

Stochastic forecasts achieve high throughput and low delay over cellular networks

Winstein, K., Sivaraman, A., and Balakrishnan, H.,
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Outlines

- Introduction
- Sprout Algorithm
- Experimental Testbed
- Evaluation
- Discussion

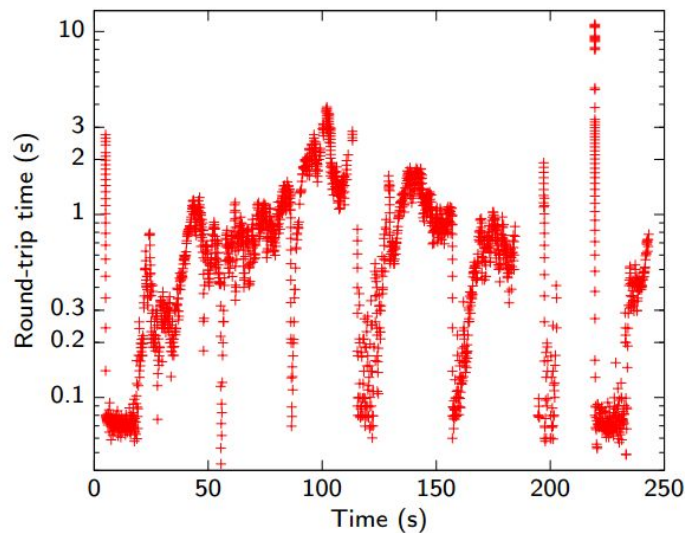
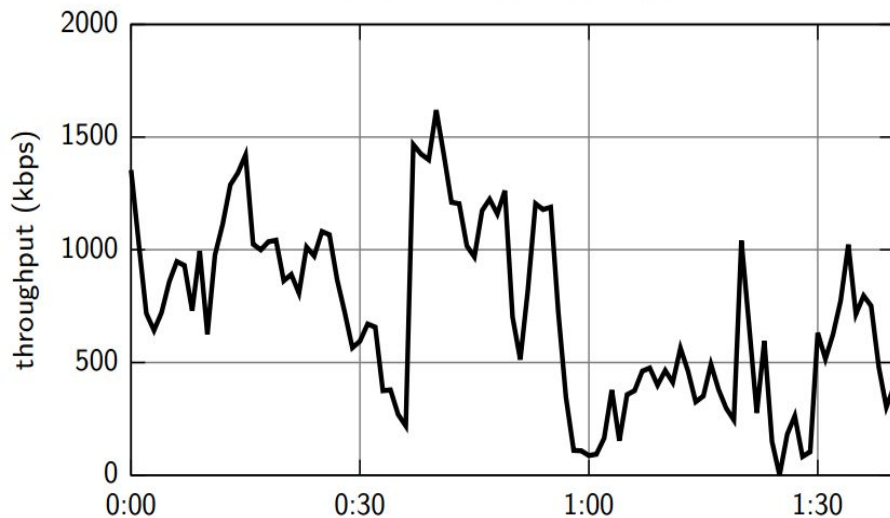
Paper Goals

- Design a protocol to cope with dramatic changes in link quality
- Maximize throughput and avoid delays while also not under-utilizing link
- Handle link outages without over buffering, and recovering gracefully

Cellular Links are not stable

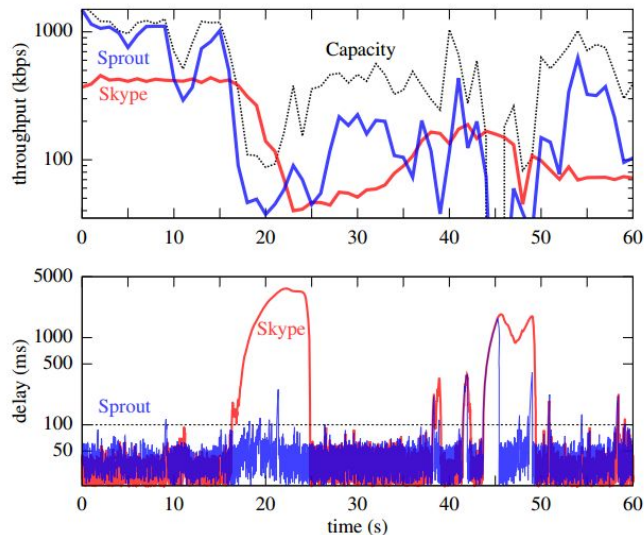
- Rapidly varying link rates
- Multi-second outages in either direction

