# Workload Generation and Emulation Environments Master 2R SL module MD

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#### Simulation Environments

- Introduction
- Operating System Emulation
- Network Emulation
- Grid Emulation
- World-wide Platform "Emulation"

# 2 Generating Synthetic Platforms

- Topology
- Compute resources
- Background conditions

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#### $Simulation = model \ implementation$

/ more abstract	Mathematical Simulation	Based solely on equations
	Discrete-Event Simulation	Abstraction of system as a set of dependant actions and events (fine- or coarse- grain)
less abstract	Emulation	Trapping and virtualization of low-level application/system actions

#### Simulation = model implementation

Network

		Network
more abstract	Mathematical Simulation	Macroscopic: flows in a pipe (coarse-grain d.e simulation + math. simulation)
	Discrete-Event Simulation	Microscopic: packet-level (fine-grain d.e. simulation)
less abstract	Emulation	Actual flows go through some network

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#### Simulation = model implementation

#### CPU

more abstract	Mathematical Simulation	Macroscopic: flows in a pipe (coarse-grain d.e simulation + math. simulation)
	Discrete-Event Simulation	Microscopic: Cycle-accurate simulation (fine-grain d.e. simulation)
less abstract	Emulation	Virtualization via another CPU/virtual machine

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#### Simulation = model implementation

#### Application

more abstract	Mathematical Simulation	$\label{eq:macroscopic:application} \mbox{Macroscopic: application} = \mbox{analytical "flow"}$
	Discrete-Event Simulation	Less Macroscopic: set of abstract tasks with resource needs and dependancies
less abstract	Emulation	Virtualization (emulation of actual code with trapping of application generated events)

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Simulation can combine all paradigms at different levels. Today, we will mainly talk about emulation though.

- Xen, User Mode Linux
- ModelNet, Emulab/DummyNet

PlatetLab

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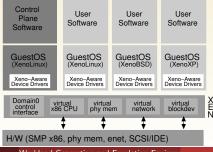
# **Operating System Emulation**

Goal run many instances of operating systems on the same physical machine.

Difficulties

- Performance Isolation (scheduling priority, memory demand, network traffic, disk accesses).
- OS compatibility (full/partial virtualization)
- Performance overhead

Common Tools User Mode Linux, Vserver, VMware, Xen An example: Xen

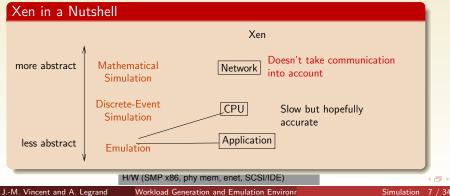


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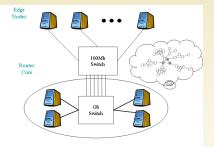
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# ModelNet

ModelNet is a UCSD/Duke project [VYW<sup>+</sup>02] lead by Amin Vahdat. Applications are supported by emulation and virtualization: Actual application code is executed on "virtualized" resources A key tradeoff in ModelNet is scalability versus accuracy. ModelNet accounts for network but not for CPU:

- Resource gethostnames, sockets are wrapped
- CPU Plain mapping, slowdown is not taken into account
- Network one emulator running FreeBSD, a gigabit LAN, some host machines with IP aliasing for the virtual nodes → emulation of heterogeneous links



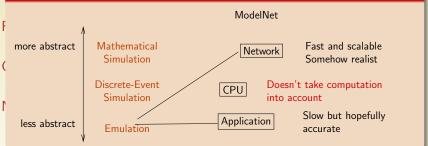
Similar ideas have been used in other projects (Emulab [WLS+02], DummyNet, Panda [KBM+02], ...)

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ModelNet in a Nutshell



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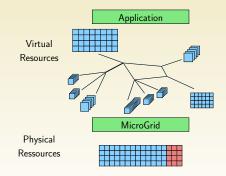
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# MicroGrid

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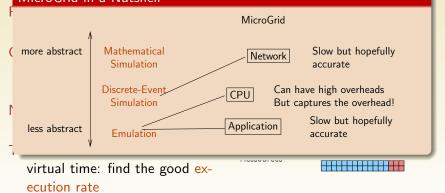
Resource gethostnames, sockets, GIS, MDS, NWS are wrapped

- CPU Direct execution on a fraction of physical CPU: find a good mapping
- Network Packet-level simulation (parallel version of MaSSF)
- Time Synchronize real time and virtual time: find the good execution rate



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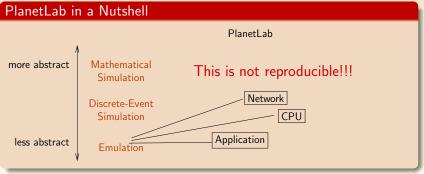
PlanetLab is an open platform for developping, deploying, and accessing planetary-scale services.

