



# **A Game-Theoretic Approach Towards Congestion Control in Communication Networks**

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# OUTLINE

- ❖ INTRODUCTION
- ❖ CURRENT CONGESTION CONTROL POLICIES IN A ROUTER OR SWITCH
- ❖ DWS(DIMISHING WEIGHT SCHEDULERS)
- ❖ PROPERTIES OF DWS
- ❖ SIMULATION RESULTS
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# INTRODUCTION(1/3)

## ❖ Network Congestion Control

⌘ User-end : end-to-end congestion control( e.g. TCP)

⌘ Router or Switch : resource sharing policy( e.g. RED, ECN...)

❖ Most of the end-to-end congestion control are *“voluntary”*.

❖ During congestion, router and switch mechanisms can *“absolutely”* determine the sharing of resources.

# INTRODUCTION(2/3)

- ❖ The authors using a game-theoretic approach can show that all current router and switch mechanisms either *“encourage”* congestion or *“oblivious”* to it.
- ❖ The authors also show that in the presence of selfish users, all such scheme will inevitably lead to *“congestion collapse”*.



# INTRODUCTION(3/3)

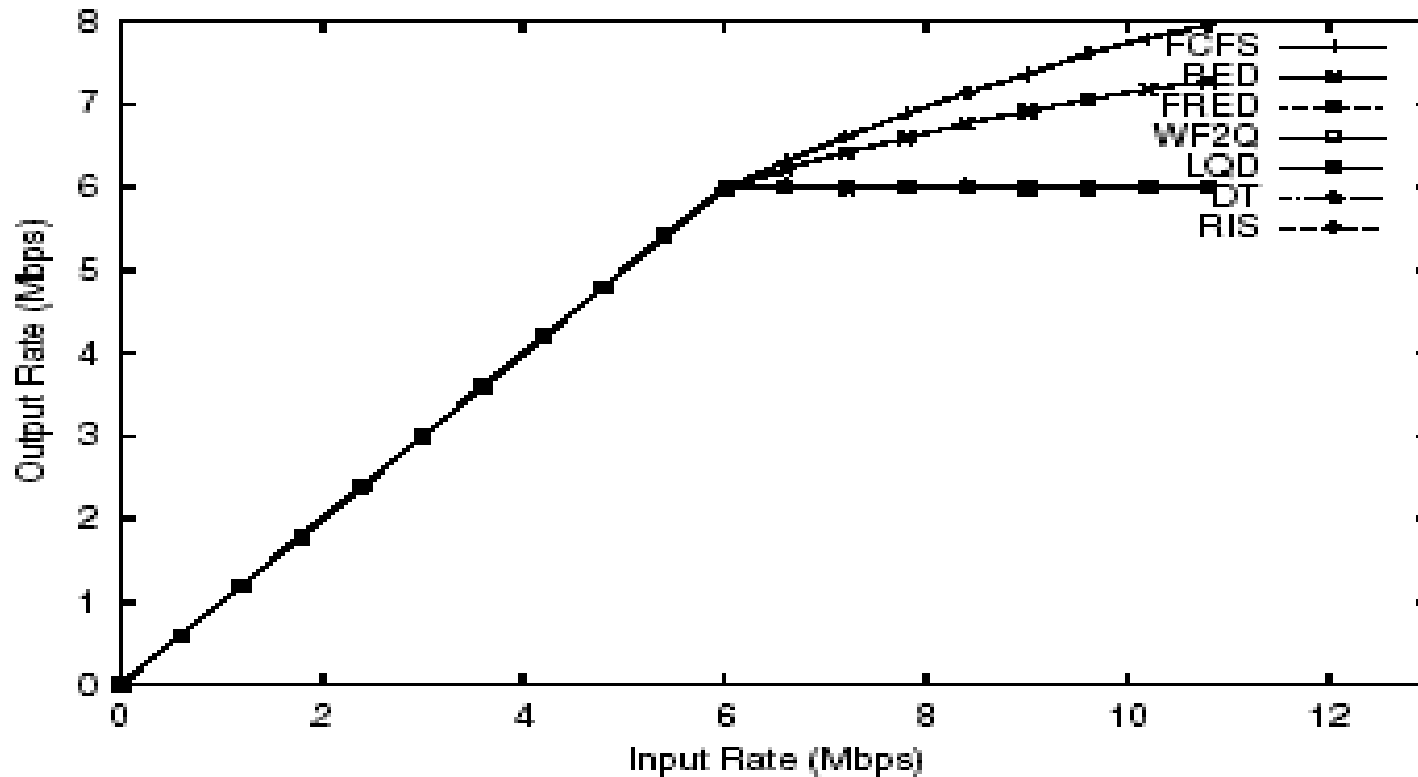
- ❖ The authors of this paper propose a game-theoretic approach towards congestion control which is called “*DWS*” .
- ❖ The crux of the approach is to deploy buffer management policies at switches and routers that “*punish misbehaving flows*” and “*encourage well behaved flows*” .

