An experimental study of the learnability of congestion control

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- Formulate a mental model of the target network
- Decide on the protocol's goal
- Design a protocol to achieve this goal
- Can either be implicit or explicit

But, the model is always wrong!

- Lost throughput due to stochastic loss
- Bufferbloat when queues are incorrectly sized
- Incast in datacenters



Our work

Can we formalize this design process ...

Our work

- Can we formalize this design process ...
- and quantify the consequences of model mismatch?

Contributions

- Formalize learnability in the context of congestion control
- Use it to answer:
 - Do we need to know the link speed exactly?
 - What is the cost of backwards compatibility?
 - Do we need to know the topology exactly?

Approach

- Specify a training scenario.
 - Topology
 - Locations of senders and receiver
 - Application workload
 - Buffer size and queuing discipline
- Specify an objective function.
- Synthesize protocol automatically.
- Evaluate on a testing scenario inside ns-2

Automated protocol synthesis

- Find best protocol, given an imperfect network model
- ▶ The problem is hard to solve in general
- Rely on Remy¹ to produce congestion-control protocols.

¹KW and Hari Balakrishnan, TCP ex Machina: Computer-Generated Congestion Control, SIGCOMM 2013 4 日) 4 周) 4 達) 4 達)

Caveats and non-goals

- Very simple, controlled experiments
- Results could change with better protocol-design tools
- Not trying to understand Remy's internals

Training scenario:

Link speed 32 Mbits/sec

Minimum RTT 150 ms

Topology Dumbbell

Number of senders 2

Workload 1 sec ON/OFF times

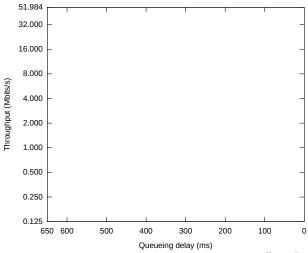
Buffer size 5 BDP

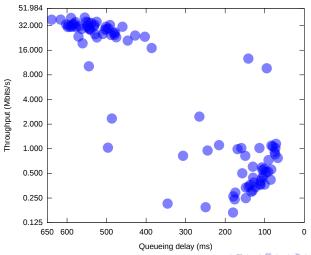
Objective function $\sum \log(\text{throughput}) - \log(\text{delay})$

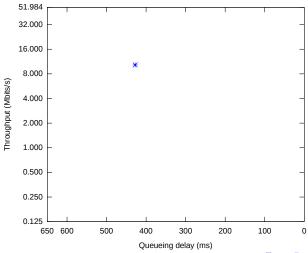
Testing scenario identical to training scenario

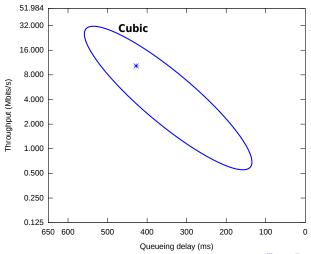


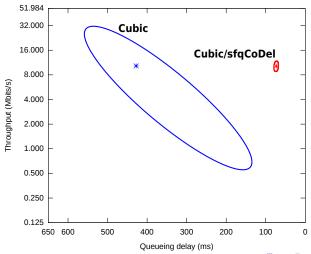
- A hypothetical centralized scheme (CEN)
 - ightharpoonup Every time a sender turns ON/OFF, solve $\sum \log$ (throughput)
 - Set each sender's rate using obtained solution
 - Zero queuing delay

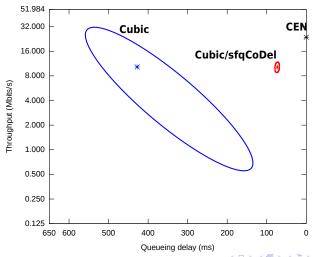


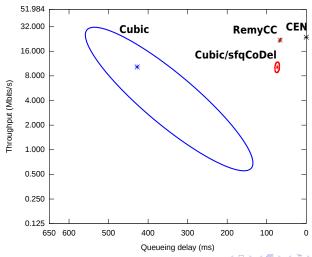










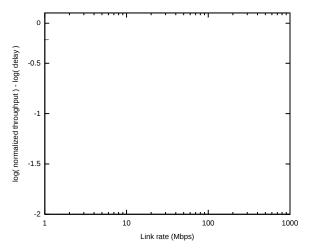


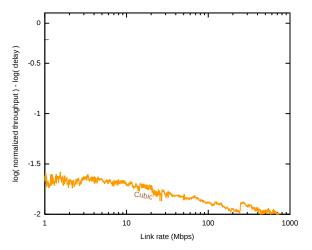
RemyCC	Link rates	RTT	Senders	ON/OFF time	Topology
1000×	1-1000 Mbps	150 ms	2	1 sec	Dumbbell
100×	3.2-320 Mbps	150 ms	2	1 sec	Dumbbell
10×	10-100 Mbps	150 ms	2	1 sec	Dumbbell
2x	22-44 Mbps	150 ms	2	1 sec	Dumbbell

