Peeking into Black Boxes: Automated Fuzzing of Router Resource Usage

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Motivation

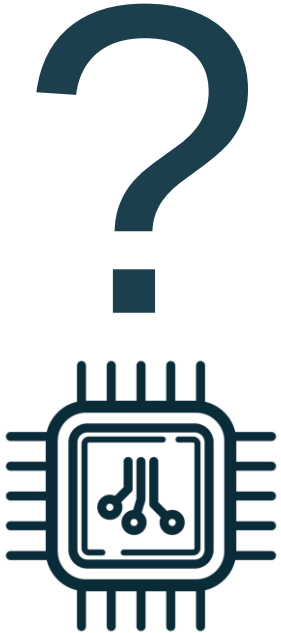
➔Managing changes in critical networking infrastructures requires thorough testing

➔Router resource testing often is a blind spot in test plans (black box)

- Where are the router‘s limits (e.g. TCAM space)?
- How much headroom is left for future innovation?

➔Reasons:

- Vendors are tight-lipped on hardware resources
- Complex topic (even for vendors)



Our Approach

➔Fuzzing/fuzzy testing in security research [1]

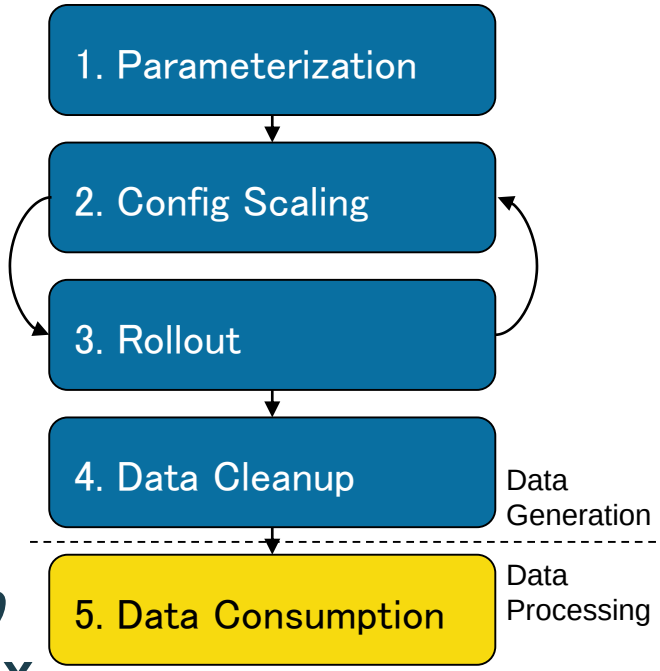
- Automated security testing
- Generate random inputs, monitor implementation behavior

➔We are not looking for security issues but explore the HW limits of the router under test

1. Generate masses of guided, valid configuration changes
2. Measure router behavior: runtime errors and exposed hardware counters
3. Correlate configuration and measurements, identify scaling behavior and possible bottlenecks

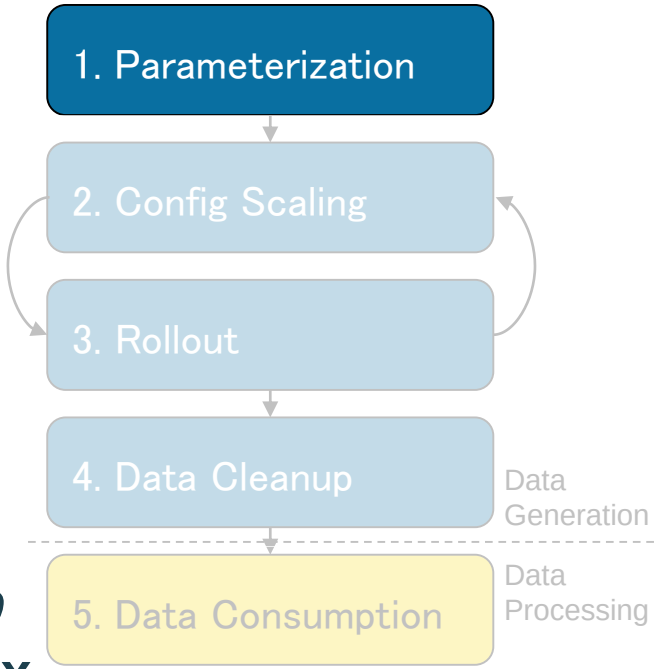
[1] https://en.wikipedia.org/wiki/Fuzzing

