Automatically Correcting Networks with NEAt

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Networks are so complex it's hard to make sure they're doing the right thing.

What is a network supposed to do?



- no untrusted traffic entering a secure zone
- the preference of one path over another
- loop & black hole avoidance

A Typical Enterprise Network



Source: http://www.cisco.com/c/en/us/td/docs/solutions/Enterprise/Medium_Enterprise_Design_Profile/MEDP/chap5.html

Lots of problems arise today



89% of operators are <u>not</u> certain their configuration changes are safe. [Kinetic NSDI'15]

Networks are so complex it's hard to make sure they're doing the right thing.

Let's automate.

What to automate?

Automatically *identifying* errors in networks (Verification)





What if we can automatically correct networks on the fly?



What if we can automatically correct networks on the fly?



What if we can automatically correct networks on the fly?



Network Abstraction



Operating on the data plane simplifies our work

- Diagnose problems as close as possible to actual network behavior
- Data plane is a "narrower waist" than configuration

Goal: Improve upon a manual effort with transparency in both <u>performance</u> and <u>architecture</u>.

Challenge I: Repair speed

- Based on real-time verification technique
- Derive fixes via linear optimization, with min. changes
- Topology limitation & graph compression

Challenge 2: Zero/minimal architecture/application changes

- Minimal changes
- Pass-through mode
- Interactive mode









Application Mode





Policy as Graphs

Graphs are *<u>neat</u>*

- Network state synthesis \rightarrow viewing the network as a whole.

A policy graph is defined on a packet header pattern

• ip dst 10.0.1.0/24, port 443.

Reachability	m = 1
Bounded path length (shortest path)	m = l n = path_length
Multipath/Resillience	m = k (k > 1)
lsolation	m = 0







Policy as Graphs (Cont'd)



Policy

• X

Service Chaining



Load balancing



Use *policy graphs* to express both <u>qualitative</u> and <u>quantitative</u> reachability constraints

Cast the problem as an optimization problem:

- Map forwarding graph to policy graph
- Minimize # of changes





Correction

Engine

Cast the problem as an optimization problem:

- Map forwarding graph to policy graph
- Minimize # of changes
- boolean variable x_{i, j, p, q}:
 - topology edge (i, j) policy edge (p, q)
- s.t., policy level reachability (p, q)



Correction Engine



Repair - Generalized Reachability

Basic Reachability

$$\forall (i,j) \qquad x_{i,j} \ge \sum_{(p,q) \in E_{\mathscr{O}c}} \frac{x_{i,j,p,q}}{N(E_{\mathscr{O}})}$$

$\forall (j, No x tight oppose for a constraint oppose for$

 $\forall (j,i) \qquad x_{i,j} + x_{j,i} \le 1$

 $\forall (p,q), \forall i \in T:$

 $\begin{cases} \sum_{j \in NB_T(i)} x_{i,j,p,q} = 1\\ \sum_{j \in NB_T(i)} x_{j,i,p,q} = 0 \end{cases}$

$\begin{cases} \textbf{Flow}_{T}(\textbf{conservation} \\ \sum_{j \in NB_{T}(i)} x_{j,i,p,q} = 1 \end{cases} \text{ if } i = q \end{cases}$

 $\left\{\sum_{j\in NB_T(i)} (x_{i,j,p,q} - x_{j,i,p,q}) = 0\right\}$

otherwise

if i = p

 Isolation $\begin{cases} \sum_{j \in NB_T(i)} x_{i,j,p,q} = 0 \\ Flow sinks^q \text{ at } DROP \text{ node} \end{cases}$ $\left\{\sum_{j\in NB_T(i)} x_{i,j,p,q} = 0\right\}$ if i = q Service Chaining $\begin{cases} \sum_{j \in NB_{T}(i)} x_{i,i,p,q} = 0 \\ \sum_{j \in NB_{T}(i)} x_{j,i,p,q} = 0 \\ \end{bmatrix} \text{ waypoint} \text{ order}_{q,i} \end{cases}$ •Bounded or Equal Path Length $\sum_{(i,i)\in F} path length <= n$ MultiPath (Resilience) $\sum_{j \in NB_T(i)} x_{i,j,p,q} \ge m$ if i = p $\sum_{j \in NB_T(i)} x_{j,i,p,q} = 0$ $\sum_{j \in MB_T} of_{i,j} paths \geq m_{if} n_{i} = q$ $\sum_{j \in NB_T(i)} x_{j,i,p,q} \ge m$ Load Balancing Flow distribution propagates

Missing-abything?eedom

The preceding algorithm operates on a loop-free graph.

