

# Online Capacity Maximization in Wireless Networks

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Joint work with  
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# Problem

## Situation

- Wireless network
- Request  $(s_i, r_i)$  from sender  $s_i$  to receiver  $r_i$ :

$$s_i \longrightarrow r_i$$

Objective: Accept as many requests as possible

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## Situation

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- Request  $(s_i, r_i)$  between  $s_i$  and  $r_i$ :

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- Request  $(s_i, r_i)$  between  $s_i$  and  $r_i$ :

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Competitive Analysis: Acceptance based on previous requests only

# Wireless networks

Characteristics:

$(s_i, r_i)$

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- Receiver can interpret the signal if SINR (signal-to-interference-plus-noise-ratio) constraint holds:

$$\frac{S}{N + I} \geq \beta$$

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- Metric space:  $d$ -dimensional Euclidean space

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# Power Assignments

$$S = \frac{(s_i, r_i) P}{\text{distance}_{ij}^\alpha}$$
$$\frac{S}{N+1} \geq \beta$$

Power Assignments:

- Distance-based:  $P = \phi(\text{distance}_{ij})$
- Polynomial:  $P = \text{distance}_{ij}^{r\alpha}$  for a fix  $r \in \mathbb{R}$ 
  - $r = 0$  (uniform)
  - $r = 1$  (linear)
  - $r = \frac{1}{2}$  (square root)

# Greedy Algorithm

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## Algorithm 1 (GREEDY)

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Accept if SINR constraint still holds at each accepted receiver

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sender → receiver

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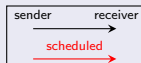
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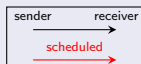
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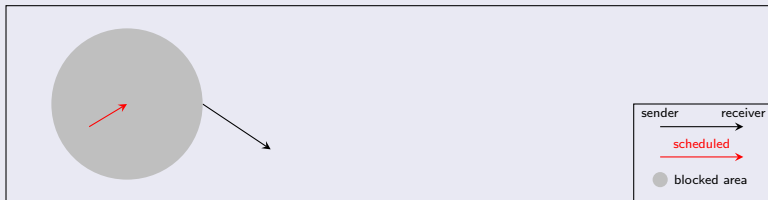
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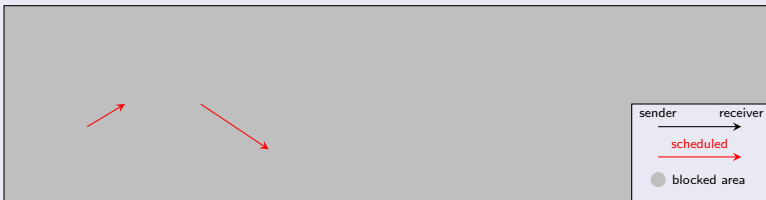
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→ Conflict between correctness and competitive ratio

→ Worst-case scenario

# Safe Distance Algorithm

$$P = \text{distance}_{ij}^{r_i \alpha}$$

$$\text{distance}_{ij} \in [1, \Delta]$$

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**Algorithm 2 (SAFE-DISTANCE)**

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### Theorem

SAFE-DISTANCE has competitive ratio

$$O(\Delta^d), \quad \text{if } r \in [0, 1]$$

$$O(\Delta^{d \cdot \max\{r, 1-r\}}), \quad \text{else}$$



# Square-root power assignment

$$P = \text{distance}_{ij}^{r\alpha}$$
$$\text{distance}_{ij} \in [1, \Delta]$$

$$r \in (0, 1)$$

- Short requests can be scheduled in longer ones
- Use  $m$  length classes

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## Theorem

MULTI-CLASS SAFE-DISTANCE has competitive ratio

$$O\left(\Delta^{\frac{d}{2} + \epsilon}\right)$$

for any constant  $\epsilon > 0$

# Lower bound

$$P = \text{distance}_{ij}^{r\alpha}$$

$$\text{distance}_{ij} \in [1, \Delta]$$

## Theorem

Every deterministic online algorithm has at least competitive ratio

$$\Omega\left(\Delta^{d \cdot \max\{r, 1-r\}}\right)$$

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## Proof (Idea).

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- Online Algorithm must accept the first request
- It blocks some area around sender and receiver
- Offline Algorithm can place requests into these blocked areas



# Channels

Use multiple channels

→ Each for a specific request length

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## Corollary

MULTI-CLASS SAFE-DISTANCE with  $\log \Delta$ -channels has competitive ratio

$$O(\log \Delta)$$

# Randomization

$$P = \text{distance}_{ij}^{r,\alpha}$$
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RANDOM SAFE-DISTANCE has competitive ratio

$$O(\log \Delta)$$

# Conclusion

- GREEDY is not competitive
- Usage of safe distance leads to local effect
- MULTI-CLASS SAFE-DISTANCE approaches lower bound
- RANDOM SAFE-DISTANCE is  $O(\log \Delta)$ -competitive

# Conclusion

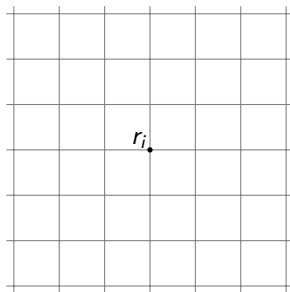
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**Thank you for your attention**

# Safe distance computation

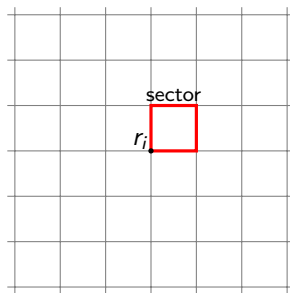
$r_i$

# Safe distance computation

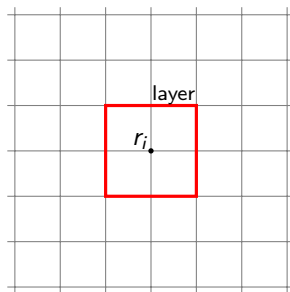




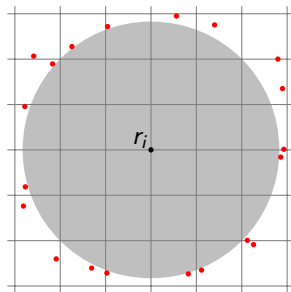
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