# CURRENT CONGESTION CONTROL POLICIES(1/4)

Legend	Queuing discipline	Buffer management policy
<b>FIFO</b>	<b>FIFO</b>	<b>Shared buffers, drop tail</b>
<b>DT</b>	<b>FIFO</b>	<b>Dynamic threshold</b>
<b>RED</b>	<b>FIFO</b>	<b>Random early drop</b>
<b>WF2Q</b>	<b>Worst case fair weighted fair queuing</b>	<b>Per flow buffers, drop tail</b>
<b>LQD</b>	<b>FIFO</b>	<b>Longest queue tail drop</b>
<b>FRED</b>	<b>FIFO</b>	<b>Flow RED</b>
<b>RIS</b>	<b>Rate inverse scheduling</b>	<b>Per flow buffers, drop tail</b>

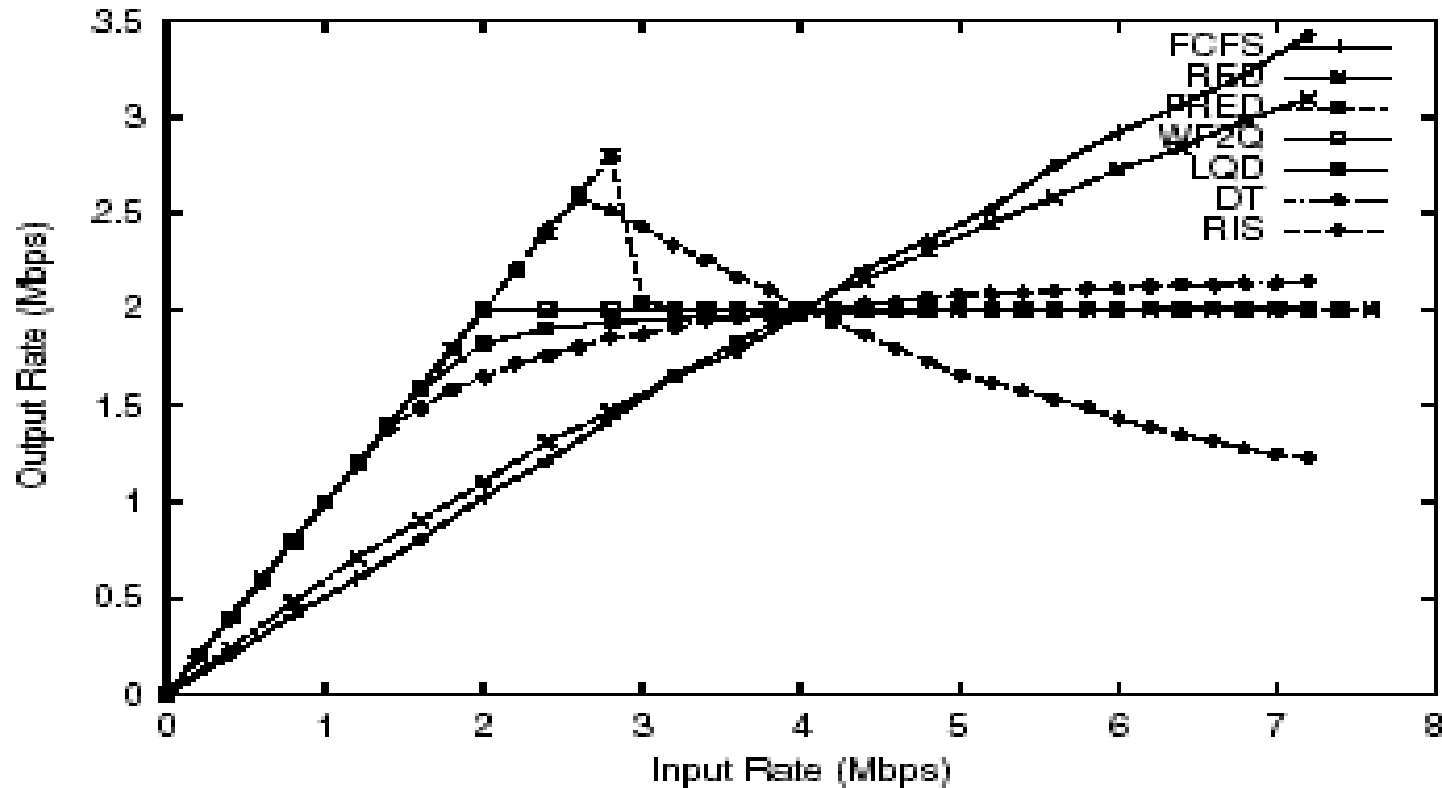
# CURRENT CONGESTION CONTROL POLICIES(2/4)