Planetary-scale 350 machines, 140 sites, 20 countries Distribution Virtualization each user can get a slice of the platform. Unbundled Management

- OS defines only local (per-node) behavior (global (networkwide) behavior implemented by services)
- multiple competing services running in parallel (shared, unprivileged interfaces)

Unstable like the real world Convenient to try P2P applications or demonstrate the feasability of a middleware.

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#### Two goals of simulations:

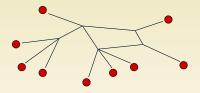
- Simulate platforms beyond the ones at hand
- Perform sensitivity analyses

### Need: Synthetic platforms

- Examine real platforms
- Discover principles
- Implement "platform generators"

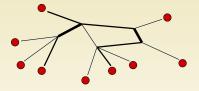
Network Topology

▶ Graph



#### Network Topology

- Graph
- Bandwidth and Latencies

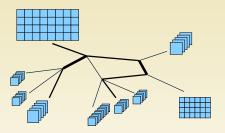


### Network Topology

- Graph
- Bandwidth and Latencies

Compute Resources And other resources "Background" Conditions

CPU capacity

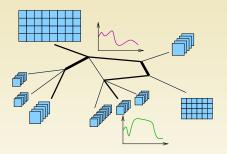


### Network Topology

- Graph
- Bandwidth and Latencies

Compute Resources And other resources "Background" Conditions

- CPU capacity
- Load

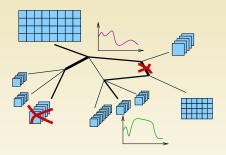


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#### What is Representative and Tractable?

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The network community has wondered about the properties of the Internet topology for years

- The Internet grows in a decentralized fashion with seemingly complex rules and incentives
- Could it have a mathematical structure?
- Could we then have generative models?

Three "generations of graph generators"

- "Plain" Random
- Structural
- Degree-based

Waxman [Wax88] Dots are randomly placed on a square of side cand are randomly connected with a probability  $P(u, v) = \alpha e^{-d/(\beta L)}$ ,  $0 < \alpha, \beta \leq 1$  where d is the Euclidean distance between u and vand  $L = c\sqrt{2}$ .

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Exponential Dots are randomly placed and are connected with a probability  $P(u, v) = \alpha e^{-d/(L-d)}$ .

Brain-dead N dots are randomly chosen (using a uniform distribution) in a square. Then they are randomly connected with a uniform probability  $\alpha$ .

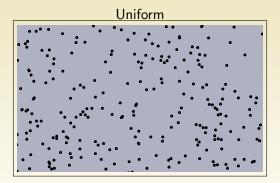
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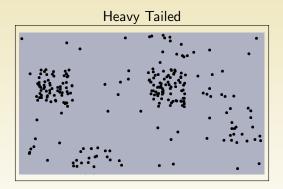
Locality [ZCD97] This model is due to Zegura. Dots are randomly placed and are connected with a probability

$$P(u, v) = \begin{cases} \alpha & \text{if } d < L \times r \\ \beta & \text{if } d \geqslant L \times r \end{cases}$$

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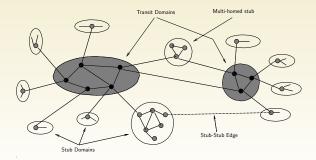
# Node placement





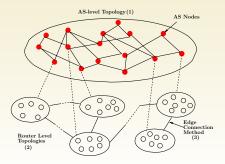
#### Top-Down

# N-level [ZCD97] Starting from a connected graph, at each step, a node is replaced by another connected graph (Tiers, GT-ITM).



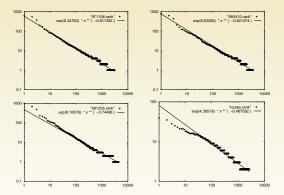
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N-level [ZCD97] Starting from a connected graph, at each step, a node is replaced by another connected graph (Tiers, GT-ITM).
 Transit-stub [ZCD97] 2-levels of hierarchy and some additional edges (GT-ITM, BRITE).



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#### Power Law (rank exponent).

Given a graph, the degree  $d_v$  of a node v is proportional to the rank of the node  $r_v$  to the power of a constant  $\mathcal{R}$ .

