

University of Delaware
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Information & Inference in the Wireless Physical Layer

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Information & Inference in the Wireless PHY



Wireless Networks: Layers

Application (APP)



Web Browsing,
Voice, etc.

Network (NET)



Routing,
Flow Control,
etc.

Medium Access Control (MAC)



Scheduling,
Access Control,
etc.

Physical (PHY)



Data
Transmission

Information & Inference in the Wireless PHY



Research Trends in Wireless Nets

- The Past 25 Years: Key Developments at the PHY
 - CDMA
 - OFDM
 - UWB
 - MUD
 - MIMO
 - Turbo

Information & Inference in the Wireless PHY



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- The Past 25 Years: Key Developments at the PHY
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- Today: Focus on Interactions Among Nodes & Across Layers
 - *Among Nodes*:
 - Competition
 - Collaboration
 - Cooperation
 - *Across Layers*:
 - MAC-PHY
 - NET-PHY
 - APP-PHY

Information & Inference in the Wireless PHY



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 - Among Nodes:
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 - Across Layers:
 - MAC-PHY
 - NET-PHY
 - APP-PHY ← Information Transmission & Statistical Inference

Information & Inference in the Wireless PHY



Today's Talk: Four Problems in the Wireless PHY Motivated by the APP

- *PHY Security in Wireless Communication Networks*

Motivated by Secure Information Transmission

- *Distributed Learning*

Motivated by Statistical Inference in Wireless Sensor Networks

- *Finite-Blocklength Capacity*

Motivated by Multimedia Information Transmission

- *Message Delivery in Small-World Networks*

Motivated by Social Networking (Information & Inference)

Information & Inference in the Wireless PHY



Physical Layer Security in Communication Networks

Secure Information Transmission

[Joint work with [Yingbin Liang](#), [Shlomo Shamai](#), et al.]

Information & Inference in the Wireless PHY



Motivation: Exploiting the PHY

- Key Techniques for Improving Capacity & Reliability:
 - *MIMO*
 - *Cooperation & Relaying*
 - *Cognitive Radio*

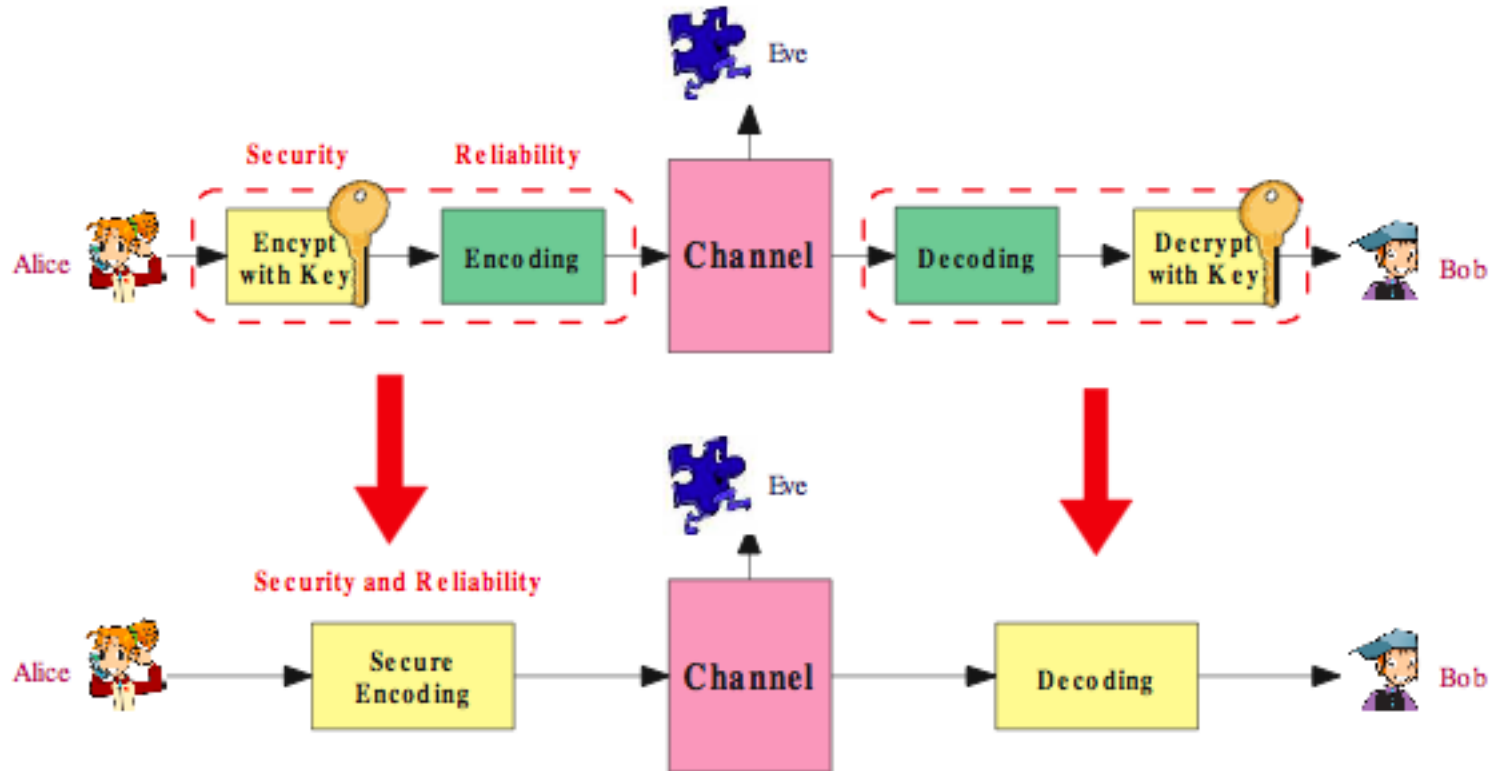
Motivation: Exploiting the PHY

- Key Techniques for Improving Capacity & Reliability:
 - MIMO
 - Cooperation & Relaying
 - Cognitive Radio
- What About Security?
 - Traditionally a higher-layer issue (APP or Presentation)
 - Encryption can be complex and difficult without infrastructure
 - *Information theoretic security* examines the fundamental ability of the PHY to provide security
 - Caveat: This is still largely a theoretical issue



Physical Layer Security

Joint Encoding for Security and Reliability

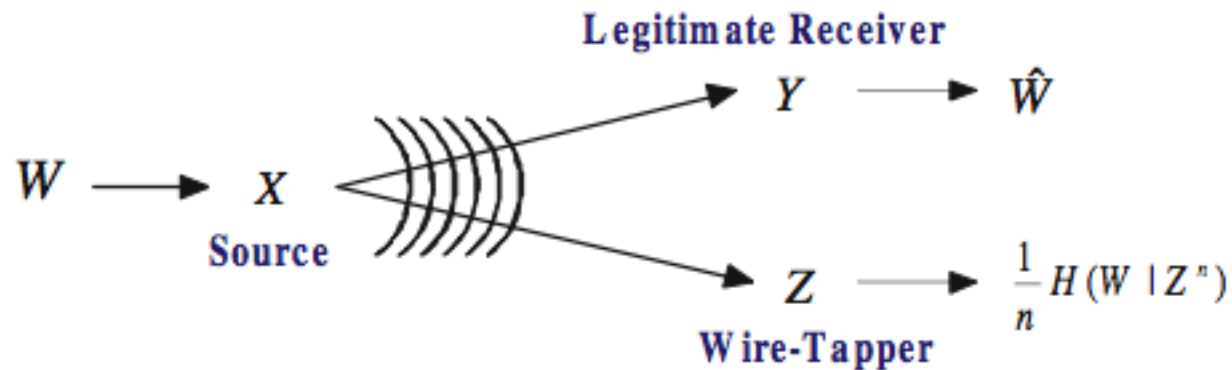


PHY Security in Wireless Communication Nets



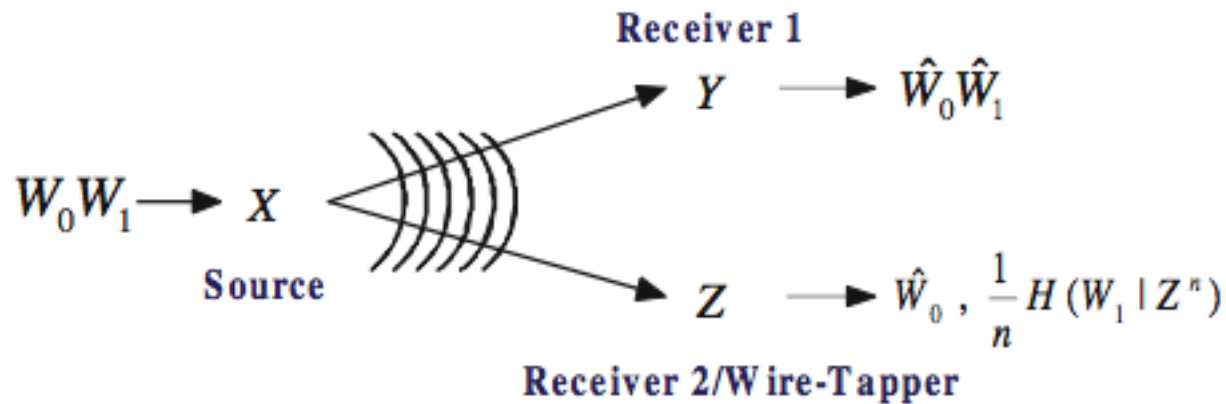
A (Very) Brief History

- Shannon [BSTJ'49]: For **cipher**, need $H(K) > H(S)$.
- Wyner [BSTJ'75]: For the **wire-tap channel**



the wire-tapper must be **degraded**.