All other flows sending at 1 Mbps



# CURRENT CONGESTION CONTROL POLICIES(3/4)

All other flows sending at 4 Mbps





# CURRENT CONGESTION CONTROL POLICIES(4/4)

- ❖ For FIFO, RED, and DT resource sharing policies, the only Nash Equilibrium is when the input rates approach infinity. Which can *“encourage congestion causing behavior.”*
- ❖ For WF2Q, LQD, and FRED, where each user’s input rate is more than fair rate constitutes a Nash Equilibrium. Which are *“oblivious to congestion causing behavior.”*

# DWS(DIMISHING WEIGHT SCHEDULERS)

- ❖ It is provided in the Appendix.
- ❖ If a flow is experiencing losses, then decreasing the flow's input rate by a sufficiently small amount will either increase its output rate, or leave it unchanged.
- ❖ If a flow is not experiencing losses, then increasing the flow's input rate by a sufficiently small amount will either increase its output rate, or leave it unchanged.

# PROPERTIES OF DWS

## ❖ Single link :

☞ With DWS scheduling, the fair rate is *“the unique”* Nash Equilibrium for the system.

☞ For DWS scheduling, Nash Equilibrium and Stackelberg Equilibrium *“coincide”*.

## ❖ Arbitrary network of links :

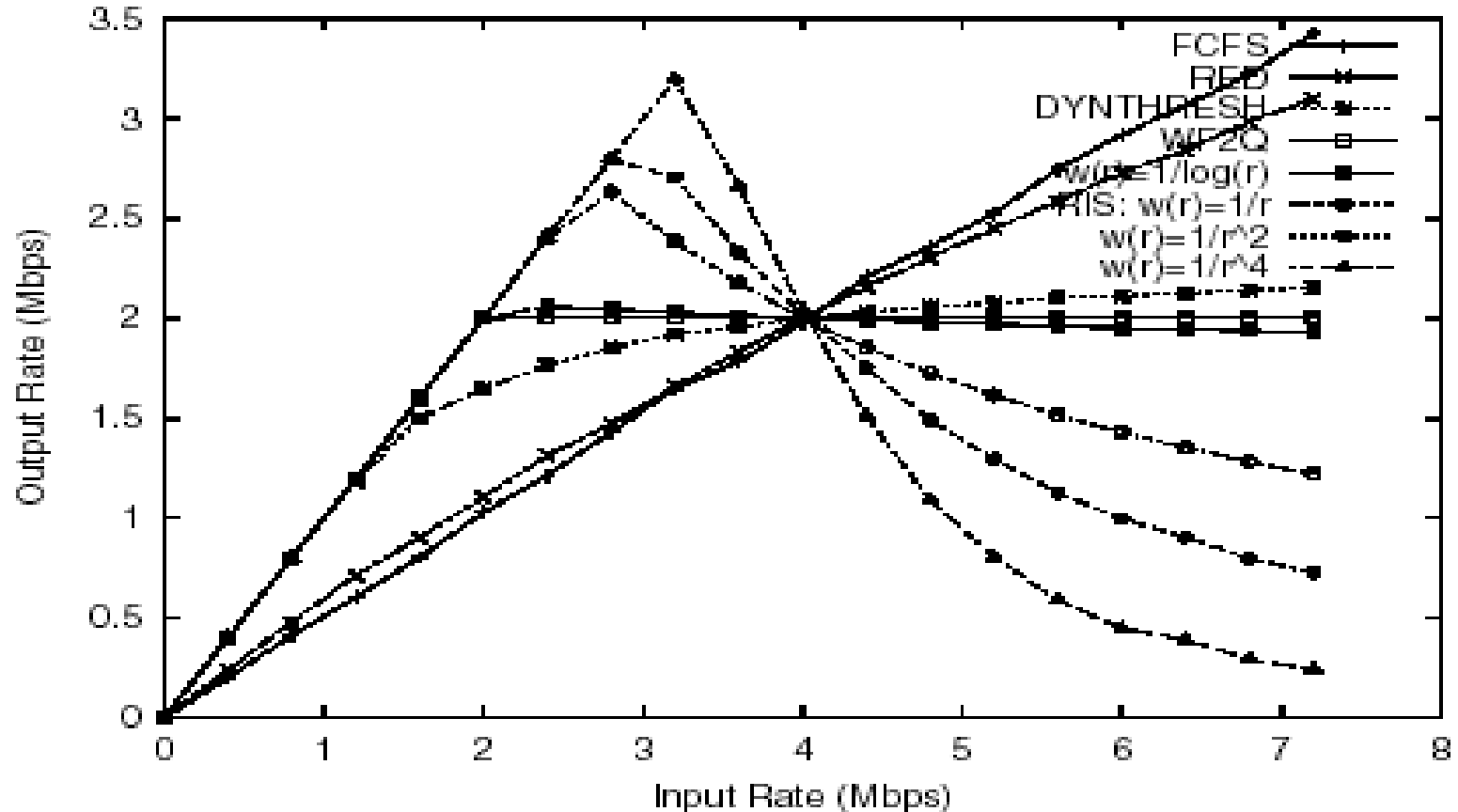
☞ *“The max-min fair rates”* constitute a Nash as well as Stackelberg Equilibrium which there are no losses in the system.

