

NN4SysBench: Characterizing Neural Network Verification for Computer Systems

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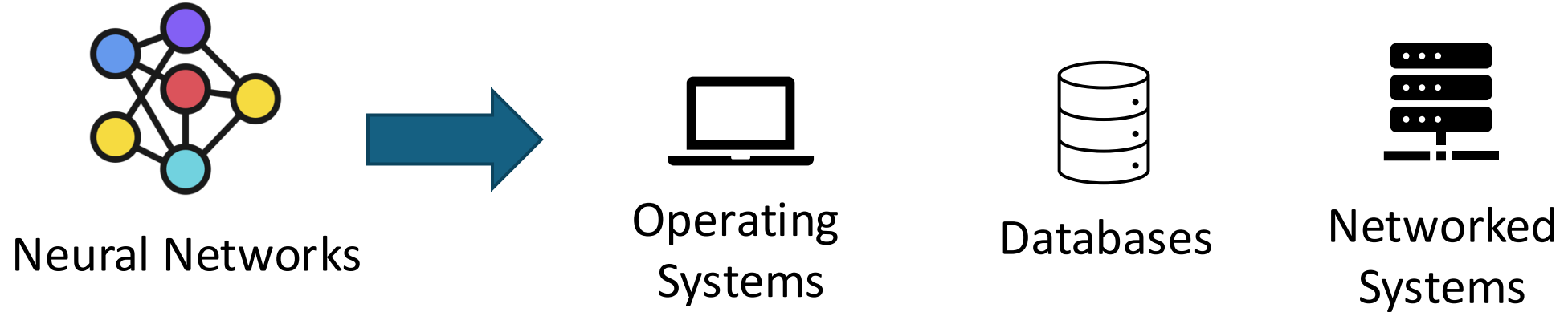
Tübingen AI Center

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University**



I ILLINOIS

Neural Networks for computer systems (NN4Sys)



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Neural Networks for computer systems (NN4Sys)



For example, NN for

- OS scheduler => better job completion time
- database index => smaller memory footprint, faster lookup
- network routing => more efficient packet routing

However, NN4Sys...

- ..., despite improving the **average performance**, ...
- ...may do much worse in the **worst-case scenarios**.

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- ...may do much worse in the **worst-case scenarios**.

For example, NN for

- OS scheduler => job starvation
- database index => pointing to wrong positions
- network routing => blackholes (loops in the chosen path)

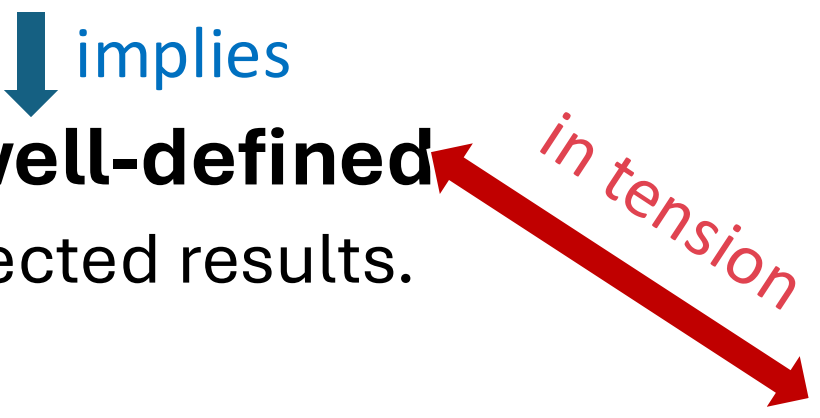
The problem is fundamental

- NNs are complicated **black boxes**.
 - Hard to understand what a NN has learned.

