

# Traffic Pattern-based Adaptive Routing in Dragonfly Networks

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# Motivation

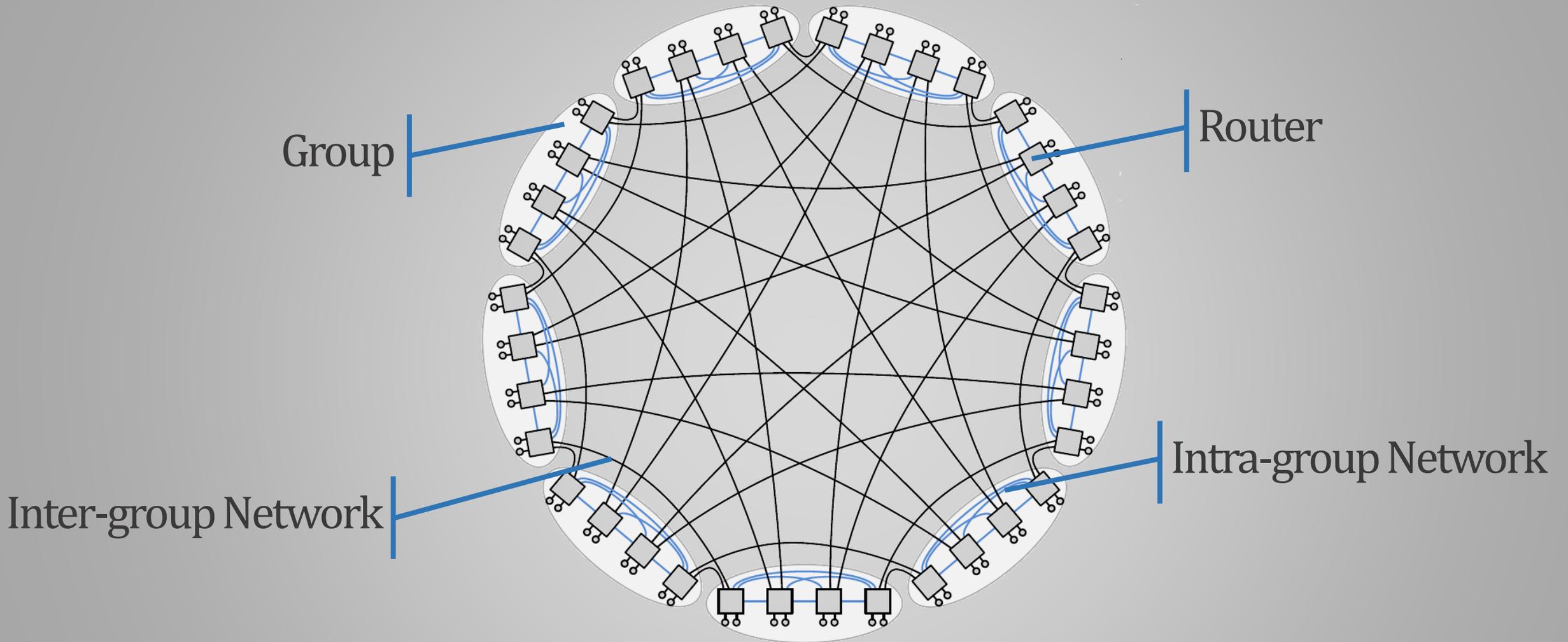
Dragonfly has been known as a potential topology for next generation of HPC systems

Effective routing in Dragonfly depends on the traffic pattern: minimal routing for uniform traffic and non-minimal routing for adversarial traffic

Adaptive routing is required to achieve good performance under various traffic patterns which chooses between minimal and non-minimal paths based on respective queue lengths

We will show that the available methods have some limitations and propose a traffic pattern-based adaptive routing to address these issues

# Dragonfly

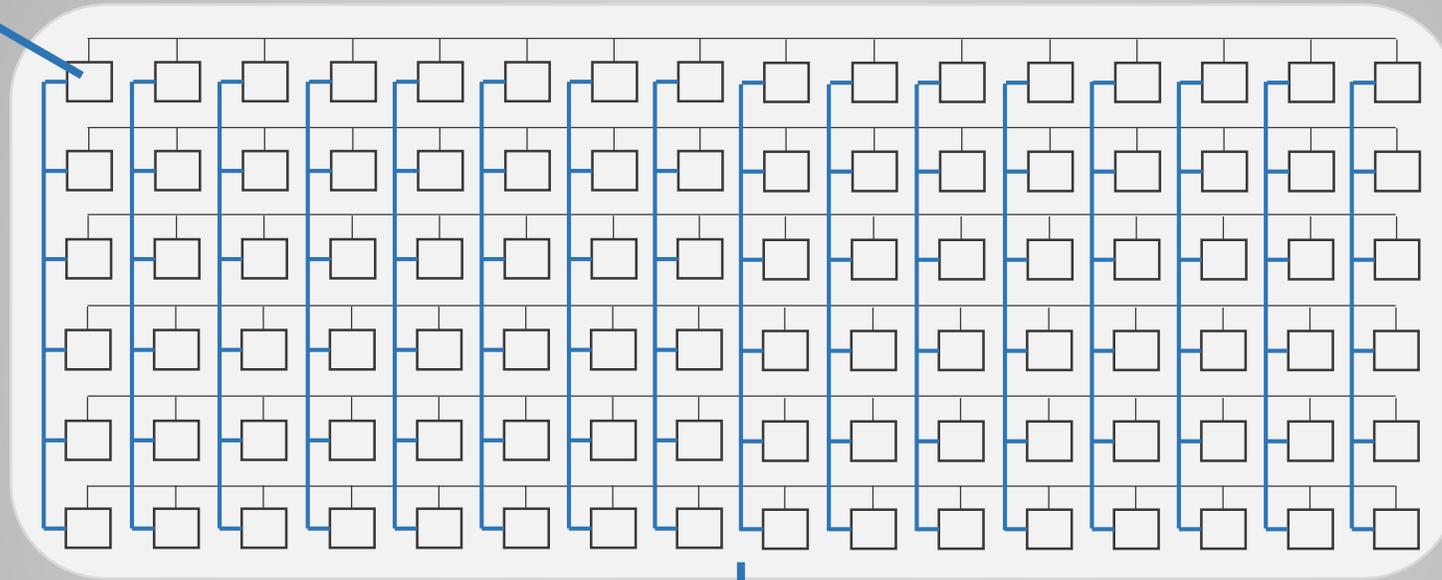


[Garcia et al, INA-OCMC '13]

