



STANFORD

*Lecture 14*

# **MAC GDFEs and Design Measures**

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# Announcements & Agenda

## ■ Announcements

- Section 5.4
- PS7 – last homework, 2 weeks, double weight.

## ■ Agenda

- ZF/MMSE Convergence Conditions (hold over from L13)
- MAC and GDFE Comparison (Sec 5.4.1)
- Tonal MAC with DMT (Section 5.4.2)
  - Tonal GDFE
  - SWF
- Designs with weighted sums (Section 5.4.3)

## ■ Problem Set 7 = PS7 (due June 5 or 7)

1. 5.16 A tonal channel
2. 5.17 GDFE MAC Design
3. 5.18 Dual computations
4. 5.19 GDFE BC design via duality
5. 5.20 IC with/without GDFE



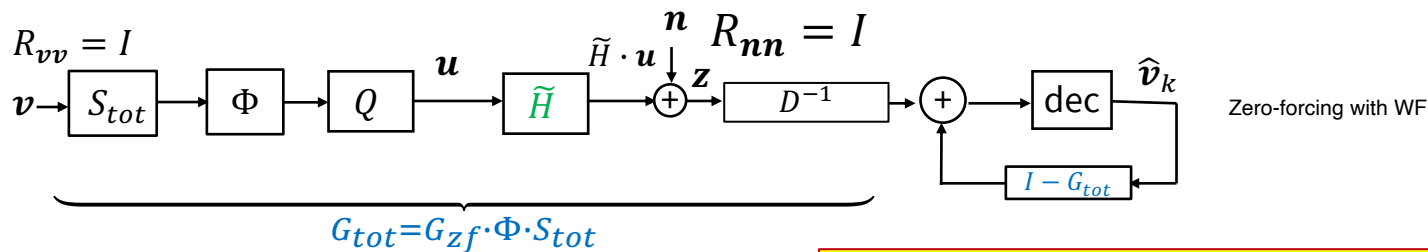
# ZF/MMSE convergence conditions

Section 5.3.5

# Use Zero Forcing GDFE with Water-fill?

- To Show:** MMSE GDFE's triangular matrix  $G$  equals the triangular factor of  $\tilde{H} = D \cdot G_{zf} \cdot Q^*$ ?
  - IF the input is water-fill and nonsingular (so resampled):**
- $D^{-1} \cdot \tilde{H} = F \cdot \Lambda \cdot M^*$  with energies  $diag\{\mathcal{E}\} = K - \Lambda^{-2} \rightarrow R_{uu} = M \cdot (K - \Lambda^{-2}) \cdot M^* = Q \cdot \Phi \cdot \Phi^* \cdot Q^*$ 
  - To find  $\Phi$ , Cholesky-Factor  $Q^* \cdot R_{uu} \cdot Q = \Phi \cdot \Phi^*$ .
  - Define monic triangular:  $G_{tot} = G_{zf} \cdot \Phi \cdot S_{tot}$  and note the MSWF and ZF-GDFE with receiver diagonal  $D^{-1}$  is:

Find  
this



always  
nonsingular

$\tilde{H} = F \cdot \Lambda \cdot M^* = D \cdot G_{zf} \cdot Q^*$
Cholesky: $Q^* \cdot R_{uu} \cdot Q = \Phi \cdot \Phi^*$

**This zero-forcing actually has  $G_{tot}$ , not  $G_{zf}$ , as feedback; however, designers often use flat input  $S_{tot} = \bar{\mathcal{E}}_x \cdot I$ , where  $K - \Lambda^{-2} \approx \bar{\mathcal{E}}_x \cdot I$  which sets  $G_{tot} = G_{zf}$  ( $M = Q$ , so  $\Phi = I$ ).**

**Equivalently, the input  $v$  as  $N \rightarrow \infty$  has  $S_{tot} \rightarrow$  constant, so then exactly true.**

**So maybe just use ZF and do only 1 rq factorization (not 2 Cholesky's).**



# The two water-fill receivers

## Rearrange Water-fill

$$R_{uu} = M \cdot [K \cdot I - \Lambda^{-2}] \cdot M^* = Q \cdot \Phi \cdot \Phi^* \cdot Q^*$$

Noting

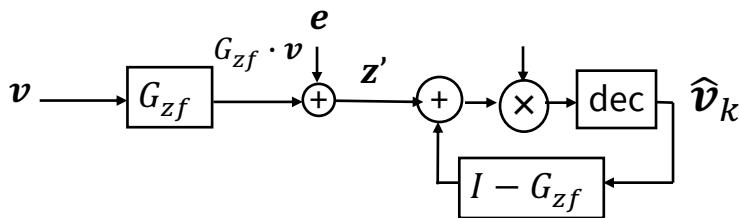
$$M \cdot \Lambda^{-2} \cdot M^* = \tilde{H}^{-1} \cdot D^2 \cdot \tilde{H}^{-*} = R_f^{-1}$$

$$K \cdot I = Q \cdot \Phi \cdot \Phi^* \cdot Q^* + Q \cdot G_{zf}^{-1} \cdot G_{zf}^{-*} \cdot Q^*$$

$$K \cdot G_{zf} \cdot G_{zf}^* = \underbrace{G_{tot} \cdot S_{tot}^{-2} \cdot G_{tot}^*}_{R_f} + I$$

**This is  $R_b^{-1}$  !**

----- Equivalent MMSE with WF (same  $G = G'_{zf}$ ):  
Performance still not same, MMSE slightly better



**$G = G_{zf}$  so water-filling leads to MMSE having same feedback as the water-fill zero-forcing, at least the flat-energy approximation to wf**

**However,  $G_{tot}$ , is really the ZF cascade when non flat or for finite symbol length.**

- The  $K$  calculation needs to be positive (or increase  $K$  so slightly positive and then scale down the resulting energies); the water-fill input was not full rank so there is a loss; can be small in wireless.
- However, MMSE still has (slightly) higher SNR so use of  $G = G_{zf}$  with waterfill as feedback, not  $G_{tot}$ , is highest SNR.
  - Note carefully on previous slide that the ZF-GDFE feedforward filter is not the same as MMSE-GDFE, even if feedback is same.



# Worst-Case Noise equates ZF and MMSE

- Easy proof: WCN diagonalizes the primary-user receivers from BC in Chapter 2.

- Step 1: Separates noise-whitened-noise-matched channel's triangular part

- $\tilde{H} \triangleq R_{wcn}^+ \cdot H = R_{zf} \cdot Q_{zf}^* = \begin{bmatrix} 0 & R_1 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} q \\ Q_1^* \end{bmatrix}$

$R_1$  is triangular part  
 $Q_1^*$  is corresponding column set

- Step 2: cascade that triangular part with Cholesky of rotated input:

- $R_{\tilde{x}\tilde{x}} = Q_1^* \cdot R_{xx} \cdot Q_1 = \Phi \cdot \Phi^*$  where  $\Phi$  is triangular Cholesky factor.
- $A = Q_1 \cdot \Phi$ .

- Step3: find the channel gains/SNR and feedback section:

- The cascade of receiver triangular inverse is  $D_A \cdot G_{zf} = R_1 \cdot \Phi$ .



# Example of WCN's RCVR Diagonalization

```
>> H =  
    0.9000    1.0000     0     0  
     0    0.9000    1.0000     0  
     0     0    0.9000    1.0000  
H=(1/sqrt(.181))*H;  
>> Rxx=eye(4);  
>> [Rwcn,b]=wcnoise(Rxx,H,1);  
>> b = 4.8024  
>> Htilde=inv(Rwcn)*H;  
>> [R,Q]=rq(Htilde);  
>> R =  
     0  -2.7323  -1.4039   0.0071  
     0     0  -2.7077  -1.2346  
     0     0     0   -3.0719  
Q1=Q(:,2:4);  
Rxxrot=Q1'*Rxx*Q1;  
Rup=R(:,2:4);  
Phibar=lohc(Rxxrot);  
DA=diag(diag(Rup*Phibar));  
>> G=inv(DA)*Rup*Phibar  
    1.0000   0.5138  -0.0026  
     0     1.0000   0.4559  
     0     0     1.0000  
>> A=Q1*Phibar;  
>> sri=inv(sqrtm(Rwcn));
```