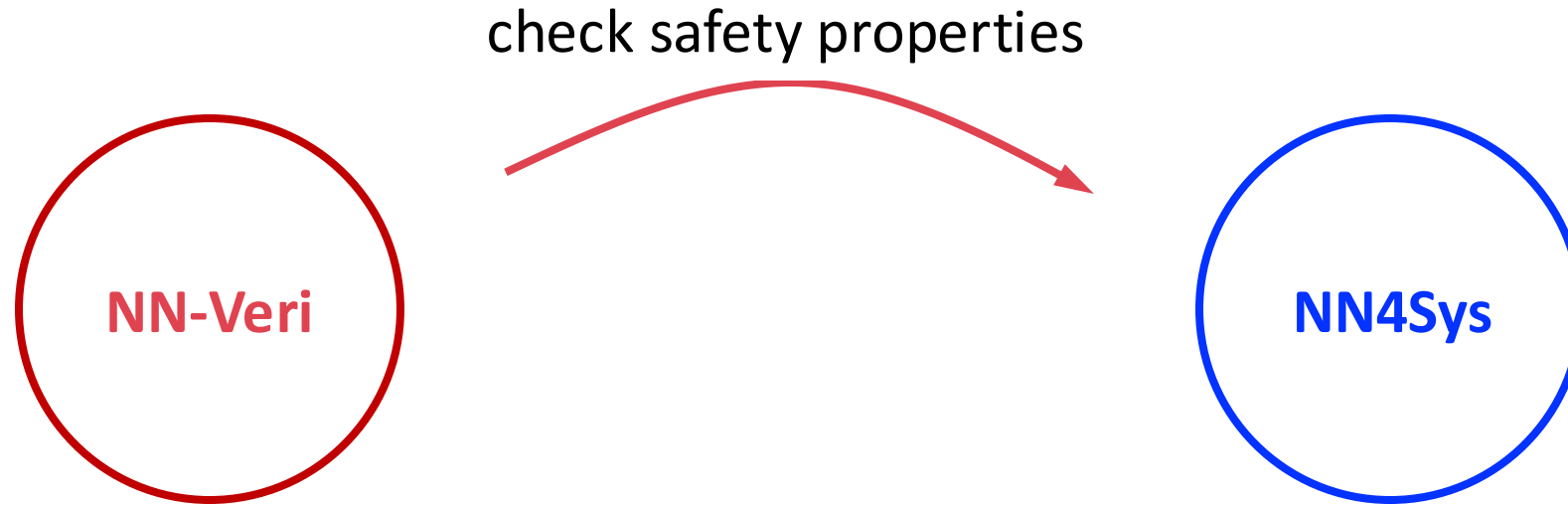
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- ↓ implies
- NN's behaviors are **not well-defined**.
 - NNs may produce unexpected results.

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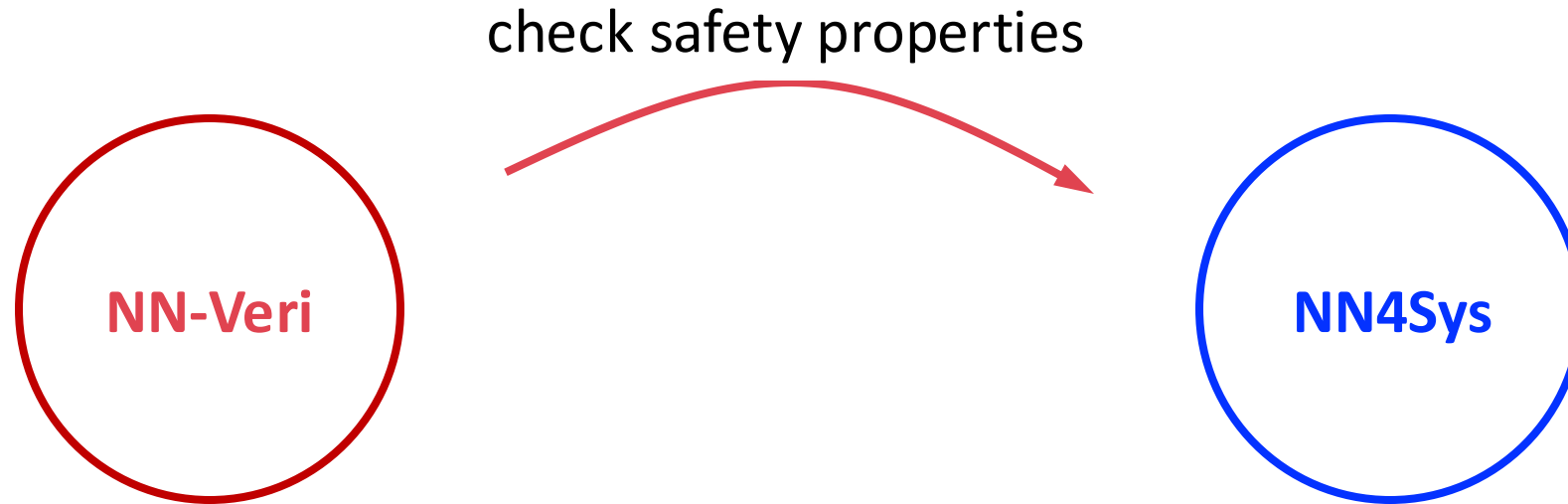
- NNs are complicated **black boxes**.
 - Hard to understand what a NN has learned.
 - NN's behaviors are **not well-defined**.
 - NNs may produce unexpected results.
 - Meanwhile, computer systems require **safety properties**.
- 

