



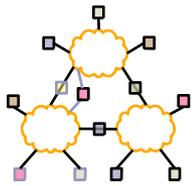
CE693: Adv. Computer Networking

L-13 Network Topology

Acknowledgments: Lecture slides are from the graduate level Computer Networks course taught by Srinivasan Seshan at CMU. When slides are obtained from other sources, a reference will be noted on the bottom of that slide. A full list of references is provided on the last slide.

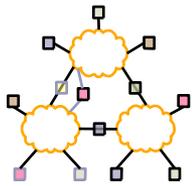


Today's Lecture



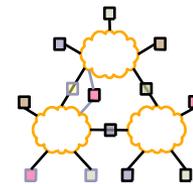
- Structural generators
- Power laws
- HOT graphs
- Assigned reading
 - On Power-Law Relationships of the Internet Topology
 - A First Principles Approach to Understanding the Internet's Router-level Topology

Outline



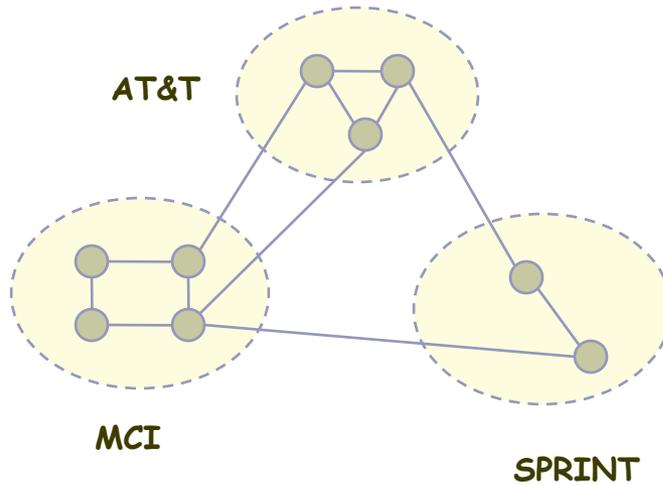
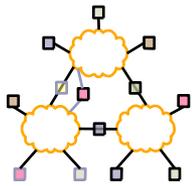
- **Motivation/Background**
- Power Laws
- Optimization Models

Why study topology?

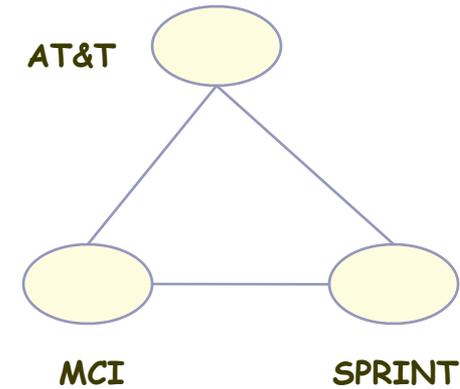


- **Correctness** of network protocols typically independent of topology
- **Performance** of networks critically dependent on topology
 - e.g., convergence of route information
- Internet **impossible** to replicate
- **Modeling of topology** needed to generate test topologies

Internet topologies

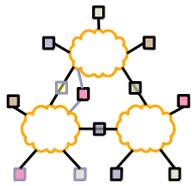


Router level



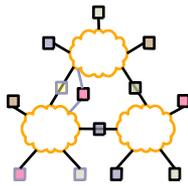
Autonomous System (AS) level

More on topologies..

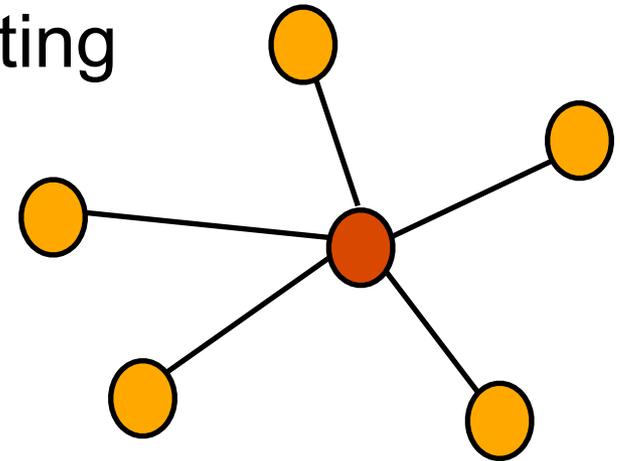


- Router level topologies reflect **physical connectivity** between nodes
 - Inferred from tools like *traceroute* or well known public measurement projects like Mercator and Skitter
- AS graph reflects a **peering relationship** between two providers/clients
 - Inferred from inter-domain routers that run BGP and public projects like Oregon Route Views
- Inferring both is difficult, and often **inaccurate**

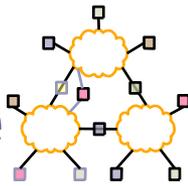
Hub-and-Spoke Topology



- Single hub node
 - Common in enterprise networks
 - Main location and satellite sites
 - Simple design and trivial routing
- Problems
 - Single point of failure
 - Bandwidth limitations
 - High delay between sites
 - Costs to backhaul to hub

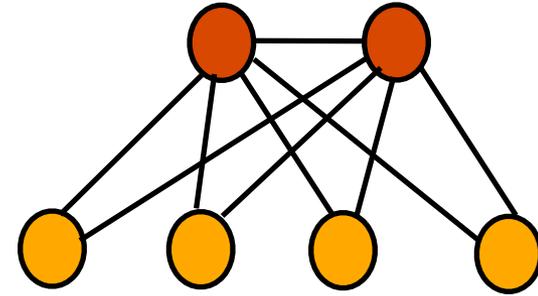


Simple Alternatives to Hub-and-Spoke



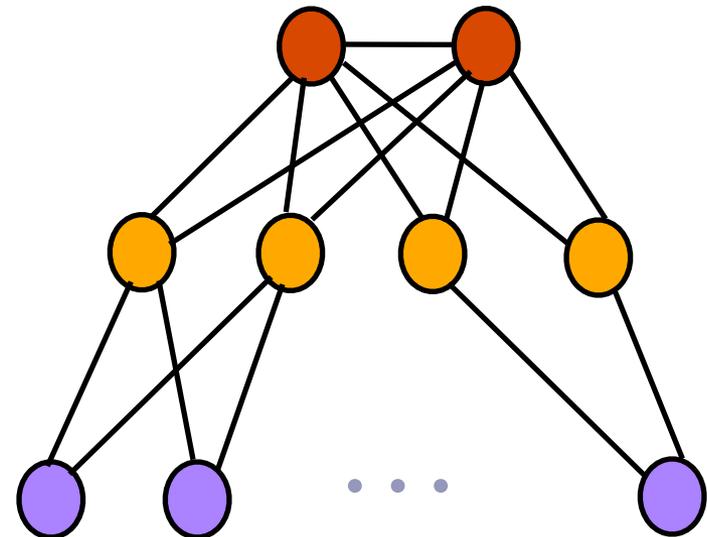
- Dual hub-and-spoke

- Higher reliability
- Higher cost
- Good building block

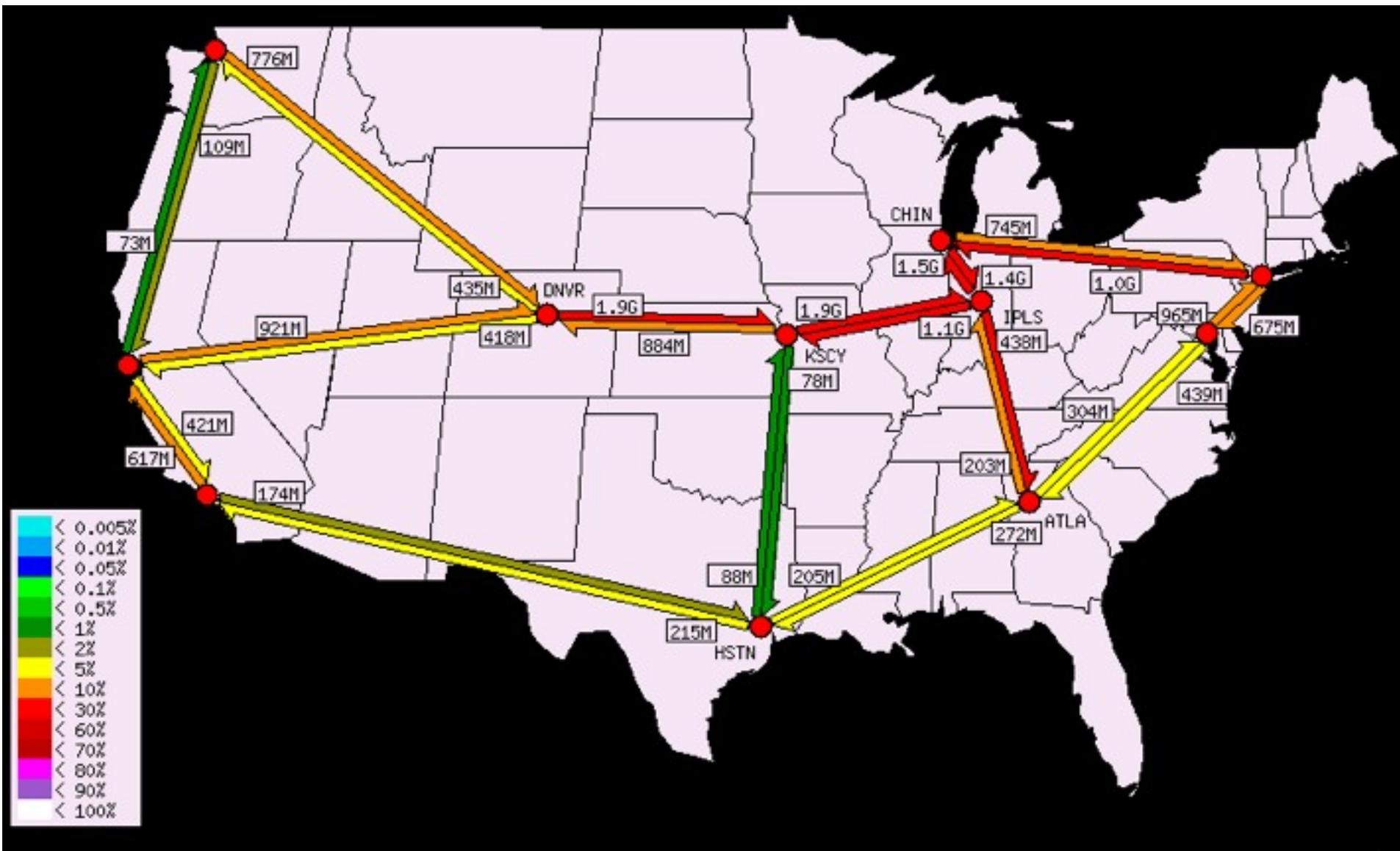
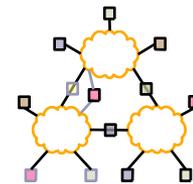


