

# Once is Never Enough: Foundations for Sound Statistical Inference in Tor Network Experimentation

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#### Anonymous Communication with Tor

- Separates identification from routing
- Provides unlinkable communication
- Protects user privacy and safety online



#### **Motivation: Tor Experimentation**

#### Anonymous Communication with Tor

- Separates identification from routing
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#### Research to Improve Tor Performance

- A faster Tor means privacy is accessible to more humans
- Many ways to improve performance
- Need methods and tools to help us safely conduct Tor experiments
- We want experimentation results to be accurate and dependable so they can help inform real world decisions



(1) Model a Tor test network configuration

#### (2) Use Shadow<sup>1</sup> to run Tor experiments

#### (1) Model a Tor test network configuration

- How many clients? How many relays? How to sample relays?
- What are their node characteristics (location, bandwidth, rate limits, relay position)?

#### (2) Use Shadow<sup>1</sup> to run Tor experiments

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- <sup>(2)</sup> Previously done by considering the state from a single consensus (1 hour)
- <sup>©</sup> We consider the state of the network *over time* when sampling relays for test networks

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⊗ Large RAM and computational requirements
 ⊚ We reduced RAM usage by 64% and run time by 94%, enabling larger-scale experiments

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#### (3) Analyze and compare experimental results

<sup>©</sup> Previously, one experiment done with vanilla Tor and each research variant

<sup>©</sup> We present methods for quantifying the statistical significance across a set of experiments

## Outline

#### (1) Model a Tor test network

- (2) Use Shadow to run Tor experiments
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### Step (1) Modeling a Tor Test Network

- (1) We use data available on metrics.torproject.org to make informed modeling decisions
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#### (2) Simulate multiple users in each Tor client process to save RAM

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- (1) Model a Tor test network
- (2) Use Shadow to run Tor experiments
- (3) Analyze and compare experimental results



- Conducted performance audit of Shadow using the Linux perf tool
  - Fixed several performance bottlenecks
  - Added feature to shorten Tor bootstrapping
  - Enabled run-time optimizations
- Improved Shadow networking
  - Fixed non-determinism bugs
  - Improved network stack

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Table 2: Scalability improvements over the state of the art

Model	Scale s*	RAM	<b>Bootstrap Time</b>	<b>Total Time</b>	$\Omega^{\circ}$
			3 days, 11 hrs. 17 hrs.	35 days, 14 hrs. 2 days, 2 hrs.	
This work <sup>‡</sup>	100%	3.9 TiB	2 days, 21 hrs.	8 days, 6 hrs.	310

\* 31%:  $\approx$ 2k relays and  $\approx$ 250k users; 100%: 6,489 relays and 792k users

 $^{\circ}$   $\Omega$ : ratio of real time / simulated time in steady state (after bootstrapping)

<sup>†</sup> Using 8×10-core Intel Xeon E7-8891v2 CPUs each running @3.2 GHz.

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CCS'18 [38] <sup>†</sup>	31%	2.6 TiB	3 days, 11 hrs.	35 days, 14 hrs.	1850		
This work $^{\dagger}$	31%	932 GiB	17 hrs.	2 days, 2 hrs.	79		
This work <sup>‡</sup>	100%	3.9 TiB	2 days, 21 hrs.	8 days, 6 hrs.	310		

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Supports larger test networks



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- (1) Model a Tor test network
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#### Running experiments involves two levels of sampling:

- 1. Sampling a test network model at some scale  $\leq 100\%$
- 2. Simulating the sampled test network with a seed

#### Previous work uses one simulation for each research variant

• Ignores sampling error

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- 1. Sampling a test network model at some scale  $\leq 100\%$
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#### Previous work uses one simulation for each research variant

• Ignores sampling error

#### One simulation is never enough

- We need repeated sampling of test networks (not just sim seeds)
- We quantify the sampling error by computing CIs over empirical CDFs
- Allows us to make statistical arguments for the observed results



#### **Estimating the True CDF with CIs**

Each simulation in each test network produces an empirical CDF for our metric of interest





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We use the mean to estimate the true CDF, and sampling and measurement error to compute CIs



Demonstrate how to apply our methods with an example

Hypothesis:

• Increasing user traffic load by 20% will decrease Tor performance for existing clients

#### Experiment setup

- 100% and 120% traffic loads
- 1%, 10%, and 30% scale factors
- 420 total simulations

# How does network scale affect the conclusions we can draw from the results?

Scale	Load	Number of Simulations
1%	100%	100
1%	120%	100
10%	100%	100
10%	120%	100
30%	100%	10
30%	120%	10



Scale = 1% Cls overlap even with 100 sims





Scale = 1%

CIs overlap even with 100 sims



Scale = 10%

Cls are close with 5 sims,



Scale = 1%Scale = 10% Scale = 30% Cls overlap even with 100 sims CIs clearly separate with Cls are close with 5 sims, either 5 or 10 sims but separate with 10 sims 0.99-0.99 0.99 scale) CDF (log scale) CDF (log scale) 0.98 0.98 0.98 0.97 0.97 0.97 (log 0.96 0.96 0.96 0.95 0.95 0.95 0.94 0.93 0.92 0.91 0.99 0.94 0.94 CDF 0.93 0.92 0.91 0.9-0.93 0.92 0.91 0.9- $\ell = 1.0, n = 5$ . . . . . . . . . . . . . . . . True  $\ell = 1.0, n = 10$ True Estimated True  $\ell = 1.0, n = 10$  $\ell = 1.0, n = 100$  $\ell = 1.0, n = 5$ 0.8 0.8 0.8 Estimated Estimated 0.7 0.7 0.7 *ℓ*=1.0, *n*=100 *ℓ*=1.2, *n*=5 *ℓ*=1.0, *n*=10 0.6 0.6 0.6 0.5 *ℓ*=1.2, *n*=10 0.5 ℓ=1.2, *n*=10 0.5  $\ell = 1.2, n = 5$ 0.4 0.4 0.2 0.0 0.4 0.3 0.2 0.0 0.3 0.2 0.0*ℓ*=1.2, *n*=100 *ℓ*=1.2, *n*=100 *ℓ*=1.2, *n*=10 30 355040 500 20252030 40 60 2030 51015100 100 Time to Last Byte (s) Time to Last Byte (s) Time to Last Byte (s)





#### More simulations needed at smaller scales, fewer at larger scales to reach a certain CI precision



#### Summary



	Primary Contributions	Main Results
S	(1) New methods for constructing Tor test networks considering the state of the network over time rather than at a static point	<ul> <li>We create many test Tor networks in Shadow</li> <li>→ a Tor network with up to 6,489 relays</li> <li>→ traffic of up to 792k simultaneously active users</li> </ul>
	(2) New/improved experimentation tools, optimized to run Tor faster and at larger scales than previously possible	<ul> <li>We enhanced Shadow to reduce:</li> <li>→ RAM usage by 64%</li> <li>→ run time by 94%</li> </ul>
	(3) New methodology for conducting statistical inference using results collected from experiments in smaller-scale Tor networks	Running multiple simulations in independently sampled Tor networks → necessary for statistical significance
<b>`</b>	(4) Demonstrated how to apply our methodologies to conduct sound Tor performance research	<ul> <li>To reach a desired precision requires:</li> <li>→ more simulations in smaller-scale networks</li> <li>→ fewer simulations in larger-scale networks</li> </ul>

rtifacts: https://neverenough-sec2021.github.io

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