NN-Verification to the rescue



NN-Verification can provide a “lower bound” for NN4Sys.

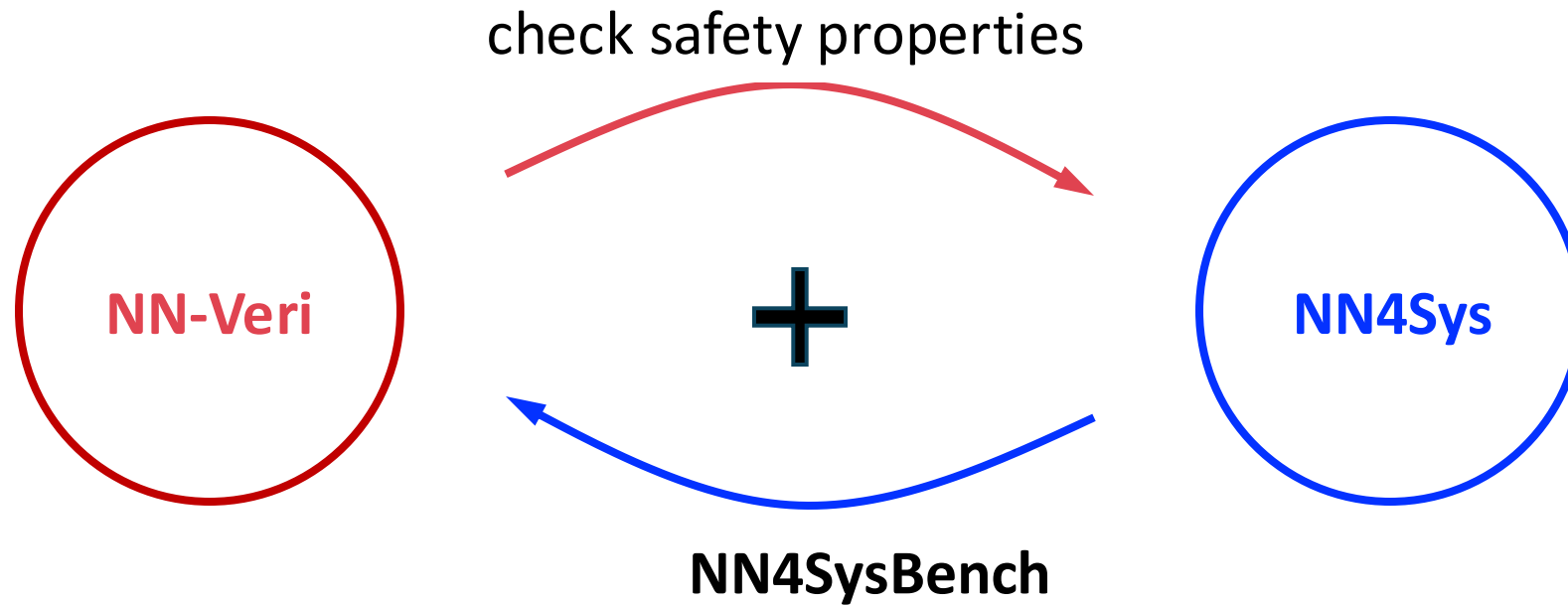
NN-Verification to the rescue



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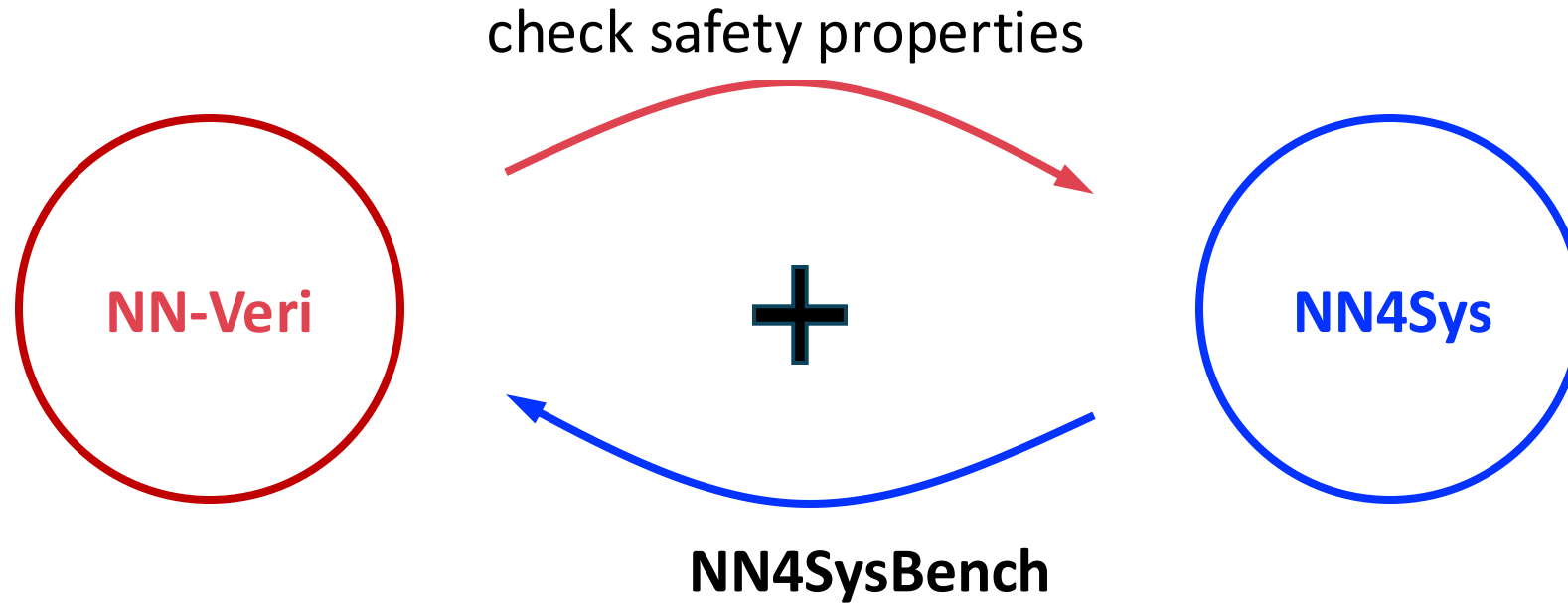
- Crucially, NN-Verification works well for NN4Sys because
 - NN4Sys has simple models;
 - NN4Sys has clear semantics.

Our vision: NN4Sys + NN-Verification



- We argue: NN4Sys + NN-verification whenever possible

Our vision: NN4Sys + NN-Verification

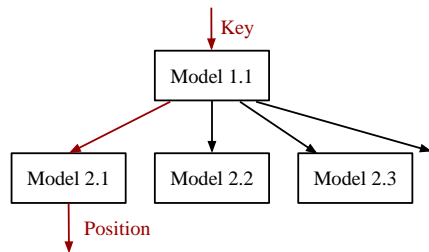


- We argue: NN4Sys + NN-verification whenever possible
- However, NN4Sys have **characteristics**...
...that today's verifiers do not support well.

