

Statistical Approach to NoC Design

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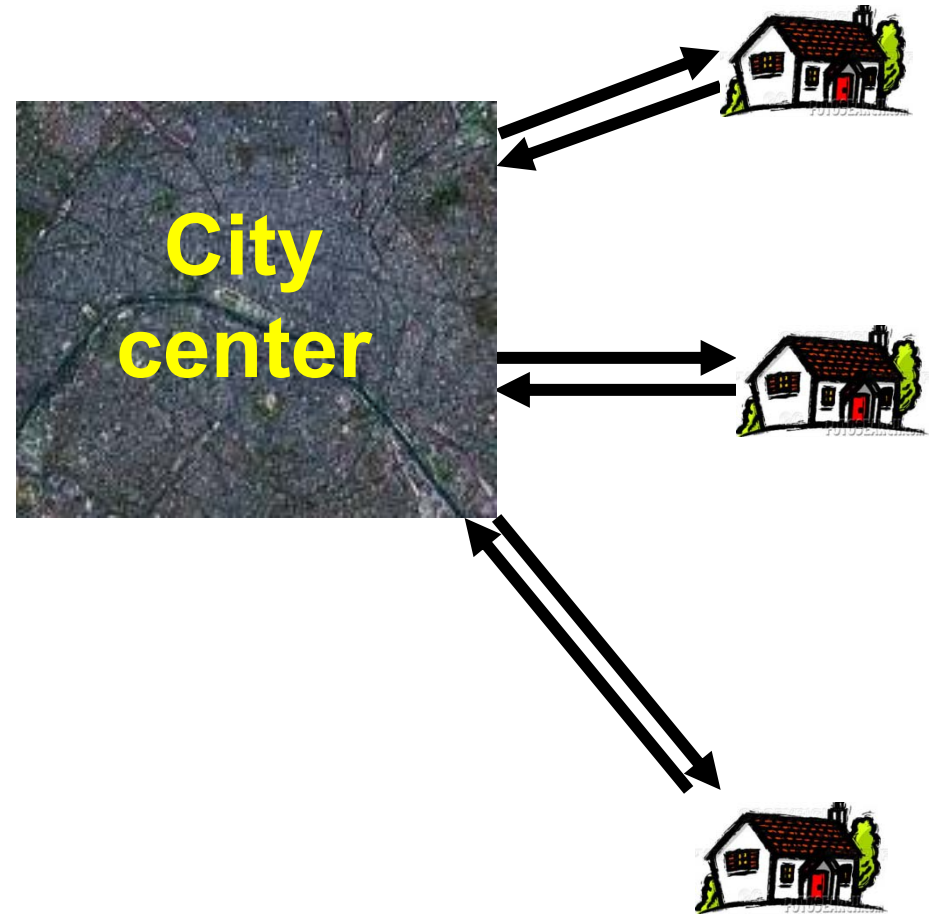
Problem

The traffic matrix in NoCs is often-changing and unpredictable

⇒ makes NoCs hard to design

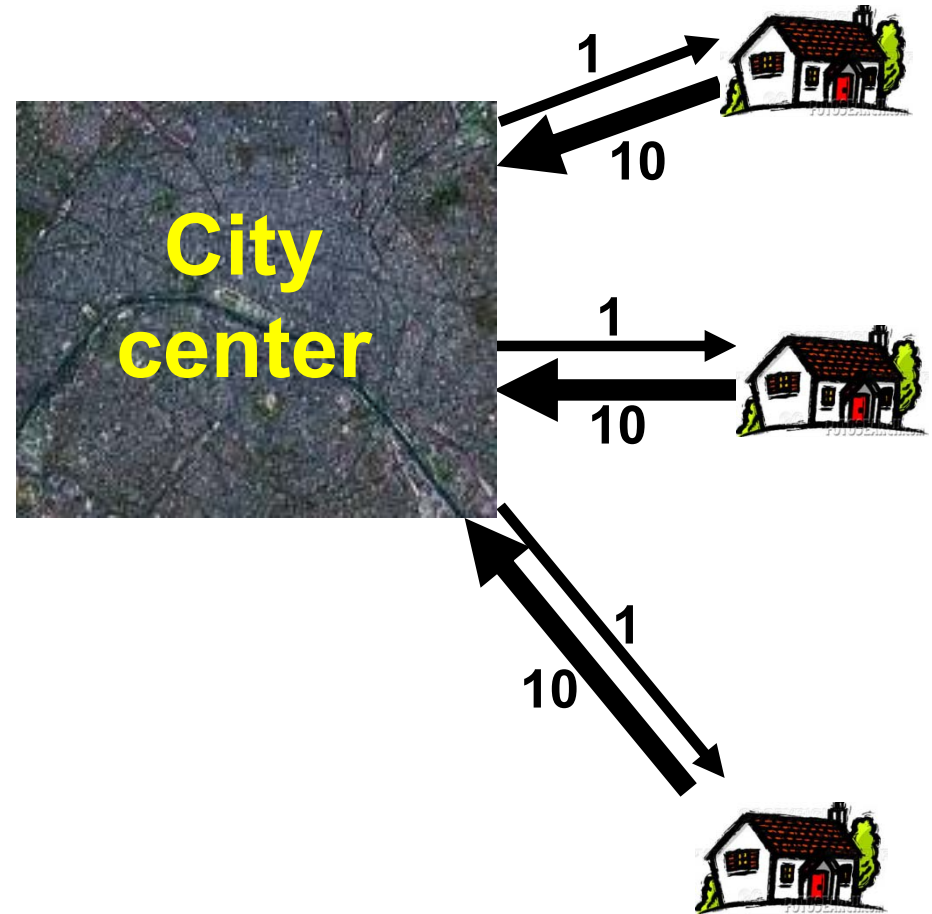
Road Capacities

- **Goal:** Design road capacities between a city and its suburbs
- Let's model the traffic matrices...



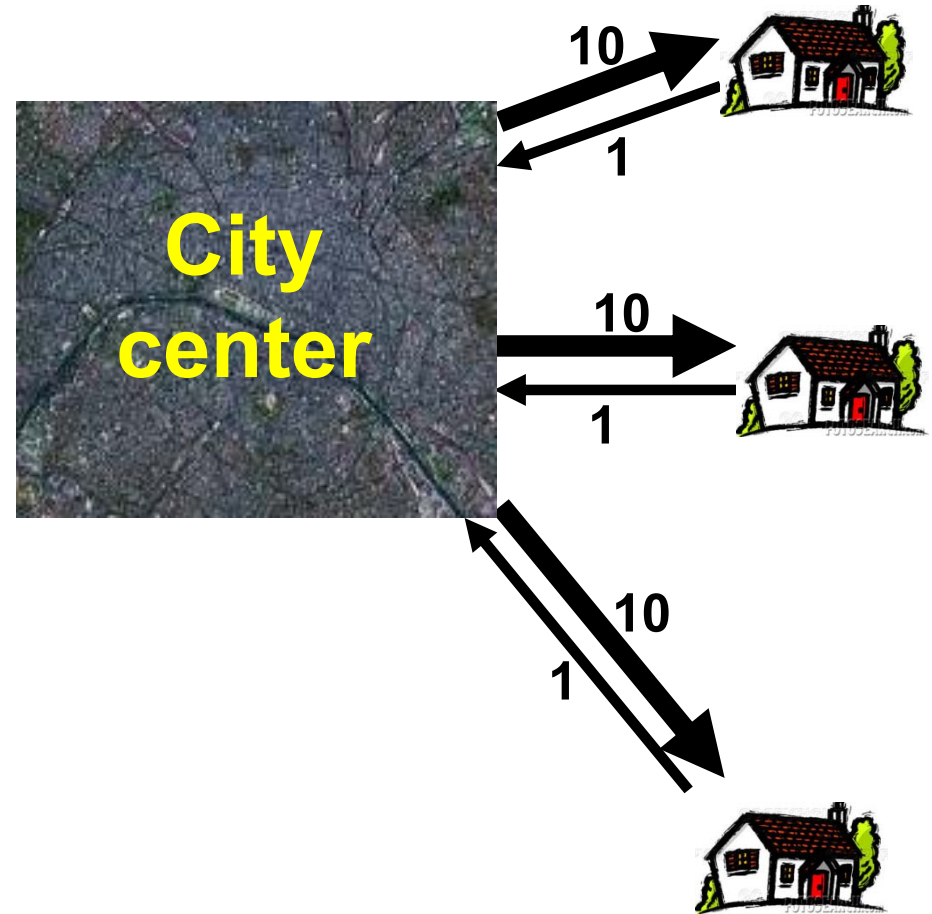
Road Capacities

- **Morning peak:** most traffic towards city



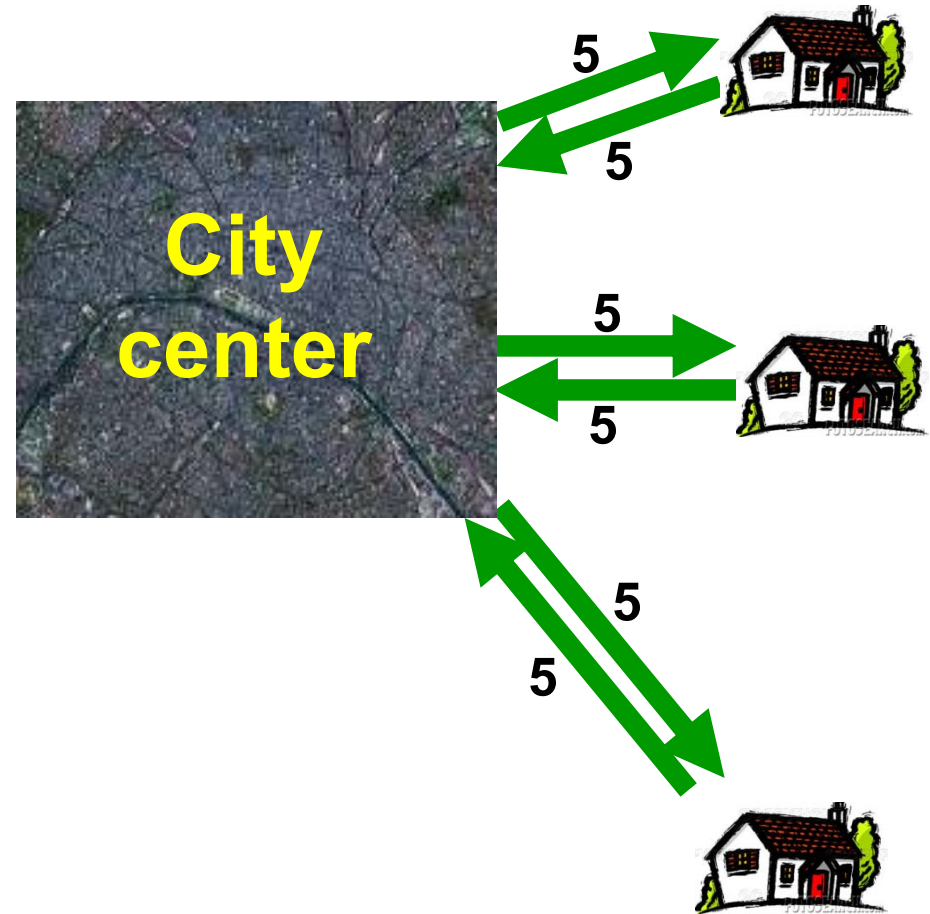
Road Capacities

- **Morning peak:** most traffic towards city
- **Afternoon peak:** most traffic leaving city