Verizon LTE uplink throughput



Transport Protocols Today

- Deal with rate variations reactively.
- Slow to decrease their transmission rates
- Don't fully utilize capacity

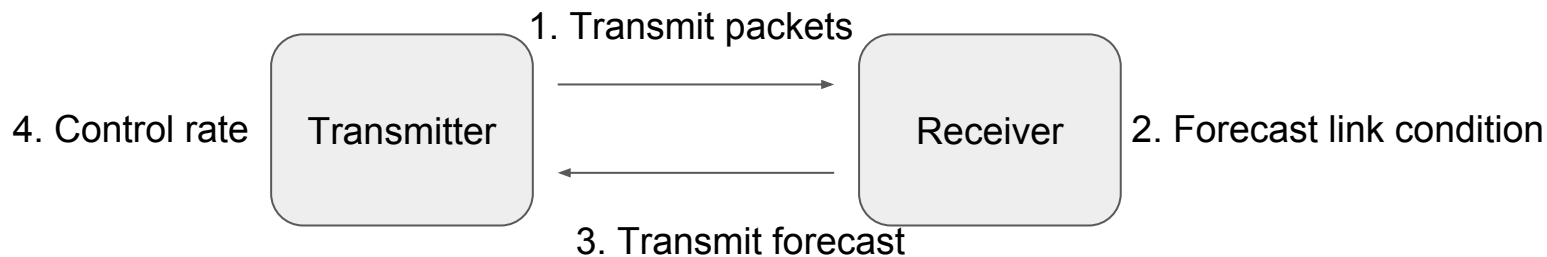


Cellular Links

- Mobile base station runs users' queues in round robin fashion
- End-to-end delay tends to be self-inflicted
- Traditional bandwidth delay product buffering breaks down
- Active Queue Management schemes are difficult to configure

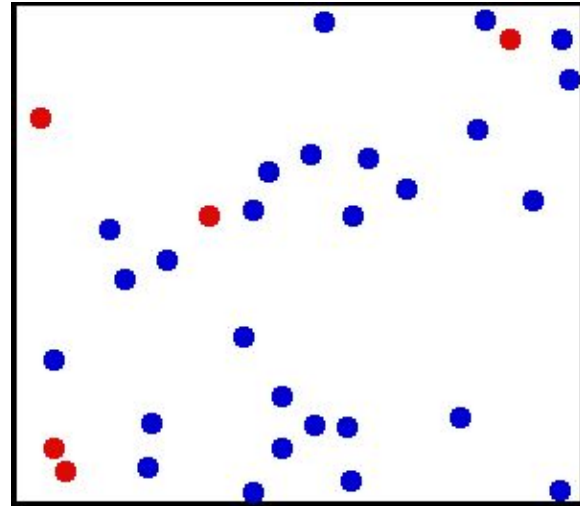
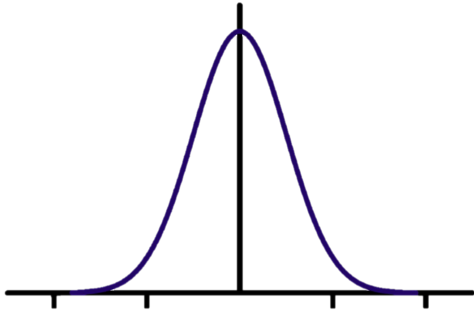
Sprout Algorithm Overview

- Goal
 - Achieving highest possible throughput, while preventing packets from waiting too long in a network queue



