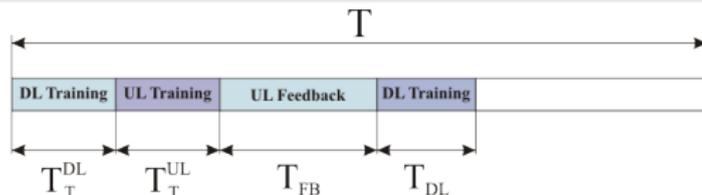
- In a distributed approach we use time multiplexing to allow all BSs to estimate all the DL channels
- Each BS has to estimate $\mathbf{H}_i = [\mathbf{H}_{i1}, \dots, \mathbf{H}_{iK}]^{N_r \times KN_t}$, then each BS needs $\tau_{FB}^o = KN_r$ samples.
- The total length of the output feedback phase is:

$$T_{FB}^o \geq K^2 N_r$$

- OFB length is the same as channel FB length
- OFB does not reduce FB duration but reduces overhead due to partial elimination of silent periods

- CSIR is usually neglected
- Some schemes for arbitrary time-varying channels assume that Rx's know all channel matrices at all time: impossible to realize in practice
- An additional DL training phase is required to build the Rx filters

Downlink Training Phase



- Once the BFs have been calculated (Centralized/Distributed) they are used in the DL transmission
- Each MU applies a ZF RX filter to suppress interference
- To design Rx filter 2 approaches are possible

(a) DL Training

(b) Analog Transmission of Rx's

(a) Each BS sends BF'd pilots to estimate ch-BF cascade or Rx

$$T_{DL} \geq \sum_k d_k$$

(b) The entire Rx filter \mathbf{F}_k is transmitted to the MU_k using analog signaling

$$T_{DL} \geq \sum_k \frac{N_k d_k}{\min\{N_k, M_k\}}$$

In the end: Sum Rate (high SNR)

- SR

$$\mathcal{R}^{PCSI} = \sum_{k,n} \underbrace{\left(1 - \frac{\sum T_i}{T}\right)}_{\text{reduced data channel uses}} \ln(|\mathbf{f}_{kn} \mathbf{H}_{kk} \mathbf{g}_{kn}|^2 \underbrace{\rho / \left(1 + \sum_i \frac{b_{kni}}{T_i}\right)}_{\text{SNR loss}}),$$

$$T_i \geq T_{i,\min}$$

Assume $b_{kni} = b_i$ for what follows.

- Fixing $\sum_i T_i = T_{\text{ovrhd}}$, optimal $T_i = T_{\text{ovrhd}} \sqrt{b_i} / (\sum_i \sqrt{b_i})$.
- Optimizing over T_{ovrhd} now

$$T_{\text{ovrhd}} = \frac{\sqrt{T} (\sum_i \sqrt{b_i})}{\sqrt{\mathcal{R}^{PCSI}}}$$

- Usually TDD transmission scheme is used to simplify the DL CSI acquisition at the BS side
- BS_k learns the DL channel \mathbf{H}_{ki} , $\forall i$ through reciprocity
- MU_i do not need to feedback \mathbf{H}_{ki} to BS_k but this channel is required at $BS_{j \neq k}$
- In **Distributed Processing** reciprocity does NOT help in reducing channel feedback overhead \implies TDD equivalent FDD
- In **Centralized Processing** reciprocity makes channel feedback NOT required
- In what follow we concentrate on FDD transmission strategy

Further Optimizing DoF

- data Tx stage (as good as perfect CSI):
 - Can FB increase DoF with perfect CSIT?
According to [HuangJafar:IT09] and [VazeVaranasi:ITsubm11] **NO** for $K = 2$ MIMO IFC; $K > 2$ is OPEN.
 - If not in general, then use of OFB is mainly (only) for CSI acquisition, not for augmenting DoF in presence of CSIT
- CSI acquisition stages:
 - Optimize number of streams/number of active antennas for small T : if less channel to learn then more time to Tx data, even if on reduced number of streams
 - Instead of going from $K = 1$ to full K immediately, could gradually increase number of interfering links (and their CSI acquisition) from 1 to K .
 - When T gets too short: delayed CSIT approaches.
 - A single (the largest) MIMO link can start transmitting right away w/o CSIT (possibly w/o CSIR also).

From Practical to More Optimal at Finite SNR

- When (analog) channel FB is of extended (non-minimal) duration, BF's can get computed and some DL transmission could start while FB is still going on. No need to wait until all CSI is gathered before transmission can get started.
- rate constants: partial CSI T_x/R_x design, diversity issues (optimized IA)
- optimization of training duration/power

- can OFB increase dof w perfect CSIT for $K \geq 3$?
- need to handle CSIR also in delayed CSIT approaches
- users with difference coherence time
- full duplex operation (2-way communications)
- minimum reciprocity:
coherence times equal on UL and DL, feasible dof same on UL and DL
- real IFC system: doubly selective