

# Modeling and Analysis of Interconnected Systems in the Presence of Uncertainty

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How should we simulate the  
Global Financial System?

Ideally, what should be the  
“Architecture of the Financial System?”

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## Modeling of Interconnected Systems

Dynamical View: Space of Random Trajectories

Not Equilibrium View

Tearing  $\implies$  Open Sub-systems;

Modeled to extent possible

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# Theory of Interconnection of Uncertain Systems

Composition of Stochastic Kernels to create Joint Probability distribution

Probability Theory allows us to compose “Heterogeneous” stochastic systems

Special kind of interconnection: **feedback** through a stochastic dynamical system

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**Stability of Interconnected System** is obtained through “lossless” interconnection of dissipative systems

**Feedback Interconnection of Economic System and Financial System**

For example, we could study the stability of a Complex Power System using the methodology.

Part of Modeling: Parameterization and Identification of Parameters from Data

Measurement and data generation

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## Model Uncertainty and Robustness

### Example:

Recent work in Finance to deal with Model Uncertainty (Knightian Uncertainty)

Also, Levy Processes replacing Brownian Motion

De Finetti's ideas of Coherence

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## Lessons from Power Systems

In Cascading Power Transmission, component outages propagate non-locally

Using data from many Cascading Failure Simulations, one can construct a Markovian Influence Graph (which is not the Grid topology) where Cascading Power Outages propagate locally

## Model Reduction

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Blackout data suggest that the frequency of large blackouts is governed by a power law

Power law makes the risk of large blackouts consequential, and is consistent with the Power System operating at a critical point due to steady increase in loading, maximum use of the grid and other factors

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Real-Time Pricing in Electrical Power Systems  
may cause Instability due to Feedback

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# Section 1

## Architecture: Definition

Architecture is the organization of Distributed Algorithms in hardware and software for performing complex heterogeneous functions.