Network Path Model

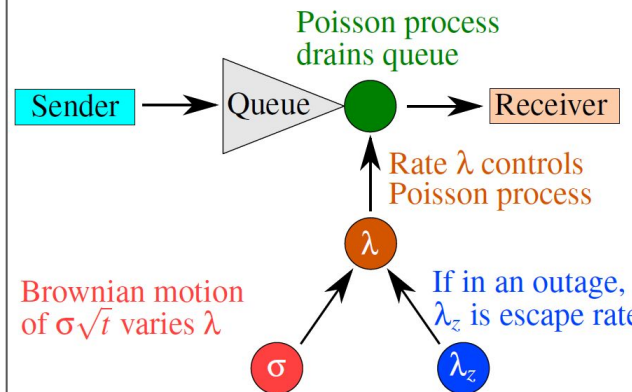
- Doubly-Stochastic process
 - Poisson process
 - Brownian Motion



Network Path Model (Cont'd)

- λ of Poisson process (arrival rate)
 - λ is discretized into 256 level
 - λ changes according Brownian motion with noise power σ (random walk to nearby levels)
 - Update interval: 20 ms
- Outage escape rate to model “sticky” property
 - If $\lambda = 0$, it tends to stay at outage (Exponential distribution λ_z)

Figure 3: Sprout’s model of the network path. A Sprout session maintains this model separately in each direction.



Update λ Distribution

1. Evolves λ by

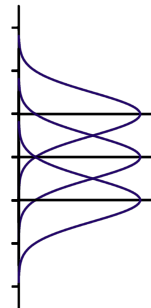
- Applying Brownian motion for λ level $\neq 0$
- Applying both Brownian motion and outage escape rate bias for λ level = 0

2. Observes the received bytes (k) to further update distribution of λ

$$F(x) \leftarrow \mathbb{P}_{\text{old}}(\lambda = x) \frac{(x \cdot \tau)^k}{k!} \exp[-x \cdot \tau]$$

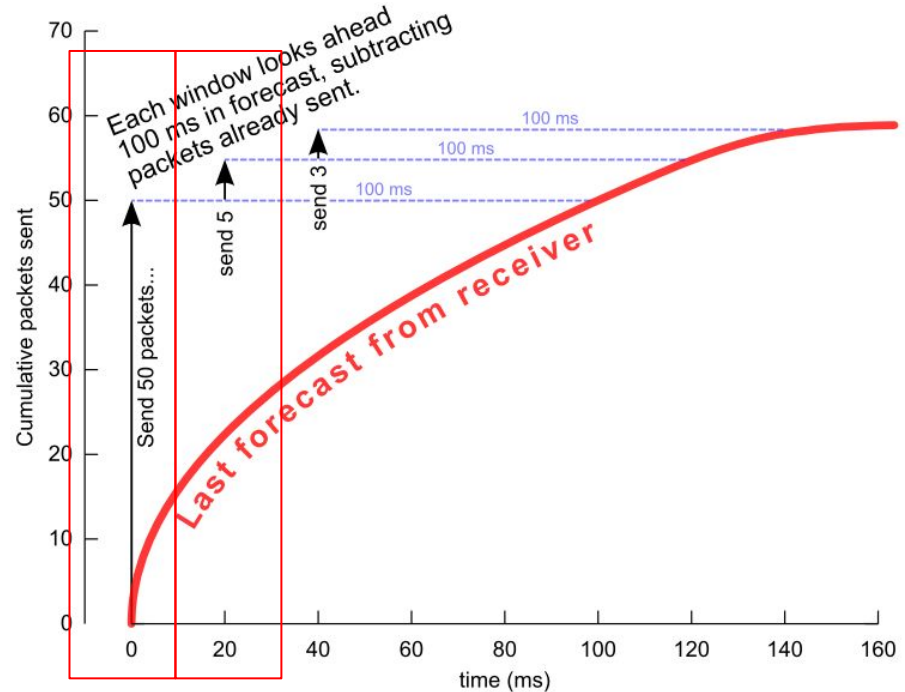
3. Normalizes the distribution of λ : $\mathbb{P}_{\text{new}}(\lambda = x) \leftarrow \frac{F(x)}{\sum_i F(i)}$

