

# Symmetric vs. Sum Capacity of Rayleigh MAC

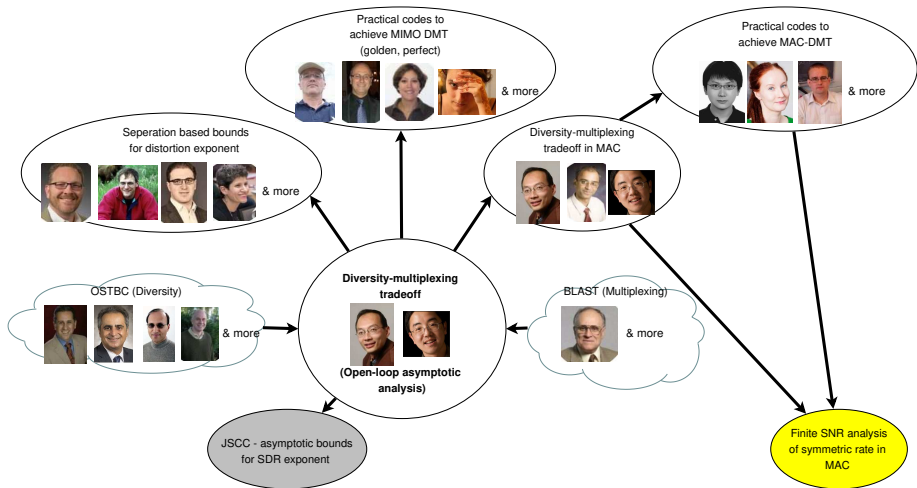
or

Probability of Achieving Fairness for Free

Elad Domanovitz

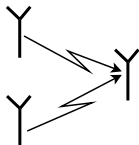
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# Channel Model

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



- CSI at Rx
- 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

- Sum capacity:  $C_{\text{sum}} = \log \left( 1 + \sum |h_i|^2 \right)$

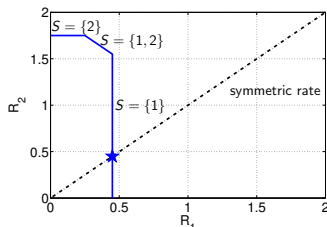
- Capacity region (set of constrains):

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

- Symmetric capacity:

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

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



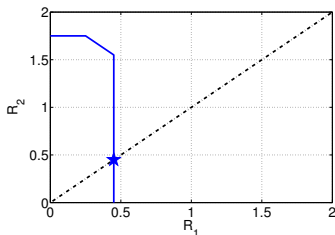
**Need to analyze  
the bottleneck !**

# Symmetric vs. Sum Capacity

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

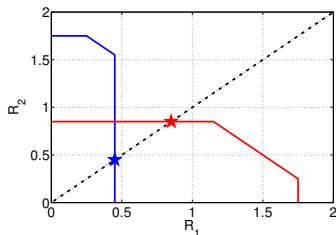
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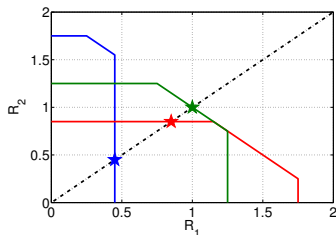
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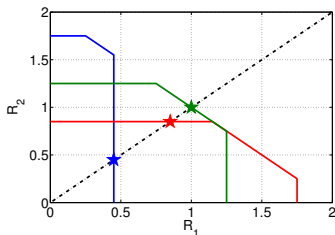
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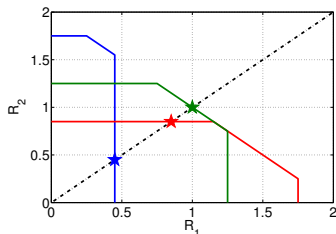
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Well, there are three faces so... 1/3?