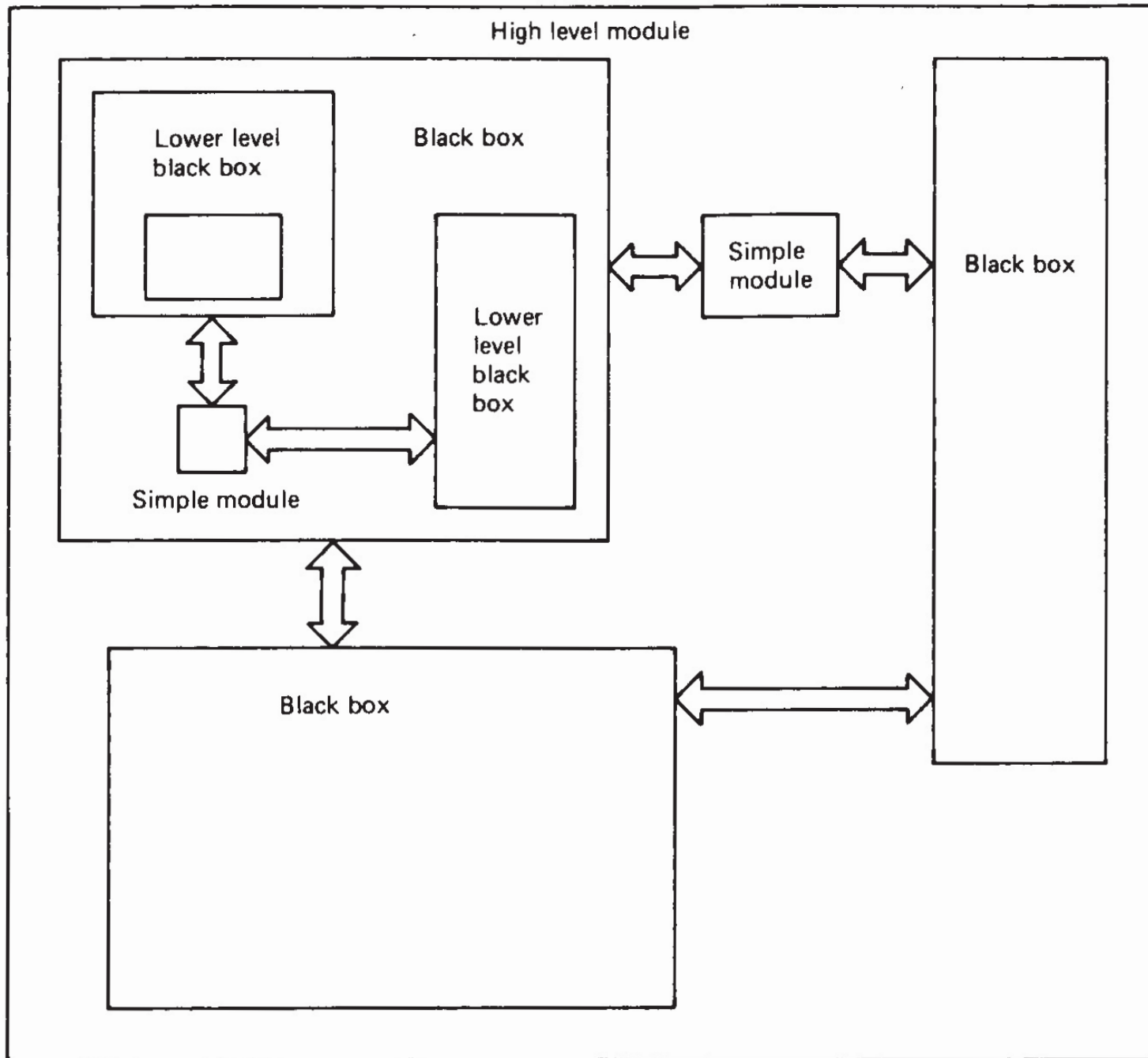
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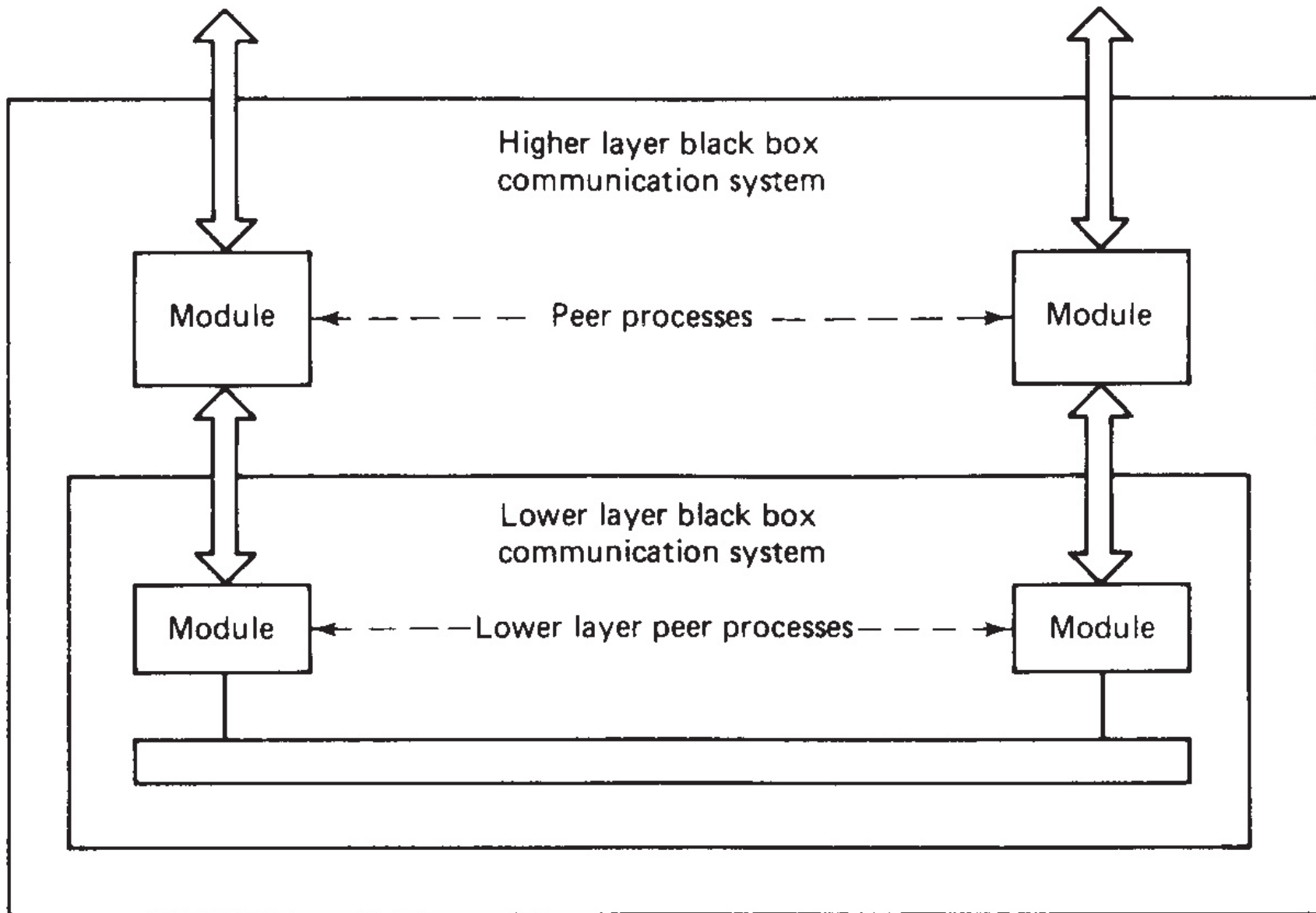
## Section 2