Fuzzing Framework Design

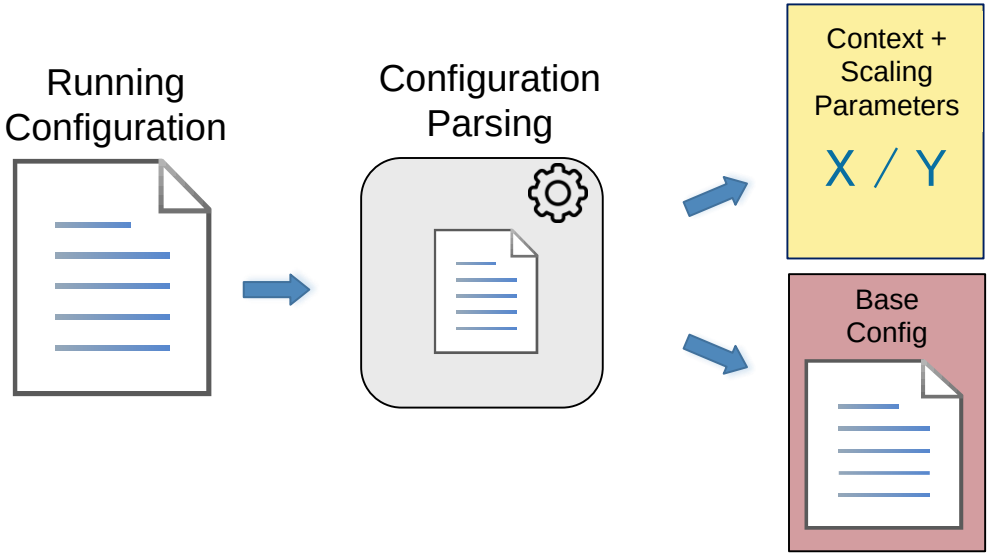


- ➔ Modular five stage framework
  - Based on Python and Jinja2 for templating
  - Flexible and adaptable (e.g. different vendors)
- ➔Use existing production configuration
  - Extend as required by test case
  - Evaluate scalability of single routers (i.e. vertical)
- ➔ Design goals
  - Automation of repetitive steps (>1000 configurations)
  - Visualization + Identification of bottlenecks
  - User support in all stages

Framework Design: Parameterization



- ➔ User provides production configuration
- ➔ Extract scaling parameters



Framework Design: Config Scaling

1. Parameterization
2. Config Scaling
3. Rollout
4. Data Cleanup
5. Da

Context

```
"e_pol": {
  1: {
    "descr": "This is policy 1"
  },
  . . .
  678: {
    "descr": "Last policy"
  }
}
```

From parameterization

+

Jinja Template

```
forwarding
  {% for policyid, policy in e_pol.items() %}
  new rule {{ policyid }}
    description {{ policy.description }}
  done
done
```

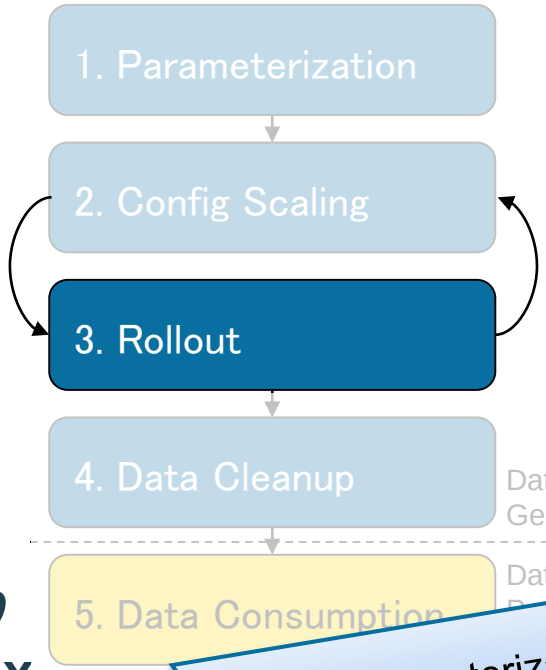
User supplied

Rendered Extension Config

```
forwarding
 new rule 1
   description "This is policy 1"
 done
 . . .
 new rule 678
   description "Last policy"
 done
```