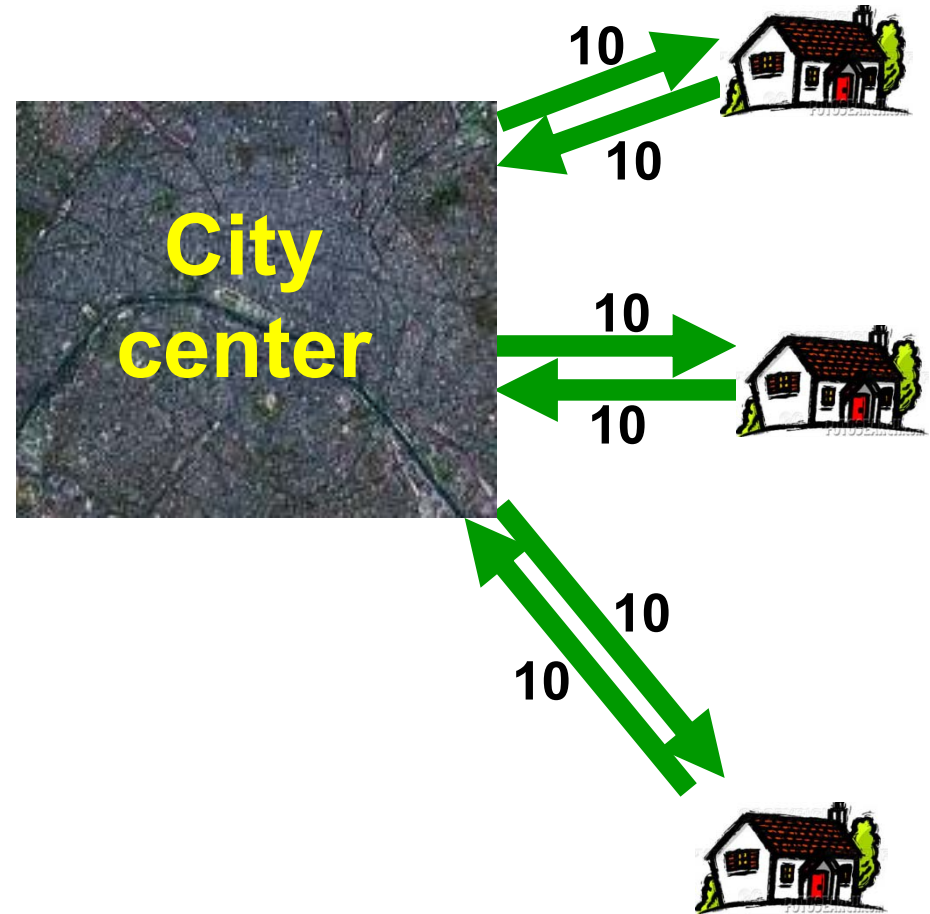
Solution (1): Average-Case

- Solution (1): plan for **average-case**
 - i.e. allocate capacity of ~ 5 for each link.
 - $\lambda < \mu$
- Problem: traffic jam during many hours, every day.



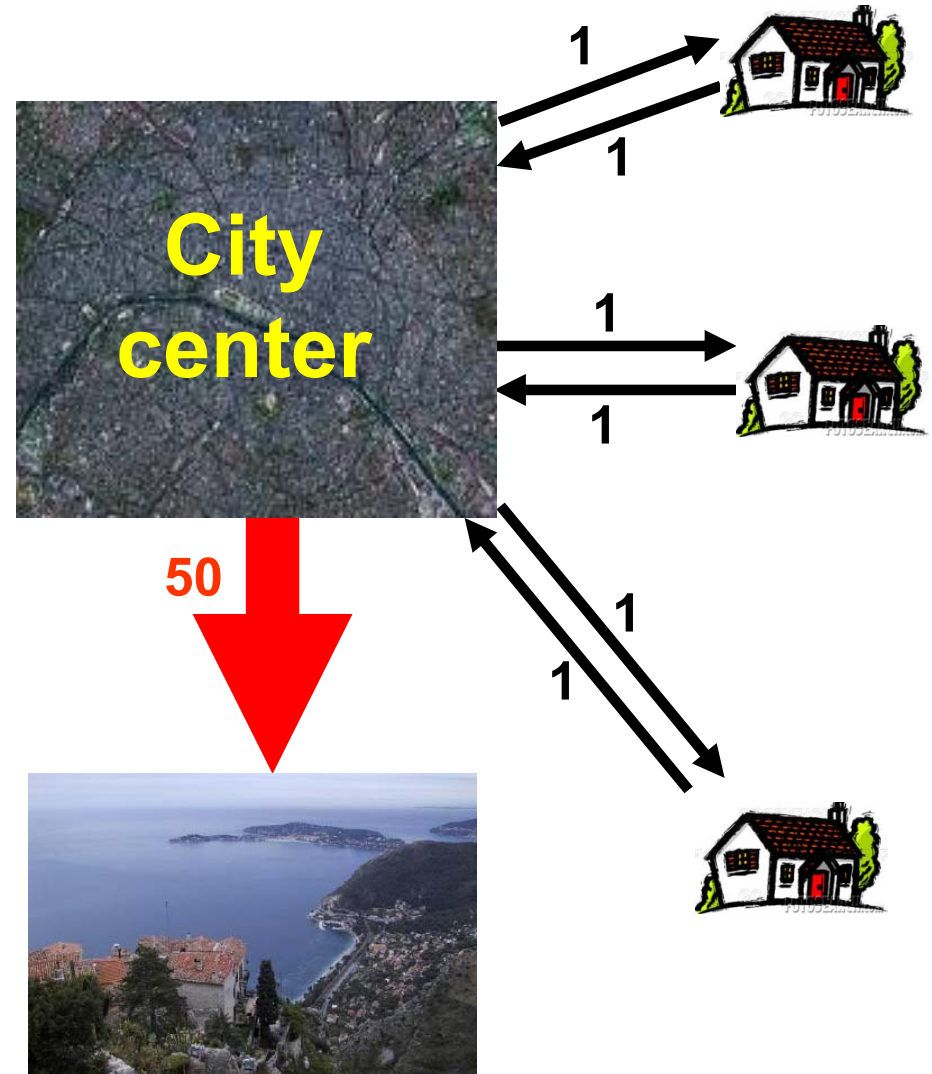
Solution (2): Worst-Case

- Solution (2): plan for **worst-case**
 - i.e. allocate capacity of 10 for each link.



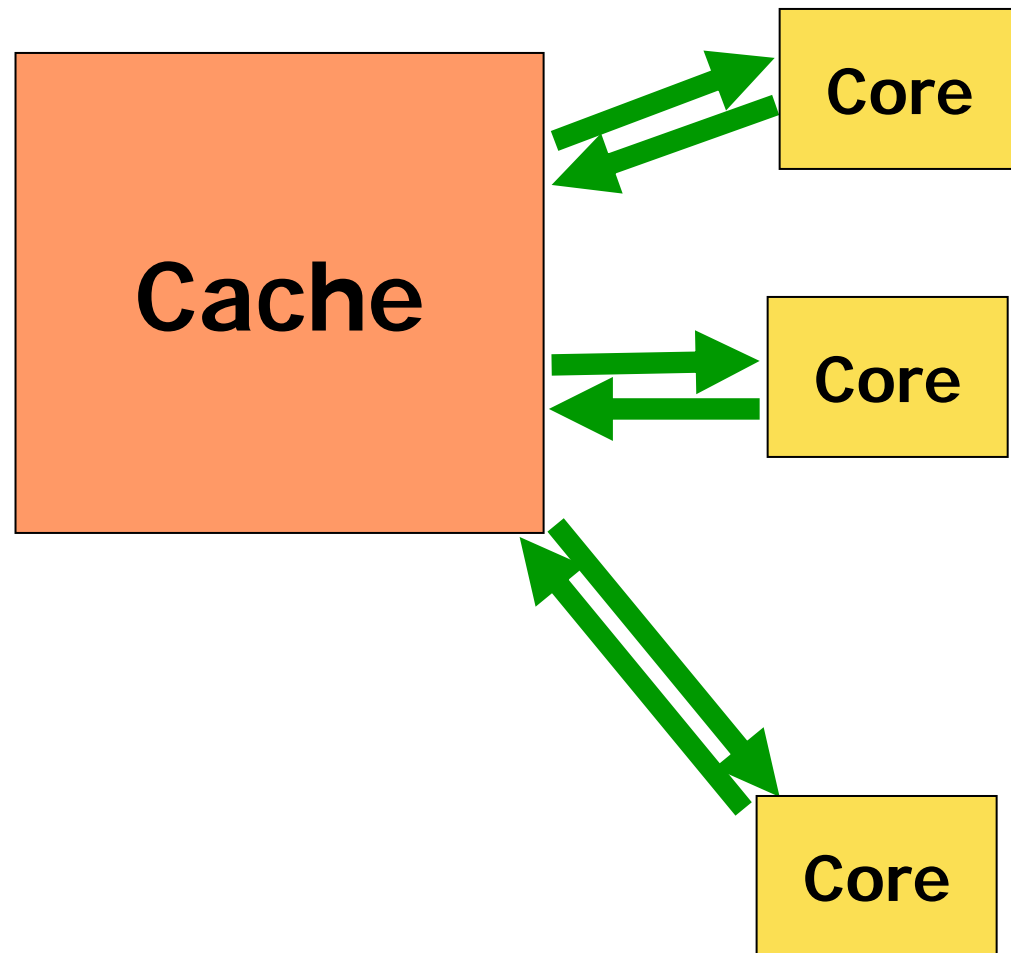
Holidays

- Problem: traffic burst in the holidays
⇒ excessive resources
- Solution (3): **statistical approach**
 - Enough capacity for 99% of the time
 - Allow for occasional congestion



Back to the NoC world

- Similar problems in NoC design process
 - City → Shared cache
 - Suburbs → Cores
 - Many possible traffic matrices: writing, reading, etc.



Motivation

**Tradeoff:
more resources
vs. better guarantee**

➤ **“Easy” solution: Worst-Case Guarantee**

- E.g., no-congestion guarantee
time

*Better when no
congestion is
allowed*

➤ **Objective: Statistical Guarantee**

- E.g., no-congestion guarantee
time

*Better when
power/area are most
important*

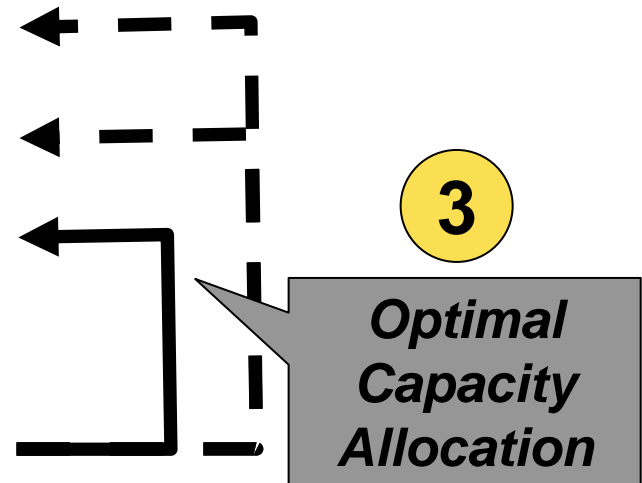
Statistical Approach to NoC Design

Given:

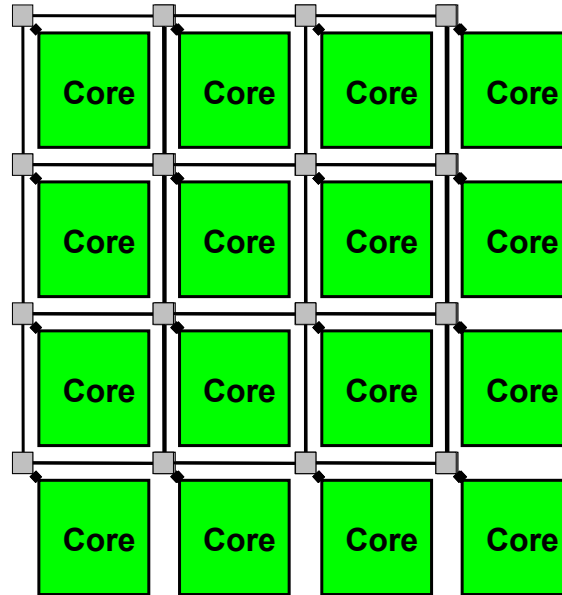
- Traffic matrix distribution
- Topology
- Routing
- Link capacities

- Compute congestion guarantee

- “95% of traffic matrices will receive enough capacity”



CMP (Chip Multi-Processor)



How can we model the future traffic matrix distribution?