# SIMULATION RESULTS(1/7)

- ❖ The previous section imply that *“the best behavior”* for a user is to send traffic at its *“max-min fair rate”*. However, a user will not know its max-min fair rate.
- ❖ If a link with DWS scheduling is modeled as a game, then TCP-like end user algorithms seem to be reasonable rules to play the game.
- ❖ The authors illustrate that *“TCP indeed converge to their max-min fair rate”*.

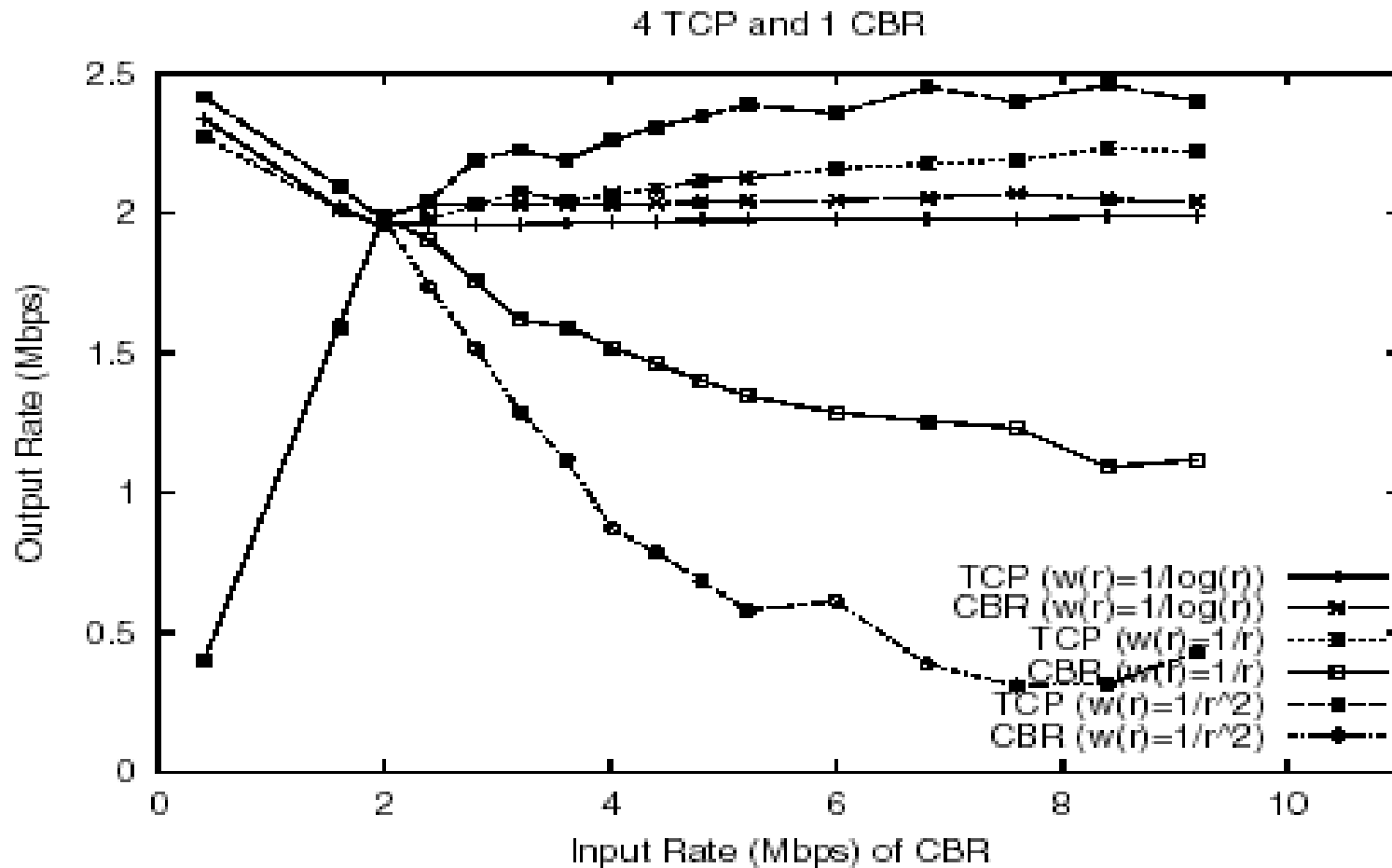
# SIMULATION RESULTS(2/7)