**Rwcn is nonsingular here**

**Finding ZF modulator G**

```
>> [snrGDFEu, GU, WU, S0, MSWMFU,b,bbar] = computeGDFE(sri*H, A, 2, 4)  
snrGDFEu = 1.1340 dB  
  
GU =  
    1.0000   0.5826  -0.0030  
     0     1.0000   0.5160  
     0     0     1.0000  
  
WU =  
    0.1339     0     0  
   -0.0676   0.1317     0  
    0.0244  -0.0470   0.1031  
  
S0 =  
    8.4657     0     0  
     0   8.5956     0  
     0     0  10.7006  
  
MSWMFU =  
   -0.3657  -0.0128   0.0054  
   -0.0125  -0.3560  -0.0125  
    0.0047  -0.0111  -0.3164  
>> MSWMFU*sri' =  
   -0.3660  -0.0000   0.0000  
   -0.0000  -0.3565   0.0000  
    0.0000  -0.0000  -0.3167  
>> b' = 1.5408  1.5518  1.7098  
>> bbar = 1.2006  
>> sum(b) = 4.8024 (checks)
```

**Diagonal !  
(so ZF = MMSE GDFE)**

**Only need  
RQ fact and lohc**

- See Example 5.3.6 with non-white Rxx



# Some Final Comments

- The GDFE is canonical – capacity rate is reliably achievable with  $\Gamma = 0$  (or capacity less shaping loss).
- GDFE can have error propagation (limited to  $\bar{N}$ ) if  $\Gamma > 0$  dB.
  - Unless it is VC ( $\sim$ DMT), which is ML decoder uniquely among all GDFEs.
  - Other GDFE's becoming increasingly less favorable performance relative to VC/DMT as gap grows.
- The DMT form benefits from FFT algorithms so also more cost effective than the others.
- By Separation Theorem, Coded-OFDM can capture the DMT benefits also without error propagation.
  - But will lose more rapidly lose performance relatively if input is not water filling.
- The MMSE-DFE is limiting (stationary) case of the CDFE and can be canonical.
  - Set of MMSE-DFE's for each of which PWC holds, which
  - has unlimited error propagation (use precoder) and also degrades more rapidly for nonzero-gap codes

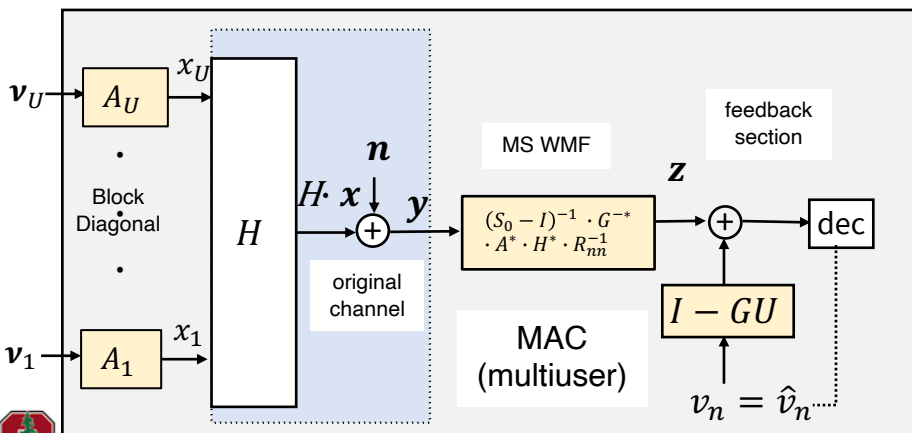
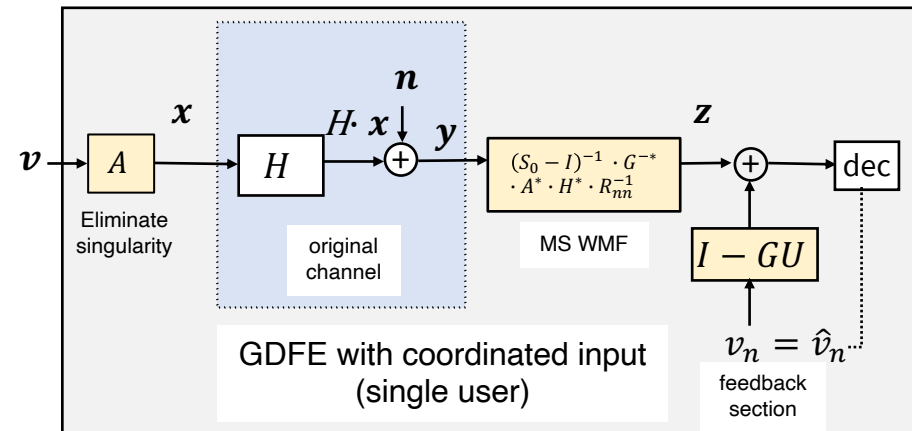
**Eventual Global Conclusion: Use DMT (wireline) or C-OFDM (wireless) on almost all difficult single-user transmission systems.**





# MAC and GDFE Comparison

# The MMSE MAC vs MMSE GDFE



## ■ GDFE is:

- Designed for "single-user"  $H$ ;  $A = R_{xx}^{1/2}$ ,
- MMSE:  $R_b^{-1} = R_{xx}^{-*/2} \cdot H^* \cdot R_{nn}^{-1} \cdot H \cdot R_{xx}^{-1/2} + I = G \cdot S_0 \cdot G^*$ ,
- Canonical (decisions correct),
- Input has only  $\text{trace}\{R_{xx}\} \leq \mathcal{E}_x$  energy constraint, &
- Rate-independent of dimensional order.

## ■ MAC has:

- **block-diag**  $R_{xx}$ , with  $\text{trace}\{R_{xx}\} \leq \mathcal{E}_x$ ,
  - only in energy-sum case, and otherwise
- input energies  $\text{trace}\{R_{xx}(u)\} \leq \mathcal{E}_u$ ,
- separated locations so  $A_u = R_{xx}^{1/2}(u)$ ;  $A = R_{xx}^{1/2}$ ,
- MMSE:  $R_b^{-1} = R_{xx}^{-*/2} \cdot H^* \cdot R_{nn}^{-1} \cdot H \cdot R_{xx}^{-1/2} + I = G \cdot S_0 \cdot G^*$
- Canonical performance (decisions correct)
  - Rates per user – order shifts sum rate among users.



# A Scalar Example Revisited

## MAC 80/60 channel

```
>> H=[80 60];
>> Rxx=0.5*eye(2); (equal energy both dim/users)
>> A=[sqrt(.5) 0'; 0 sqrt(.5) ];
>> Lxu=[1 1];
>> cb=2;

>> [b, GU, WU, S0, MSWMFU] = mu_mac(H, A, Lxu, cb);

b = 5.8222 0.3218
GU = 1.0000 0.7500 MSWMFU = 0.0177
      0 1.0000      0.0236
S0 = 1.0e+03 *
      3.2010 0
      0 0.0016

>> sum(b) = 6.1440
>> 10*log10(2^(6.1440)-1) = 18.4334 dB
```

## GDFE – remove singularity

```
>> [F,L,M]=svd(H);
>> [F,L,M]=svd(H)
F = 1
L = 100 0
M =
      0.8000 -0.6000
      0.6000 0.8000
>> 0.5*log2(1+ 0.5*L(1)^2) = 6.1440
```

All energy on pass space

```
>> 0.5*log2(1+ 1*L(1)^2) = 6.64 > 6.144
```

- For 6.64, (single-user) input is:

$$\mathbf{x} = \begin{bmatrix} .800 \\ .600 \end{bmatrix} \cdot v$$

- $v$  goes to both channel input dimensions (not MAC).
- All GDFE's with this input  $R_{xx} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  perform same
  - and trivially have  $G = 1$ .



# Or use computeGDFE.m

```
>> [snrGDFEu, GU, WU, S0, MSWMFU, b, bbar,  
snrGLEu] = computeGDFE(H, A, cb)
```

```
snrGDFEu = 18.4334 dB
```

```
GU =
```

```
1.0000 0.7500  
0 1.0000
```

```
WU =
```

```
0.0003 0  
-1.3333 1.7783
```

```
S0 = 1.0e+03 *
```

```
3.2010 0  
0 0.0016
```

```
MSWMFU =
```

```
0.0177  
0.0236
```

```
b =
```

```
5.8222  
0.3218
```

```
bbar = 3.0720
```

```
snrGLEu = 16.8125 dB
```

```
>> sum(b) = 6.1440
```

No Lx  
input

```
>> [F,L,M]=svd(H);
```

```
>> Lx=2;
```

```
[snrGDFEu, GU, WU, S0, MSWMFU, b, bbar, snrGLEu] =  
computeGDFE(H, M(:,1), cb, Lx)
```

```
snrGDFEu = 19.9566
```

```
GU = 1
```

```
WU = 1.0000e-04
```

```
S0 = 10001
```

```
MSWMFU = 0.0100
```

```
b = 6.6439
```

```
bbar = 3.3220
```

```
snrGLEu = 19.9566 (dB) – linear is same for VC
```

(same energy, but more going to best mode)

Not a MAC,  
A is not diag.