# Cray Cascade

Radix-48 Aries  
8 processing  
nodes/router

All-to-all inter-group network

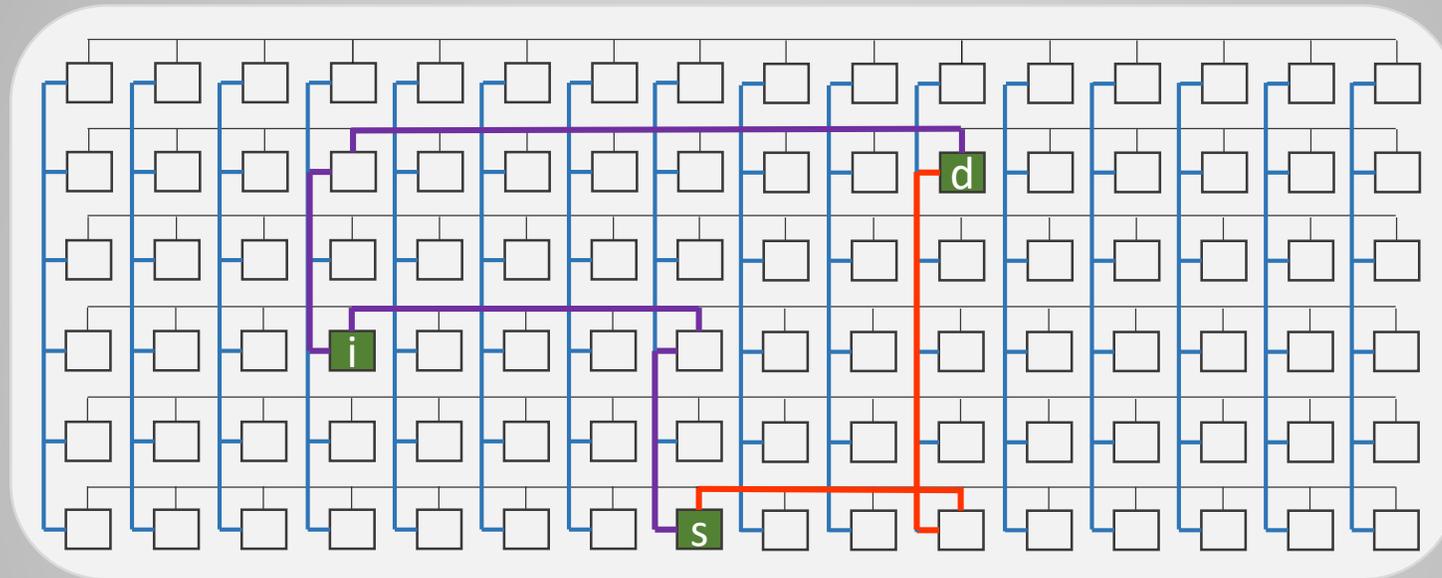


2D HyperX 16x6  
Intra-group network

# Dragonfly Routing

## Basic intra-group routing

— Minimal  
— VLB

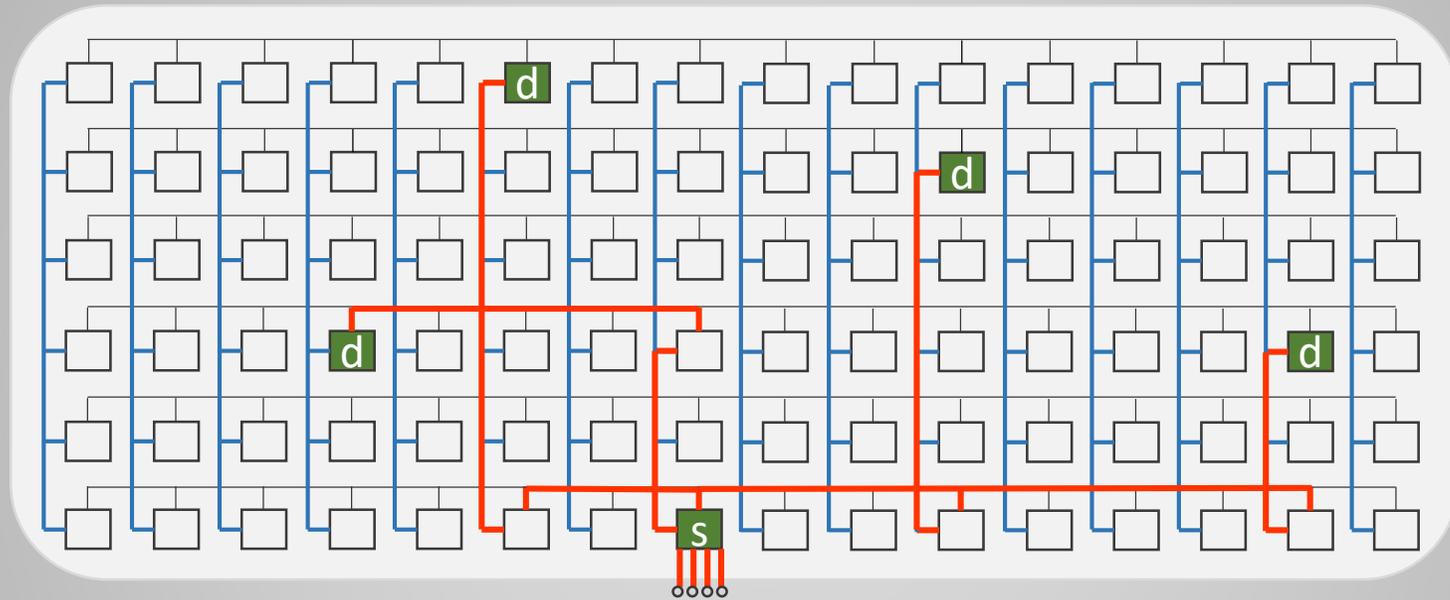


Minimal path length  $\leq 2$  hops

VLB path length  $\leq 4$  hops

# Dragonfly Routing

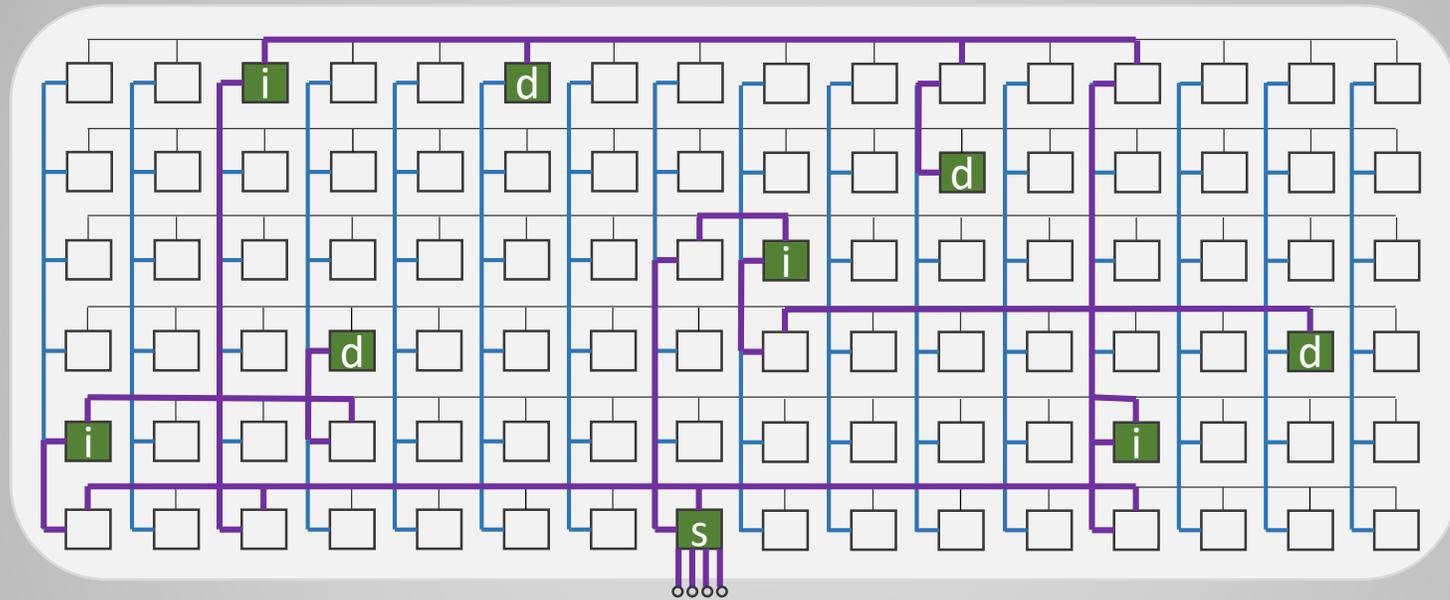
MIN with uniform random traffic



4 packets  
8 links used

# Dragonfly Routing

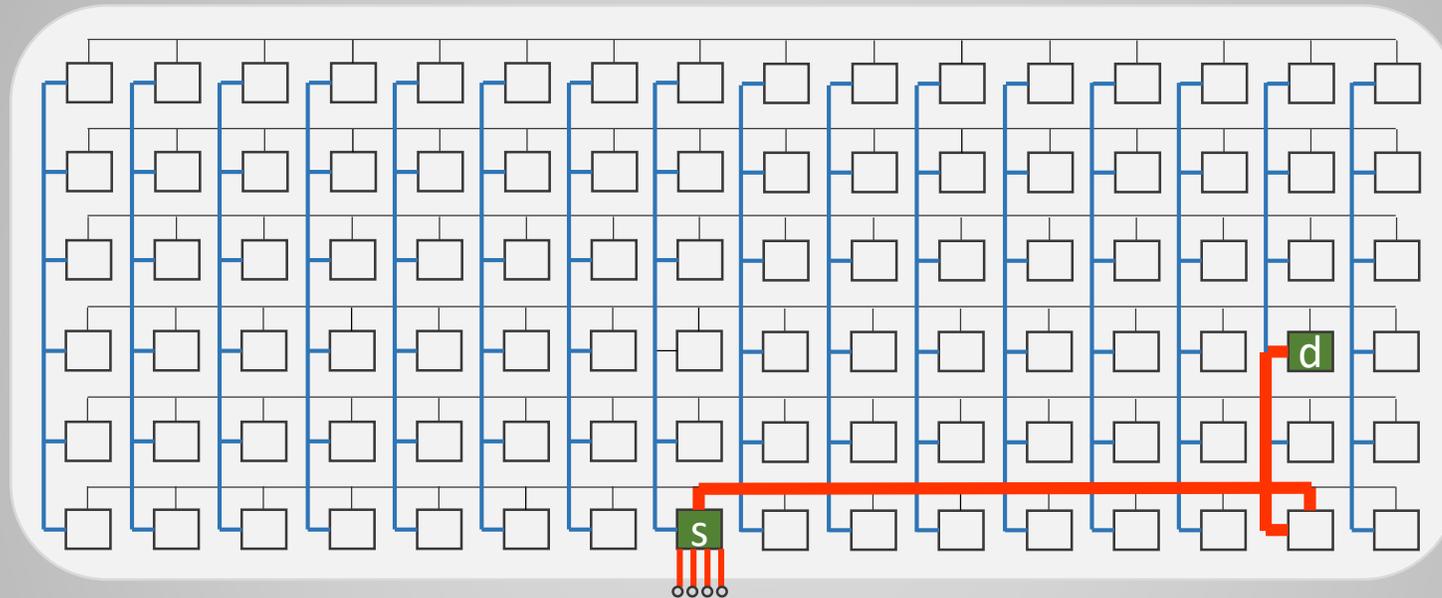
VLB with uniform random traffic



