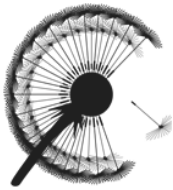


Impact of power-law topology on IP-level routing dynamics: simulation results

Amélie Medem, Clémence Magnien, Fabien Tarissan
CNRS and UPMC Sorbonne Universités
LIP6 Laboratory

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30 mars 2012
Orlando FL, Florida

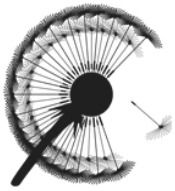




Studying the Internet topology

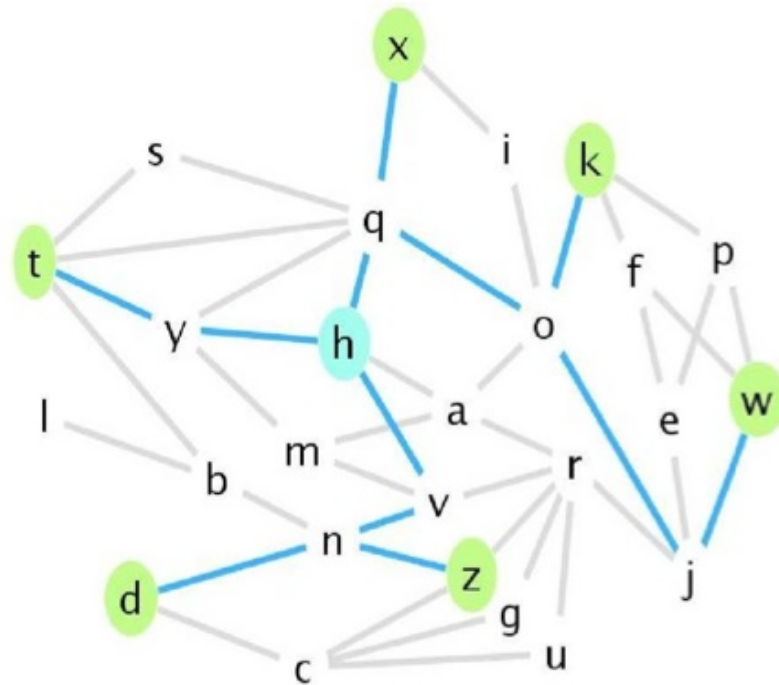
- Mapping the Internet topology is important
 - Future network protocols hard to test on real Internet
- Understanding how the Internet topology evolves in times is equally or even more important
 - Study its dynamics

Goal: IP-level routing topology and its dynamics

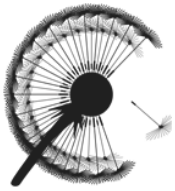


Ego-centered view

- **Focus:** the IP routing topology around a single node



Ego-centered view: what a node sees from the Internet



Outline

- (1) Measure the Internet IP topology around a node

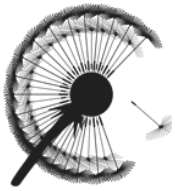
What a machine sees from the internet ?

- (2) Extract dynamic behaviors

How the topology measured evolves in time ?

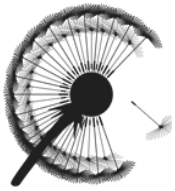
- (3) Confront them with the dynamics in artificial graphs

How to explain the observed dynamics ?



Outline

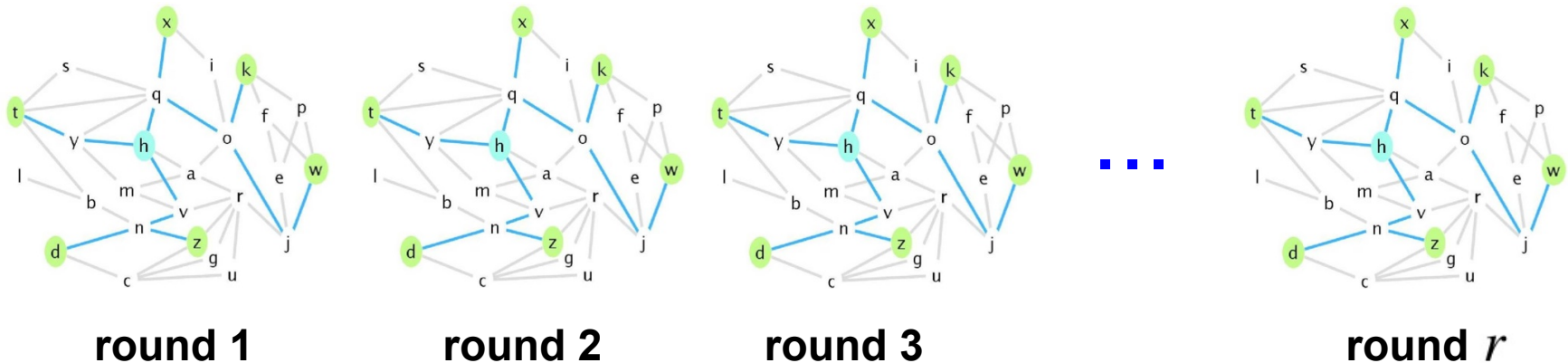
- (1) Measure the Internet IP topology around a node
- (2) Extract dynamic behaviors
- (3) Confront them with the dynamics of artificial graphs



