#### Symmetric vs. Sum Capacity of Rayleigh MAC

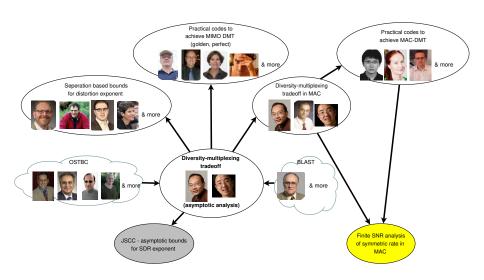
or

Probability of Achieving Fairness for Free

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February 14th, 2018
2018 Information Theory and Applications Workshop





#### Channel Model

• MAC: 
$$y = \sum_{i=1}^{N} h_i x_i + z$$



- CSI at Rx
- ullet Equal average transmission power per antenna: P=1
- $z \sim \mathcal{CN}(0,1)$
- $h_i \sim \sqrt{\text{SNR}} \cdot \mathcal{CN}(0,1)$  and i.i.d. (symmetric setting)



#### **Definitions**

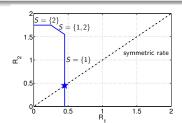
- ullet Sum capacity:  $C_{
  m sum} = \log \left( 1 + \sum |h_i|^2 
  ight)$
- Capacity region:

$$C(\mathbf{h}) = \sum_{i \in S} R_i \le \log \left(1 + \sum_{i \in S} |h_i|^2\right), \ S \subseteq \{1, \dots, N\}$$

Symmetric capacity:

$$C_{\text{sym}} = \max_{\mathbf{R} \in C(\mathbf{h})} \min(R_1, \dots, R_N) = \min_{S \subseteq \{1, \dots, N\}} \frac{1}{|S|} \log \left( 1 + \sum_{i \in S} |h_i|^2 \right)$$

$$C_{\Sigma - \text{sym}} = N \cdot C_{\text{sym}}$$

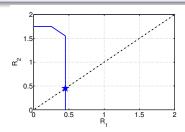


# Need to analyze the bottleneck!

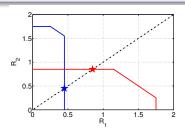


- $C_{\Sigma-\mathrm{sym}} = C_{\mathrm{sum}} \Rightarrow$  fairness comes for free!
- But what are the chances of that happening?
- Q1: Probability that  $C_{\Sigma-\mathrm{sym}} = C_{\mathrm{sum}} = \log\left(1 + \sum_{i=1}^{N} |h_i|^2\right)$  is?
- ullet We analyze the probability given  $\mathcal{C}_{\mathrm{sum}}$
- ullet Let's start with a concrete example:  $\mathcal{C}_{\mathrm{sum}}=2$

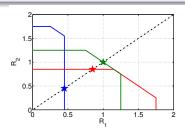
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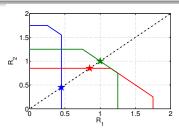
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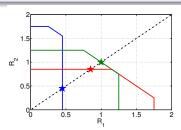
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Well, there are three faces so... 1/3?



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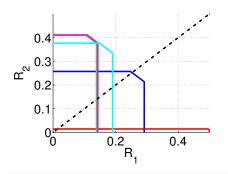


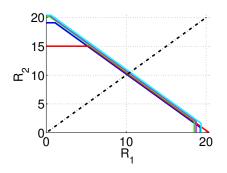
#### Correct answer, explanation can be improved...



#### What is Known

- $C_{\Sigma-\mathrm{sym}} \leq C_{\mathrm{sum}}$
- ullet (Implicitly from the MAC-DMT): SNR  $o \infty \ \Rightarrow \ \mathit{C}_{\Sigma \mathrm{sym}} \overset{w.h.p.}{ o} \ \mathit{C}_{\mathrm{sum}}$





 $\bullet$  Our goal: analyze the (finite SNR) distribution of  $\textit{C}_{\Sigma-\mathrm{sym}}$  given  $\textit{C}_{\mathrm{sum}}$ 



# Goal & Applications

**Q2:** What is 
$$\Pr\left(C_{\Sigma-\text{sym}} < R | C_{\text{sum}} = c_{\text{sum}}\right)$$
?

#### **Applications**

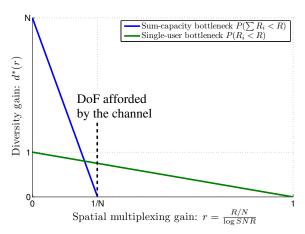
- Simple MAC transmission protocol
  - Receiver learns channel gains of active users
  - Calculates Symmetric capacity and notifies transmitters to each transmit at rate R/N where  $R < C_{\Sigma-\mathrm{sym}}$ :
    - Trivial rate allocation
    - Minimal feedback
- Rayleigh open-loop outage probability
  - N active users
  - $\triangleright$  All users (when they are active) transmit at a common target rate R/N
  - Outage probability is then given by  $\mathbb{E}_{C_{\mathrm{sum}}}[\Pr(C_{\Sigma-\mathrm{sym}} < R | C_{\mathrm{sum}})]$