Traffic Matrix Distribution

- **Measure** on real large CMP networks?
- **Problems**
 - Their future applications are unknown
 - Their future traffic types are unknown: Between processors? Cache accesses? Control traffic?
 - (Chicken-and-egg problem: traffic might depend on what architecture offers)

Traffic Matrix Distribution

- **General model:** any core can communicate with any core - but with a bounded core input/output rate.
 - Example: each core may send/receive up to 1 Gbps.
 - Used in CMPs [Murali et al. '07] – but also backbone networks [Dukkipati et al. '05], interconnection networks [Towles and Dally '02], VPN networks [Duffield et al. '99], routers [McKeown et al. '96]

Core 1 sends → $T = \begin{pmatrix} 0 & 0.5 & 0.1 & 0.2 \\ 0 & 0 & 0 & 0.6 \\ 0.2 & 0.1 & 0 & 0 \\ 0.1 & 0.1 & 0.3 & 0 \end{pmatrix} \leq 1$

- Reminder: just one model among many...

Statistical Approach to NoC Design

Given:

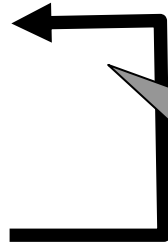
- Traffic matrix distribution
- Topology
- Routing
- Link capacities

- Compute congestion guarantee**

- “95% of traffic matrices will receive enough capacity”

3

*Optimal
Capacity
Allocation*

A diagram showing a feedback loop. A grey box labeled '3' contains the text 'Optimal Capacity Allocation'. A black arrow points from the box to the left, then down, then left again, ending at the red text 'Compute congestion guarantee' in step 2.

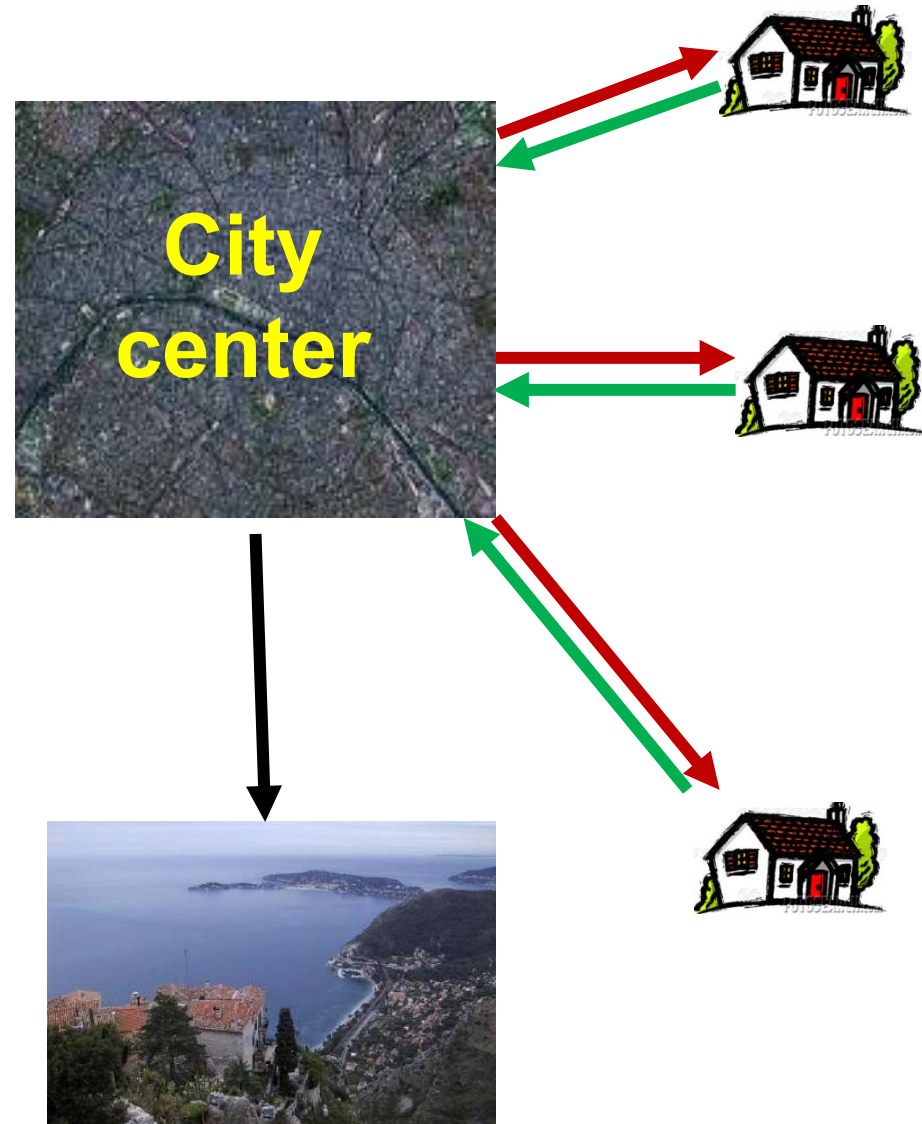
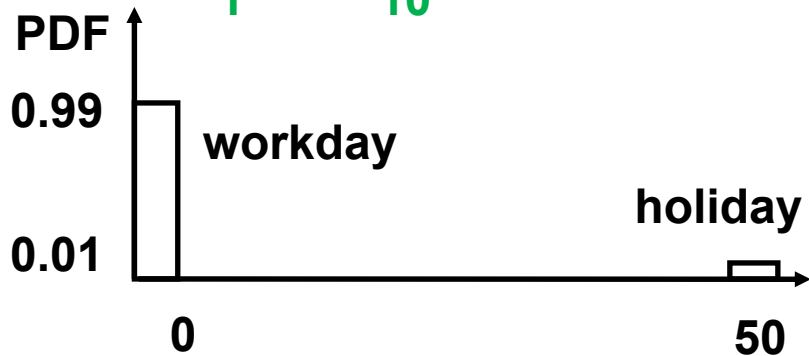
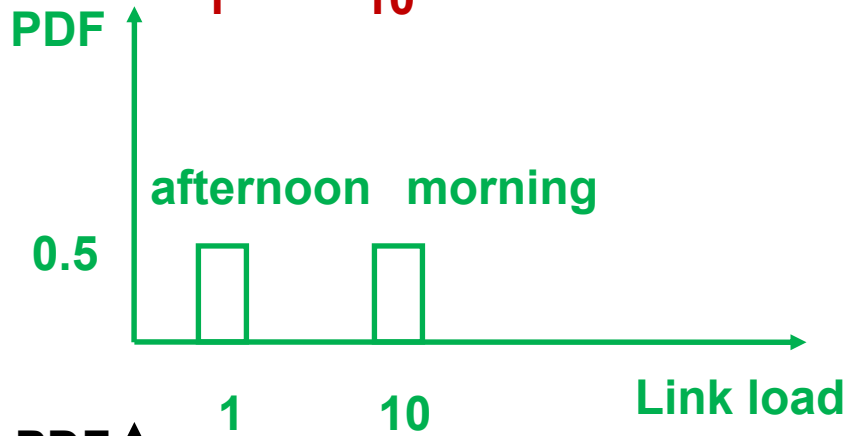
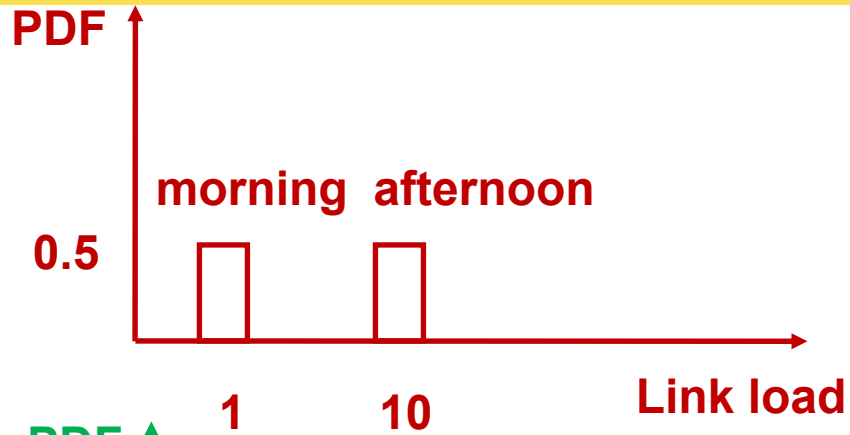
Compute Congestion Guarantee

Show that “**95%** of all traffic matrices will receive enough capacity on link ℓ ”

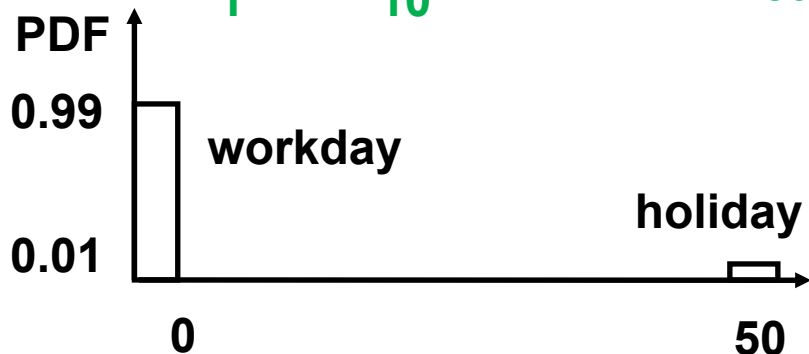
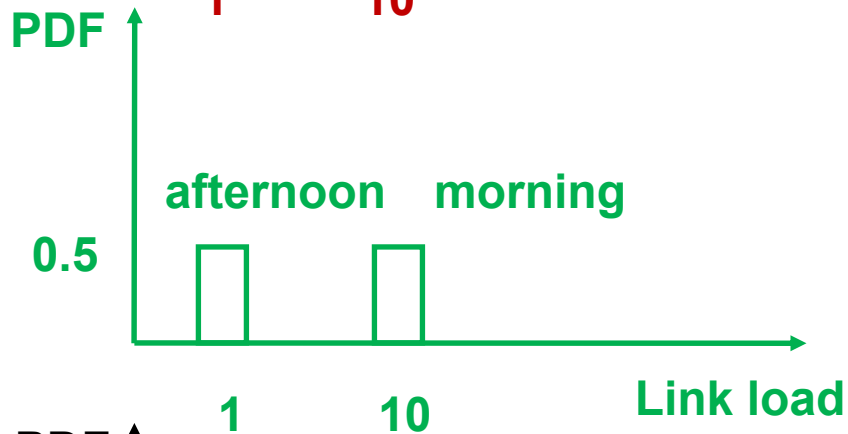
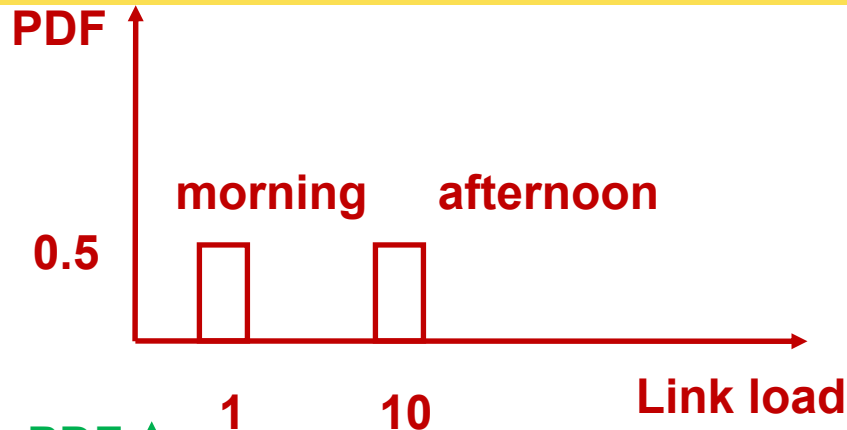


1. Compute ***Traffic-load distribution Plot*** (or ***T-Plot***) for link ℓ
2. Show that the load on link ℓ is less than its capacity for **95%** of traffic matrices.