4 packets  
16 links used

# Dragonfly Routing

MIN with adversarial traffic

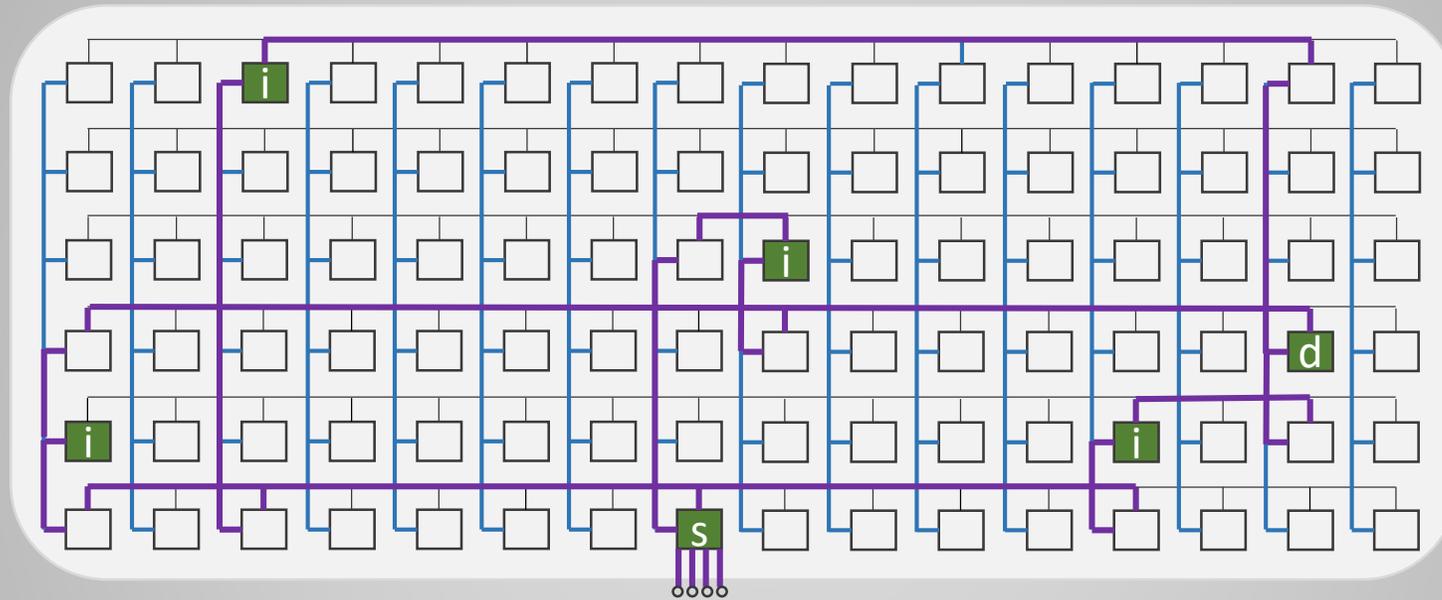


4 packets

2 links used, max bandwidth =  $1/4$

# Dragonfly Routing

VLB with adversarial traffic

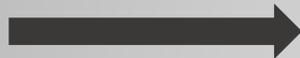
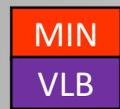


4 packets

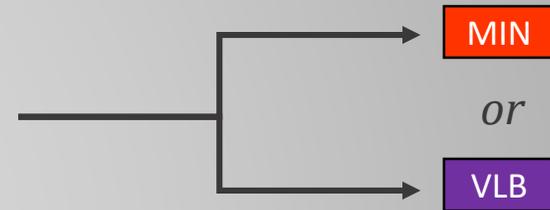
16 links used, max bandwidth = 1

# Dragonfly Routing

Adaptive routing



$$Q_{min} \times H_{min} \leq Q_{vlb} \times H_{vlb} + T$$



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**CHOOSE**  
MIN path  
VLB path

---

**SELECT**  
least loaded  
path

---

**FORWARD**

# Dragonfly Routing

## Adaptive routing

$$Q_{min} \times H_{min} \leq Q_{vlb} \times H_{vlb} + T$$

T

Bias towards selecting MIN path

T

Bias towards selecting VLB path

*“T is used to balance the performance under uniform random and worst case traffic patterns”*

[Jiang et al, ISCA'09]

*“Value of T needs to be determined empirically”*

[Jiang et al, ISCA'09]

... so the performance of UGAL  
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**But how do we identify the traffic pattern???**