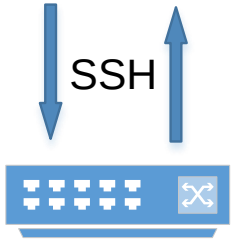
Inlined python

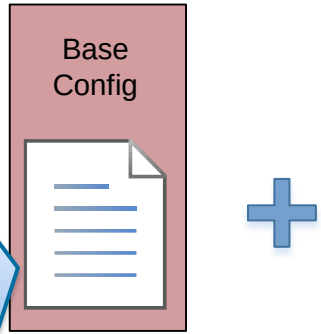
Framework Design: Configuration Rollout



→ Rollout

- Apply configuration on test router
- Gather logs (accepted/rejected, errors, hardware counter dumps, …)
- Create clean test environment (reboot or rollbacks)





From scaling

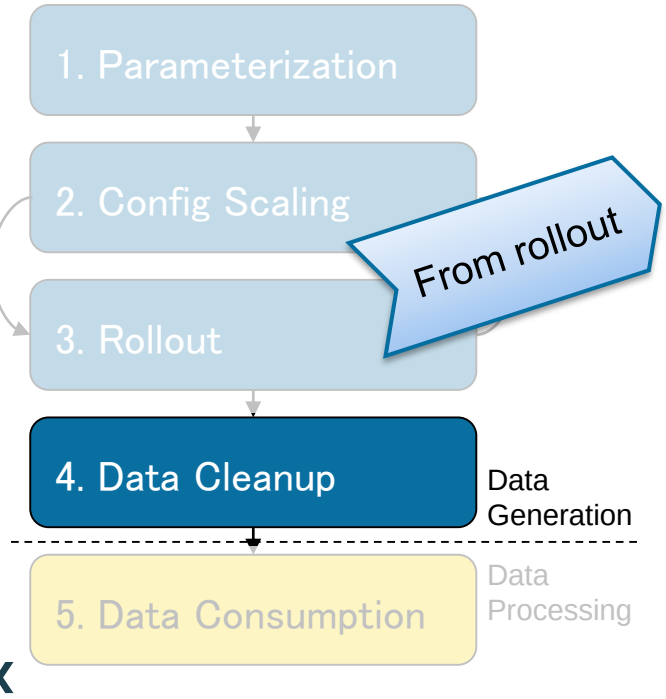
From parameterization

Execution Log

Error Log

Resource Dumps

Framework Design: Data Cleanup



➔Extract data

- Execution log
- Error codes
- Resource usage

➔Create data pool

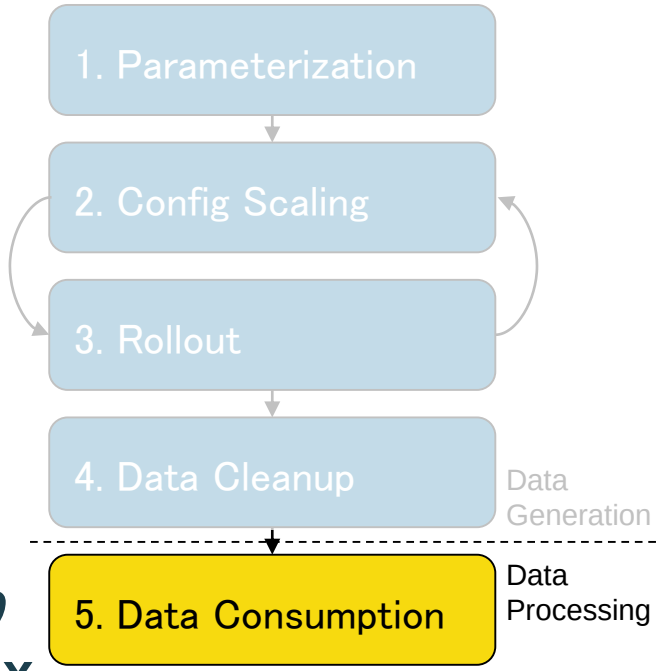
- Combine data from all runs

➔Standardize output

- (see right)