# Symmetric vs. Sum Capacity

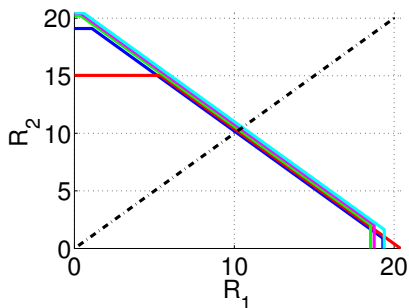
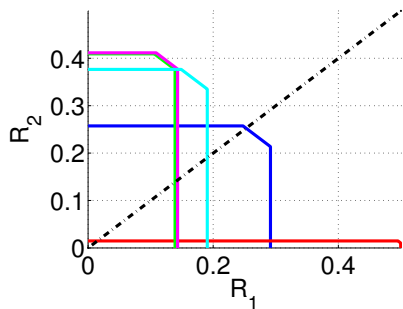
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**Correct answer, explanation can be improved...**

# What is Known

- $C_{\Sigma\text{-sym}} \leq C_{\text{sum}}$
- (Implicitly from the MAC-DMT):  $\text{SNR} \rightarrow \infty \Rightarrow C_{\Sigma\text{-sym}} \xrightarrow{w.h.p.} C_{\text{sum}}$



- Our goal: analyze the (finite SNR) distribution of  $C_{\Sigma\text{-sym}}$  given  $C_{\text{sum}}$

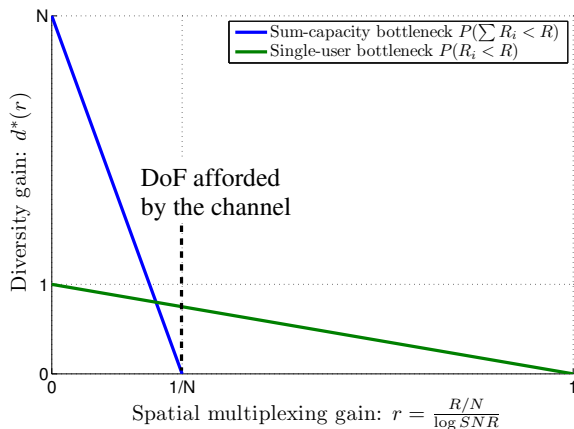
**Q2: What is  $\Pr(C_{\Sigma\text{-sym}} < R | C_{\text{sum}} = c_{\text{sum}})$  ?**

## Applications

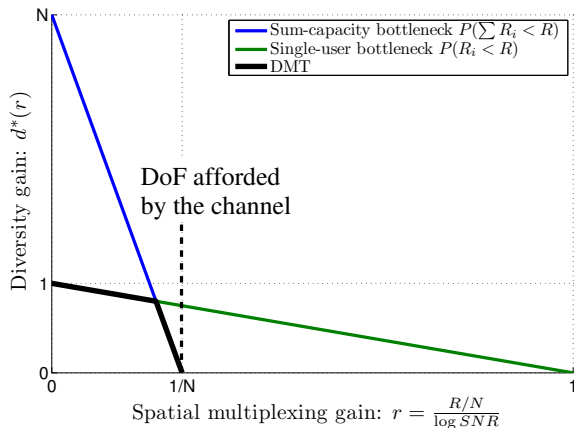
- Simple MAC transmission protocol
  - ▶ Receiver learns channel gains of active users
  - ▶ Calculates Symmetric capacity and notifies transmitters to each transmit at rate  $R/N$  where  $R < C_{\Sigma\text{-sym}}$ :
    - ★ Trivial rate allocation
    - ★ Minimal feedback
- Rayleigh open-loop outage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