NN4SysBench: a benchmark suite for **NN-Verification**
whose benchmarks are from impactful **NN4Sys**

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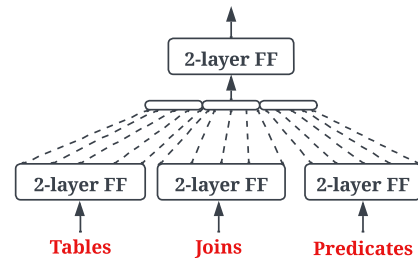
learned index



specifications

$$\forall k \in [\mathcal{K}[i], \mathcal{K}[i+1]],$$
$$F(k) \in [DB(\mathcal{K}[i])-\epsilon, DB(\mathcal{K}[i+1])+\epsilon]$$

learned cardinality estimation



specifications

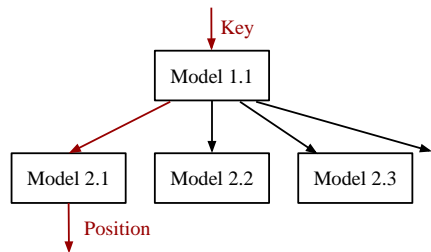
1. $F(q) \leq |R|$.
 2. $F(q) \geq 0$.
 3. $F(q) \geq F(q \wedge \theta_i), i \in [1, d]$.
- ...

supervised learning

learned bloom filter

NN4SysBench: a benchmark suite for **NN-Verification** whose benchmarks are from impactful **NN4Sys**

learned index

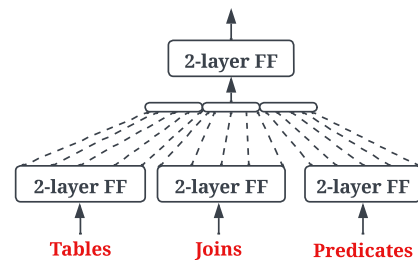


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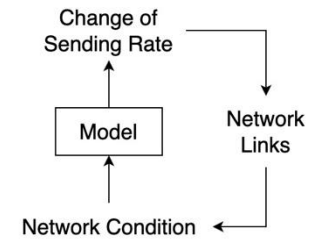
supervised learning

learned bloom filter

learned adaptive bitrate

learned distributed system scheduler

learned Internet congestion control



specifications

- Basic properties: in good (or bad) network condition, model does not decrease (increase) packet sending rates.
- ...

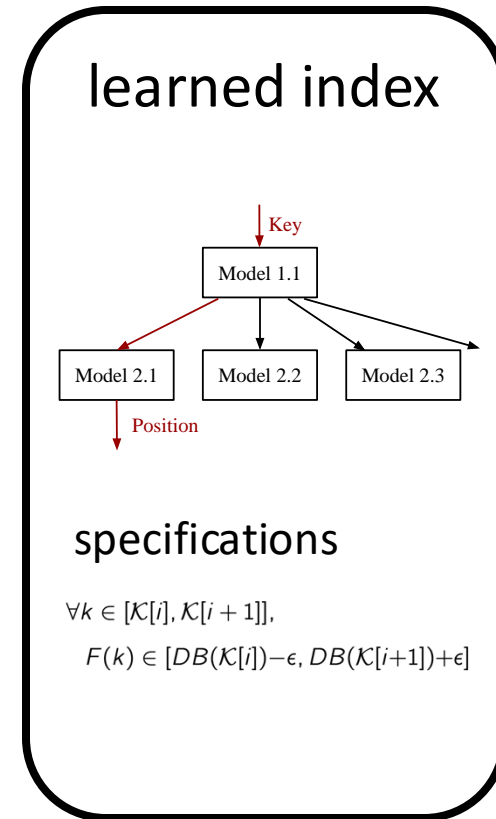
reinforcement learning

An (incomplete) list of NN4Sys characteristics

1. small number of input dimensions
2. large number of specification entries
3. hierarchical models
4. temporal specification
5. monotonicity specification