**Better to use mu\_mac with a MAC,  
than to play with cb & Lx on computeGDFE,  
which is really for single user GDFEs.**

**Similarly: use computeGDFE on single user.  
ComputeGDFE also provides linear output.**

Correct comparison with GDFE notes the A input has 2 real dimensions  
VC resets the Lx to 2 as optional 4<sup>th</sup> computeGDFE input.



# MAC Loss

- MAC Loss – ratio of single-user capacity SNR to MAC maximum-rate-sum SNR (for  $[H \quad R_{nn}]$ ).

$$\gamma_{MAC} \triangleq \frac{2^{2 \cdot \bar{c}} - 1}{2^{2 \cdot \bar{c}_{e-sum}} - 1}$$

- For the previous example:  $\gamma_{MAC} = \frac{2^{6.64} - 1}{2^{6.322} - 1} = 1.5 \text{ dB}$ .

```
>> [~, btsum, ~]=macmax(1,[80 60],[1 1],1,2)
btsum = 6.3220
```

- Clearly  $0 \leq \gamma_{MAC} \leq 1$ .



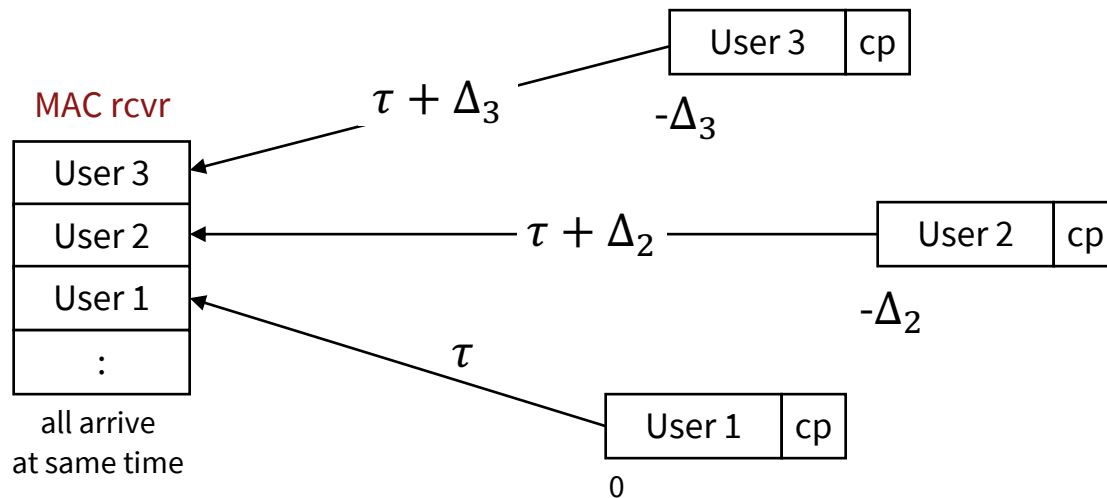
# Tonal MAC (with DMT)

## Section 5.4.2

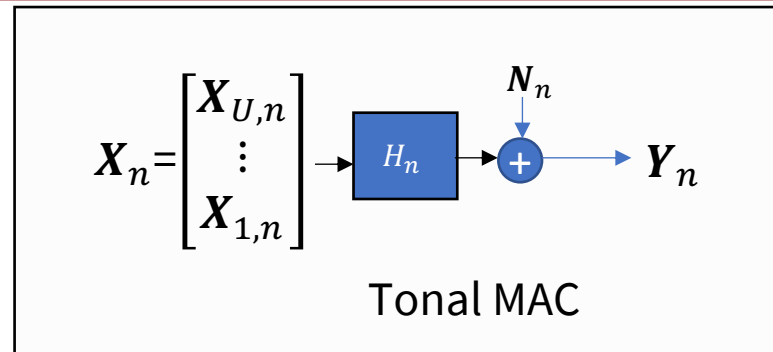
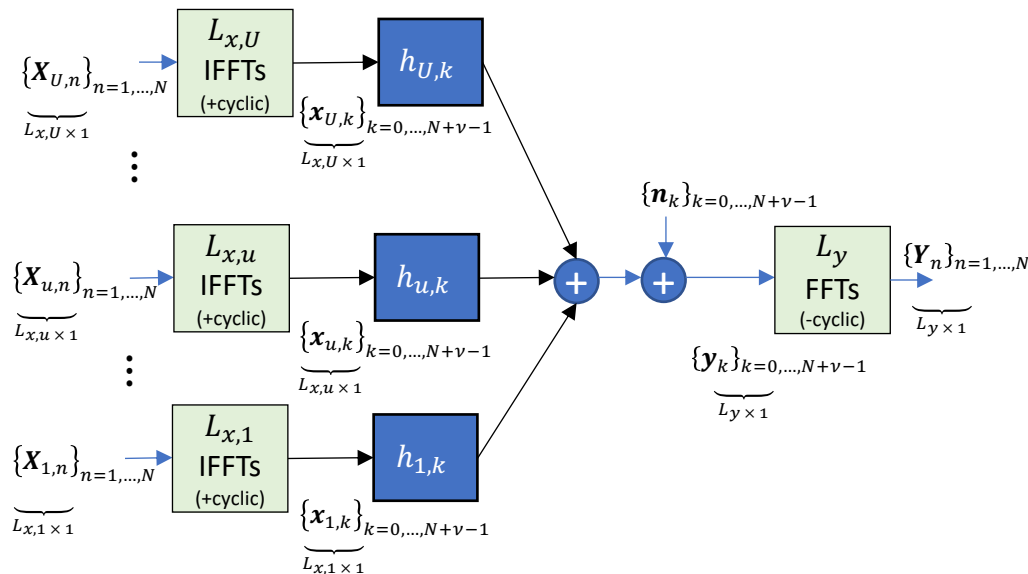
Note Section 5.4.1's two-user ISI-GDFE is interesting, but largely becomes superfluous with the tonal vector-DMT system.

# Align Receiver DMT Symbols for MAC

- So far, examples have largely been space time (with a few antennas).
- In practice, there usually is also a temporal (time-freq) C-OFDM or DMT system **also** present.



# Vector DMT/OFDM with MAC



$$R_{XX}(u, n) = \mathbb{E} \left[ X_{u,n} \cdot X_{u,n}^* \right]$$

$$\sum_n \text{trace} \{ R_{XX}(u, n) \} \leq \epsilon_u \quad \forall u = 1, \dots, U$$

$$\text{Esum-MAC: } \sum_{u=1}^U \sum_{n=0}^{\bar{N}} \text{trace} \{ R_{XX}(u, n) \} \leq \epsilon_x$$