Time	0	20	40
Level 0	1/256	0.01	0.03
Level 1	1/256	0.02	0.01
...
Level 255	1/256	0.01	0.02



Forecast and Control Packet Delivery Rate

- Predict the link capacity by finding 5th percentile of λ distribution
- Forecast 8 ticks (160 ms)
 - Without the second step
- Transmitter will control the packet transmission based on the forecast and estimated packets in queue



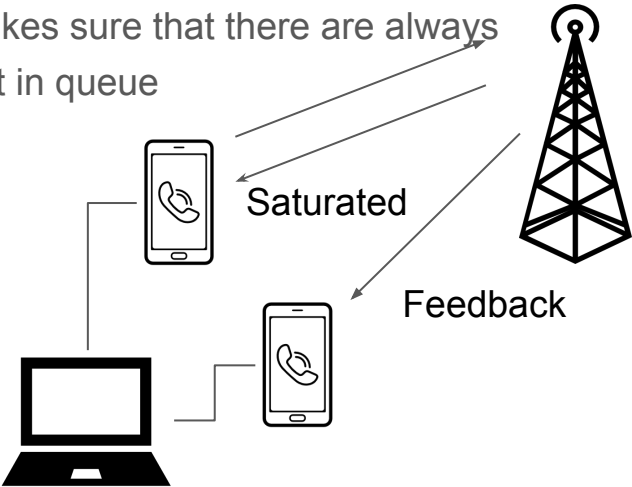
Experimental Testbed - Data Collection

- **Devices**

- A laptop to generate data
- One cell phone for data transmission and another for feedback

- **Saturator**

- A program that makes sure that there are always packets to transmit in queue

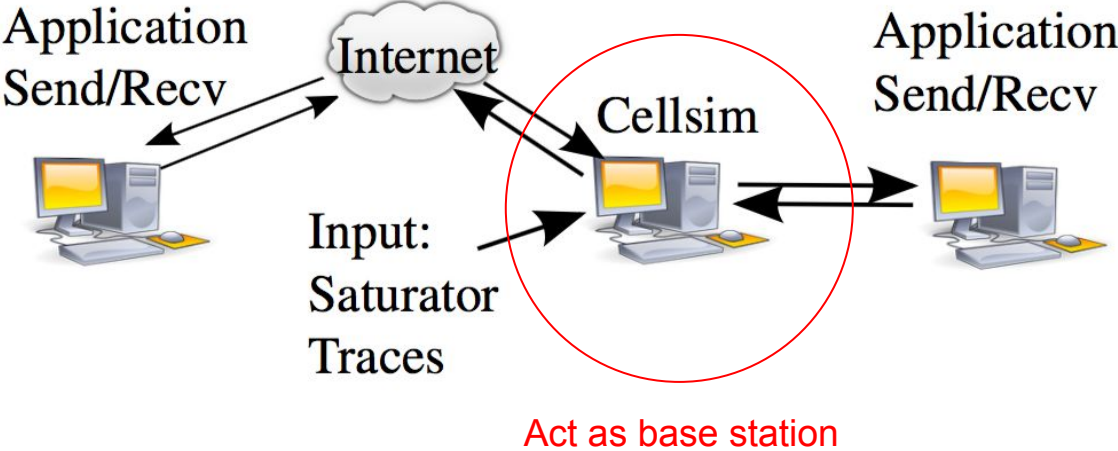


Trace Records

Packet Received Time	Packet Sent Time
1 ms	3 ms
2 ms	6 ms
4 ms	7 ms
...	...

Experimental Testbed - Cellsim

Figure 5: Block diagram of Cellsim



Trace Records

Packet Received Time	Packet Sent Time
1 ms	3 ms
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...	...