First check for and remove loops before repairing other type violations.

Objective: Minimize changes

- Remove the minimal # of rules.
- Affect few packets as possible.
 - E.g. remove a rule matching 10.0.0.1/32 over one for 10.0.0.0/8.



Correction

Engine

Scalability Challenge and Solution

Scalability challenge

- # of variables $\approx |E(G_{topo})| \times |E(G_{policy})|$
- Easily exceeds 100k

Solution: ?



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Scalability Challenge and Solution

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Solution

- Topology Limitation
- Graph Compression w.r.t policy
 - Key: Compressed graph == original graph
 - Bisimulation Based Graph Compression

Prototype implementation in Python

- Use Gurobi within optimization engine
- Pass-through mode: proxy
- Interactive mode: XML-RPC API

Datasets:

- Synthetic fat-tree configurations
- SDN applications
- 244-router enterprise network trace

Application End-to-End Delay

Pox + Mininet

- Learning switch app (pass-through)
- Load balancing app (interactive)



244 routers, one million forwarding rules

Policy: loop freedom & reachability

Issues found and repaired:

- Loops caused by default route
- Load balancing shouldn't be turned on

Synthesizer (NetGen) as repair tool

#TopoLinks	NEAt	NetGen	NetGen-C
96	5.9ms	743.2ms	513.2ms
324	7.2ms	4404.0ms	1160.8ms
768	9.0ms	16337.7m	2056.3ms

NEAt as synthesizer

#TopoLinks	NEAt	NetGen
96	921.7ms	7.1 min
324	16.3s	381.7min
768	2.9min	173.2hrs

NEAt, a system analogous to a smartphone's autocorrect.

- Casting the problem as an optimization problem
- Millisecond to second repair speed
- Generic policy support

Future work:

- Evolving & richer policies
- Different optimization goals
- Repair relevance study



Graph Compression

w.r.t policy Key: Compressed graph == original graph

Major building block:

Bisimulation Based Graph Compression*





*Query preserving graph compression. SIGMOD 2012. W. Fan et al

Graph Compression (Cont'd)

Customized policy-preserving compression

Topology	Compression Ratio
Fattree (6750 hosts, 1125 switches)	99.38%
Enterprise (236 routers)	88.98%

- Incremental Compression
- •Repair compressed graphs

 $\underbrace{\mathsf{M}}_{\substack{j \in NB_{T}cp(i) \\ x_{j,i,p,q}}}^{j} \underbrace{\mathsf{M}}_{j,i,p,q}^{(x_i,d_{p,q},w_{ij},h_{ij})} = \underbrace{\mathsf{M}}_{m}^{m} \mathsf{constraint}$

- Mapping back
- Proved Policy Perseverance





Model packet space as a set of Equivalence Classes

Equivalence class (EC): Packets experiencing the same forwarding actions throughout the network.

Fwd'ing rules Equiv classes

Model forwarding behavior of each EC as a directed graph





Preventing errors at <u>run-time</u>

- Allow arbitrary SDN applications to run on top
- Not restricted to any programming language
- <u>Influence</u> updates, rather than synthesize from scratch

Graphs are *neat*

- Networks are graphs
- Model network forwarding behaviors as directed graphs
- Represent operator intents as a <u>policy graph</u>

Discovering repairs

 Equivalent to modifying network state graph so that there exists a mapping between it and the policy graph







Con	figuration	Configuration	Data-plane
Con	trol plane	Prediction is difficult: • Various configuration	Closer to actual network behavior
Da	ta plane	languagesDynamic distributedprotocols	Unified analysis for multiple control-plane protocols
Network behavior	Misses control-plane bugs	Can catch control-plane bugs	
		Test prior to deployment	Only detects bugs that are present in the data plane

Operating on the data plane simplifies our work

- Diagnose problems as close as possible to actual network behavior
- Data plane is a "narrower waist" than configuration