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

## Shouldn't be too bad...

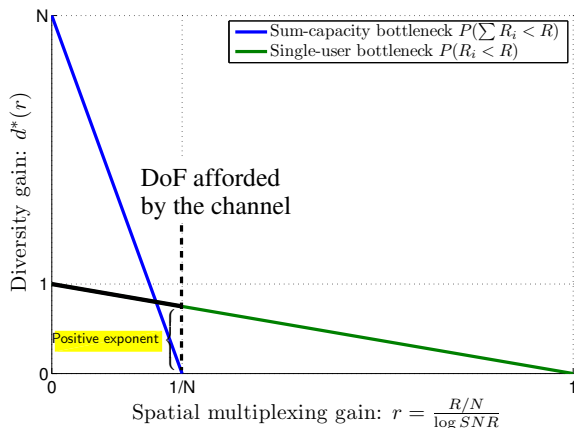


## Shouldn't be too bad...



# Moral 1 from the Symmetric MAC-DMT

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



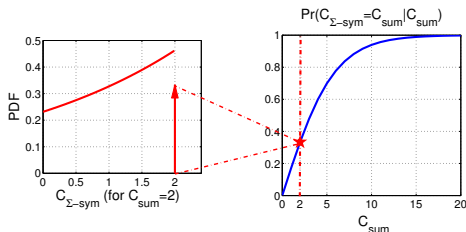
# Bottom Line - Two-User Rayleigh MAC

## Theorem 1

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

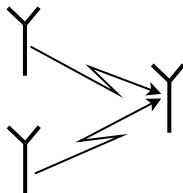
$$\Pr(C_{\Sigma\text{-sym}} < R | C_{\text{sum}}) = 2 \cdot \frac{2^{R/2} - 1}{2^{C_{\text{sum}}} - 1}; \quad 0 \leq R \leq C_{\text{sum}}$$

$$\begin{aligned} \Pr(C_{\Sigma\text{-sym}} = C_{\text{sum}} | C_{\text{sum}}) &= 1 - \Pr(C_{\Sigma\text{-sym}} < C_{\text{sum}} | C_{\text{sum}}) \\ &= 1 - 2 \cdot \frac{2^{C_{\text{sum}}/2} - 1}{2^{C_{\text{sum}}} - 1} \end{aligned}$$





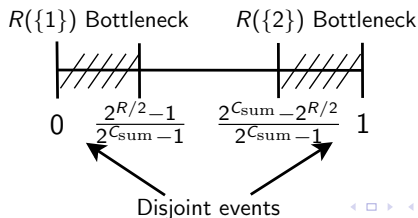
# Sketch of Proof: Two-User Rayleigh MAC



- $h_i \sim \mathcal{CN}(0, \text{SNR})$  and i.i.d  $\Rightarrow |h_i|^2 \sim \exp(\text{SNR})$
- Normalize:  $u_i = \frac{1}{\sqrt{2^{C_{\text{sum}}}-1}} h_i$
- **Given**  $C_{\text{sum}}$ 
  - ▶  $|u_1|^2 + |u_2|^2 = 1 \Rightarrow$  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

- $C_{\Sigma\text{-sym}} = N \min_{S \subseteq \{1, \dots, N\}} R(\{S\}) = N \min_{S \subseteq \{1, \dots, N\}} \frac{1}{|S|} \log \left( 1 + \sum_{i \in S} |h_i|^2 \right)$
- Two users, given  $C_{\text{sum}}$ :  $C_{\Sigma\text{-sym}} = \min(2R(\{1\}), 2R(\{2\}), \cancel{C_{\text{sum}}})$
- $R(\{i\}) = \log \left( 1 + |u_i|^2 (2^{C_{\text{sum}}} - 1) \right)$
- $\Pr(C_{\Sigma\text{-sym}} < R | C_{\text{sum}}) =$   
 $\Pr \left( |u_1|^2 < \frac{2^{R/2} - 1}{2^{C_{\text{sum}} - 1}} \right) + \Pr \left( |u_1|^2 > \frac{2^{C_{\text{sum}} - 2^{R/2}}}{2^{C_{\text{sum}} - 1}} \right)$
- $\Rightarrow \Pr(C_{\Sigma\text{-sym}} < R | C_{\text{sum}}) = 2 \frac{2^{R/2} - 1}{2^{C_{\text{sum}} - 1}}$



# General $N$ : The Bottleneck

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

$$\Pr(R(\{S\}) < R|C_{\text{sum}}) = \Pr\left(\frac{|S|}{N} \log\left(1 + (2^{C_{\text{sum}}} - 1) \sum_{i \in S} |u_i|^2\right) < R \mid \sum |u_i|^2 = 1\right)$$

- ▶ 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)
- $\Rightarrow$  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  $C_{\text{sum}}$ , the outage probability for a set  $S \subseteq \{1, 2, \dots, N\}$  is

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

where  $0 \leq R \leq C_{\text{sum}}$  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