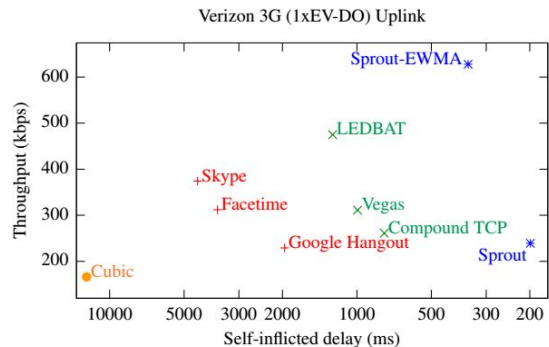
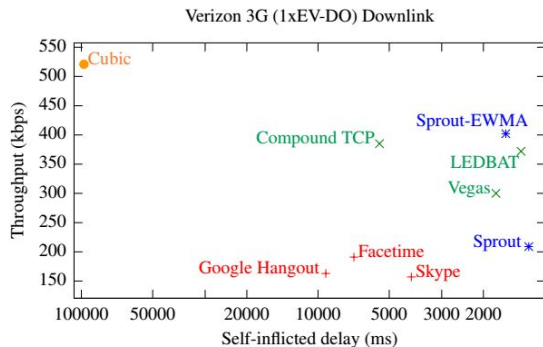
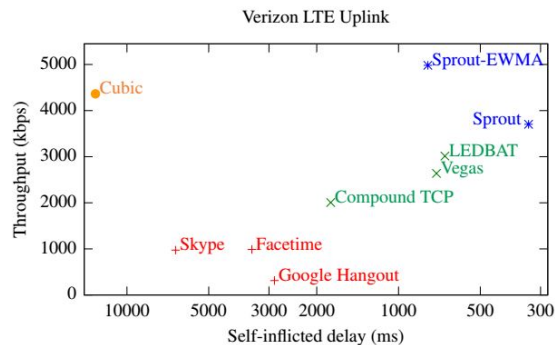
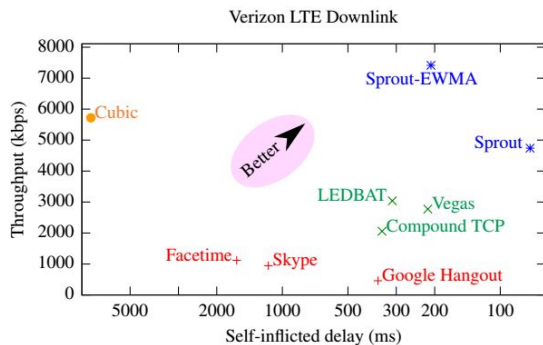
Sprout Exponentially Weighted Moving Average

- Used to compare benefits of Sprout's forecasting
- Still uses packet arrival times
- Doesn't do inference, but passes values to exponentially-weighted moving average (EWMA) function.
- Mostly results in higher throughput and delay compared to Sprout