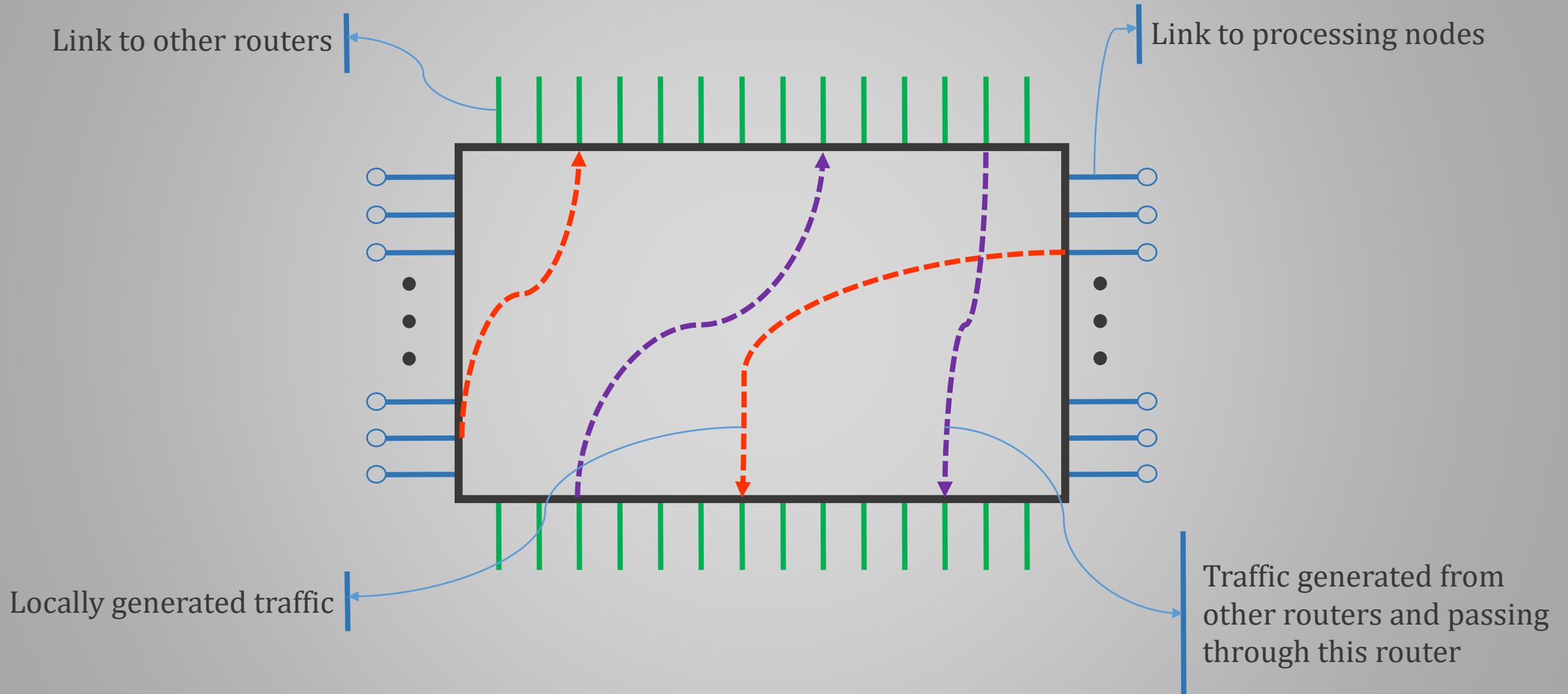
# Why to Identify Traffic Pattern

Minimal routing works best under load balanced or uniform random traffic

Non-minimal routing is desirable when adversarial traffic is observed

By identifying these traffic patterns, we can decrease the number of false routing decisions made by the adaptive routing scheme

# Observed Traffic at Each Router





# Quantifying Traffic Pattern

Local traffic pattern can be quantified using *localimpact*

$$\text{Localimpact} = \text{DestC}_i/h$$

<i>Injection Rate</i>	<i>Pattern</i>	<i>DestC<sub>i</sub></i>	<i>Localimpact</i>
0.1	UR	1	0.02
	ADV	90	1.80
0.44	UR	4	0.08
	ADV	396	7.92
0.9	UR	8	0.16
	ADV	∞	∞



<i>Localimpact &lt; low<sub>l</sub></i>	<i>Benign</i>
<i>Localimpact &gt; high<sub>l</sub></i>	<i>Adversarial</i>
<i>otherwise</i>	<i>Mixed</i>

# Quantifying Traffic Pattern

Globally generated traffic

$Port\_thr_j$



*Uniform Random,  $h = 50$ , injection rate = 0.4*

$Port\_thr_j$



*Adversarial,  $h = 50$ , injection rate = 0.4*

Count the number of packets generated from other routers and passing through each port over the past  $h$  cycles

# Quantifying Traffic Pattern

Global traffic pattern can be quantified using *globalimpact*

$$\text{Globalimpact} = \text{Port\_thr}_j/h$$

<i>Injection Rate</i>	<i>Pattern</i>	<i>Port_thr<sub>j</sub></i>	<i>Globalimpact</i>
0.1	UR	2.24	0.04
	ADV	5.45	0.11
0.44	UR	9.86	0.20
	ADV	33.7	0.67
0.9	UR	20.5	0.41
	ADV	∞	∞



<i>Globalimpact &lt; low<sub>g</sub></i>	<i>Benign</i>
<i>Globalimpact &gt; high<sub>g</sub></i>	<i>Adversarial</i>
<i>otherwise</i>	<i>Mixed</i>

# Traffic Pattern Based Adaptive Routing

Based on *localimpact* and *globalimpact*, our routing scheme operates in nine operating regions

		<i>globalimpact</i>		
		<i>benign</i>	<i>mixed</i>	<i>adversarial</i>
<i>localimpact</i>	<i>benign</i>	<i>benign</i>	<i>mixed</i>	<i>adversarial</i>
	<i>mixed</i>	<i>benign</i>	<i>mixed</i>	<i>adversarial</i>
	<i>adversarial</i>	<i>benign</i>	<i>mixed</i>	<i>adversarial</i>

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**We can tailor T values to  
each operating region**

# Traffic Pattern Based Adaptive Routing

UGAL(T) | Larger T  $\rightarrow$  prefer minimal path  
Smaller T  $\rightarrow$  prefer non-minimal path

		<i>globalimpact</i>		
		<i>benign</i>	<i>mixed</i>	<i>adversarial</i>
<i>localimpact</i>	<i>benign</i>	<i>MIN/UGAL(64)</i>	<i>MIN/UGAL(64)</i>	<i>UGAL(48)</i>
	<i>mixed</i>	<i>UGAL(-4)</i>	<i>UGAL(-20)</i>	<i>UGAL(-40)</i>
	<i>adversarial</i>	<i>UGAL(-48)</i>	<i>UGAL(-64)</i>	<i>VLB</i>

| By observing higher local and global impact, routing moves toward using non-minimal paths to avoid congestion