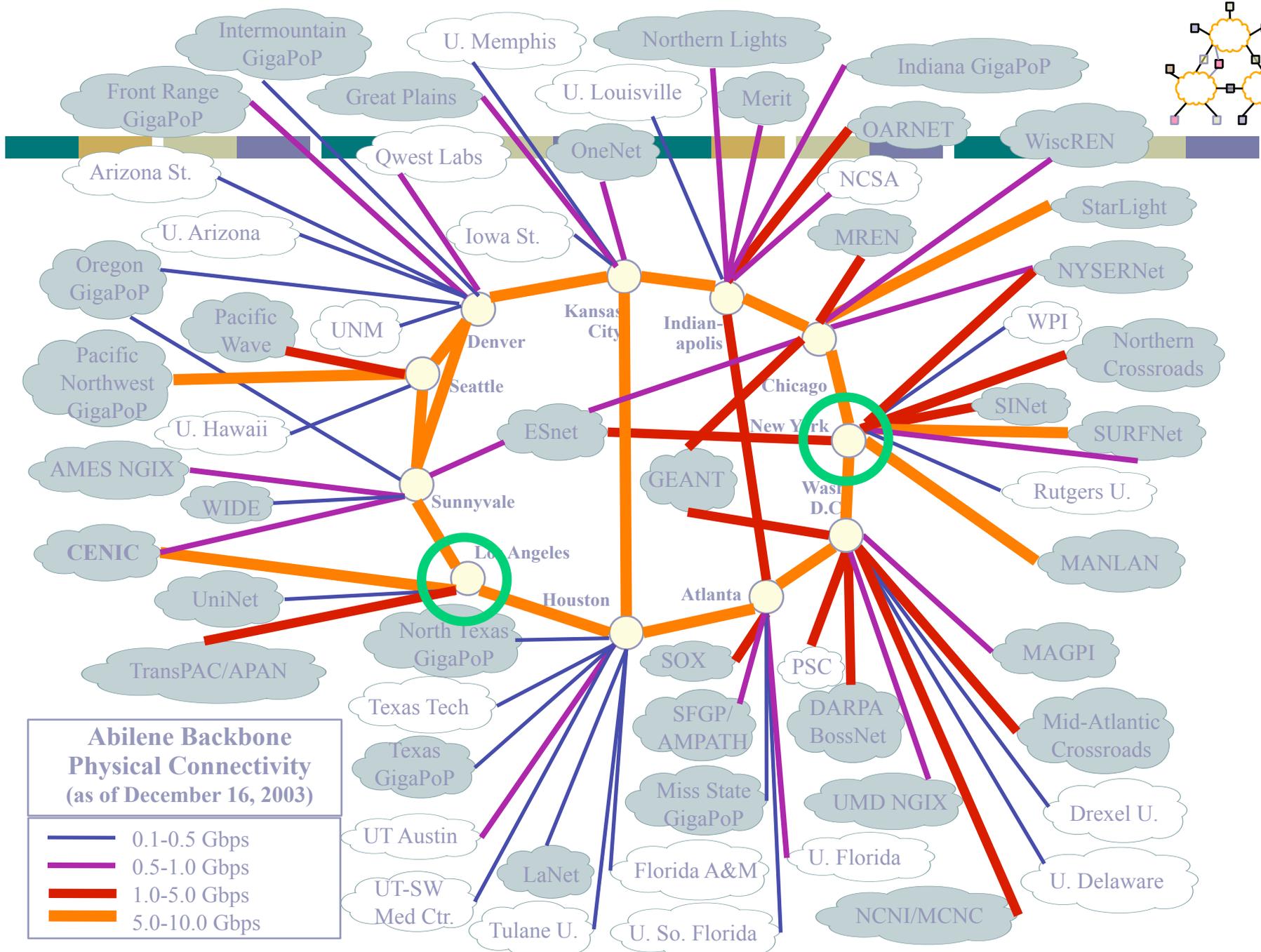
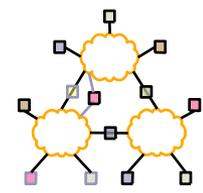
- Levels of hierarchy

- Reduce backhaul cost
- Aggregate the bandwidth
- Shorter site-to-site delay



Abilene Internet2 Backbone

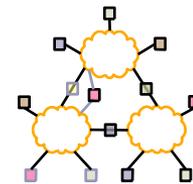




Abilene Backbone Physical Connectivity (as of December 16, 2003)

- 0.1-0.5 Gbps
- 0.5-1.0 Gbps
- 1.0-5.0 Gbps
- 5.0-10.0 Gbps

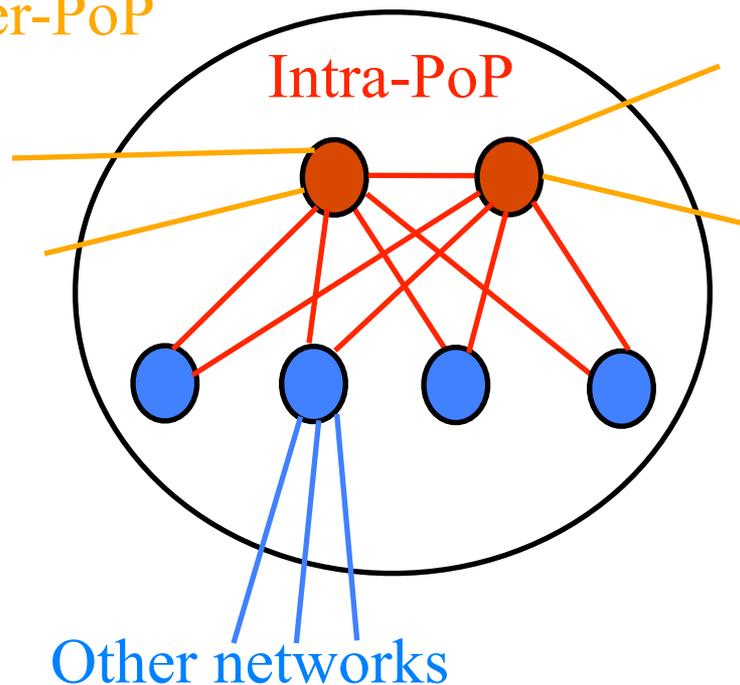
Points-of-Presence (PoPs)



- Inter-PoP links
 - Long distances
 - High bandwidth
- Intra-PoP links
 - Short cables between racks or floors
 - Aggregated bandwidth
- Links to other networks
 - Wide range of media and bandwidth

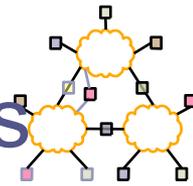
Inter-PoP

Intra-PoP



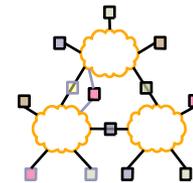
Other networks

Deciding Where to Locate Nodes and Links



- **Placing Points-of-Presence (PoPs)**
 - Large population of potential customers
 - Other providers or exchange points
 - Cost and availability of real-estate
 - Mostly in major metropolitan areas
- **Placing links between PoPs**
 - Already fiber in the ground
 - Needed to limit propagation delay
 - Needed to handle the traffic load

Trends in Topology Modeling



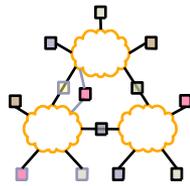
Observation

- Long-range links are expensive
- Real networks are not random, but have obvious hierarchy
- Internet topologies exhibit power law degree distributions (Faloutsos et al., 1999)
- Physical networks have hard technological (and economic) constraints.

