Congestion and Centrality in Data Networks

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Motivation

Congestion in a Manhattan Network Delay and Total Number of Packets in the

Network

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Betweenness Centrality

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Motivation Congestion in a Manhattan Network Delay and Total Number of Packets in the Network Mean Field Approximation

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Introduction

In a "simple" data/packet network the onset of congestion (a dynamical characteristic) depends on the average of all shortest path lengths (a topological characteristic).

Ohira and Sawatari, 1998; Fukś and Lawniczak 1999; Solé and Valverde 2001; Woolf *et. al* 2002

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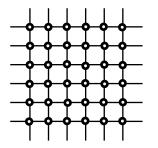
Introduction

- In a "simple" data/packet network the onset of congestion (a dynamical characteristic) depends on the average of all shortest path lengths (a topological characteristic).
- Is the above result valid for all network's topologies?
- Can this result help us when simulating very large networks?

Ohira and Sawatari, 1998; Fukś and Lawniczak 1999; Solé and Valverde 2001; Woolf *et. al* 2002

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Congestion in a Manhattan Network



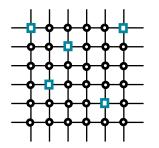
- Consider a Manhattan-toroidal network with S nodes
- Each node contains a queue where packets can be stored in transit (if the node is busy)

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Congestion in a Manhattan Network



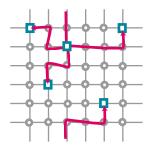
- Consider a Manhattan-toroidal network with S nodes
- Each node contains a queue where packets can be stored in transit (if the node is busy)
- ► The proportion of sources/sinks of traffic is ρ ∈ (0, 1], i.e.#sources = ρS
- ► Each traffic source generates, on average, the same amount of traffic *λ*

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Congestion in a Manhattan Network



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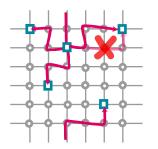
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 The packets are sent through the shortest and/or *less busy* route



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Congestion in a Manhattan Network



Packet generation, hop movement, queue movement and updating of the routing table occurs at one time step.

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- ► The proportion of sources/sinks of traffic is ρ ∈ (0, 1], i.e.#sources = ρS
- Each traffic source generates, on average, the same amount of traffic λ
- The packets are sent through the shortest and/or *less busy* route
- If one node is busy (queue busy), then another route is chosen

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Delay and Congestion

• τ_{sd} is the journey time from s to d



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Delay and Congestion

- τ_{sd} is the journey time from s to d
- ▶ For low load, $\tau_{sd} \approx \ell_{sd}$
- For higher load, $\tau_{sd} \approx \ell_{sd} +$ delays due to the queuing.



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Delay and Congestion

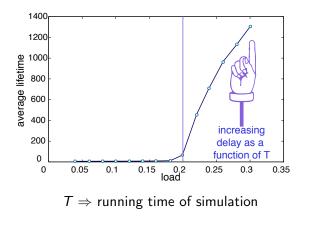
- τ_{sd} is the journey time from s to d
- ▶ For low load, $\tau_{sd} \approx \ell_{sd}$
- ▶ For higher load, $\tau_{sd} \approx \ell_{sd}$ + delays due to the queuing.
- ► If the traffic load increases even further, then at the critical load \u03c0_c, the queues of some nodes will grow unbounded and the average delay time diverge.
- At this critical load, we consider that the network is congested.

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Delay and Congestion



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Total Number of Packets in the System and Congestion

Total number of packets:



$$N(t) = \sum_{i=1}^{S} Q_i(t)$$

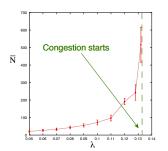
- $Q_i(t)$ is size of queue *i*.
- At the free flow state, $\bar{N} = \lim_{T \to \infty} 1/T \sum_{T} N(T)$ is finite

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- At the free flow state, $\bar{N} = \lim_{T \to \infty} 1/T \sum_{T} N(T)$ is finite
- At the congestion point, the queues of the congested nodes start growing unbounded $\Rightarrow \bar{N} \rightarrow \infty$

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Definitions and Assumptions

- ► The network is represented by the graph G = (V, E) where V is the set of nodes (vertices) and E is the set of links (edges).
- The total number of nodes is denoted by S.
- The graph is undirected and connected



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- The total number of nodes is denoted by S.
- The graph is undirected and connected
- ► The minimum distance between vertices s ∈ V and d ∈ V is denoted by ℓ_{sd} (shortest path between s and d.)
- The characteristic path length

$$\bar{\ell} = \frac{1}{S(S-1)} \sum_{v \in \mathcal{V}} \sum_{d \in \mathcal{V} \setminus v} \ell_{sd}$$

(sometimes $\overline{\ell}$ is referred as the diameter of the network).