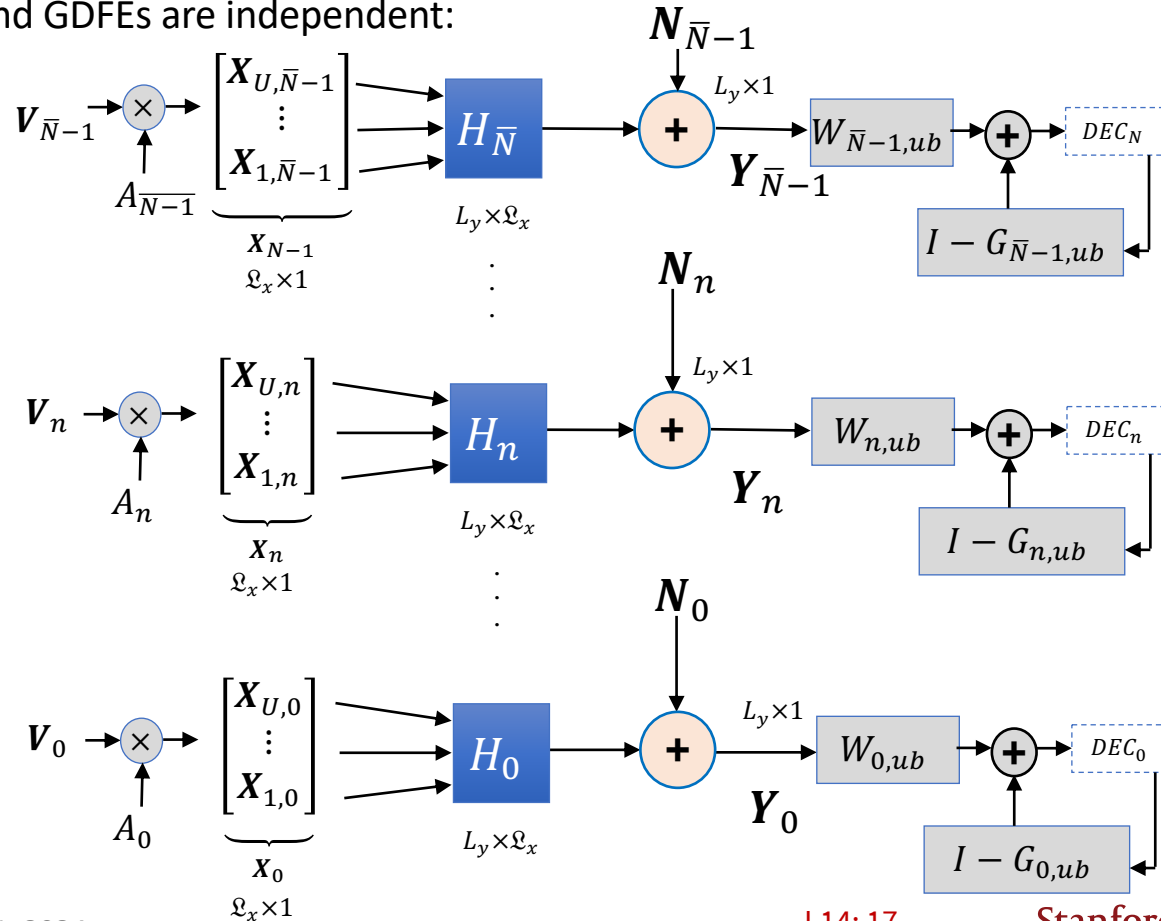
- Symbol boundaries align through cyclic-extension (guard period) uses (even with different channel delays, may increase  $v$ ).
- Basically, an IFFT per transmit-antenna-user – symbol boundary delayed/advanced so that receiver FFT captures all  $U$ .
- Discrete-time MT MAC, indeed all SVD's, Cholesky's, and QR factorizations become “frequency-dependent” (finite  $n$  replaces  $\lim_{n \rightarrow \infty} [ ]$ ).





# Tonal GDFEs with MAC

- Discrete tonal modulators and GDFEs are independent:



- Put an “ $n$ ” index on all the GDFE design equations.

$$b_{u,\ell} = \sum_{n=0}^{\bar{N}-1} b_{u,\ell,n}$$

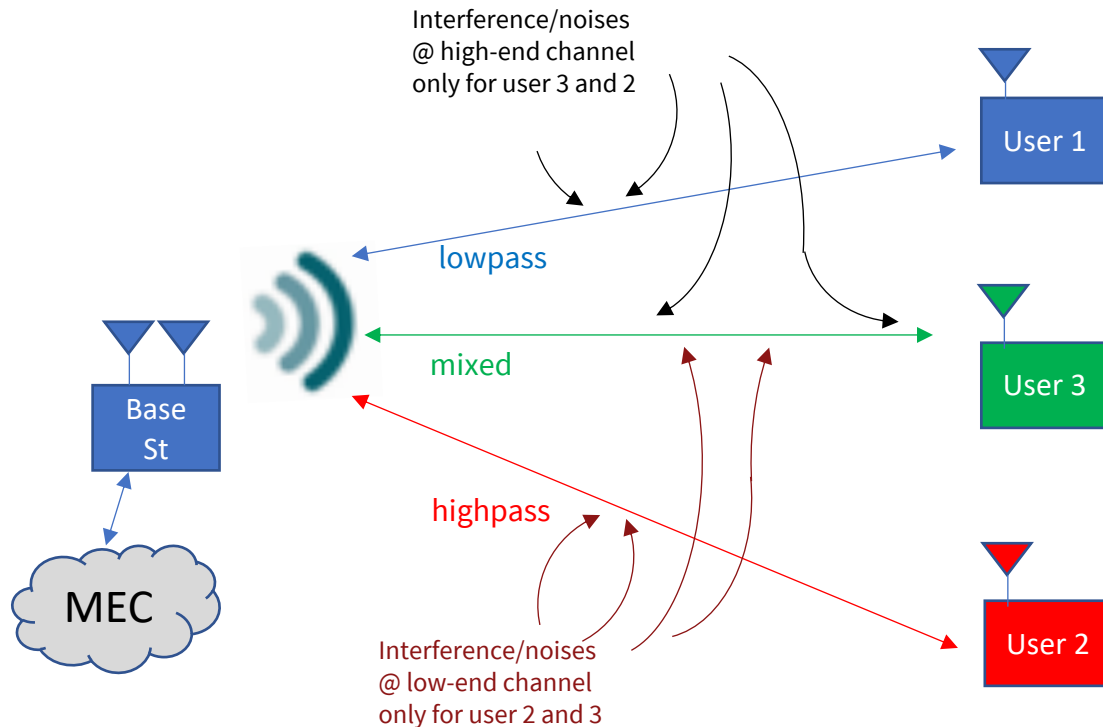
$$b_{u,n} = \sum_{\ell=1}^{L_{x,u}} b_{u,\ell,n}$$

$$b_u = \sum_{\ell=1}^{L_{x,u}} \sum_{n=0}^{\bar{N}-1} b_{u,\ell,n}$$

$$b = \sum_{u=1}^U \sum_{\ell=1}^{L_{x,u}} \sum_{n=0}^{\bar{N}-1} b_{u,\ell,n}$$



# Example complex BB channel



**More users than antennas**

This allows up to 64 resource Blocks (tones) for each user, all in same channel.

- Illustrates many effects
  - ISI and crosstalk

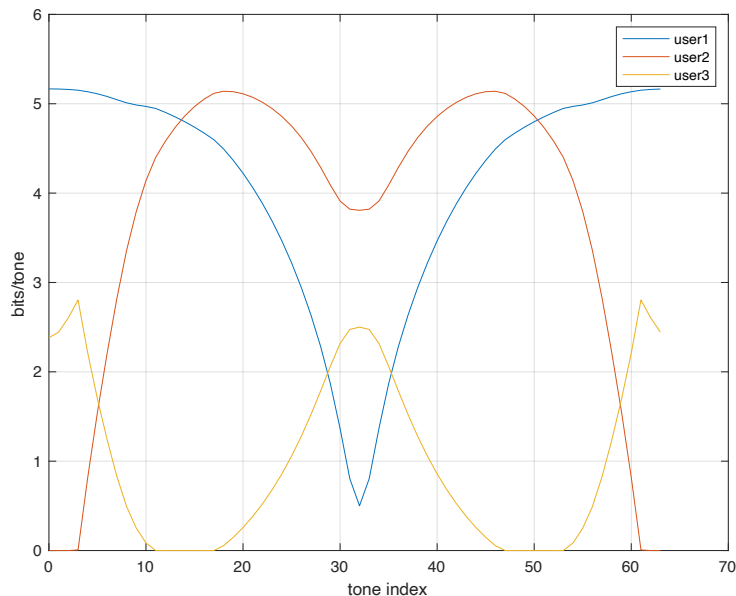
$$H(D) = \begin{bmatrix} 1 + .9 \cdot D & -.3 \cdot D + .2 \cdot D^2 & .8 \\ .5 \cdot D - .4 \cdot D^2 & 1 - D - .63 \cdot D^2 + .648 \cdot D^3 & 1 - D \end{bmatrix}$$



# Example has ISI and MIMO together

- There are 3 user channels .
- Any tone will maximally have rank  $\rho_H=2$ ;  $N_0 = .01$ .

$$H(D) = \begin{bmatrix} 1 + .9D & -.3D + .2D^2 & .8 \\ .5D - .4D^2 & 1 - D - .63D^2 + .648D^3 & 1 - D \end{bmatrix}$$



```
h=cat(3,[1 0.8; 0 1 1],[.9 -.3 0; .5 -1 -1],[0 .2 0; .4 -.63 0],[0 0 0; 0 .648 0])*10;
h(:,:,1)=
    10    0    8
     0   10   10
h(:,:,2)=
     9    -3    0
     5   -10   -10
h(:,:,3)=
     0    2.0000    0
    4.0000 -6.3000    0
h(:,:,4)=
     0    0    0
     0    6.4800    0
N=8;
H = fft(h, N, 3)
>> H = fft(h, N, 3)
H(:,:,1) =
    19.0000 + 0.0000i   -1.0000 + 0.0000i    8.0000 + 0.0000i
     9.0000 + 0.0000i    0.1800 + 0.0000i    0.0000 + 0.0000i
H(:,:,2) =
    16.3640 - 6.3640i   -2.1213 + 0.1213i    8.0000 + 0.0000i
     3.5355 - 7.5355i   -1.6531 + 8.7890i    2.9289 + 7.0711i
And 6 more values, see text
```

← Increase to 64 – look ahead at MAC with equal energy every dimension.