All other flows sending at 4 Mbps

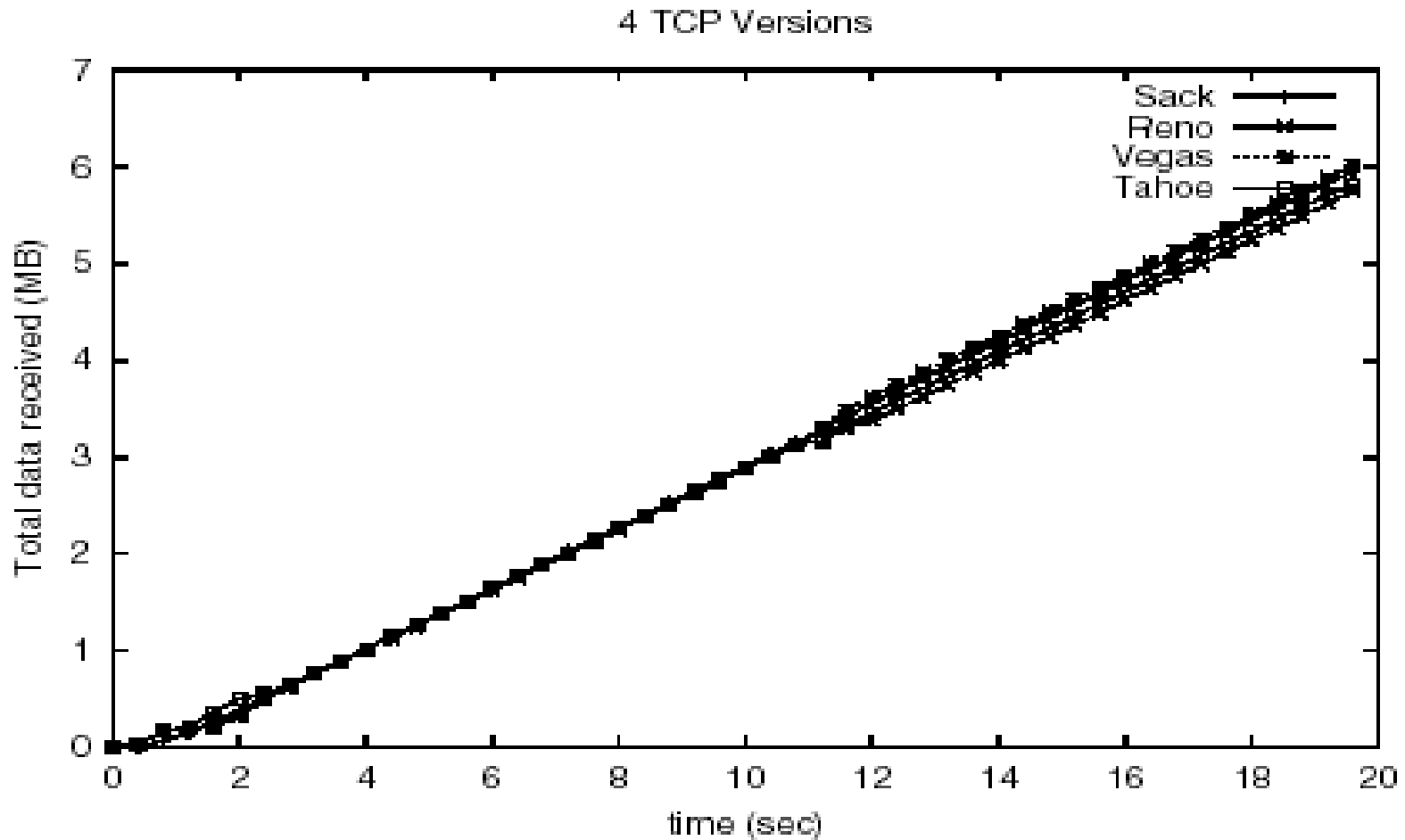




