

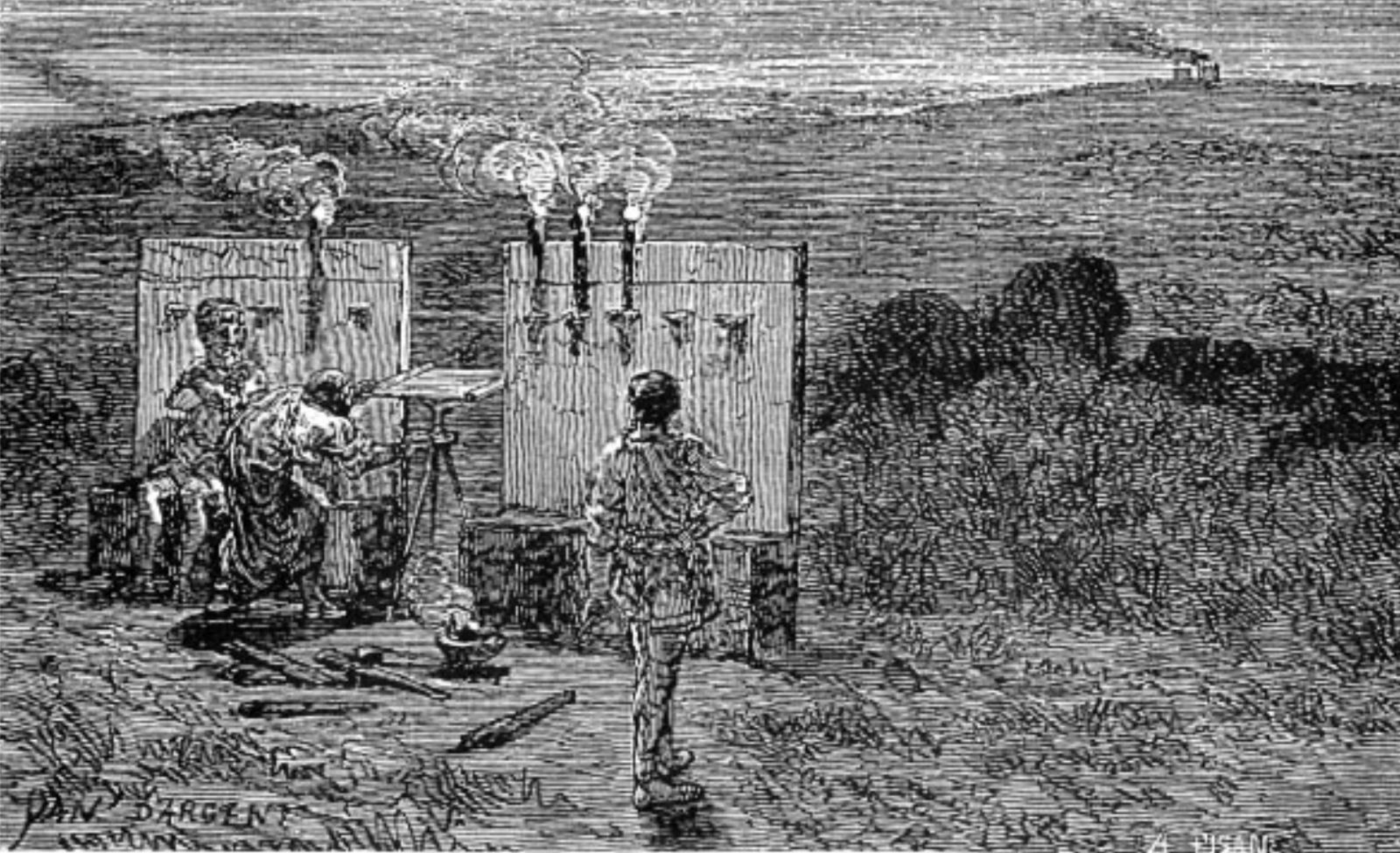
# *Wireless Algorithms*



*Roger Wattenhofer*

... an oxymoron?

...as old as society



more recently?



## **Wireless Communication**

EE, Physics

Maxwell Equations

Simulation, Testing

'Scaling Laws'

## **Network Algorithms**

CS, Applied Math

[Geometric] Graphs

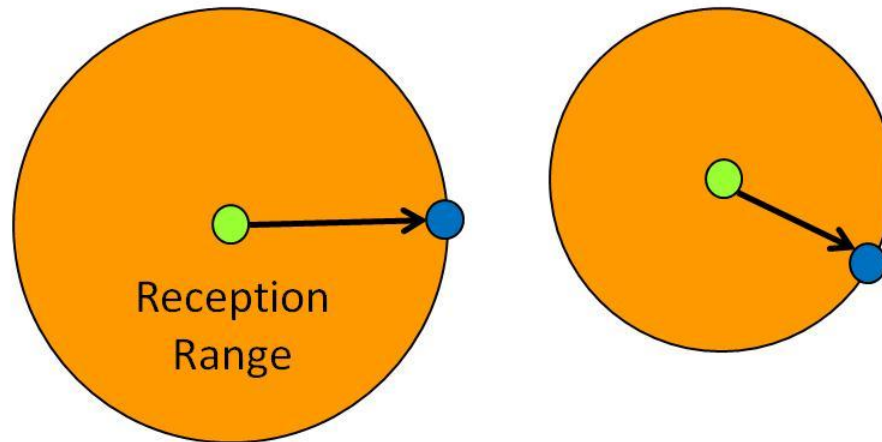
Worst-Case Analysis

Any-Case Analysis

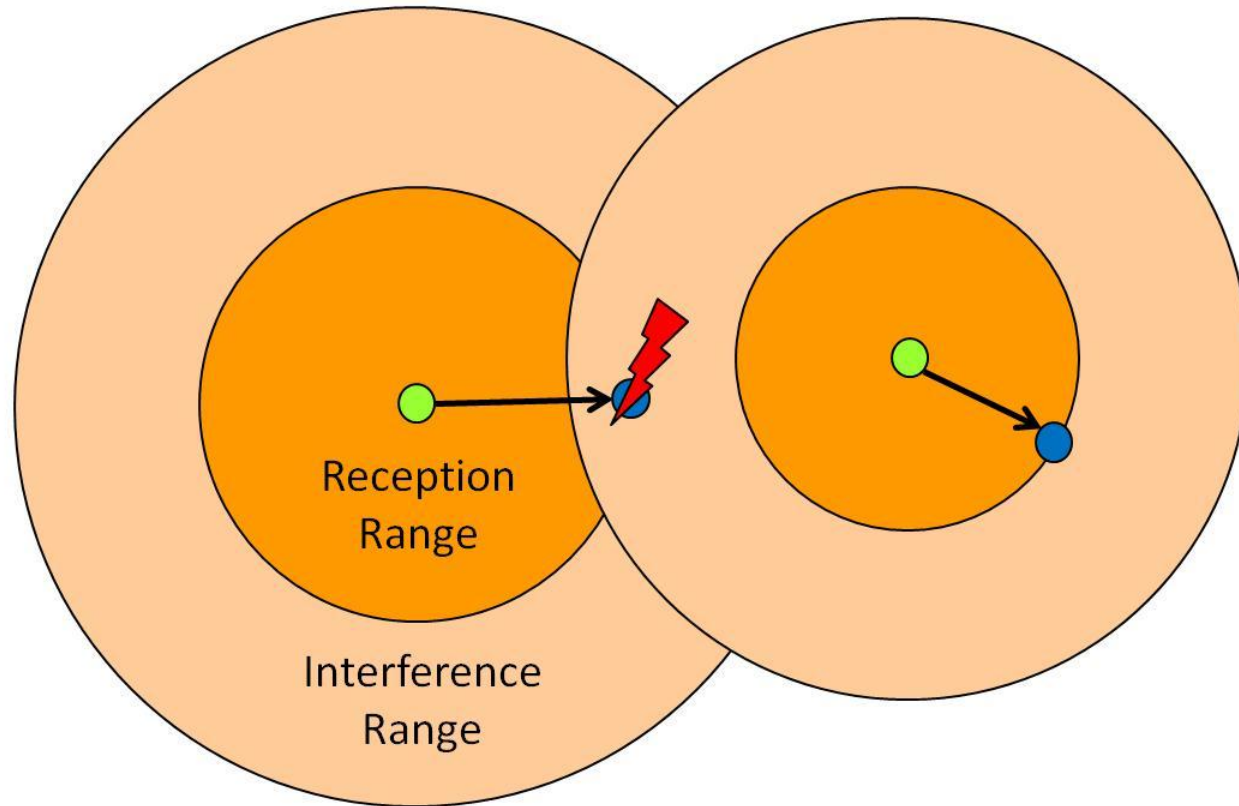
# CS Models: e.g. Disk Model (Protocol Model)



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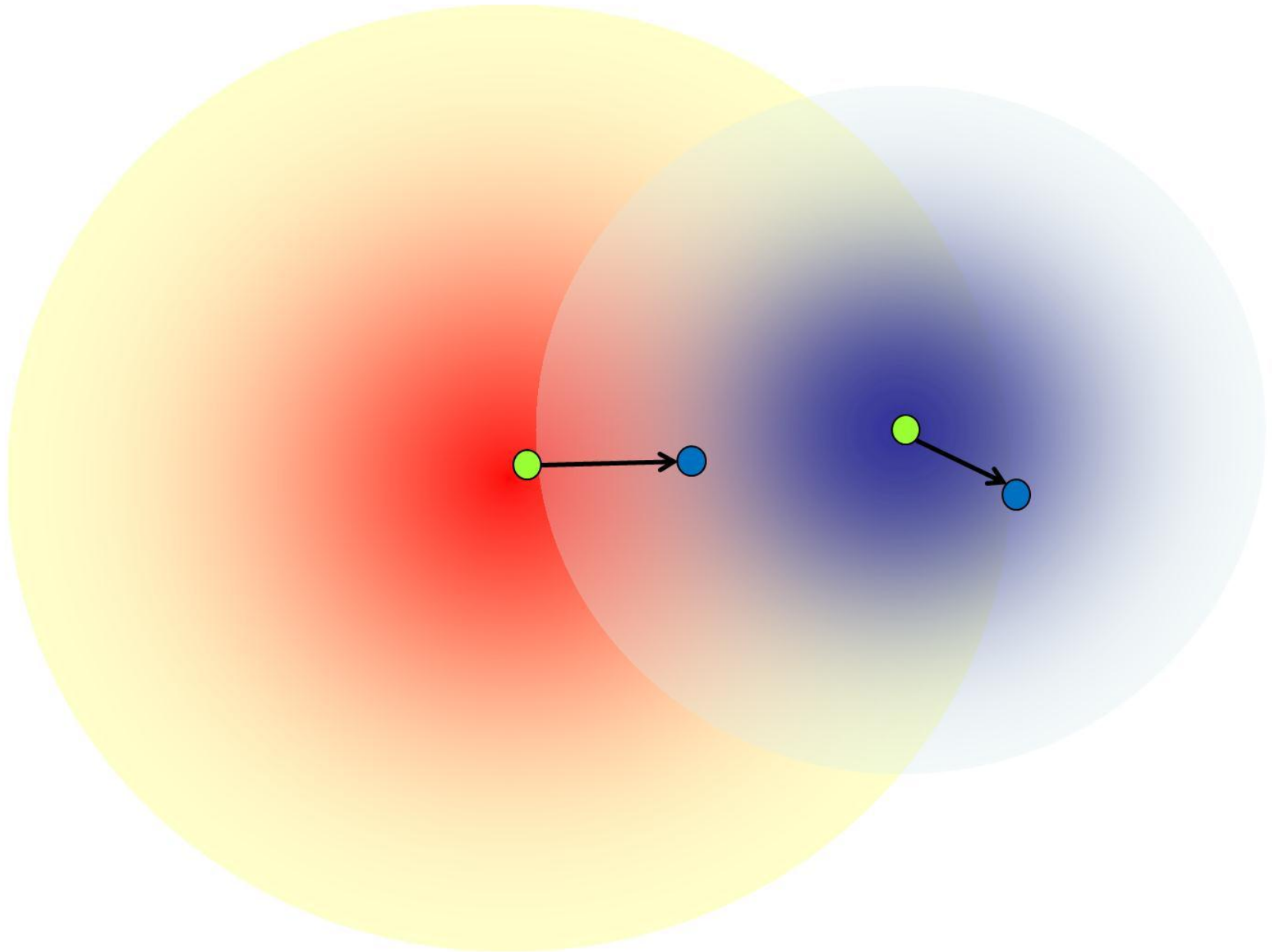




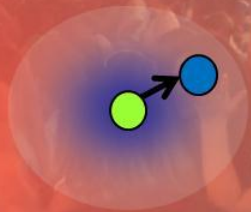




EE Models: e.g. SINR Model (Physical Model)



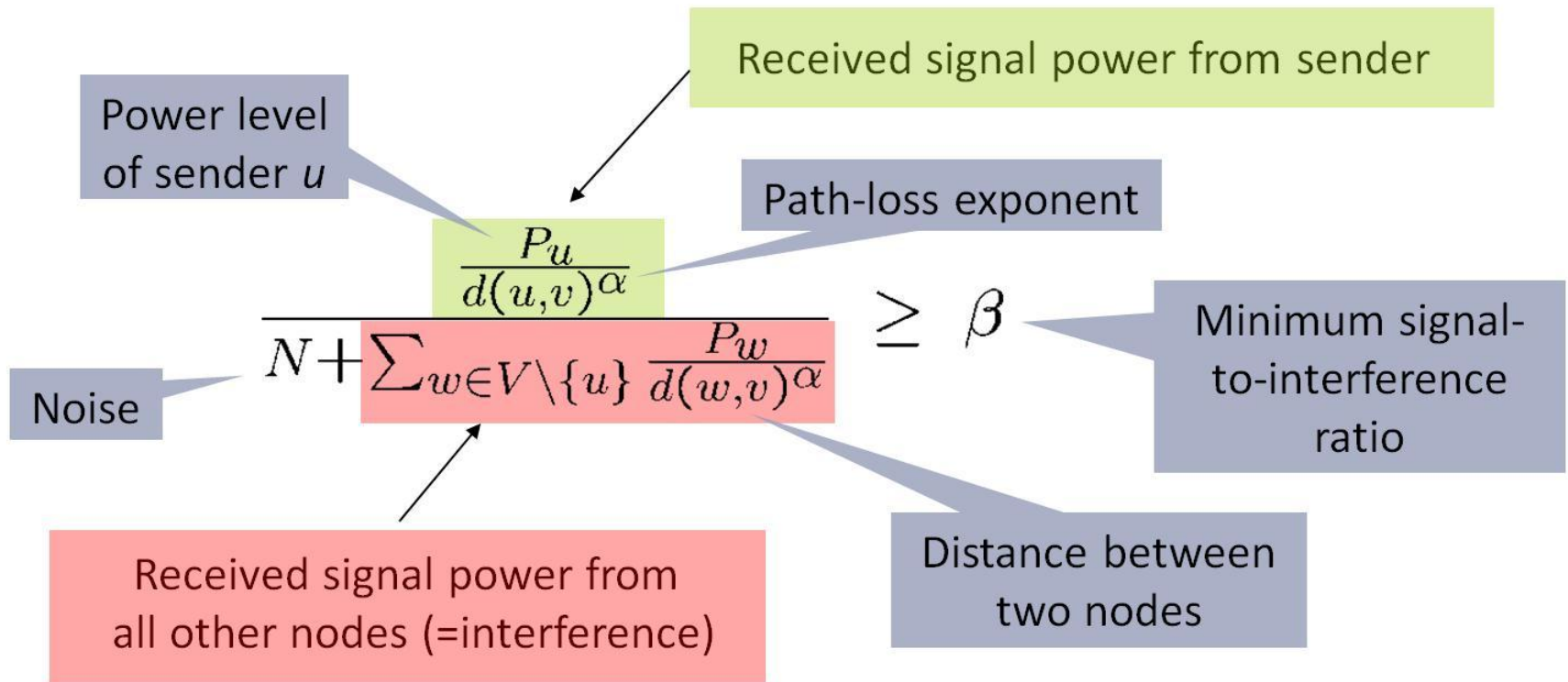




## Signal-To-Interference-Plus-Noise Ratio (SINR) Formula

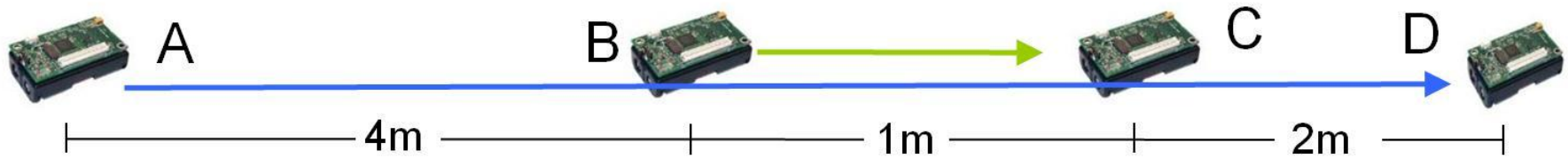
$$\frac{\frac{P_u}{d(u,v)^\alpha}}{N + \sum_{w \in V \setminus \{u\}} \frac{P_w}{d(w,v)^\alpha}} \geq \beta$$

# Signal-To-Interference-Plus-Noise Ratio (SINR) Formula



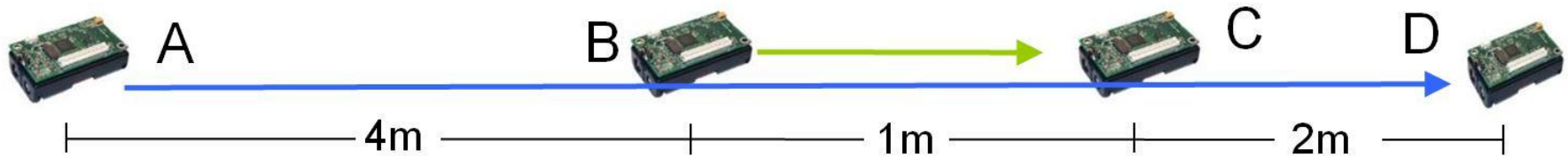


## Example: Protocol vs. Physical Model

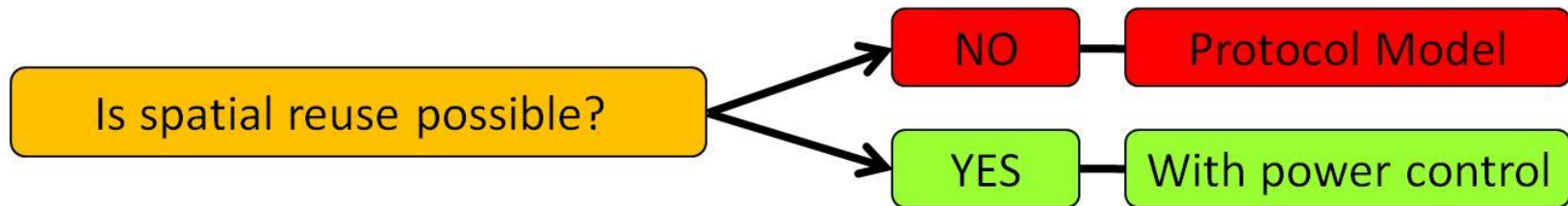


Assume a **single frequency** (and no fancy decoding techniques!)