# Actual MAC/GDFE calculations for L14:17

## White-Input Tonal GDFE

```
Nmax=32;
U=3;
Ly=2;
cb=1;
Lxu=[1 1 1];
bsum=zeros(1,Nmax);
for index=1:Nmax
    i=2*index;
    H = fft(h, i, 3);
    GU=zeros(U,U,i);
    WU=zeros(U,U,i);
    S0=zeros(U,U,i);
    Bu=zeros(U,i);
    MSWMFU=zeros(U,Ly,i);
    AU=zeros(3,3,i);
    for n=1:i
        AU(:,n)=sqrt(i)/sqrt(i+3)*eye(3);
    end
    for n=1:i
        [Bu(:,n), GU(:,n), WU(:,n), S0(:,n), MSWMFU(:,n)] = ...
            mu_mac(H(:,n), AU(:,n), Lxu, cb);
    end
    bvec=sum(Bu');
    Bsum(index) = sum(bvec);
end
bvec = 445.1264 412.8794 132.7477
sum(bvec) = 990.7535
```

```
>> GU(:,23) =
    1.0000 + 0.0000i -1.3365 + 0.3447i -0.3093 + 0.3130i
    0.0000 + 0.0000i  1.0000 + 0.0000i  0.8108 + 0.4218i
    0.0000 + 0.0000i  0.0000 + 0.0000i  1.0000 + 0.0000i
>> MSWMFU(:,15) =
    0.0454 + 0.0341i -0.0105 + 0.0249i
    0.0183 + 0.0179i  0.0275 - 0.0468i
    0.0589 + 0.0199i  0.0179 - 0.0416i
>> SNRs = 10*log10(diag(S0(:,29)).^(64/67)-eye(3)) =

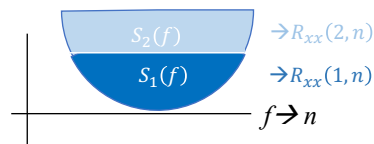
    12.1946
    20.7691
     8.8939
```

**Feedback is sizeable**



# Simultaneous Water Filling with DMT

- Frequency  $f \rightarrow$  tone index  $n$ .



See also Section 2.7.4.2 and Lecture 8.

1. Find all noise and crosstalk:

- $R_{noise}(u, n) = \sum_{i \neq u} H_{i,n} \cdot R_{XX}(i, n) \cdot H_{i,n}^* + R_{NN}(n)$

2. Create a noise-equivalent that includes all other users as noise (no order, all):

- $\tilde{H}_{u,n} = R_{noise}^{-1/2}(u, n) \cdot H_{u,n} = F_{u,n} \cdot \Lambda_{u,n} \cdot M_{u,n}^*$

3. Water-fill each user:

- $\mathcal{E}_{u,l,n} + \frac{1}{g_{u,l,n}} = K_u \quad \forall n, l$  with  $g_{u,l,n} = \lambda_{u,l,n}^2$

4. Form resulting input autocorrelation matrices (energy distribution with  $L_{x,u} = 1$ )

- $R_{XX}^o(u, n) = M_{u,n} \cdot \text{Diag}\{\mathcal{E}_{u,n}\} \cdot M_{u,n}^* \quad \forall n = 0, \dots, \bar{N} - 1$

**With MT and  $L_y=1, n \rightarrow \infty$   
There is always an FDM SWF solution.**



# SWF.m versus macmax.m

```
>> help SWF
function [Rxx, bsum , bsum_lin] = SWF(Eu, H, user_ind, Rnn, cb)
```

Simultaneous water-filling MAC max rate sum (linear and nonlinear GDFE)  
The input is space-time domain h, and the user can specify a temporal block symbol size N (essentially an FFT size).

Inputs:

Eu U x 1 energy/SAMPLE vector. Single scalar equal energy all users  
any  $(N/N+nu)$  scaling should occur BEFORE input to this program.  
H The FREQUENCY-DOMAIN  $L_y \times \text{sum}(L_x(u)) \times N$  MIMO channel for all users.  
N is determined from size(H) where  $N = \# \text{ tones}$   
(equally spaced over  $(0,1/T)$  at  $N/T$ .  
if time-domain h,  $H = 1/\text{sqrt}(N) \cdot \text{fft}(h, N, 3)$ ;  
user\_ind The start index for each user, in the same order as Eu  
The Lxu vector of each user's number of antennas is computed internally. % U is determined from user\_ind  
Rnn The  $L_y \times L_y \times N$  noise-autocorrelation tensor (last index is per tone)  
cb cb = 1 for complex, cb=2 for real baseband

Outputs:

Rxx A block-diagonal psd matrix with the input autocorrelation for each user on each tone. Rxx has size  $(\text{sum}(L_x(u)) \times \text{sum}(L_x(u)) \times N$ .  
sum trace(Rxx) over tones and spatial dimensions equal the Eu  
bsum the maximum rate sum.  
bsum bsum\_lin - the maximum sum rate with a linear receiver  
b is an internal convergence sum rate value, not output

This program is modified version of one originally supplied by student  
Chris Baca

**Energy-Vector  
MAC**

```
function [Rxx, bsum , bsum_lin] = macmax(Esum, h, Lxu, N , cb)
```

Simultaneous water-filling Esum MAC max rate sum (linear & nonlinear GDFE)  
The input is space-time domain h, and the user can specify a temporal block symbol size N (essentially an FFT size).

This program uses the CVX package

the inputs are:

Esum The sum-user energy/SAMPLE scalar in time-domain.  
This will be increased by  $cb \cdot N$  by this program.  
Each user energy should be scaled by  $N/(N+nu)$  if there is cyclic prefix  
This energy is the trace of the corresponding user Rxx (u)  
The user energy is computed as the sum of the Eu components internally.  
h The TIME-DOMAIN  $L_y \times \text{sum}(L_x(u)) \times N$  channel for all users  
Lxu The number of antennas for each user  $1 \times U$   
N The FFT size (equally spaced over  $(0,1/T)$  at  $1/(NT)$ .  
cb cb = 1 for complex, cb=2 for real baseband

the outputs are:

Rxx A block-diagonal psd matrix with the input autocorrelation for each user on each tone. Rxx has size  $(\text{sum}(L_x(u)) \times \text{sum}(L_x(u)) \times N$ .  
sum trace(Rxx) over tones and spatial dimensions equal the Eu  
FREQUENCY DOMAIN  
bsum the maximum rate sum, when cb=2, this is effectively over lower half of tones, or equivalently the  $1/2 \cdot \log_2$  form of data rates are summed  
bsum bsum\_lin - the maximum sum rate with a linear receiver  
  
b is an internal convergence (vector, rms) value, but not sum rate

**Energy-Sum  
MAC**

**Up to N on h**

**This & bcmax  
updated at  
Website.**

- SWF is frequency domain input (useful with non-white noise psd), has separate user energies, and uses no CVX.
- macmax is time-domain (and uses Lxu instead of user\_ind), has single sum energy, and uses CVX.