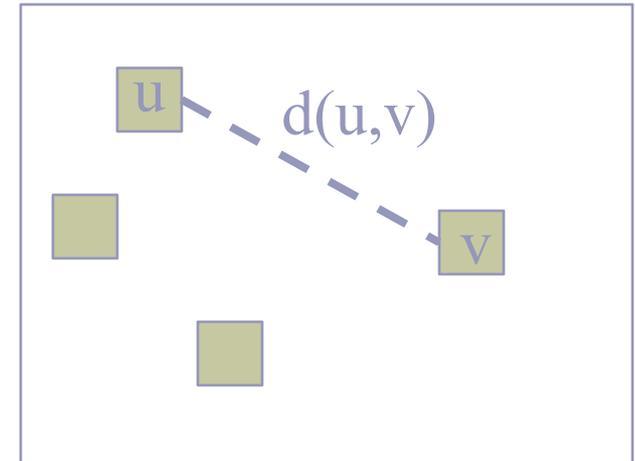
Modeling Approach

- Random graph (Waxman88)
- Structural models (GT-ITM Calvert/Zegura, 1996)
- Degree-based models replicate power-law degree sequences
- Optimization-driven models topologies consistent with design tradeoffs of network engineers

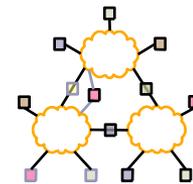
Waxman model (Waxman 1988)



- Router level model
- Nodes placed at random in 2-d space with dimension L
- Probability of edge (u,v) :
 - $ae^{-d/(bL)}$, where d is Euclidean distance (u,v) , a and b are constants
- Models locality

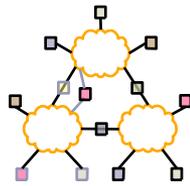


Real world topologies

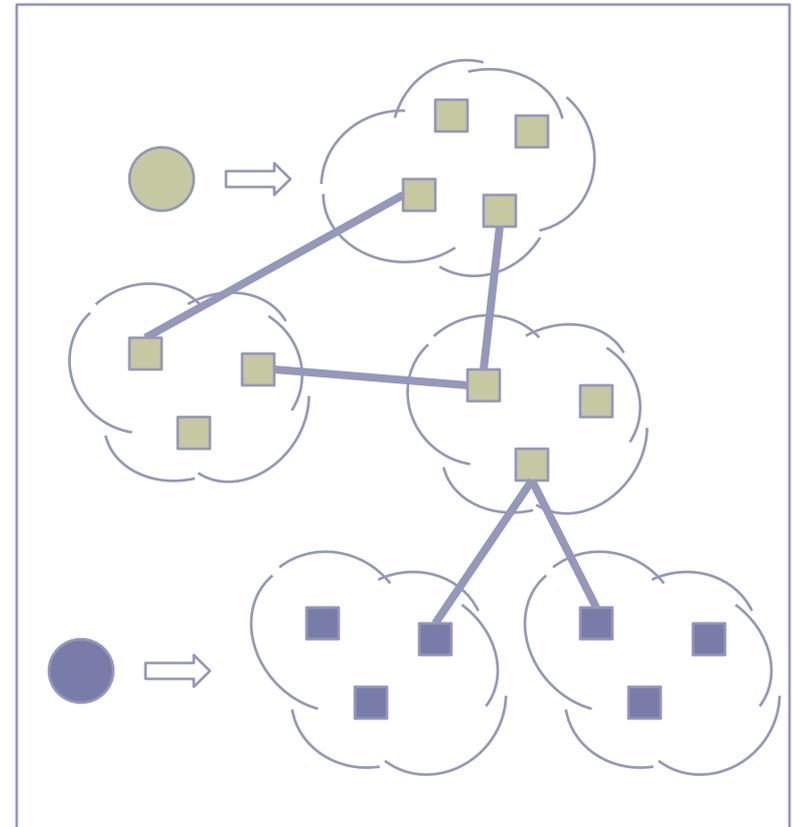


- Real networks exhibit
 - Hierarchical structure
 - Specialized nodes (transit, stub..)
 - Connectivity requirements
 - Redundancy
- Characteristics incorporated into the Georgia Tech Internetwork Topology Models (GT-ITM) simulator (E. Zegura, K. Calvert and M.J. Donahoo, 1995)

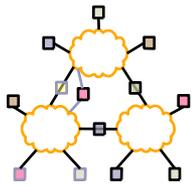
Transit-stub model (Zegura 1997)



- Router level model
- Transit domains
 - placed in 2-d space
 - populated with routers
 - connected to each other
- Stub domains
 - placed in 2-d space
 - populated with routers
 - connected to transit domains
- Models hierarchy



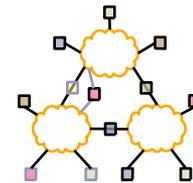
So...are we done?



- No!
- In 1999, Faloutsos, Faloutsos and Faloutsos published a paper, demonstrating **power law relationships** in Internet graphs
- Specifically, the **node degree distribution** exhibited power laws

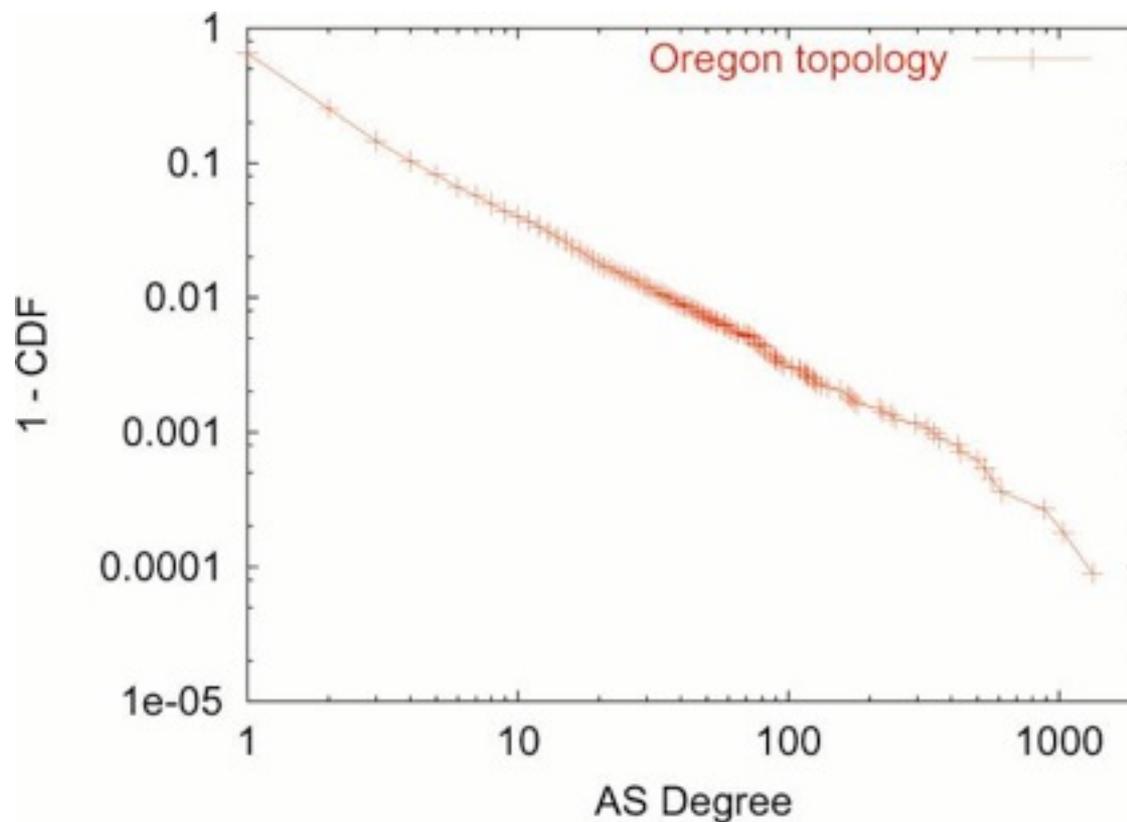
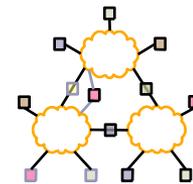
That Changed Everything.....