Measures

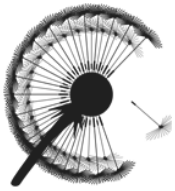
- **Tracetree:** traceroute-like measurement tool
 - A routing tree of IP paths from 1 source to destinations
- **Radar:** periodic measure with **Tracetree**

Series of routing trees



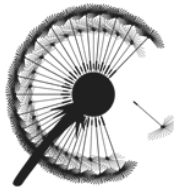
Sources: mostly **PlanetLab** (> 150); Destinations : **random**; Measurement frequency : **15 mn.** or less

Web site: "A Radar for the Internet—Publicly available datasets." <http://data.complexnetworks.fr/Radar/>



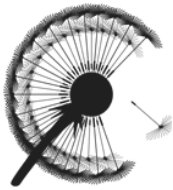
Outline

- (1) Measure the Internet IP topology around a node
- (2) Extract dynamic behaviors
- (3) Confront them with the dynamics in artificial graphs



Dynamic behaviors

- Two dynamic behaviors
 - IPs discovery
 - Pattern of occurrence of IPs

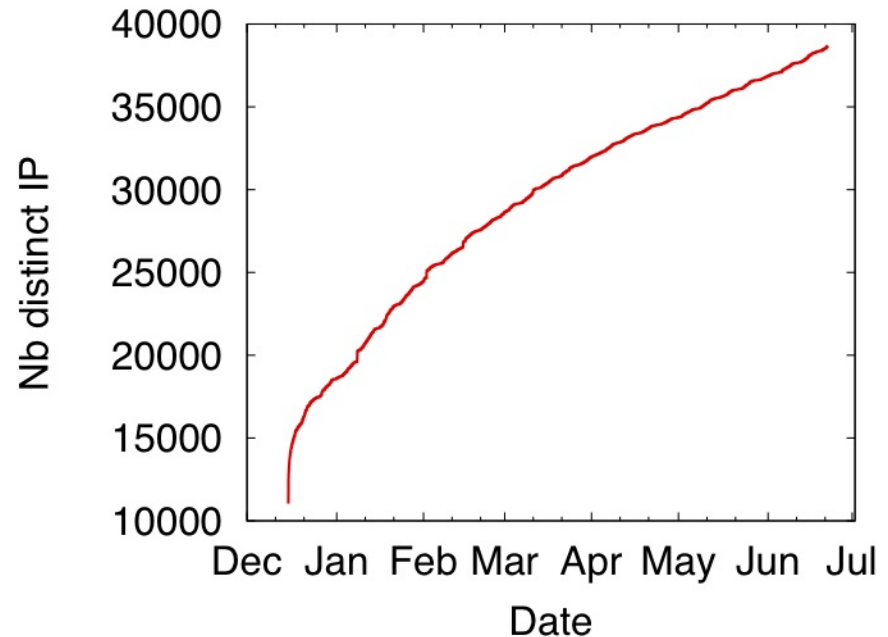


IPs discovery

■ Stabilization ?

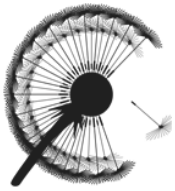
- **woolthorpe**
- 7 months
- 3,000 destinations
- 17,450 rounds

Number of distinct IPs observed since the beginning



New IPs are persistently discovered

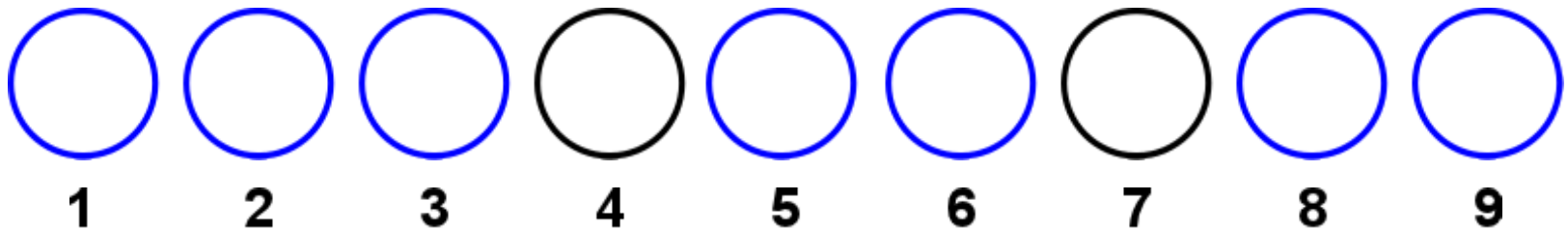
- High rate of discovery: ~150 new IPs per day
- Not due do measurement artifacts