# Example: Protocol vs. Physical Model





Assume a **single frequency** (and no fancy decoding techniques!)



Let  $\alpha=3$ ,  $\beta=3$ , and  $N=10\text{nW}$

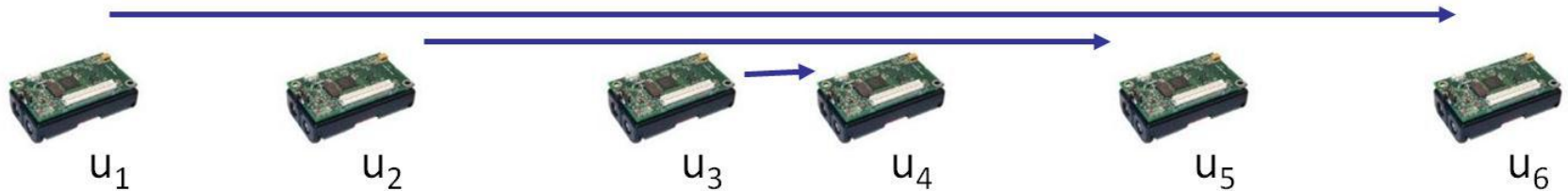
Transmission powers:  $P_B = -15\text{ dBm}$  and  $P_A = 1\text{ dBm}$

SINR of A at D: 
$$\frac{1.26\text{mW}/(7\text{m})^3}{0.01\mu\text{W} + 31.6\mu\text{W}/(3\text{m})^3} \approx 3.11 \geq \beta$$
 

SINR of B at C: 
$$\frac{31.6\mu\text{W}/(1\text{m})^3}{0.01\mu\text{W} + 1.26\text{mW}/(5\text{m})^3} \approx 3.13 \geq \beta$$
 

# This works in practice!

... even with very simple hardware (sensor nodes)



Time for transmitting 20'000 packets:

	Time required	
	standard MAC	"SINR-MAC"
Node $u_1$	721s	267s
Node $u_2$	778s	268s
Node $u_3$	780s	270s

	Messages received	
	standard MAC	"SINR-MAC"
Node $u_4$	19999	19773
Node $u_5$	18784	18488
Node $u_6$	16519	19498

**Speed-up is almost a factor 3**

# The Capacity of a Wireless Network

# Measures for Capacity

## Throughput capacity

- Number of packets successfully delivered per time
- Dependent on the traffic pattern
- *E.g.: What is the maximum achievable rate, over all protocols, for a random node distribution and a random destination for each source?*

## Transport capacity

- A network transports one **bit-meter** when one bit has been transported a distance of one meter.
- *What is the maximum achievable rate, over all node locations, and all traffic patterns, and all protocols?*

## Convergecast capacity

- *How long does it take to get information from all nodes to a sink*

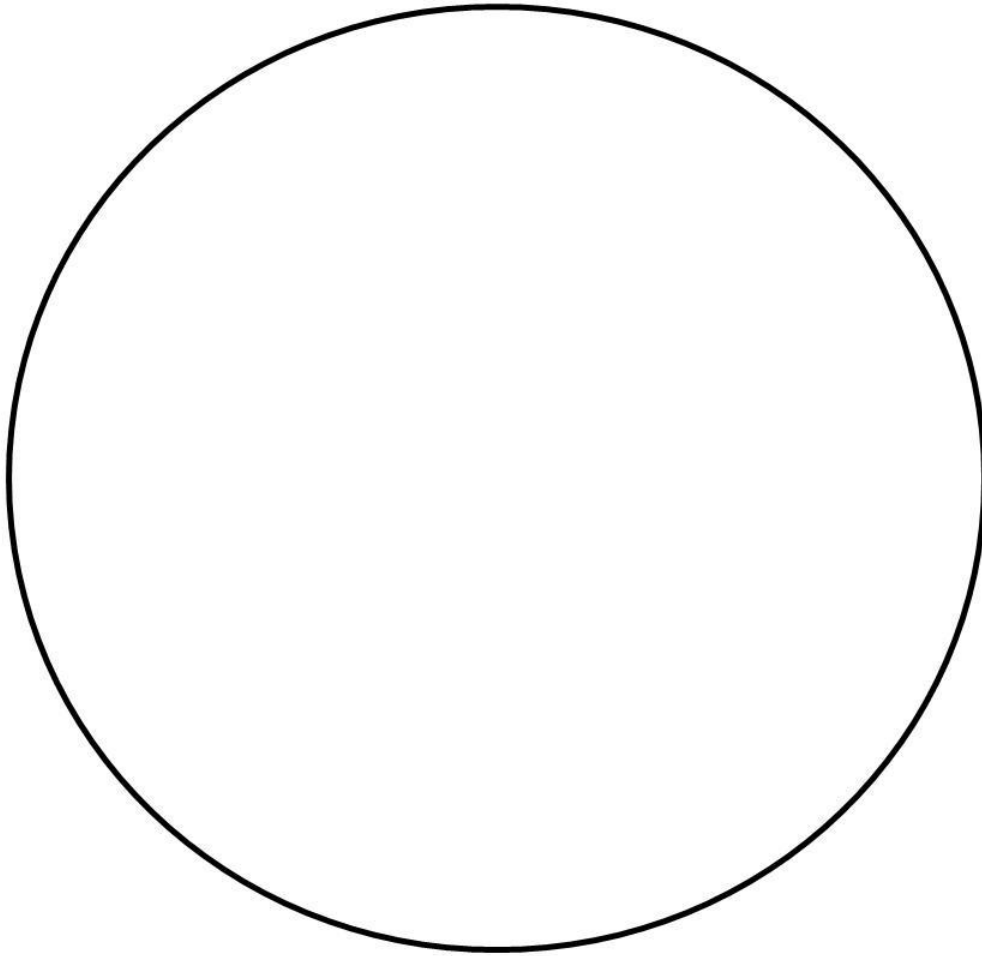
Many more...



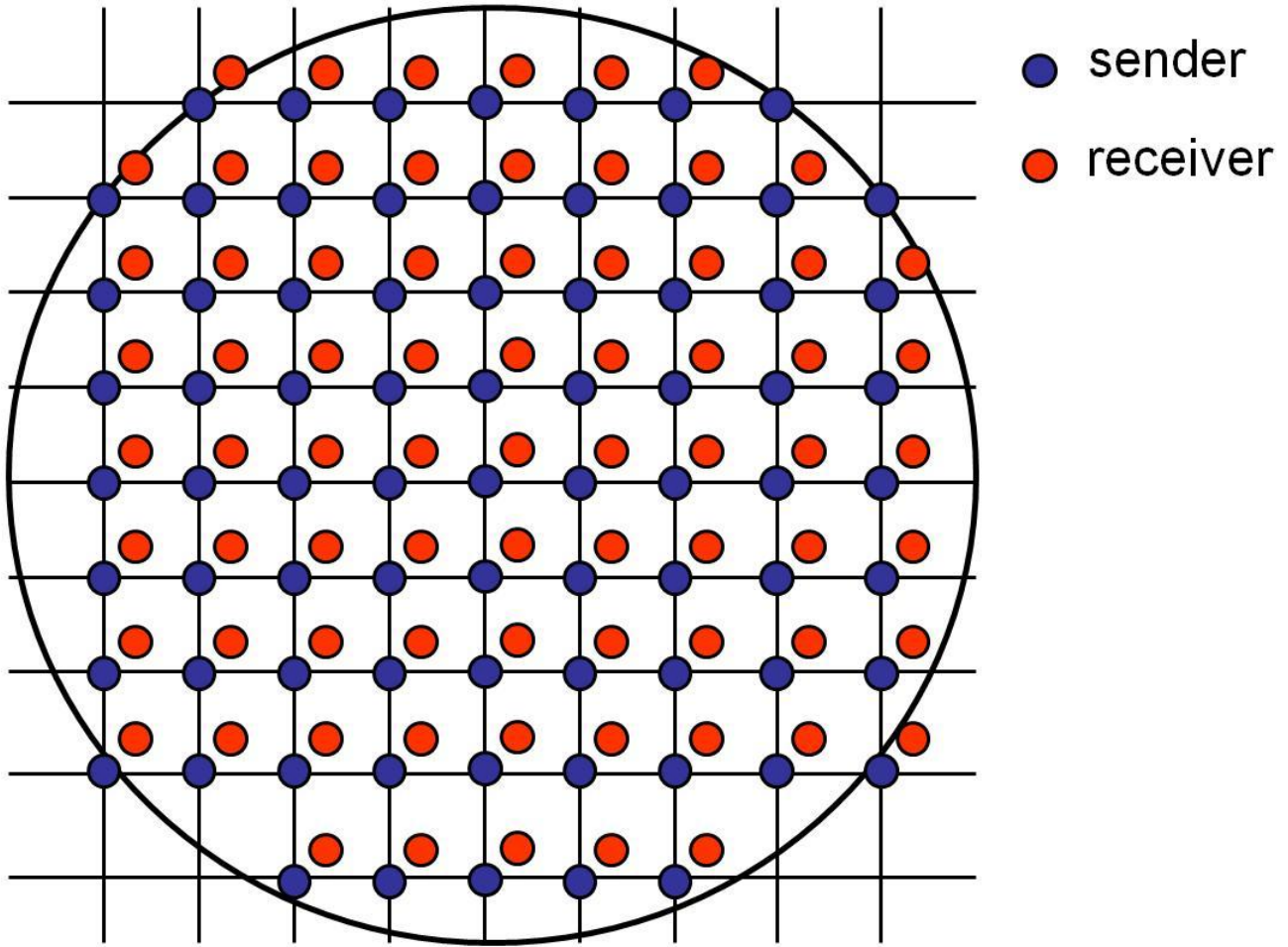
# Transport Capacity

- $n$  nodes are arbitrarily located in a unit disk.
- We adopt the [protocol model](#) with  $R=2$ , that is a transmission is successful if and only if the sender is at least a factor 2 closer than any interfering transmitter. In other words, each node transmits with the same power, and transmissions are in synchronized slots.
- [Quiz: What configuration and traffic pattern will yield the highest transport capacity?](#)
- Idea: Distribute  $n/2$  senders uniformly in the unit disk. Place the  $n/2$  receivers just close enough to senders so as to satisfy the threshold.

## Transport Capacity: Example

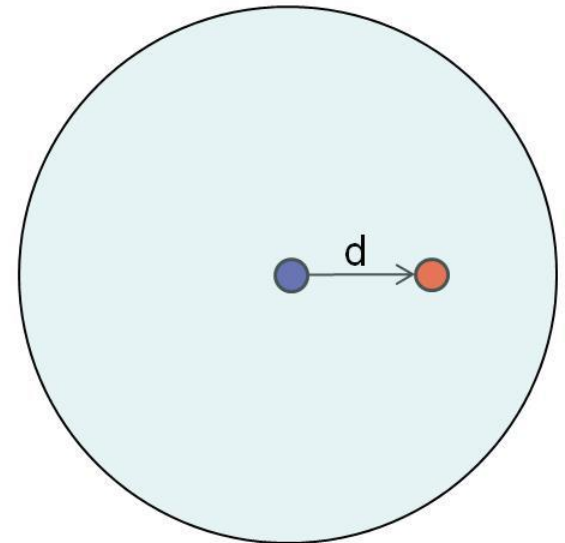


# Transport Capacity: Example



# Transport Capacity: Understanding the example

- Sender-receiver distance is  $\Theta(1/\sqrt{n})$ . Assuming channel bandwidth  $W$  [bits], transport capacity is  $\Theta(W\sqrt{n})$  [bit-meter], or per node:  $\Theta(W/\sqrt{n})$  [bit-meter]
- Can we do better by placing the source-destination pairs more carefully? No, having a sender-receiver pair at distance  $d$  inhibits another receiver within distance up to  $2d$  from the sender. In other words, it kills an area of  $\Theta(d^2)$ .
- We want to maximize  $n$  transmissions with distances  $d_1, d_2, \dots, d_n$  given that the total area is less than a unit disk. This is maximized if all  $d_i = \Theta(1/\sqrt{n})$ . So the example was asymptotically optimal.
  - BTW, a fun [open](#) geometry problem: Given  $k$  circles with total area 1, can you always fit them in a circle with total area 2?



## More capacities...

- The throughput capacity of an  $n$  node **random network** is  $\Theta\left(\frac{W}{\sqrt{n \log n}}\right)$
- There exist constants  $c$  and  $c'$  such that  $\lim_{n \rightarrow \infty} \Pr\left[c \frac{W}{\sqrt{n \log n}} \text{ is feasible}\right] = 1$   
 $\lim_{n \rightarrow \infty} \Pr\left[c' \frac{W}{\sqrt{n \log n}} \text{ is feasible}\right] = 0$
- Transport capacity:
  - Per node transport capacity decreases with  $\frac{1}{\sqrt{n}}$
  - Maximized when nodes transmit to neighbors
- Throughput capacity:
  - For random networks, decreases with  $\frac{1}{\sqrt{n \log n}}$
  - Near-optimal when nodes transmit to neighbors
- In one sentence: **local communication is better**

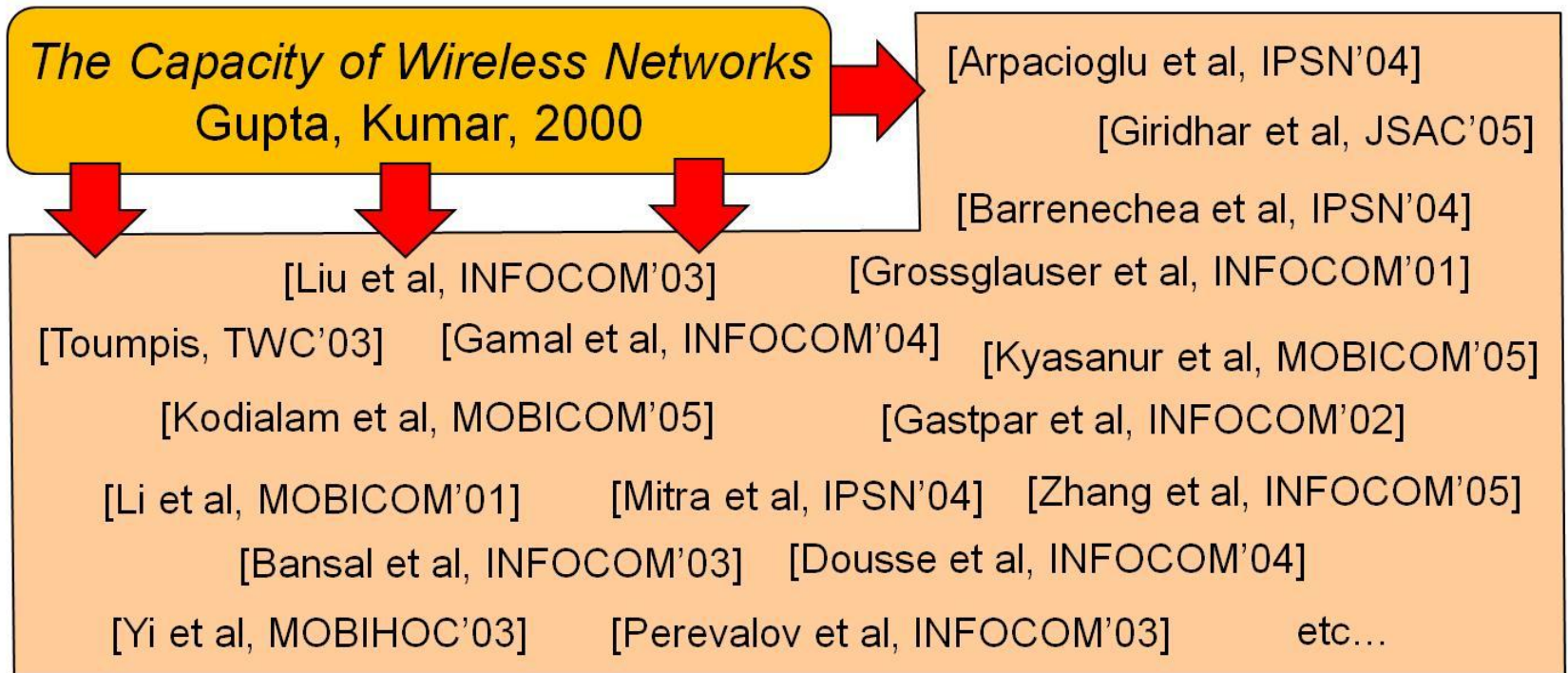


## Even more capacities...

- Similar claims hold in the physical (SINR) model as well...
- There are literally thousands of results on capacity, e.g.
  - on random destinations
  - on power-law traffic patterns
  - communication through relays
  - communication in mobile networks
  - channel broken into subchannels
  - etc.