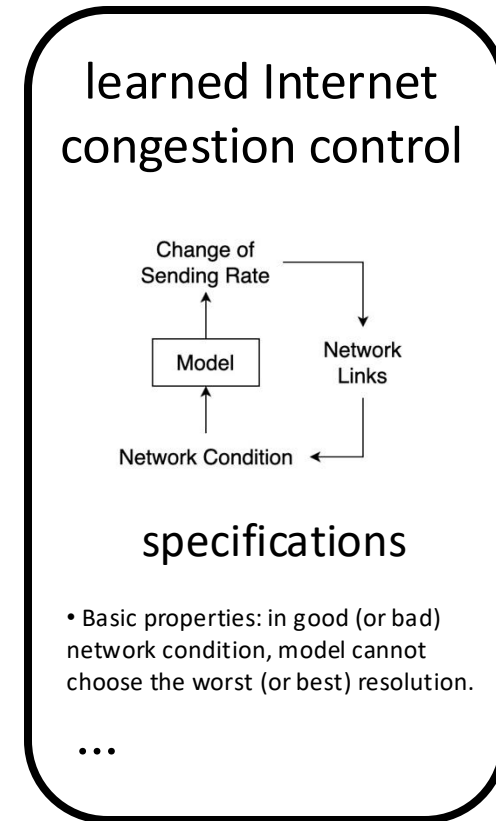
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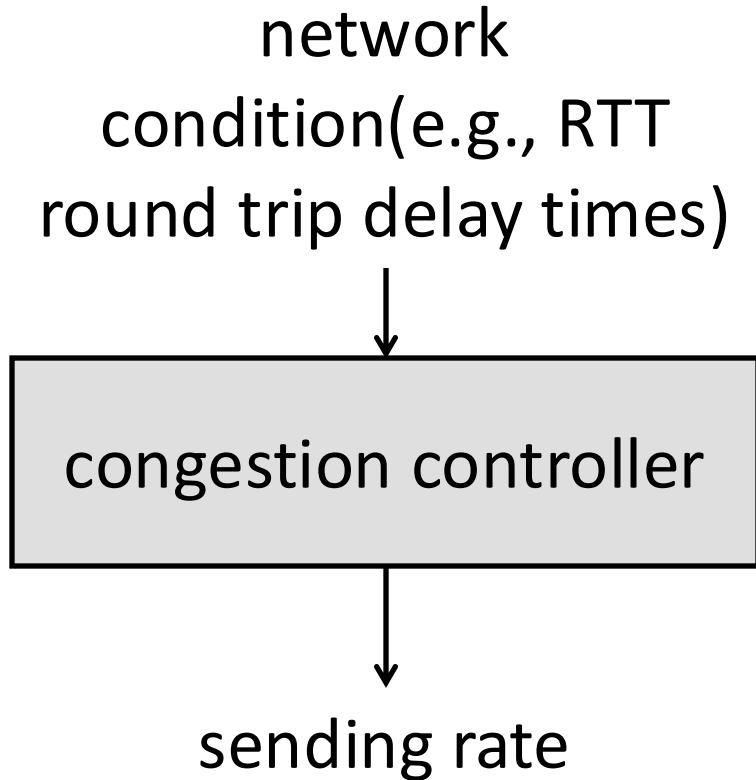


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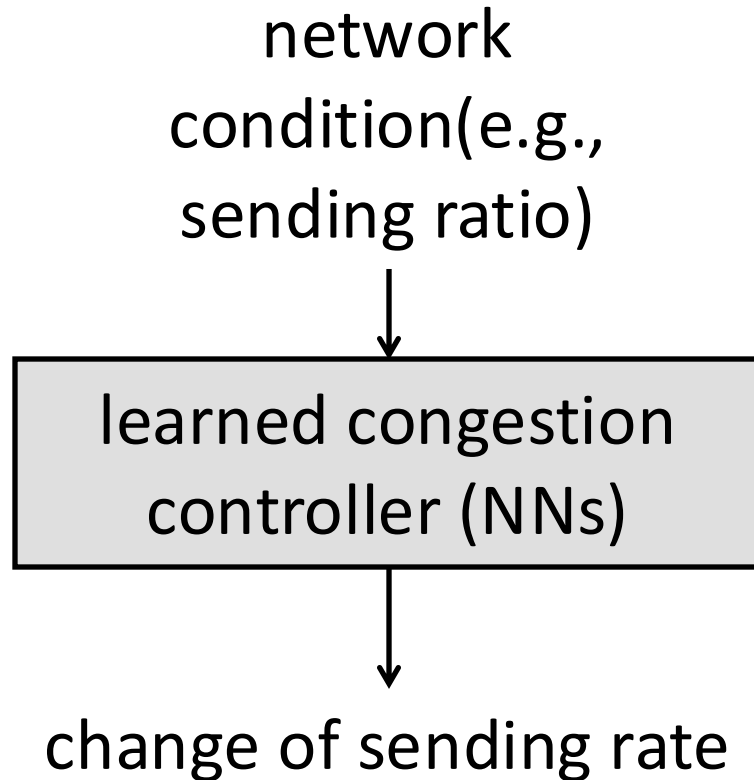