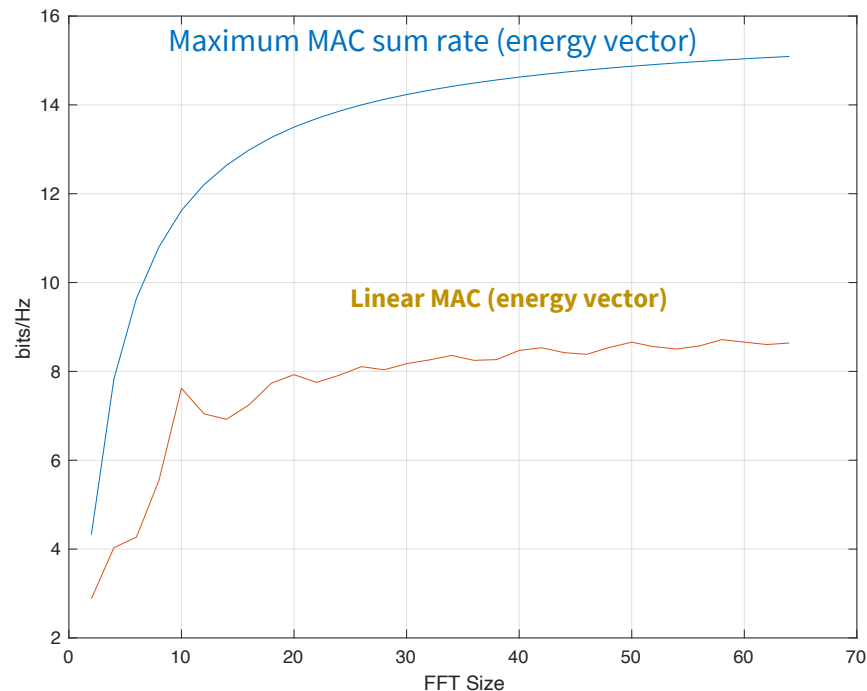
# Max rate-sum Example

```

h=cat(3,[1 0 .8 ; 0 1 1],[.9 -.3 0 ; .5 -1 -1],[0 .2 0 ; .4 -.63 0],[0 0 0 ; 0 .648 0])*10;
bsum=zeros(1,Nmax);
bsumlin=zeros(1,Nmax);

for index=1:Nmax
    i=2*index; % (don't need to plot a point for every number of tones)
    H = fft(h, i, 3);
    Rnn=zeros(Ly,Ly,i);
    for n=1:i
        Rnn(:,n) = eye(2);
    end
    [Rxx, bsum(index), bsumlin(index)] = SWF(i/(i+3)*[1 1 1], H, [1 1 1], Rnn(:,n), 1);
    bsum(index)=bsum(index)/(i+3);
    bsumlin(index)=bsumlin(index)/(i+3);
end
bsum(32)*67= 1011.1 > 990.8
bsumlin(32)*67 = 578.8502
plot(2*[1:Nmax], bsum,2*[1:Nmax],bsumlin)
    
```

- Even with 3 users > 2 antennas, linear loses much.
  - ~ 20 dB (from “link budget”)
- **Linear curve variation** is because  $\bar{N}$  is finite and the simultaneous water filling is not necessarily best solution under linear restriction
- When would “linear receiver be best?”



**If vector-coding could be used,  
But not possible on MAC in general.  
Discrete slush-packing awaits...**

**The linear  
max-sum prob  
is not convex,  
See OSB in Sec 5.6**



# SWF energy/Rxx distribution

```

Rxx(:,1)
1.4504  0  0
0  0  0
0  0  1.3050
Rxx(:,2) =
1.4512  0  0
0  0  0
0  0  1.3120
Rxx(:,3) =
1.4528  0  0
0  0  0
0  0  1.3274
.....
Rxx(:,9) =
1.4419  0  0
0  0.0670  0
0  0  1.3303
Rxx(:,11) =
1.3170  0  0
0  1.0228  0
0  0  0.5748
Rxx(:,15) =
1.3889  0  0
0  1.4116  0
0  0  0.1513
Rxx(:,26) =
0.1384  0  0
0  1.4184  0
0  0  1.3192
    
```

```

.....
Rxx(:,27) =
0  0  0
0  1.4939  0
0  0  1.3767
Rxx(:,28) =
0  0  0
0  1.4885  0
0  0  1.3761
Rxx(:,31) =
0  0  0
0  1.4229  0
0  0  1.3689
Rxx(:,32) =
0  0  0
0  1.3394  0
0  0  1.3606
Rxx(:,39) =
0  0  0
0  1.4939  0
0  0  1.3767
Rxx(:,40) =
0.1384  0  0
0  1.4184  0
0  0  1.3192
Rxx(:,51) =
1.3889  0  0
0  1.4116  0
0  0  0.1513
Rxx(:,52) =
1.3871  0  0
0  1.3929  0
0  0  0.1700
    
```

```

Rxx(:,53) =
1.3678  0  0
0  1.3329  0
0  0  0.243
Rxx(:,57) =
1.4419  0  0
0  0.0670  0
0  0  1.3303
.....
Rxx(:,58) =
1.4577  0  0
0  0  0
0  0  1.3736
Rxx(:,59) =
1.4573  0  0
0  0  0
0  0  1.3693
Rxx(:,60) =
1.4567  0  0
0  0  0
0  0  1.3633
Rxx(:,61) =
1.4557  0  0
0  0  0
0  0  1.3548
Rxx(:,62) =
1.4544  0  0
0  0  0
0  0  1.3428
Rxx(:,63) =
1.4528  0  0
0  0  0
0  0  1.3274
    
```

$$\rho_{H,n} < U ??$$

With  $\bar{N} > 1$ , there can be some tones that use all 3 dimensions. **Sep Theorem** applies only over a single user's tones.

These are equivalent to time-shared (dimension shared) of 2-user-only tones, but with same order on all tones.

Freq sharing cannot happen when  $\bar{N} = 1$  &  $\rho_H < U$ .

**Also  $L_y > 1$ ,  
So FDM not assured.**





# macmax (new at website, also bcmax)

```
[RxxEsum, bsumEsum , bsum_linEsum] = macmax(3*64/67, h, [1 1 1], 64 , 1);  
  
bsumEsum = 1011.3 > 1011.2           (just slightly)  
bsum_linEsum = 571.7289 < 578.85     (no guarantee that linear version is best)  
  
>> sum(real(Rxx),3) =  
61.1343    0    0  
    0 61.1343    0  
    0    0 61.1343  
>> 64^2/67 = 61.1343 checks on each dimension  
>> trace(sum(real(Rxx),3)) = 183.4030 (clearly 3x single dimensional energy)
```

**Any order can be used for same rate sum  
(in this case maximum)**

**But, the same order needs to be used  
on ALL tones.**

**Secondary user COMPONENTS are 0.**

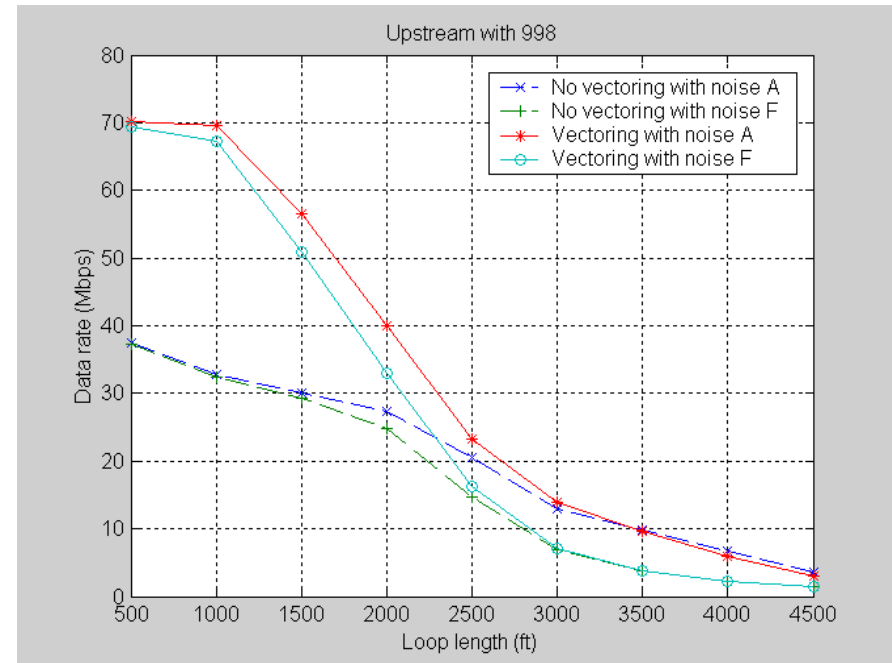
**(there is no additional time-sharing  
of two vertices for a SWF/macmax point)**

- The RxxEsum are very similar to those from SWF.m.
- While many tones individually zero one user (consistent with secondary-component concept).
  - It is not the same user for all such tones.
  - Some tones energize all 3 users.
- $\sum_n \rho_{H,n} > U$  , significantly so. This means there is effectively dimension-sharing occurring over the 64 tones, at least for the rate-sum max.



# Gain is larger when crosstalk is larger

- Binders of copper wires (think ethernet or your/neighbors' cable of telephone wires) crosstalk.
  - Highly variable with twisting (even measuring point can lead to 20dB or more variation if moved an inch or two).
  - Probabilistic models (like wireless' distributions) are also used.
  - Average xtalk is larger on SHORTER wires because xtalk coupling increases with frequency.
    - Shorter wires use higher frequencies that are less attenuated
- Example is vectored VDSL (upstream MAC).
- Each user has its own "link" that terminates (upstream) on a common receiver – by default all primary users (no time-sharing needed).
  - "perfect massive MIMO" all (used) tones (plot is for 25 links)
  - Can see up to  $U=384$  links vectored (predates "massive MIMO" in invention and use by 10 years).
- The GDFE cancels the crosstalk.
- It exhibits diagonal dominance too. Why?
  - So typically no feedback section is used.
- Actually, some "mgfast" (ITU G.9711) to 5 Gbps multiuser and 10 Gbps "fastback" (ITU G.9702), 2 pairs – 3 channels, single user can use some GDFE's at lowest frequencies.