Pattern of occurrence of IPs

- **2 values** to quantify the occurrence of IPs
 - **Observation number:** The total of distinct rounds in which it occurs
 - **Block number:** The number of groups of consecutive rounds in which it is observed

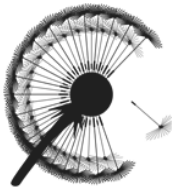
9 rounds



An IP_a appears in rounds: 1, 2, 3, 5, 6, 8, 9

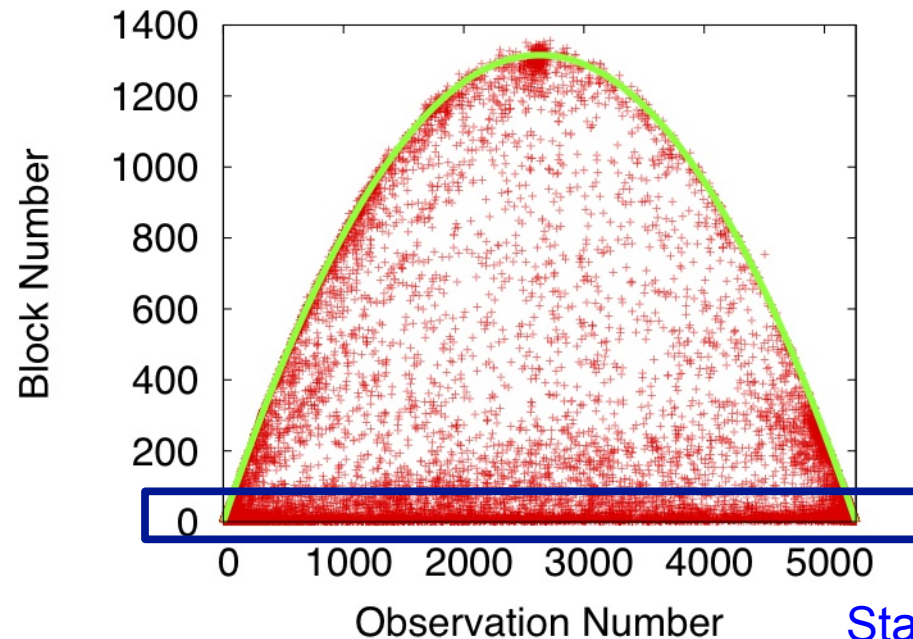
Observation number of $IP_a = 7$

Block number of $IP_a = 3$



Pattern of occurrence of IPs

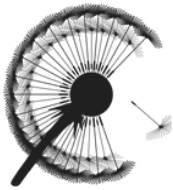
- Observation number vs. block number



(1) A parabola

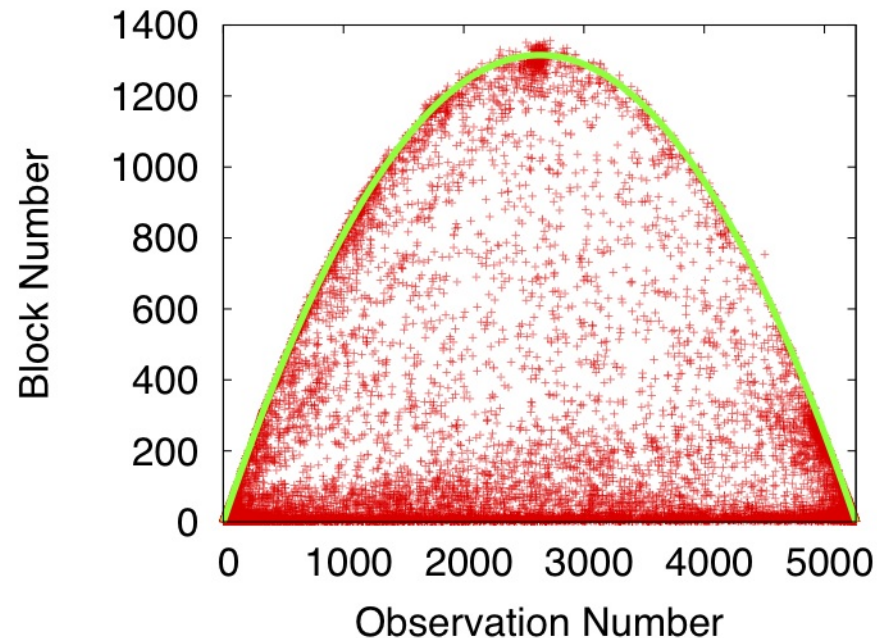
Stable IPs:
observed on consecutive rounds

(2) A large set of points close to the x -axis

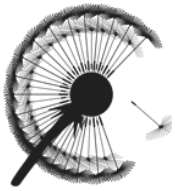


Pattern of occurrence of IPs

- The parabola, explained through [Load balancing](#)