Traditional congestion control



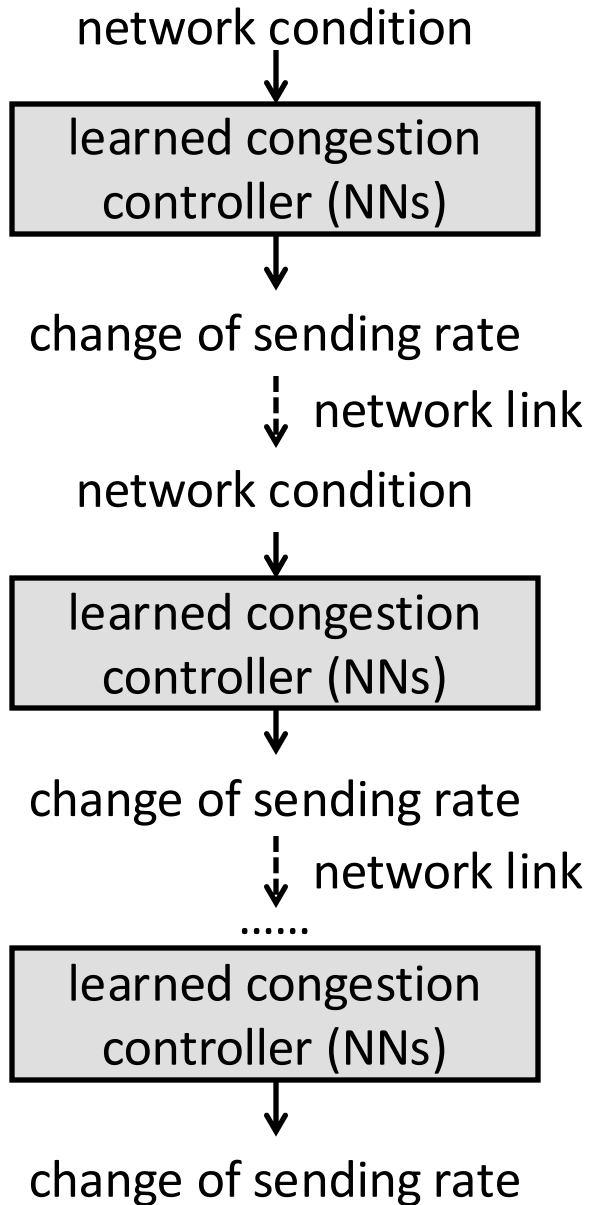
- Classic congestion controller: Cubic, BBR