Outline



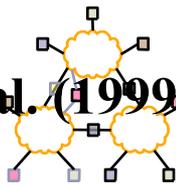
- Motivation/Background
- **Power Laws**
- Optimization Models

Power laws in AS level topology



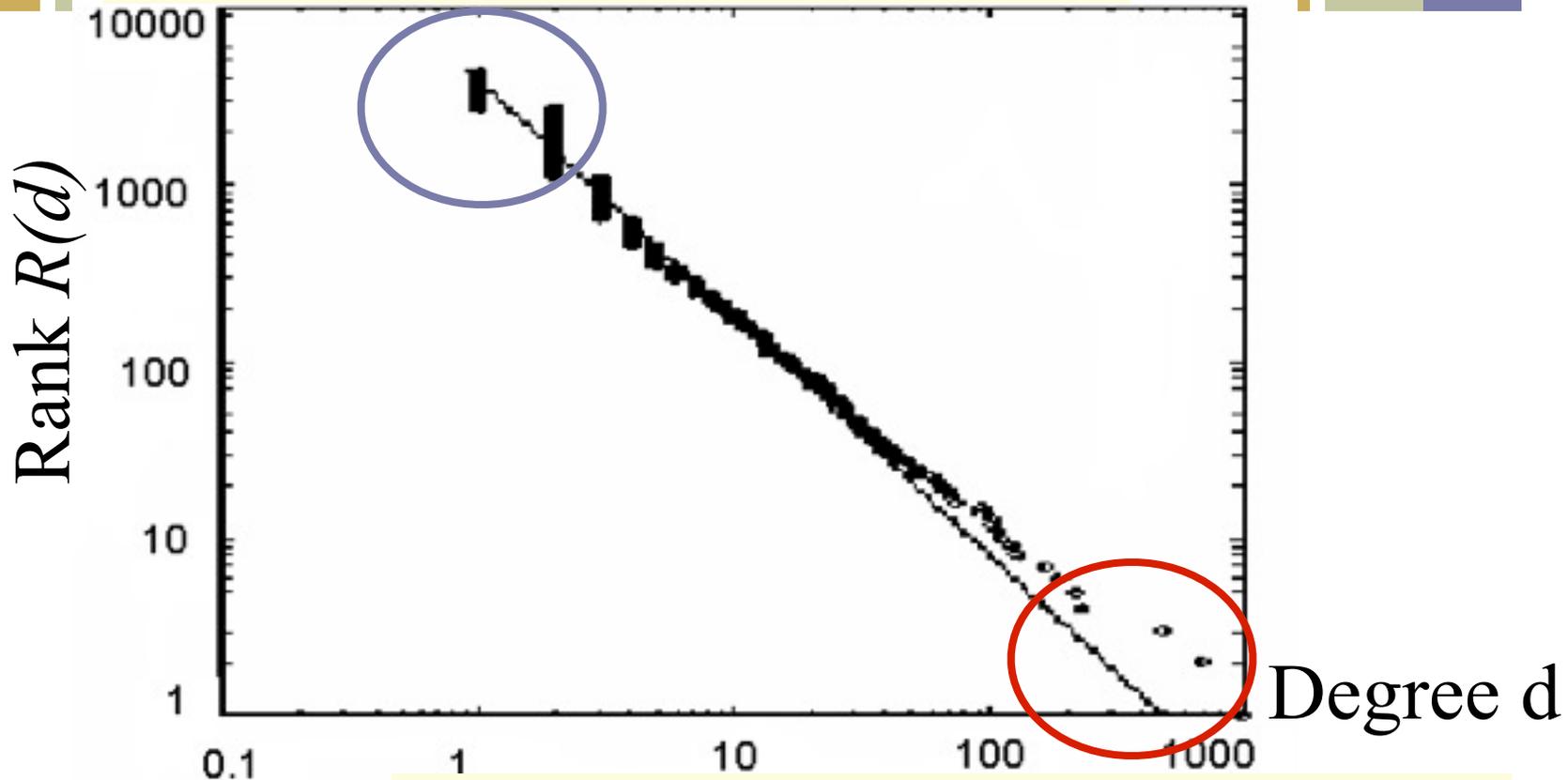
Power Laws and Internet Topology

Source: Faloutsos et al. (1999)



Most nodes have few connections

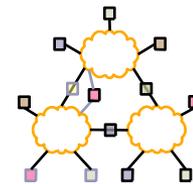
$$R(d) = P(D > d) \times \#nodes$$



A few nodes have lots of connections

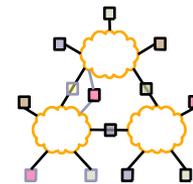
- Router-level graph & Autonomous System (AS) graph
- Led to active research in *degree-based* network models

GT-ITM abandoned..



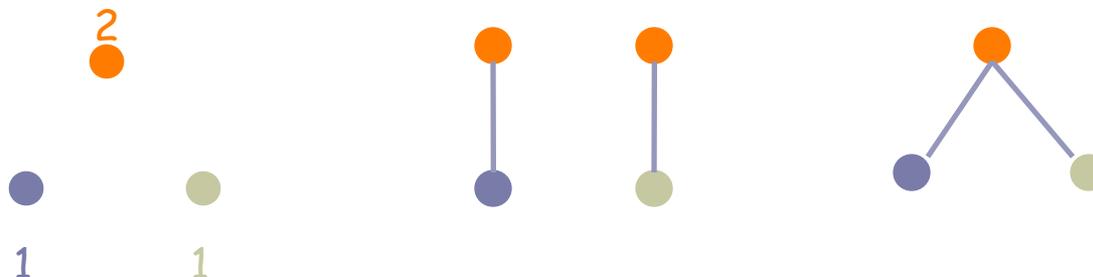
- GT-ITM did not give power law degree graphs
- New topology generators and explanation for power law degrees were sought
- Focus of generators to match degree distribution of observed graph

Power law random graph (PLRG)



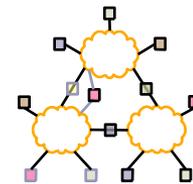
- Operations

- assign degrees to nodes drawn from power law distribution
- create k_v copies of node v ; k_v degree of v .
- randomly match nodes in pool
- aggregate edges



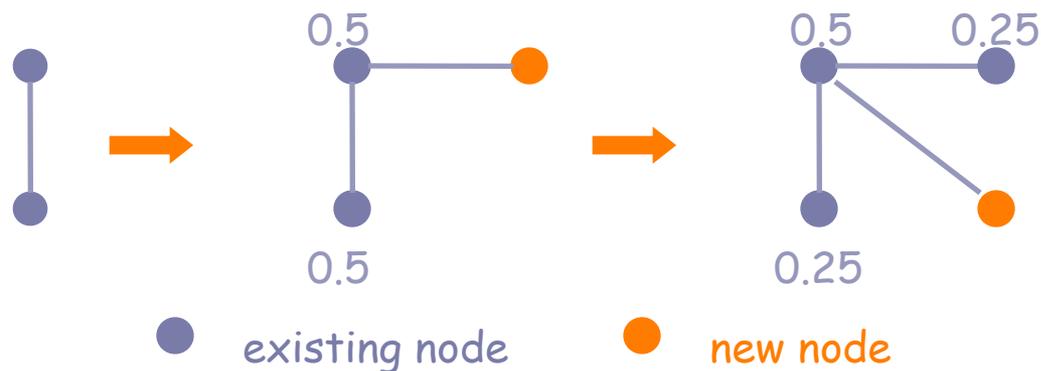
may be disconnected, contain multiple edges, self-loops

- contains unique giant component for right choice of parameters



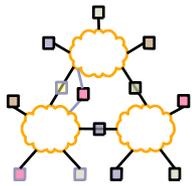
Barabasi model: fixed exponent

- incremental growth
 - initially, m_0 nodes
 - step: add new node i with m edges
- linear preferential attachment
 - connect to node i with probability $k_i / \sum k_j$

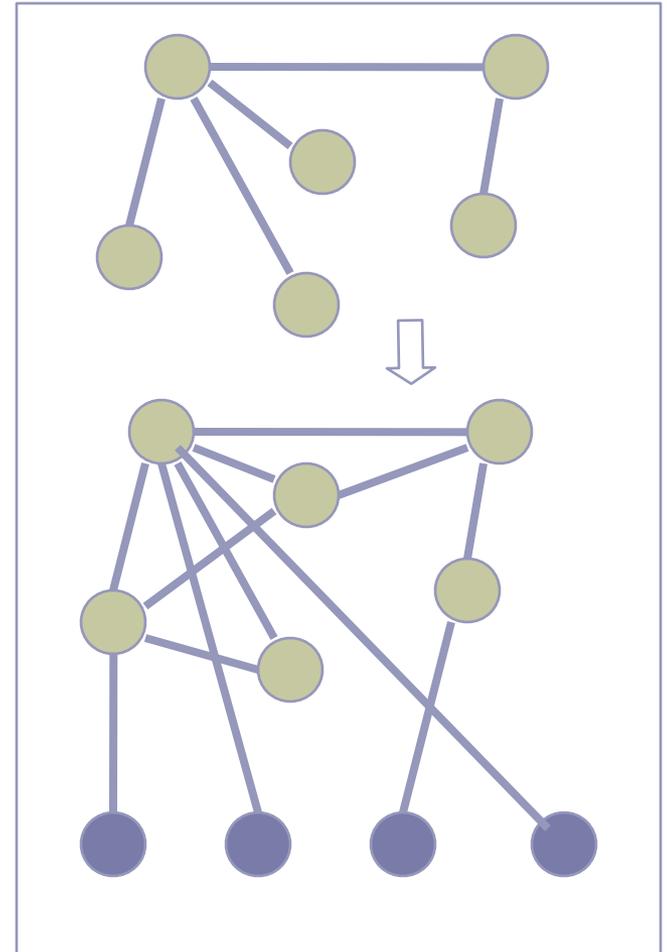


may contain multi-edges, self-loops

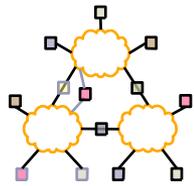
Inet (Jin 2000)