# Wireless – uplink Cellular or Wi-Fi

- The C-OFDM systems are as in Lecture 6 (Wi-Fi and cellular).
- A single MAC receiver uses common FFTs (one for each, max #, of spatial stream/dimension with MIMO).
- They share common frequencies.
- Usually, no feedback sections are used ... yet, so linear setting from computeGDFE provides performance.

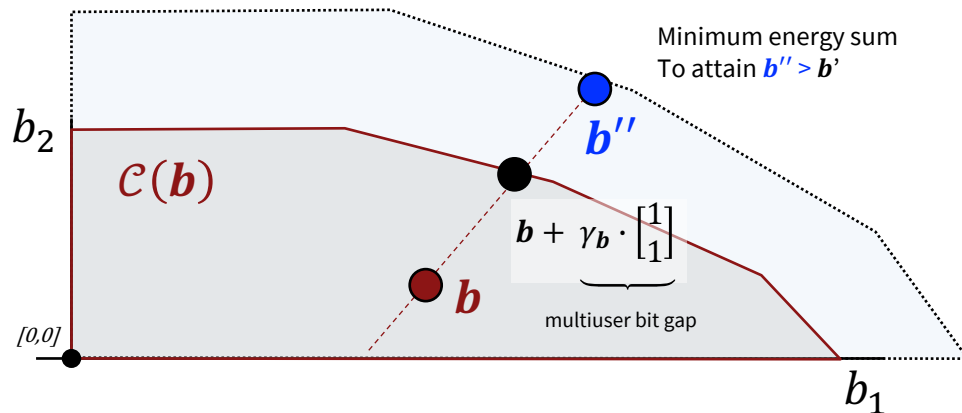
**OK – All good, but what is  $R_{xx}(u)$  when we don't maximize a rate sum??**



# Designs with weighted sums

## Section 5.4.3

# Capacity region(s)



- $\mathcal{C}(\mathbf{b})$  contains all possible weighted **rate** sums  $\sum_{u=1}^U \theta_u \cdot b_u$  that meet **energy-vector** constraint  $\boldsymbol{\varepsilon} \preceq \boldsymbol{\varepsilon}_x$ .
- The max-b-sum point is “highest” (tangent to plane  $\mathbf{1}^t \cdot \mathbf{b}$ ) with  $\mathbf{b}$  in  $\mathcal{C}(\mathbf{b})$ , **but we want another  $\mathbf{b}$ !**
- If  $\boldsymbol{\varepsilon} \preceq \boldsymbol{\varepsilon}_x$ , **then  $\mathbf{b}$  is admissible.**
- If not, a design might still target the minimum energy-sum that achieves  $\mathbf{b}$ .



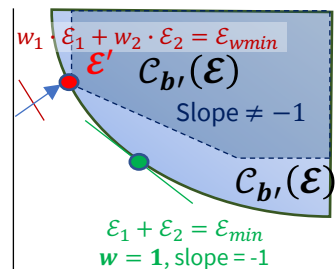
# minimum weighted energy sum

- Minimize **weighted energy** sum  $\mathcal{E}_x$ , given  $\mathbf{b}$ 
  - No energy limit (beyond minimum)
  - Energy weights  $\mathbf{w}$  given, non-negative

$$\min_{\{R\mathbf{x}\mathbf{x}(u)\}} \sum_{u=1}^U w_u \cdot \underbrace{\text{trace}\{R\mathbf{x}\mathbf{x}(u)\}}_{\mathcal{E}_u} \quad \mathcal{E}_2$$

$$ST: \quad \mathbf{b} \succeq [b_{1,\min} \ b_{2,\min} \ \dots \ b_{U,\min}]^* = \mathbf{b}_{\min}^* \succeq \mathbf{0}$$

$$\mathcal{E} \succeq \mathbf{0} \ .$$

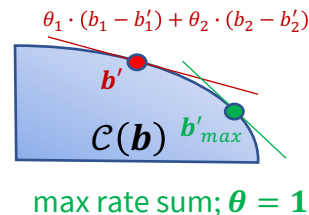


- Maximize **weighted rate** sum  $b$ , given  $\mathcal{E}$ 
  - No rate limit
  - rate weights  $\boldsymbol{\theta}$  given, non-negative

$$\max_{\{R\mathbf{x}\mathbf{x}(u)\}} \sum_{u=1}^U \theta_u \cdot b_u$$

$$ST: \quad \mathcal{E} \preceq [\mathcal{E}_{1,\max} \ \mathcal{E}_{2,\max} \ \dots \ \mathcal{E}_{U,\max}]^* = \mathcal{E}_{\max}^* \preceq \mathbf{0}$$

$$\mathbf{b} \succeq \mathbf{0} \ .$$



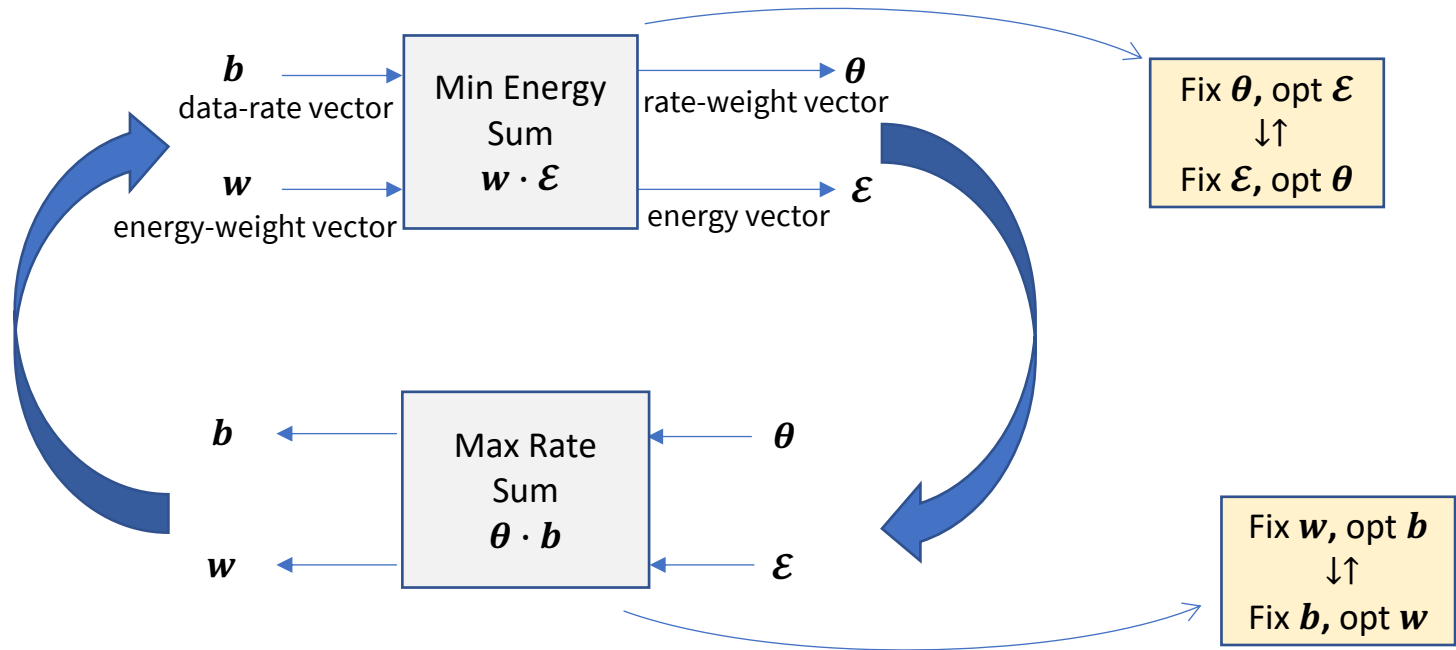
- These are “dual” problems

$$L_{\min E}(R\mathbf{x}\mathbf{x}, \mathbf{b}, \mathbf{w}, \boldsymbol{\theta}) = \max_{\boldsymbol{\theta}} \min_{R\mathbf{x}\mathbf{x}} \underbrace{\sum_{u=1}^U [w_u \cdot \text{trace}\{R\mathbf{x}\mathbf{x}(u)\} + \theta_u \cdot b_u]}_{\text{common term}} - \theta_u \cdot b_{\min,u}$$

$$L_{\max R}(R\mathbf{x}\mathbf{x}, \mathbf{b}, \mathbf{w}, \boldsymbol{\theta}) = \min_{\mathbf{w}} \max_{R\mathbf{x}\mathbf{x}} \underbrace{\sum_{u=1}^U [w_u \cdot \text{trace}\{R\mathbf{x}\mathbf{x}(u)\} + \theta_u \cdot b_u]}_{\text{common term}} - w_u \cdot \mathcal{E}_{\max,u}$$



# Basic Solution Cycles



- Each of these “boxes” (subnetworks) can be intense calculation, but ( $\sim$ , see L14: 31) convex and convergent.
- The overall recursive cycling also converges if  $\mathbf{b} \in \mathcal{C}(\mathbf{b})$ .