Broadcast Channel with Confidential (BCC) Messages



Csiszár & Körner [IT'78]: Discrete Memoryless BCC

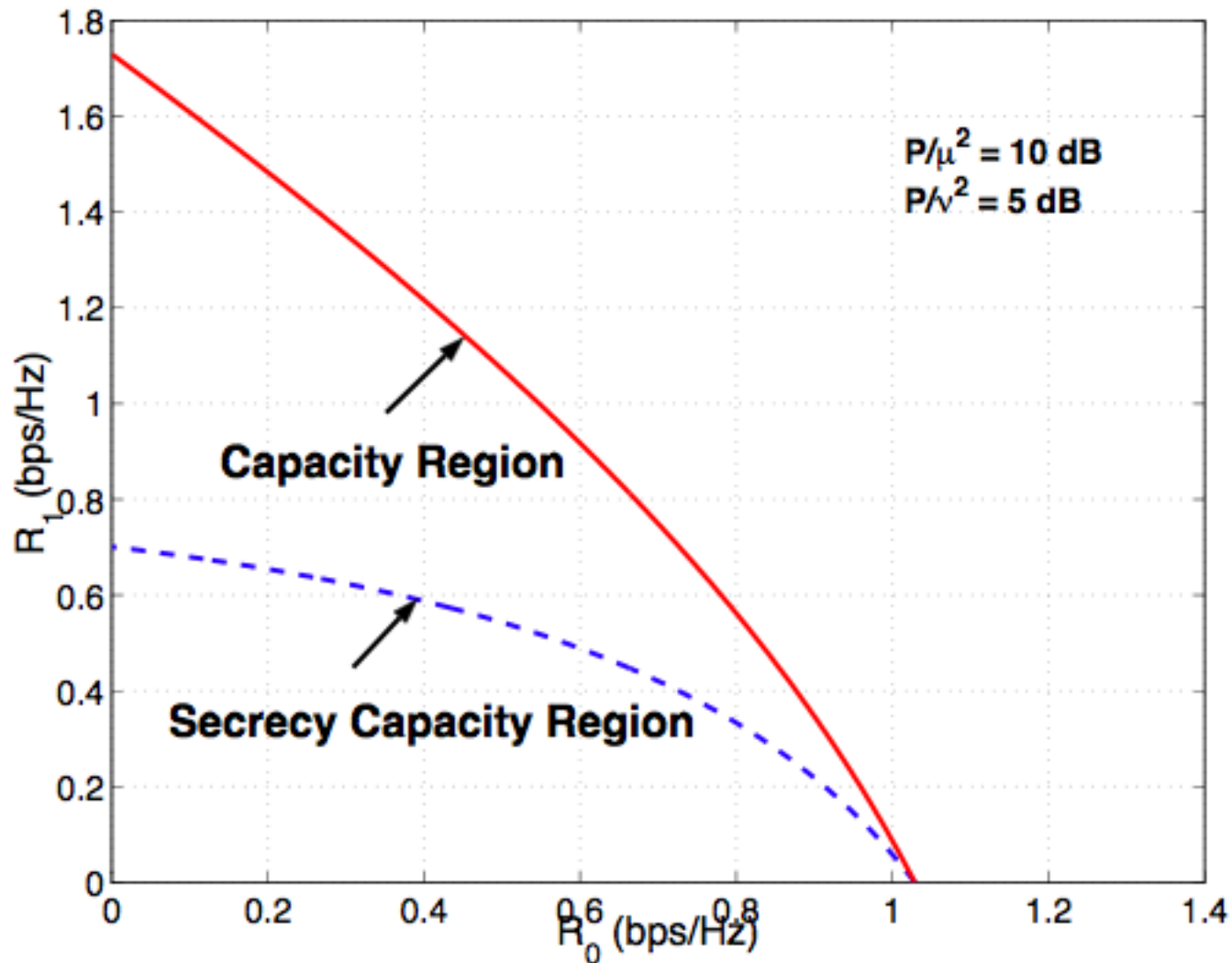
Liang, Poor & Shamai [IT'08]:

- **Gaussian** BCC
 - secrecy-capacity region
- **Fading** BCC
 - secrecy-capacity region
 - **exploit fading** to achieve secrecy
 - optimal power allocation