# Practical relevance?

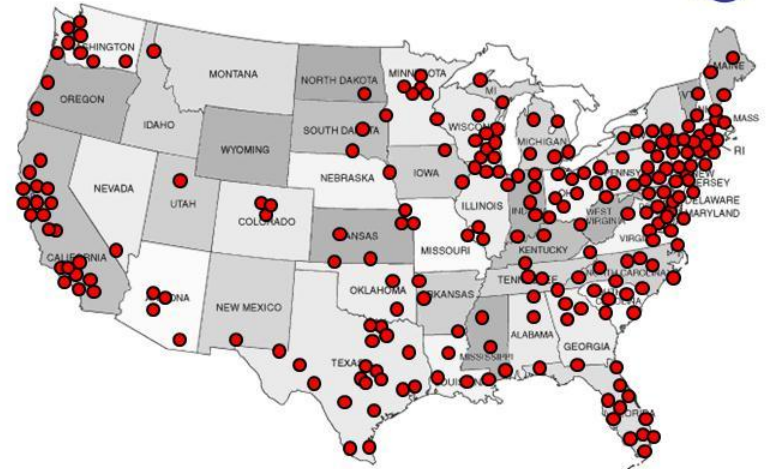
- Efficient access to media, i.e. **MAC** layer!
- This (and related) problem is studied theoretically:



# Network Topology?

- All these capacity studies make very **strong assumptions** on node deployment, topologies
  - randomly, uniformly distributed nodes
  - nodes placed on a grid
  - etc.

What if a network looks differently...?



# 'Scaling Laws'

## "Classic" Capacity

How much information can be transmitted in nice networks?



# 'Scaling Laws'

## "Classic" Capacity

How much information can be transmitted in **nice** networks?



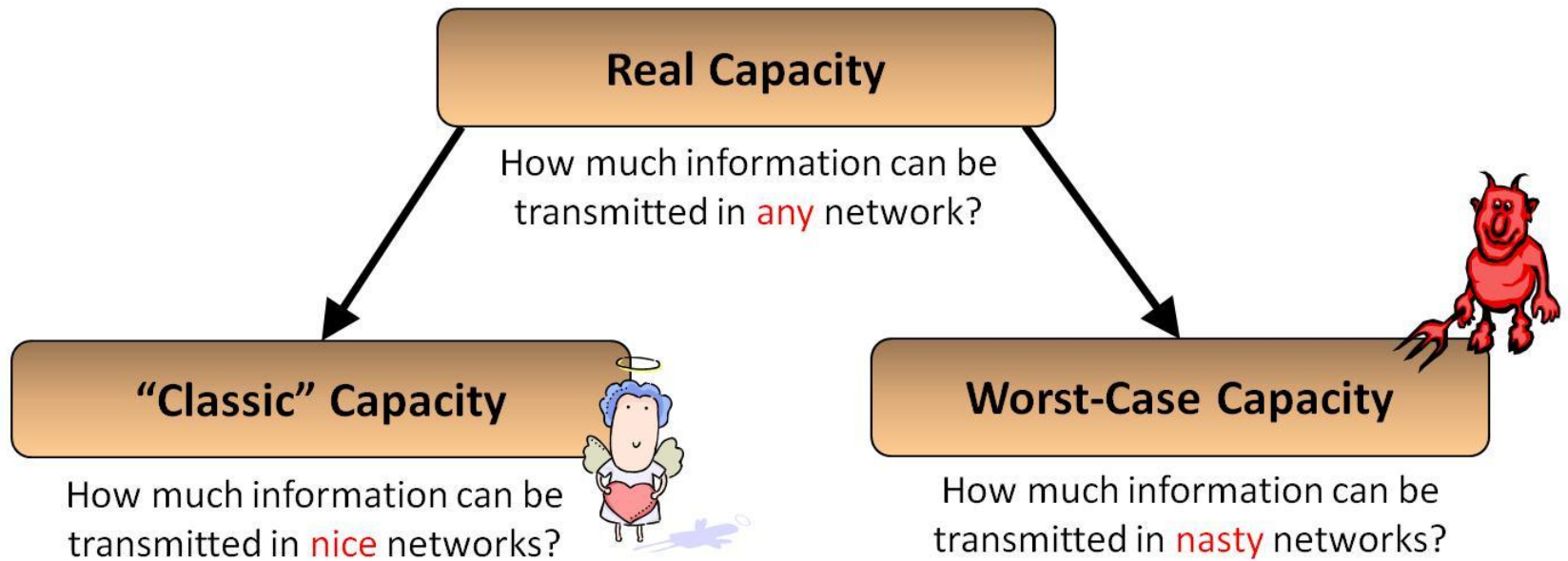
## Worst-Case Capacity

How much information can be transmitted in **nasty** networks?



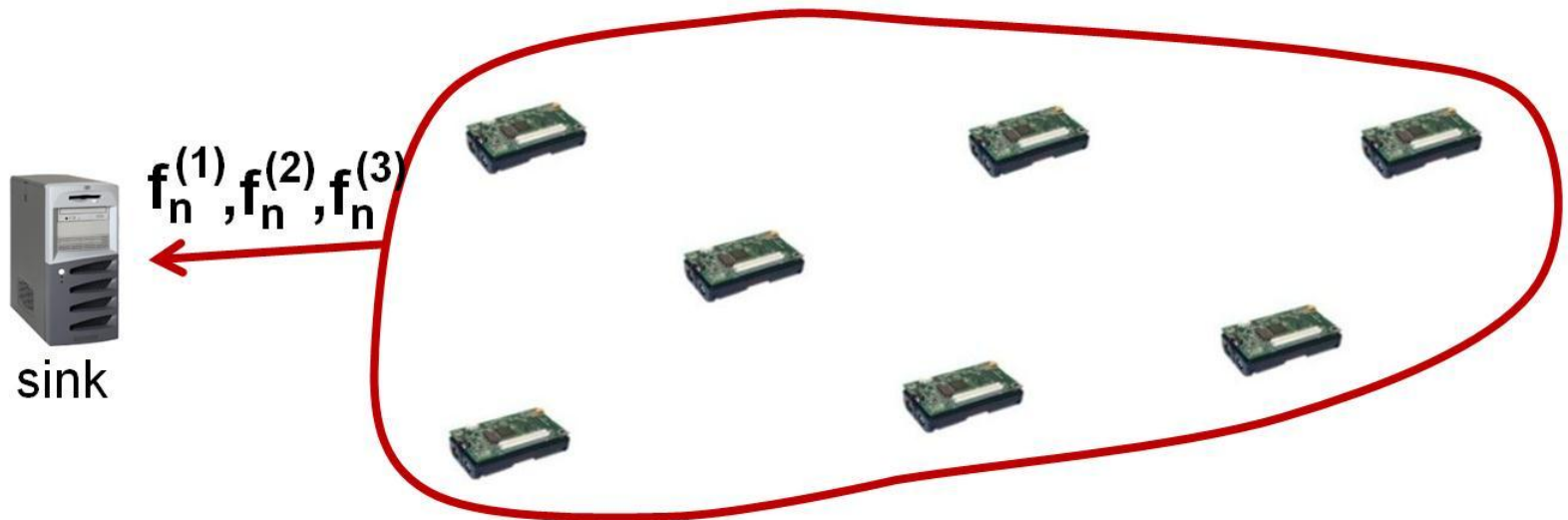


# 'Scaling Laws'



# Convergecast Capacity in Wireless Networks

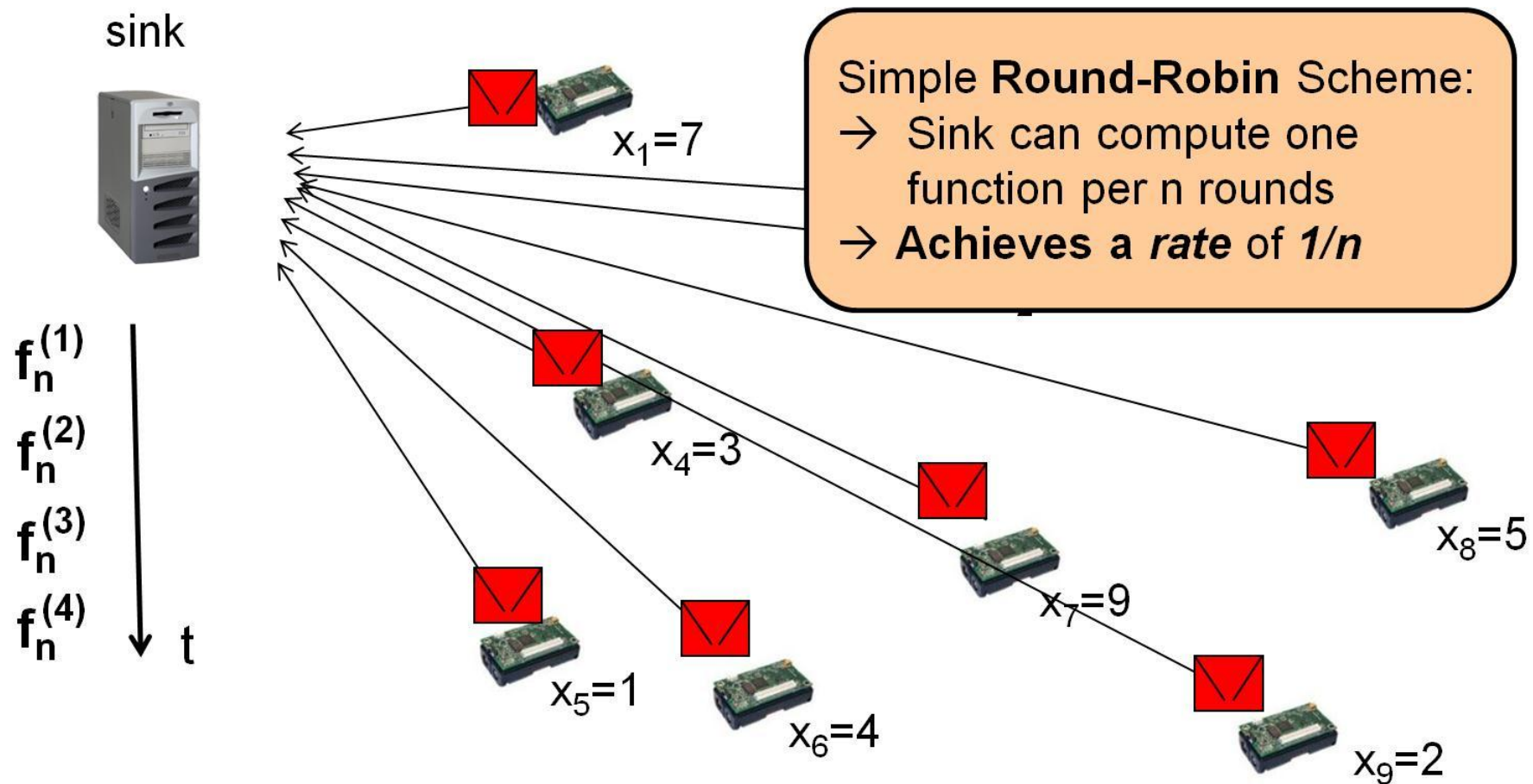
- Data gathering & aggregation
  - Classic application of sensor networks
  - Sensor nodes periodically sense environment
  - Relevant information needs to be transmitted to **sink**
- Functional Capacity of Sensor Networks
  - Sink periodically wants to compute a **function**  $f_n$  of sensor data
  - At what **rate** can this function be computed?



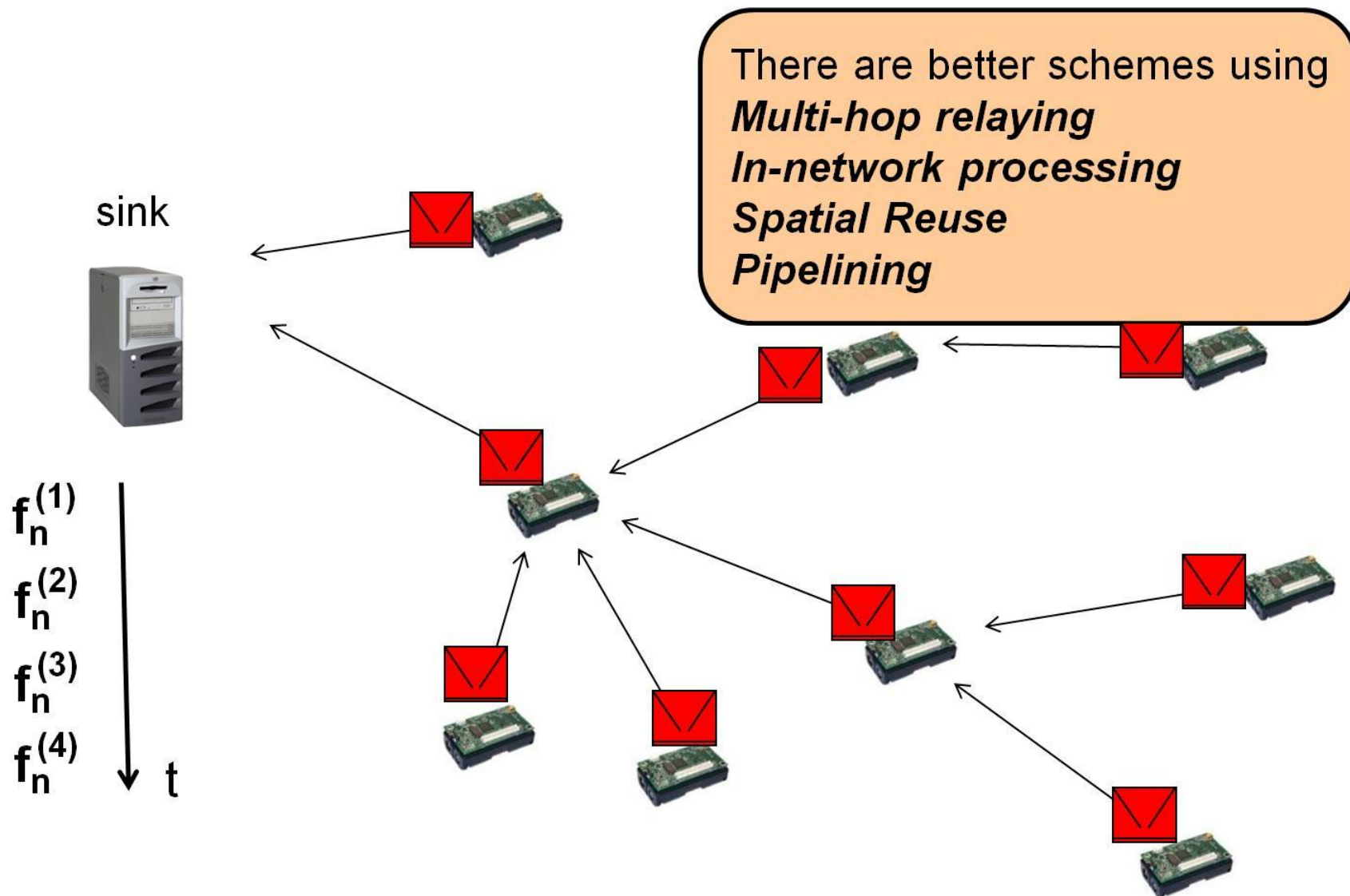
# Convergecast Capacity in Wireless Networks

Example: simple **round-robin scheme**

→ Each sensor reports its results directly to the root one after another



# Convergecast Capacity in Wireless Networks



# Convergecast Capacity in Wireless Networks

At what **rate** can sensors transmit data to the sink?  
Scaling-laws  $\rightarrow$  how does rate decrease as  $n$  increases...?

$$\Theta(1/n)$$

$$\Theta(1/\sqrt{n})$$

$$\Theta(1/\log n)$$

$$\Theta(1)$$



# Convergecast Capacity in Wireless Networks

At what **rate** can sensors transmit data to the sink?  
Scaling-laws  $\rightarrow$  how does rate decrease as  $n$  increases...?

$$\Theta(1/n)$$

$$\Theta(1/\sqrt{n})$$

$$\Theta(1/\log n)$$

$$\Theta(1)$$

Answer depends on:

Function to be computed

Coding techniques

Network topology

...

Only perfectly compressible functions  
(max, min, avg, ...)

No fancy coding techniques

# Convergecast Capacity in Wireless Networks

Networks Model/Power	Max. rate in arbitrary, worst-case deployment	Max. rate in random, uniform deployment
Protocol Model		
Physical Model (w/ power control)		

# Convergecast Capacity in Wireless Networks

[Moscibroda, W, 2006]

Worst-Case Capacity

[Giridhar, Kumar, 2005]

Best-Case Capacity

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Networks Model/Power	Max. rate in arbitrary, worst-case deployment	Max. rate in random, uniform deployment
Protocol Model	$\Theta(1/n)$	$\Theta(1/\log n)$
Physical Model (w/ power control)	$\Omega(1/\log^3 n)$	$\Omega(1/\log n)$

Exponential gap  
between protocol and  
physical model!

## The Price of Worst-Case Node Placement

- Exponential in protocol model
- Polylogarithmic in physical model  
(almost no worst-case penalty!)