Mean Field Approximation

Little's Law

"The average number of customers in a queuing system is equal to the average arrival rate of customers to that system, times the average time spent in the system", Kleinrock 1975

Formulation

$$\frac{d N(t)}{d t} = \rho S \lambda - \frac{N(t)}{\tau(t)}.$$



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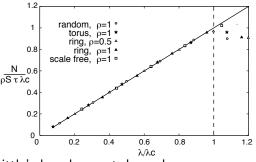
• $\rho S \lambda$ is the average arrival rate to the queues per unit of time,

- au(t) is the average time spent in the system, and
- $N(t)/\tau(t)$ is the number of packets delivered per unit of time.



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Mean Field Approximation. Little's Law



← From the steady state solution

 $\rho S\lambda - \frac{N}{\tau} = 0$

Little's law does not depends on

- the arrival distribution of packets to the queue
- or the service time distribution of the queues.
- Also it does not depends upon the number of queues in the system or upon the queuing discipline within the system.



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Mean Field Approximation. Congestion

Estimating the time delay

- If the load is low, the delay is time is given by the length of the shortest path, then the average delay is the average of the shortest paths \(\bar{\tau}\) < \(\bar{\ell}\).</p>
- If the load is high, the delay time is the length of the shortest path plus the time a packet spends on the queues



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Mean Field Approximation. Congestion

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- If the load is high, the delay time is the length of the shortest path plus the time a packet spends on the queues
- if we assume that, on average each queue contains \bar{N}/S packets

$$au(t) pprox ar{ au} pprox ar{\ell}(1+ar{Q}) = ar{\ell}\left(1+rac{ar{N}}{S}
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Mean Field Approximation. Congestion

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$$au(t) pprox ar{ au} pprox ar{ heta}(1+ar{ extsf{Q}}) = ar{ heta}\left(1+rac{ar{ extsf{N}}}{ets}
ight)$$

we are approximating the delay time with the queue's average size



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Mean Field Approximation. Congestion

Estimating the critical load λ_c

From the steady state solution dN(t)/dt = 0 the traffic load generated is

$$\lambda = rac{1}{
ho ar{\ell} (1+S/ar{N})}$$

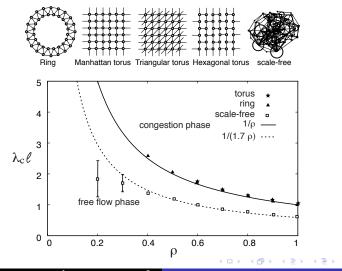
▶ At the congestion point the average number of packets diverges, i.e. $\bar{N} \to \infty$ so the critical load is

$$\lambda_c = rac{1}{
ho \overline{\ell}}$$



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Congestion and Centrality in Data Networks

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Betweenness Centrality

- Consider the journey time between two nodes in the network where there is at least two shortest paths between the nodes
- The journey time of two shortest paths with the same length can be very different due to the different patterns of usage of the routes
- The reason is that some nodes are more "prominent" because they are highly used when transferring packet-data.
- A way to measure this "importance" is by using the concept of node betweenness centrality (also called *load* or just betweenness).



Centrality: Definitions

- ▶ The number of shortest paths from $s \in \mathcal{V}$ to $d \in \mathcal{V}$ is denoted by σ_{sd}
- ► The number of shortest paths from s to d that some v ∈ V lies on is denoted by σ_{sd}(v)
- The pair-dependency of a pair s, d ∈ V on an intermediary v ∈ V is

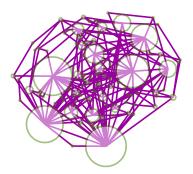
$$\delta_{sd}(v) = \frac{\sigma_{sd}(v)}{\sigma_{sd}}$$

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Definitions Betweenness Centrality and Congestion Extensions

Betweenness/load/Betweenness Centrality

$$\mathcal{C}_B(v) = \sum_{s \in \mathcal{V}} \sum_{d \in \mathcal{V} \setminus s} \delta_{sd}(v), \quad v \in \mathcal{V}$$



A small modification

 $C_B(v) = \mathcal{C}_B(v) - 1$



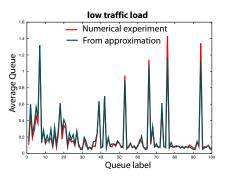
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An improvement to τ