PHY Security in Wireless Communication Nets



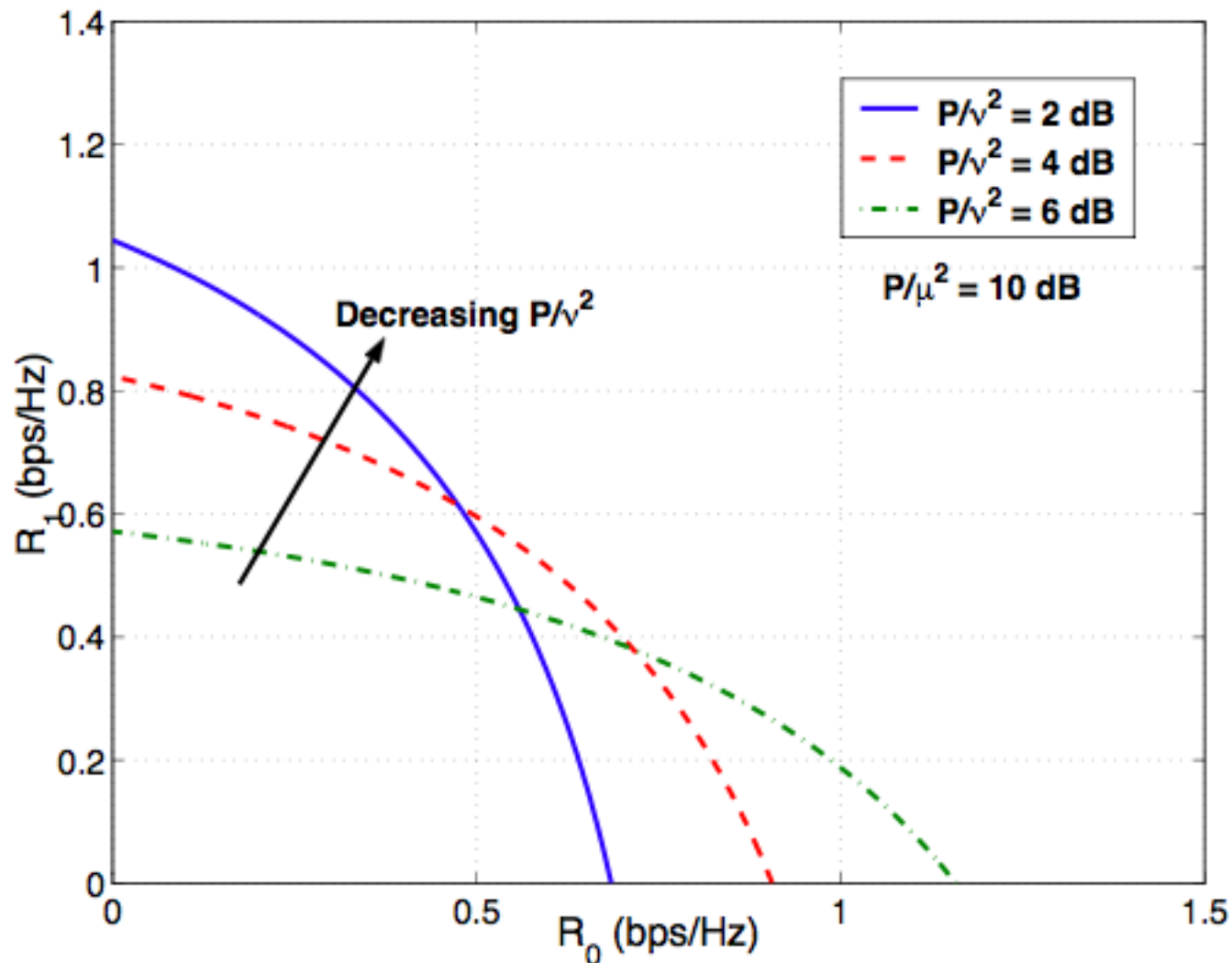
Gaussian BCC



PHY Security in Wireless Communication Nets



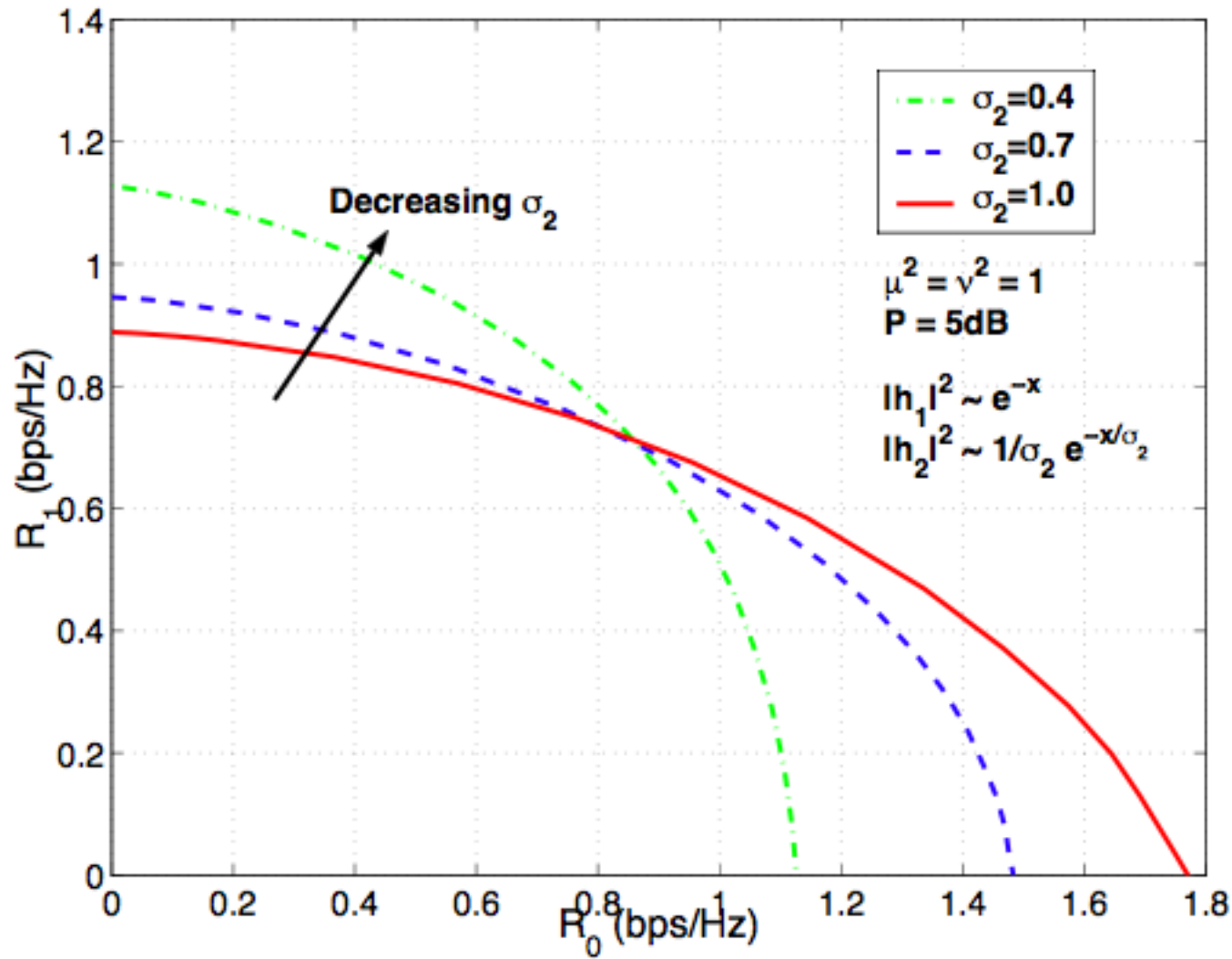
Gaussian BCC: Secrecy Capacity Regions



PHY Security in Wireless Communication Nets



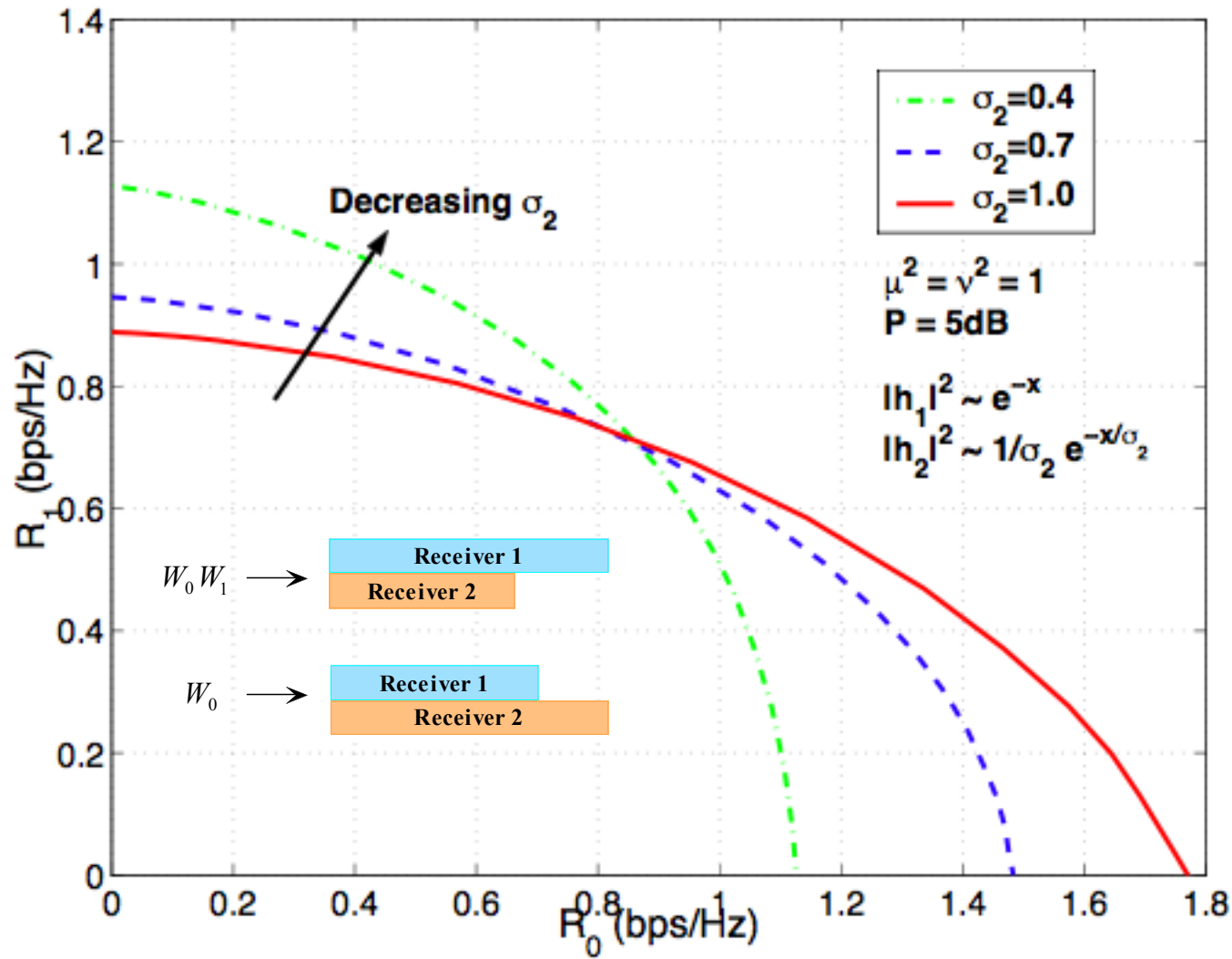
Fading BCC: Secrecy Capacity Regions



PHY Security in Wireless Communication Nets



Fading BCC: Secrecy Capacity Regions

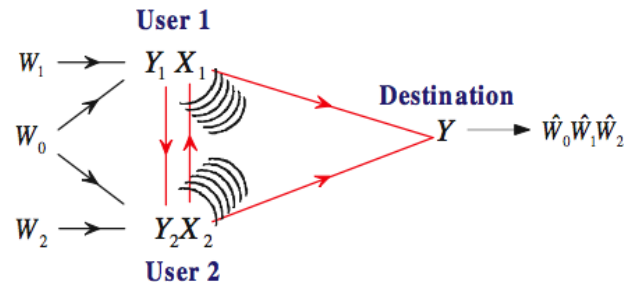


PHY Security in Wireless Communication Nets

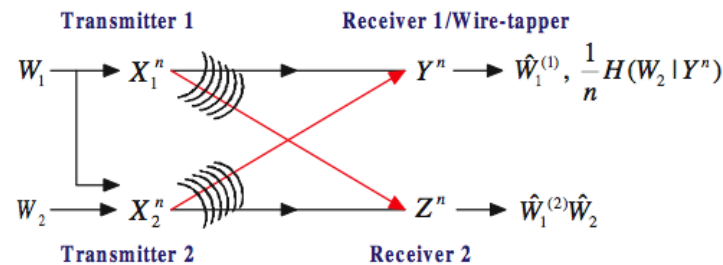


Other Channels of Interest

- Multiple-Access Channel [w/ Liang - IT'08 (Gaussian); w/ Liu, Liang - IT'11 (fading)]:



- Interference Channel [w/ Liang, Someck-Baruch, Shamai, Verdú - IT'09 (cognitive) & w/ Koyluoglu, El Gamal, Lai - IT'11 (interference alignment)]:

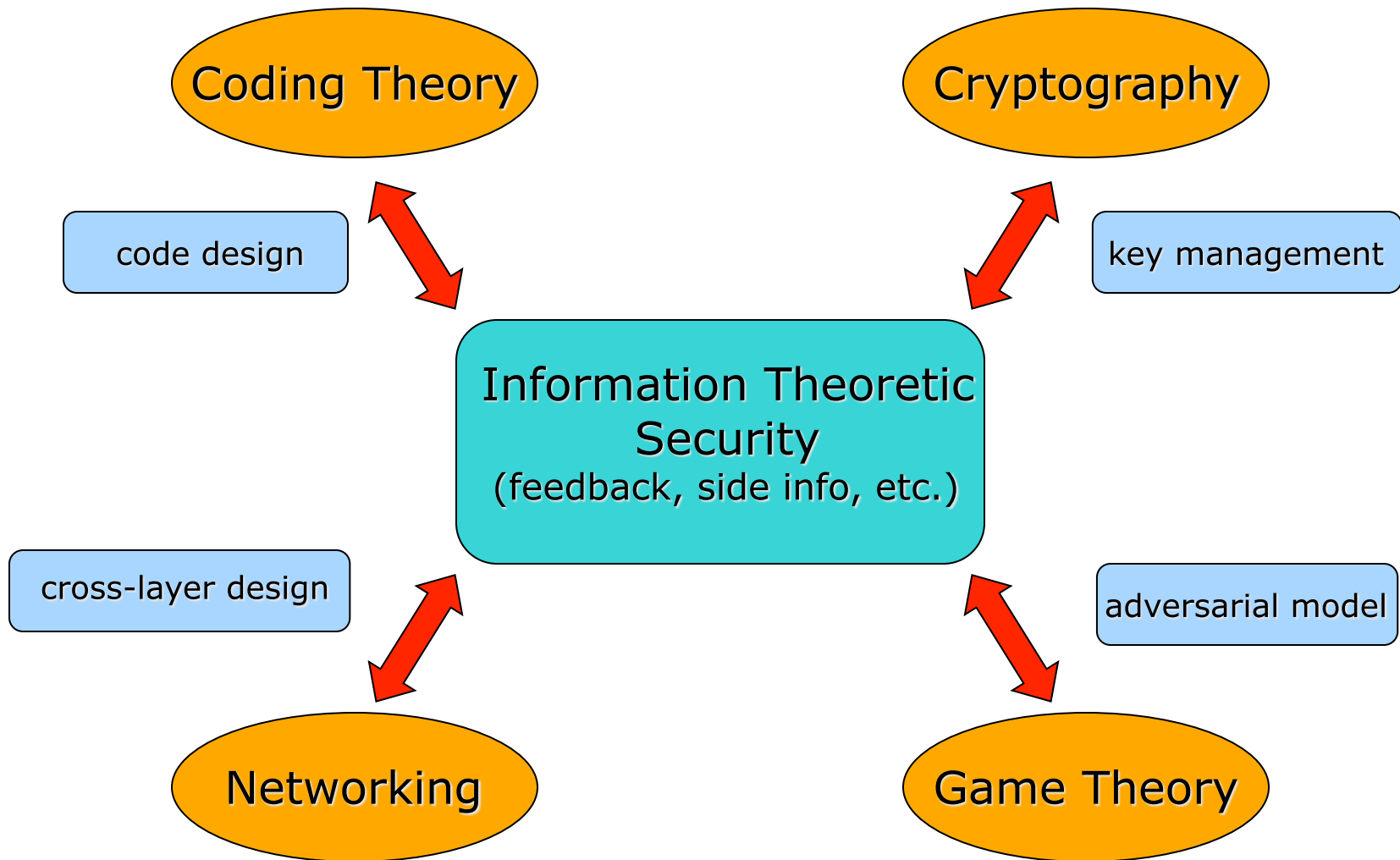


- Relay Channels [e.g., w/ Aggarwal, Sankar, Calderbank - JWCN'09 & w/ Kim - IT'11]: Source and relay cooperate to improve security.
- MIMO [e.g., w/ Liu, Liu, Shamai - IT'10]: Use of multiple transmit & receive antennas allows simultaneous secure broadcast without rate penalty.

PHY Security in Wireless Communication Nets



A Rich Area



Liang, Poor & Shamai, *Information Theoretic Security* (Now '09)

Liu & Trappe, Eds., *Securing Wireless Communications at the Physical Layer* (Springer '10)

Bloch & Barros, *Physical Layer Security* (CUP '11)

PHY Security in Wireless Communication Nets



Distributed Learning Inference in Wireless Sensor Networks

[Joint work with Joel Predd, Sanjeev Kulkarni, et al.]

Information & Inference in the Wireless PHY



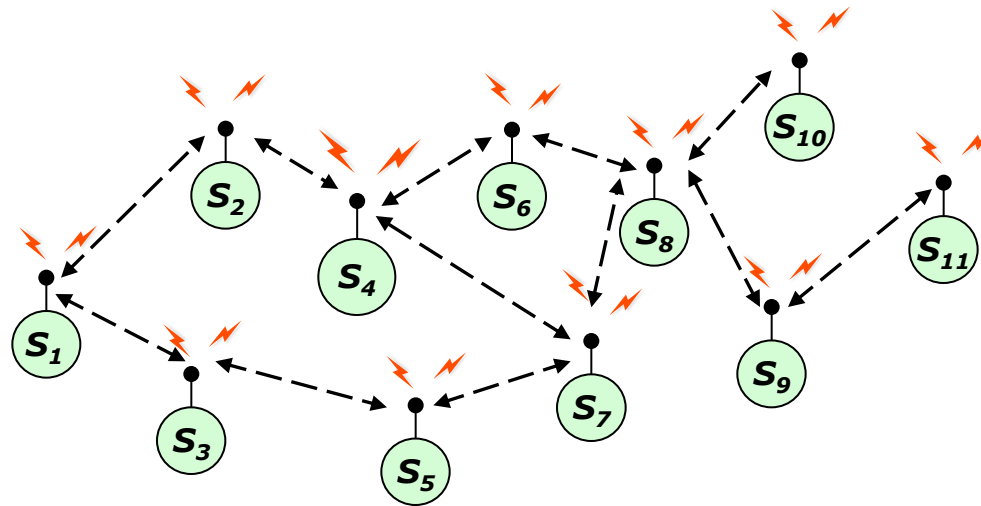
Sensor Field



Distributed Learning

A Model for Dist'd Learning in WSNs

"A distributed sampling device with a wireless interface"



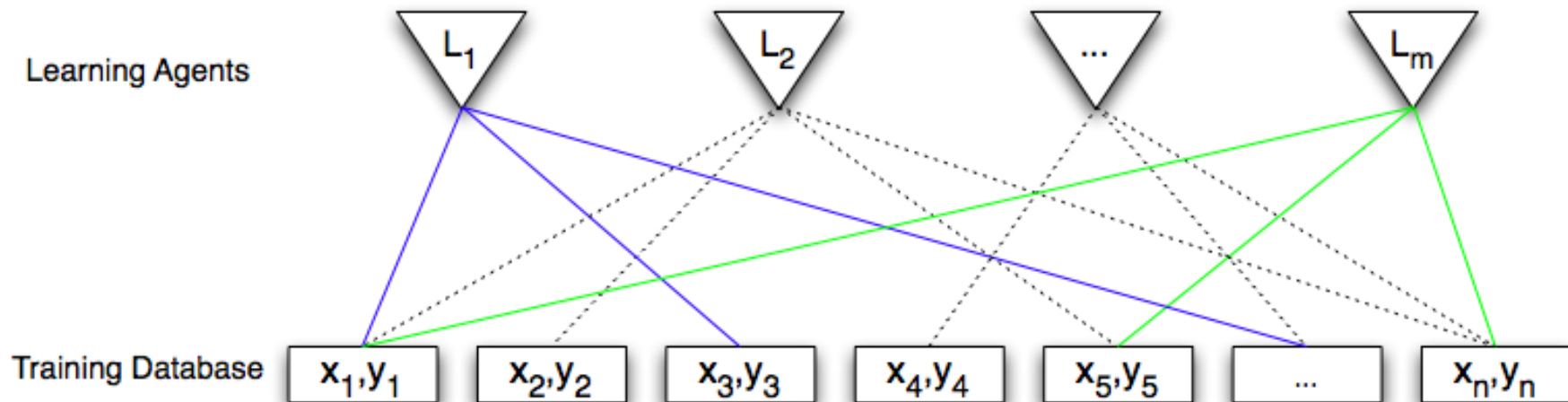
- Each sensor measures a **subset** of a large data set
- Each sensor can access all **neighboring** sensors' measured data.

Distributed Learning



A General Model

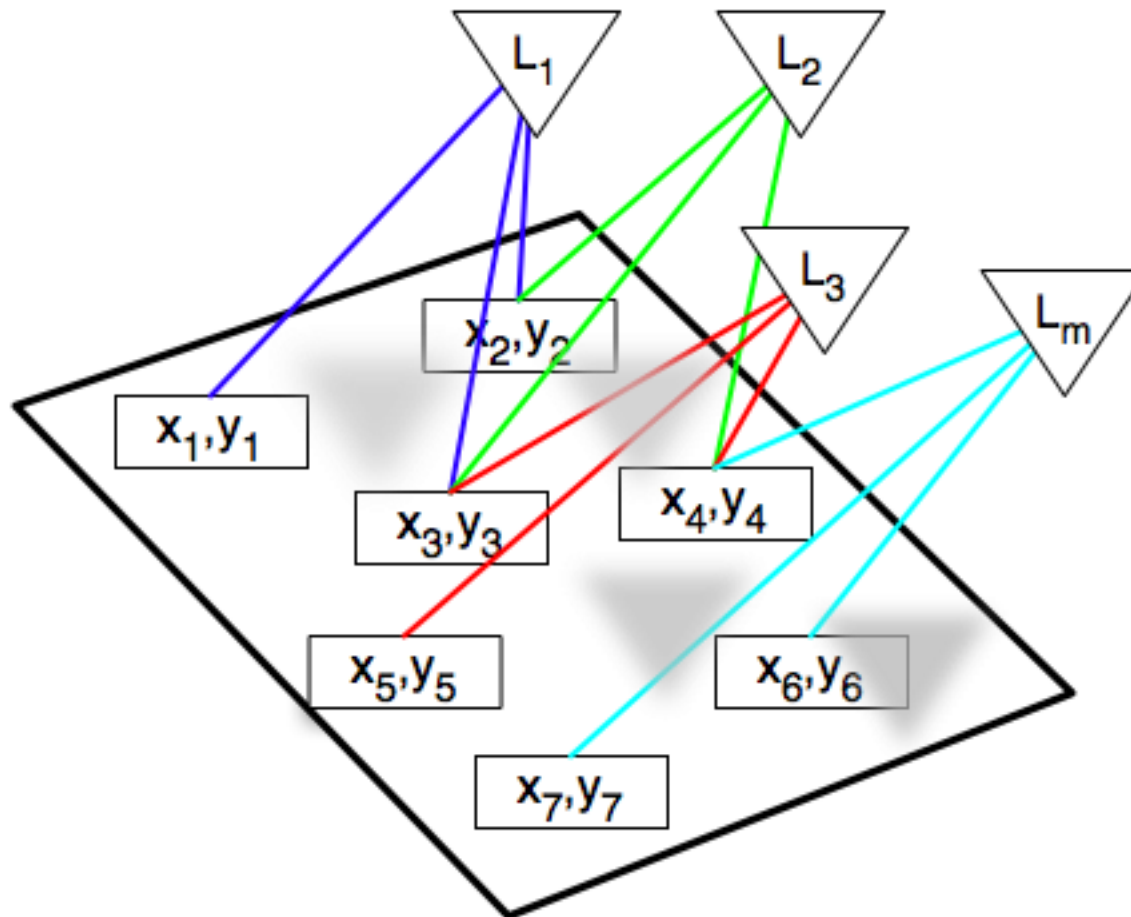
- m learning agents (i.e., sensors)
- n training data $S = \{(\mathbf{x}_i, y_i)\}_{i=1}^n$