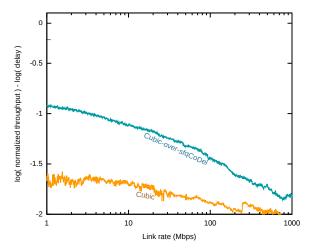
Table: Training scenarios for forwards-compatibility experiment

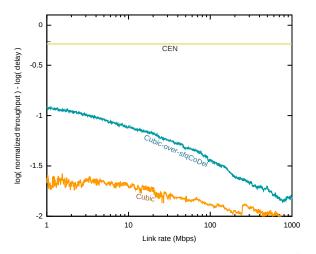
Link rates	RTT	Senders	ON/OFF time	Topology
1–1000 Mbps	150 ms	2	1 sec	Dumbbell

Table: Testing scenarios for forwards-compatibility experiment

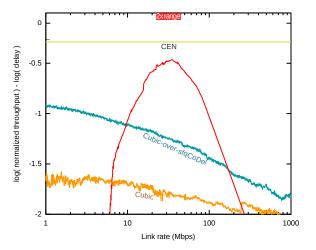


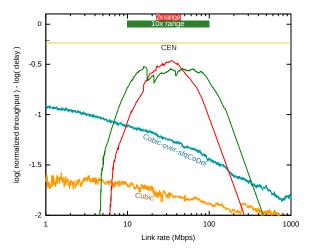


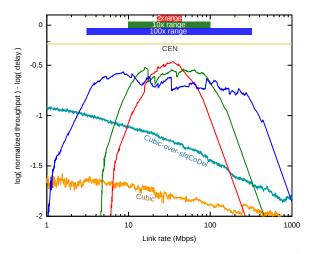




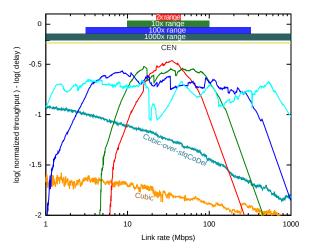












Can we design a RemyCC that is TCP aware?

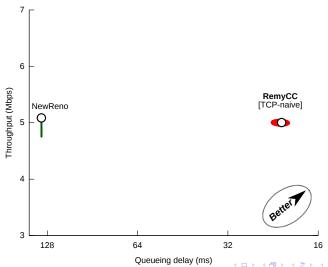
RemyCC	Link rates	RTT	Senders	ON/OFF time	Topology
TCP-aware	9–11 Mbps	100 ms	2 Remy	5 sec ON/OFF	Dumbbell
			1 Remy, 1 AIMD	5 sec ON, 10 ms OFF	
TCP-naive	9–11 Mbps	100 ms	2 Remy	5 sec ON/OFF	Dumbbell
				5 sec ON, 10 ms OFF	Dumbben

Table: Training scenarios

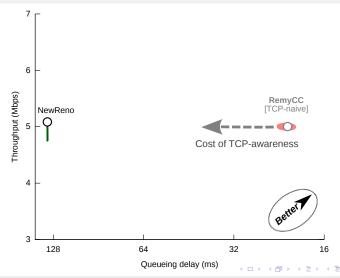
Link rates	RTT	Senders	ON/OFF time	Topology
10 Mbps	100 ms	2 TCP-aware	5 sec ON, 10 ms OFF	Dumbbell
10 Mbps	100 ms	2 TCP-naive	5 sec ON, 10 ms OFF	Dumbbell
10 Mbps	100 ms	TCP-aware, AIMD	5 sec ON, 10 ms OFF	Dumbbell
10 Mbps	100 ms	TCP-naive, AIMD	5 sec ON, 10 ms OFF	Dumbbell
10 Mbps	100 ms	2 AIMD	5 sec ON, 10 ms OFF	Dumbbell

Table: Testing scenarios

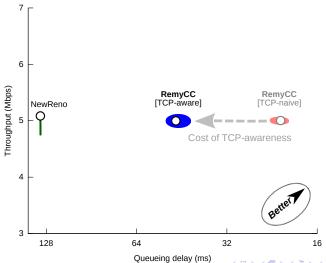
RemyCC competing against itself



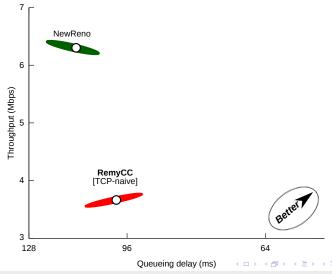
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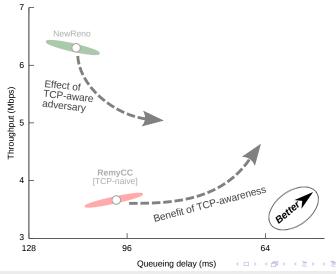
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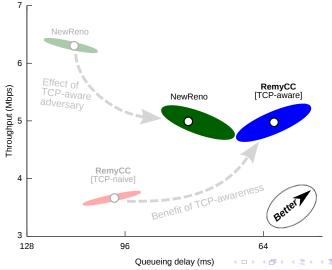
RemyCC competing against TCP NewReno