# Evaluation Methodology

## NETWORK

1 Group of a Cray Cascade machine  
16x6 2D HyperX, a=96, p=18

## SIMULATION

Booksim, 4 VCs, VC buffer size = 32  
Single flit packets

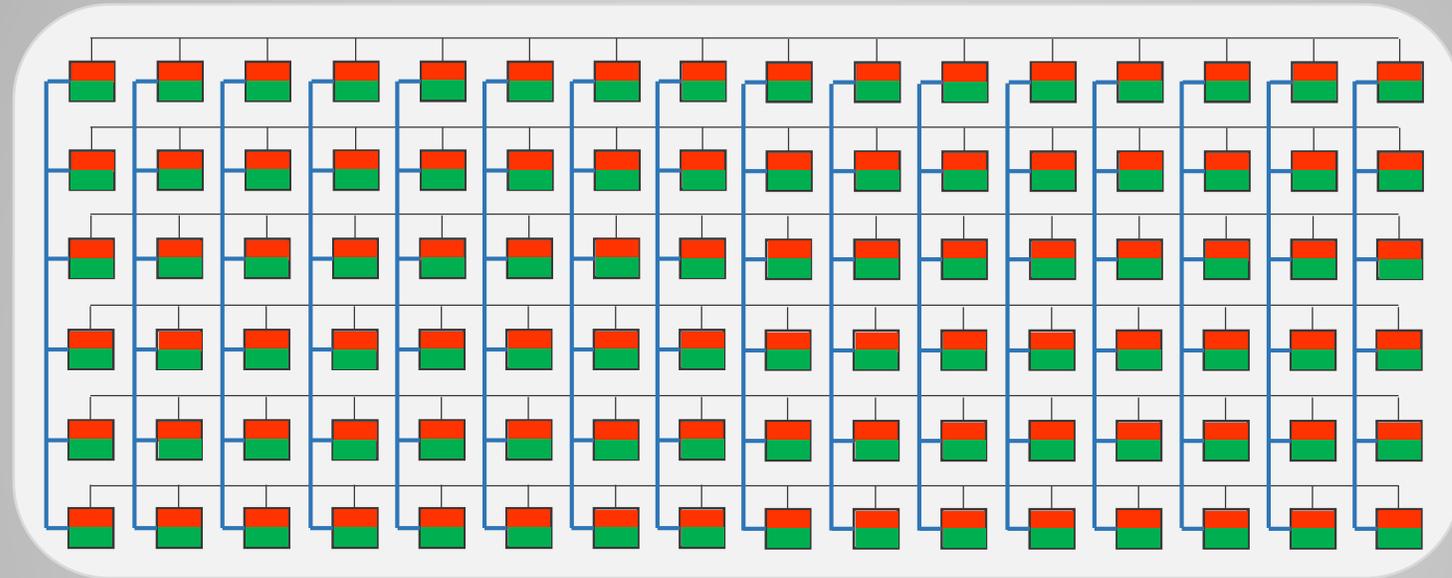
## ROUTING

MIN, VLB, UGAL-L, UGAL-G, TPR

## TRAFFIC

Uniform Random, Shift<sub>1</sub>, NLC\_URADV, RLC\_URADV  
Only intra-group communication

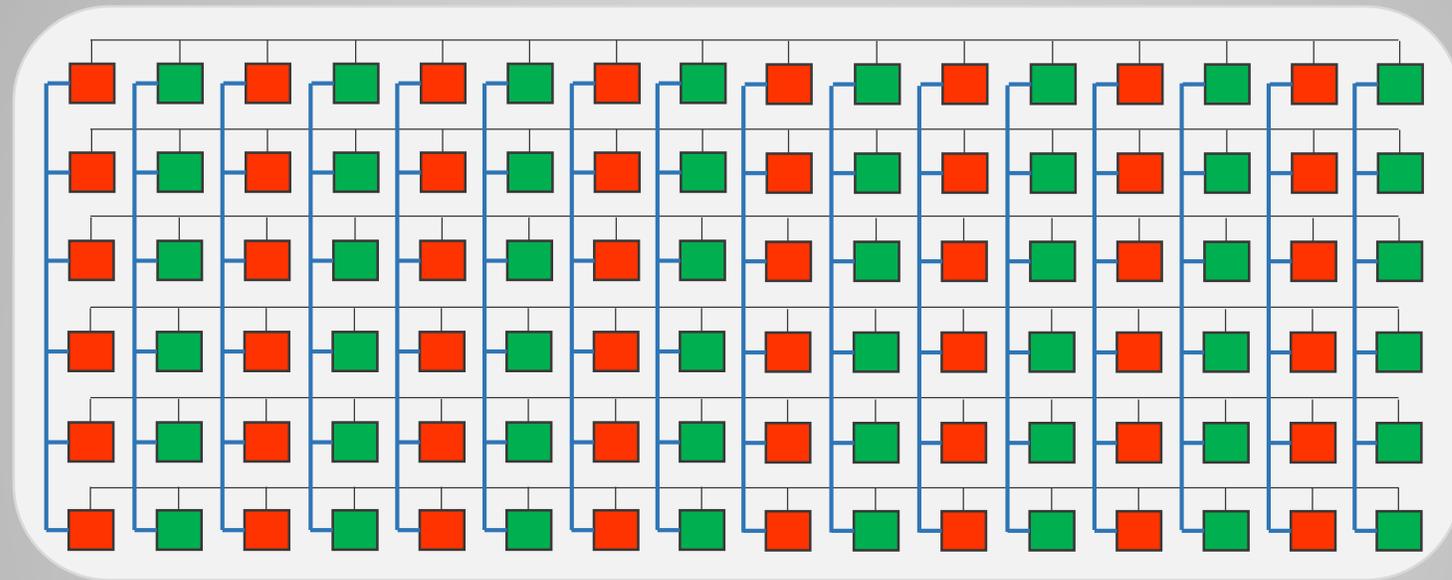
# Node-Level Combined Traffic



UR  
Shift

NLC\_URADV(50,50)