Distributed Learning



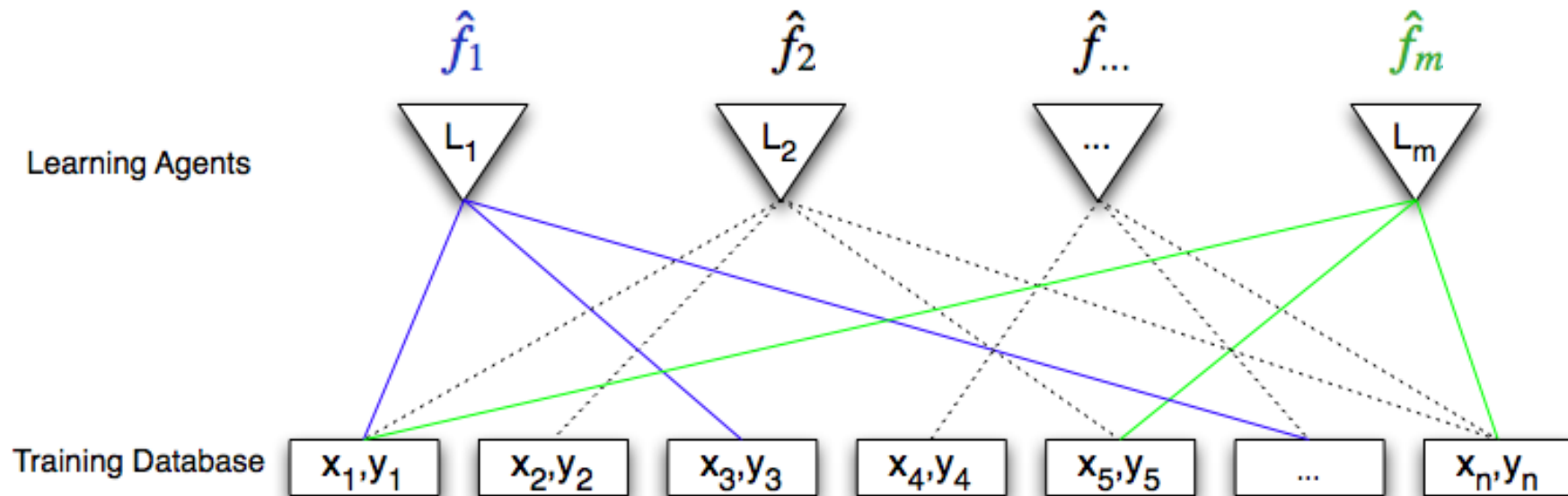
Example: Spatio-Temporal Field Estimation



Distributed Learning



"Local" Learning: A Natural Approach



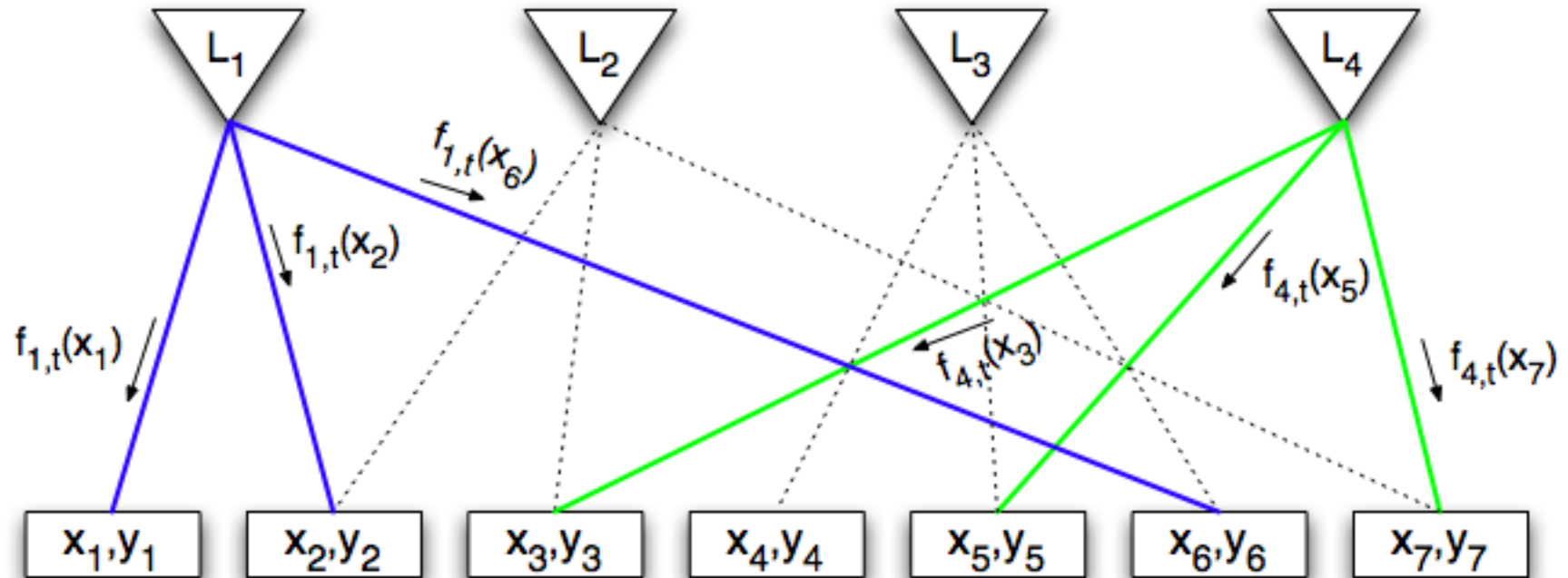
- Each learns the field with its locally available data.
- This is generally locally incoherent – e.g., $\hat{f}_1(\mathbf{x}_1) \neq \hat{f}_m(\mathbf{x}_1)$

Distributed Learning



A Collaborative Algorithm

[w/ Predd, Kulkarni, IT'09]



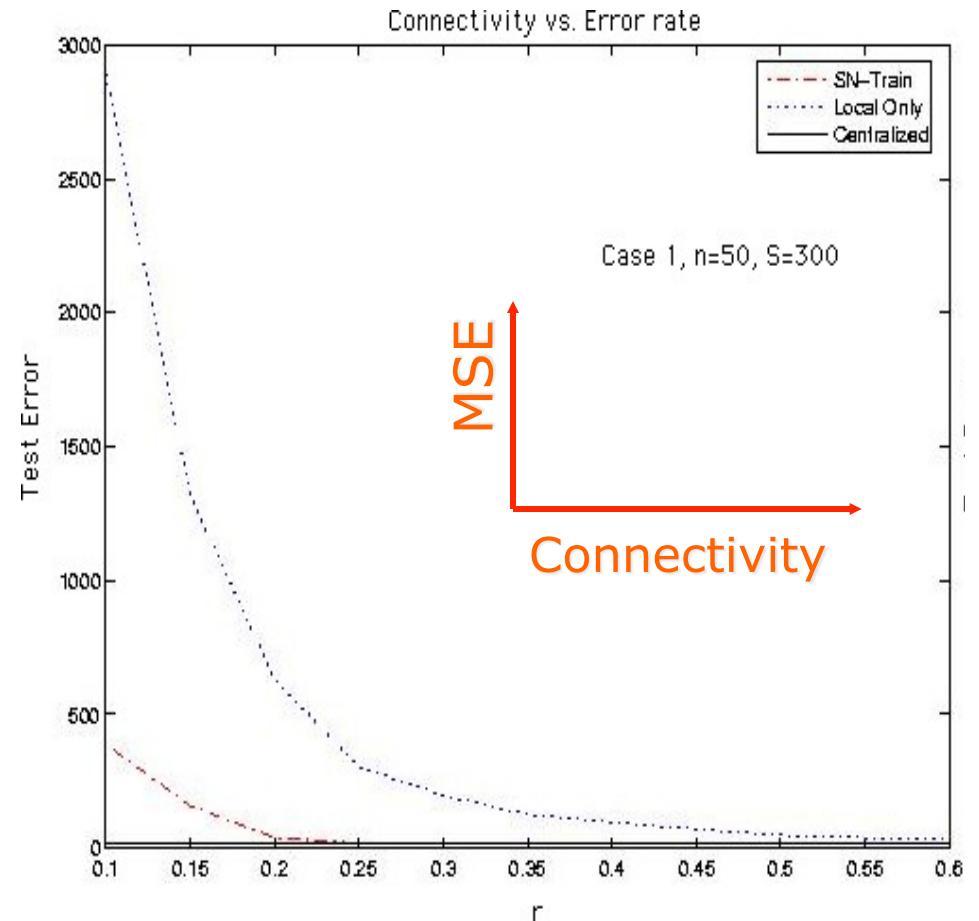
- Message passing is used to update the database.
- Nice properties & combines coherence with locality.