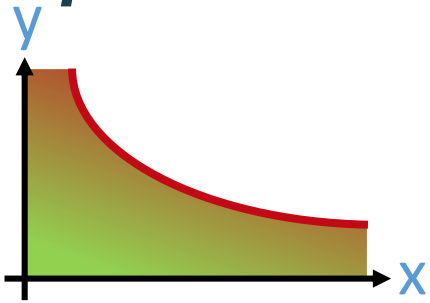
```
{
  "desc": "Resource X",
  "linecard": "20",
  "allocated": null,
  "scaling_x": 111,
  "scaling_y": 222,
  "error": "42",
  "errordesc": “Resource X exhausted“
}
```

Framework Design: Data Consumption



➔Visualization

- Generate plots on measured errors
- Visualize resource usage



➔Predictive modelling

- Decision tree model
- Provides flow-chart like predictions whether a scaled configuration will exceed available resources
- Useful for management and predictive maintenance

Case Study: QoS+ACLs for Traffic Filtering

→ Use Case

- Drop DDoS/unwanted traffic at IXP
- How many QoS policies+ACLs per port can we apply before running out of resources?
- How does resource usage scale?
- What are the bottleneck resources?

→ Test Setup

- Complex service router with multiple line cards
- Production configuration of a multiple Tbps/>100 ports router as a base configuration
- Generated extension configurations scale #QoS policies and #ACLs per policy

In the following, axes of plots are obfuscated due to NDAs.

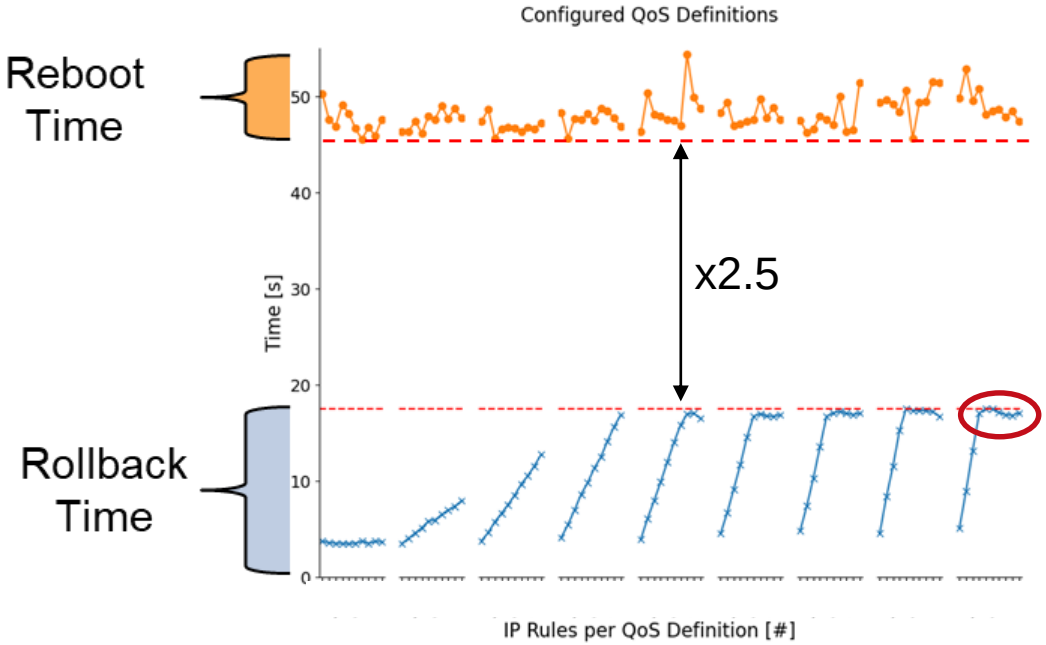
Timescale of Experiments

Step Size (#QoS and #ACLs/QoS)	**10**	**20**	**40**	**80**
Time	23h11m	5h44m	100m	23m
#Data Points	2295	598	168	48