- $\Pr(C_{\Sigma\text{-sym}} < R | C_{\text{sum}}) = \Pr\left(\min_{S \subseteq \{1,2,\dots,N\}} R(\{S\}) < R | C_{\text{sum}}\right)$
- All sets with the same cardinality have the same outage probability
- $P_{\text{out}}(k, R) \triangleq \Pr(R(\{|S| = k\}) < R | C_{\text{sum}})$
- Union bound can be used to bound overall probability

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

For a  $1 \times N$  Rayleigh MAC with sum capacity  $C_{\text{sum}}$ , the outage probability can be bounded as

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

# Upper and Lower Bounds

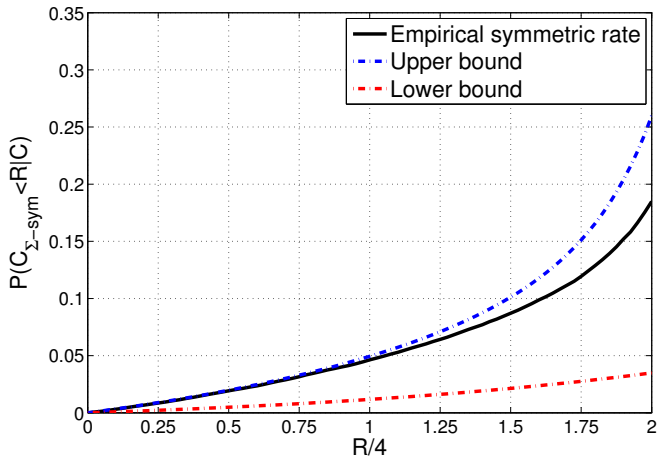


Figure: Bounds vs. Empirical error probability for  $1 \times 4$  channel with  $C_{\text{sum}}/4 = 2$

# Practical Scheme (NOMA): Two-User Example

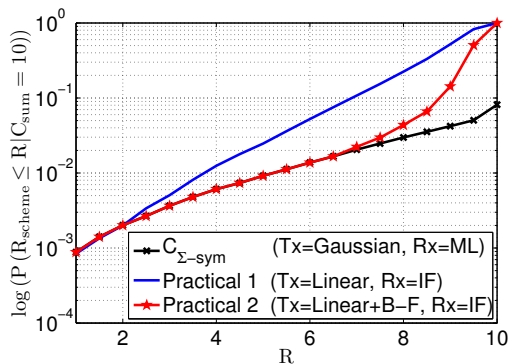


Figure: ML vs. IF for a two-user i.i.d. Rayleigh fading MAC with  $C_{\text{sum}} = 10$ .

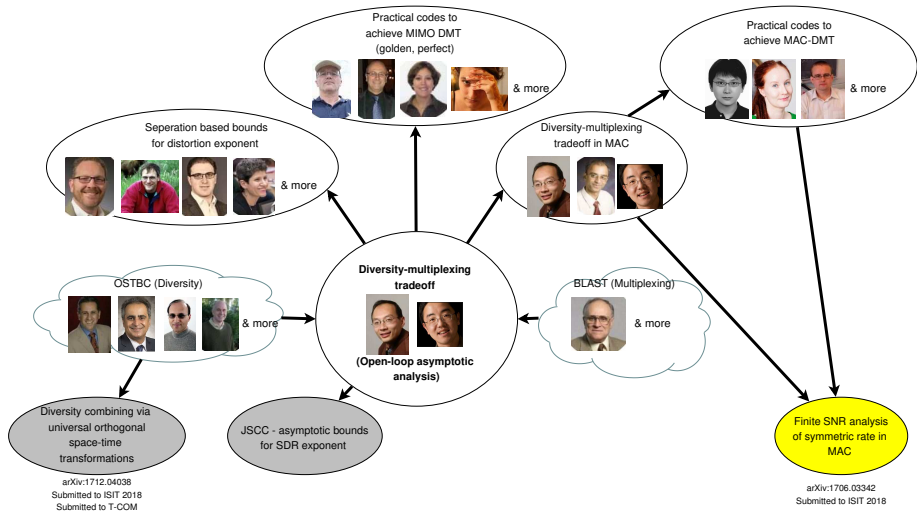
## Ingredients

### Integer-forcing (Rx)

- Beyond this talk
- All users use the same linear code
- Not sufficient

### MAC-DMT (Tx)

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



Slides available in <https://domanovi.github.io/>