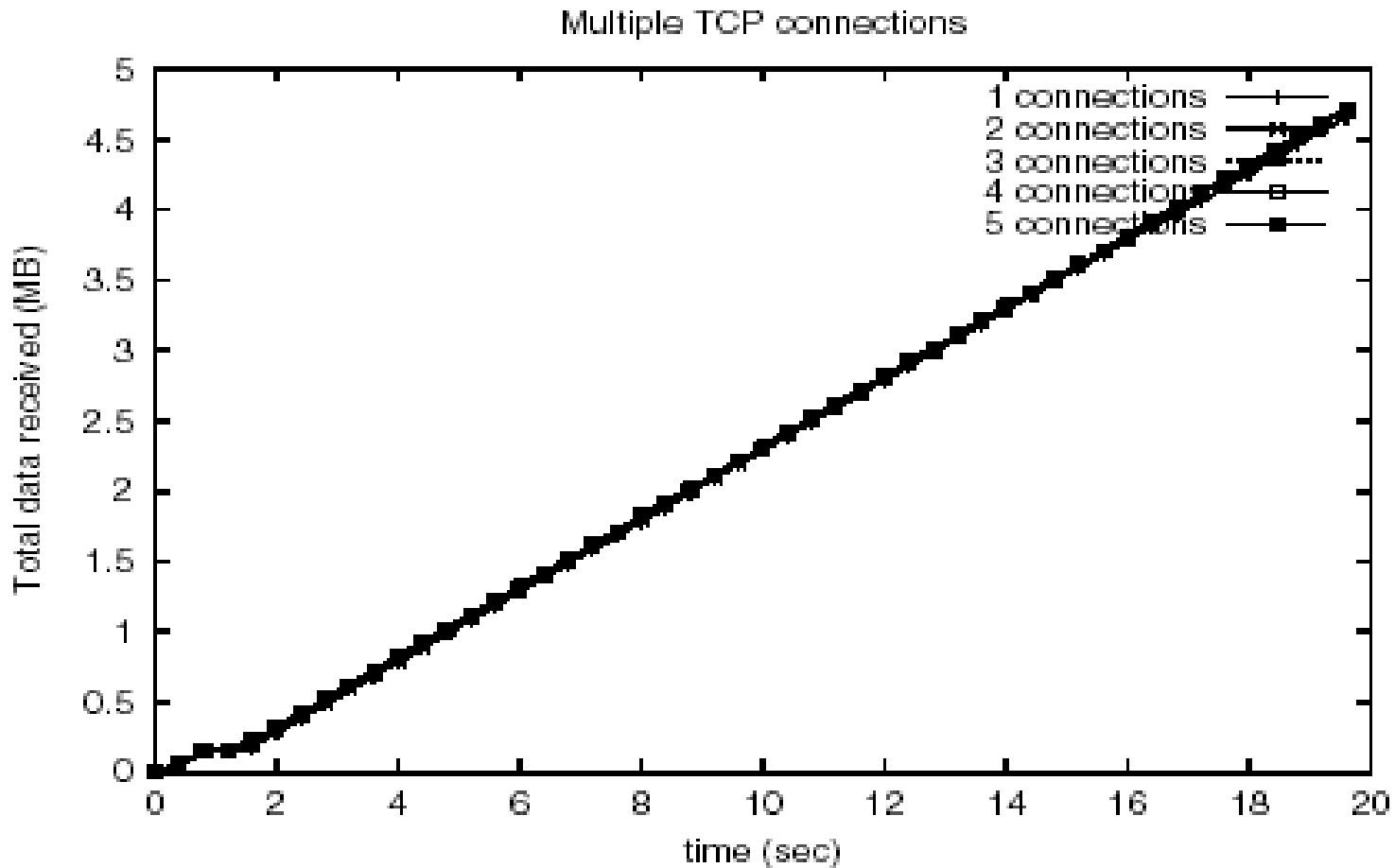
# SIMULATION RESULTS(3/7)