## **Wireless Communication**

EE, Physics

Maxwell Equations

Simulation, Testing

'Scaling Laws'

## **Network Algorithms**

CS, Applied Math

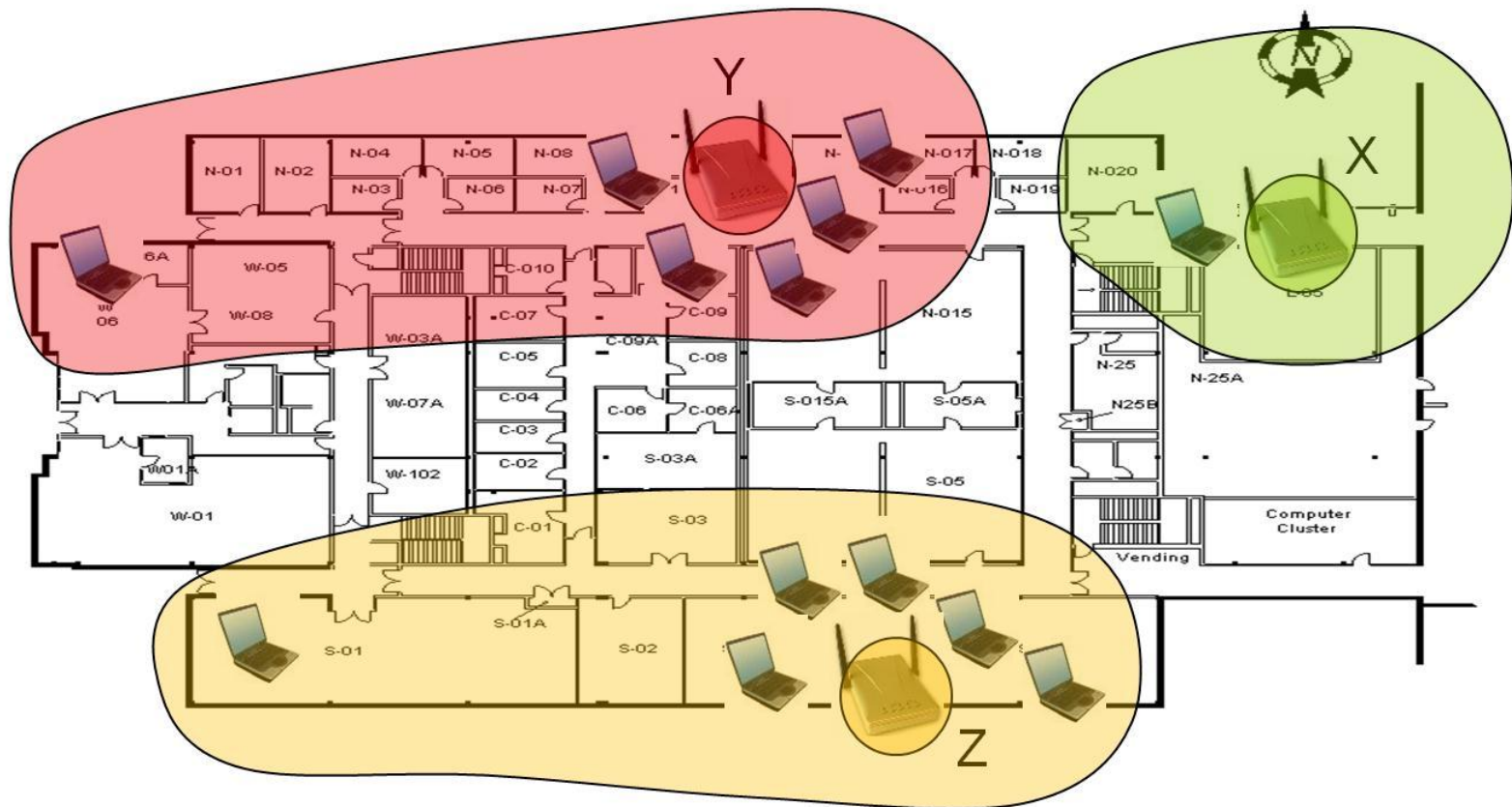
[Geometric] Graphs

Worst-Case Analysis

Any-Case Analysis

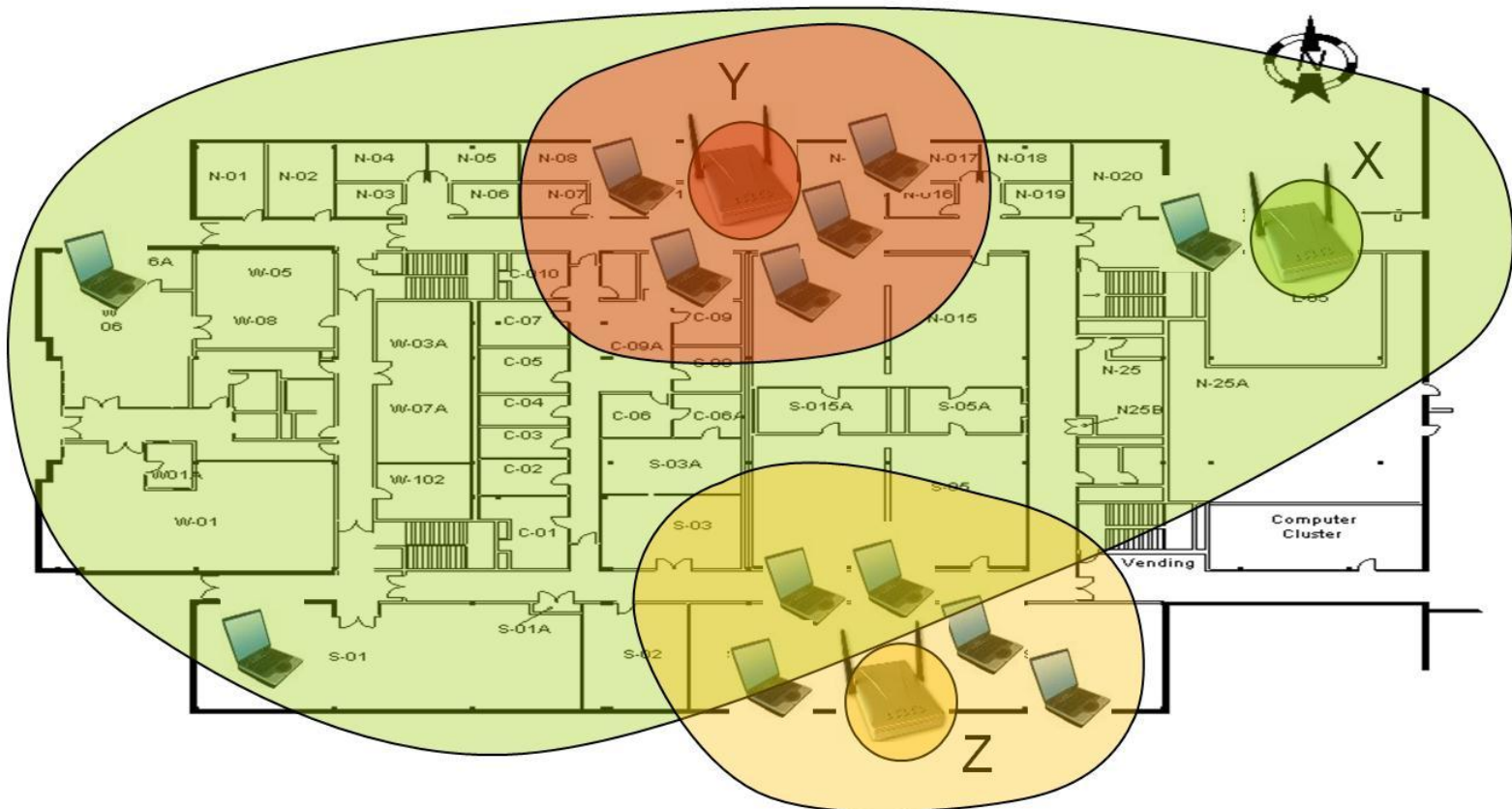
# Possible Application – Hotspots in WLAN

Traditionally: clients assigned to (more or less) closest access point  
→ far-terminal problem → hotspots have less throughput



# Possible Application – Hotspots in WLAN

Potentially better: create hotspots with very high throughput  
Every client outside a hotspot is served by one base station  
→ Better overall throughput – increase in capacity!



## **Wireless Communication**

EE, Physics

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## **Network Algorithms**

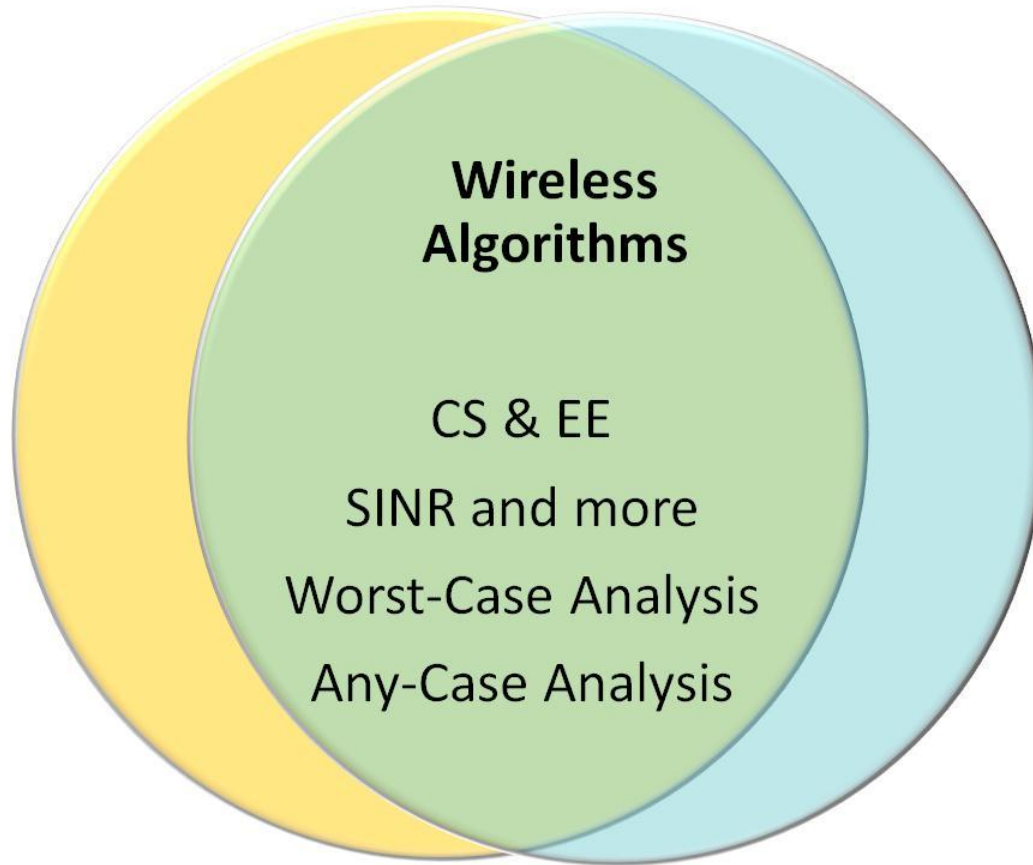
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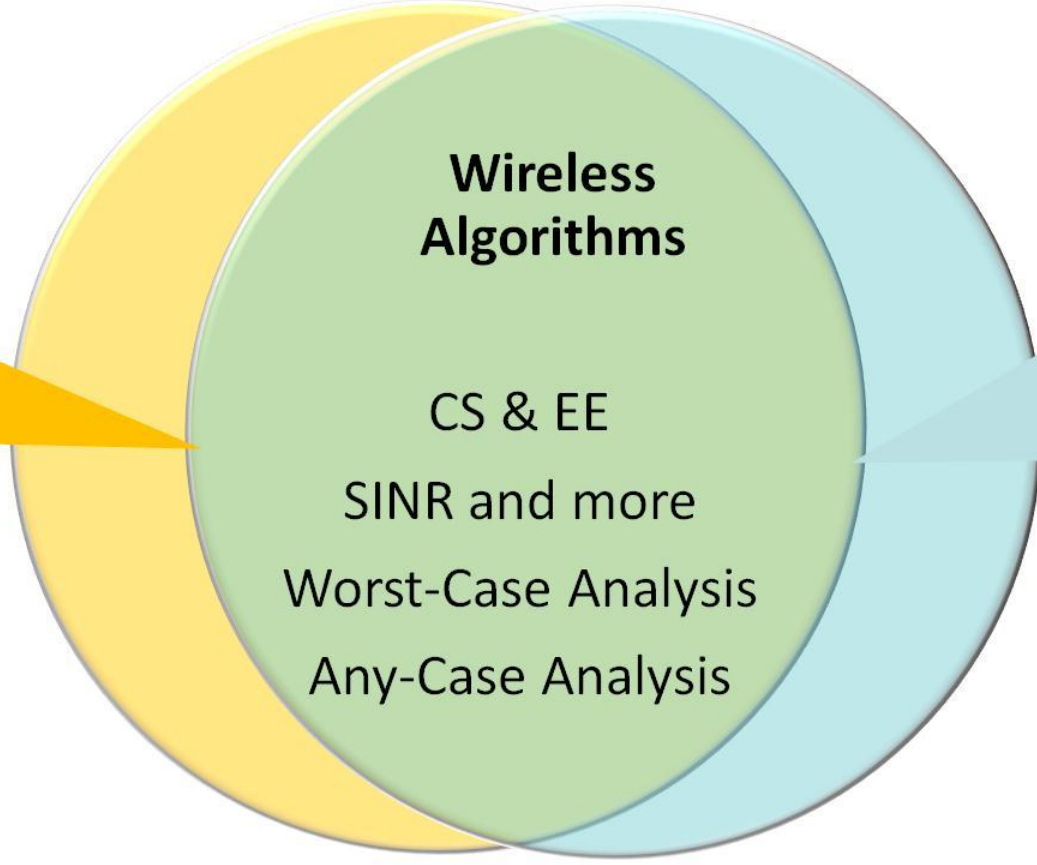
Any-Case Analysis





[Grossglauser, Tse, 2002]

Mobility increases the capacity  
of ad hoc wireless networks



**Wireless  
Algorithms**

CS & EE

SINR and more

Worst-Case Analysis

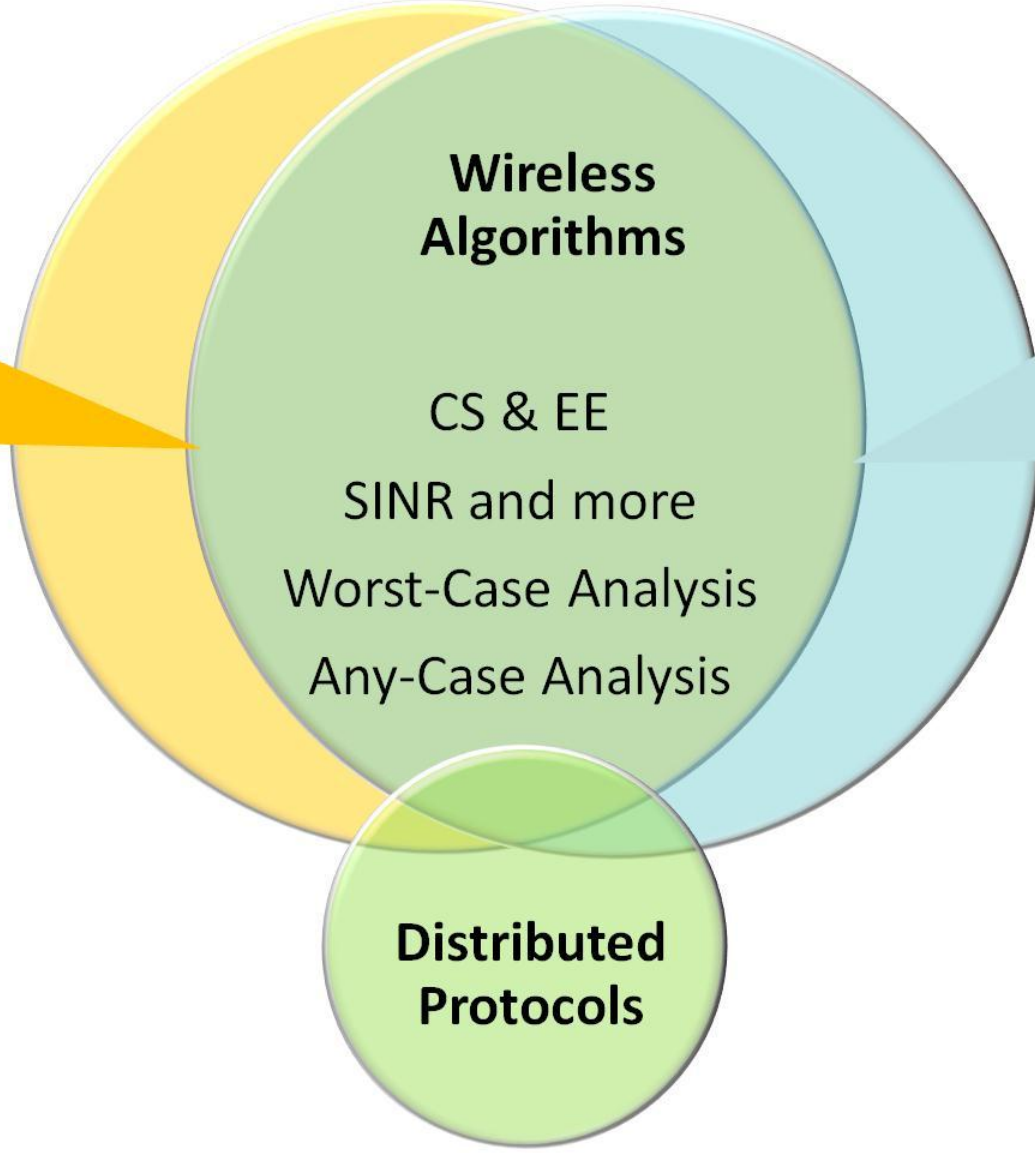
Any-Case Analysis

On the time-complexity of broadcast  
in multi-hop radio networks [Bar-

Yehuda, Goldreich, Itai, 1992]