➔ Total runtime

- Configuration scaling, rollout, data collection, environment cleanup
- Environment cleanup is the bottleneck (see next slide)

Environment Cleanup: Rollback vs. Reboot

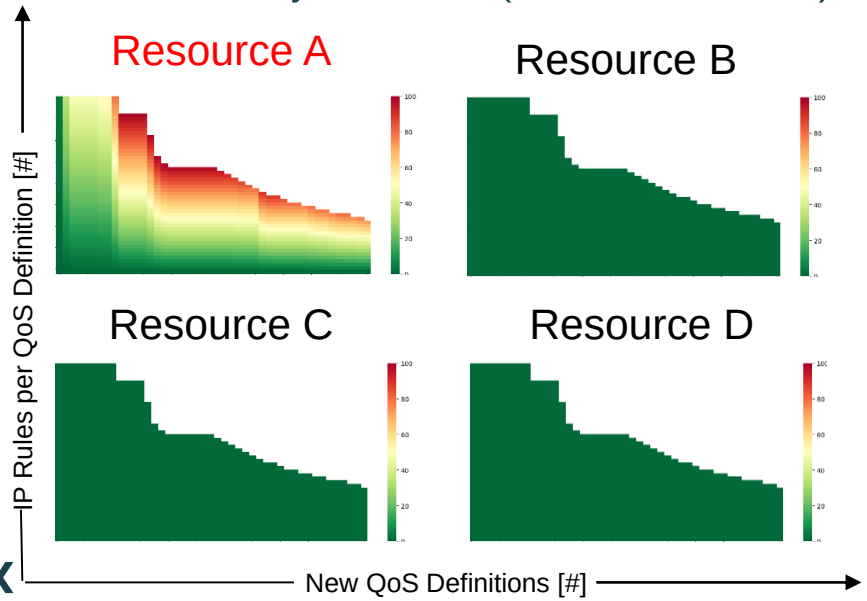


- At least 2.5 times speedup with rollbacks compared to reboots per measurement
- Rollback time depends on size of extension configs
- Plateaus (◯) indicate large failing configs

Resource Usage

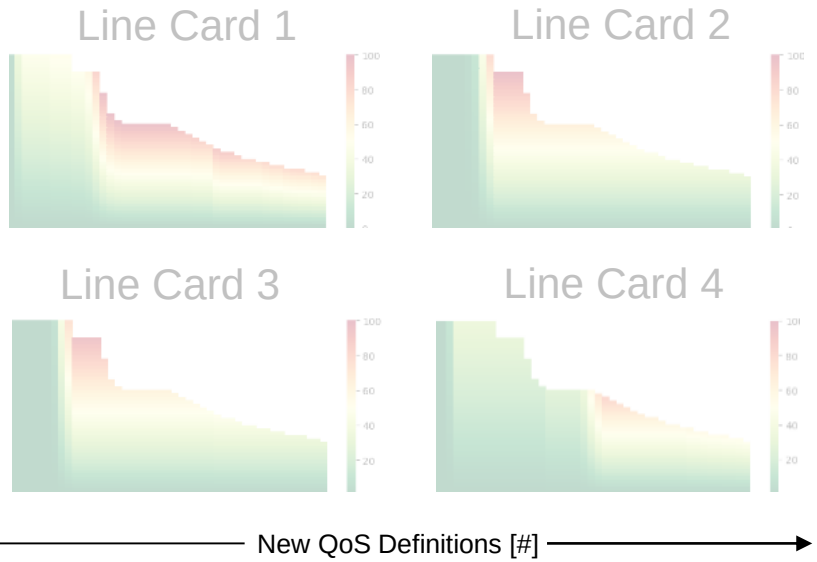
➔Aggregation across all line cards

- Identify bottleneck (max. HW counters)