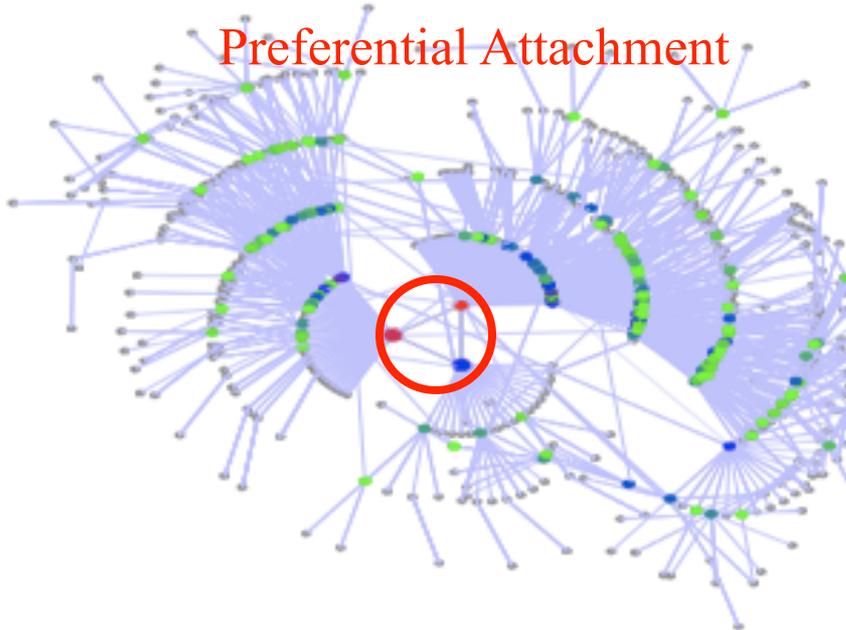
- Generate degree sequence
- Build spanning tree over nodes with degree larger than 1, using preferential connectivity
 - randomly select node u not in tree
 - join u to existing node v with probability $d(v)/\sum d(w)$
- Connect degree 1 nodes using preferential connectivity
- Add remaining edges using preferential connectivity



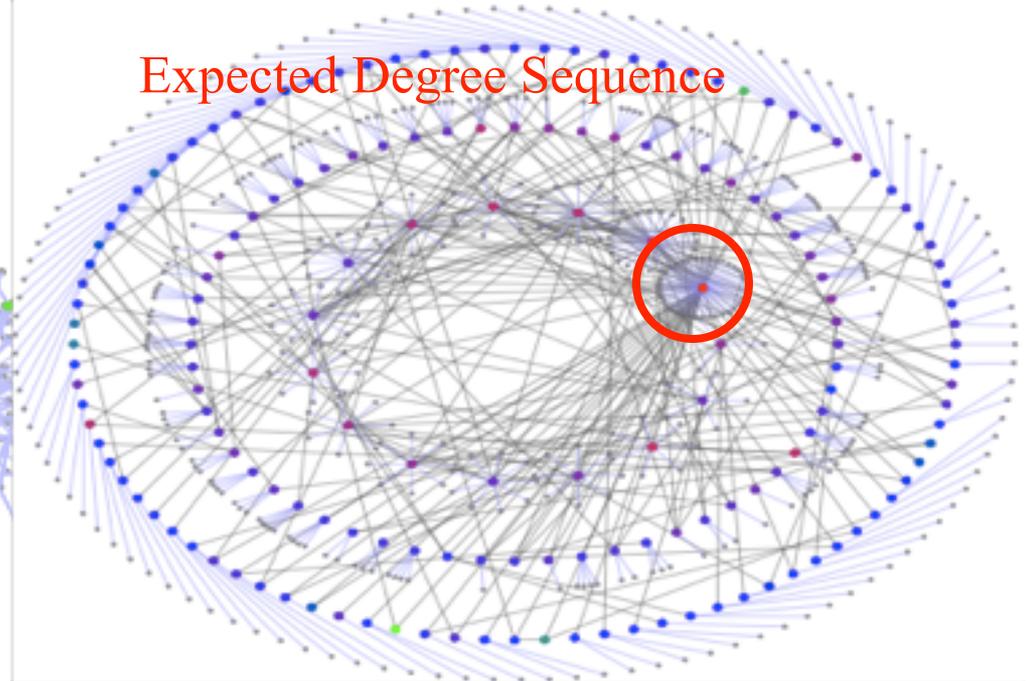
Features of Degree-Based Models



Preferential Attachment

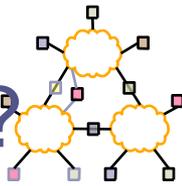


Expected Degree Sequence



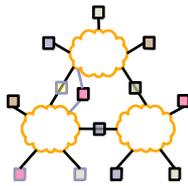
- Degree sequence follows a power law (by construction)
- High-degree nodes correspond to highly connected central “hubs”, which are crucial to the system
- Achilles’ heel: robust to random failure, fragile to specific attack

Does Internet graph have these properties?



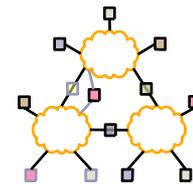
- No...(There is **no Memphis!**)
- Emphasis on degree distribution - **structure ignored**
- Real Internet very **structured**
- Evolution of graph is highly **constrained**

Problem With Power Law

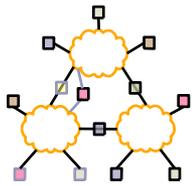


- ... but they're descriptive models!
- No correct physical explanation, need an understanding of:
 - the driving force behind deployment
 - the driving force behind growth

Outline

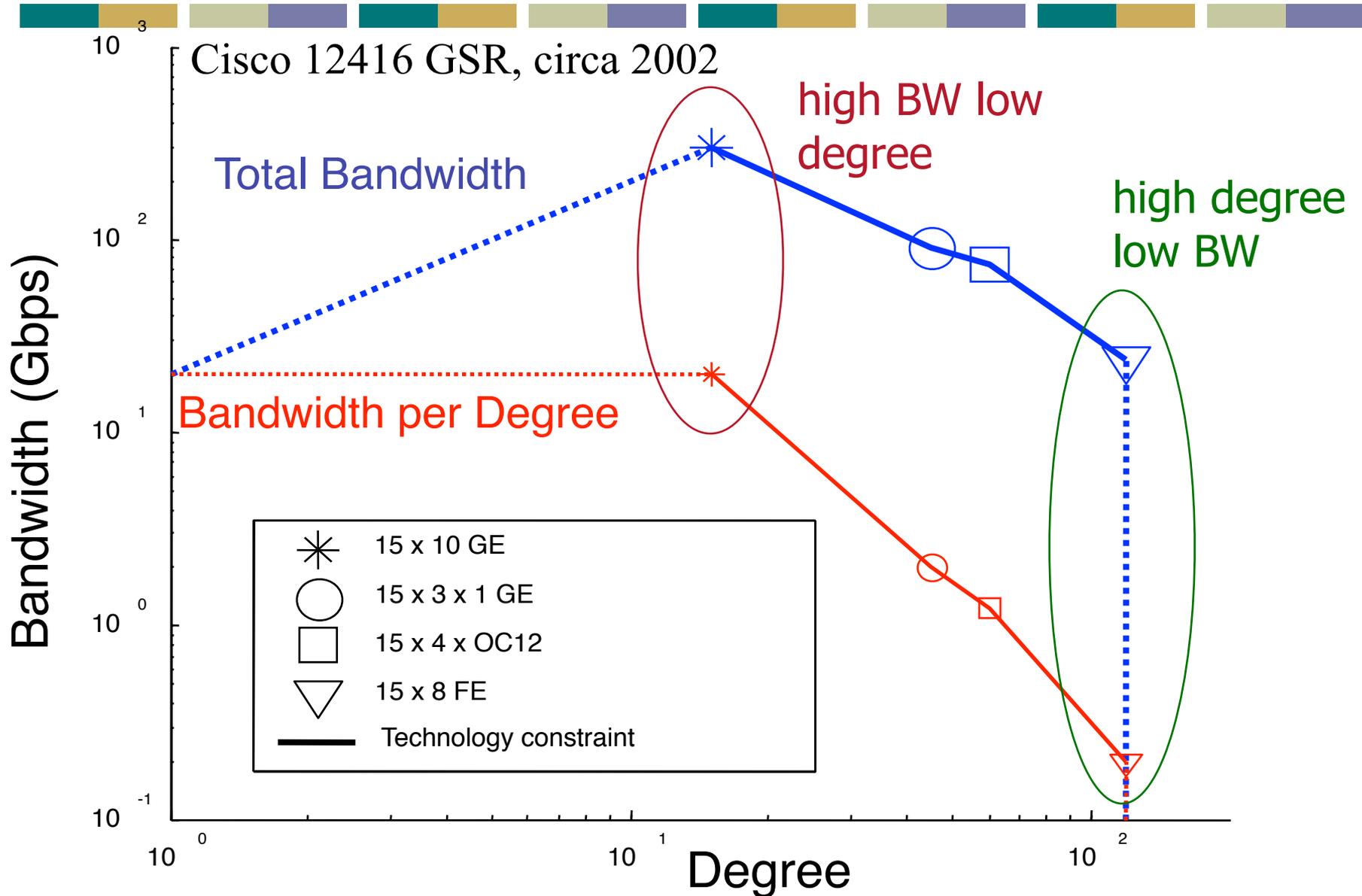
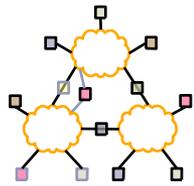