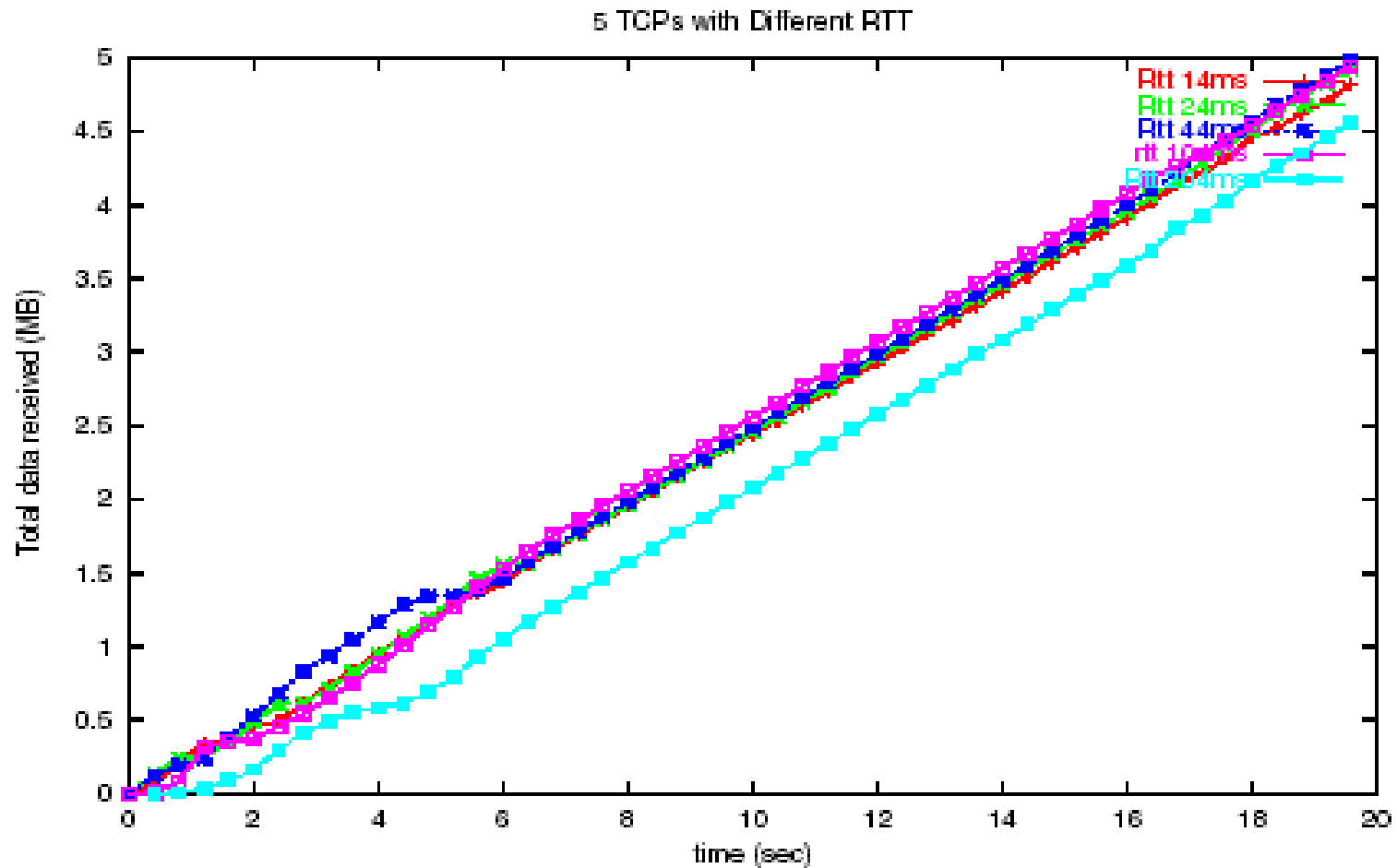
# SIMULATION RESULTS(4/7)