➔Drill down per line card for Resource A

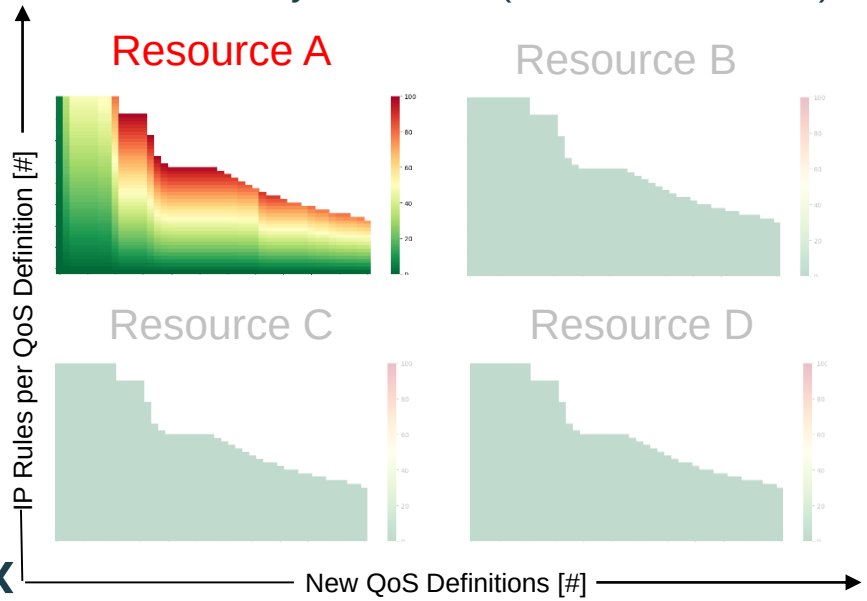
- Identify bottleneck hardware module



Resource Usage

➔Aggregation across whole device

- Identify bottleneck (max. HW counters)



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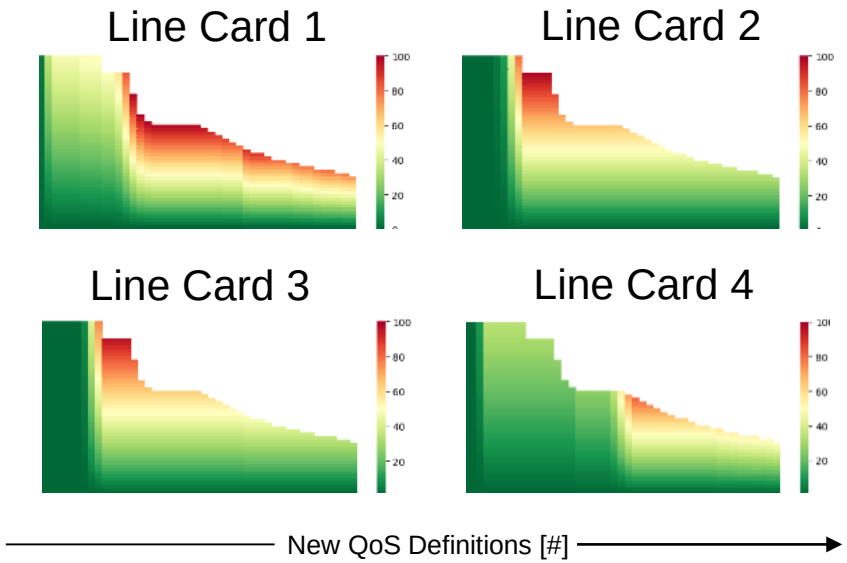
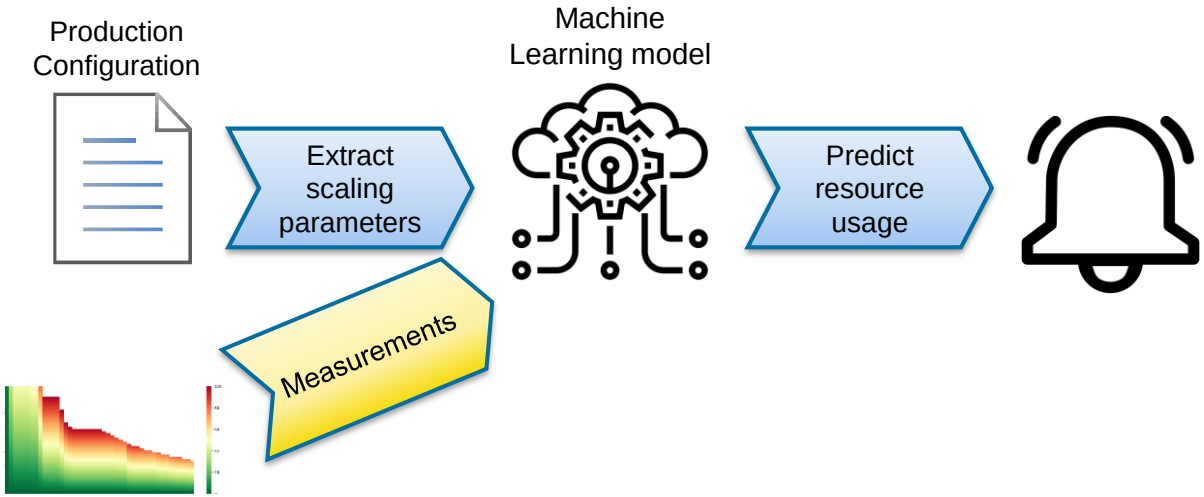