### Architecture of the Internet

“Layering” the Financial System

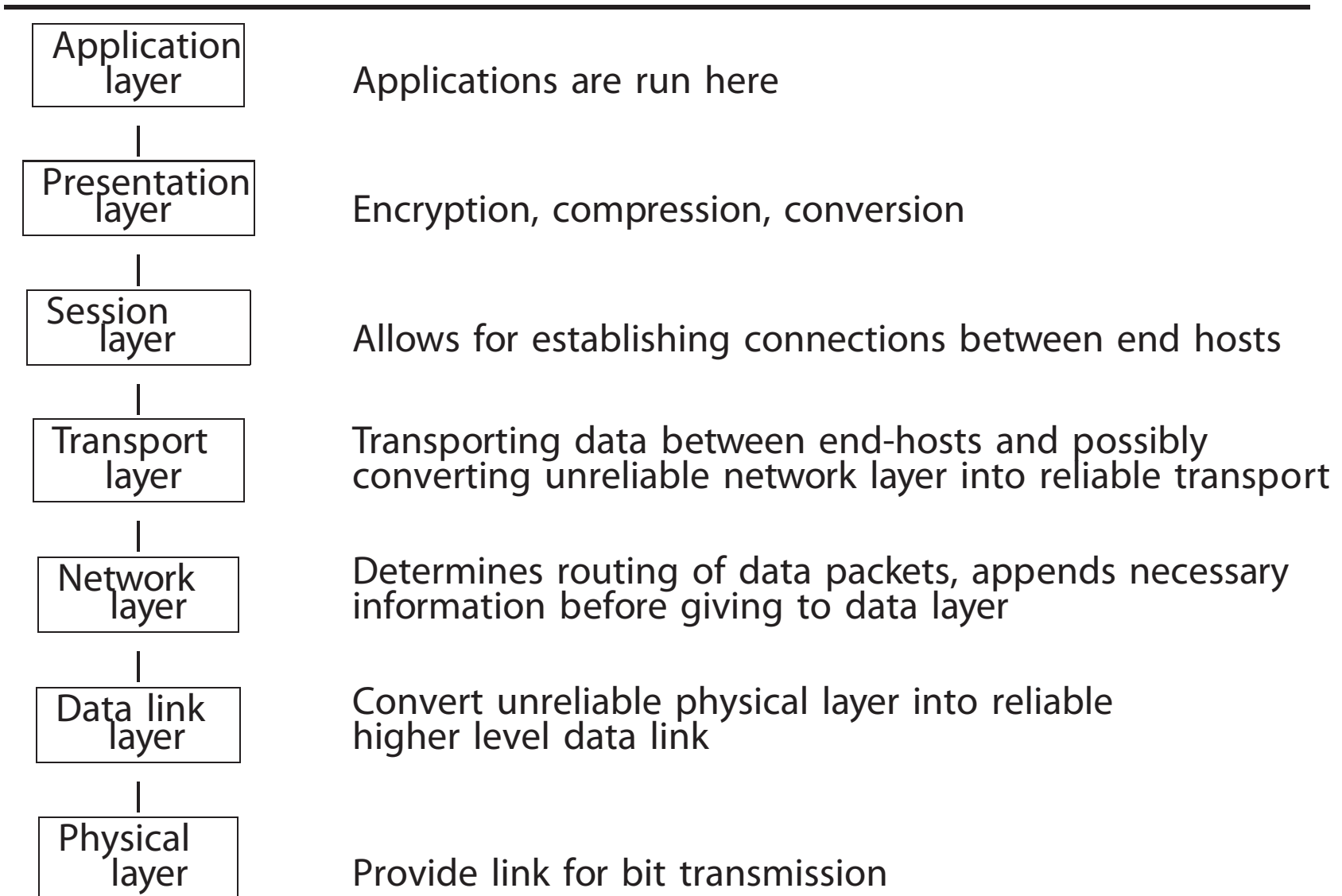