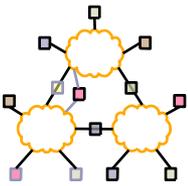
- Motivation/Background
- Power Laws
- Optimization Models



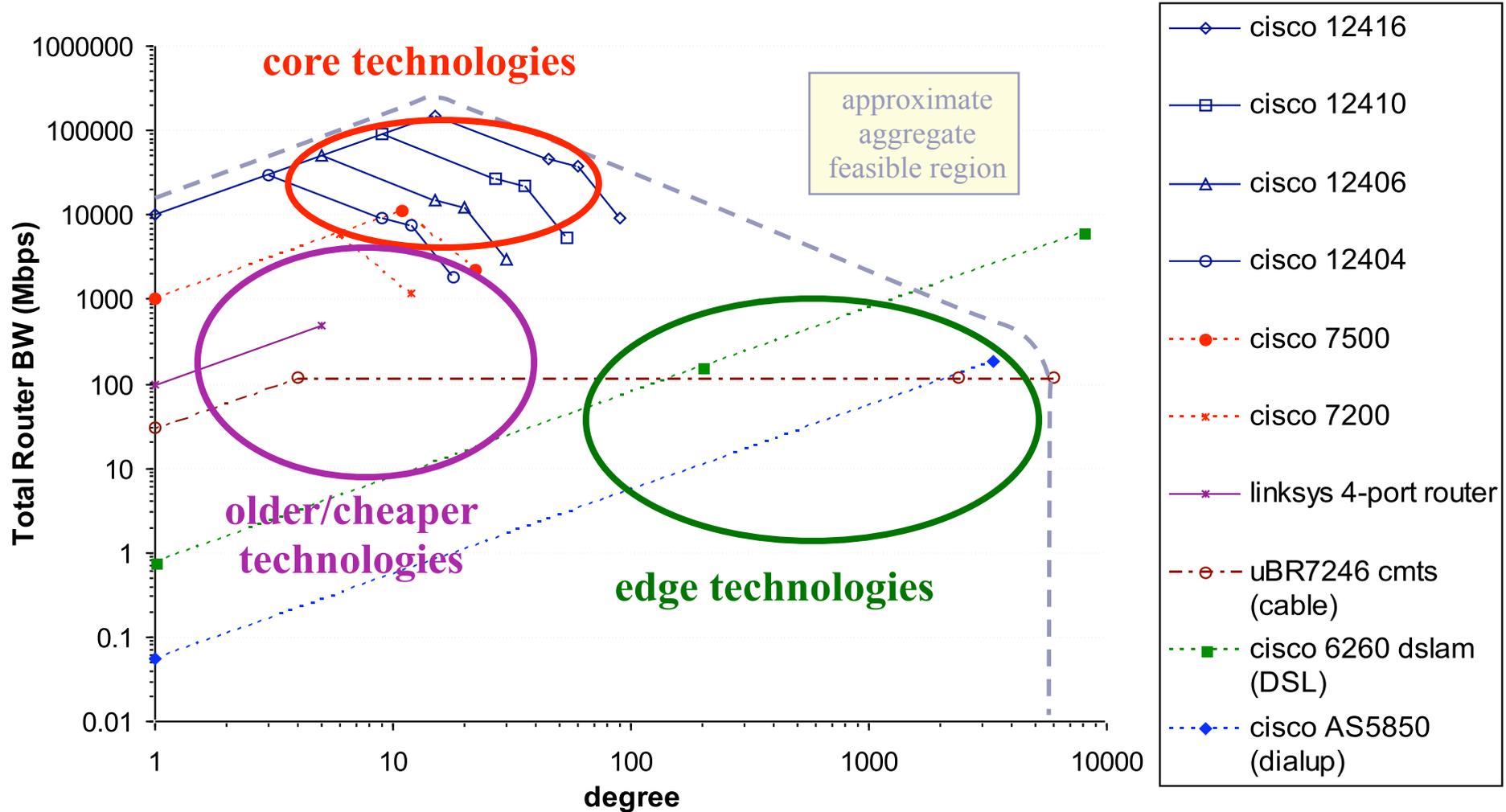
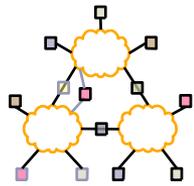
- Consider the explicit design of the Internet
 - Annotated network graphs (capacity, bandwidth)
 - Technological and economic limitations
 - Network performance
- Seek a theory for Internet topology that is explanatory and not merely descriptive.
 - Explain high variability in network connectivity
 - Ability to match large scale statistics (e.g. power laws) is only secondary evidence

Router Technology Constraint



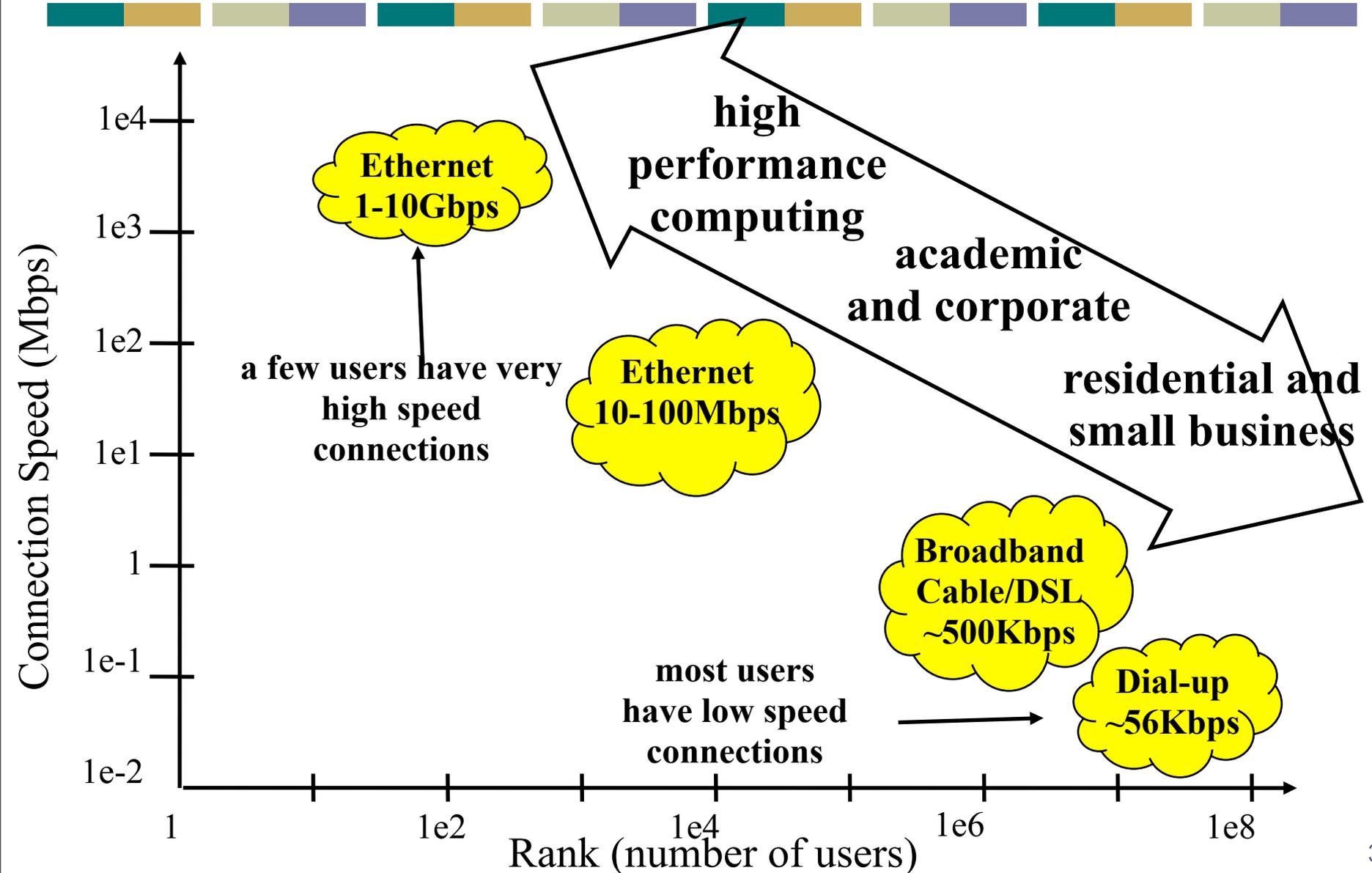
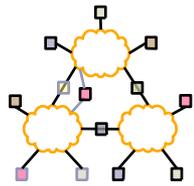


Aggregate Router Feasibility

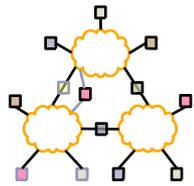


Source: Cisco Product Catalog, June 2002

Variability in End-User Bandwidths



Heuristically Optimal Topology



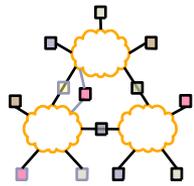
Mesh-like core of fast, low degree routers

Cores

High degree nodes are at the edges.