Monitoring without Measurements



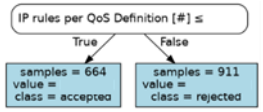
➔ Can we use the datasets to approximate resource usage without measurement?

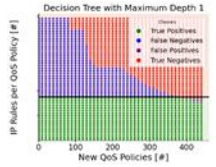
➔ Predictive maintenance can help mitigating problems even before deployment

Decision Tree Resource Model

- ➔Decision Tree ML-model
- ➔Train/test (70/30) split of measured data
- ➔Prediction accuracy > 98%
- ➔Tree depth allows tuning understandability vs. accuracy trade-off

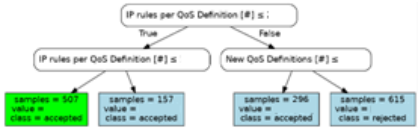
Maximum Depth 1

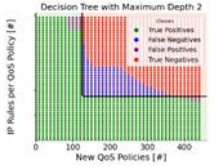




Accuracy
77.93%

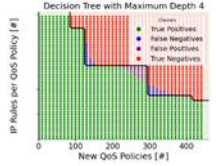
Maximum Depth 2





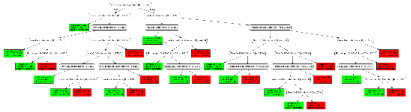
93.04%

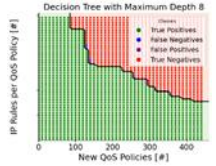
Maximum Depth 4



98.22%

Maximum Depth 8





98.67%

Summary and Conclusions

➔Deploying new features to critical infrastructure often requires resource testing

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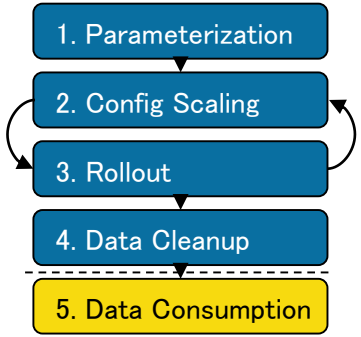
➔Our method/framework supports resource testing automation

- Generates more than 2000 data points in < 24 hours
- Identifies bottleneck per router and per line card
- Creates accurate and human readable prediction models



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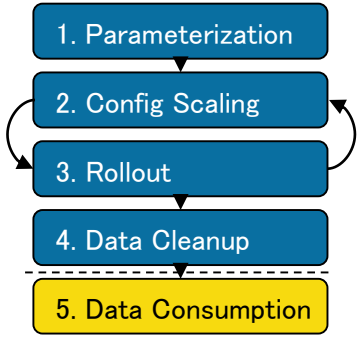

➔The general method has been used in practice by DE-CIX

- For assessing configuration changes and for dimensioning products
- For assessing the accuracy of simulated router instances vs. real-world hardware
- For validating vendor claims on HW capabilities

Thank You for Your attention!

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