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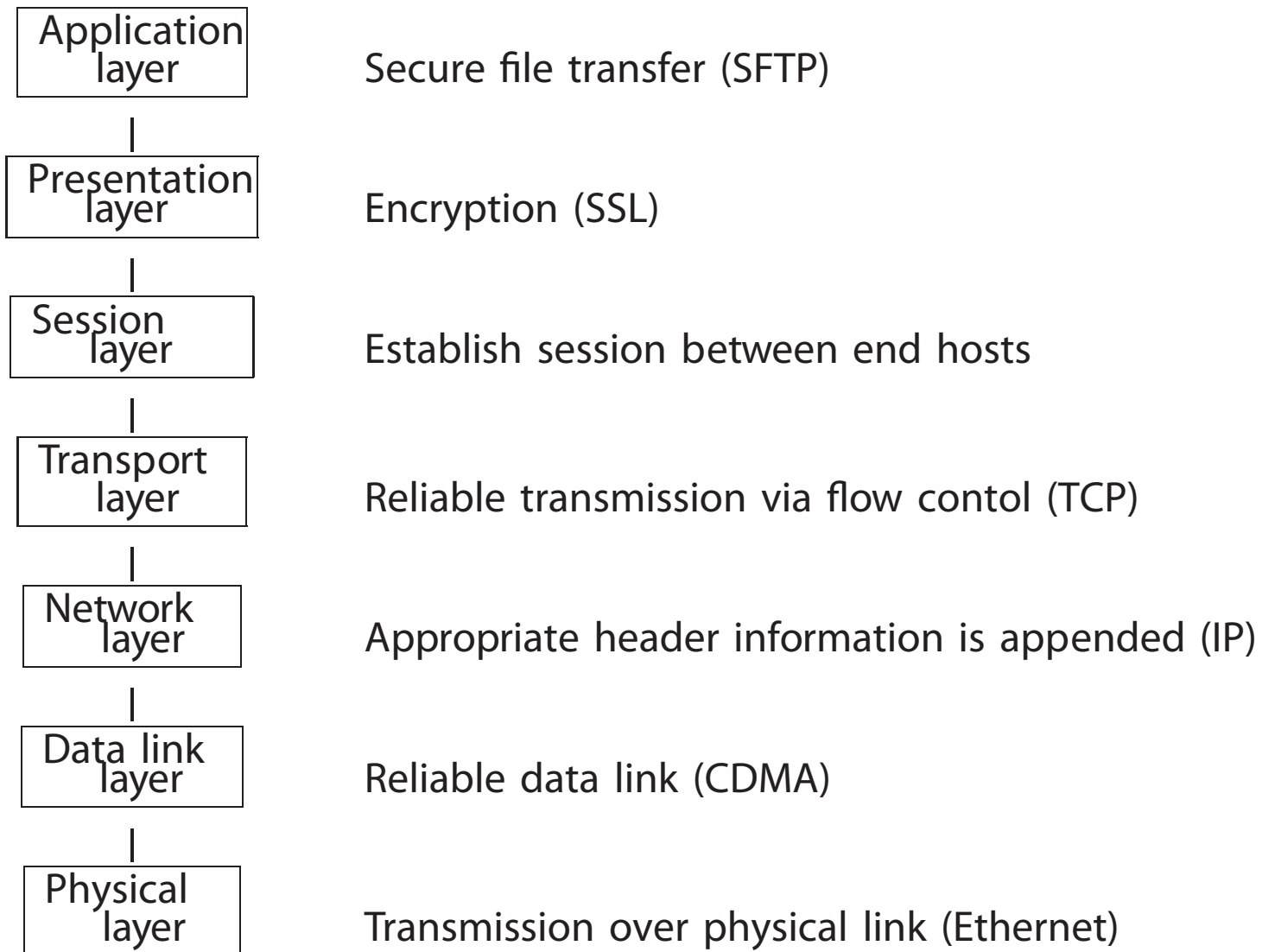
## Layered Architecture