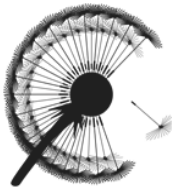


- Given r rounds of measurements
 - IPs on [load-balancing paths](#) has Prob p of being observed
 - Observation number: rp
 - Block number: $rp(1 - p)$



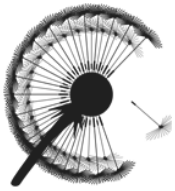
Outline

- (1) Measure the Internet IP topology around a node
- (2) Extract dynamic behaviors
 - Hard to explain them without knowing the Internet
- (3) Confront them with the dynamics in artificial graphs



Our approach

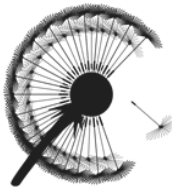
- **Goal:** Simulate **Tracetree** measurements on artificial graphs
 - Generate an initial undirected graph **G**
 - Power-law graph, Erdős-Rényi graph ...
 - Simulate **Tracetree** on **G** to create a routing tree **T**
 - Shortest-path model, ...
 - Incorporate on **G** well-known route change factors
 - Repeat previous steps to simulate periodic **Tracetree**



Route changes

- Well-known **factors of route changes**
 - Load balancing
 - Route evolution

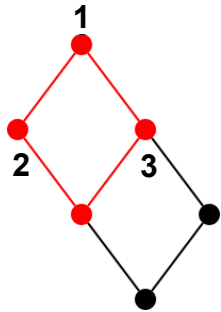
How to model these factors ?



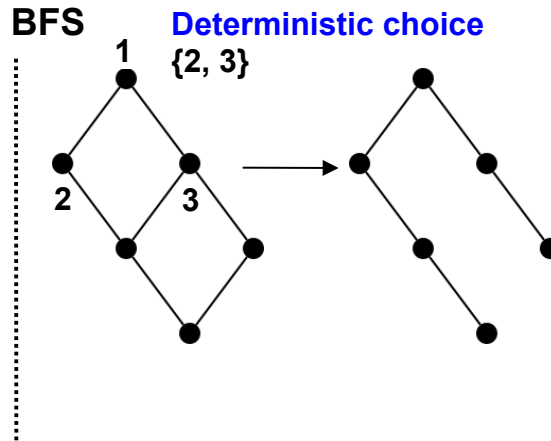
Modeling load balancing

- **How to model Load balancing ?**
- Tracetree measurement simulation
 - Shortest path model
 - Routing tree → BFS from the source