Link T-Plot

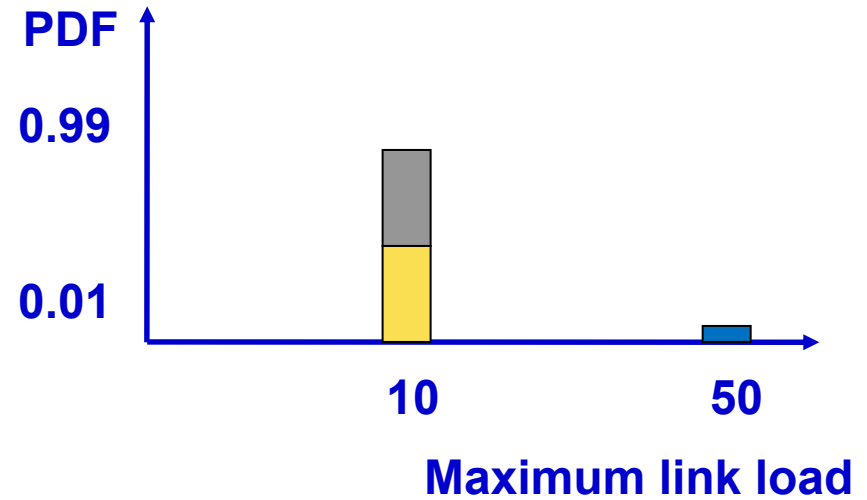
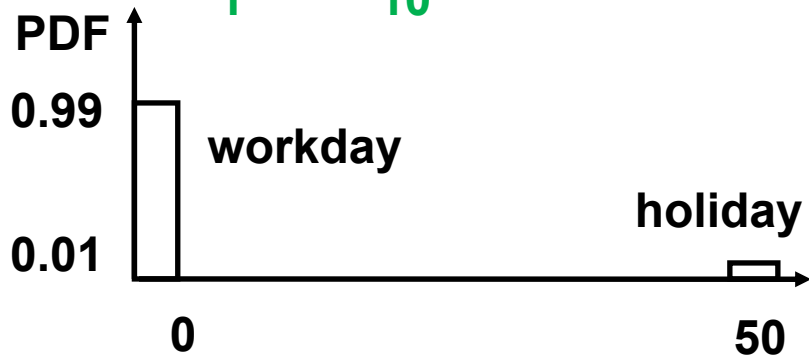
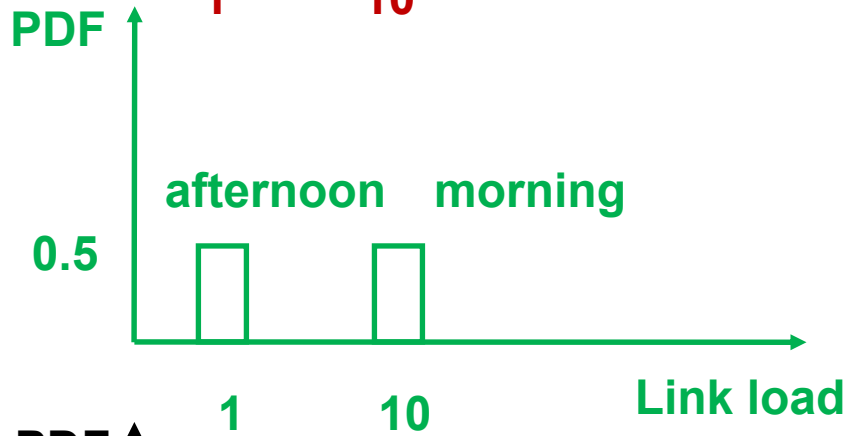
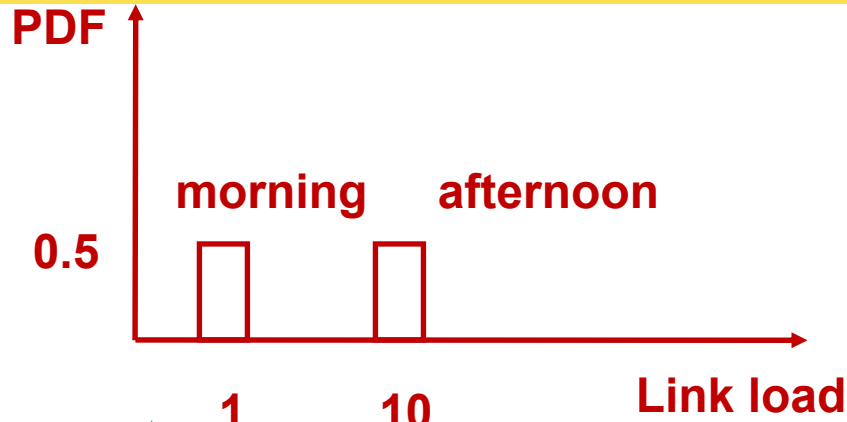


Global T-Plot

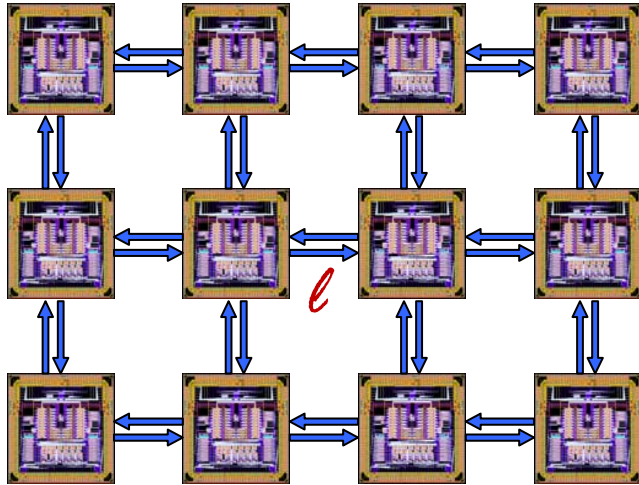


- We want to provide statistical guarantees on the whole network
 - i.e. on **all** links
- **Global T-Plot:** for each traffic matrix T , measure maximum load among **all** links

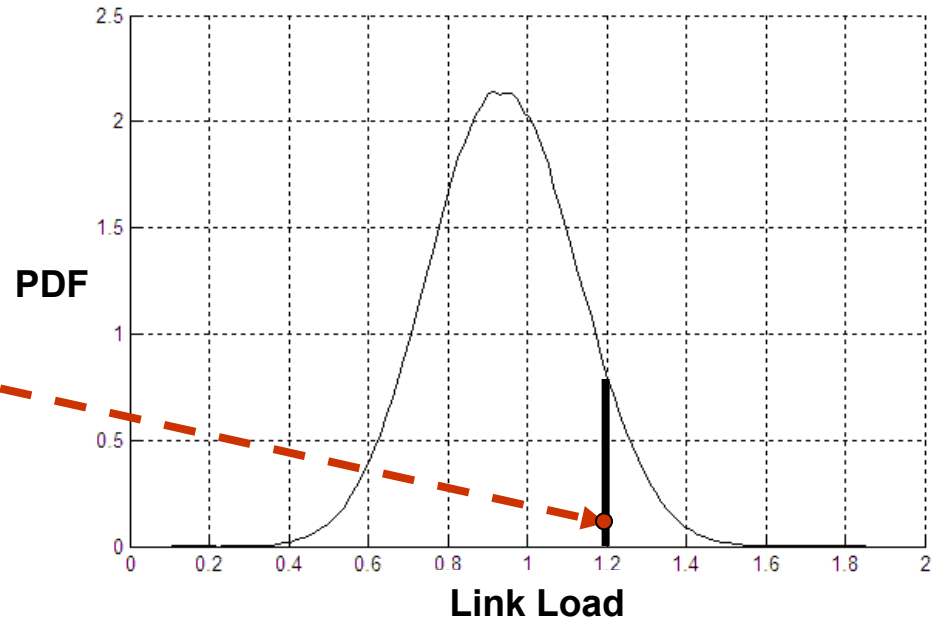
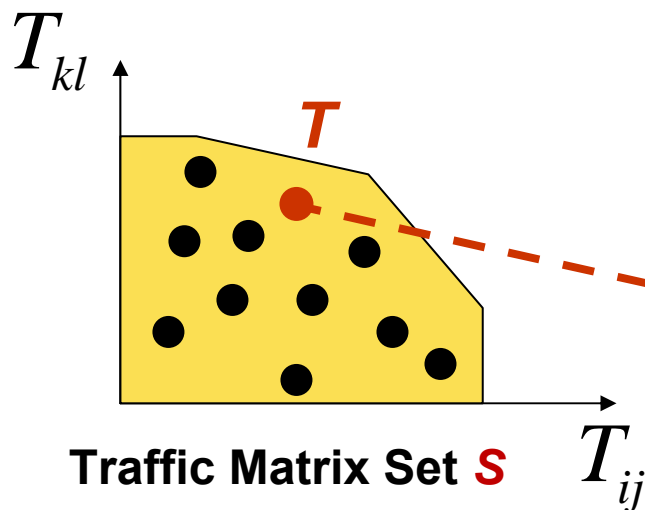
Global T-Plot