# SIMULATION RESULTS(5/7)

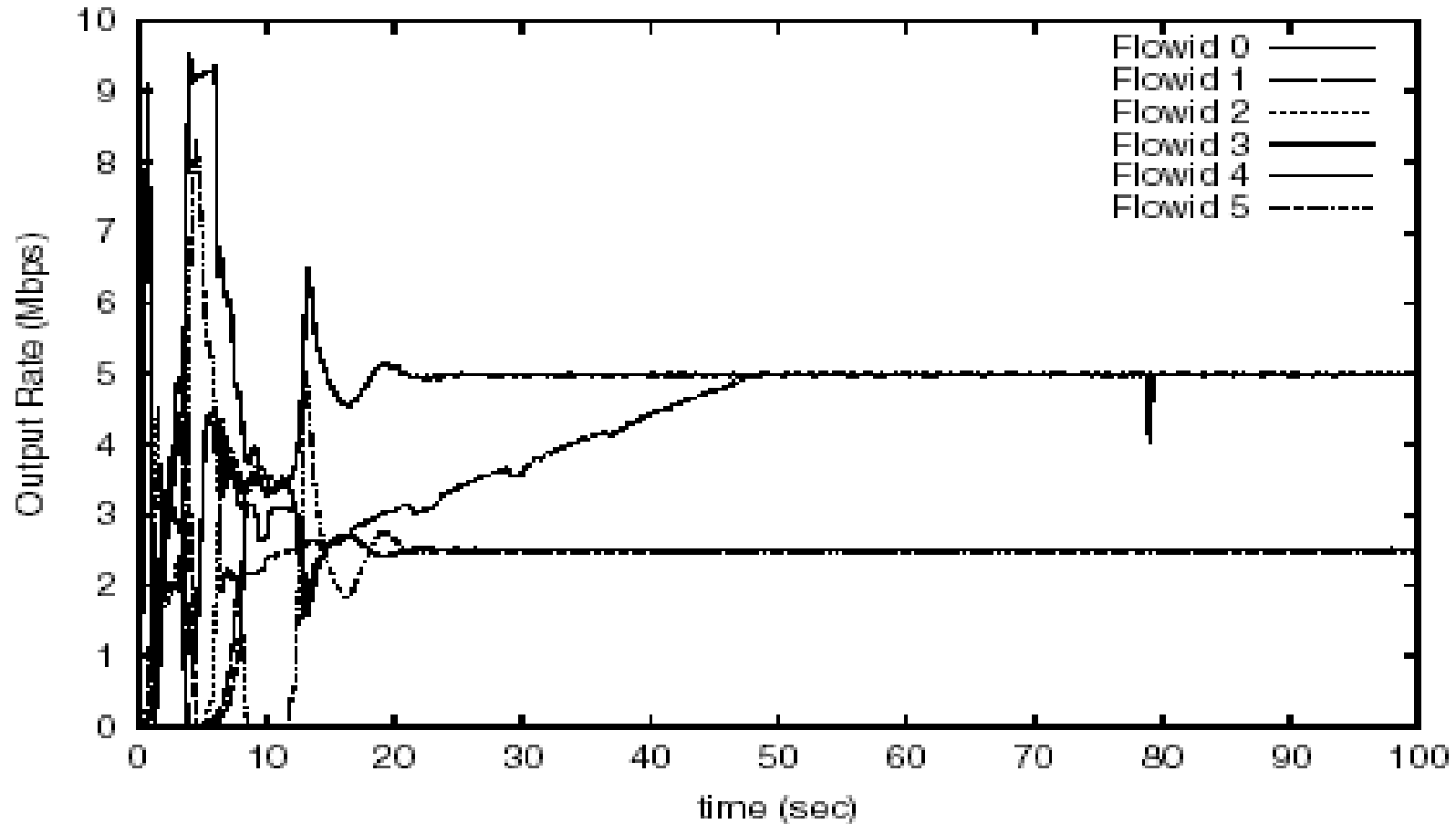


# SIMULATION RESULTS(6/7)



# SIMULATION RESULTS(7/7)

RIS scheduling in a Network





# CONCLUSIONS(1/5)

- ❖ Using the techniques of game theory, the authors showed that “*the current congestion control mechanisms*” in the router or switch either “encourage congestion causing behavior” or “*are oblivious to it*”.
- ❖ The authors proposed a scheduling algorithm by the name “*DWS*” and showed that it “*encourage congestion avoiding behavior*” and “*punish behaviors that lead to congestion*”.

## CONCLUSIONS(2/5)

- ❖ The authors showed that for a single link with DWS scheduling, *“fair rates constitute the unique Nash and Stackelberg Equilibrium”*.
- ❖ They also showed that for an arbitrary network with DWS scheduling at every link, *“the max-min fair rates constitute a Nash as well as Stackelberg Equilibrium”*.

## CONCLUSIONS(3/5)

- ❖ Although the max-min rate constitute Nash and Stackelberg Equilibrium, it is not clear how users can estimate their max-min fair rates.
- ❖ For above, *“a decentralized distributed scheme is required”*. It must be stable and will indeed converge to the max-min fair rates when DWS are deployed in the network.

# CONCLUSIONS(4/5)

- ❖ Using simulations the authors showed that in a network with DWS, most of the TCP variants are able to estimate their max-min rate well, *“irrespective of their versions and RTT”*.
- ❖ They also showed that with DWS, *“the TCP users indeed get rewarded in the presence of unresponsive CBR flows”* which get punished.



# CONCLUSIONS(5/5)

- ❖ The proposed model requires per-flow queuing and scheduling in the core routers, which may not be very easy to implement.
- ❖ However, this paper presents a significantly different view of resource sharing and congestion control on communication networks.