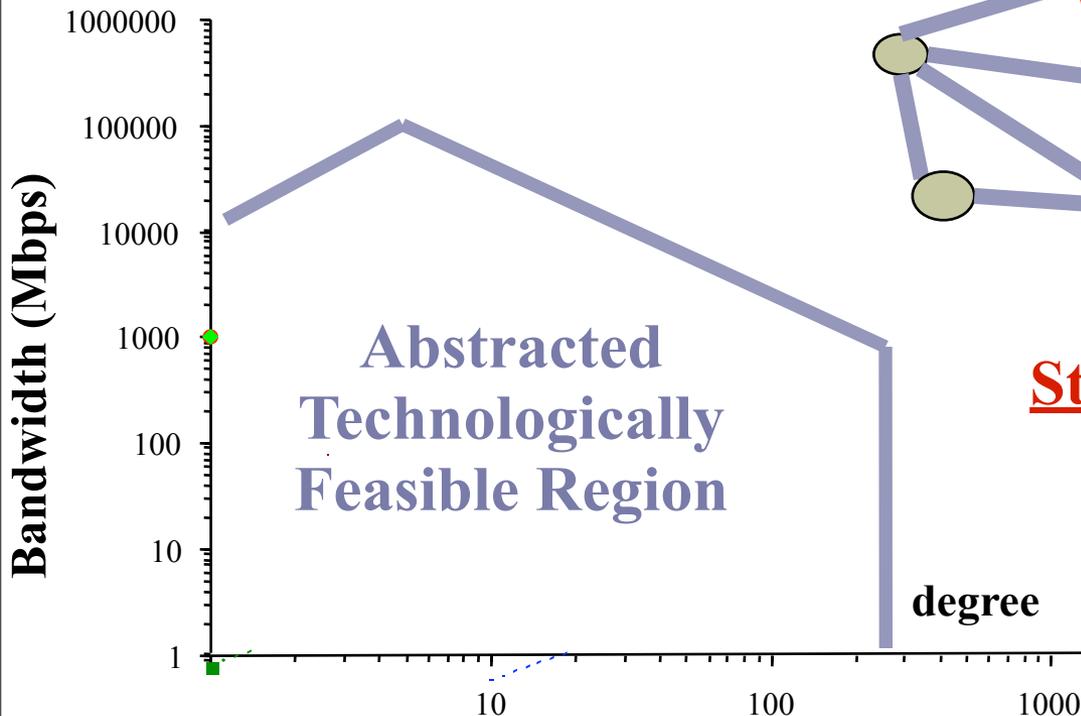
Hosts

Comparison Metric: Network Performance

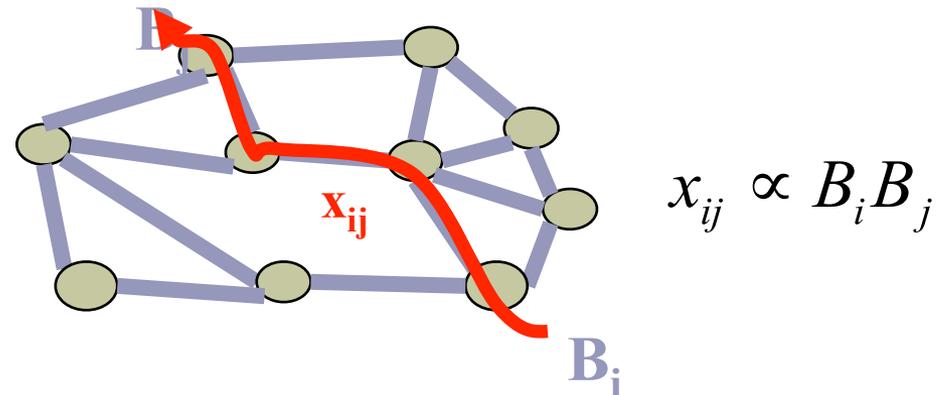


Given realistic technology constraints on routers, how well is the network able to carry traffic?

Step 1: Constrain to be feasible



Step 2: Compute traffic demand

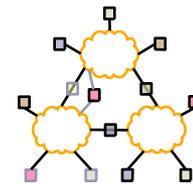


Step 3: Compute max flow

$$\max_{\alpha} \sum_{i,j} x_{ij} = \max \sum_{i,j} \alpha B_i B_j$$

$$s.t. \sum_{i,j:k \in r_{ij}} x_{ij} \leq B_k, \forall k$$

Likelihood-Related Metric

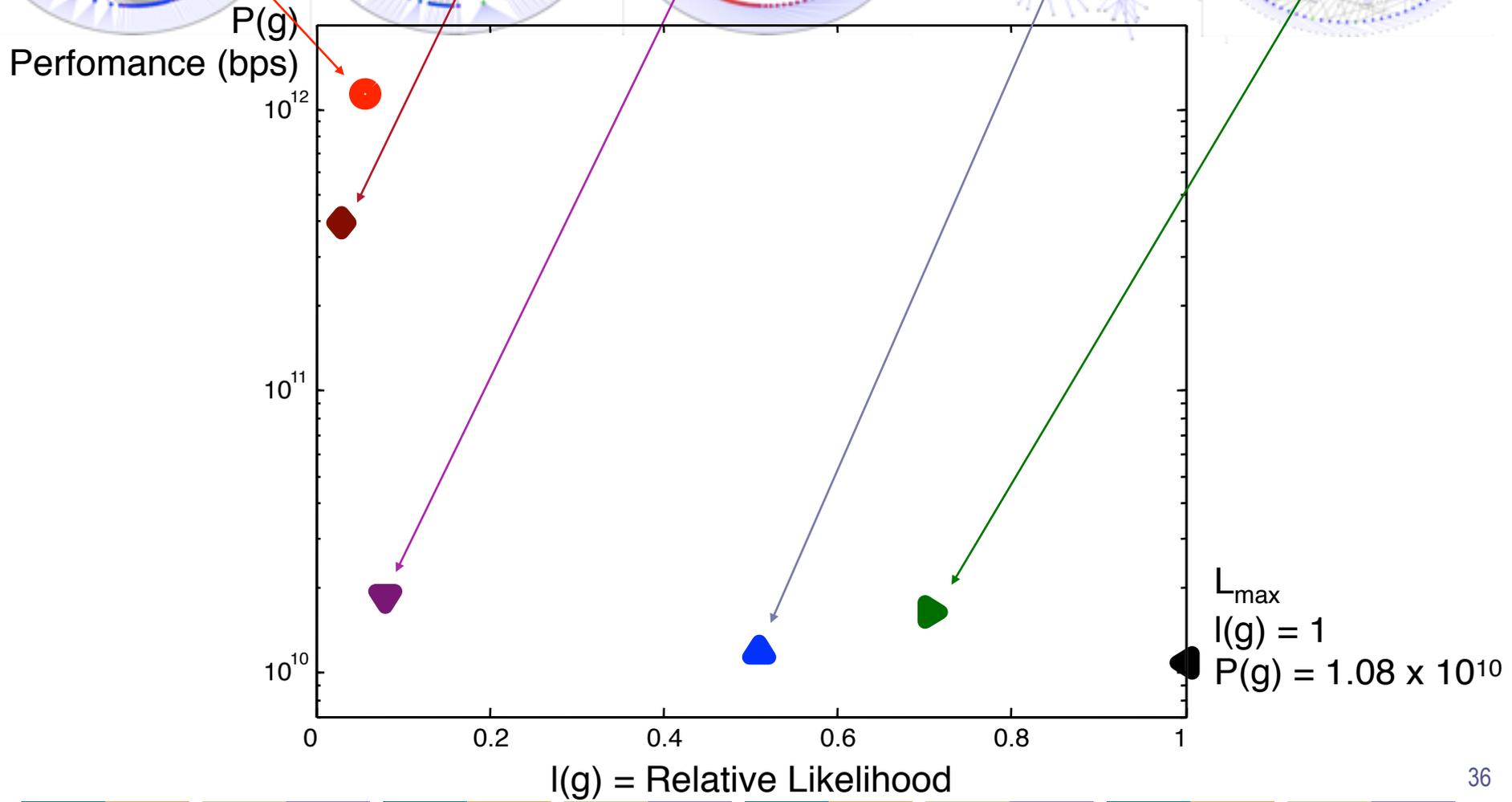
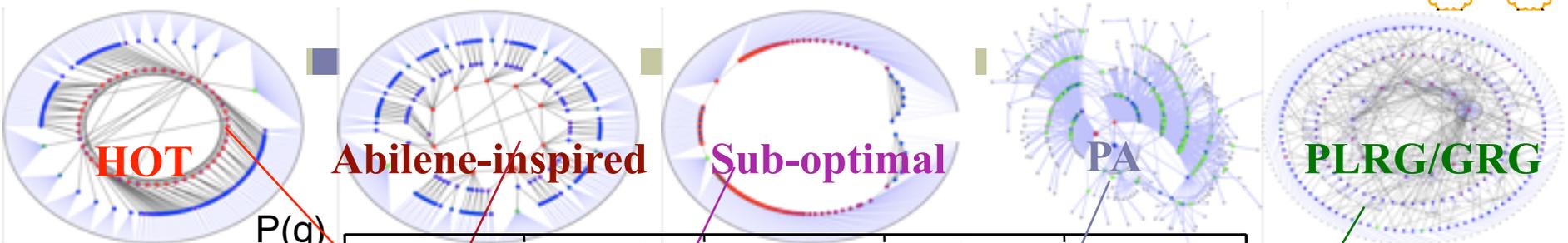
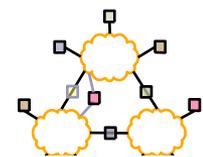