Local T-Plots in NoCs

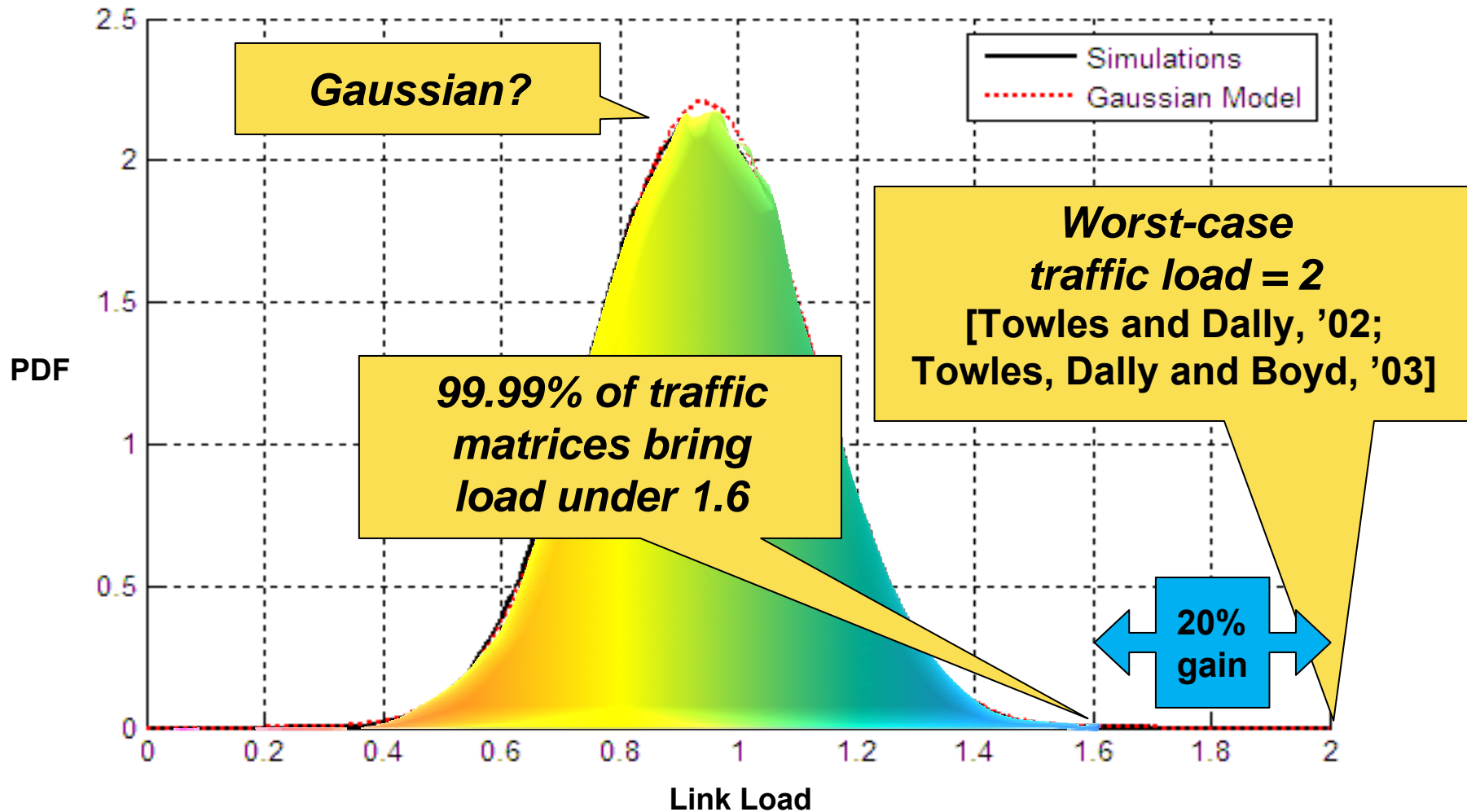


- Given:
 - Traffic matrices $T \in \mathcal{S}$
 - 3x4 mesh topology
 - XY routing
 - Link l
- Find T-Plot on l



T-Plot

➤ Close-up view:

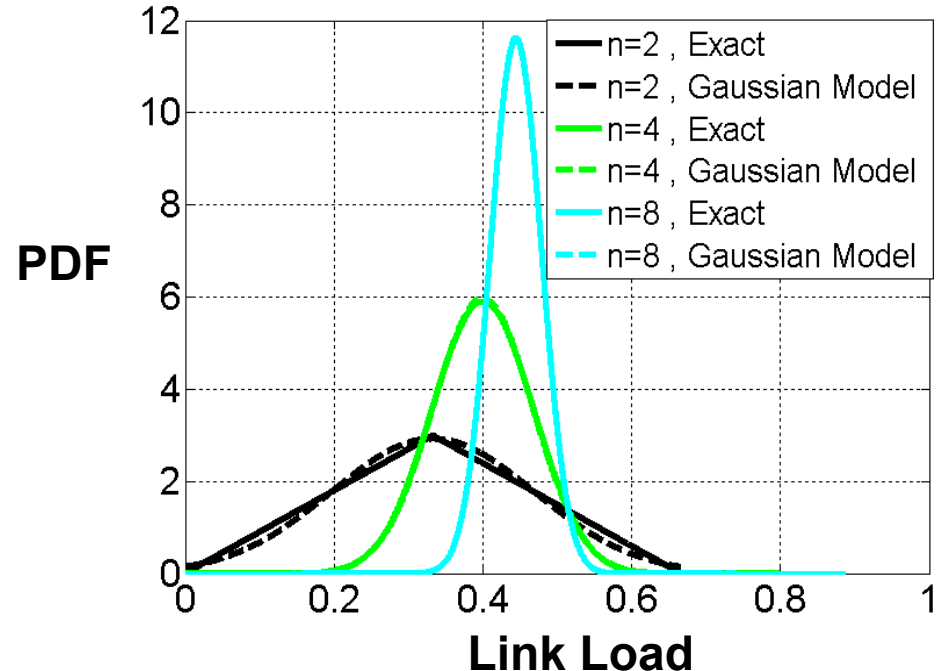
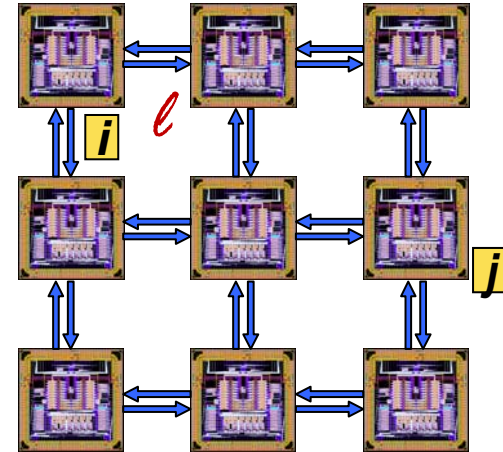


Are T-Plots Gaussian?

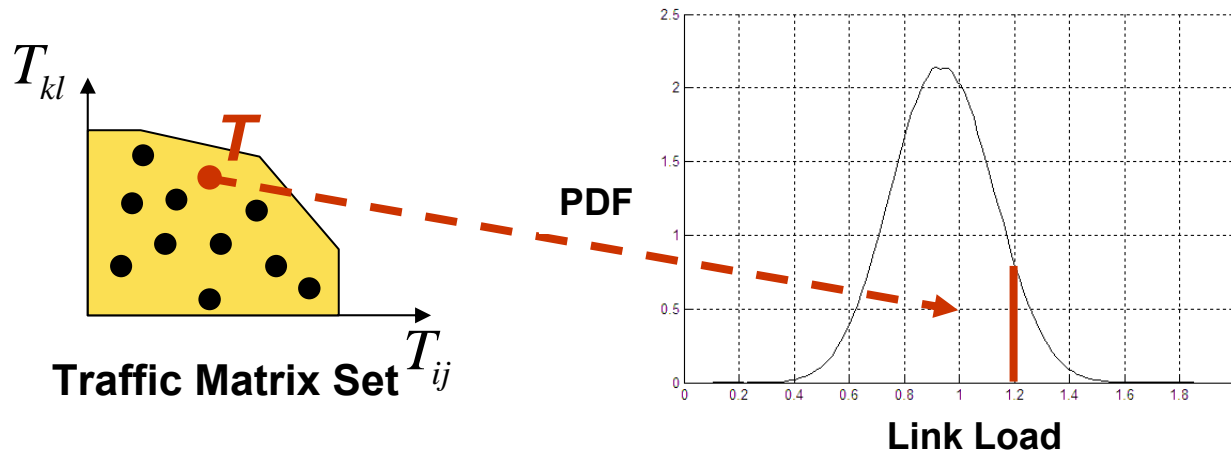
- Some T-Plots may be well approximated as Gaussian.
- **Intuition (Central Limit Theorem):** when N grows, the sum of N i.i.d. (independent and identically distributed) random variables converges to a Gaussian distribution.

Gaussian T-Plot Example

- Example: $n \times n$ mesh with XY routing.
- Assume i.i.d traffic from i to j
- **Theorem:** As n grows, T-Plot for ℓ converges to Gaussian.

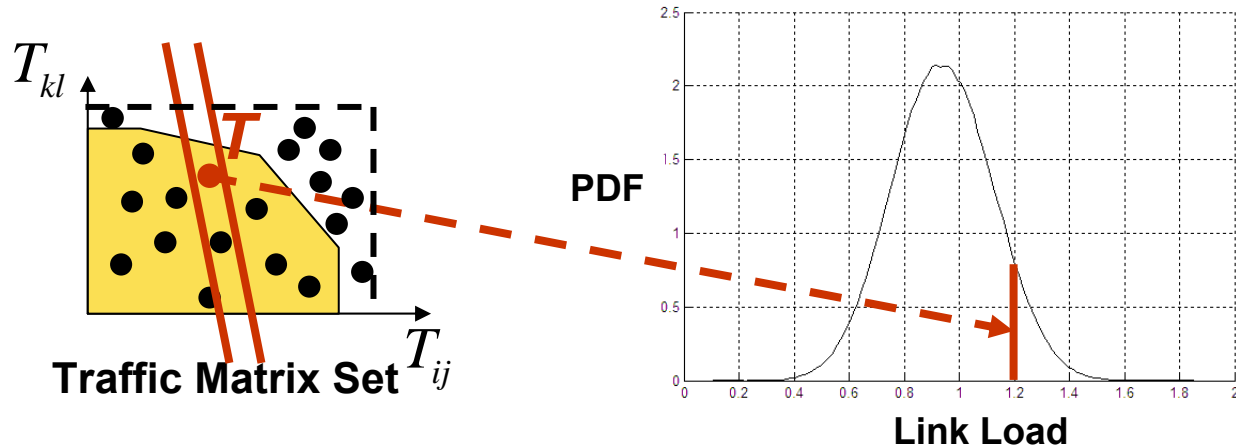


Computing the T-Plot



- **Theorem:** for an arbitrary graph and routing, computing the *T-Plot* is **#P-complete**.
- #P-complete problems are at least as hard as NP-complete problems.
 - NP: “Is there a solution?”
 - #P: “How many solutions?”