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**Wireless Algorithms**

CS & EE

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Worst-Case Analysis

Any-Case Analysis

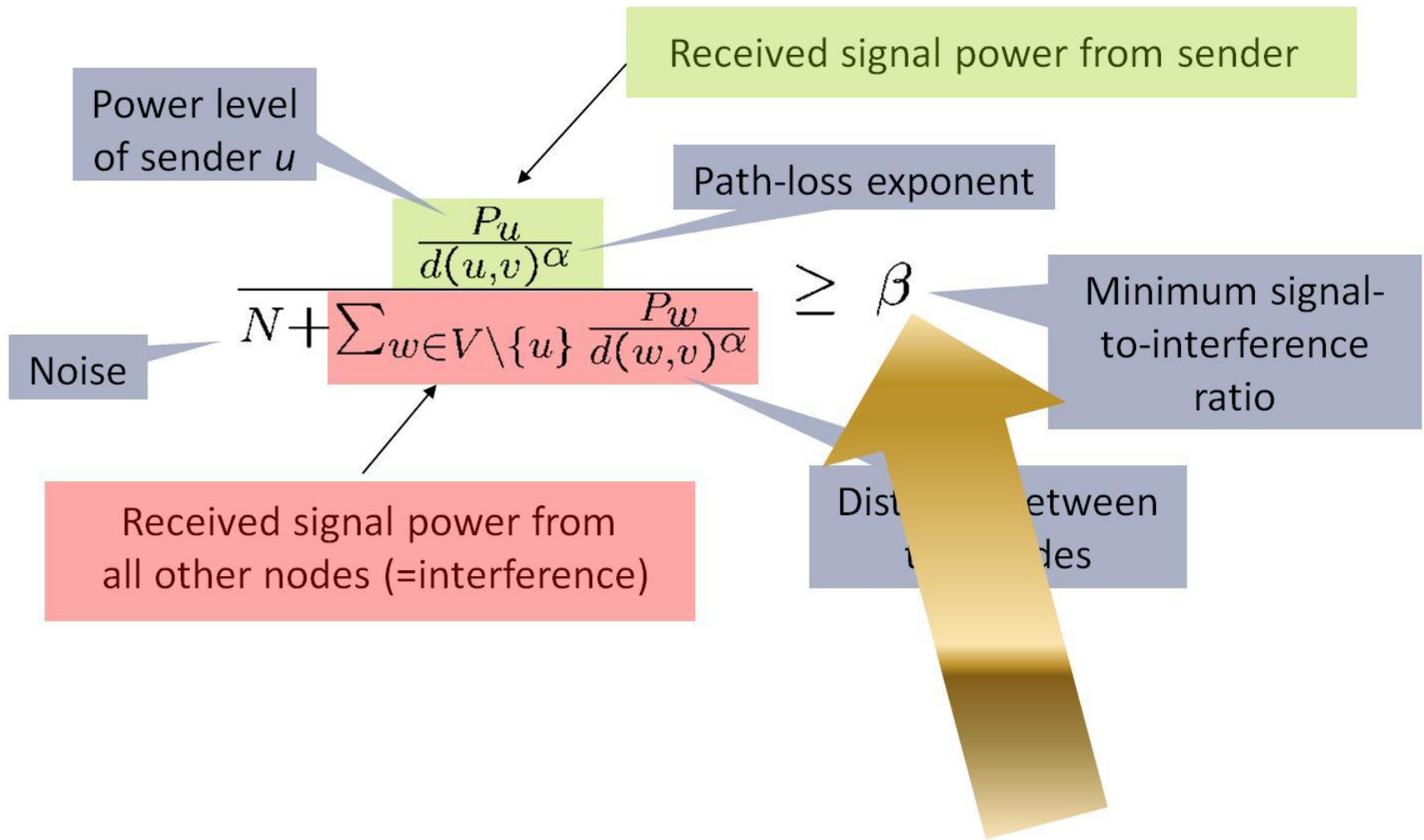
**Distributed Protocols**

On the time-complexity of broadcast  
in multi-hop radio networks [Bar-  
Yehuda, Goldreich, Itai, 1992]

# Wireless Communication 101



# Signal-To-Interference-Plus-Noise Ratio (SINR) Formula





Ratio  $\beta$  depends on receiver (hardware, software, parameters)

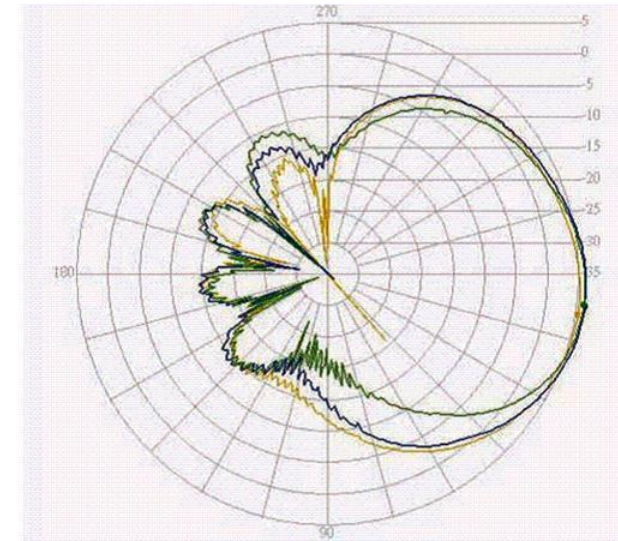
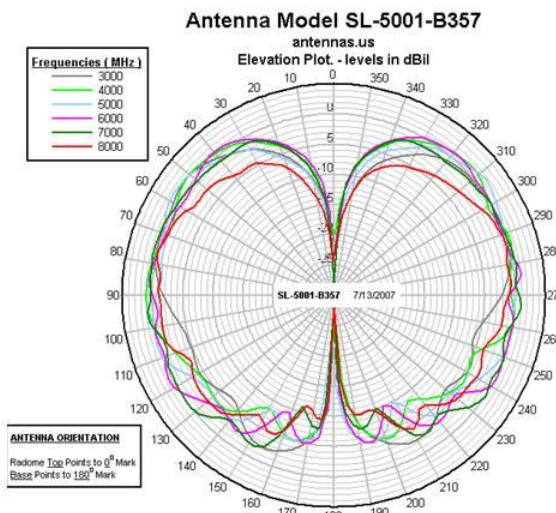
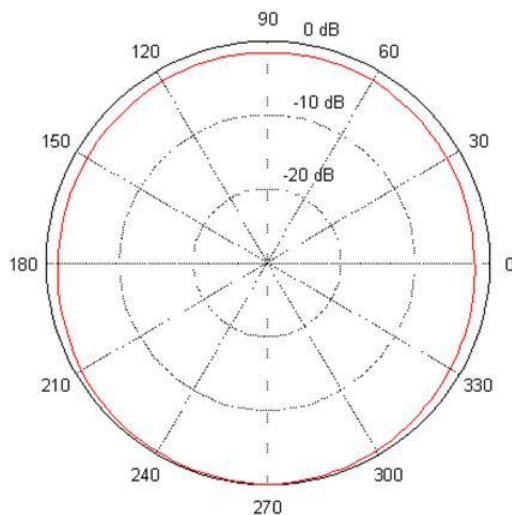
- Simple solutions have  $\beta > 10$ 
  - But  $\beta < 1$  is possible (thanks to forward error correction)

## Ratio $\beta$ depends on receiver (hardware, software, parameters)

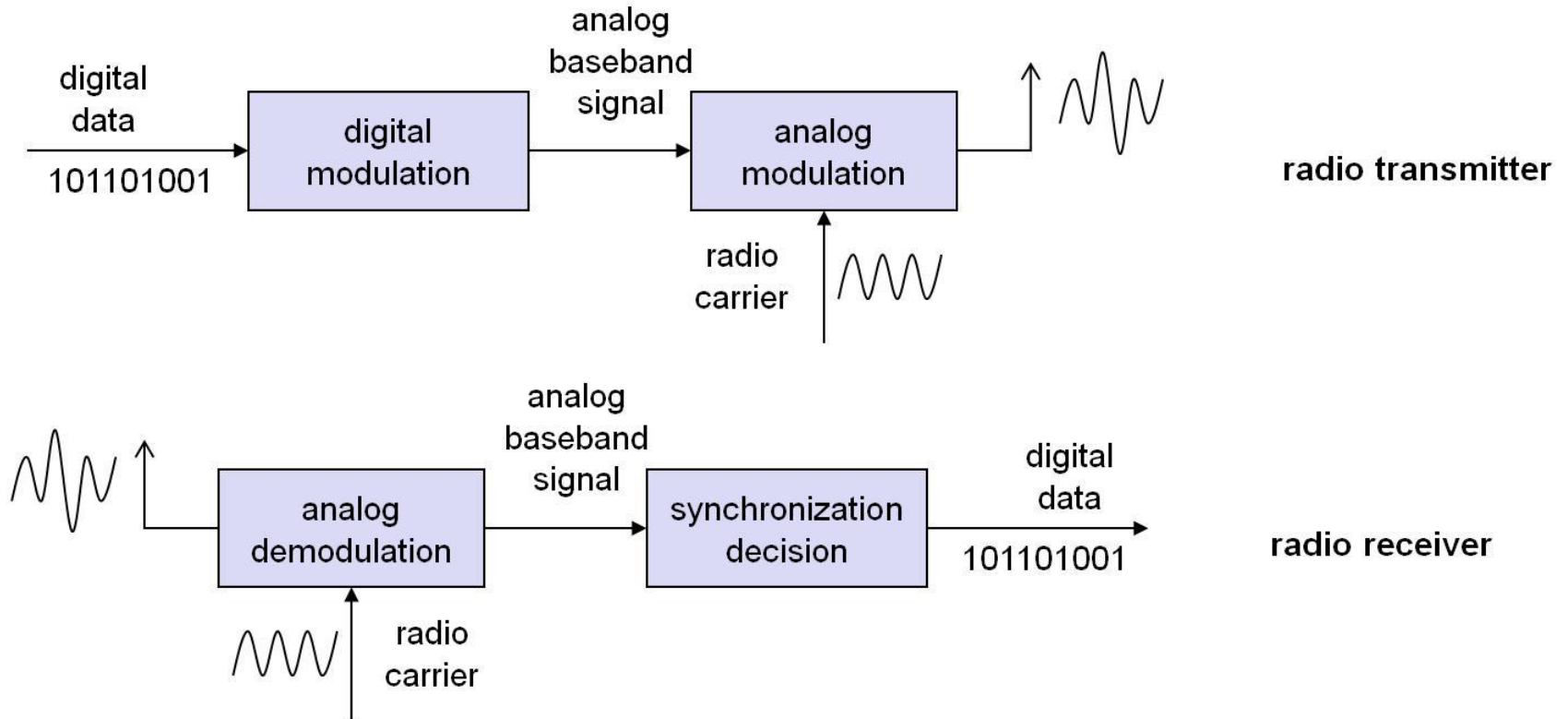
- Simple solutions have  $\beta > 10$ 
  - But  $\beta < 1$  is possible (thanks to forward error correction)
- Algorithmically speaking, the exact value of  $\beta$  does not really matter, thanks to SINR robustness
  - [Halldorsson, W, 2009] and [Fanghänel, Kesselheim, Räcke, Vöcking, 2009]
  - Model not only robust with regard to  $\beta$ , but also with regard to other constant factor disturbances, for instance, wind, constant antenna gain, etc.
  - Concretely: If we adapt model by factor  $\phi$ , results will change at most by factor  $\phi^2$ .

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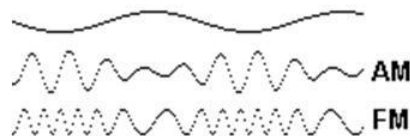
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# Modulation and demodulation



Modulation in action:



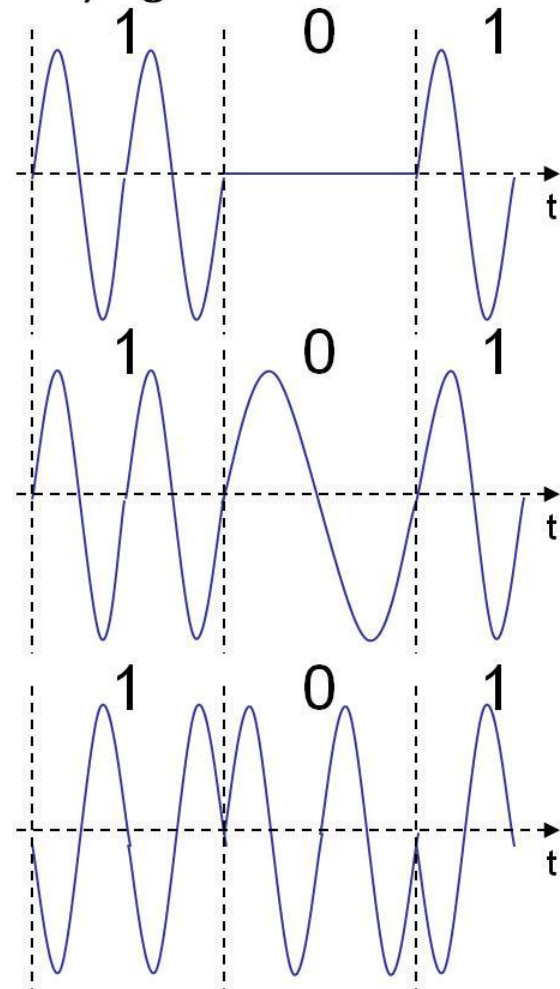
# Digital modulation

- Modulation of digital signals known as Shift Keying

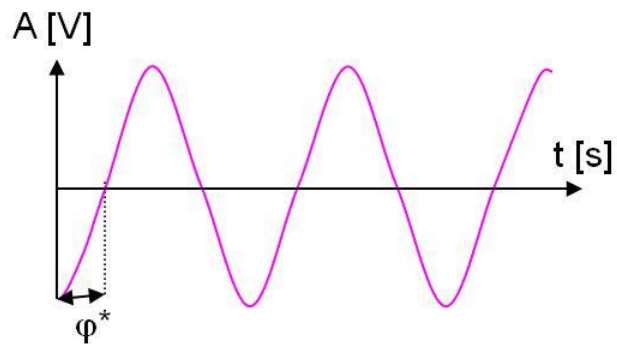
- Amplitude Shift Keying (ASK):
  - very simple
  - low bandwidth requirements
  - very susceptible to interference

- Frequency Shift Keying (FSK):
  - needs larger bandwidth

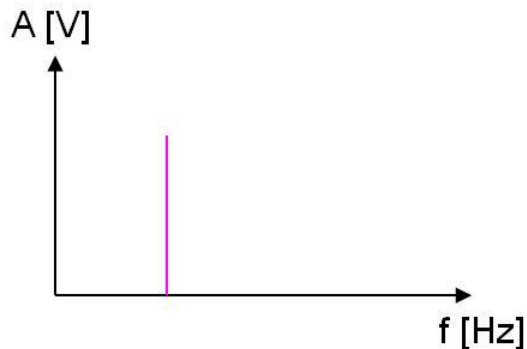
- Phase Shift Keying (PSK):
  - more complex
  - robust against interference



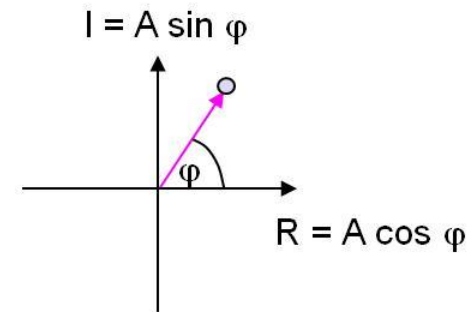
# Phase Shift Keying 101



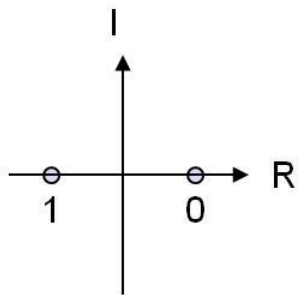
amplitude domain



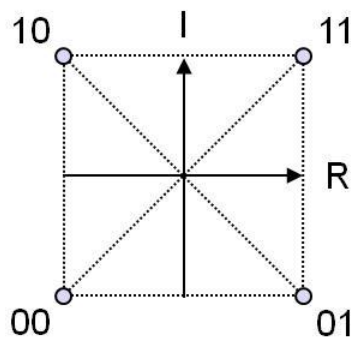
frequency spectrum



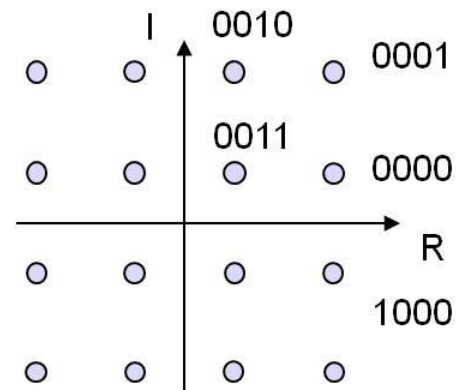
phase state diagram



BPSK (robust, satellites)