Learned congestion control



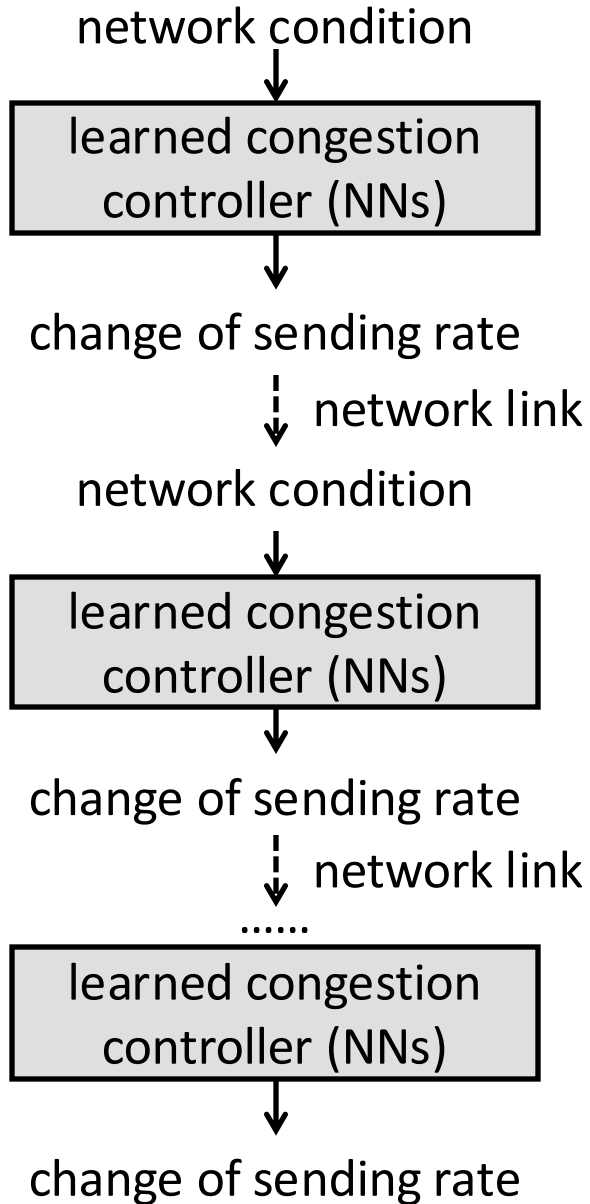
- Classic congestion controller: Cubic, BBR
- Learned congestion control: use NNs to replace the congestion control algorithm

Specifications of learned congestion control



- When the network condition changes from bad to good, the sender **eventually** increases packet sending rates.

Specifications of learned congestion control

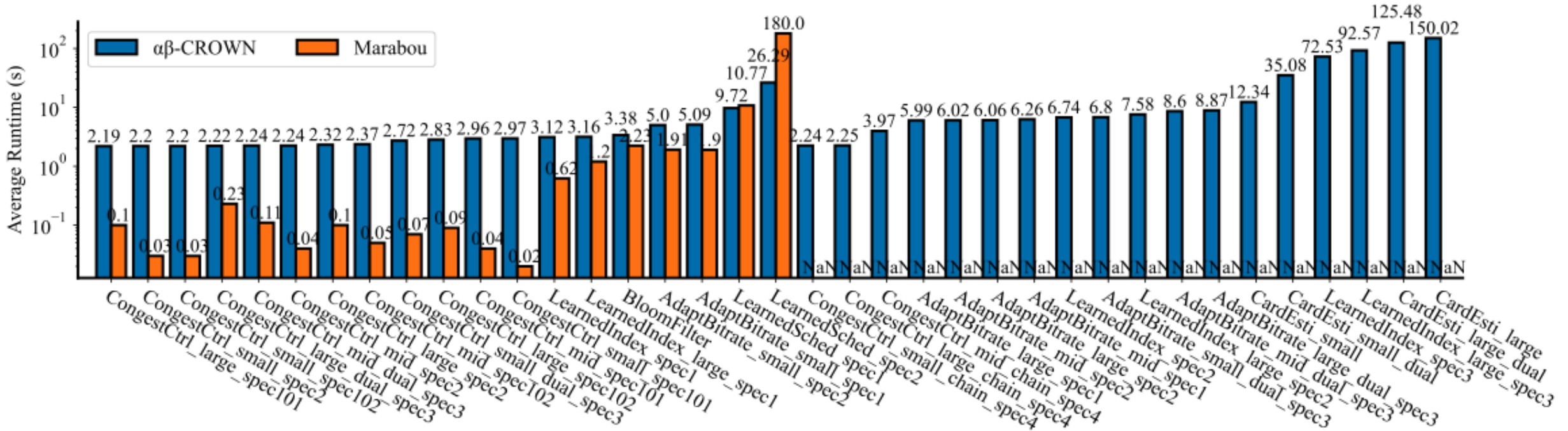


- When the network condition changes from bad to good, the sender **eventually** increases packet sending rates.

multiple steps

characteristic 4: temporal specification

Evaluation on 2 verifiers: $\alpha\beta$ -crown and marabou



Verification runtime for different benchmarks

NN4SysBench: Characterizing Neural Network Verification for Computer Systems

Contribution: Previous version adopted by the Verification of Neural Networks Competition (VNN-Comp) for 3 consecutive years

Code, paper and contact info at website: <https://shuyilinn.github.io/BenchmarkWeb/>