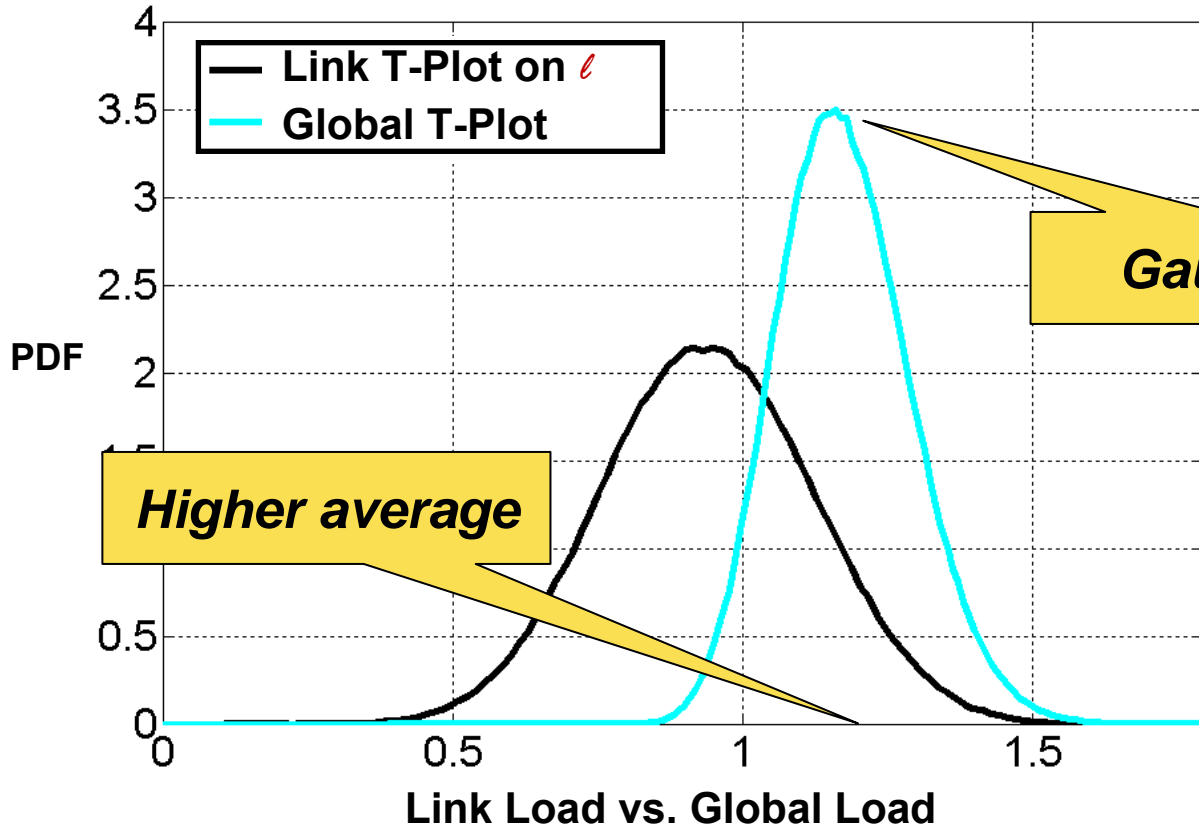
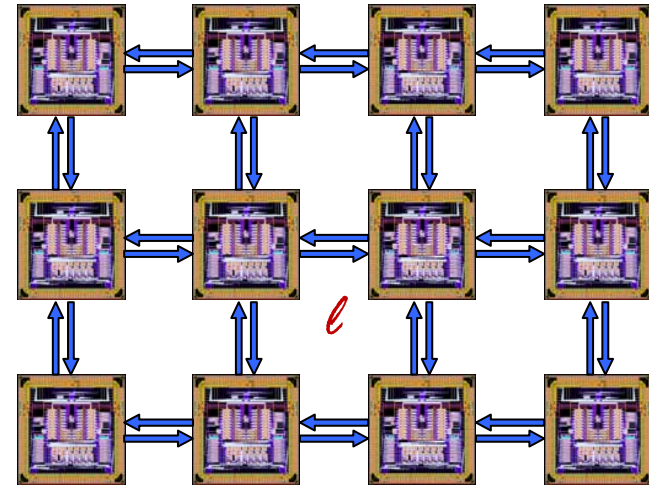
Computing the T-Plot



- Idea: use Monte Carlo simulations to approximate T-Plots
 - Plot many points, and count number of points to approximate density.

Link T-Plot Vs Global T-Plot

- **Global T-Plot:** for each traffic matrix T , measure maximum load among **all** links



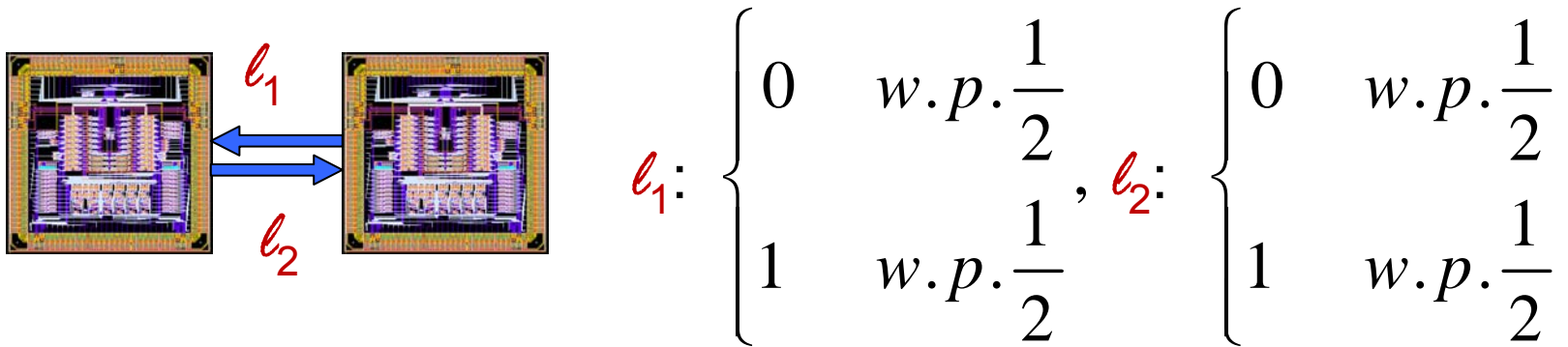
Gaussian?

Higher average

**Model?
Assume
independent
links...**

Link-Independent Model: Simple Example

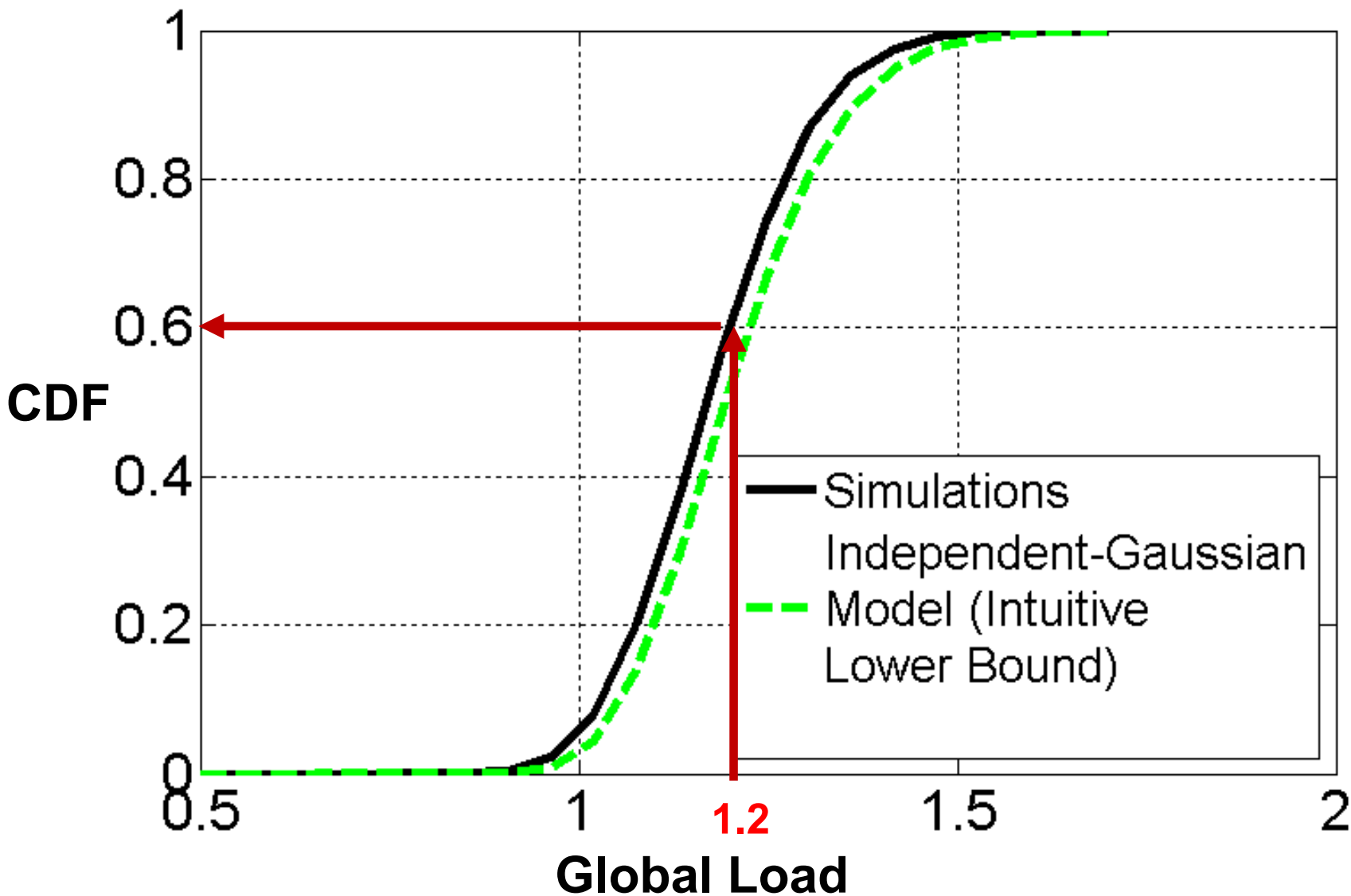
➤ Assume:



➤ If l_1 and l_2 independent:

Global Load:
$$\begin{cases} 0 & w.p. \frac{1}{4} \\ 1 & w.p. \frac{3}{4} \end{cases}$$

Independent-Gaussian Model



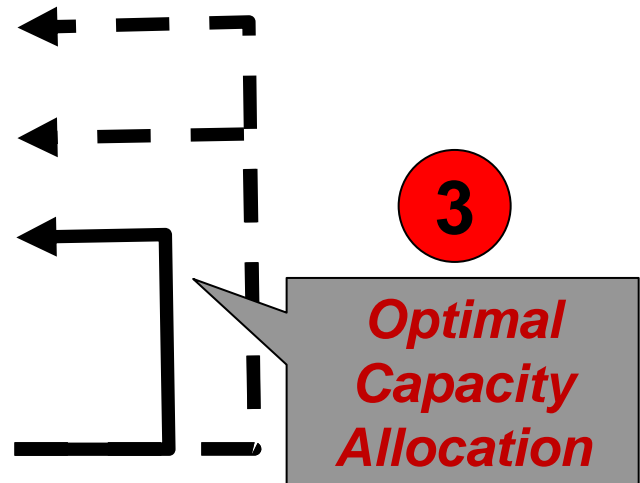
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Given:

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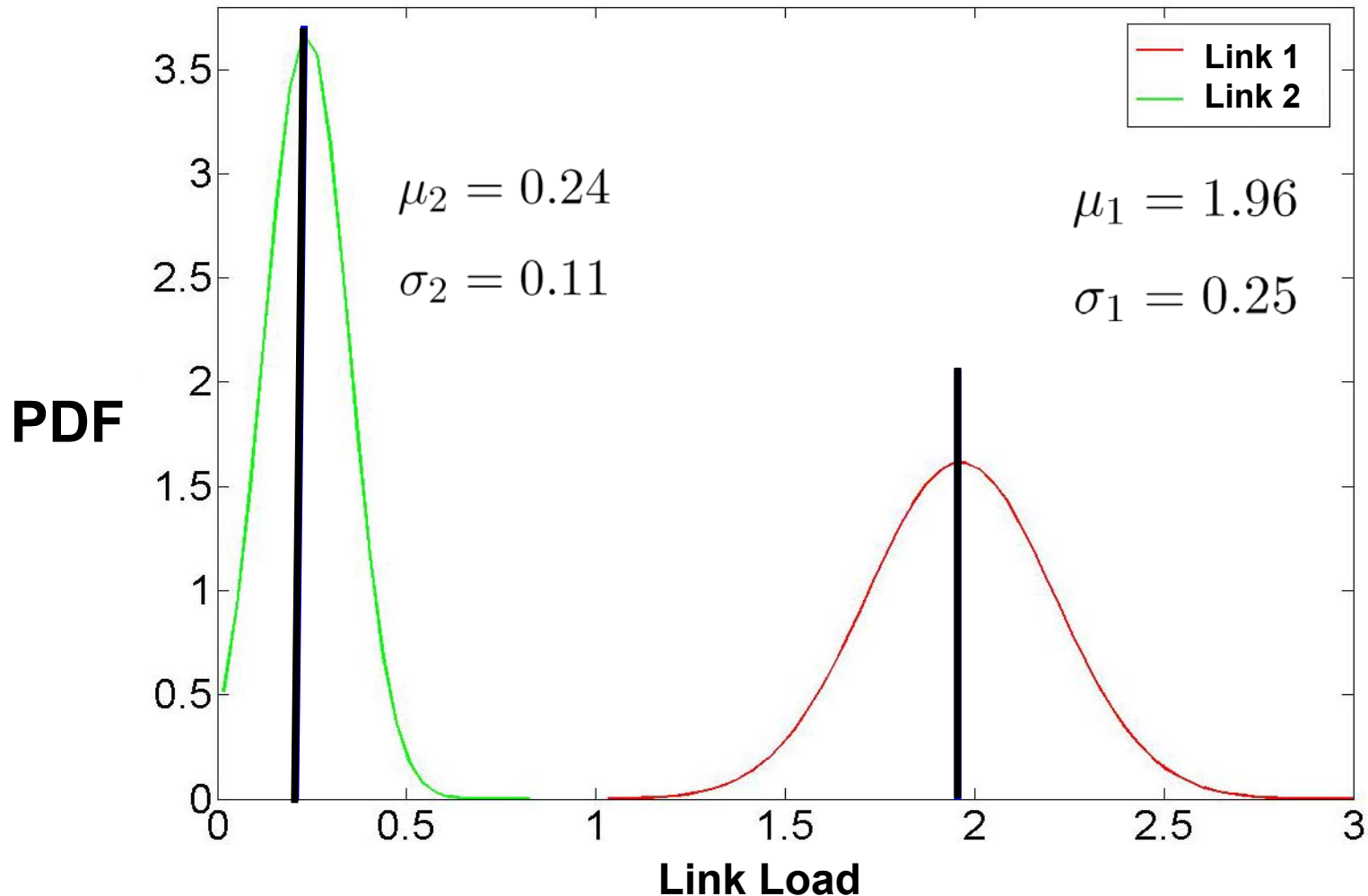
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- “95% of traffic matrices will receive enough capacity”



Capacity Allocation Intuition

- **Idea:** given a total capacity, distribute it so that the overflow probability on each link is **the same**.



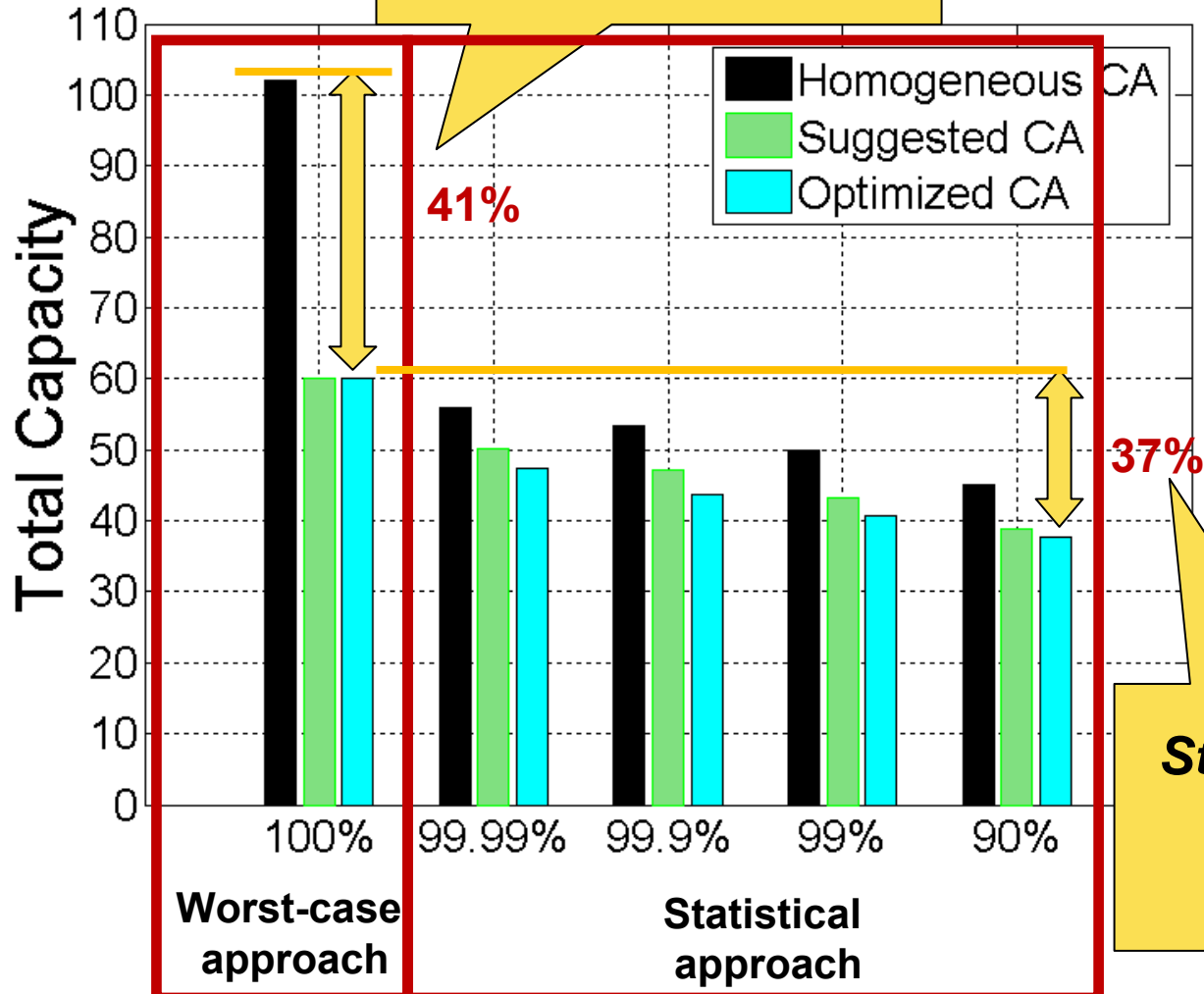
Capacity Allocation Intuition

- **Idea:** given a total capacity, distribute it so that the overflow probability on each link is **the same**.
- **Theorem:** if Gaussian-independent model holds, with same standard-deviation σ on all links, then **this scheme is optimal**.

Capacity Allocation Performances

- Total needed capacity vs Capacity Allocation (CA)

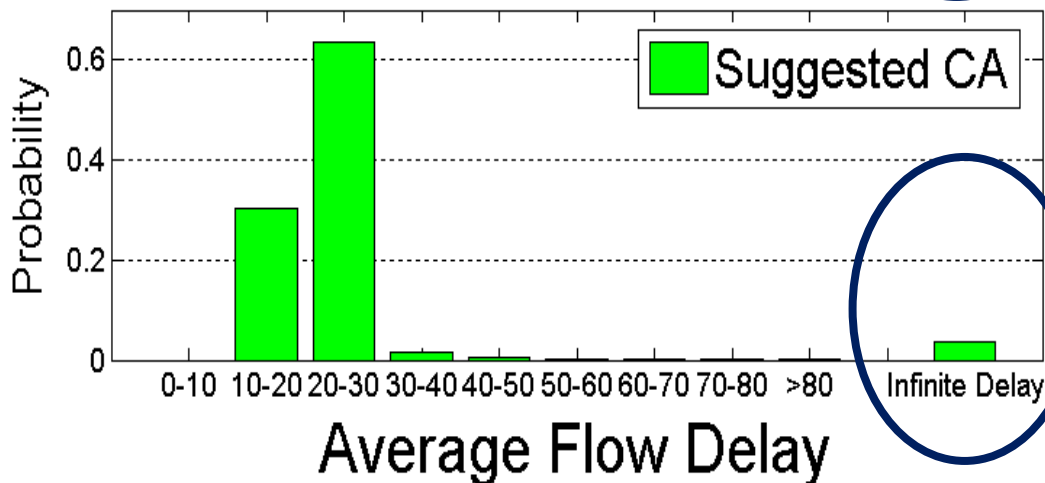
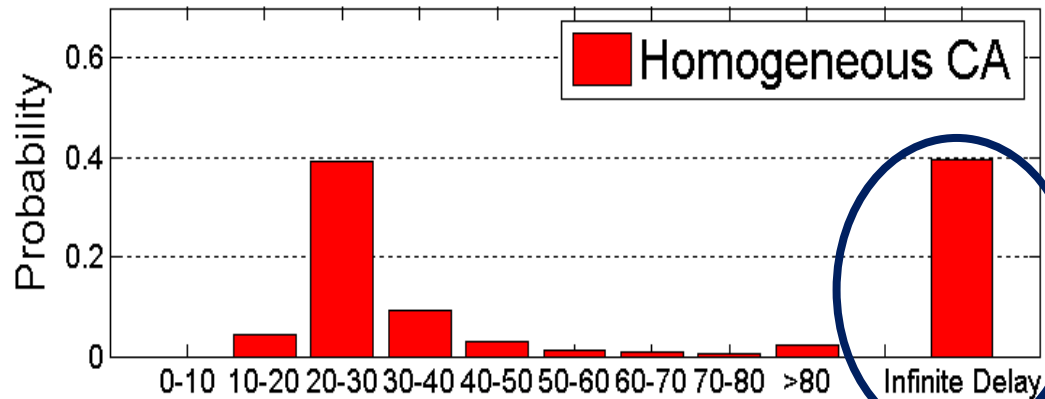
Worst-case capacity saving with different link capacities



Statistical gain with 90% guarantee

Average Flow Delay Model

- Assume M/M/1 delay model
- Average flow delay distribution over all traffic matrices, for two different Capacity Allocation (CA) schemes:



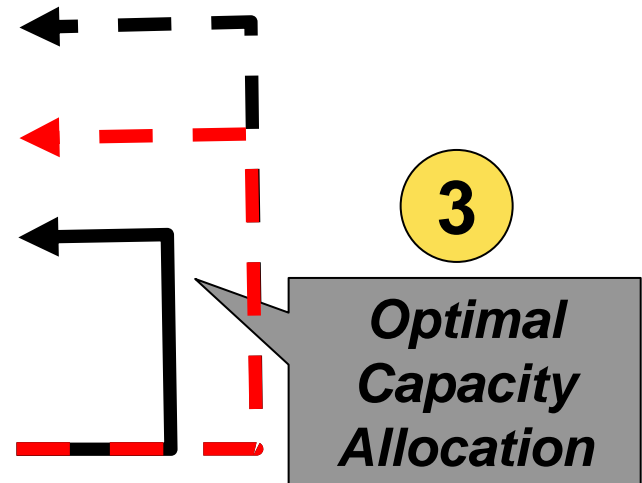
Statistical Approach to NoC Design

Given:

- Traffic matrix distribution
- Topology
- **Routing**
- Link capacities

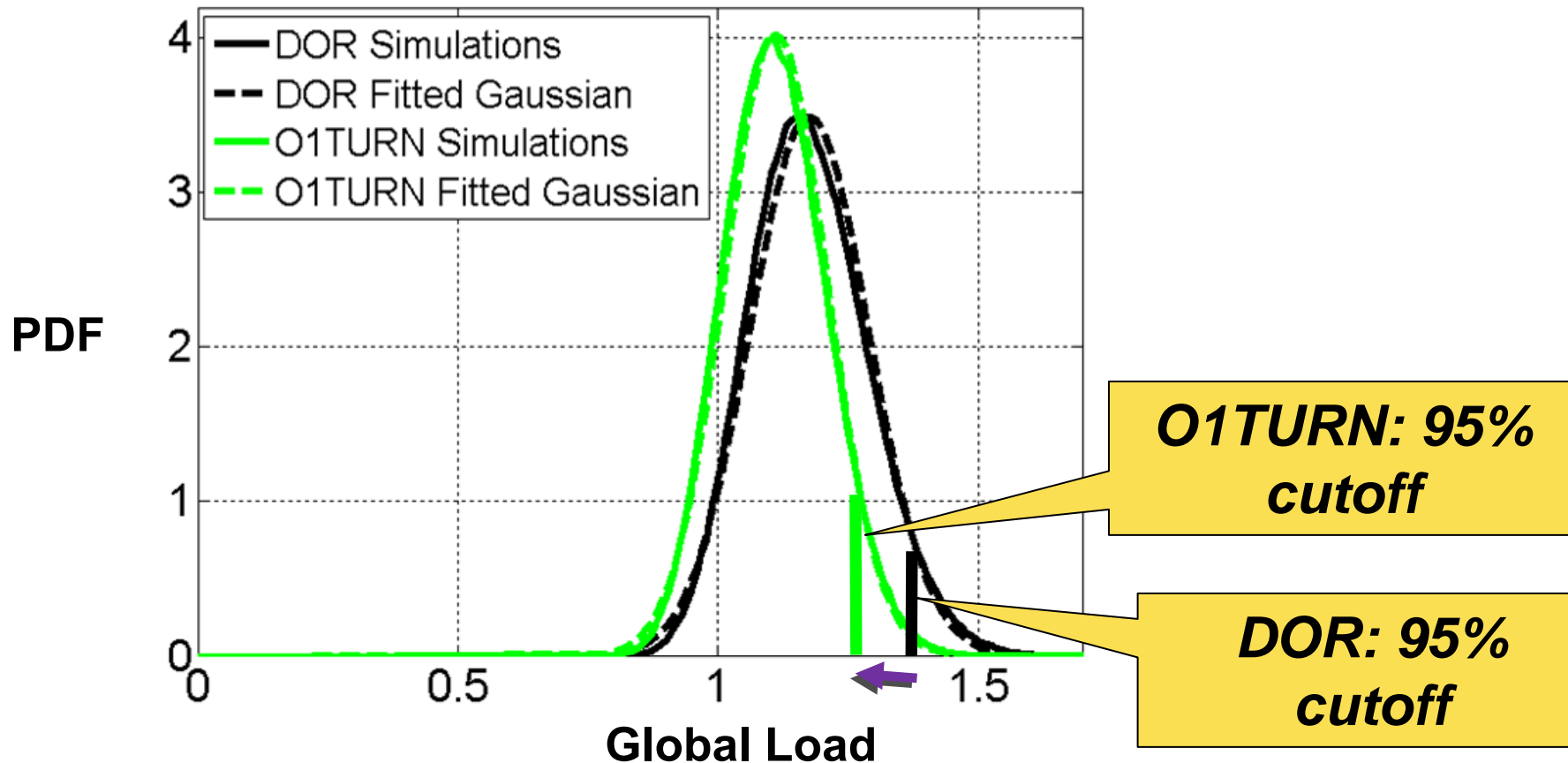
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- “95% of traffic matrices will receive enough capacity”



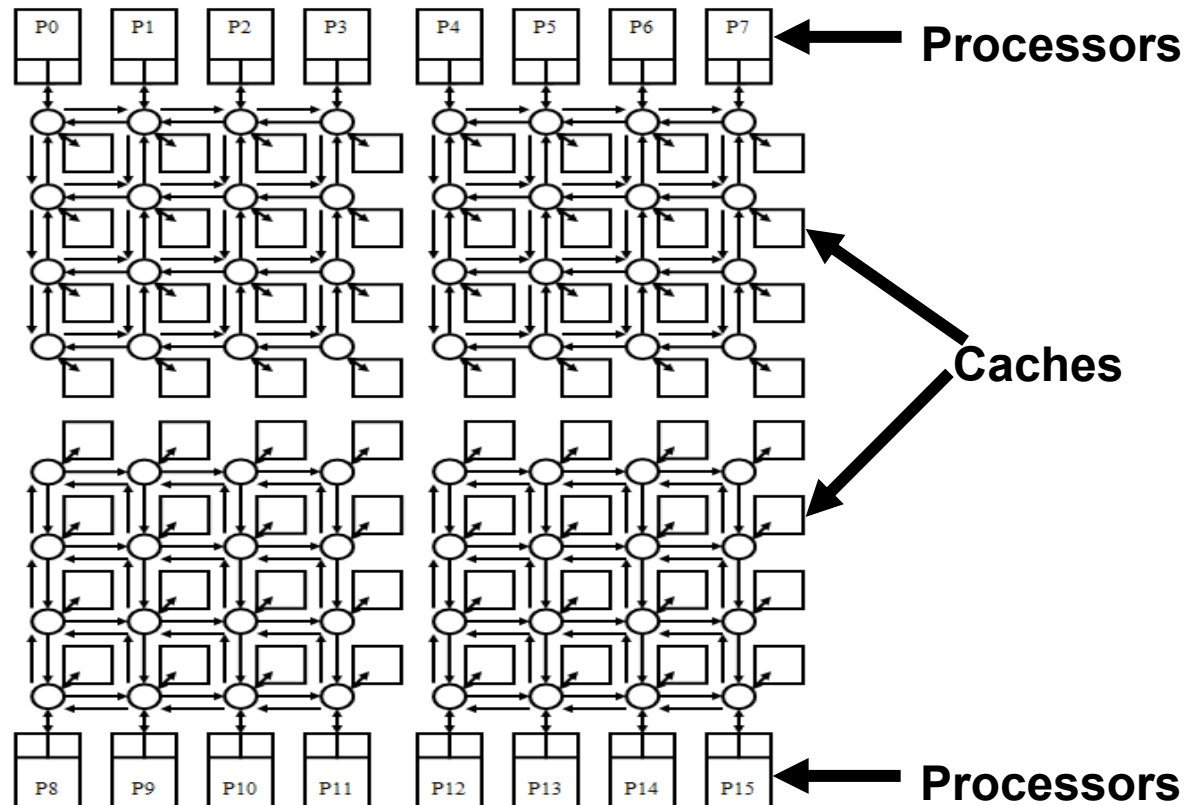
Comparing routing algorithms

- DOR (XY) and O1TURN ($\frac{1}{2}$ XY, $\frac{1}{2}$ YX) on 3x4 mesh:



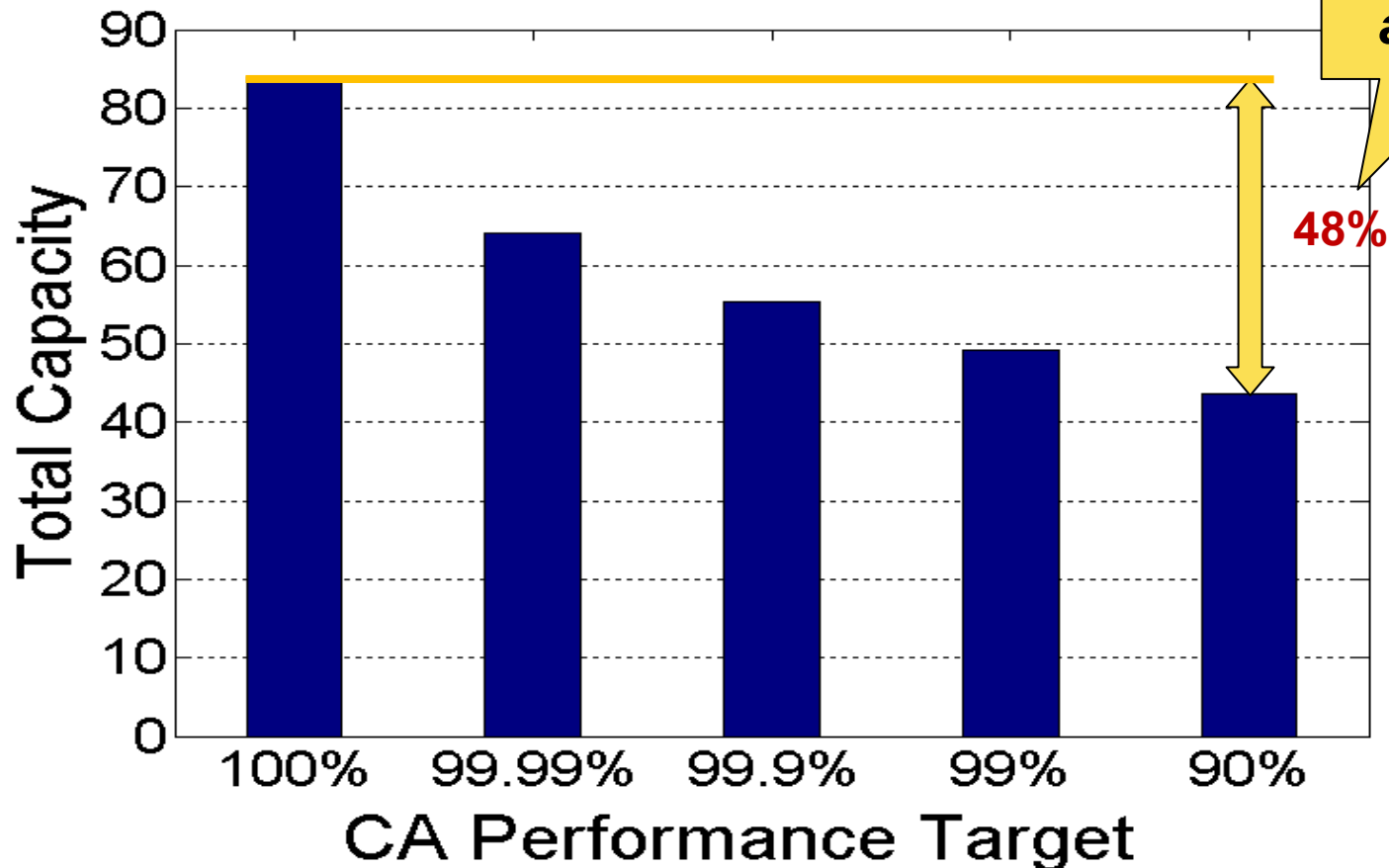
NUCA network

- More complex CMP architectures show similar results
 - **NUCA** (Non-Uniform Cache Architecture) with sharing degree 4.
 - Traffic model: each core (cache) may only send/receive traffic to/from caches (cores) in its sub-network.



NUCA network – Total capacity

- Total capacity required for various Capacity Allocation (CA) targets.



Gain of statistical approach

Summary

- Statistical approach to NoC design
 - Deals with several traffic matrices
 - Can optimize capacity allocation and routing
- Can be applied to any
 - Traffic matrix distribution
 - Topology
 - Oblivious routing
 - Link capacities

Thank you.