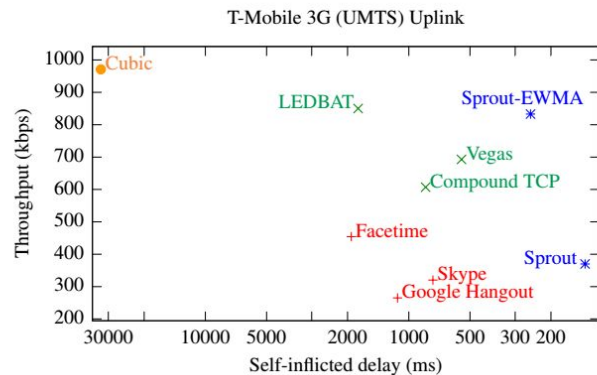
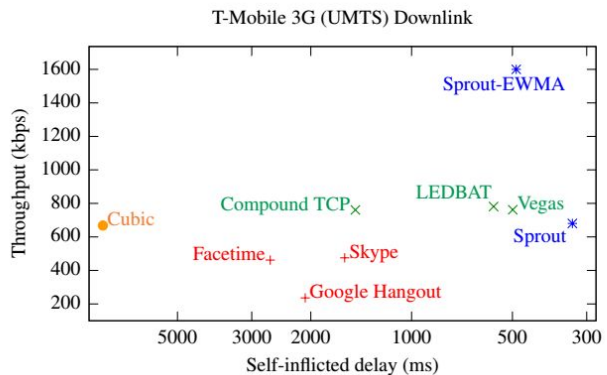
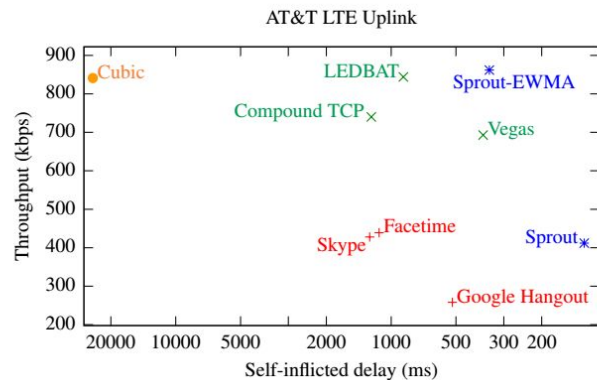
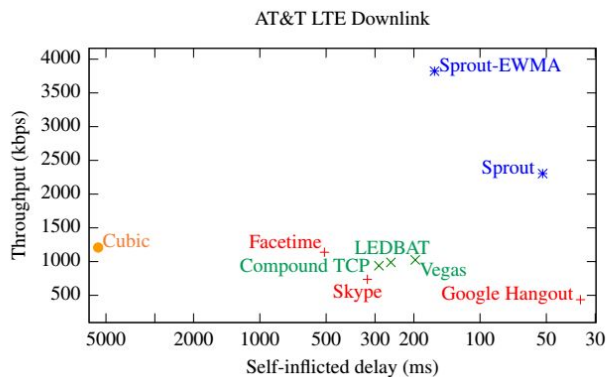
Evaluation Metrics

- Throughput: total bits received / duration of experiment
- Self-inflicted delay: difference in end to end delay between a perfect protocol and the chosen protocol