Distributed Learning



Experiment

- 50 sensors uniform in $[-1, 1]$
- Sensor i observes $y_i = f(x_i) + n_i$
 - $\{n_i\}$ is i.i.d. $N(0,1)$
 - regression function f is linear
 - i and j are neighbors: $|x_i - x_j| < r$
- Sensors employ linear kernel



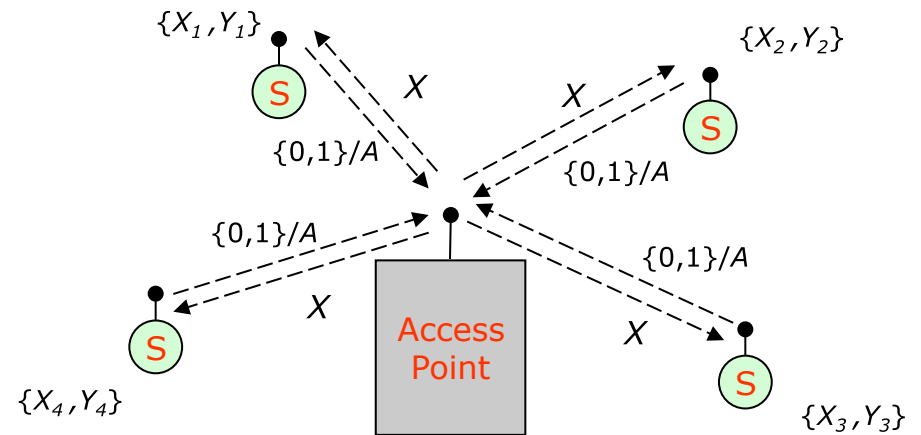
Distributed Learning



Related Results

- Consistency w. Limited Capacity

[w/ Predd, Kulkarni - IT'06]



- Judgment Aggregation

[w/ Osherson et al. - *Decision Analysis* '08, '11]

- Attribute Distributed Data

[w/ Zheng et al. - SP'11]

Distributed Learning



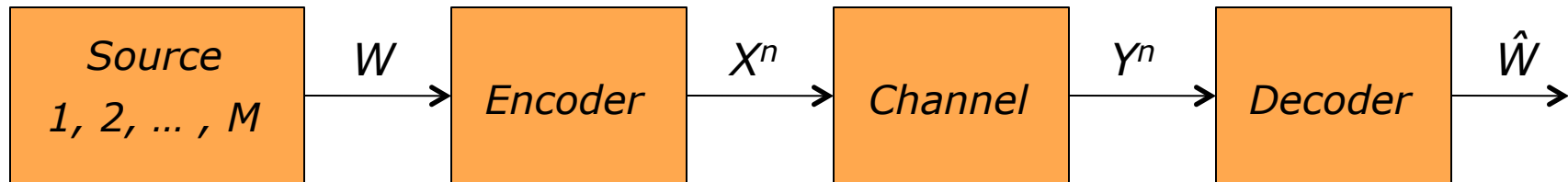
Finite-Blocklength Capacity Multimedia Information Transmission

[Joint work with Yury Polyanskiy & Sergio Verdú]

Information & Inference in the Wireless PHY



A Fundamental Problem



- (n, M, ϵ) code: $P(W \neq \hat{W}) \leq \epsilon$
- Fundamental limit: $M^*(n, \epsilon) = \max\{M: \exists \text{ an } (n, M, \epsilon) \text{ code}\}$
- Shannon: As $n \rightarrow \infty$, $\epsilon \rightarrow 0$

$$\frac{\log M^*(n, \epsilon)}{n} \rightarrow C \quad (\text{capacity})$$

- In many apps (e.g., multimedia) n and ϵ are noticeably finite.

Finite-Blocklength Capacity



Finite n and ε

[w/ Polyanskiy, Verdú, IT'10 & IT'11]

- *Bounds:*

- *Shannon-Feinstein* (1954/57); *Gallager* (1965)
- *Random coding union* (2008); *dependence testing* (2008)

- *Approximation:*

- *Strassen* (1962) – *discrete memoryless channels*
- *New bounds yield* (2008/09) – *sharper for DMCs; Gaussian; fading*

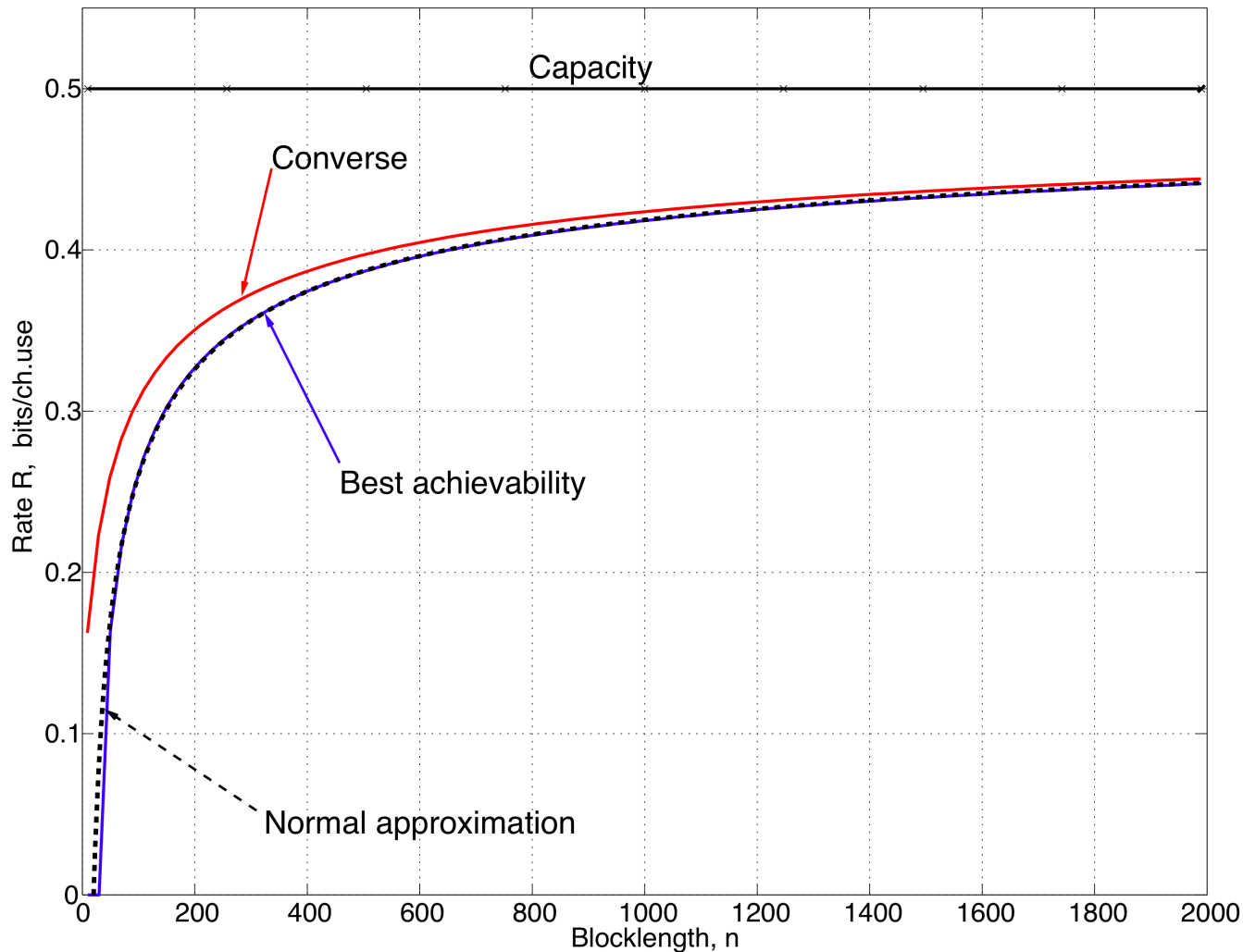
$$\log M^*(n, \varepsilon) = n C - \sqrt{nV} Q^{-1}(\varepsilon) + O(\log n)$$

$$V = \text{Var}[i(X^*, Y^*)] \quad (\text{"dispersion"})$$

Finite-Blocklength Capacity



Ex: AWGN ($SNR = 0$ dB; $\epsilon = 10^{-3}$)