#### Let's recall the DMT of the MAC



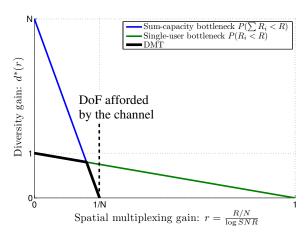
# Moral 1 from the Symmetric MAC-DMT

#### Shouldn't be too bad...



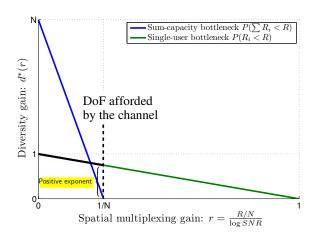
# Moral 1 from the Symmetric MAC-DMT

#### Shouldn't be too bad...



#### Moral 1 from the Symmetric MAC-DMT

# But, in our analysis/protocol we know $C_{\text{sum}}$ ...



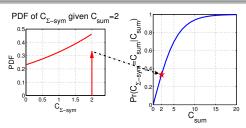
## Bottom Line - Two-User Rayleigh MAC

#### Theorem 1

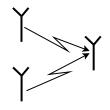
For a  $1 \times 2$  Rayleigh MAC with sum capacity  $C_{\text{sum}}$ :

$$\Pr(C_{\Sigma-\text{sym}} < R | C_{\text{sum}}) = 2 \cdot \frac{2^{R/2} - 1}{2^{C_{\text{sum}}} - 1}; \ 0 \le R \le C_{\text{sum}}$$

$$\begin{split} \Pr(\textit{C}_{\Sigma-\mathrm{sym}} = \textit{C}_{\mathrm{sum}} | \textit{C}_{\mathrm{sum}}) &= 1 - \Pr(\textit{C}_{\Sigma-\mathrm{sym}} < \textit{C}_{\mathrm{sum}} | \textit{C}_{\mathrm{sum}}) \\ &= 1 - 2 \cdot \frac{2^{\textit{C}_{\mathrm{sum}}/2} - 1}{2^{\textit{C}_{\mathrm{sum}}} - 1} \end{split}$$



# Sketch of Proof: Two-User Rayleigh MAC



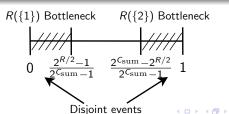
- $h_i \sim \mathcal{CN}(0,\mathsf{SNR})$  and i.i.d  $\Rightarrow |h_i|^2 \sim exp(\mathsf{SNR})$
- Normalize  $u_i = \frac{1}{\sqrt{2^{C_{\text{sum}}}-1}}h_i$
- Given  $C_{\text{sum}}$ 
  - $|u_1|^2 + |u_2|^2 = 1 \Rightarrow \text{zero-sum game}$
  - $|u_i|^2$  given  $|u_1|^2 + |u_2|^2 = 1$  is uniformly distributed over [0,1] (conditioning property of Poisson process)



# Sketch of Proof: Two-User Rayleigh MAC

• Recall: 
$$C_{\Sigma-\mathrm{sym}} = \min_{S \subseteq \{1,2,...,N\}} \frac{N}{|S|} \log \left(1 + \sum_{i \in S} |h_i|^2\right)$$

- For  $1 \times 2$ ,  $C_{\Sigma-\text{sym}} = \min(R(\{1\}), R(\{2\}), C_{\text{sum}})$
- $R({i}) = 2 \log (1 + |u_i|^2 (2^{C_{\text{sum}}} 1))$
- $\Pr\left(C_{\Sigma-\text{sym}} < R | C_{\text{sum}}\right) =$  $\Pr\left(|u_1|^2 < \frac{2^{R/2}-1}{2^{C_{\text{sum}}}-1}\right) + \Pr\left(|u_1|^2 > \frac{2^{C_{\text{sum}}}-2^{R/2}}{2^{C_{\text{sum}}}-1}\right)$
- $\Rightarrow \Pr\left(C_{\Sigma-\text{sym}} < R | C_{sum}\right) = 2\frac{2^{R/2}-1}{2^{C_{\text{sum}}}-1}$