 Characterise the node usage using the normalised betweenness centrality

$$\hat{C}_B(w) = \frac{C_B(w)}{\sum_{v \in \mathcal{V}} C_B(v)}$$

• approximate the average queue size using $\bar{Q}_w \approx \hat{C}_B(w)\bar{N}$



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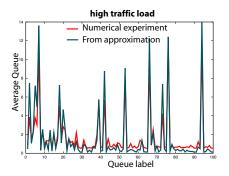


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An improvement to au

the approximation to the average delay time is

$$ar{ au} pprox ar{\ell} + rac{1}{S(S-1)} \sum_{s \in V} \sum_{d \in \mathcal{V} \setminus v} \left(\sum_{v \in \mathcal{R}_{sd}} \hat{\mathcal{C}}_B(v) ar{N}
ight) = ar{\ell} + Dar{N}$$

where $v \in \mathcal{R}_{sd}$ is the set of nodes visited by the route • ! we are taking the average of averages



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where $\textit{v} \in \mathcal{R}_{\textit{sd}}$ is the set of nodes visited by the route

- ! we are taking the average of averages
- \blacktriangleright Using the new approximation to $\bar{\tau}$

$$\lambda_c = \frac{1}{\rho S D}$$

Definitions Betweenness Centrality and Congestion Extensions

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An improvement to λ_c

The equation

$$\lambda_c = \frac{1}{\rho S D}$$

- simplifies to $\lambda_c = 1/(\rho \bar{\ell})$ in the case of regular networks.
- this is obtained by exploiting the property that

$$\ell_{\textit{sd}} = \sum_{w \in \mathcal{W}} \delta_{\textit{sd}}(w) - 1$$

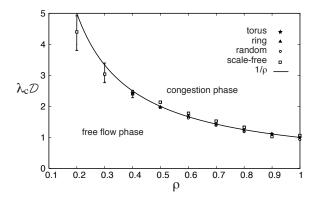
where \mathcal{W} is the set of nodes visited by the shortest paths from s to d.

Definitions Betweenness Centrality and Congestion Extensions

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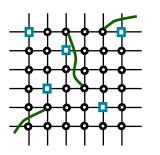
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Betweenness Centrality



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Fukś and Lawcnizack Observation



- Take a Manhattan toroidal network
- add a few new random links
- the onset of congestion occurs more readily when adding these new links that in the original network
- this is because, the new links "attract" traffic, the nodes containing the extra links congest more easily

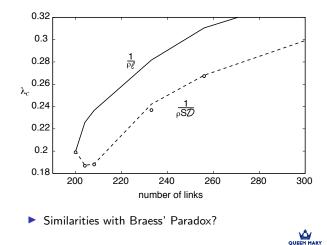
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Definitions Betweenness Centrality and Congestion Extensions

Fukś and Lawcnizack Observation

- The original network has 200 links
- we compare the prediction using $\lambda_c = 1/(\rho \ell)$ and $\lambda_c = 1/(\rho SD)$



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Another improvement

The queue discipline is $\ensuremath{\mathsf{M}}\xspace/\ensuremath{\mathsf{D}}\xspace/\ensuremath{\mathsf{1}}\xspace$

The average of the queue is approximated by

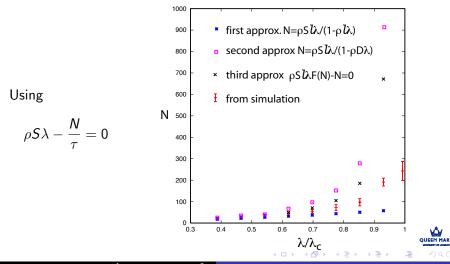
$$ar{Q}_i = \Lambda_i + rac{\Lambda_i^2}{2(1-\Lambda_i)} pprox ar{N} D_i$$

The average delay is

$$\tau_i = \frac{\Lambda_i}{2(1 - \Lambda_i)} \approx \frac{1 + \bar{N}D_i - \sqrt{1 + (\bar{N}D_i)^2}}{2(\sqrt{1 + (\bar{N}D_i)^2} - \bar{N}D_i)}$$



Number of Packets in the Network

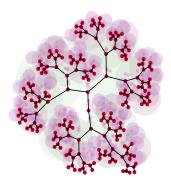


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Limitations

Limitations



- The prediction doesn't work for trees
- The delay time approximation is poor

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