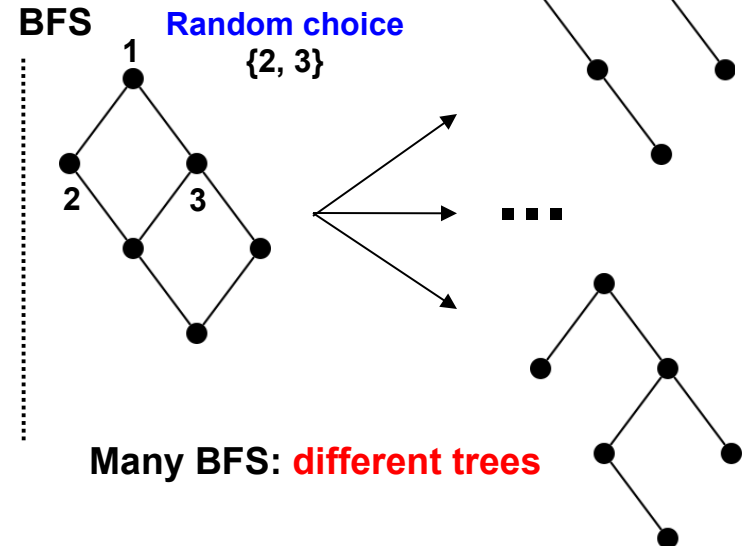
- **Random BFS**



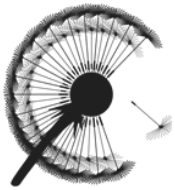
Load balancing



Many BFS: **same tree**



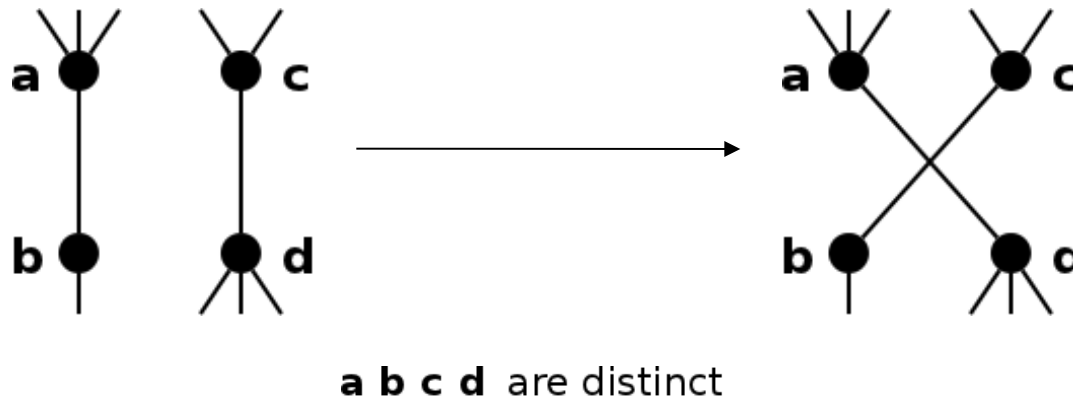
Many BFS: **different trees**



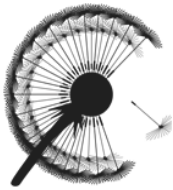
Modeling route evolution

■ How to model route evolution ?

- Possible approaches:
 - Realistic or not: link rewiring, nodes added or removed
- **Link rewiring or swap**

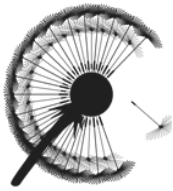


Main interest: the degree distribution of nodes is conserved



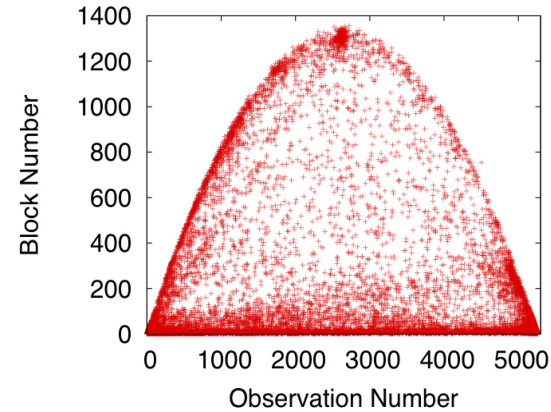
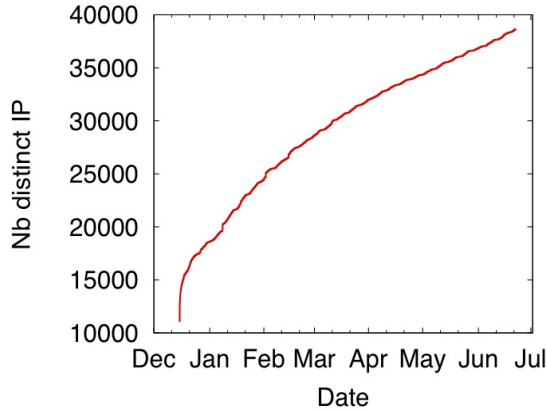
Simulations

- **Goal:** Simulate **Tracetest** measurements on artificial graphs
 - (1) Generate a **Power-law (PL) graph** $G = (V, E)$
 - Nodes: n ; Exponent: α
 - (2) Select 1 source and d destinations
 - Uniformly, **at random**
 - (3) Simulate **Tracetest** from source to destinations in G
 - Shortest paths, **Random BFS** $\rightarrow T_1$
 - (4) Simulate route evolution : Inject s **link swap** in G
 - (5) **Repeat steps 3 and 4** $\rightarrow T_2, T_3, T_4, \dots, T_r$

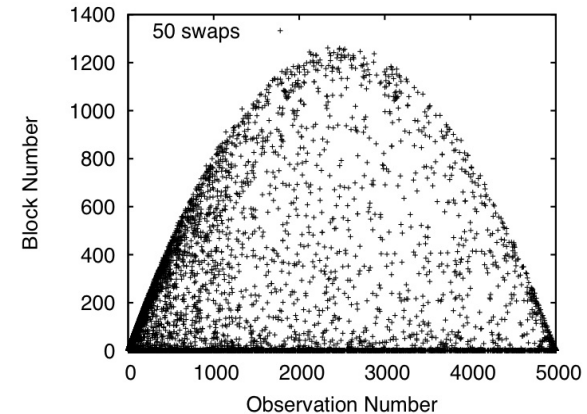
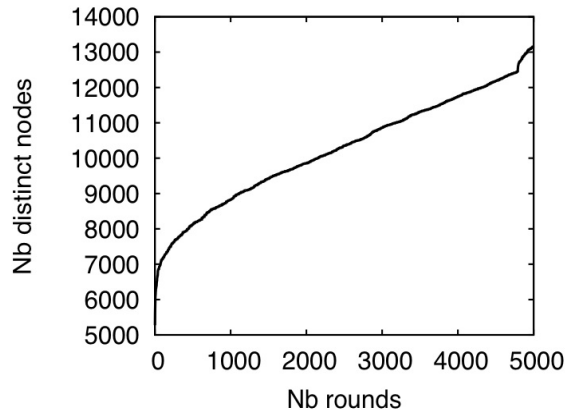


Reproducing the dynamics

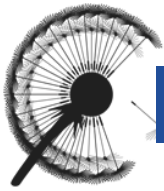
■ Internet



■ Simulations: Power graph with $n=500,000$; $\alpha=2.3$; with $d=3,000$ **50 swaps**