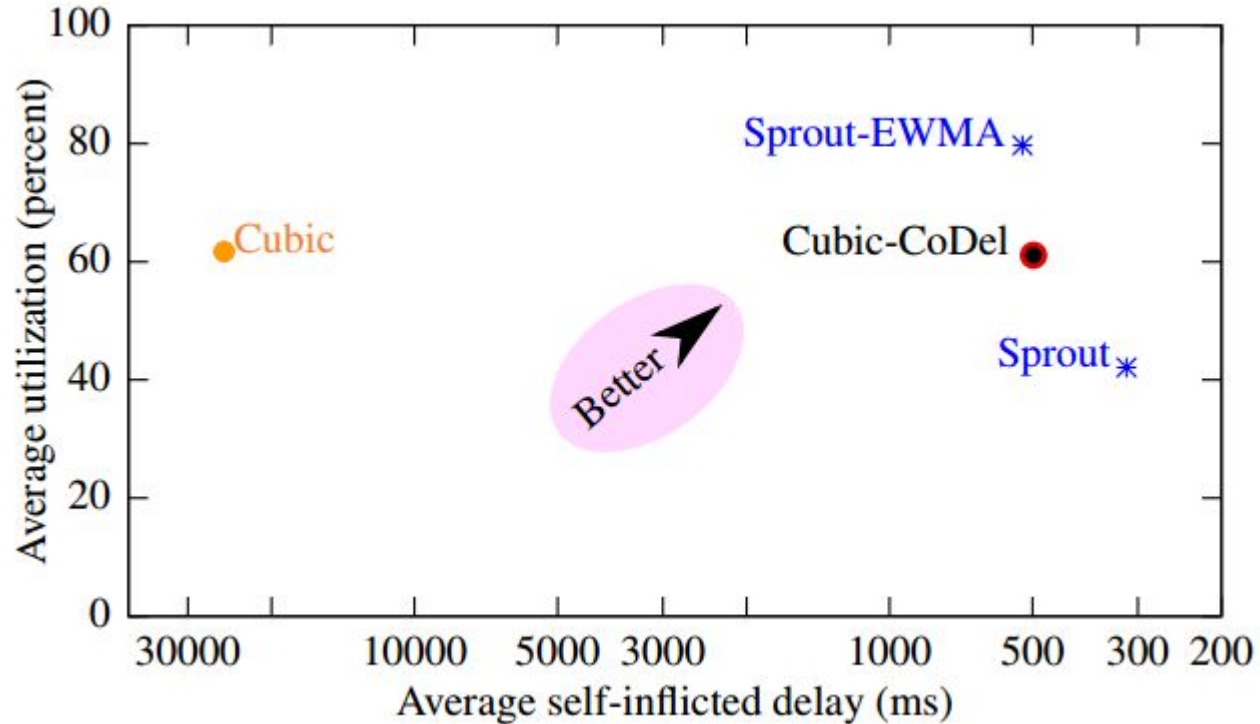
Results Over Different Networks



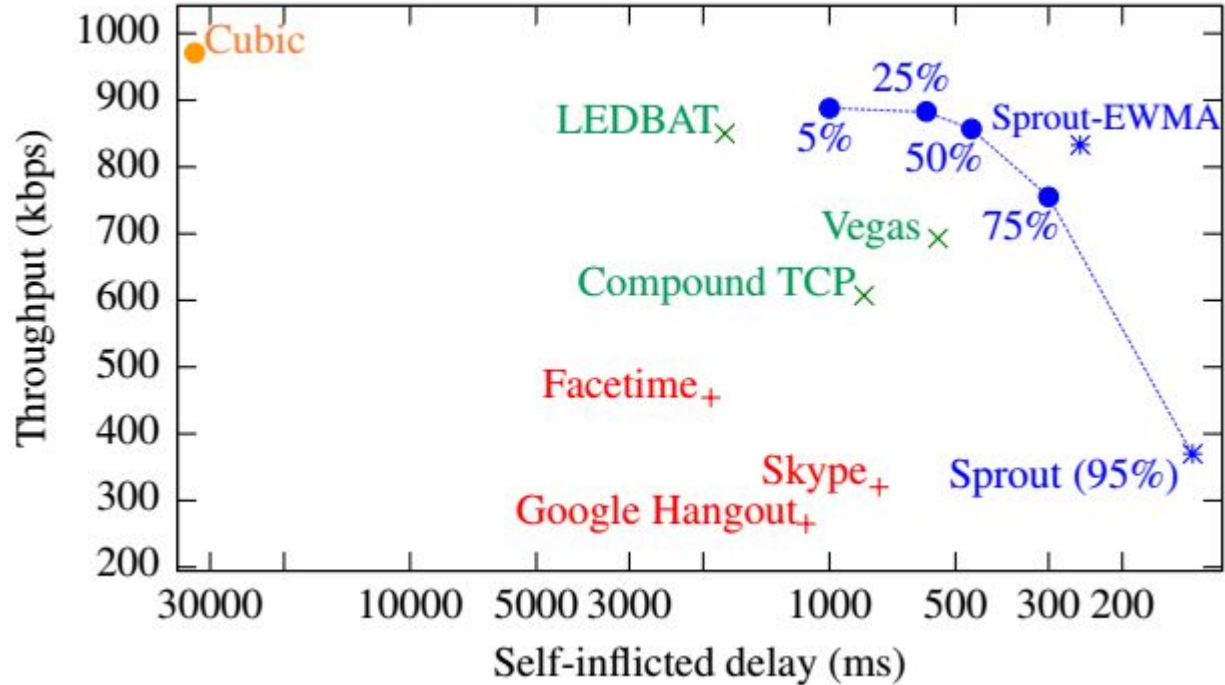
Results Over Different Networks (cont)



Results Compared to AQM



Confidence Parameter Effects



What's Novel and “aha”?

- Sprout can outperform CoDel
- Adoption of Brownie Motion Model

How to extend or adopt this work?

- Extend
 - Short term performance?
 - Trying different models other than Brownian motion
 - Multiple Sprouts at the same time
 - More traffic scenario should be considered
- Shortcomings
 - Self-similar property may still holds, is Poisson enough?
 - No explicit explanation of why adopting brownian motion