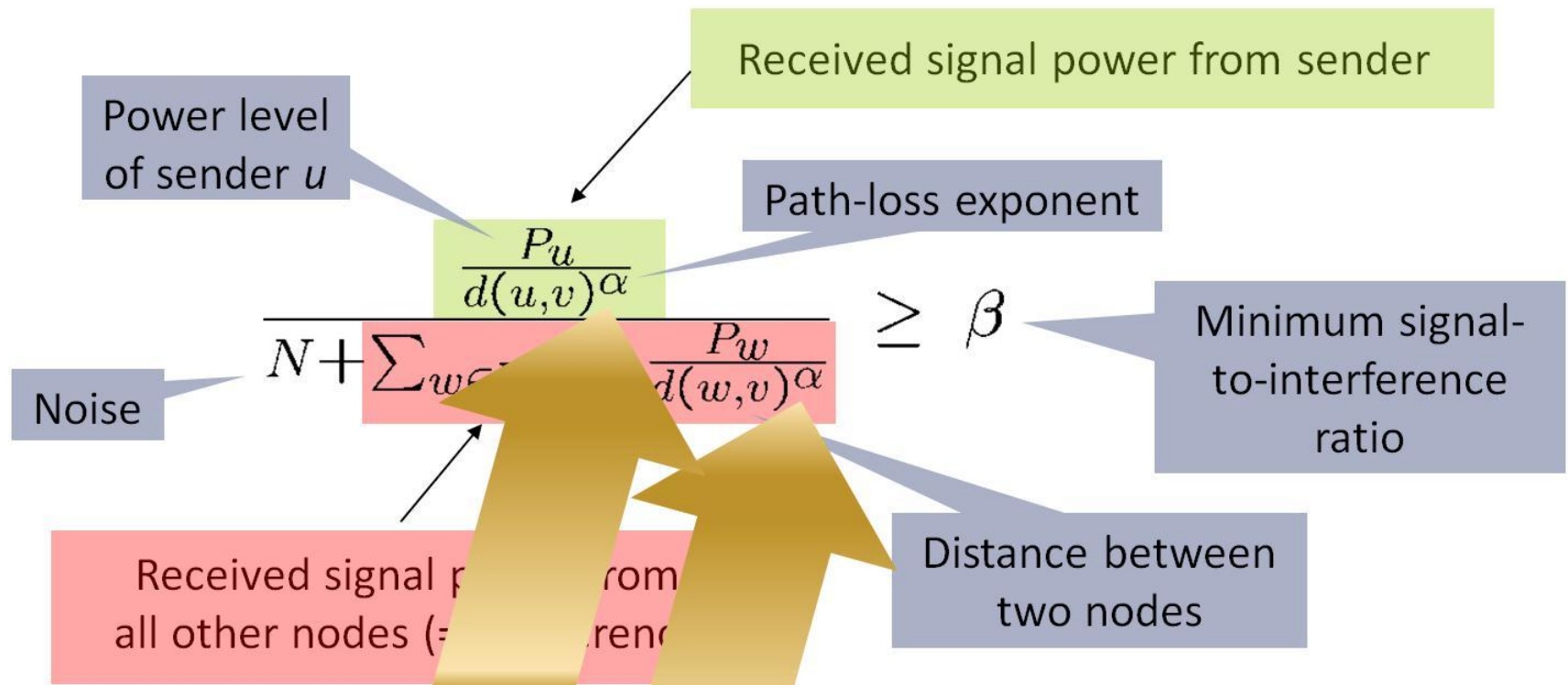
QPSK



QAM (large  $\beta$ )

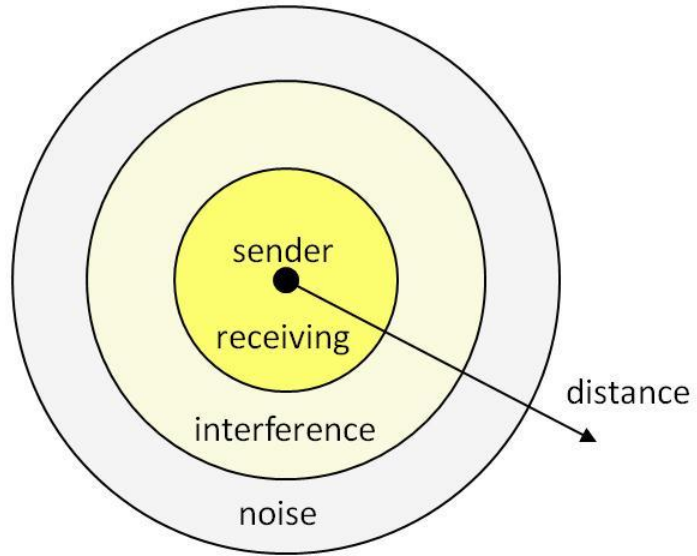


# Signal-To-Interference-Plus-Noise Ratio (SINR) Formula

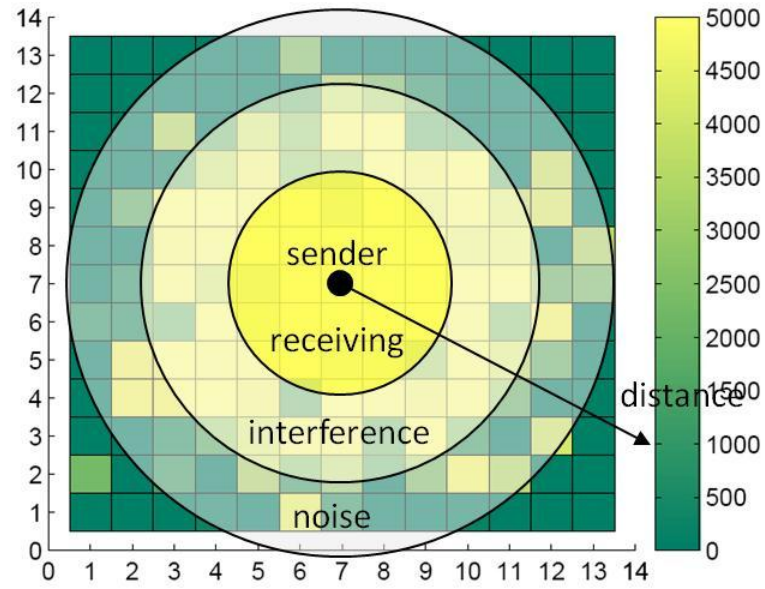


(BTW:  $d^\alpha$  has nothing to do with energy consumption)

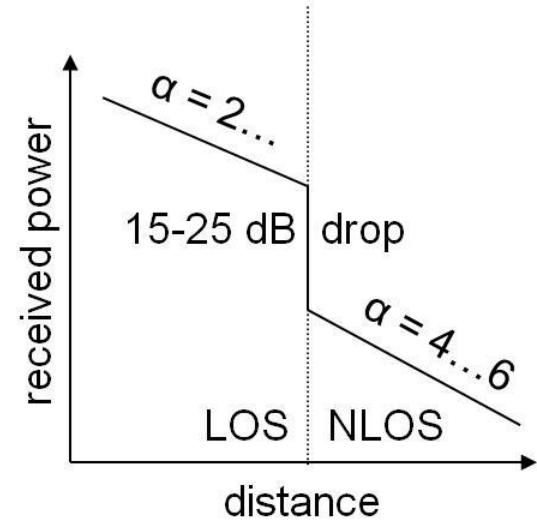
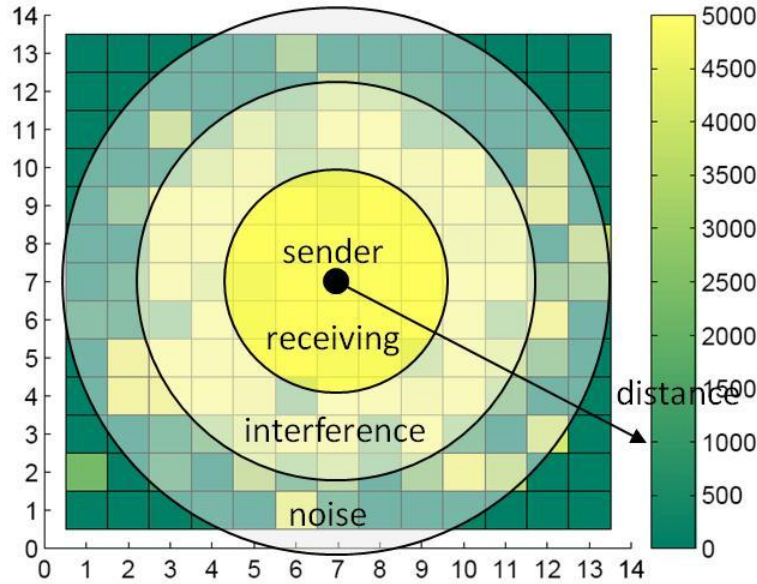
# Path-loss-exponent $\alpha$



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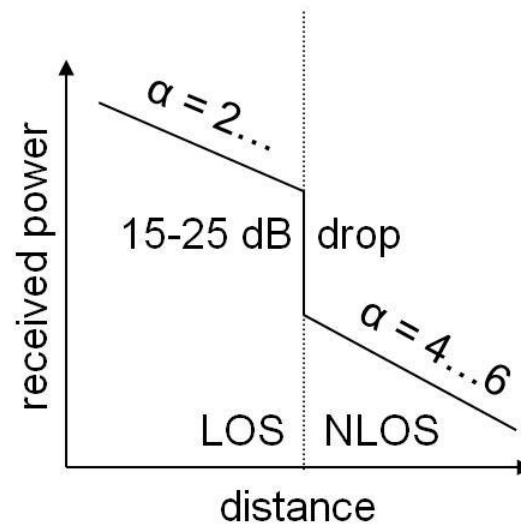
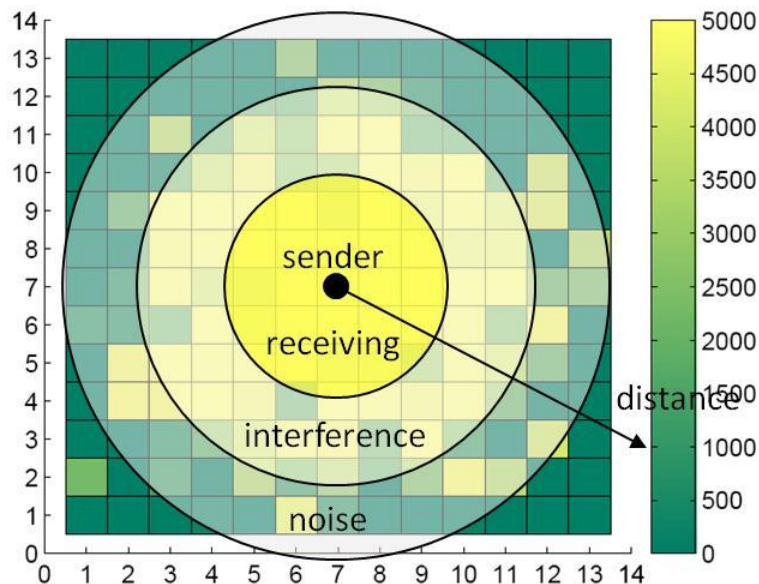
# Path-loss-exponent $\alpha$



$$P_r = \frac{P_s G_s G_r \lambda^2}{(4\pi)^2 d^2 L}$$

$$P_r = \frac{P_s G_s G_r h_s^2 h_r^2}{d^4}$$

# Path-loss-exponent $\alpha$



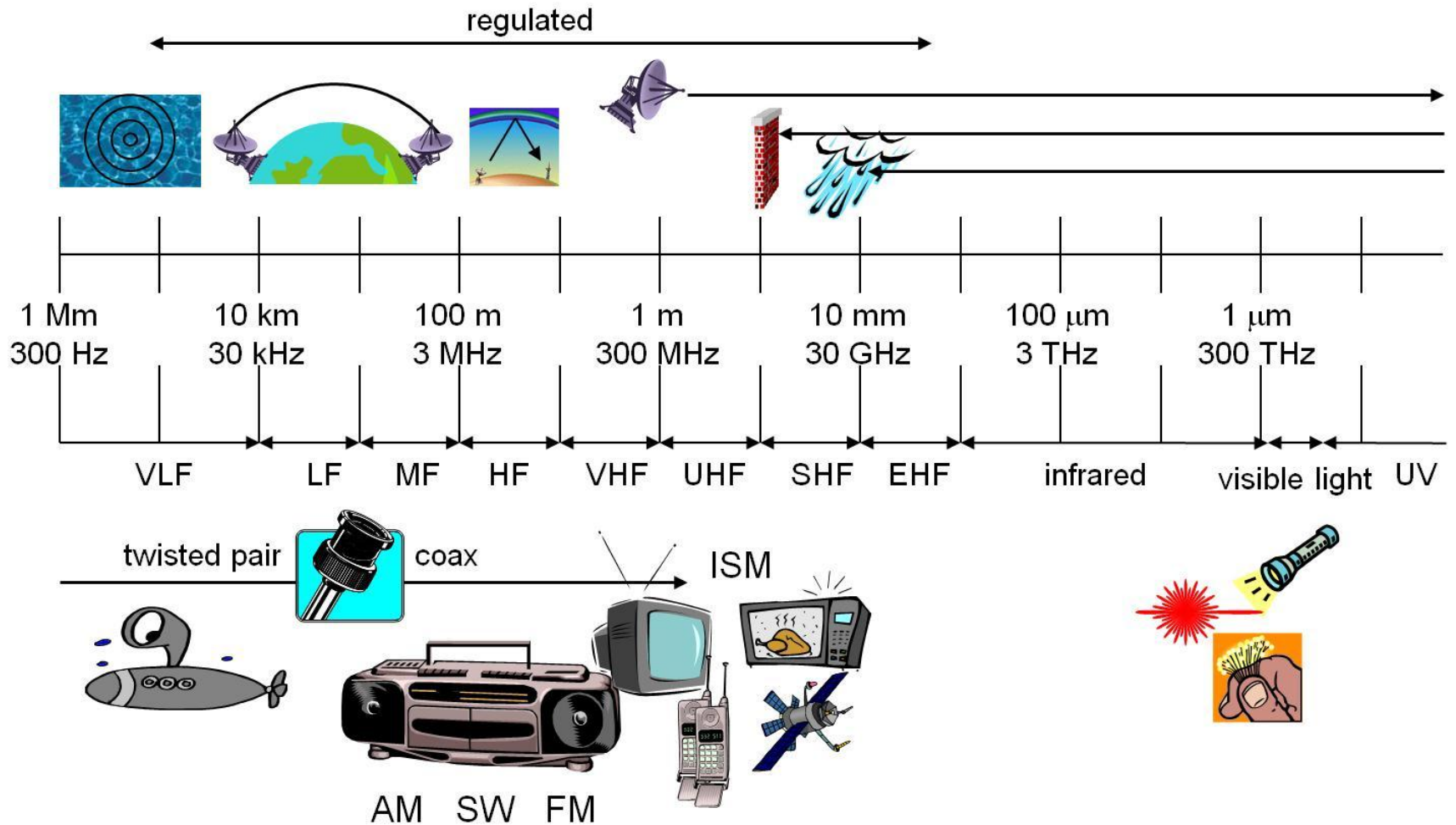
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$\alpha \geq \text{Dimension}$   
2<sup>nd</sup> law of thermodyn.



# Wireless Propagation Depends on Frequency





## Path-loss-exponent $\alpha$ : Near-Field Effects

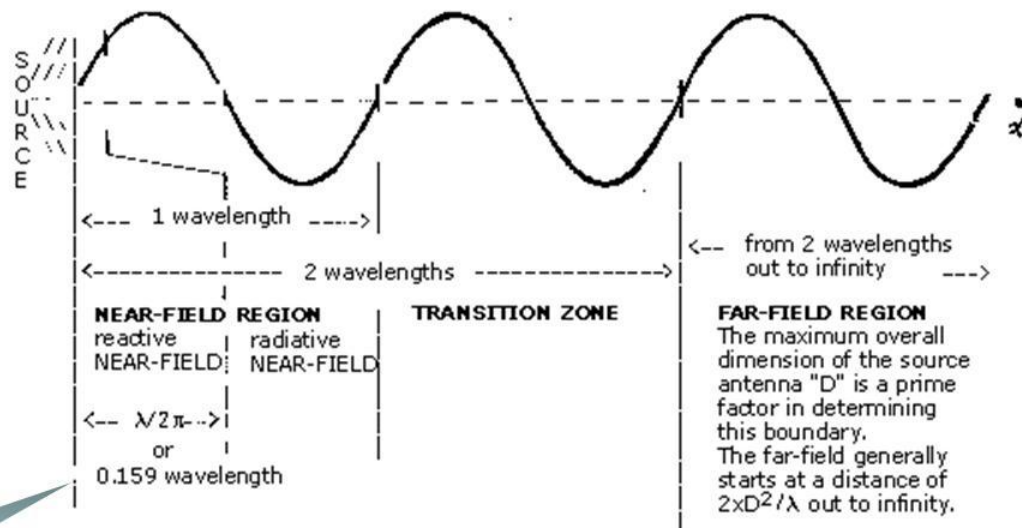
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$d \ll 1?$

# Path-loss-exponent $\alpha$ : Near-Field Effects

$$P_r = \frac{P_s G_s G_r \lambda^2}{(4\pi)^2 d^2 L}$$

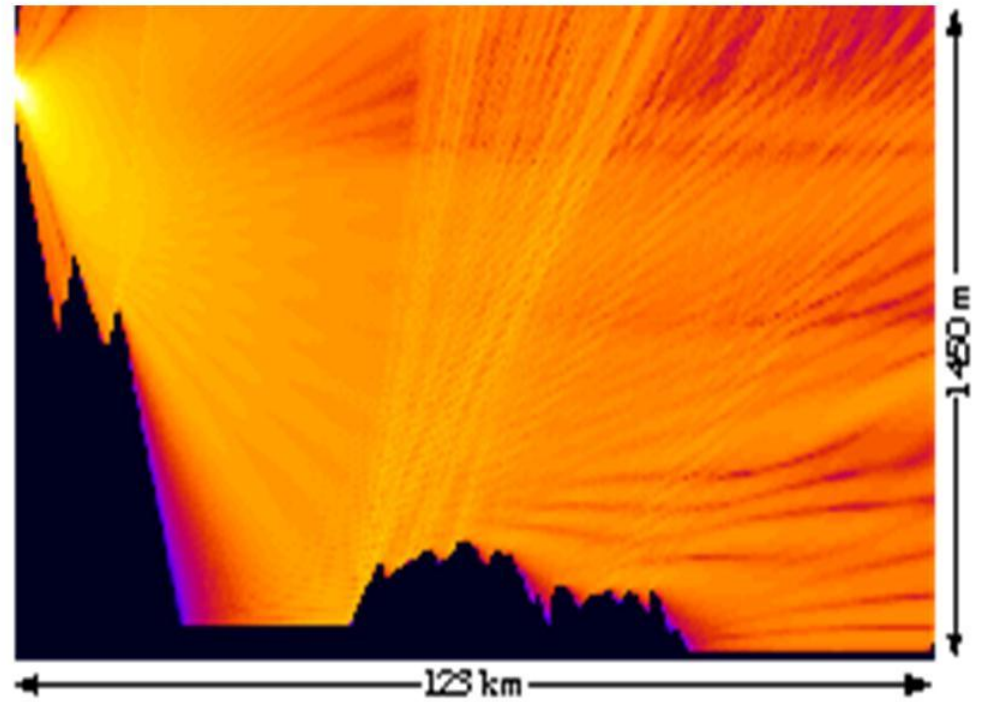
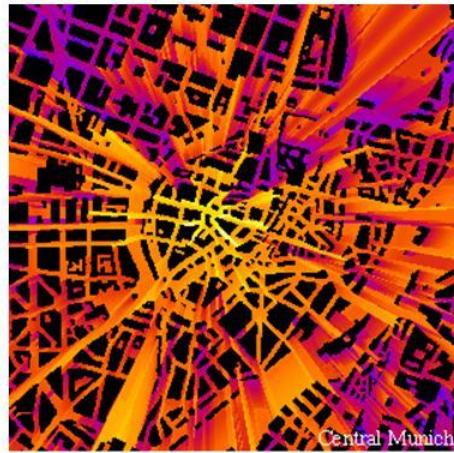
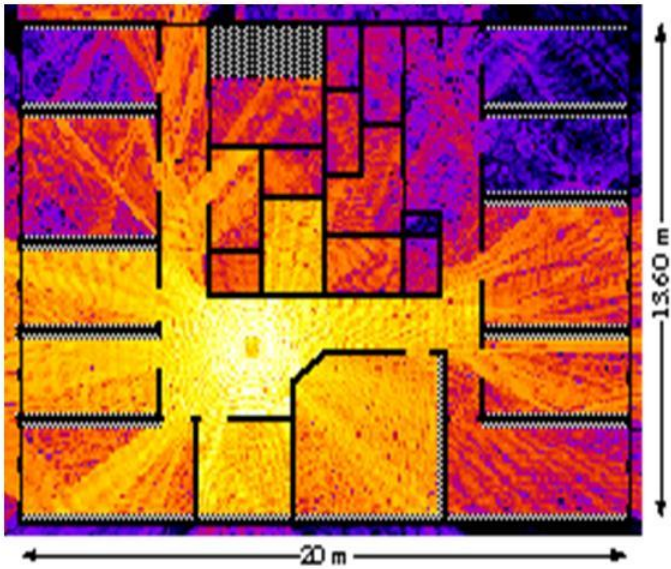
$d \ll 1?$



1cm!!