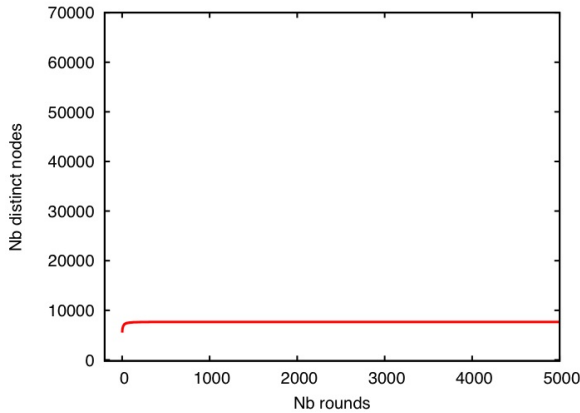


Qualitative similar behavior as observed in real Internet data !

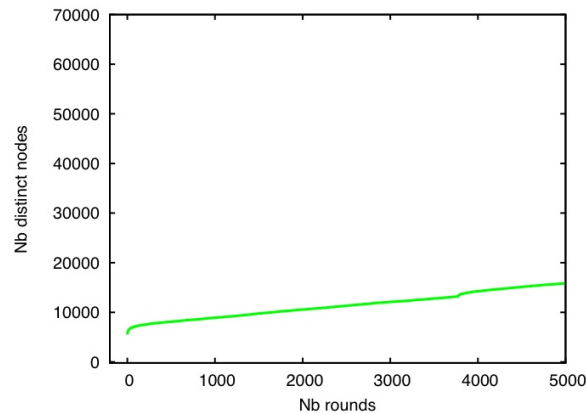


Impact of simulation parameters

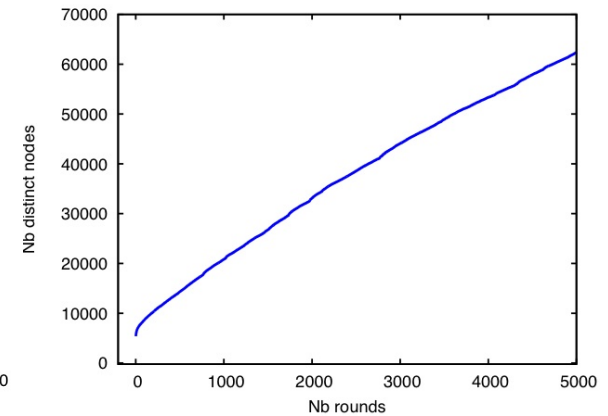
Impact of the number of swaps



0 swaps



100 swaps

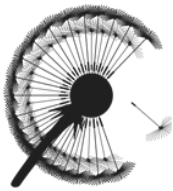


1,000 swaps

More swaps → Faster node discovery

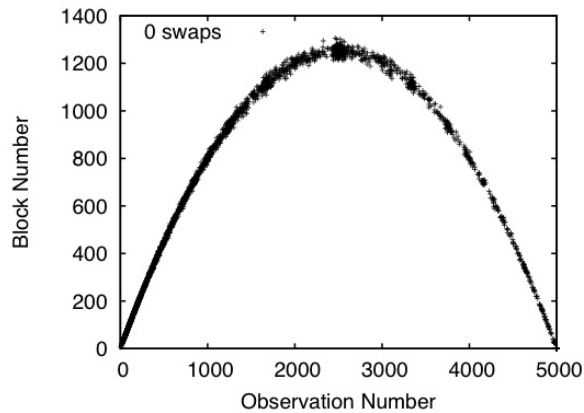
No swaps → Stabilization

The persistent discovery of IPs: due to route evolution

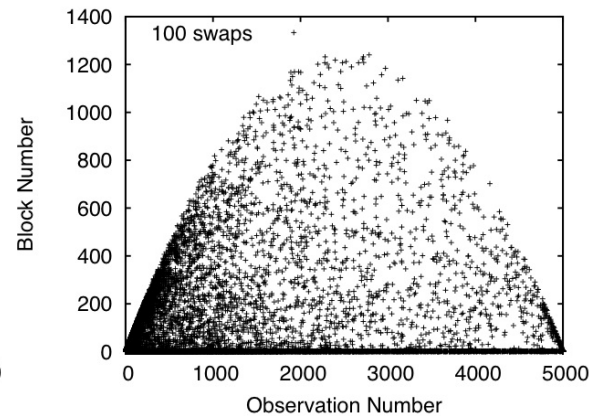


Observation vs. block

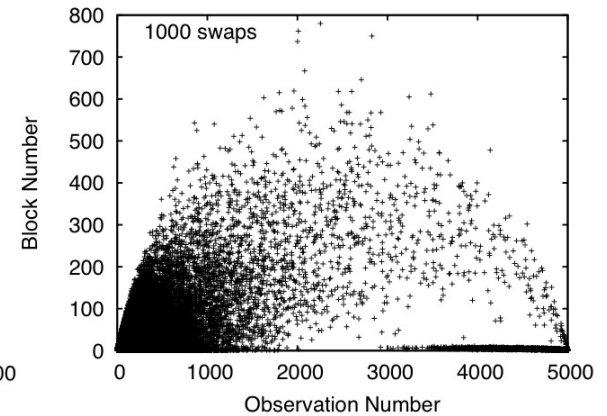