#### General N: The Bottleneck

- When N > 2:
  - There are more possible bottlenecks to check (but remember the DMT moral...)
  - Need to analyze

$$\begin{split} & \Pr\left(R(\{S\}) < R \big| \mathcal{C}_{\text{sum}}\right) = \\ & \Pr\left(\frac{|S|}{N} \log \left(1 + \left(2^{\mathcal{C}_{\text{sum}}} - 1\right) \sum_{i \in S} |u_i|^2\right) < R \mid \sum |u_i|^2 = 1\right) \end{split}$$

- Possible bottlenecks  $\{S\}$  are no longer disjoint
- Tool for analysis
  - Given  $C_{\text{sum}}$ ,  $u_i$  can be viewed as elements from a row taken from a unitary matrix drawn from the CUE (Haar measure)
  - Edelman 05' Singular value distribution of a truncated unitary matrix (eigenvalues have Jacobi/MANOVA distribution)
- ⇒ lower and upper bounds



#### General N: The Bottleneck

#### Theorem 2 - distribution of a specific set

For a  $1 \times N$  Rayleigh MAC with sum capacity  $C_{\mathrm{sum}}$ , the outage probability for a set  $S \subseteq \{1,2,\ldots,N\}$  is

$$\begin{split} & \Pr\left(R(\{S\}) < R|C_{\text{sum}}\right) = \\ & \Pr\left(\frac{|S|}{N}\log\left(1 + (2^{C_{\text{sum}}} - 1)\sum_{i \in S}|u_i|^2\right) < R \mid \sum |u_i|^2 = 1\right) = \\ & \frac{\mathcal{B}(\frac{2^{R|S|/N} - 1}{2^{C_{\text{sum}}} - 1}; |S|, N - |S|)}{\mathcal{B}(1; |S|, N - |S|)} \end{split}$$

where  $0 \le R \le C_{\text{sum}}$  and  $\mathcal{B}(x; a, b) = \int_0^x u^{a-1} (1-u)^{b-1} du$  is the incomplete beta function.



#### General N: The Bottleneck

• 
$$\Pr\left(C_{\Sigma-\text{sym}} < R | C_{\text{sum}}\right) = \Pr\left(\min_{S \subseteq \{1,2,...,N\}} R(\{S\}) < R | C_{\text{sum}}\right)$$

- All sets with the same cardinality have the same outage probability
- $P_{\mathrm{out}}(k,R) \triangleq \Pr(R(\{|S|=k\} < R|C_{\mathrm{sum}}))$
- Union bound can be used to bound overall probability

#### Theorem 3 - lower and upper bound for N Rayleigh MAC

For a  $1 \times \textit{N}$  Rayleigh MAC with sum capacity  $\textit{C}_{\mathrm{sum}}$ , the outage probability can be bounded as

$$\max_{k} P_{\text{out}}(k, R) \leq \Pr\left(C_{\Sigma-\text{sym}} < R | C_{\text{sum}}\right) \leq \sum_{k=1}^{N} {N \choose k} P_{\text{out}}(k, R)$$



#### Upper and Lower Bounds

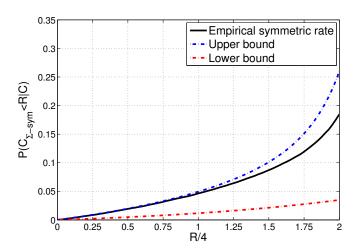


Figure: Bounds vs. Empirical error probability for  $1\times 4$  channel with  $\textit{C}_{\mathrm{sum}}/4=2$ 

# Practical Scheme (NOMA): Two-User Example

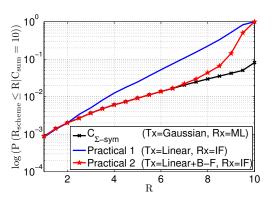


Figure: ML vs. IF for a two-user i.i.d. Rayleigh fading MAC with  $C_{\rm sum}=10$ .

#### Ingredients

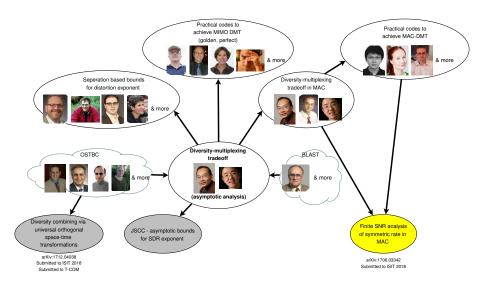
#### Integer-forcing (Rx)

- Beyond this talk
- Use same linear code
- Not sufficient

#### MAC-DMT (Tx)

- Hollanti, et. al., '11 (uncoded, asymptotic)
- Same linear code ⇒ need to use (different) "space"-time modulation
- Badr, et. al. '08





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