### $d_v \propto r_v^{\mathcal{R}}$

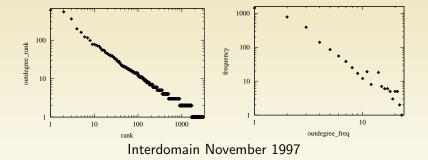
	Nov. 97	Apr. 98	Dec. 98	Router 95
$\mathcal{R}$	0,81	0,82	0,74	0,48

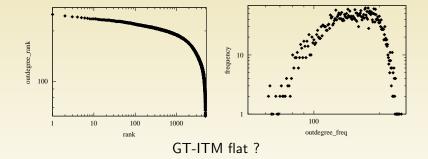
#### Incremental growth and affinity lead to Power Laws [BA99].

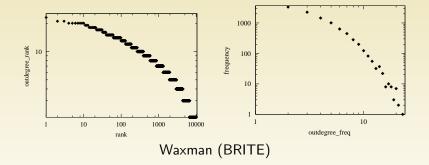
Incremental growth and affinity lead to Power Laws [BA99]. Nodes are incrementally added. The probability that v is connected to u depends on  $d_u$ :

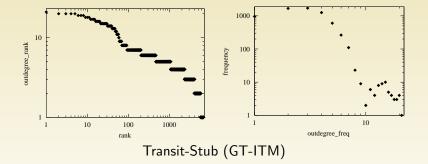
$$P(u,v) = \frac{d_u}{\sum_k d_k}$$

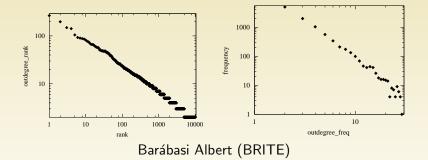
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Observation [TGJ+02]:

- AS-level and router-level have similar characteristics
- Degree-based generators are significantly better at representing large scale properties of the Internet than structural ones.
- Hierarchy seem to arise from degree-based generators.

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- ► For a 100 node platform, power laws do not make sense and structural generators may be more appropriate.

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We need some additional informations (e.g. routing, bandwidth, latency, sharing capacity, ...).

Topology generators only produce a graph: We need link characteristics as well

- Model physical characteristics.
  - Some "models" in topology generators
  - Need to simulate background traffic
  - No accepted model for generating background traffic
  - Simulation can be very costly

Model end-to-end performance. Models ([LS01]) or Measurements (NWS, ...). Go from path modeling to link modeling?

Turns out to be a difficult question (DARPA workshop on network modeling).

Maybe none of this matters?

- Fiber inside the network mostly unused
- Communication bottleneck is the local link
- Appropriate tuning of TCP or better protocols should saturate the local link
- Don't care about topology at all!
- Maybe none of this matters for my application (no network contention)

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#### What resources do we put at the end- points?

"Ad-hoc" generalization Look at the TeraGrid. Generate new sites based on existing sites

#### Statistical modeling

- Examing many production resources
- Identify key statistical characteristics
- Come up with a generative/predictive model

Many Grid resources are clusters. What is the "typical" distribution of clusters?

"Commodity Cluster synthesizer" [KCC04a]

- Examined 114 production clusters (10K+ procs)
- Came up with statistical models
  - Linear fit between clock-rate and release-year within a processor family
  - Quadratic fraction of processors released on a given year
- Validated model against a set of 191 clusters (10K+ procs)
- Models allow "extrapolation" for future configurations
- Models implemented in a resource generator

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#### Probabilistic models

- Naive: experimental distributed availability and unavailability intervals
- Sophisticated: Weibull distributions [NBW05]
- Traces NWS, Desktop Grid resources [KCC04b]

Workload models e.g. batch schedulers

- Traces
- ▶ Models [LF03]: job inter-arrival times (Gamma), amount of work requested (Hyper-Gamma), number of processors requested: Compounded (2<sup>p</sup>, 1, ...)
- Adversary ?

# A Sample Synthetic Grid ?

- Generate a 5,000 node graph with BRITE
- Annotate latency according to BRITE's Euclidian distance method (scaling to obtain the desired network diameter)
- Annotate bandwidth based on a set of end-to-end NWS measurements
- Pick 30% of the end-points
- Generate a cluster at each end-point according to Kee's synthesizer for Year 2006
- Model cluster load with Feitelson's model with a range of parameters for the random distributions
- Model resource failures based on Inca measurements on Tera-Grid

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