Impact of the number of swaps



0 swaps

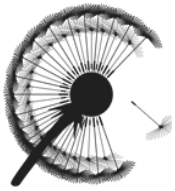


100 swaps



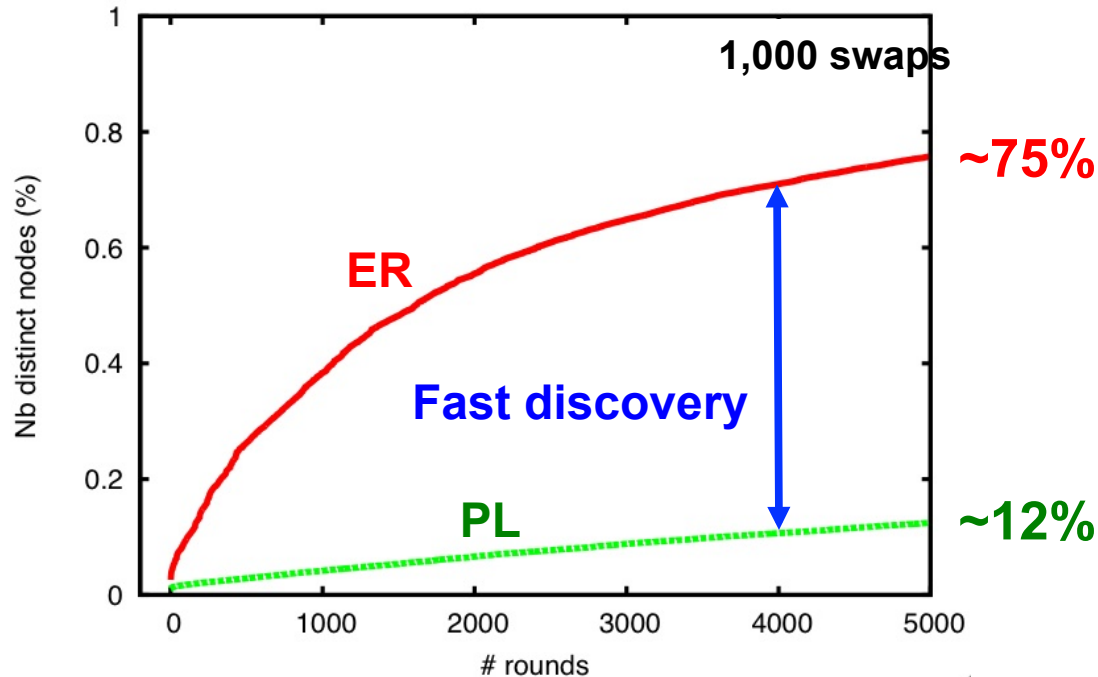
1,000 swaps

The parabola vanishes when the number of swaps increases
Points close to the parabola : due to load-balancing nodes



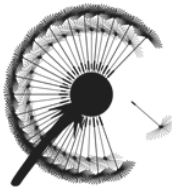
Power law vs. Erdős-Renyi

Quantitative difference between PL and ER graphs



Intuitions:

- (1) Degree-1 nodes ?
- (2) Average distance ?



Power law vs. Erdős-Renyi

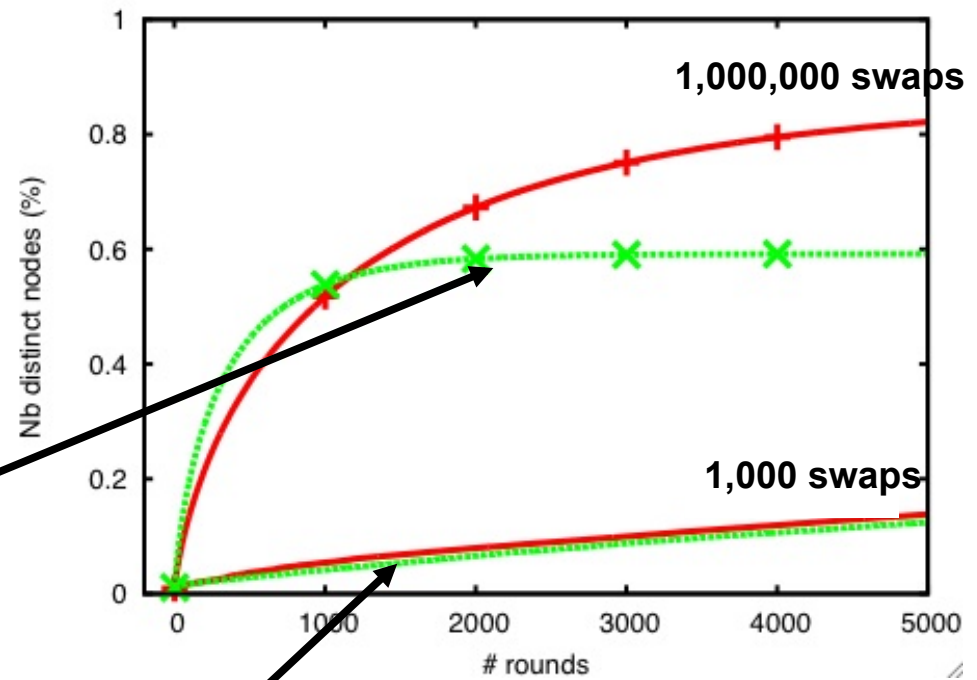
(1) Degree-1 nodes : Large fraction of nodes in PL graphs !!!