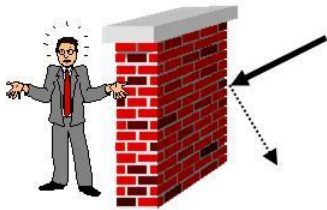
... in other words, algorithmic papers should rule out near-field effects

# Real World Examples

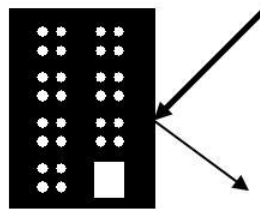


# Attenuation by objects

- Shadowing (3-30 dB):
  - textile (3 dB)
  - concrete walls (13-20 dB)
  - floors (20-30 dB)
- reflection at large obstacles
- scattering at small obstacles
- diffraction at edges
- fading (frequency dependent)



shadowing



reflection



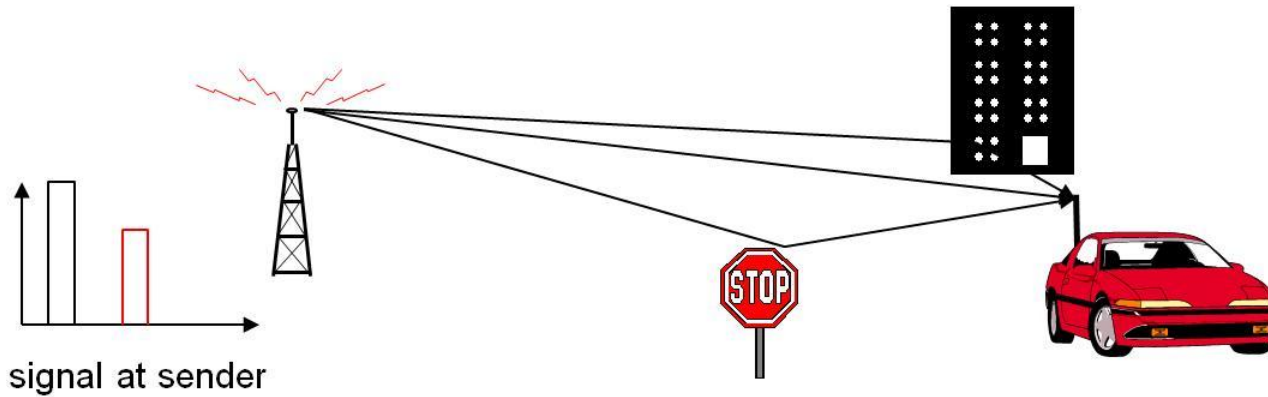
scattering



diffraction

# Multipath

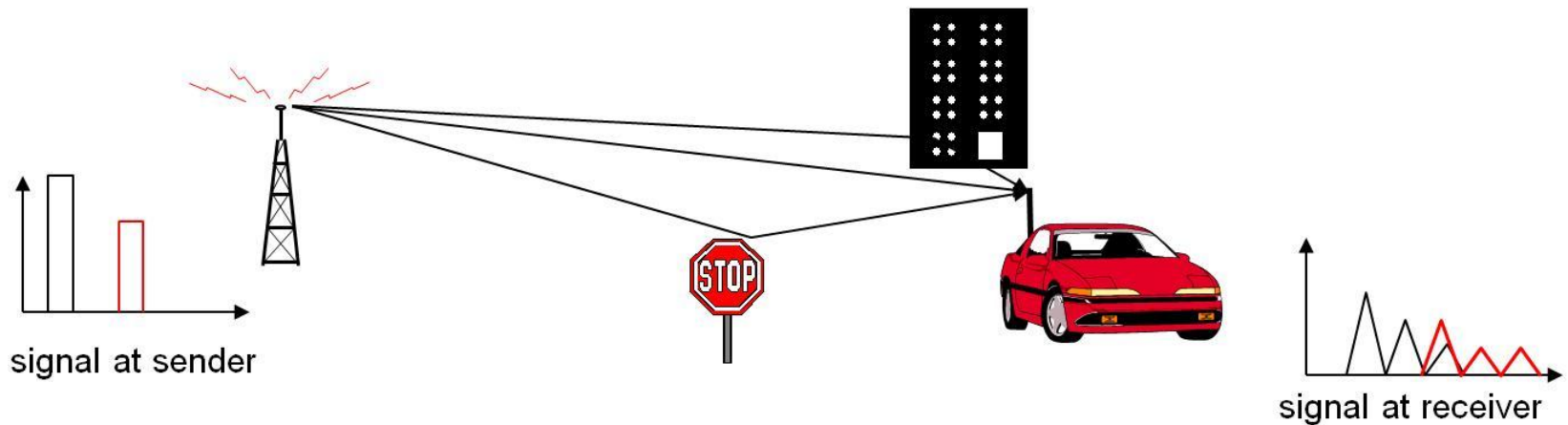
- Signal can take many different paths between sender and receiver due to reflection, scattering, diffraction



- Time dispersion: signal is dispersed over time

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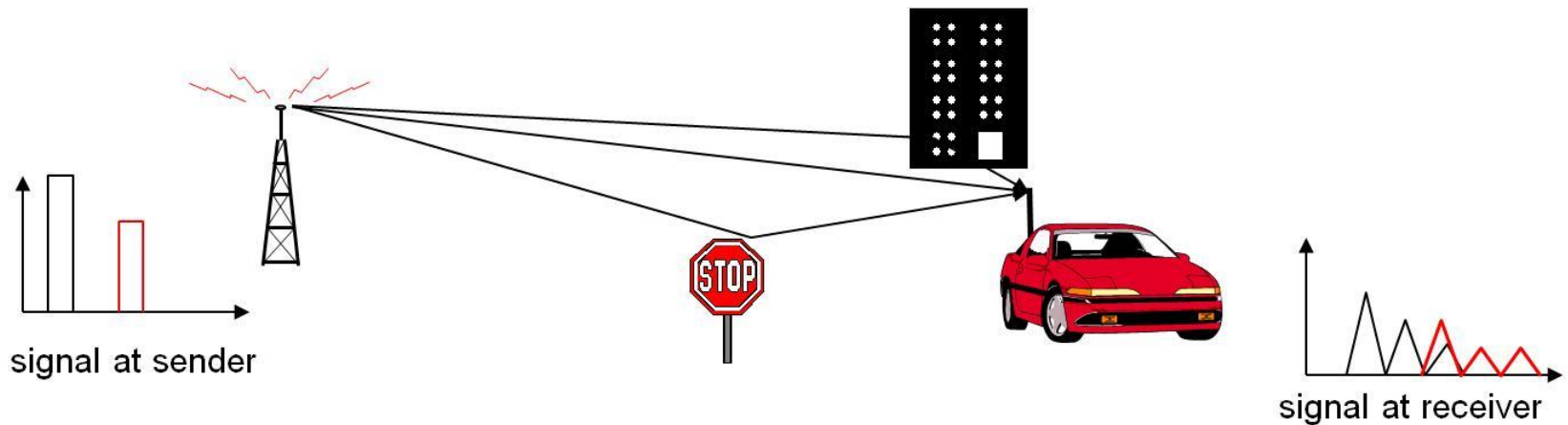


- Time dispersion: signal is dispersed over time
- Interference with “neighbor” symbols:  
Inter Symbol Interference (ISI)

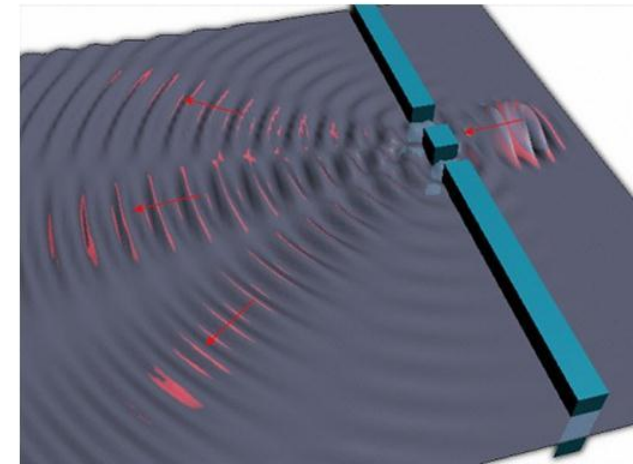


# Multipath

- Signal can take many different paths between sender and receiver due to reflection, scattering, diffraction



- Time dispersion: signal is dispersed over time
- Interference with “neighbor” symbols: Inter Symbol Interference (ISI)
- The signal reaches a receiver directly and phase shifted. Distorted signal depending on the phases of the different parts



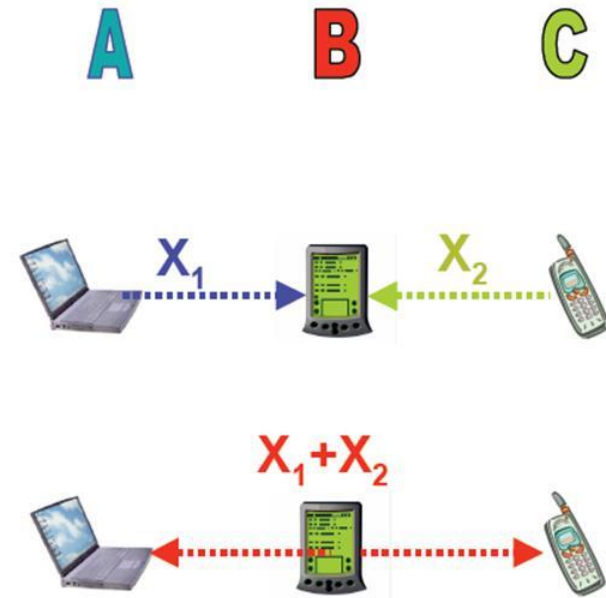
There's much more...



There's much more...



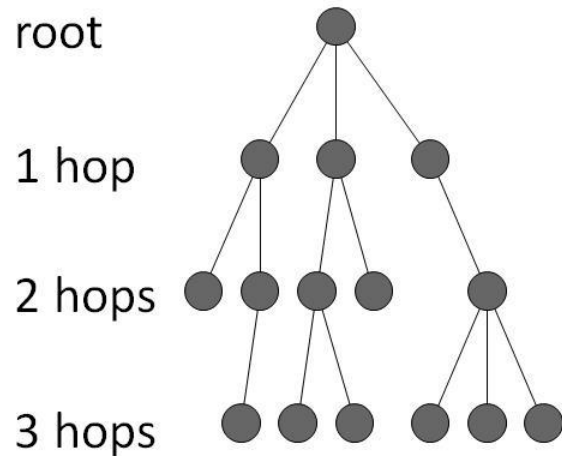
### Analog Network Coding



Advanced Algorithms?

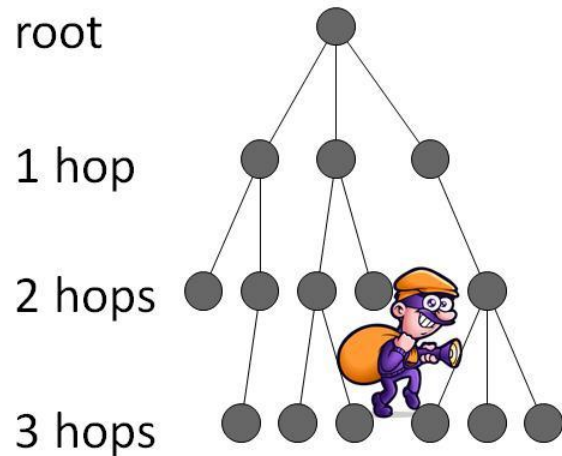
# Quiz: How to Build a Multi-Hop Alarm System

Problem: More than 1 node may sense problem at the same time.  
Potentially we have a massive interference problem!



# Quiz: How to Build a Multi-Hop Alarm System

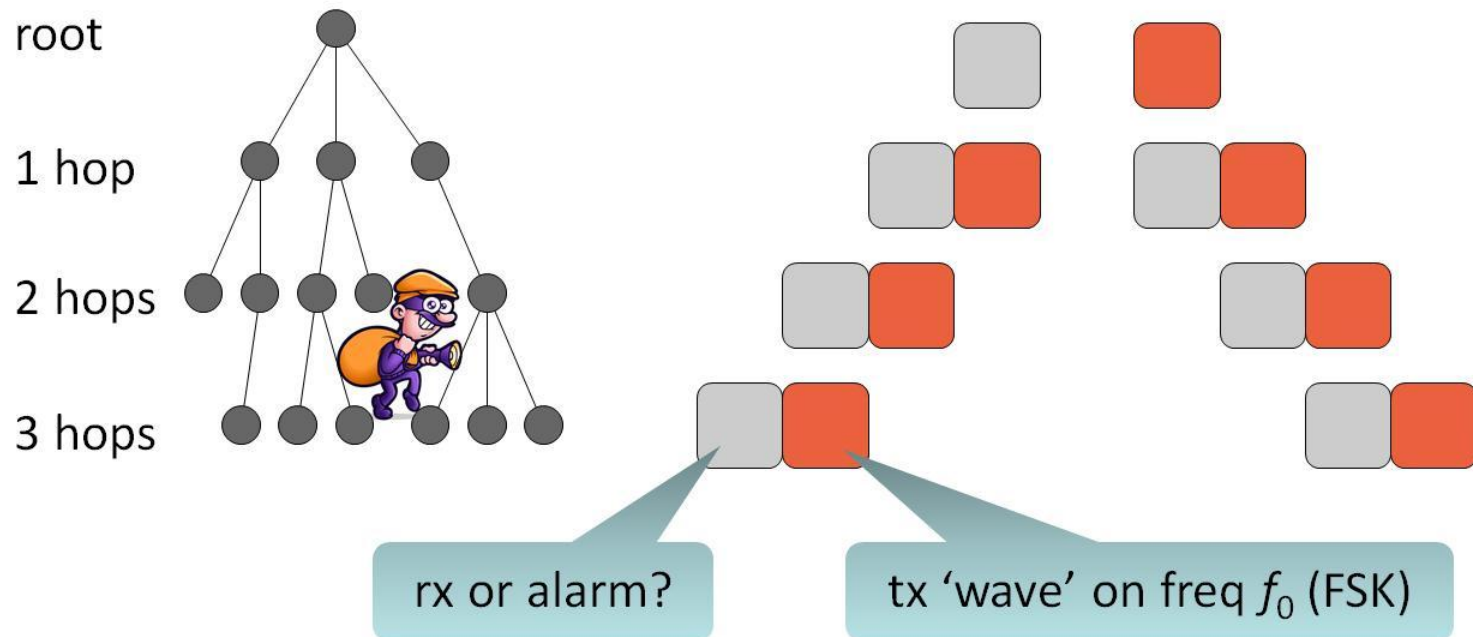
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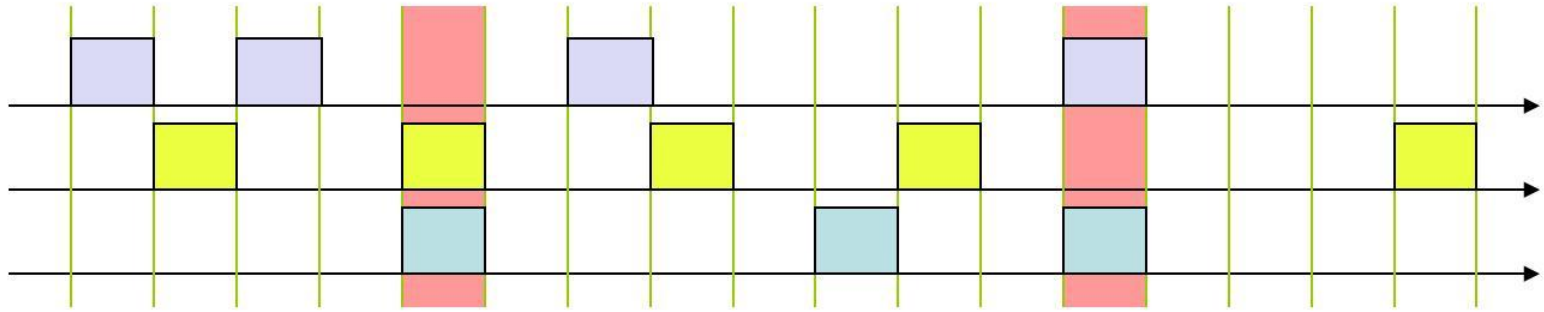
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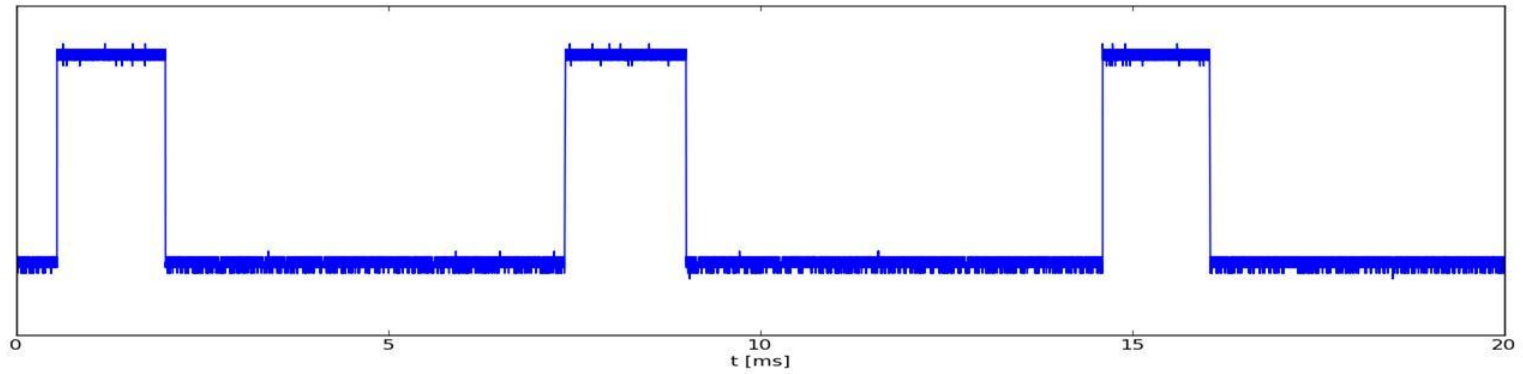
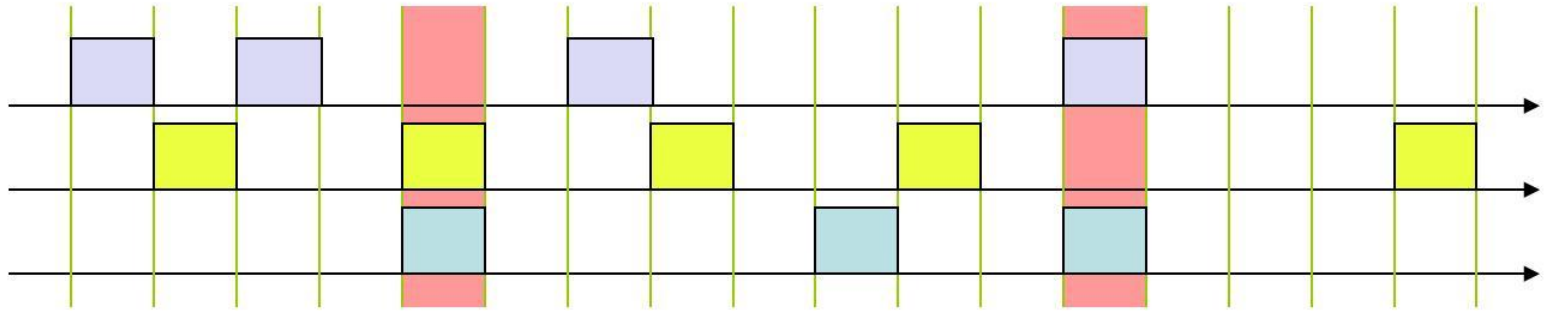