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## Layered Architecture: An Example

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## References

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### (1) Two papers

- “The Design Philosophy of The DARPA Internet Protocols,” by D. Clark
- “Architectural Considerations for a New Generation of Protocols,” by D. Clark and D. L. Tennenhouse

### (2) Data Networks, by Bertsekas–Gallager

- Chapter 1.3

### (3) Expository article

- MIT Tech Review, Dec 05/Jan 06 issue, article, p. 62–69
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## Features of Layered Architectures for Performing Heterogeneous Functions

- Structure of Layers: by analogy.
    - Two adjacent layers are “conditionally independent” give the interface linking the two layers
  - Interconnection view
  - Partial order on layers
  - Regulation in Financial systems  $\equiv$  Interfaces
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→ Higher order layers are “quotients” of lower order layers equipped with an equivalence relation.

For example, in Vision, where image is defined as a function  $f : \mathcal{M} \subset \mathbb{R}^2 \rightarrow \mathbb{R}$ , the equivalence relation is induced by a group action, the semi-direct, direct product of the Euclidian Group on  $\text{dom}(f) \subset \mathbb{R}^2$ , appropriate group on  $\text{Range}(f)$  and the gauge group on the space of Images, resulting in a quotient

(cf: the work of Stefano Soatto)

→ Quotient is often finite: Symbols

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- Layered Structure depends on functions that the system is required to perform and properties the system should possess (for example, robustness of performance against uncertainties).
  - “Autonomy of Layers” vs. Adaptation Possibilities
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## Section 3

# What the Frog's Eye Tells the Frog's Brain

J.Y. Lettvin, H.R. Maturanat, W.S. McCulloch II, and W.H. Pitts

Research Laboratory for Electronics

MIT

*Proc. Inst. of Radio Engr., 1959.*

## Conclusions

What are the consequences of this work? Fundamentally, it shows that the eye speaks to the brain in a language already highly organized and interpreted, instead of transmitting some more or less accurate copy of the distribution of light on the receptors. As a crude analogy, suppose that we have a man watching the clouds and reporting them to a weather station. If he is using a code, and one can see his portion of the sky too, then it is not difficult to find out what he is saying.

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# Methodological Advances

Statistical Inference and Optimization: High-dimensional Statistics

(Parameter space  $\gg$  size of data)

Compressed Sensing

Manifold Learning: High-dimensional data often lies on or near low-dimensional manifold

Integration of Heterogeneous Data Sets

Structure in Data: Number of connected components, homology groups

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## Methodological Advances (continued)

Data Assimilation: Bayesian Inference

Equivalent to Free-Energy Minimization

Pattern Recognition (Empirical Risk Minimization)

Clustering

Data to Models

Network Vulnerability and Security

Systemic Risk

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# MicroEconomics vs. MacroEconomics

Need:

Sufficiently rich behavioral modeling of heterogeneous interacting economic agents of Micro level, leading to emergent phenomena at Macro level

Gibbs Measures and Phase Transitions

Modeling of Human Behavior

Linguistic Description: Minimalist View of Noam Chomsky

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