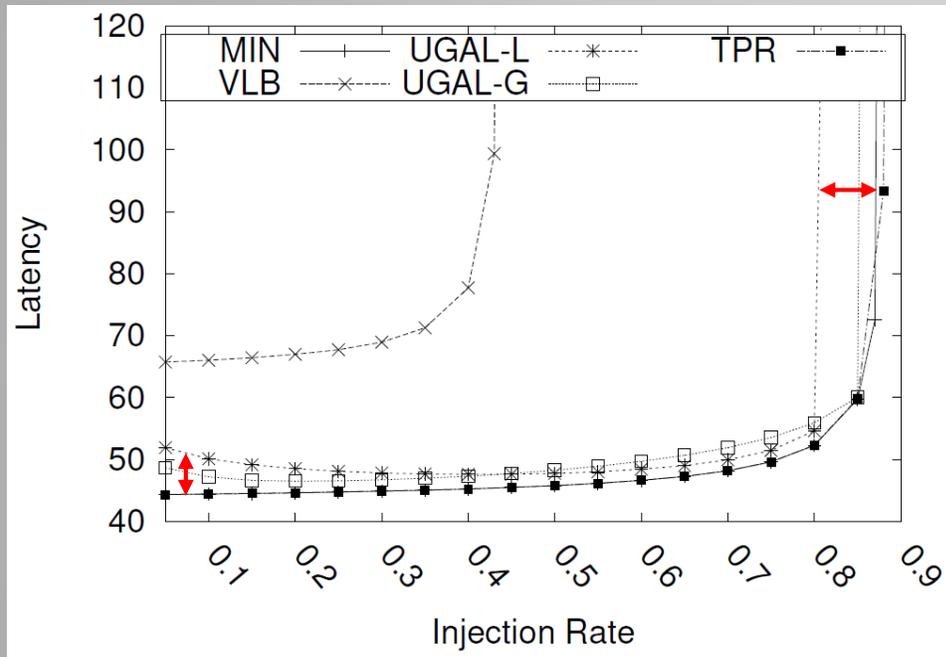
# Router-Level Combined Traffic



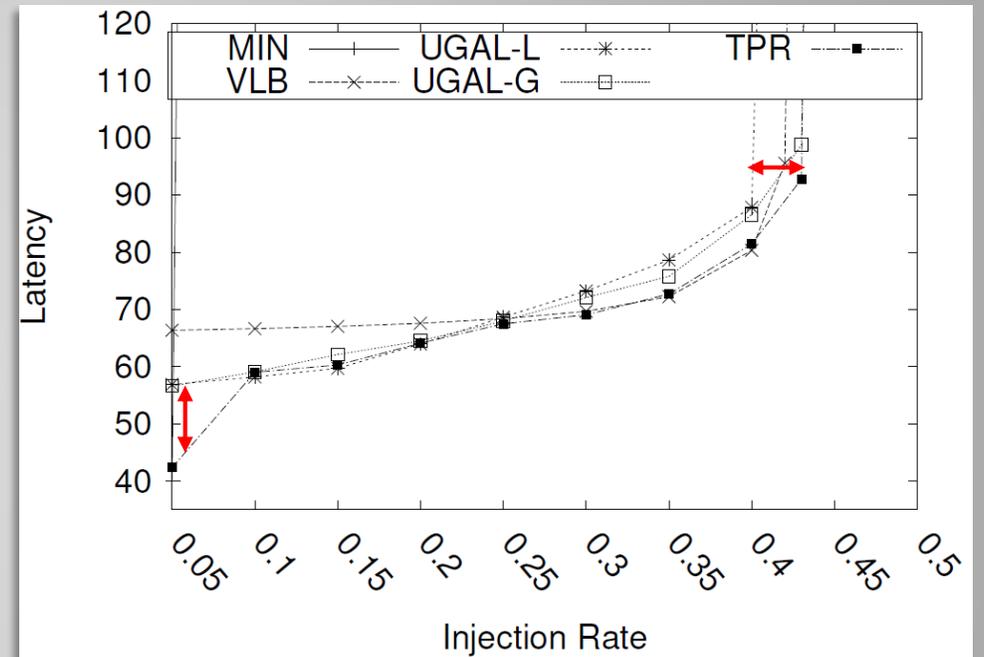
UR  
Shift

RLC\_URADV(50,50)