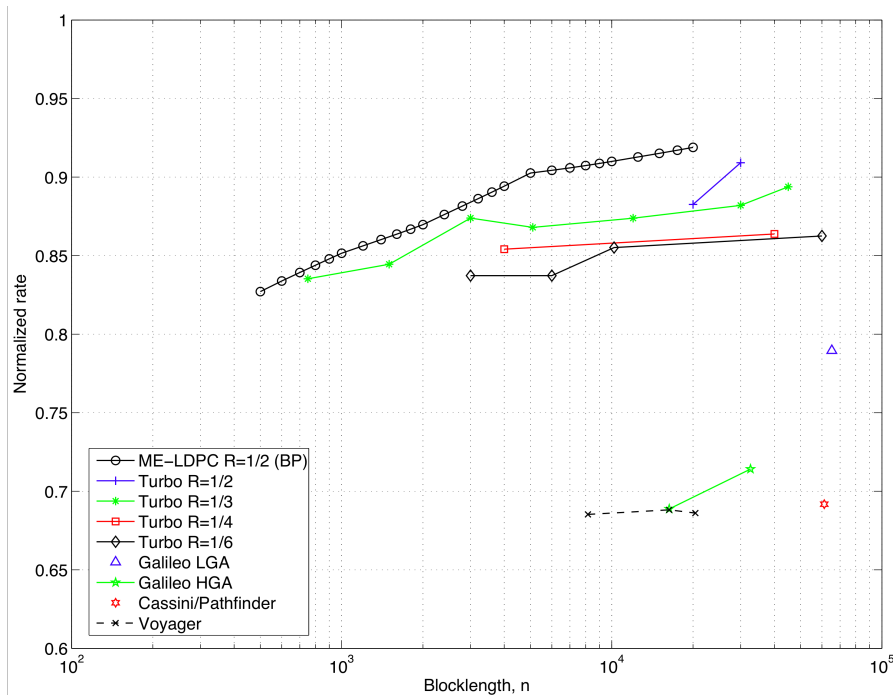
Finite-Blocklength Capacity



Some Applications

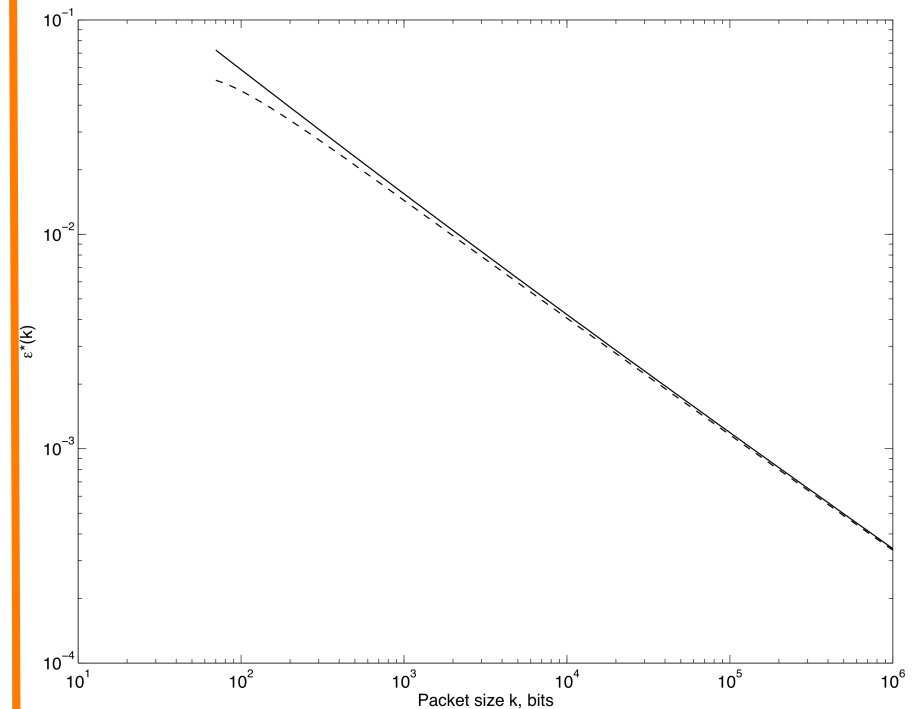
Analysis of Codes

(normalized to the approx.; $\varepsilon = 10^{-4}$)



ARQ: Optimal ε vs. n

(AWGN; SNR = 0 dB)



More generally: information theory for finite n ?

Finite-Blocklength Capacity



*Message Delivery in Small
World Networks
Social Networking
(Information & Inference)*

[Joint work with [Hazer Inaltekin](#) & [Mung Chiang](#)]

Information & Inference in the Wireless PHY



Message Delivery in Small World Social Networks

- Milgram's 1967 experiment:

$$\text{“ } \mathbb{E}[\text{ Path Length }] = 6. \text{”}$$

- Two striking conclusions:

- people are connected through short chains of acquaintances
- these chains can be found via local information

- **Analysis** can help explain this

Message Delivery in Small-World Nets

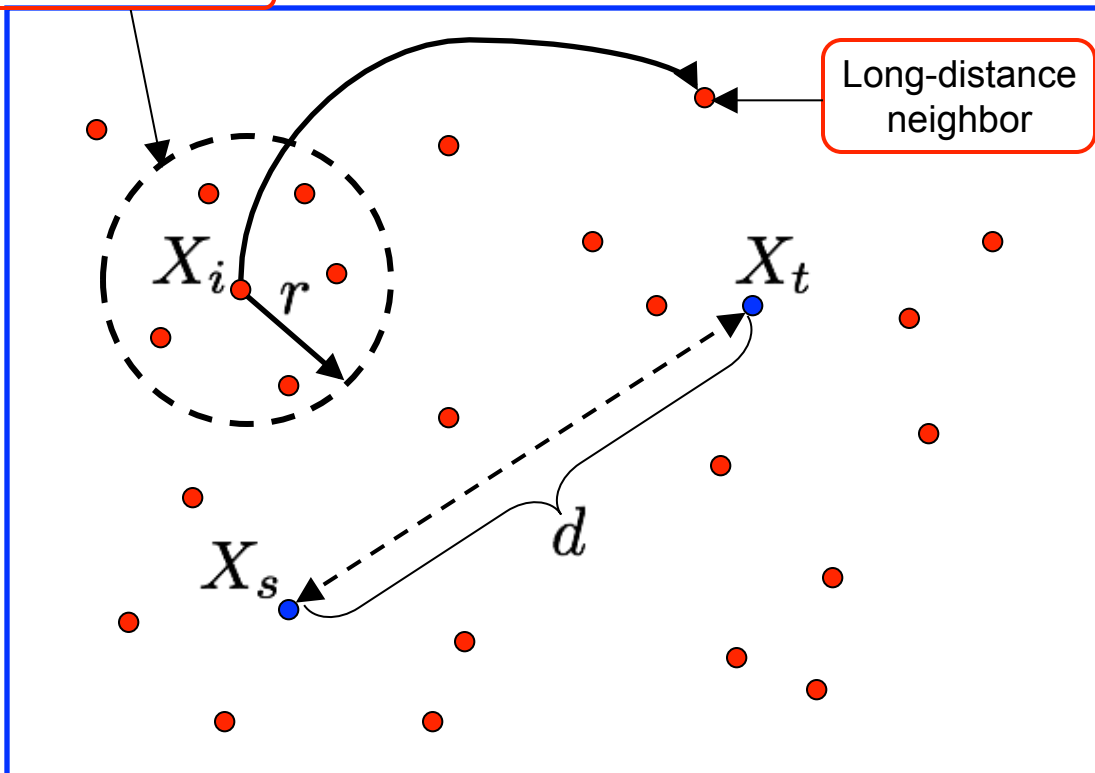


Random Geographic Graph Model

[w/ Inaltekin, Chiang – IT'10]

Local neighbors

Long-distance neighbor



- Source and target nodes are placed at arbitrary positions.
- Their separation is d .
- n other **relay** nodes are distributed uniformly over the domain.
- Each node has local communication range r .
- Each node has one long-distance neighbor.
- Greedy geographic routing.

E.g.:

- Social networks: Granovetter, *Am. J. Sociology*'78
- Ad hoc networks: Reznik, Kulkarni, Verdú, *Comm. Inf. Syst.*'04

Message Delivery in Small-World Nets



Average Message Delivery Time, $g(d)$, in the Continuum Limit ($n \rightarrow \infty$)

- Define

$$g_0 = g(0), g_1 = g(d) \text{ for } 0 < d < r,$$

$$g_k = g(d) \text{ for } (k-1)r \leq d < kr.$$

- Recursive equation:

$$g_{k+1} = 1 + \left(1 - \alpha(k-1)^2\right) g_k + \alpha \sum_{i=1}^{k-1} (2i-1) g_i \quad \alpha = \frac{\pi r^2}{R^2 - \pi r^2}$$

$R = \text{domain dimension}$

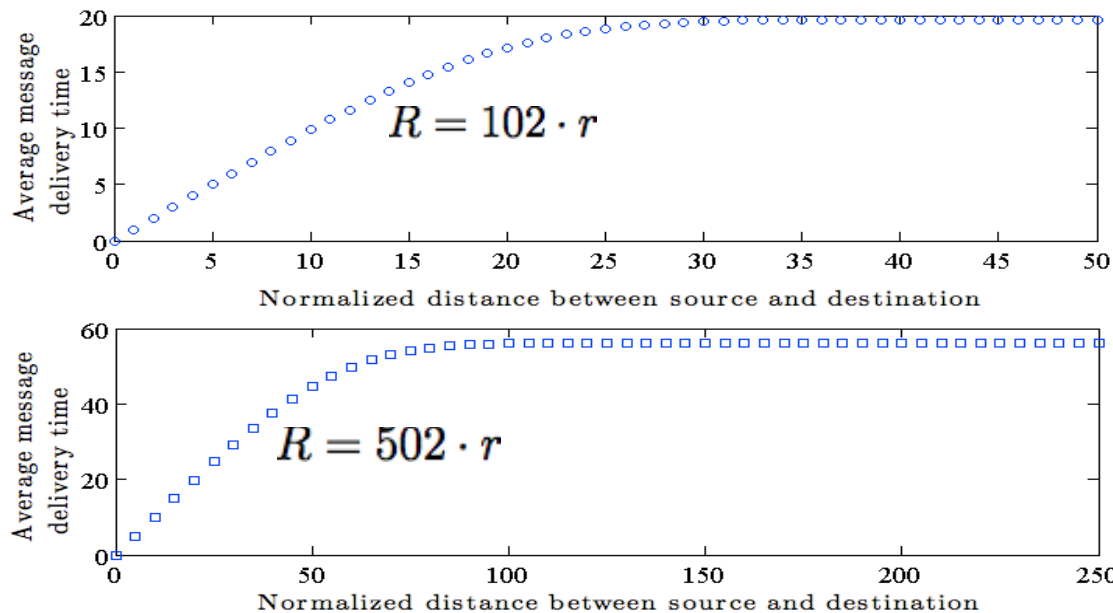
- Solution for the recursion:

$$g_{k+1} = 1 + \sum_{j=1}^k \prod_{i=1}^j \beta_i \text{ for } k \geq 1. \quad \beta_k = 1 - \alpha(k-1)^2$$

Message Delivery in Small-World Nets



Numerical Examples: Average Message Delivery Time



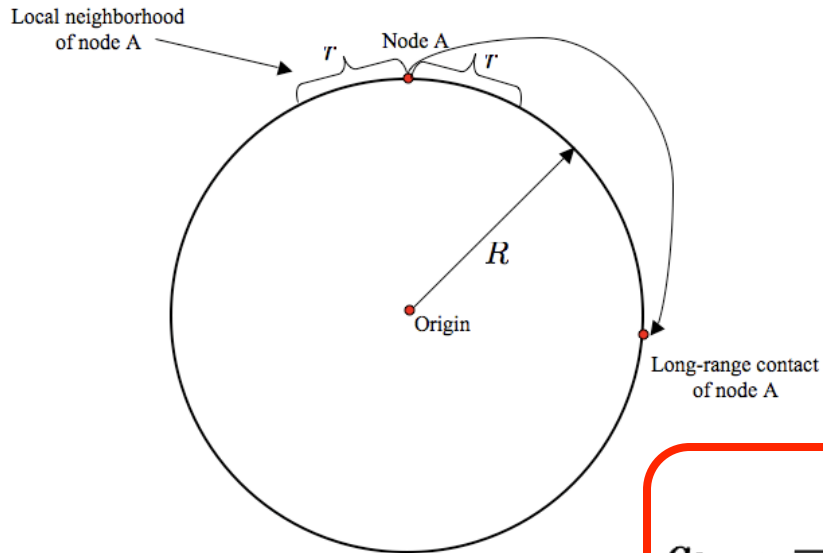
- Effects of Short-cuts on Packet Delay:
 - **short distances:** message delivery grows **linearly**
 - **long distances:** message delivery time **saturates** to a constant
 - **Observation of Travers & Milgram [Sociometry '69]:** "Chains which converge on the target principally by using geographic information reach his **hometown or surrounding areas readily**, but once there often **circulate** before entering **target's circle of acquaintances.**"