Define the metric $L(g) = \sum_{i,j \text{ connected}} d_i d_j$ ($d_i = \text{degree of node } i$)

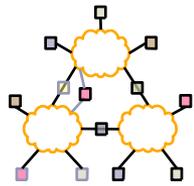
- Easily computed for any graph
- Depends on the structure of the graph, not the generation mechanism
- Measures how “hub-like” the network core is
- For graphs resulting from probabilistic construction (e.g. PLRG/GRG),

$$\text{LogLikelihood (LLH)} \propto L(g)$$

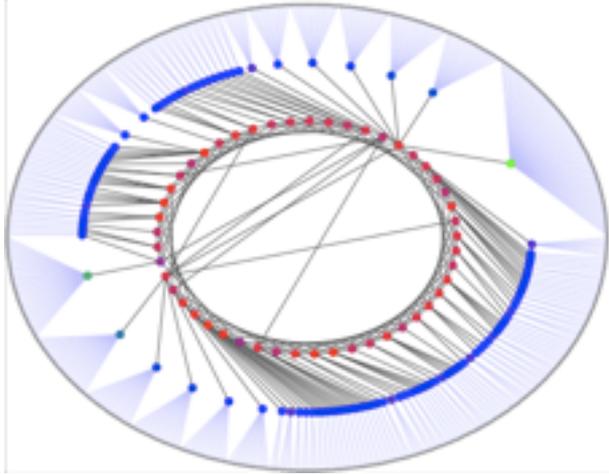
- Interpretation: How likely is a particular graph (having given node degree distribution) to be constructed?



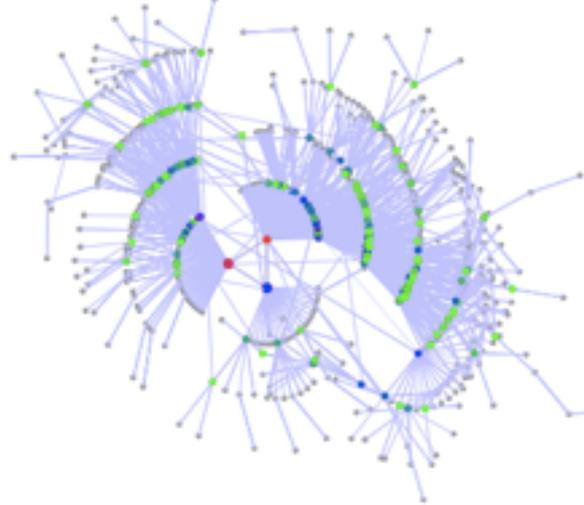
Structure Determines Performance



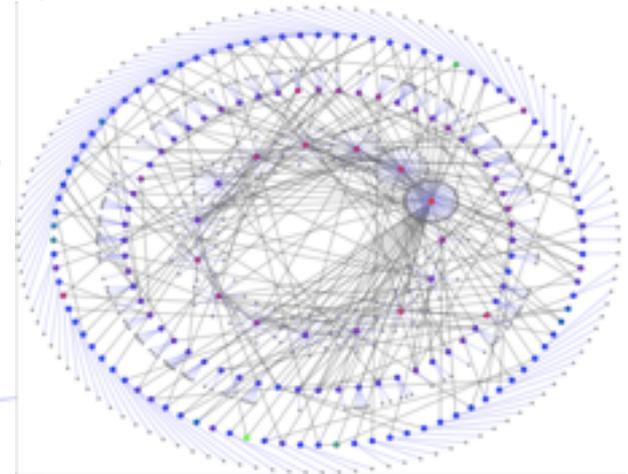
HOT



PA



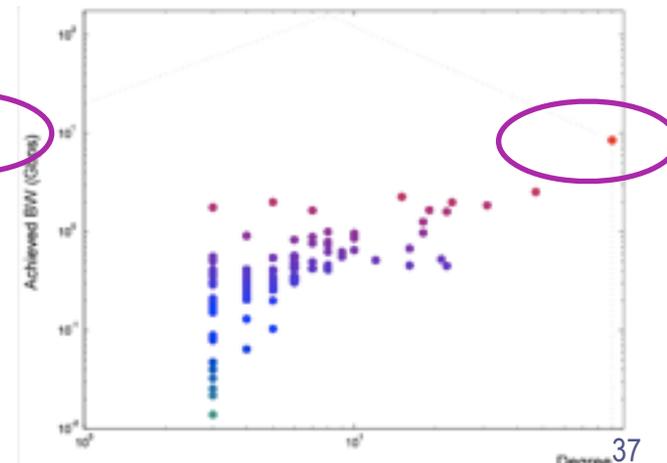
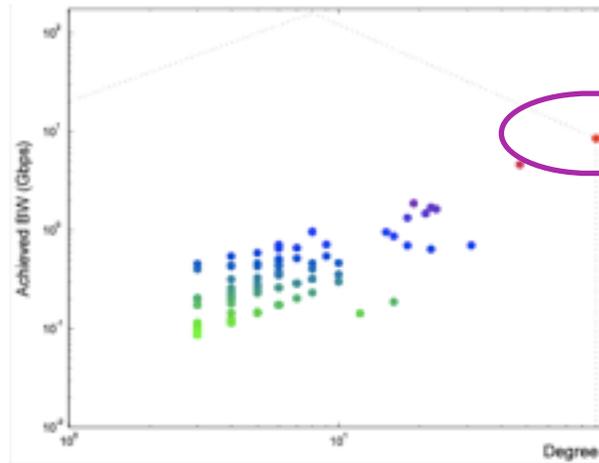
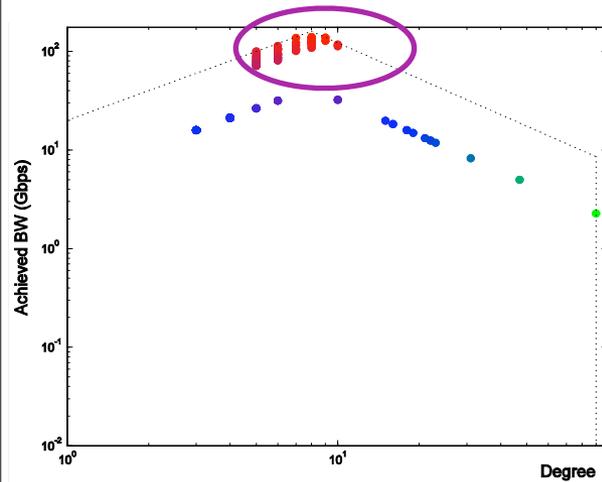
PLRG/GRG



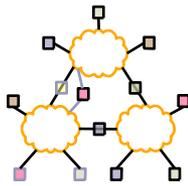
$$P(g) = 1.13 \times 10^{12}$$

$$P(g) = 1.19 \times 10^{10}$$

$$P(g) = 1.64 \times 10^{10}$$

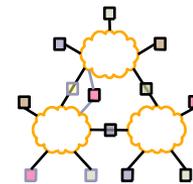


Summary Network Topology



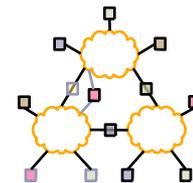
- Faloutsos³ [SIGCOMM99] on Internet topology
 - Observed many “power laws” in the Internet structure
 - Router level connections, AS-level connections, neighborhood sizes
 - Power law observation refuted later, Lakhina [INFOCOM00]
- Inspired many degree-based topology generators
 - Compared properties of generated graphs with those of measured graphs to validate generator
 - What is wrong with these topologies? Li et al [SIGCOMM04]
 - Many graphs with similar distribution have different properties
 - Random graph generation models don't have network-intrinsic meaning
 - Should look at fundamental trade-offs to understand topology
 - Technology constraints and economic trade-offs
 - Graphs arising out of such generation better explain topology and its properties, but are unlikely to be generated by random processes!

The elephant in the room...



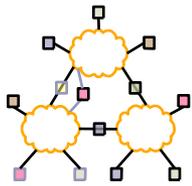
- How good is the underlying data on which these studies are based?
- E.g., sampling bias → traceroute of shortest paths on random graph can produce power-law distribution [Lakhina03]
 - Similar issues with AS-level view
- Router level data is very noisy

Better Measurements?



- Rocketfuel [sigcomm02]
 - Better router alias resolution
 - Detailed maps based on multiple viewpoints
- RouteViews and BGP collection efforts

Next Lecture



- Overlay networks
- Challenges in deploying new protocols
- Required readings:
 - Active network vision and reality: lessons from a capsule-based system
- Optional readings:
 - Resilient Overlay Networks
 - Future Internet Architecture: Clean-Slate Versus Evolutionary Research