- Unless source/destinations, difficult for them to be discovered
- Not router nodes

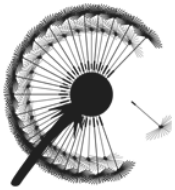
PL original

PL wo degree-1 nodes

Flat phase,
99.9% of
remaining nodes
have degree 1



Same evolution



Power law vs. Erdős-Renyi

(2) The average distance is smaller in PL than in ER graphs !!!

- PL graphs produce smaller routing trees than ER
 - On avg: 5,363 vs. 12,868 ($n=500,000$; $\alpha=2.3/1,000,000$ links)

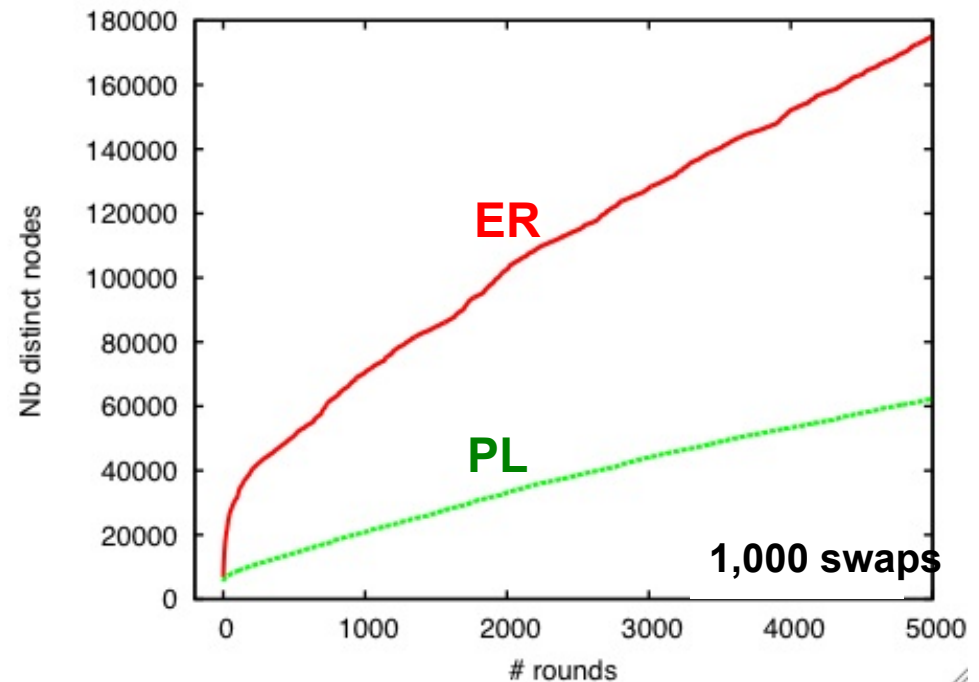
Same average distance

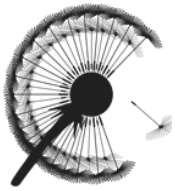
ER (8,000,000 links)

PL (1,000,000 links)

Evolution, still faster in ER

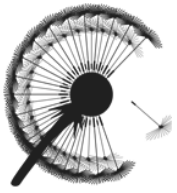
Ongoing work !

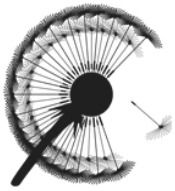




Summary

- **Study the dynamics of Internet IP routing topology around a node**
 - Two main behaviors to characterize the dynamics
 - A model (IP topology, dynamics, **Tracetree**) for explanation
 - Observed dynamics reproduced on power-law graphs
 - Observations quantitatively different in Erdős-Rényi graphs
 - Degree-one nodes, Average distance
- **Perspectives**
 - Integrate other dynamics
 - Node adding/remove, link adding/remove
 - Test other topologies (realistic topologies)
 - Perform theoretical analysis (quantify the slopes of curves)





Dynamic IPs

Stable destinations