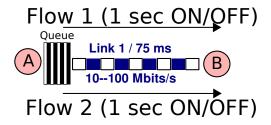
RemyCC competing against TCP NewReno



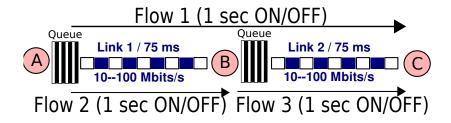
RemyCC competing against TCP NewReno

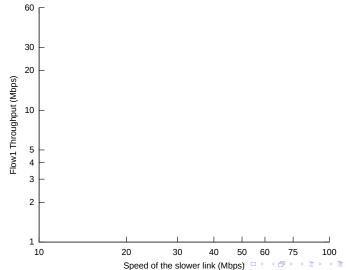


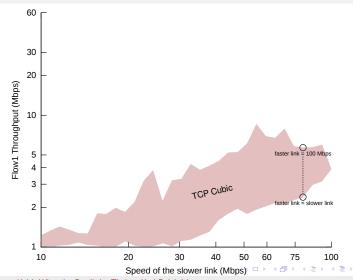
One bottleneck

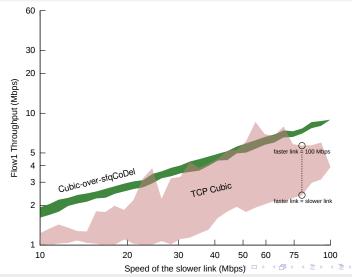


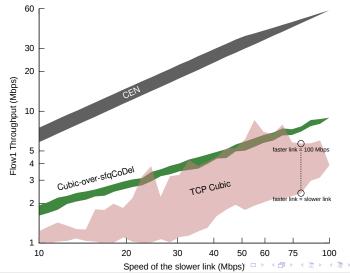
Two bottlenecks

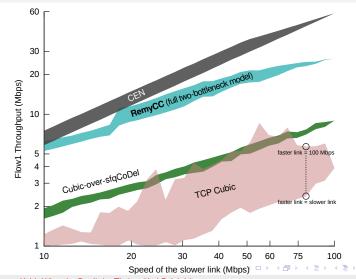


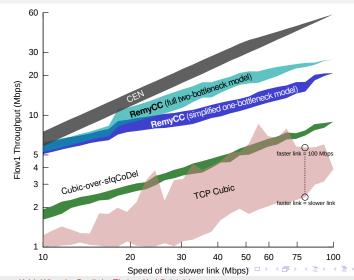












Related Work

- Probably approximately correct learning
- Transfer learning
- Machine-generated congestion control

Limitations and future work

- Generalizability to more complex topologies?
- Better characterization of gap from optimal
- Do results change if we learn in-network behavior as well?
- Model mismatches between simulation and the real world

Backup slides

Can applications with different objectives coexist?

Tpt. Sender: A throughput-intensive sender

$$log(throughput) - 0.1 * log(delay)$$
 (1)

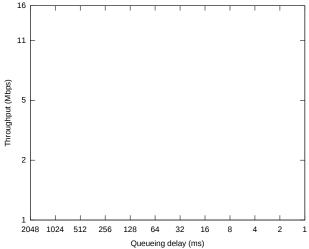
Lat. Sender: A latency-sensitive sender

$$log(throughput) - 10.0 * log(delay)$$
 (2)

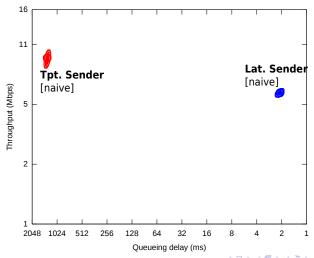
Running over a FIFO queue



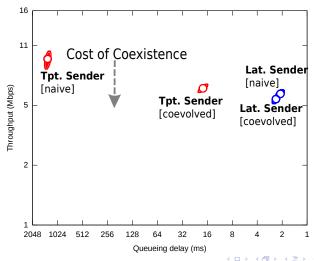
Training for diversity has a cost ...



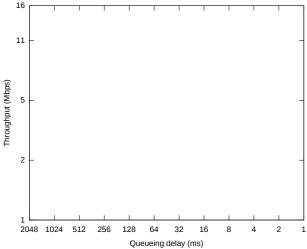
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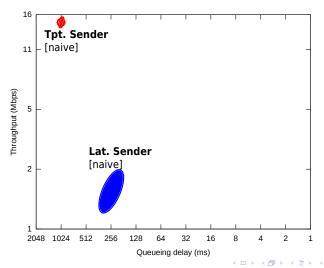
Training for diversity has a cost ...



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