Use Convex, ML/AI methods ...



# Tonal Lagrangian

- Minimize (over  $R_{\mathbf{X}\mathbf{X}}(u)$ ) weighted sum at any given (think temporary)  $\boldsymbol{\theta}$  where  $\mathbf{b}$  and  $\mathbf{w}$  are the specified values:

$$L(R_{\mathbf{X}\mathbf{X}}, \mathbf{b}, \mathbf{w}, \boldsymbol{\theta}) = \sum_{n=0}^{\bar{N}-1} \left\{ \underbrace{\sum_{u=1}^U \left[ w_u \cdot \text{trace} \{ R_{\mathbf{X}\mathbf{X}}(u, n) \} - \sum_{u=1}^U \theta_u \cdot b_{u,n} \right]}_{L_n(R_{\mathbf{X}\mathbf{X}}(n), \mathbf{b}_n, \mathbf{w}, \boldsymbol{\theta})} \right\} + \theta_u \cdot b_u ,$$

With fixed  $\boldsymbol{\theta} \geq \mathbf{0}$   
each tone can  
be individually  
minimized

- which produces then for tone  $n$ :

$$L_{min}(\boldsymbol{\theta}, n) \triangleq \min_{\{R_{\mathbf{X}\mathbf{X}}(u, n)\}, b_{u,n}} L_n(R_{\mathbf{X}\mathbf{X}}(n), \mathbf{b}_n, \mathbf{w}, \boldsymbol{\theta}) .$$

- Then, max over  $\boldsymbol{\theta}$

$$L^* = \max_{\boldsymbol{\theta}} \sum_{n=0}^{\bar{N}-1} L_{min}(\boldsymbol{\theta}, n) \triangleq \max_{\boldsymbol{\theta}} L_{min}(\boldsymbol{\theta}) ,$$

- and satisfy tonal GDFE (achievable region) constraint

$$\mathbf{b}_n \in \left\{ \mathbf{b}_n \mid 0 \leq \sum_{\mathbf{u} \subseteq U} b_{\mathbf{u},n} \leq \log_2 \left| \left( \sum_{u=1}^U \tilde{H}_{u,n} \cdot R_{\mathbf{X}\mathbf{X}}(u, n) \cdot \tilde{H}_{u,n}^* \right) + I \right| \right\} = \mathcal{A}_n (\{R_{\mathbf{X}\mathbf{X}}(n)\}, \bar{H}_n).$$

(5.2)





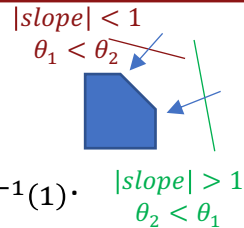
# The tonal achievable-region constraint

- Maximum  $\sum_{u=1}^U \theta_u \cdot b_{u,n}$  occurs at  $\mathcal{A}_n(R_{\mathbf{X}\mathbf{X}}(u, n), \mathbf{b}_n)$  vertex (think slope -1 line and pentagon),
  - Given  $R_{\mathbf{X}\mathbf{X}}(u, n) \rightarrow R_{\mathbf{X}\mathbf{X}}(u)$ ; equivalently max  $\sum_{u=1}^U \theta_u \cdot b_u$  occurs at  $\mathcal{A}(R_{\mathbf{X}\mathbf{X}}(u), \mathbf{b})$  vertex.
- That max-weighted-sum vertex has specific  $\theta_u$ , which must satisfy  $\theta_{\pi^{-1}(U)} \geq \theta_{\pi^{-1}(U-1)} \geq \dots \geq \theta_{\pi^{-1}(1)}$ .
  - Proof see Theorem 5.4.1, and also discussion in Sec 5.4.4.1. (Same order all tones)
  - Alternative to testing all orders – optimum order is inferred from the (converged) real vector  $\theta$ .
- The user data rates in  $\mathcal{A}_n(R_{\mathbf{X}\mathbf{X}}(u, n), \mathbf{b}_n)$  must satisfy the (sum of) **tonal-GDFE constraint(s)**:

$$b_{u,n} = \log_2 \left\{ \frac{|R_{yy}(u, n)|}{|R_{yy}(u-1, n)|} \right\} = \log_2 \left| \sum_{i=1}^u \tilde{H}_{\pi^{-1}(i), n} \cdot R_{\mathbf{X}\mathbf{X}}(\pi^{-1}(i), n) \cdot \tilde{H}_{\pi^{-1}(i), n}^* + I \right| - \log_2 \left| \sum_{i=1}^{u-1} \tilde{H}_{\pi^{-1}(i), n} \cdot R_{\mathbf{X}\mathbf{X}}(\pi^{-1}(i), n) \cdot \tilde{H}_{\pi^{-1}(i), n}^* + I \right|$$

- For given  $\theta$ , min weighted rate sum over  $R_{\mathbf{X}\mathbf{X}}(u, n)$  minimizes convex sum

$$\sum_{u=1}^U \theta_u \cdot b_{u,n} = \sum_{u=1}^U \left\{ \underbrace{[\theta_{\pi^{-1}(u)} - \theta_{\pi^{-1}(u+1)}]}_{\delta_{\pi^{-1}(u)} \leq 0} \cdot \log_2 \left| \sum_{i=u}^U \tilde{H}_{\pi^{-1}(i), n} \cdot R_{\mathbf{X}\mathbf{X}}(\pi^{-1}(i), n) \cdot \tilde{H}_{\pi^{-1}(i), n}^* + I \right| \right\}.$$



# Equal Theta

- Successive equal theta values
  - can happen, often!
  - This usually happens when there are secondary user components.
- The corresponding rate-sum difference term(s) is (are) zero.
- Only the sum rate of the corresponding users can be varied  $b_{\pi^{-1}(u)} + b_{\pi^{-1}(u)+1}$  is optimized.
- The corresponding vertices for swapping the order (more generally varying when 3 or more) need to be “vertex-shared” in a proportion that causes the desired data rate to be achieved.
- Complicated program in matlab to do this, although concept is straightforward.

**Coming Attraction: The Stanford minPMAC program(s)**





# End Lecture 14

# Two iterated steps

- $R_{\mathbf{X}\mathbf{X}}(u, n)$  step: With the given (current)  $\boldsymbol{\theta}$ ,  $\mathbf{w}$ ,  $\{b_{u,n}\}$ , minimize the (neg) weighted rate sum over  $R_{\mathbf{X}\mathbf{X}}(u, n)$ 
  - Each tone separately and sum

$$L_{min}(\boldsymbol{\theta}, n) = \underbrace{\sum_{u=1}^U \left[ w_u \cdot \text{trace} \{ R_{\mathbf{X}\mathbf{X}}(u, n) \} - \sum_{u=1}^U \theta_u \cdot b_{u,n} \right]}_{L_n(R_{\mathbf{X}\mathbf{X}}(n), \mathbf{b}_n, \mathbf{w}, \boldsymbol{\theta})}$$

e.g.  $L_{k+1} = L_k - \mu \cdot (\nabla^2 L_k)^{-1} \cdot \nabla L_k$       Weighted steepest descent (“Newton”)

- Order step: With the given (current)  $R_{\mathbf{X}\mathbf{X}}(u, n)$ ,  $\mathbf{w}$ ,  $\{b_{u,n}\}$ , maximize the Lagrangian over  $\boldsymbol{\theta}$

$$L(\boldsymbol{\theta}) = \sum_{n=1}^{\bar{N}} L_{min}(\boldsymbol{\theta}, n)$$

Initialize (first time only) with FM SWF for given  $\mathbf{b}$   
This is the “find the vertex set” – elliptic algorithm, see text