Message Delivery in Small-World Nets

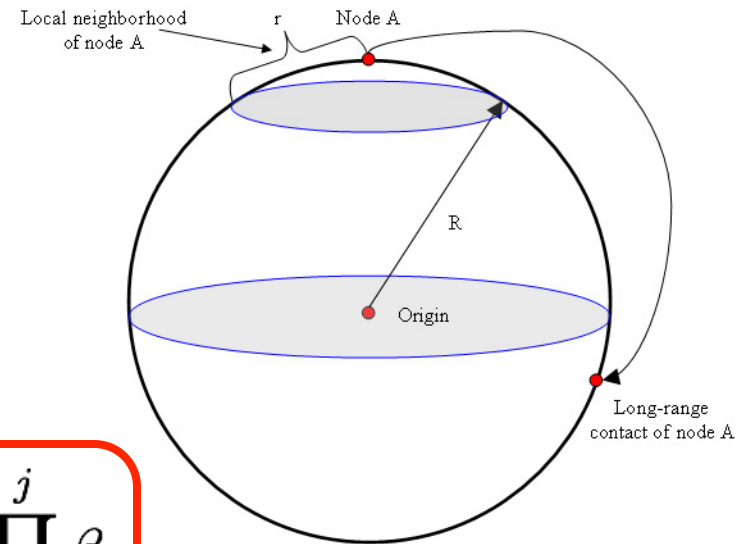


Other Network Topologies

On the Circle



On the Sphere



$$g_{k+1} = 1 + \sum_{j=1}^k \prod_{i=1}^j \beta_i$$

$$\beta_i = \frac{\pi - i\theta}{\pi - \theta}$$

$$\theta = \frac{r}{R}$$

$$\beta_i = \frac{\cos(\theta) + \cos((i-1)\theta)}{1 + \cos(\theta)}$$

$$\theta = \frac{r}{R}$$

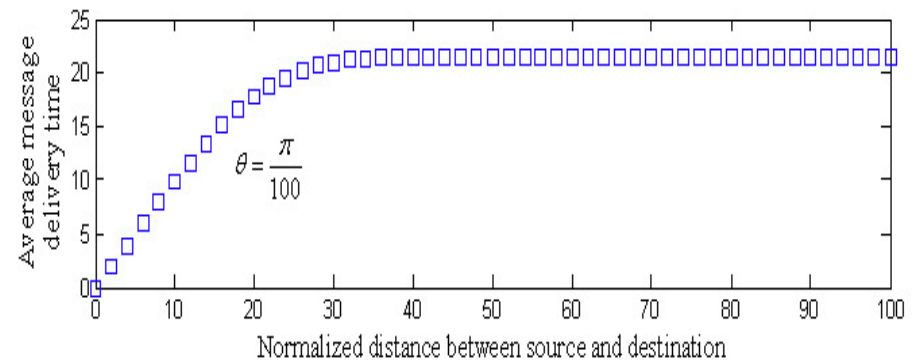
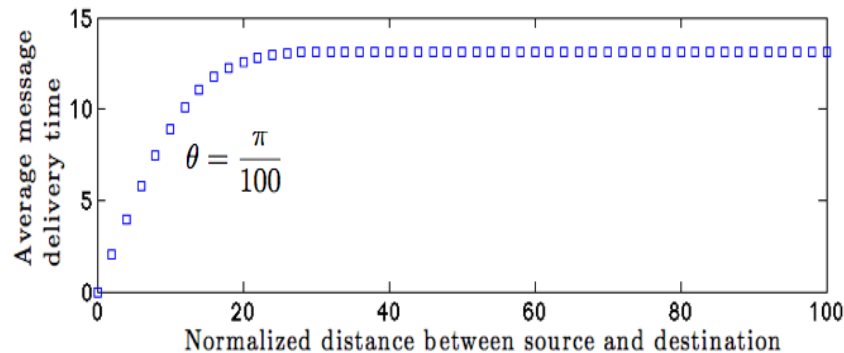
Message Delivery in Small-World Nets



Numerical Examples: Other Network Topologies

On the Circle

On the Sphere

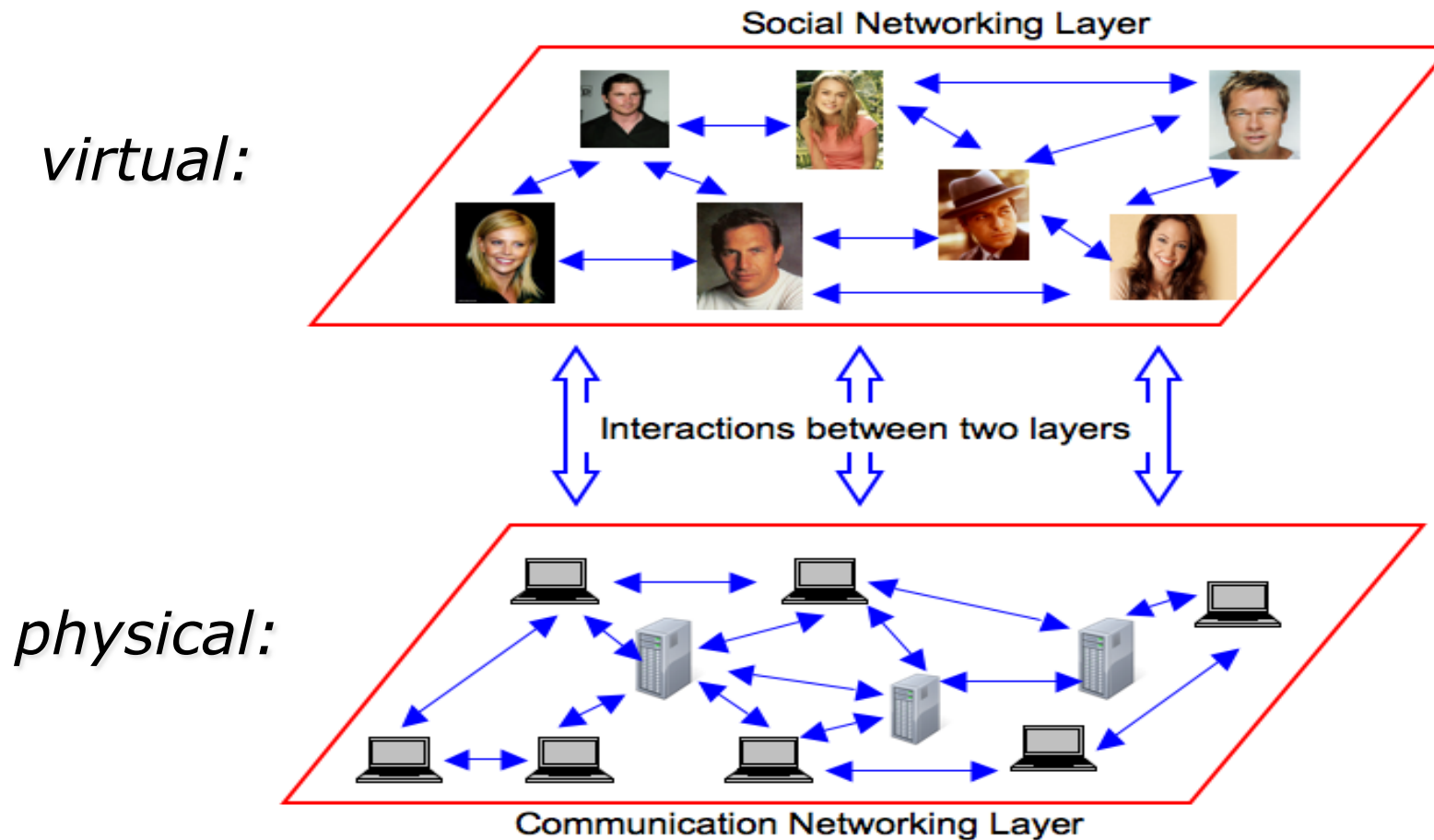


Generalization to other topologies, other connection models, etc.:
[w/ Inaltekin, Chiang – IT'10 & J. Math. Soc., to appear]

Message Delivery in Small-World Nets



Social Overlay/Communication Underlay



*Social overlay imposes new structure (e.g., **trust**).*

Message Delivery in Small-World Nets



Summary: Four Problems in the Wireless PHY Motivated by the APP

- *PHY Security in Wireless Communication Networks*

Motivated by Secure Information Transmission

- *Distributed Learning*

Motivated by Statistical Inference in Wireless Sensor Networks

- *Finite-Blocklength Capacity*

Motivated by Multimedia Information Transmission

- *Message Delivery in Small-World Networks*

Motivated by Social Networking (Information & Inference)

Information & Inference in the Wireless PHY



The background of the slide is a solid dark blue color. Overlaid on this background are several overlapping, wavy white lines that create a sense of depth and movement, resembling a stylized landscape or a series of ripples. The lines are more prominent in the upper and right portions of the slide.

Thank you!