...followed by verification (1% false positives outdoors)

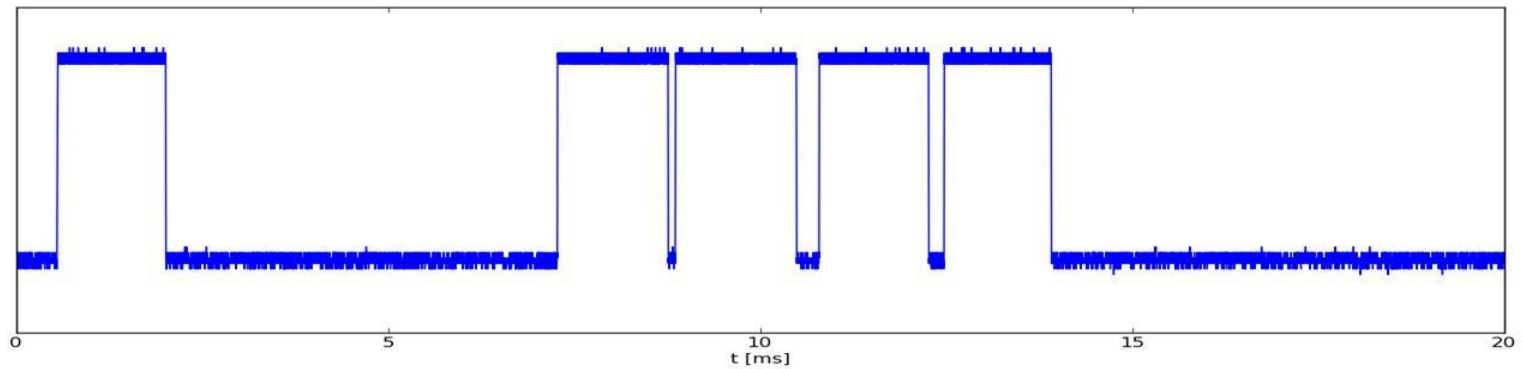
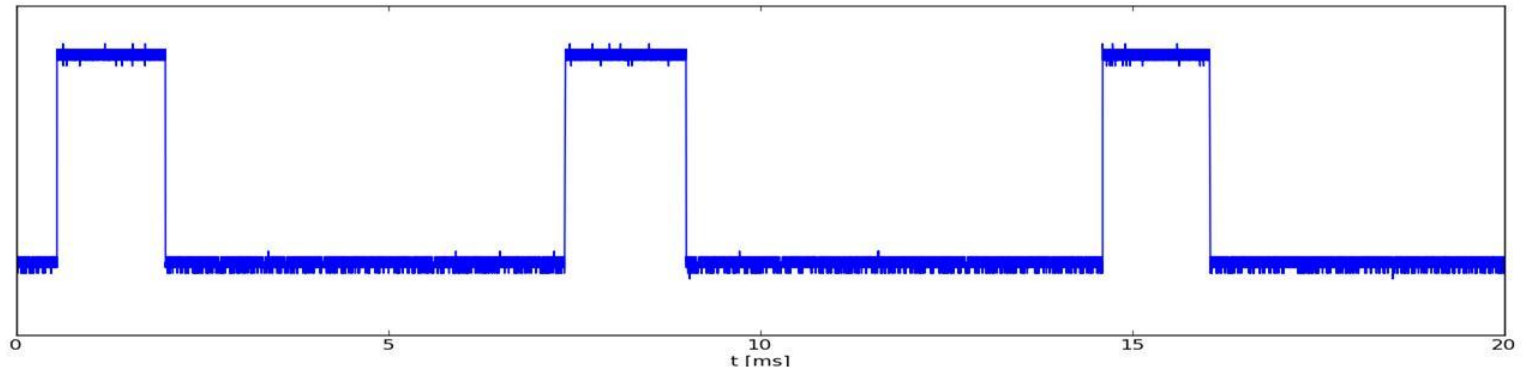
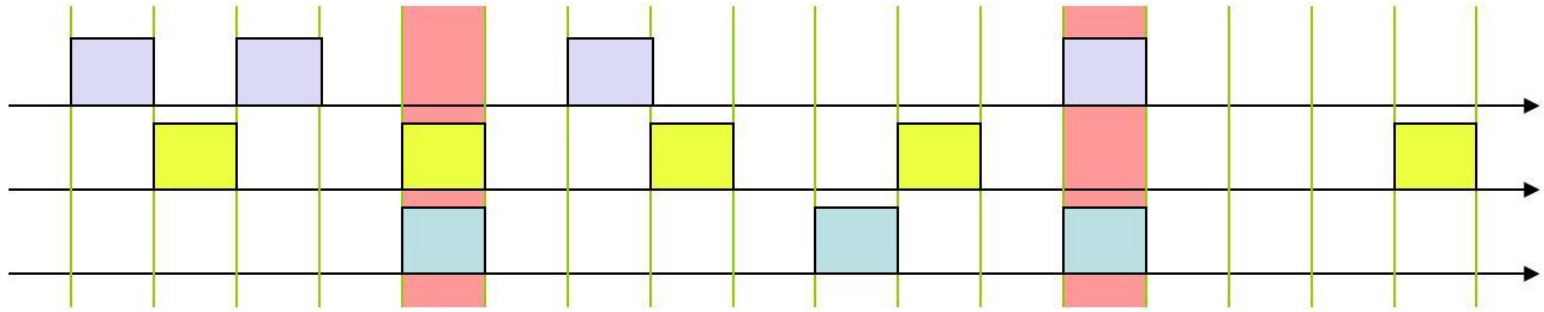
# Media Access: Theory and Practice



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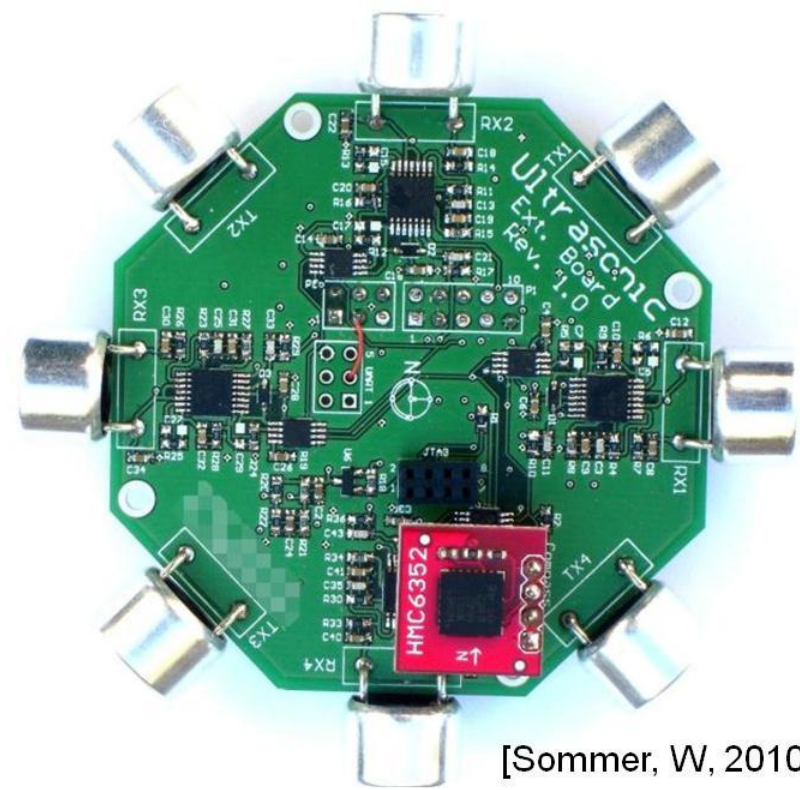


# Media Access: Theory and Practice

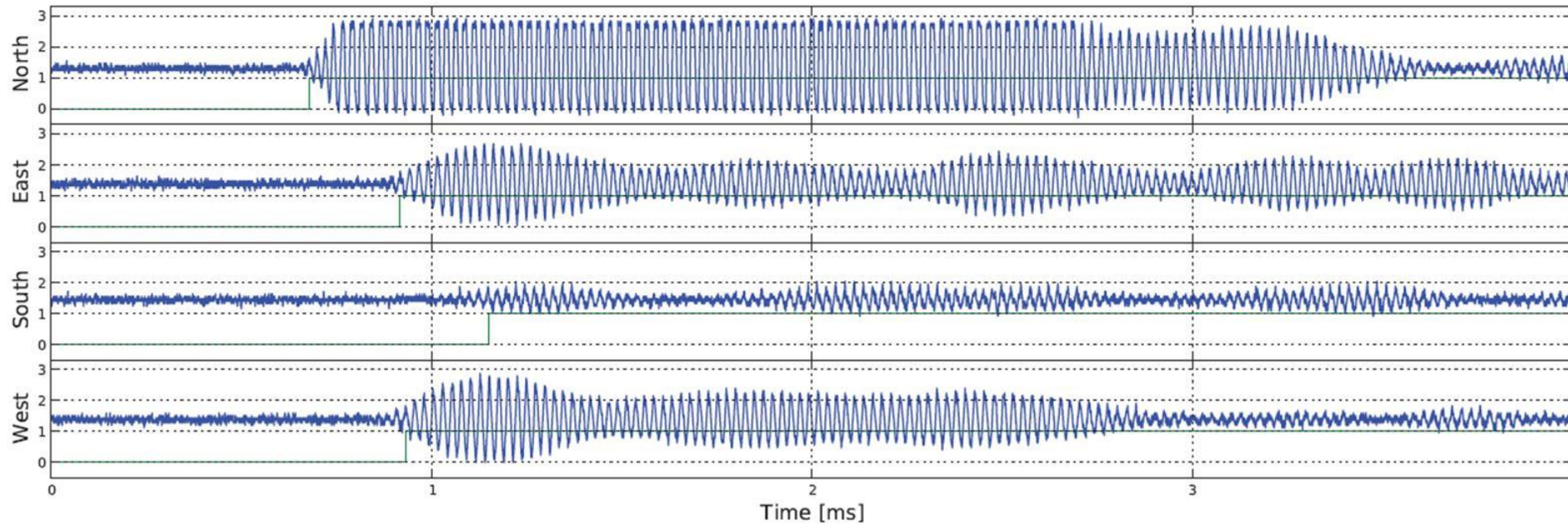


# Ultrasound

(A different kind of communication)

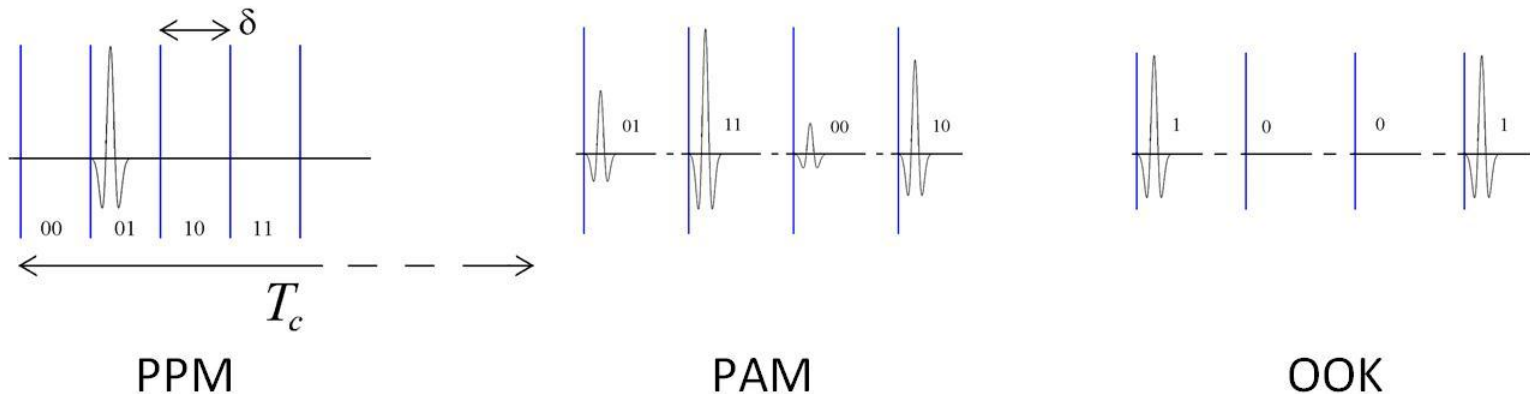


[Sommer, W, 2010]



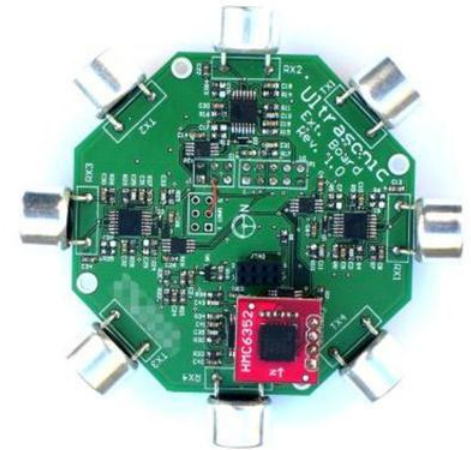
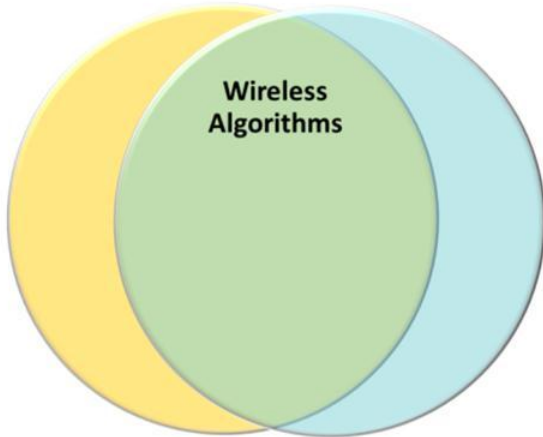
# Ultra-Wideband (UWB)

- An example of a new physical paradigm.
- Discard the usual dedicated frequency band paradigm.
- Instead share a large spectrum (about 1-10 GHz).
- Modulation: Often pulse-based systems. Use extremely short duration pulses (sub-nanosecond) instead of continuous waves to transmit information. Depending on application 1M-2G pulses/second





# Summary



# *Thank You!*

*Questions & Comments?*



*Roger Wattenhofer*