# Evaluation Results

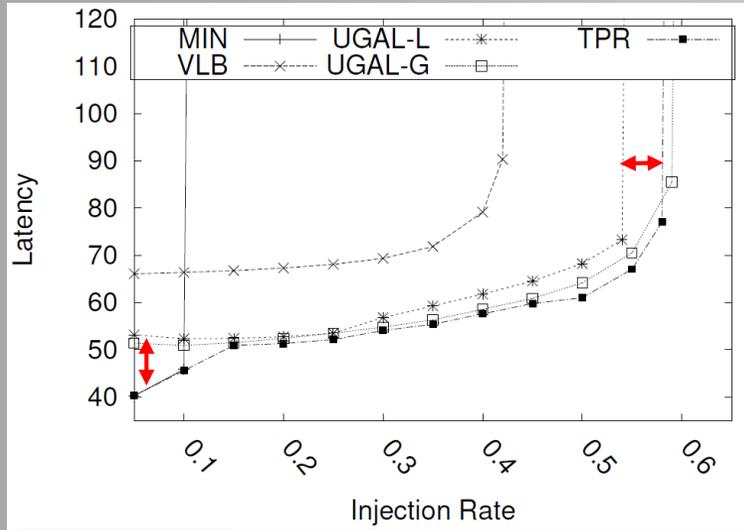


Uniform Random

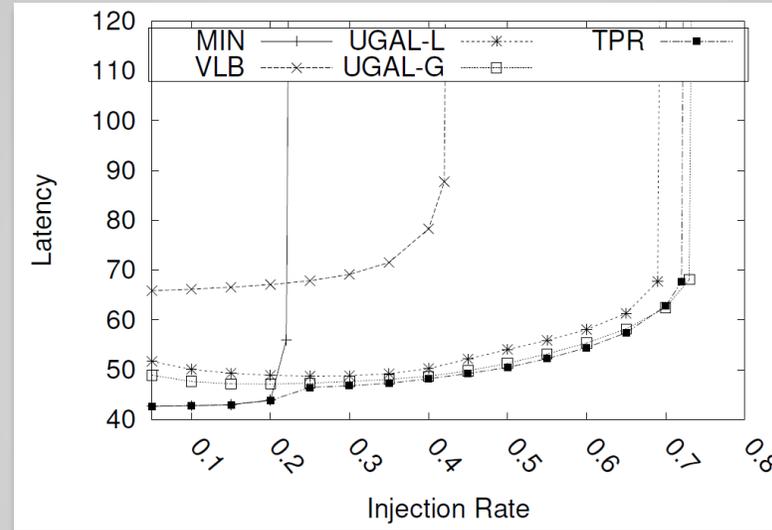


Shift<sub>1</sub>

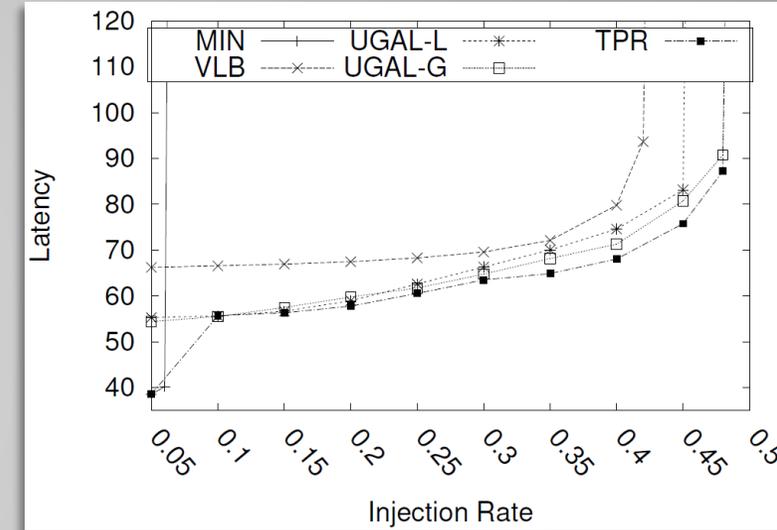
# Evaluation Results



NLC\_URADV  
(50,50)

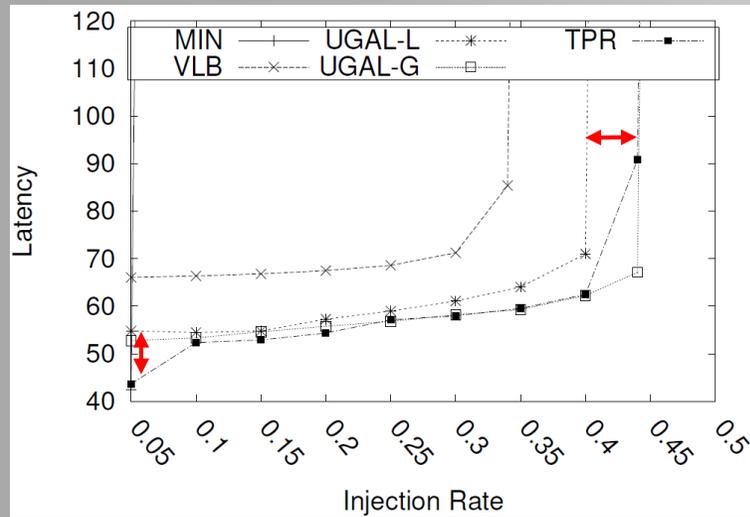


NLC\_URADV  
(80,20)

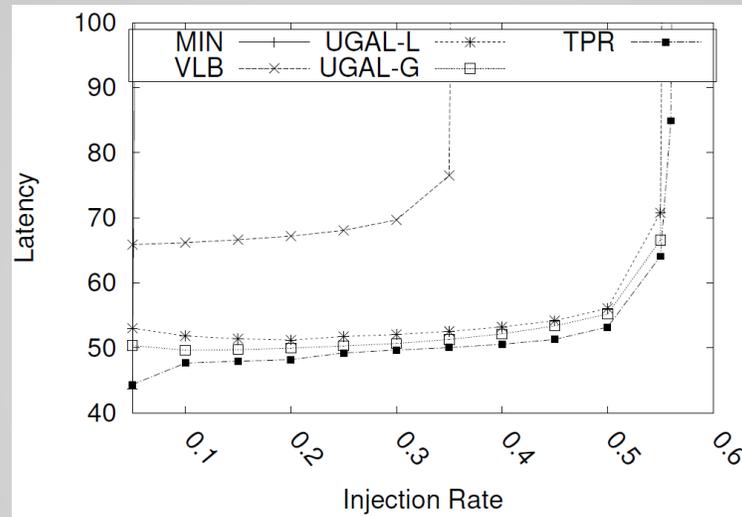


NLC\_URADV  
(20,80)

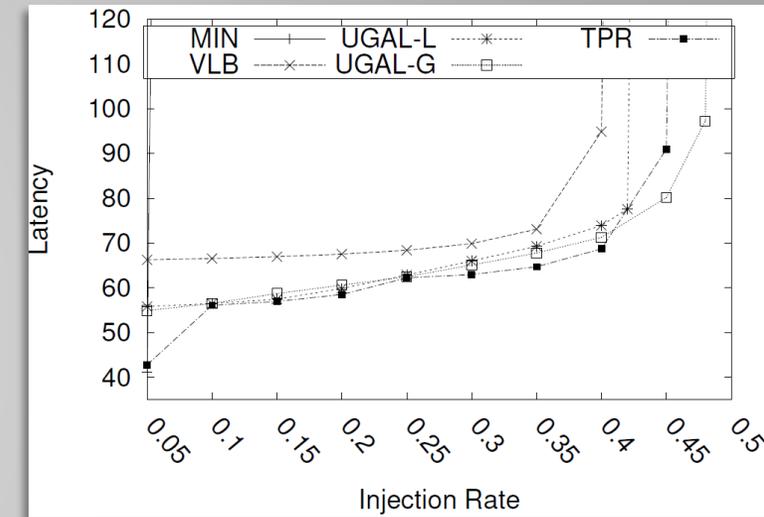
# Evaluation Results



RLC\_URADV  
(50,50)



RLC\_URADV  
(80,20)



RLC\_URADV  
(20,80)

# Conclusion

By identifying local and global traffic conditions, TPR achieves the best latency results among all evaluated routing schemes

TPR improves the throughput performance of UGAL-L for almost every traffic pattern considered in this study

The same proposed method, can improve the performance of other similar routing schemes including Piggyback, Reservation and Progressive adaptive routing



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